

**THREE ESSAYS ON ECONOMIC AND POLITICAL
INSTITUTIONS**

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**UNIVERSIDAD DEL CEMA
2006**

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DEDICATION

To my wife, Patricia

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PREFACE

This thesis is a study on the behavioral incentives and control mechanisms defined by economic and political institutions. The three essays are empirical quantifications of the economic effects implied by different institutional arrangements. We examine the economic implications of legal and political structures using economic theory, econometric analysis and hypothesis testing.

The first essay is a study of incentives for resource use under different contractual arrangements in the Argentinean agricultural production. Using a transaction costs approach, we contend that in a context of modern agriculture, with well defined property rights, agricultural contracts must balance costs and benefits, aligning tenant and landlord incentives towards a similar objective. The study debates the potential effects of tenancy status and duration of contracts, over soil conservation and input use. We present empirical evidence about the effects over the soil and input use in tenant and owner-operator farms using farm level data from the 2002 National Agricultural Census of Argentina. Contrary to the conventional wisdom, the empirical results do not support a general and clear negative effect for tenancy arrangements. Our intuition is that the interaction among specific characteristics of farmers, natural resources endowment and institutional environment are more important factors than the land tenancy or contract type itself.

The second essay looks at the relationship between electorally motivated fiscal policy cycles and separation of powers. Previous empirical work on electoral cycles implicitly assumes the executive has full discretion over fiscal policy. In contrast, we contend that an unaligned legislature may have a moderating role under separation of powers. Focusing on the budget surplus, we find that stronger effective checks and balances explain why cycles are weaker in developed and established democracies. Once the discretionary component of executive power is isolated, there are significant cycles in all democracies. Hence, what we add to the ongoing debate about the factors behind conditional Political Budget Cycles is a study of the role of effective checks and balances that reduce the discretion of the executive.

The third essay presents evidence of electorally-motivated changes in the budget balance, public expenditures, composition of public expenditures and provincial revenues in Argentine provinces. The empirical study is made using panel data analysis

for 22 Argentine provinces during the period 1985-2001. Results show that conditioning on the alignment of provincial and federal executives (same political party in power) there is evidence of systematic changes in fiscal policies around elections. The observed changes support the predictions of rational opportunistic models of Political Budget Cycles. In election years, total provincial expenditures increase in aligned provinces, without affecting the fiscal balance, because to the increased discretionary transfers from the federal government supporting the provincial incumbent federal revenues. By contrast, deficit increases for unaligned provinces. In addition, expenditure shifts toward current spending and away from capital spending for unaligned provinces in electoral years.

Finally, I wish to thank the members of my Thesis Committee Professors Jorge M.Streb (Chairman), Germán Coloma, Marcos Gallacher and Juan Jorge Medina for their insightful suggestions. A special acknowledgement I owe to Professor Streb for helpful, deep and extensive discussions on issues contained (and not) in this study. I also wish to thank the Director of the Department of Economics, Professor Mariana Conte Grand, for their warm and continuous encouragement. The Instituto de Economía y Sociología at INTA (Instituto Nacional de Tecnología Agropecuaria) has provided important support for developing this study. I am very grateful for it.

Chapter 1

Contracts, Transaction Costs and Agricultural Production in the Pampas

I. Introduction

Land tenure and contractual arrangements are controversial issues in the Pampean agriculture. The early colonial pattern of large land holdings dictated the system of sharecropping agriculture with immigrant farmers that developed after 1860. Some authors describe the sharecropper system as a rational economic arrangement that favored landlords and tenants. There was no collusion on the part of large landholders to bar access to land by the newly arrived European farmers and the land market was open and competitive without legal or economic barriers to entrance (Cortés Conde, 1995). However, other authors contend that tenure regime and sharecropping arrangements had negative social and productive consequences (Scobie, 1964; Ferrer, 1965).

Eventually, through inheritance or sale, many of the very large “estancias” (cattle ranch) were broken up, but the original pattern of land occupation resulted in larger landholdings than in similar regions of the U.S. as the “Corn Belt”. In spite of these differences in land tenure arrangements, Gallacher et. al. (2003) suggests that a similar overall performance is observed in agricultural production in both countries because farmers are efficient in resource allocation (including land tenure arrangements).

During the last 15 years, Argentine agricultural production has been rising and land rental (both fixed rent and sharecropping) is a growing practice, implying a greater separation between the property and control of land. In this paper, our objective is to show that land tenure arrangements in the Pampean agriculture align interests and incentives in an efficient way. Using a transaction costs approach, we present empirical evidence about the effects of agricultural contracts (fixed rent, sharecropping) on soil conservation practices and input use (fertilizers).

The organization of the paper is the following. Section II briefly reviews the literature and presents the theoretical background. Section III describes the characteristics of the study area and the data set from the 2002 National Agricultural Census (NAC). Section IV presents the econometric analysis. Section IV has the final comments.

II. Literature Review and Theoretical Background

The analysis of fixed rent and sharecropping contracts has been extensively developed in the literature. In a fixed rent contract, the tenant pays a fixed amount not related to farm production. On the other hand, the sharecropping contract allows the tenant only a fraction of the total product (e.g., between 65 and 70% in Pampean grain production).

In a sharecropping contract the tenant has incentives to under-utilize inputs, this is widely known as the sharecropping problem in its “Marshallian” version (Johnson, 1950). Several reasons have been proposed to explain the use of sharecropping contracts. One incorporates risk in the analysis of contracts (Stiglitz 1988). Under this explanation, sharecropping is a way to share risk between the landlord and the tenant and the sharecropping contract appears as a choice in order to avoid risks. Therefore, the tenant shares not only the product but also part of the risk associated with agricultural production. One reply to this argument is that if there are no restrictions to make multiple fixed rent contracts, it is possible to avoid risk diversifying the use of fixed contracts (Newbery 1977). Alternatively, the “moral hazard” approach suggests that efficient contracts balance the exchange between the costs associated with the risk and the benefits derived from generating optimal incentives for both parties. A sharecropping contract can be seen as a result of this balance (Stiglitz 1988).

These models have been considerably developed in the literature and empirically applied in the study of the contractual relations. Empirical results are mixed about efficiency under fixed and sharecropping contracts, with studies focused on developing countries with traditional agricultural sectors.

Our analysis will concentrate on the relationship between the landlord and the tenant using a theoretical framework associated with the transaction costs approach. Cheung (1968) shows that with well defined property rights the type of contracts does not affect efficient resource allocation. A critical assumption is the absence of transaction costs and in particular the inexistence of monitoring costs in the use of inputs or the effort made by the tenant.

Following Allen and Lueck (2002), we do not consider risk and we add to the analysis the use of specific characteristics of land. Specifically, the soil attributes are treated as an additional input in the production process. When a producer carries out the production in his own land, he manages the resources taking into account the present and future implications of his decisions. By contrast, a tenant with a fixed rent contract will only worry about his current results. Then, if greater yields could be obtained by

putting aside adequate soil management, applications of fertilizers or other practices, the tenant has incentives in this direction.

In sharecropping contracts, if effort is observed imperfectly and there are monitoring costs, there would be incentives to underutilize inputs by the tenant. This implies, also that he may have less incentives to use the soil attributes excessively or to carry out actions with potential harmful effects over the natural resources. This could be seen as a potential benefit of sharecropping contracts (Allen and Lueck 2002).

However, this does not imply that sharecropping is always the most convenient arrangement for the landlord. Transaction costs are important in controlling and dividing the output, because the tenant has incentives to underreport the quantity of crop to the landlord. Of course, the landlord is aware of this problem and he will do all that he can to avoid this behavior.

The relative advantage of a fixed rent contract is to avoid the quantity control. However, it presents the problem of over-utilization of soil attributes. The sharecropping contract reduces the incentives to dig the soil, but it has costs related to the control of quantity and quality of crop.

Some other factors can lead the actions of tenants and landlords to the optimal use of the resources. For example, repeated transactions can build a reputation and reduce the costs of control. If transactions are less frequent but the landlord has good knowledge of the activity, he can reduce the monitoring costs. If control of production is relatively more costly, then he can opt for fixed rent contracts.

It is often argued that short-term contracts do not generate adequate incentives for both the conservation of resources and investments. However, when an owner-operator decides how to manage his land, he has as an inter-temporal profit-maximizing objective. When he considers the option of renting the land to a tenant, the analysis cannot be different. The landlord surely is aware of the incentives that the tenant has to make an over use of the soil attributes in the short term.

Our working hypothesis is that the design of the contracts should align interests of tenants and landlords, minimizing transaction costs. The contract design should make the actions converge in such a way that the results for a tenant will be similar to those of an owner or landlord-operator. However, a greater alignment of interests tends to increase the complexity and the costs of the contracts. Longer contracts may stimulate the conservation of assets and soil, but at the same time require more detailed conditions that are costly to control and enforce.

Pampean agricultural production is based mostly on short-term contracts, but the transactions are repeated and frequent. The incentive to build a reputation can act as an alternative mechanism for long-term contracts. A repeated short term contract can have implicit renovation if the tenant carries out the expected actions, but it is revoked easily if not. Our intuition is that we should not observe a systematic bias in resource allocation between annual tenants (fixed rent or sharecropping) *vis à vis* landlord-operators¹.

Hence, our principal conjecture is that contractual arrangements must balance costs and benefits, aligning incentives towards an objective similar to a producer-landlord with full interest in maximizing and conserving his wealth. We do not expect major differences between owner operators and tenants in input use or natural resources conservation.

If the crop share contracts do not give full incentives for the optimal use of inputs, and there are monitoring and control costs, some differences in input use (e.g., fertilizers) could be found in case of fixed rent contracts with respect to sharecropping contracts.

These conjectures are empirically tested in section IV.

III. Study Area and Data

The geographic focus of this study is the central-eastern region of Argentina known as the Pampas, one of the most productive agricultural areas in the world and of major importance to the Argentine economy (85% of the total grain production). Wheat and corn have been the principal crops for the last 100 years and soybean is a more recent crop.

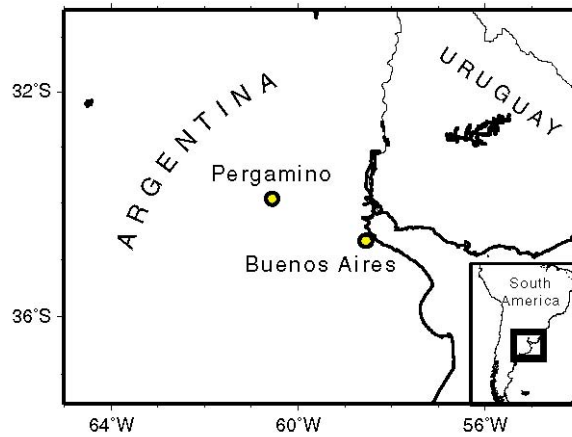
The empirical analysis is carried out using farm level data from the National Agricultural Census 2002 (NAC 2002) of Pergamino County, Province of Buenos Aires.

Cropping systems include maize, soybean, wheat-soybean double crop and characteristic rotations include maize and soybean. Pergamino is representative of the

¹ In Pampean agriculture, annual contracts prevail. According to current legislation, all agrarian contracts must be signed for three years and registered in courts. Even though detailed statistical information is not available on the fulfilment of this requirement, it is a well-know fact in the rural media that the majority of contracts do not comply with this formality. The evidence points out that for different reasons, surely linked to the transaction costs, farmers have opted to set up informal contracts. In this sense, we consider that the actual legal framework is neutral for the selection of the contracts and the productive decisions.

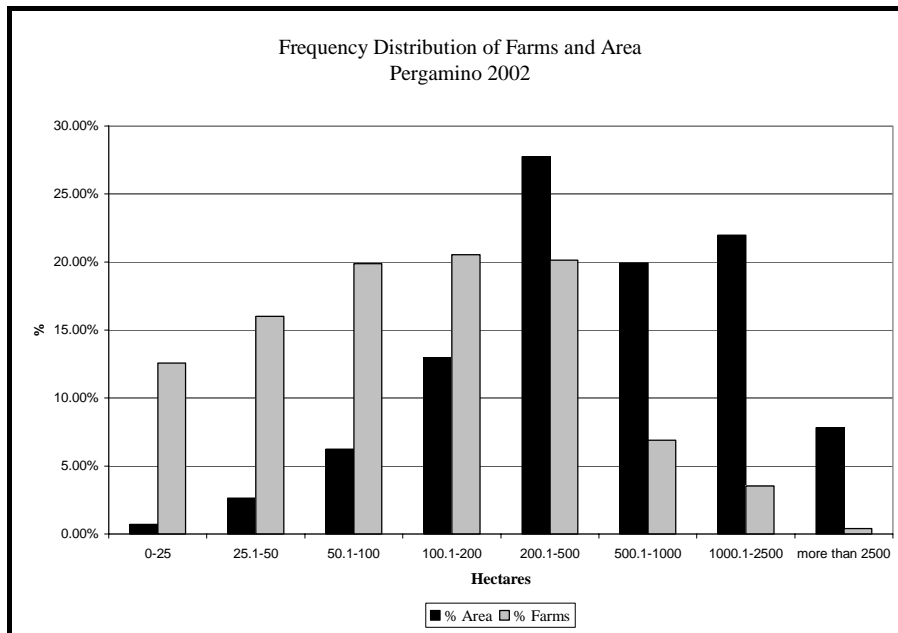
Pampas and it presents characteristics of modern agriculture (defined property rights, modern inputs and technological knowledge) which makes it comparable, for example, with the American “Corn Belt”. Figure 1 displays the location of the study area.

Figure 1. The study area.



The available micro data includes productive, economic and management variables for 1117 farms in a total area of 285,992 hectares, averaging 256 hectares per farm. Figure 2 presents the frequency distribution of farms and area.

Figure 2. Frequency Distribution of Farms and Area



IV. Econometric Analysis

We carry out estimations of binary election models (Probit) to explain the utilization of soil conservation practices (no tillage or reduced tillage). Selection models (Tobit) are used to explain the cultivated area with no till practices and total fertilized area. From the 1117 farms we considered those that produced at least some of the four principal crops (soybeans, corn, wheat and sunflower). The result is a total of 944 observations available for the study. Dependent and independent variables and their definitions are presented in Table 1.

Table 1. Definition of Variables

<i>Dependent Variables:</i>
CONS: Dummy variable that assumes the value of one if on the cultivated area some conservationist practice is carried out (e.g. no tillage, reduced tillage,).
AREANT: Area with no till practice as percentage of the total area cultivated with soybeans, corn, wheat and sunflower.
TOTFERT: Total fertilized area as percentage of the cultivated area in corn, wheat and sunflower.
CWFERT: Fertilized area in corn and wheat crops as percentage of the total implanted area with corn and wheat.
SSFERT: Fertilized area in soybeans and sunflower crops as percentage of the total implanted area with soybeans and sunflower.
<i>Independent Variables:</i>
T: Dummy variable for tenant farms. It assumes the value of one if the ratio between the own cultivated area (OA) and the total area of the farm (ATOT) is less or equal to 0.20 ($0.20 \geq OA/ATOT$).
TF: Dummy variable for type T farms with fixed rent contract. It assumes the value of one if the rent is fixed in a proportion greater or equal than 0.80 with respect to the total rented area.
TS: Dummy variable for type T farms with sharecropping contract. It assumes the value of one if the sharecropped area is a proportion greater or equal than 0.80 with respect to the total area rented.
T_OTHER: Dummy variable that takes value of one for type T farms that belong neither to the category TF nor to the category TS ($T_OTHER = T - TF - TS$).
OWN_TEN: Dummy variable for farms that combine own land and tenancy. It assumes the value of one if $0.20 < OA/ATOT < 0.80$.
CULTA: Total cultivated area with wheat, corn, soybeans and sunflower, in thousands of hectares.
SOY: Dummy variable that assumes the value of one if the farm produces only soybeans
SUMM: Dummy variable that assumes the value of one if the farm carries out just summer crops (corn-soybeans)
PLOWS: Total number of plows (or similar equipment)
TRACT: Dummy variable that takes the value one if the farm has one or more tractors.
DIR: Total number of machinery for direct seed-planting (no till machinery)
SERV: Total area contracted for services of plowing and soil preparation in thousand of hectares
MAINT: Total area contracted for maintenance work and conservation of the crops, in thousands of hectares.
EDU: Education of the producer measured in a scale between 1 and 7.5 (1: no education; 7.5: complete college education) (this variable assumes the value of zero when the farm is some type of partnership or corporation).
EDUD: Dummy variable that assumes the value of one when the variable EDU assumes the value of zero (it controls for possible bias in the coefficient associated with EDU due to the inclusion of zeros)
RESID: Dummy variable that assumes the value of one if the producer or any of the partners resides on the farm.
MANAG: Dummy variable that assumes the value of one if the farm keeps formal accounting and productive records.
PUBEXT: Dummy variable that assumes the value one if the farm receives extension services from some public organization (state or federal).
PRIVADV: Dummy variable that assumes the value of one if the farm uses some private technical advise (independent professionals, companies, NGOs)

CONS is a binary variable that identifies the use of the conservation practices (when CONS=1). A group of four continuous variables measures the relative adoption of no till practices (AREANT) and fertilizing (TOTFER), the relative fertilization in cereals (CWFER), and fertilization in oilseeds (SSFER). Table 2 presents the probit estimation for the binary dependent variable CONS.

Table 2. Probit (Conservation Practices) and Tobit (No Till - Fertilization) Estimates

Variables	CONS (1)	CONS (2)	AREANT (3)	AREANT (4)	TOTFERT (5)	TOTFERT (6)
T	-0.177 (-1.55)		-2.654 (-0.56)		-7.593 (-1.91)*	
TF		-0.251 ^A (-1.63)		-11.192 ^B (-1.75)*		-5.901 ^C (-1.13)
TS		0.119 ^A (0.59)		4.867 ^B (0.57)		-11.034 ^C (-1.47)
A_OTHER		-0.300 (-1.68)*		3.426 (0.48)		-7.845 (-1.30)
OWN_TEN	-0.103 (-0.89)	-0.106 (-0.92)	-1.986 (-0.42)	-2.035 (-0.43)	-2.347 (-0.62)	-2.342 (-0.62)
CULTA	-0.365 (-2.43)**	-0.361 (-2.39)**	-2.964 (-0.50)	-20.754 (-0.47)	-7.036 (-1.40)	7.119 (-1.42)
SUMM	-0.156 (-1.62)	-0.165 (-1.71)*	-23.440 (-5.82)***	-23.825 (-5.92)***		
SOY					-63.220 (-15.58)***	-63.122 (-15.54)***
PLOWS	-0.177 (-5.42)***	-0.177 (-5.42)***	-6.291 (-4.85)***	-6.192 (-4.78)***		
TRACT	-0.676 (-5.12)***	-0.670 (-5.04)***	-27.386 (-4.91)***	-28.174 (-5.04)***	-7.798 (-1.78)*	-7.770 (-1.77)*
DIR	0.585 (6.87)***	0.594 (6.93)***	23.749 (7.76)***	23.893 (7.80)***	4.333 (1.84)*	4.299 (1.83)*
SERV	0.552 (2.34)**	0.557 (2.36)**	20.453 (2.18)**	21.407 (2.28)**		
MAINT					5.635 (2.70)***	5.578 (2.67)***
EDU	0.107 (3.36)***	0.111 (3.50)***	4.738 (3.53)***	4.860 (3.62)***	2.593 (2.32)**	2.548 (2.27)**
EDUD	0.627 (3.20)***	0.650 (3.30)***	20.461 (2.45)**	20.322 (2.44)**	16.190 (2.36)**	16.104 (2.35)**
MANAG	0.241 (2.25)**	0.238 (2.21)**	6.436 (1.44)	7.136 (1.60)	13.906 (3.73)***	13.843 (3.69)***
PUBEXT	0.458 (2.16)**	0.452 (2.13)**	13.312 (1.57)	13.260 (1.56)	9.621 (1.37)	9.695 (1.38)
PRIVADV	0.221 (2.00)**	0.210 (1.90)*	9.725 (2.13)**	9.514 (2.08)**	4.441 (1.15)	4.569 (1.18)
RESID	-0.271 (-2.45)**	-0.270 (-2.42)**	-14.960 (-3.28)***	-14.897 (-3.28)***	-7.642 (-2.07)**	-7.673 (2.07)**
Constant	-0.224 (-1.05)	-0.236 (-1.11)	44.939 (5.00)***	44.695 (4.99)***	19.833 (2.61)***	19.956 (2.62)***
Method of Estimation	Probit	Probit	Tobit	Tobit	Tobit	Tobit
No. Observations	944	944	944	944	944	944
Censored Observations			324	324	353	353
(Dep. Var.<=0)	-	-				
Not Censored Obs.	-	-	620	620	591	591
Log-Likelihood	-524.05	-521.15	-3620.897	-3618.903	-3243.137	-3242.953
LR Test	213.44***	219.26***	240.12***	244.11***	390.10***	390.46***
Pseudo R²	0.171	0.174	0.0321	0.0326	0.0567	0.0568

Notes: z statistics in parentheses; *** Significant at the 1%; **Significant at the 5%; *Significant at the 10%; A and B: see Table 3 for Wald test of coefficients equality

The independent variables are grouped in those measuring the type of land tenancy (T, TF, TS, T_OTHER, OWN_TEN) and those that control by productive characteristics (CULTA, SOY, SUMM, SERV, MAINT), physical capital (PLOWS, TRACT, DIR), human capital and management (EDU, RESID, MANAG, PUBEXT, PRIVADV).

In the first model, the variables T and OWN_TEN were included to estimate the effect of these two forms of tenancy on CONS (controlling for covariates). In the second equation the tenancy status is distinguished by type of contract, including the variables TF, TSP and A_OTHER. Our main interest is on coefficients associated with tenancy variables (T, TF and TS), and we observe that those coefficients are not significant in any of the estimations. Only the coefficient associated with the category T_OTHER appears with negative sign and marginally significant at 10% in estimation 2. The estimated coefficient of TS has a positive sign and that of TF has a negative sign (and marginally significant at 11%). Following the theoretical conjecture that there are greater incentives to over use soil attributes in fixed rent contracts, this finding may imply a differential effect between fixed rent and crop share contracts. Table 3 (line A) presents a Wald test that contrasts the hypothesis of equality of both coefficients.

Table 3. Wald Test of Coefficients Equality (Fixed Rent and Sharecropping Contracts)

	Statistic	p-value	Statistic	p-value	
A	chi2(1)= 2.44	0.11	D	F(1, 570)= 0.66	0.42
B	F (1, 928) = 2.64	0.10	E	F(1, 929) = 2.81	0.09
C	F(1, 929) = 0.36	0.55	-	-	-

The result shows that (marginally) at 11% we can reject the null hypothesis of equality. This suggests some differential effect of greater adoption of conservation practices in cases of sharecropping contracts. So, the tenancy status appears relatively neutral in terms of conservation practices, with a slightly superior adoption of conservation practices in crop share contracts.

Regarding control variables, it is clear that the quantity and type of available machinery affects the adoption of conservation practices, since the effect of PLOWS and TRACT over CONS appears to be systematically negative and significant, while the

effect of DIR is positive and significant. Variables related with human capital and management presents a positive and significant effect over the adoption of conservation practices.

Columns 3 to 6 in Table 2, present the estimations using no till area (AREANT) and fertilized area (TOTFERT) as a percentage of the total farming area as dependent variables. Farms that do not carry out soil conservation practices or do not fertilize always have a percentage equal to zero. To address this problem of sample selection, we used models of simultaneous selection (Tobit) to perform the estimations.

Equation 3, with no till practices (AREANT) as dependent variable, shows that the estimated coefficient for variable T is not significant. On the other hand, in equation 4, the coefficient associated with fixed rent (TF) is negative and significant. This result is similar to the conservation practices equation. In the same way, we perform a Wald test to contrast equality between the estimated coefficients for TF and TS. Results (line B Table 4) suggest a greater use of no till practices for crop share tenants.

The coefficient associated with tenancy (T) is negative and significant at 10% in equation 5. However, controlling by contract type, there are no significant differential effects relative to the base category (landlords). We also tested the null hypothesis of equality between these coefficients (line C Table 4). The Wald test does not reject the null hypothesis of coefficients equality.

The effect of the dummy variable SOY, that controls farms dedicated only to soybean production, appears negative and significant in fertilization equations. Fertilization in soybeans is much less frequent, since marginal yield response is reduced. In order to control this effect we analyzed the practice of fertilization in two sub samples. One sub sample includes farms producing cereal crops and the other those producing oilseeds. Estimation results for each sub sample are presented in Table 4 (equations 7 to 10).

Equation 7, shows that the tenancy variable is significant and positive when fertilization in corn and wheat (CWFERT) is the dependent variable. Equation 8 includes dummy variables for sharecropping and fixed rent contracts, and the Wald test (line D in Table 4) suggests equality between estimated coefficients.

For oilseed crops fertilization (SSFRT, equation 9) the coefficient associated with T is negative and significant. When the effects are separated by contract type (equation 10), the negative effect on fertilization by the tenants is explained principally

by the group of sharecropping tenants. The Wald test (line E in Table 4) allows the rejection of the null hypothesis of equality.

Table 4. Tobit Estimates (Cereals and Oilseeds Fertilization)

Variables	CWFERT (7)	CWFERT (8)	SSFERT (9)	SSFERT (10)
T	15.757 (2.65)***		-44.403 (-2.33)**	
TF		12.054 ^D (1.59)		-31.592 ^E (-1.30)
TS		22.316 ^D (1.97)**		-129.616 ^E (-2.34)**
A_OTHER		17.491 (1.91)***		-28.223 (-0.98)
OWN_TEN	9.995 (1.83)*	10.032 (1.84)*	13.518 (0.82)	13.680 (0.83)
CULTA	-3.316 (-0.95)	-6.111 (-0.92)	-40.744 (-1.41)	-42.146 (-1.45)
TRACT	-2.447 (-0.35)	-2.732 (-0.39)	-25.239 (-1.39)	-27.166 (-1.50)
DIR	10.007 (3.03)***	10.049 (3.05)***	-0.068 (-0.01)	-0.316 (-0.03)
MAINT	3.033 (1.12)	3.124 (1.15)	25.986 (2.47)**	25.666 (2.43)**
EDU	3.613 (2.15)**	3.675 (2.19)**	5.124 (1.05)	4.339 (0.89)
EDUD	24.803 (2.46)**	24.937 (2.24)**	18.245 (0.60)	14.485 (0.47)
MANAG	20.514 (3.72)***	20.742 (3.73)***	54.670 (2.98)***	55.606 (3.03)***
PUBEXT	24.788 (2.42)**	24.706 (2.41)**	-5.005 (-0.16)	-4.807 (-0.16)
PRIVADV	9.158 (1.53)	8.821 (1.47)	9.492 (0.56)	10.924 (0.67)
RESID	-11.683 (-2.17)**	-11.544 (-2.15)**	-32.624 (-1.85)*	-33.204 (-1.88)*
Constant	61.878 (5.35)***	61.840 (5.26)***	-185.040 (-5.09)***	-180.928 (-4.98)***
Method of Estimation	Tobit	Tobit	Tobit	Tobit
No. Observations	584	584	943	943
Censored Observations (Var. Dep.<=0)	51	51	824	824
Not Censored Observatons	533	533	119	119
Log likelihood	-2951.476	-2951.1134	-965.845	-963.829
Likelihood Ratio Test	79.87***	80.59***	35.79***	38.92***
Pseudo R2	0.013	0.013	0.018	0.020

Notes: z statistics in parentheses; *** Significant at the 1%; **Significant at the 5%; *Significant at the 10%; C, D and E: see Table 3 for Wald test of coefficients equality.

Though the tenancy effect on fertilization is negative when all crops are considered together, it appears to be reasonable to differentiate the effect analyzing separately the cereal and oilseed crops, because they have a different marginal response. Cereal crops present a greater response to nitrogen fertilization. On the other hand, for soybeans this fertilizer has little marginal effect on yields. The application of phosphorus an element with positive residual effects for subsequent crops is more frequent.

The fertilization decision includes two criteria: sufficiency and replacement. The sufficiency criterion is to fertilize only when the level of nutrients in the soil is below the critical value. On the other hand, the replacement criterion, is to fertilize systematically, adding the quantities of nutrients that the crops extract.

We interpret the empirical findings as follows: for cereal crops, even though the tenants do not have incentives to apply the replacement criteria (because they only have a temporary property right on the land) they do have strong incentives to apply the sufficiency criterion to increase yields. Empirical results show that the effect of sufficiency criterion seems to be important, implying that tenants tend to fertilize, on average, more than the owner operators. We can conjecture that owner operators will resort to other practices that substitute the application of fertilizers in cereals (e.g. crop rotations or soybean fertilization as precedent crop).

The theoretical analysis indicates that incentives for fertilizing could be lower in sharecropping contracts. This situation is not clearly distinguished in the estimations since we do not find significant differences between coefficients. Perhaps, greater information about contracts is necessary to distinguish the effects. It is observed that in oilseed crops (soybeans) the effect of the tenancy category is clearly negative over fertilization, particularly in the case of the sharecropping contracts. In this case the sufficiency criteria may have a low impact, since the effects of fertilizers are reduced, and also there are low incentives for replacement, resulting in a clear negative effect.

Summarizing, for cereal crops the tenants (fixed rent or sharecroppers) tend to fertilize more than owners. For oilseed, due to the lower marginal response and the greater residual effect of phosphorus, a negative effect is observed for tenants, in particular for sharecroppers.

V. Final Comments

Land tenancy and contract arrangements used in the Pampean agricultural production are important and controversial issues. However, at least to the best of our knowledge, there are no studies that approach the subject with a transaction costs analytical framework and empirically contrast the conjectures. Our study debates the potential effects over soil conservation or input use of tenancy and duration of contracts. The empirical results show some differential effects but do not support a general and clear negative effect in tenant farms.

Finally, our empirical results are consistent with the theoretical conjecture that the different contract arrangements tend to minimize transaction costs, resulting in a similar resource allocation without superiority of land ownership over land rental by tenants.

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Chapter 2

Discretionary Political Budget Cycles and Separation of Powers *

I. Introduction

Without discretionary power, there is no room for political budget cycles (PBC). Unlike asymmetric information, the degree of discretion of the executive has been overlooked in the empirical literature on PBC, perhaps because theoretical papers on opportunistic cycles usually model fiscal policy in terms of a single policy maker with full discretion. However, in the U.S. two-party system Alesina and Rosenthal (1995) show how divided government is a tool to moderate the executive. A similar logic might apply in an opportunistic framework, where an opposition legislature may play a special role in moderating PBC. Indeed, Schuknecht (1996) suggests that stronger PBC in developing countries might be due to the existence of weaker checks and balances there.

Hence, what we add to the ongoing debate in Shi and Svensson (2002a, 2002b), Persson and Tabellini (2002), and Brender and Drazen (2004) about the factors behind conditional PBC is a study of the role of effective checks and balances that reduce the discretion of the executive. To measure the nominal presence of a legislative veto player, we use the Henisz (2000) political constraints index. We then construct a measure of effective checks and balances, as the product of the presence of a legislative veto player and the International Country Risk Guide (ICRG) measures of rule of law.

We focus on the behavior of the budget surplus, because it is the most sensitive indicator of aggregate PBC. We also look at the effect of checks and balances on the persistence of the budget surplus, taking into account the literature on the costs of coalition governments and divided government in terms of slower adjustment to shocks (Sachs and Roubini 1989, Alt and Lowry 1994), and more generally the suggestion in Tsebelis (2002) that more veto players imply that it is harder to change the status quo.

* This chapter is based in a research project on Political Budget Cycles developed in collaboration with Jorge M. Streb and Gustavo Torrens.

We specially thank Alejandro Saporiti for helping to start this project. Adi Brender and Allan Drazen provided their database on political budget cycles. It was great to receive insightful suggestions from Marco Bonomo and Sebastián Galiani. We benefited from comments by María Laura Alzúa, Mauricio Cárdenas, Alejandro Corbacho, Andrés Escobar, Marcela Eslava, Leonardo Hernández, Jorge Nougués, Ernesto Stein, and participants at the meetings of the LACEA Political Economy Group in Cartagena, the AAEP in Buenos Aires, and the Encontro Brasileiro de Econometria in João Pessoa, at the Conference on Monetary and International Economics of La Plata and at seminars at UdeSA, UTDT, UCEMA and UNLP.

Section II briefly reviews the empirical literature on PBC most closely connected to our study. Section III presents the theoretical framework behind this study. Section IV describes the dataset, which draws mainly on the Brender and Drazen (2004) cross-country panel of democracies, and the Henisz (2002) political constraints dataset. Section V presents econometric evidence on electoral budget cycles, isolating the discretionary PBC. Section VI has the conclusions and questions for further research.

II. Empirical literature

There is a rich empirical literature on electoral cycles in fiscal policy. Tufte (1978) provides early evidence on opportunistic fiscal cycles in the United States and other countries. Recently, there has been a wave of empirical work on aggregate PBC using panels of countries. We concentrate on the studies by Shi and Svensson (2002a, 2002b), Persson and Tabellini (2002), and Brender and Drazen (2004), which are the foundation of our research.

We describe these studies below. Briefly stated, Shi and Svensson (2002a, 2002b) find PBC are particularly pronounced in developing countries, relating this to greater corruption and less informed voters. In the subset of democratic countries, Persson and Tabellini (2002) find PBC are stronger in presidential countries and in countries with proportional elections. Brender and Drazen (2004), who also analyze democratic countries, find that new democracies have strong PBC, but in the remnant countries, whether developed or developing, and whatever their form of government, electoral rules, or level of democracy, PBC are not significant.

A. Shi and Svensson

Shi and Svensson (2002b) analyze, for a panel of 91 countries over the 1975-1995 period, the influence of a variable *ele* that takes value 1 in electoral years, and 0 elsewhere. They find that there is a pre-electoral cycle in the fiscal surplus that is much stronger in developing countries: the surplus falls 1.4 percentage points (p.p.) of GDP, against 0.6 p.p. in developed countries. The reason for this difference is not the revenue cycle, which falls 0.3 p.p. in both groups, but rather that spending rises much more strongly in developing countries. They are able to explain these differences across groups of countries in terms of larger rents for incumbents in developing countries, using as proxies either the Transparency International measure of degree of corruption, or an average of five ICRG institutional indicators (rule of law, corruption in

government, quality of the bureaucracy, risk of expropriation of private investment, and risk of repudiation of contracts).

Shi and Svensson (2002a) look at a panel of 123 countries over the 1975-1995 period. Besides the pre-electoral effects captured with *ele*, they look at the combined pre- and post-electoral effects with a variable *pbc* that equals 1 in electoral years, -1 in post-electoral years, and 0 otherwise. The variable *pbc*, which imposes the restriction that the contraction after elections is of the same magnitude as the expansion prior to elections, almost invariably turns out to be more significant in statistical terms than the *ele* variable. They again find that PBC are pervasive, and that cycles are stronger in developing countries: *pbc* has a coefficient of -1.0 in developing countries, and -0.4 in developed countries. They explain the differences in terms of a variable *sum*, a weighted average of two indicators. First, the variable *rents*, an average of the five ICRG indicators mentioned above. The rationale is that low rents (i.e., a higher value of *rents*) indicate smaller incentives to remain in power. Second, the variable *informed voters*, the product of number of radios per capita and a dummy that measures the freedom of broadcasting. The rationale is that a greater proportion of informed voters can reduce the problems of asymmetric information that allow cycles to take place. They find that the composite variable *sum* explains the differences between developing and developed cycles in regard to *ele* (however, they overlook to report the results with *pbc*).

B. Persson and Tabellini

Persson and Tabellini (2002) restrict their panel to 60 democratic countries over the 1960-1998 period. They distinguish between the pre-electoral component of electoral cycles in fiscal policy, *ele*, and the post-electoral component, *ele(+1)*, which takes value 1 in post-electoral years, and 0 elsewhere.

Though they do not test whether the differences are statistically significant, there appears to be a clear asymmetry in government expenditure, which is significantly cut the year after elections, while there is no pattern in the year before elections. On the other hand, tax cuts before elections are followed by similar hikes after elections. This pattern is reflected in the electoral behavior of the budget surplus, which falls 0.1 p.p. of GDP before elections, and rises 0.4 p.p. afterwards. Controlling for the effect of the level of democracy, they find cycles not only in the whole range of democracies (polity index from the Polity IV dataset between 1 and 10), but also in the countries with the best democratic institutions (polity index of 9 or 10).

Persson and Tabellini also analyze the effect of electoral rules and forms of government on PBC. As to electoral rules, they find a statistically significant difference in the case of spending before elections, which tends to fall in majoritarian countries, and to rise in proportional countries (though these effects are not statistically significant in themselves, the difference is). As to the form of government, the differences are more prominent. In presidential countries, the post-electoral effects of a fall in expenditure, and a rise of taxes and surplus, are stronger than in parliamentary countries, and the differences tend to be statistically significant.

C. Brender and Drazen

Brender and Drazen (2004) study a panel of 68 democratic countries over the 1960-2001 period. They concentrate on pre-electoral effects using the *ele* variable. They distinguish between new and old democracies. Countries are new democracies during the first four competitive elections, before becoming established democracies. The idea behind this is that voting may require a local learning process that matures with electoral experience, so the problems of asymmetric information may be alleviated over time.

When all countries are pooled, the electoral effect on the budget surplus of the first four competitive elections is between -1 and -1.2 percentage points of GDP, while the rest of the elections have a negligible effect on the budget surplus. When they partition the data, Brender and Drazen find that PBC are statistically significant in new democracies. On the other hand, old democracies show no evidence of cycles using the *ele* variable, whether in OECD countries or not, and whatever the level of democracy (countries with a polity index between 0 and 9, or an index of 10), the form of government (presidential or parliamentary), or the electoral rules (majoritarian or proportional).

III. Theoretical framework

Two key references on rational electoral cycles are Rogoff (1990) and Lohmann (1998a). They have different implications on the likelihood of PBC, and on the effects of PBC on the probability of reelection. Rogoff (1990) models electoral cycles in fiscal policy building on earlier work by Rogoff and Sibert (1988). Under asymmetric information, he shows that cycles can be interpreted as a signal of the competency of the incumbent. In equilibrium, only competent incumbents engage in PBC, and PBC

increase the probability of reelection. Lohmann (1998a) models electoral cycles in monetary policy. She makes the nice point that even if one abstracts from the signaling problem, there will still be cycles under asymmetric information about the policy process. The underlying issue is a credibility problem, by which the executive cannot credibly commit to not pursue expansionary policy before elections. This credibility problem carries over to fiscal policy. Shi and Svensson (2002a), in a setup that includes government debt, show that the incumbent will have an incentive to raise total expenditure and lower taxes, thereby increasing the budget deficit. In equilibrium, all types of incumbents engage in cycles, so cycles do not increase the probability of reelection.

The standard results on rational PBC not only require asymmetric information, but also a fiscal authority with discretion over fiscal policy; once one drops the assumption of a single fiscal authority, the possibility of PBC will depend on the leeway that the legislature allows the executive in pursuing electoral destabilization (Streb (2003)). This may be empirically relevant, since Alesina, Roubini, and Cohen (1997, chaps. 4 and 6) trace the lack of recent evidence on opportunistic cycles in the United States back to the fact that after 1980 many federal transfer programs have become mandatory by acts of Congress, so they cannot be easily manipulated for short run purposes.

Persson, Roland and Tabellini (1997) sparked off fruitful research on the implications of separation of powers for fiscal policy, but they did not consider its specific implications for PBC. Saporiti and Streb (2004) formally analyze the implications for PBC of considering that in constitutional democracies the process of drafting, revising, approving and implementing the budget requires the concurrence of the legislature.² In a framework of asymmetric information on the budgetary process similar to the Lohmann (1998a) timing, the moderating influence of the legislature is largest when the status quo is given by the previous period's budget. In terms of the time-consistency literature on "rules versus discretion" stemming from Kydland and Prescott (1977), which discusses how to solve the credibility problems faced by policy-makers, separation of powers is needed to make the budget rule credible, i.e., to commit the executive to not doing stimulative policies in electoral periods.

² In the case of monetary policy, Lohmann (1998b) and Drazen (2001) study how the delegation to an independent central bank can moderate electoral cycles. However, a single authority decides fiscal policy.

The interpretation we follow here is that separation of powers has a bite in the fiscal process when the executive and legislative branches are not perfectly aligned. This draws on the insight of Alesina and Rosenthal (1995) on the moderating influence of an opposition legislature. Through the metric of veto players (Tsebelis (2002)), this insight applies not only to divided government in presidential systems, but more generally to coalition governments. Coalition members start to compete among themselves for votes, so it is particularly hard for different political parties to collude close to elections. Given this interpretation, the Saporiti and Streb (2004) model has sharp empirical implications: if there is perfect compliance with the budget law, the budget rule is credible if the party of the executive's leader does not control the legislature.³ On the other hand, if there is imperfect compliance, the budget rule is never credible. Consequently, PBC should be larger either in countries with low legislative checks and balances, or low observance of the rule of law. We explore this conjecture.

IV. Data and Econometric Specification

We use the Brender and Drazen (2004) dataset. Additionally, we resort to the Henisz (2002) POLCON dataset. The precise definitions and sources of the variables used in the regressions are given in Table 1.

Brender and Drazen (2004) compile a panel data set that covers 68 developed and developing democracies, with annual observations for the period between 1960 and 2001. The sample is restricted to years in which the polity index from the Polity IV Project is non-negative, when the country is a democracy with competitive elections. They construct election dates with data from the Institute for Democracy and Electoral Assistance, the International Foundation for Electoral Systems, the Database of Political Institutions (DPI) Version 3, and several other sources.

Brender and Drazen depurate the IMF *International Financial Statistics* (IFS) fiscal series on government surplus, total expenditure, and total revenue and grants, and calculate them as percentage of GDP (drawn from the IFS). They draw on the World Bank *World Development Indicators* for control variables like per capita GDP, GDP growth rates and share of international trade.

³ This is related to the approach in Lohmann (1998b) on the conditions for independent monetary policy in Germany.

Table 1. Definition of Variables

Variable	Description	Source
<i>Texp</i>	Total government expenditure as a percentage of GDP	B&D(2004)
<i>Trg</i>	Total government revenue and grants as a percentage of GDP	B&D(2004)
<i>Bal</i>	Fiscal balance as a percentage of GDP, given by $trg - exp$	B&D(2004)
<i>lngdp_pc</i>	Natural log of GDP per capita	B&D(2004)
<i>Gdpr</i>	Annual growth rate of real GDP	B&D(2004)
<i>Trade</i>	Share of international trade as a percentage of GDP	B&D(2004)
<i>pop65</i>	Fraction of population above 65	B&D(2004)
<i>pop1564</i>	Fraction of population between 15 and 64	B&D(2004)
$\ln(1+pi)$	Natural log of 1 plus the inflation rate	IFS
<i>Polcon3</i>	Political constraints index	H(2002)
<i>vetoplayer</i>	Takes value 1 if $polcon3 \geq 2/3$, and $3/2 * polcon3$ otherwise	O.C.
<i>Law</i>	Law and Order index, combined with the ICRG Rule of Law index in the early years when the former is not available, divided by 6	H(2002) and ICRG
<i>Lawd</i>	Dummy, takes value 1 for country if $law \geq 4$ always, 0 otherwise	O.C.
<i>Checks</i>	Effective veto player, given by $vetoplayer * law$	O.C.
<i>Checksd</i>	Alternative measure of effective veto player, given by $vetoplayer * lawd$	O.C.
<i>Ele</i>	Takes value 1 in election year, 0 otherwise	B&D(2003)
<i>Pbc</i>	ele minus its lead $ele(+1)$, takes value 1 in election year, -1 in the following year, and 0 otherwise	O.C.
<i>pbc_dis</i>	Discretionary component of cycle, given by $pbc * (1 - checks)$	O.C.
<i>pbc_disd</i>	Discretionary component of cycle, given by $pbc * (1 - checksd)$	O.C.
<i>Demo</i>	Takes value 1 if Polity Index ≥ 0 .	B&D(2004)
<i>Oecd</i>	Takes value 1 if country belongs to OECD, 0 otherwise	B&D(2004)
<i>Newd</i>	Takes value 1 if country is new democracy, 0 otherwise	B&D(2004)
<i>Pres</i>	Takes value 1 if form of government is presidential, 0 if parliamentary	B&D(2004)
<i>Prop</i>	Takes value 1 if electoral rule is proportional, 0 if majoritarian	B&D(2004)

Notes: B&D(2003) refers to Brender and Drazen (2003), and similarly for B&D(2004); H(2002), to Henisz (2002); IFS, to the IMF *International Financial Statistics*; O.C., to variables that are our own construction.

From the Henisz (2002) POLCON dataset, we use the political constraints index *polcon3*. This index takes into account the extent of alignment across the executive and

legislative branches of government, and was designed by Henisz (2000) to measure the political constraints facing the executive when implementing a policy. More alignment increases the feasibility of policy change and implies less political constraints for the executive. The minimum is a value of 0, which implies no constraints and absolute political discretion for the executive. As the value of *polcon3* increases, more political constraints are implied. With a single legislative chamber, *polcon3* may reach a maximum of 2/3; while with two chambers the maximum is 4/5, when neither of the chambers is aligned with the executive.⁴

The Henisz (2000) political constraints measure is derived in a spatial model under the assumption that the status quo policy is uniformly distributed over the policy space [0,1]. Instead, based on the approximation that in many countries the status quo policy is given by the previous budget, and the fact that a legislature can prevent PBC provided that the status quo is given by the previous non-electoral year budget (Saporiti and Streb (2004)), our variable of interest is whether a legislative veto player exists or not. Hence, we define a variable *vetoplayer* that rescales *polcon3*, dividing it by 2/3, and which equals 1 for values of *polcon3* equal to 2/3 or more, because values of 2/3 or more imply that the executive faces at least one veto player. In consequence, *vetoplayer* varies in the [0,1] interval.

We do not have a direct measure of adherence to the budget law. Instead, the POLCON dataset reports the ICRG index on Law and Order, which measures the degree of rule of law based on a scale from 0 (low) to 6 (high) characterizing the strength and impartiality of the legal system and the general observance of the law. In earlier years when the Law and Order index is not available, we use instead the ICRG Rule of Law index.⁵ We divide these indices by 6, so *law* varies in the [0,1] interval.

Our measure of effective checks and balances is *checks=vetoplayer*law*, which combines *vetoplayer* with *law* to capture both the legislative checks and balances and the degree of compliance with the law. This is our main variable to condition PBC.

Following the theoretical framework and previous empirical literature on electoral cycles in fiscal policy, a relation between a given fiscal variable y in country i and year t ($y_{i,t}$) and the electoral cycle can be described as follows:

⁴ Henisz (2000, 2002) has another measure of political constraints, *polcon5*, that takes into account whether the country is a federal system or not, and whether the judicial system is independent or not. Federalism might be double-counted there, since it is already included in a second chamber of a legislature (Tsebelis (2002), chap. 8).

⁵ When there are overlapping observations, Rule of Law is an unbiased predictor of Law and Order, since the intercept is zero and the coefficient is 1. Therefore, we use the more recent series on Law and Order, supplementing it with Rule of Law when the former has missing observations.

where $E_{i,t}$ is a dummy election variable, $x_{i,t}$ is a vector of m controls, $z_{i,t}$ is a proxy

$$y_{i,t} = \sum_{j=1}^k \beta_j y_{i,t-j} + \sum_{j=1}^m \gamma_j x_{j,i,t} + \delta_E E_{i,t} + \lambda z_{i,t} + \eta z_{i,t} E_{i,t} + \varphi z_{i,t} y_{i,t-1} + \mu_i + \varepsilon_{i,t} \quad , \quad (1)$$

variable for effective checks and balances conditioning the electoral policy

manipulations, μ_i is a specific country effect, and the term $\varepsilon_{i,t}$ is a random error that is assumed i.i.d. This specification represents a dynamic panel model, where the dependent variable is a function of its own lagged levels, a set of controls and the electoral timing conditioned by effective checks and balances.

Estimates are performed using two methods, Fixed Effects (FE) and Generalized Method of Moments (GMM) for dynamic models of panel data using the procedure developed by Arellano and Bond (1991).

V. Empirical Evidence

We now turn to the evidence on aggregate PBC, focusing on the budget surplus. We first introduce effective checks and balances, to isolate the influence of discretionary executive power on PBC. We then look at the sensitivity of the results when restricted to developed or developing countries. To make sure the impact of executive discretion on electoral cycles is not driven by a larger degree of uninformed and inexperienced voters, we then contrast, at one corner, developed countries that are established democracies with, at the other, less developed countries that are new democracies. Finally, we partition these subsets according to form of government and electoral rules.

We use the same control variables as Brender and Drazen (2004), except for the use the growth rate of real GDP to control for cyclical effects.⁶ We additionally control for the effect of inflation and its square, $\ln(1+pi)$ and $\ln(1+pi)sq$, to account for issues like lack of indexation of tax bases and tax collection lags. We exclude Sweden from the sample, due to a jump in the fiscal series in the early 1990s, so our panel is reduced to 67 countries (see Table 2). The data is annual, though monthly data would be ideal, since the estimates with annual data are downward biased and may lead to underestimate the size of PBC.⁷ Descriptive statistics are presented in Table 3.

⁶ The use of the output gap measured with the Hodrick-Prescott filter does not affect the results. Since a lagged budget surplus term is included, this captures the negative effects of low growth (and hence a recession with below-trend output) on future budget surpluses.

⁷ As Akhmedov and Zhuravskaya (2004) show for Russia, the effects of PBC are strongest in the months closest to elections, and shifts of opposite sign in fiscal policies around elections partly cancel out with low frequency (quarterly or annual) data. In the Latin American environment where inflation is a means of taxation, Stein, Streb and Ghezzi (2004) also find that the manipulation of nominal exchange rate policy follows a short-run PBC, where on average the changes are concentrated in the four months up to elections, and the four months that follow (the

Table 2. Country Characteristics

Country	<i>oeed</i>	<i>newdpres</i>	<i>prop</i>	Years with <i>demo</i> ≥ 0	<i>checks</i>	<i>checks_d</i>	
Argentina	0	1	1	1	1973-75; 83-2001	0.41	0.00
Australia	1	0	0	1	1960-2001	0.74	0.71
Austria	1	0	0	1	1960-2001	0.64	0.64
Belgium	1	0	0	1	1960-2001	0.98	0.89
Bolivia	0	1	1	1	1982-2001	0.22	0.00
Brazil	0	1	1	1	1960-63; 85-2001	0.23	0.00
Bulgaria	0	1	1	1	1990-2001	0.47	0.59
Canada	1	0	0	0	1960-2001	0.64	0.63
Chile	0	1	1	0	1960-72; 89-2001	0.56	0.62
Colombia	0	0	1	1	1960-2001	0.15	0.00
Costa Rica	0	0	1	1	1960-2001	0.38	0.56
Cyprus	0	0	1	1	1960-62; 68-2001	0.33	0.00
Czech Rep.	0	1	1	1	1990-2001	0.73	0.78
Denmark	1	0	0	1	1960-2001	0.80	0.79
Dominican Rep.	0	1	1	1	1978-2001	0.36	0.00
Ecuador	0	1	1	1	1960; 68-71; 79-2001	0.24	0.00
El Salvador	0	1	1	1	1984-2001	0.21	0.00
Estonia	0	1	1	1	1991-2001	0.49	0.00
Fiji	0	1	1	0	1970-86; 90-99	n.a.	n.a.
Finland	1	0	0	1	1960-2001	0.81	0.81
France	1	0	0	1	1960-2001	0.56	0.59
Germany	1	0	0	1	1960-2001	0.60	0.61
Greece	1	1	1	1	1960-66; 75-2001	0.37	0.00
Guatemala	0	1	1	1	1966-73; 86-2001	0.15	0.00
Honduras	0	1	1	1	1982-2001	0.18	0.00
Hungary	0	1	0	1	1990-2001	0.63	0.70
Iceland	1	0	0	1	1960-2001	0.77	0.75
India	0	0	0	1	1960-2001	0.35	0.00
Ireland	1	0	0	1	1960-2001	0.54	0.64
Israel	0	0	1	1	1960-2001	0.43	0.00
Italy	1	0	0	1	1960-2001	0.66	0.74
Japan	1	0	0	1	1960-2001	0.75	0.77
Korea	0	1	0	1	1960; 63-71; 88-2001	0.35	0.00
Lithuania	0	1	1	1	1991-2001	0.47	0.64
Luxembourg	1	0	0	1	1960-2001	0.74	0.73
Madagascar	0	1	1	1	1992-2001	0.37	0.00
Malaysia	0	0	0	0	1960-2001	0.31	0.00
Mali	0	1	1	0	1992-2001	0.20	0.00
Mauritius	0	0	0	0	1968-2001	n.a.	n.a.
Mexico	0	1	1	1	1988-2001	0.23	0.00
Nepal	0	1	0	0	1990-2001	n.a.	n.a.
Netherlands	1	0	0	1	1960-2001	0.73	0.79
New Zealand	1	0	0	1	1960-2001	0.53	0.52
Nicaragua	0	1	1	1	1990-2001	0.30	0.00

exchange rate becomes 3% more appreciated than average in the run-up to presidential elections and 3% more depreciated after, because the government first steps down on the monthly rate of depreciation and then releases it).

Table 2. Country Characteristics (Cont.)

Country	<i>oecd newdpres prop</i>				Years with <i>demo</i> ≥ 0	<i>checks</i>	<i>checks</i> <i>sd</i>
Norway	1	0	0	1	1960-2001	0.73	0.72
Pakistan	0	1	0	0	1962-68; 73-76; 88-98	0.24	0.00
Panama	0	1	1	1	1960-67; 89-2001	0.18	0.00
Papua	0	0	0	1	1975-2001	0.46	0.00
Paraguay	0	1	1	1	1989-2001	0.31	0.00
Peru	0	1	1	1	1960-67; 80-99	0.15	0.00
Philippines	0	1	1	0	1960-71; 87-2001	0.24	0.00
Poland	0	1	1	1	1989-2001	0.39	0.46
Portugal	1	1	1	1	1976-2001	0.54	0.63
Romania	0	1	0	1	1990-2001	0.47	0.00
Russia	0	1	1	1	1992-2001	0.07	0.00
Slovakia	0	1	0	1	1993-2001	0.69	0.76
Slovenia	0	1	0	1	1991-2001	0.68	0.79
South Africa	0	0	1	1	1960-91; 94-2001	0.21	0.00
Spain	1	1	1	1	1978-2001	0.56	0.71
Sri Lanka	0	0	1	1	1960-2001	0.14	0.00
Switzerland	1	0	0	1	1960-2001	0.54	0.58
Trinidad	0	0	0	0	1962-2001	0.42	0.60
Turkey	1	1	1	1	1961-70; 73-79; 83-2001	0.38	0.00
UK	1	0	0	0	1960-2001	0.48	0.53
US	1	0	1	0	1960-2001	0.61	0.59
Uruguay	0	1	1	1	1960-70; 85-2001	0.39	0.00
Venezuela	0	0	1	1	1960-2001	0.33	0.00
Total	23	36	37	55		0.45	0.33

Notes: n.a. stands for not available; *checks* and *checks**sd* are computed for years with *demo* ≥ 0 .

Table 3. Descriptive Statistics

	OECD countries					Non-OECD countries					Total				
	<i>I</i>	<i>bal</i>	<i>texp</i>	<i>trg</i>	<i>checks</i>	<i>I</i>	<i>bal</i>	<i>texp</i>	<i>trg</i>	<i>checks</i>	<i>I</i>	<i>bal</i>	<i>texp</i>	<i>trg</i>	<i>checks</i>
Old	19	-1.8	29.7	28.2	0.68	12	-2.8	25.7	22.6	0.32	31	-2.1	28.2	26.1	0.55
democracies		(3.6)	(10.3)	(9.5)	(0.15)		(4.6)	(11.0)	(9.9)	(0.18)		(4.0)	(10.7)	(10.0)	(0.23)
New	4	-5.1	27.9	22.9	0.47	32	-1.9	22.4	20.6	0.32	36	-2.4	23.4	21.0	0.34
democracies		(3.2)	(13.2)	(11.9)	(0.14)		(2.9)	(9.9)	(9.4)	(0.21)		(3.2)	(10.7)	(9.9)	(0.20)
Total	23	-2.2	29.5	27.5	0.64	44	-2.3	23.9	21.5	0.32	67	-2.2	26.6	24.4	0.45
		(3.7)	(10.7)	(10.0)	(0.17)		(3.8)	(10.5)	(9.7)	(0.20)		(3.8)	(11.0)	(10.3)	(0.24)

Note: *I* refers to number of countries in each group; standard deviation reported in parenthesis below mean values.

A. Effective Checks and Balances and Discretionary Component of PBC

We look at the influence of electoral cycles on the behavior of the budget surplus as a percentage of GDP, *bal*. We concentrate on the electoral dummy *pb**c*, which takes value

1 in electoral years, -1 in post-electoral years, and 0 otherwise. This variable is meant to capture both pre and post-electoral effects, following the approach in Shi and Svensson (2002a). It is constructed with the *ele* variable in Brender and Drazen (2003), which only takes elections when the polity index is non-negative, combined with its lead, *ele(+1)*.⁸

Persson and Tabellini (2002) remark that pre and post electoral effects may differ, so we first check if the restriction that the coefficient estimate of *ele* is equal to the coefficient estimate of minus *ele(+1)* is not rejected by the data.

Column (1) of Table 4 shows that the restriction that the pos-electoral contraction in the budget surplus as a percentage of GDP (*bal*) is of the same size as the pre-electoral expansion is not rejected by the annual data. We can interpret the effect of PBC as short-run displacements: the surplus falls below its trend, and then jumps above it, if expenditures are speeded up, and taxes postponed, around elections.

Column (2) of Table 4 shows that the electoral cycle measured by the *pb*c dummy variable shows a fall of 0.3 p.p. of GDP in the surplus before elections, and an equivalent rise after elections. The pattern observed by Shi and Svensson (2002a,b) that electoral cycles are stronger in developing countries appears here, though the difference is not statistically significant.⁹

Column (5) tests whether effective checks and balances (*checks*=*vetoplayer*law*) have a moderating influence on PBC, i.e., whether the coefficient estimate of the compound variable *pb*c_*checks*=*pb*c**checks* shows the theoretically expected positive sign. We also use in column (3) an alternative measure *pb*c_*checks*d=*pb*c**checks*d, where we define a dummy variable *law*d that takes value 1 if *law* is larger than 4 in all years that are reported for a given country, and 0 otherwise, so *checks*d=*vetoplayer*law*d. This treatment implies treating rule of law as a fixed characteristic, so each country has either low or high rule of law. This has the advantage of extending the available data to the whole period, since the data on rule of law is only available since 1982. The disadvantage is losing the variation over time of rule of law. Columns (3) and (5) of Table 4 show that either version of effective checks and balances moderate PBC, though in column (5) they do not have a significant influence by themselves (the probability value is 0.113).

⁸ Brender and Drazen (2004) adjust the election years in several countries, based on the difference between fiscal and calendar year. We prefer to stick to the original election dates in Brender and Drazen (2003).

⁹ Dividing *pb*c in column (2) of Table 1 into *pb*c_*oecd*=*pb*c**oecd* and *pb*c_*noecd*=*pb*c*(1-*oecd*), the coefficients are -0.214 ($t=-2.14$) and -0.401 ($t=-3.60$). With p-value 0.2118, an F-test cannot reject the equality of both coefficients.

Table 4. Discretionary PBC in All Democracies

Dependent variable: <i>bal</i>	(1)	(2)	(3)	(4)	(5)	(6)
<i>bal(-1)</i>	0.613 (31.57)***	0.613 (31.57)***	0.615 (30.74)***	0.615 (30.76)***	0.469 (16.17)***	0.469 (16.19)***
<i>Lngdp_pc</i>	0.463 (1.37)	0.475 (1.40)	0.578 (1.64)*	0.578 (1.64)*	0.400 (0.55)	0.406 (0.56)
<i>gdpr</i>	0.091 (5.01)***	0.091 (4.99)***	0.106 (5.57)***	0.106 (5.57)***	0.107 (4.27)***	0.107 (4.27)***
<i>trade</i>	0.003 (0.62)	0.003 (0.61)	0.001 (0.24)	0.001 (0.24)	0.010 (1.08)	0.010 (1.07)
<i>Pop65</i>	-0.031 (-0.39)	-0.034 (-0.44)	-0.062 (-0.76)	-0.062 (-0.76)	0.341 (1.85)*	0.341 (1.85)*
<i>Pop1564</i>	0.037 (0.98)	0.037 (0.99)	0.036 (0.87)	0.036 (0.87)	0.013 (0.15)	0.014 (0.16)
<i>ln(1+pi)</i>	1.504 (2.54)**	1.499 (2.53)**	1.612 (2.68)***	1.607 (2.67)***	1.555 (2.13)**	1.545 (2.12)**
<i>ln(1+pi)sq</i>	-0.095 (-0.55)	-0.091 (-0.53)	-0.107 (-0.62)	-0.105 (-0.61)	-0.187 (-0.93)	-0.184 (-0.91)
<i>checks</i>					-0.975 (-1.42)	-0.978 (-1.43)
<i>checksd</i>			-0.043 (-0.05)	-0.032 (-0.04)		
<i>Ele</i>	-0.223 (-1.75)*					
<i>ele(+1)</i>	0.371 (2.92)***					
<i>Pbc</i>		-0.297 (-3.99)***	-0.465 (-3.91)***		-0.793 (-3.30)***	
<i>Pbc_checks</i>					0.700 (1.59)	
<i>Pbc_checksd</i>			0.398 (1.82)*			
<i>Pbc_dis</i>						-0.851 (-4.61)***
<i>Pbc_disd</i>				-0.483 (-4.30)***		
<i>constant</i>	-8.065 (-2.12)*	-8.105 (-2.13)**	-8.622 (-2.06)**	-8.633 (-2.06)**	-10.236 (-1.30)	-10.345 (-1.32)
Method of estimation	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects
R ² within	0.4822	0.4820	0.4850	0.4849	0.3547	0.3546
R ² between	0.8577	0.8577	0.8601	0.8600	0.2907	0.2891
R ² overall	0.6533	0.6534	0.6589	0.6590	0.3131	0.3118
No. countries	67	67	64	64	64	64
No. observations	1575	1575	1488	1488	860	860
p-value F-test:						
<i>ele = -ele(+1)</i>	0.4733	-	-	-	-	-
<i>Pbc = -</i>	-	-	-	-	0.7061	-
<i>pbc_checks</i>	-	-	-	-	-	-
<i>pbc = -pbc_checksd</i>	-	-	0.6538	-	-	-

Notes: t statistics in parentheses; * significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level. To control for time effects, dummies are included for each five-year period from 1960-64 to 1995-99, while the years 2000-01 are the base level. These coefficients are not reported.

Our main interest is in the net effect of checks and balances, given our conjecture that at least one effective veto player will prevent PBC. Specifically, the variable that isolates what can be called the discretionary component of cycles is $pbc_dis = pbc \cdot (1 - checks)$, or $pbc_disd = pbc \cdot (1 - checksd)$. The discretionary component of

PBC is an adjustment that implies, at one extreme, that if the legislature is perfectly aligned with the executive (*vetoplayer*=0), or if the observance of rule of law is very low (*law*=0, *lawd*=0), the original *pbc* variable is unchanged. At the other extreme, if the legislature is not aligned with the executive and constitutes a veto player (*vetoplayer*=1), and there is a high value of rule of law (*law*=1, *lawd*=1), an election year would not be counted as such because the electoral cycle would be completely counteracted by the legislative checks and balances.

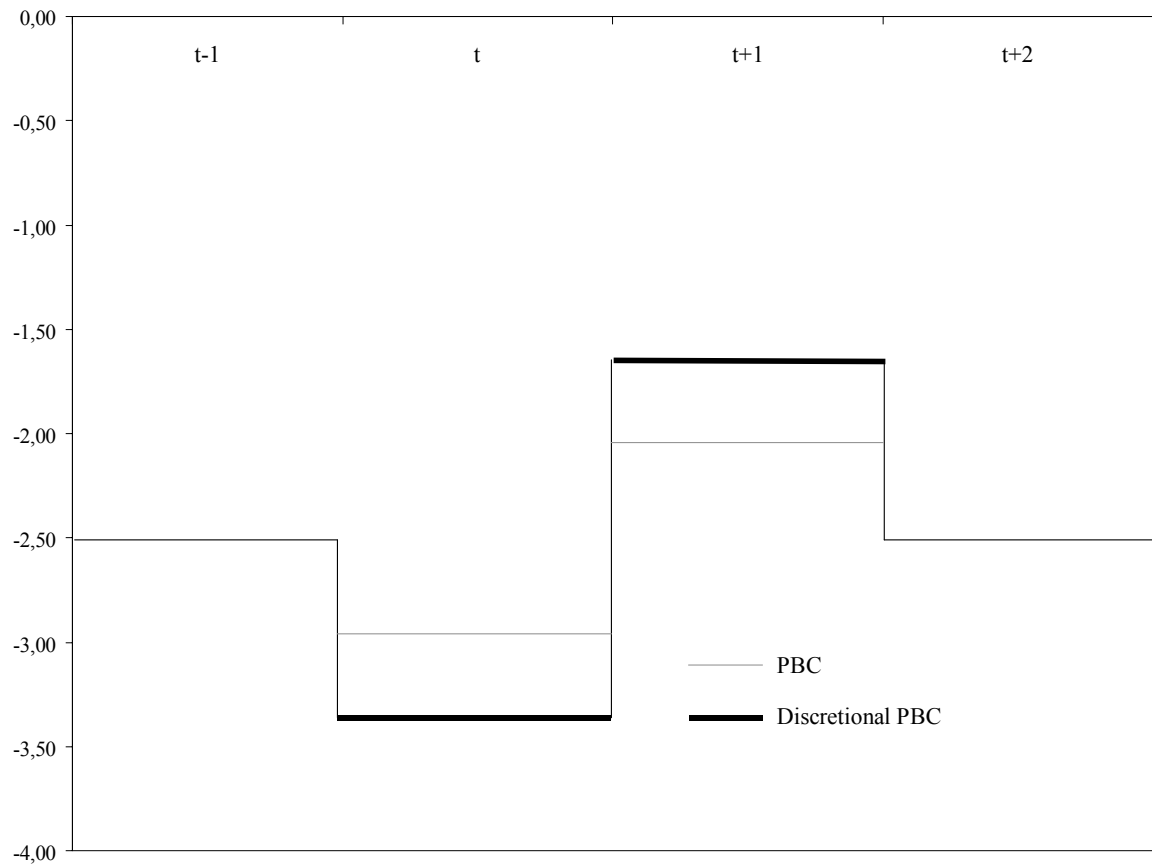
Given that the coefficients of *pbc* and *pbc_checks* (*pbc_checksd*) are of the similar magnitude but opposite sign, we formally test the hypothesis that the coefficient of *pbc* is equal to minus the coefficient of *pbc_checks* (*pbc_checksd*). The F-tests in Table 4 do not allow to reject this.¹⁰ Columns (4) and (6) present the estimates with the discretionary component of cycles.

The effects of discretionary PBC are significant at the 1% level, as are those of standard PBC in column (2). However, once we isolate the discretionary component, the estimated impact is larger for a country with no effective checks and balances: in contrast to the base estimate of 0.30 p.p. of GDP using *pbc*, the effect is 0.48 p.p. of GDP according to *pbc_disd*, and 0.85 p.p. of GDP according to *pbc_dis*. Part of the difference is due to different time periods: when *pbc_disd* in (4) is restricted to the same period as (6), the coefficient rises to 0.66 p.p. of GDP. As to the remainder, *pbc_disd* captures average rather than marginal effects, showing the influence of political constraints with switch from a low rule of law to a high rule of law country. In what follows we focus on *pbc_dis*.

There are elections on average every four years. Figure 1 depicts the time path around a year of elections *t* of the average budget surplus implied by *pbc* (-0.45) and by the discretionary component *pbc_dis* (-0.85), around the mean value of *bal*= -2.50 in the 1982-2001 period (a common set of observations are used for comparability).

¹⁰ This also avoids multicollinearity, given the pair-wise correlation of 0.90 between *pbc* and *pbc_checks*, and 0.77 between *pbc* and *pbc_checksd*.

Figure 1. Time Path of Budget Balance around Elections



B. OECD and non-OECD countries

Our aim now is to review the Schuknecht (1996) conjecture that stronger PBC in developing countries might be related to weaker checks and balances there.

Effective checks and balances are indeed smaller in developing countries: *checks* equals 0.32 in non-OECD countries, compared to 0.64 in OECD countries (see Table 3). Consequently, discretionality is larger in non-OECD countries, which implies stronger PBC in non-OECD countries: multiplying the average degree of discretionality in each group by the coefficient estimate in column (6) of Table 4 implies that PBC in developing countries have an impact of -0.6 p.p. of GDP in non-OECD countries, against -0.3 p.p. of GDP in OECD countries. This agrees with Shi and Svensson (2002a, b), though the channel is that conjectured by Schuknecht (1996): larger checks and balances moderate cycles in developed countries.

Table 5. Discretionary PBC in OECD Countries

Dependent variable: <i>bal</i>	(1)	(2)	(3)	(4)	(5)	(6)
<i>bal(-1)</i>	0.781 (35.26)***	0.781 (35.29)***	0.783 (35.38)***	0.783 (35.41)***	0.777 (23.97)***	0.777 (23.99)***
<i>lngdp_pc</i>	0.672 (1.35)	0.675 (1.36)	0.678 (1.36)	0.678 (1.37)	0.375 (0.41)	0.384 (0.42)
<i>Gdpr</i>	0.153 (6.28)***	0.153 (6.29)***	0.153 (6.31)***	0.154 (6.31)***	0.198 (4.81)***	0.199 (4.84)***
<i>Trade</i>	-0.004 (-0.55)	-0.004 (-0.55)	-0.004 (-0.57)	-0.004 (-0.57)	-0.002 (-0.11)	-0.002 (-0.12)
<i>pop65</i>	-0.023 (-0.30)	-0.023 (-0.31)	-0.026 (-0.34)	-0.026 (-0.34)	0.229 (1.42)	0.228 (1.42)
<i>pop1564</i>	0.028 (0.61)	0.028 (0.61)	0.028 (0.60)	0.278 (0.60)	-0.956 (-0.81)	-0.097 (-0.82)
<i>ln(1+pi)</i>	-2.882 (-1.24)	-2.886 (-1.24)	-2.818 (-1.22)	-2.828 (-1.22)	-2.947 (-0.76)	-2.999 (-0.77)
<i>ln(1+pi)sq</i>	2.710 (0.68)	2.721 (0.68)	2.614 (0.66)	2.629 (0.66)	5.025 (0.88)	5.161 (0.90)
<i>checks</i>					-0.441 (-0.50)	-0.456 (-0.51)
<i>checksd</i>			-0.116 (-0.16)	-0.120 (-0.17)		
<i>ele</i>	-0.225 (-1.87)*					
<i>ele(+1)</i>	0.246 (2.03)**					
<i>pbc</i>		-0.236 (-3.40)***	-0.703 (-3.15)***		-1.250 (-3.03)***	
<i>pbc_checks</i>					1.396 (2.25)**	
<i>pbc_checksd</i>			0.730 (2.20)**			
<i>pbc_dis</i>						-1.061 (-4.02)***
<i>pbc_disd</i>				-0.675 (-4.06)***		
<i>constant</i>	-8.238 (-1.42)	-8.248 (-1.42)	-8.217 (-1.42)	-8.218 (-1.42)	-0.549 (-0.05)	-0.518 (-0.05)
Method of estimation	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects
R ² within	0.7348	0.7348	0.7366	0.7366	0.7100	0.7097
R ² between	0.9821	0.9820	0.9824	0.9824	0.9770	0.9771
R ² overall	0.8466	0.8465	0.8474	0.8474	0.8504	0.8503
No. countries	23	23	23	23	23	23
No. observations	779	779	779	779	387	387
p-value F-test:						
<i>ele = -ele(+1)</i>	0.9152	-	-	-	-	-
<i>pbc = -pbc_checks</i>	-	-	-	-	0.5519	-
<i>pbc = -pbc_checksd</i>	-	-	0.8461	-	-	-

Notes: t statistics in parentheses; * significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level. To control for time effects, dummies are included for each five-year period from 1960-64 to 1995-99, while the years 2000-01 are the base level. These coefficients are not reported.

Table 6. Discretionary PBC in Non-OECD Countries

Dependent variable: <i>bal</i>	(1)	(2)	(3)	(4)	(5)	(6)
<i>bal(-1)</i>	0.483 (15.90)***	0.482 (15.91)***	0.472 (14.43)***	0.472 (14.45)***	0.199 (4.62)***	0.205 (4.75)***
<i>lngdp_pc</i>	0.021 (0.04)	0.033 (0.07)	0.172 (0.34)	0.173 (0.34)	-0.093 (-0.09)	-0.016 (-0.02)
<i>Gdpr</i>	0.057 (2.21)***	0.056 (2.18)**	0.069 (2.44)**	0.069 (2.44)**	0.056 (1.80)*	0.056 (1.80)*
<i>Trade</i>	0.010 (1.43)	0.010 (1.43)	0.010 (1.22)	0.010 (1.22)	0.029 (2.50)**	0.277 (2.37)**
<i>pop65</i>	0.301 (1.36)	0.297 (1.34)	0.188 (0.74)	0.186 (0.74)	-0.190 (-0.33)	-0.222 (-0.39)
<i>pop1564</i>	0.021 (-0.34)	0.021 (0.34)	0.027 (0.40)	0.027 (0.41)	-0.071 (-0.44)	-0.048 (-0.30)
<i>ln(1+pi)</i>	1.219 (1.57)	1.209 (1.56)	1.322 (1.64)	1.325 (1.64)	0.079 (0.09)	0.083 (0.09)
<i>ln(1+pi)sq</i>	-0.133 (-0.61)	-0.128 (-0.59)	-0.155 (-0.68)	-0.155 (-0.68)	-0.010 (-0.04)	-0.004 (-0.02)
<i>Checks</i>					-0.698 (-0.75)	-0.756 (-0.81)
<i>Checksd</i>			0.970 (0.48)	0.972 (0.48)		
<i>Ele</i>	-0.270 (-1.21)					
<i>ele(+1)</i>	0.439 (1.97)*					
<i>Pbc</i>		-0.355 (-2.67)***	-0.396 (-2.54)**		-0.317 (-0.96)	
<i>pbc_checks</i>					-0.631 (-0.75)	
<i>pbc_checksd</i>			0.191 (0.34)			
<i>pbc_dis</i>						-0.697 (-3.00)***
<i>pbc_disd</i>				-0.405 (-2.61)***		
<i>Constant</i>	-6.066 (-1.19)	-6.102 (-1.20)	-7.179 (-1.23)	-7.211 (-1.24)	0.657 (0.06)	-0.940 (-0.08)
Method of estimation	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects
R ² within	0.3232	0.3230	0.3082	0.3080	0.1933	0.1882
R ² between	0.2728	0.2764	0.3275	0.3310	0.2812	0.3196
R ² overall	0.3489	0.3504	0.3743	0.3757	0.2279	0.2438
No. countries	44	44	41	41	41	41
No. observations	796	796	709	709	473	473
p-value F-test:						
<i>ele = - ele(+1)</i>	0.6379	-	-	-	-	-
<i>pbc = - pbc_checks</i>	-	-	-	-	0.1029	-
<i>pbc = - pbc_checksd</i>	-	-	0.6957	-	-	-

Notes: t statistics in parentheses; * significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level. To control for time effects, dummies are included for each five-year period from 1960-64 to 1995-99, while the years 2000-01 are the base level. These coefficients are not reported.

Discretionary PBC do not give the slightest hint that cycles are different in OECD and non-OECD countries.¹¹ However, as a sensitivity test we divide the estimates into

¹¹ For example, breaking down *pbc_dis* in column (6) of Table 1 into *pbc_dis_oecd*=*pbc_dis***oecd* and *pbc_dis_noecd*=*pbc_dis**(1-*oecd*), the coefficients are -0.856 (t=-2.31) and -0.850 (t=-3.99). With p-value 0.9875, an F-test cannot reject the hypothesis that both coefficients are identical.

separate two groups. Table 5 shows the same estimates as Table 1, but restricted to OECD countries, while Table 6 shows non-OECD countries.

The results for OECD countries basically parallel those for the complete set of democracies. The differences appear for non-OECD countries, where our measures of legislative checks and balances in columns (3) and (5) of Table 6 are not statistically significant. Part of the explanation might be a threshold effect: when using the variable $checksd=vetoplayer*lawd$, this variable mostly takes value zero in non-OECD countries because they have low values for rule of law, so a veto player might be ineffective unless a certain minimum threshold of rule of law is surpassed.

What is clear from Tables 5 and 6 is that the model works a lot better in OECD countries, in great extent due to the larger persistence of the budget surplus in OECD countries, which makes it more predictable.

C. Budget Balance, Expenditures and Revenues: Persistence Effects

Besides the direct effect of checks and balances on the level of electoral cycles, we additionally consider their indirect effect on persistence, given the observation in Tsebelis (2002) that more veto players should lead to more persistence of policies.

Column (1) of Table 7 only differs from column (6) of Table 4 in the inclusion of the lagged term $bal(-1)_{dis}=bal(-1)*(1-checks)$, where the past surplus interacts with the current degree of discretionary power. More discretion (less effective checks and balances) decreases the persistence of budget surpluses (deficits), which can help explain why the budget balance in OECD countries has more persistence than non-OECD countries. As in the literature on the costs of coalition governments and divided government, this might be an indication of how checks and balances can make it harder for governments to adjust the budget surplus (or deficit) in any year. However, this specific variable does not explain the variation within the group of OECD or non-OECD countries, something that is discussed further in the next Sub-section.

Table 7. Discretionary PBC in All Democracies: Budget Balance, Expenditure and Revenues

Dependent variable: y	y = bal		y = exp		y = trg	
	(1)	(2)	(3)	(4)	(5)	(6)
y(-1)	0.791 (11.87)***	1.099 (9.11)***	0.720 (9.61)***	0.995 (9.15)***	0.569 (6.71)***	0.440 (5.48)***
Lngdp_pc	0.342 (0.48)	-1.74 (-0.61)	-1.737 (-1.21)	-0.474 (-0.21)	-2.091 (-1.63)	-1.379 (-0.80)
Gdpr	0.105 (4.25)***	0.159 (4.35)***	-0.111 (-2.26)**	-0.118 (-4.22)***	0.003 (0.06)	0.004 (0.20)
Trade	0.016 (1.75)*	0.027 (1.56)	0.014 (0.74)	0.041 (3.30)	0.002 (0.12)	0.017 (1.97)**
pop65	0.258 (1.42)	2.988 (0.96)	1.422 (3.89)***	5.615 (1.70)*	1.859 (5.61)***	-0.120 (-0.07)
pop1564	0.058 (0.66)	0.101 (0.11)	0.181 (1.02)	-0.685 (-0.96)	0.157 (0.98)	-2.481 (-3.22)***
ln(1+pi)	1.096 (1.52)	3.996 (2.21)**	-2.245 (-1.57)	-1.540 (-0.80)	0.044 (0.03)	-2.681 (-2.23)**
ln(1+pi)sq	-0.141 (-0.71)	-0.862 (-2.32)**	0.396 (1.00)	0.401 (1.09)	0.139 (0.39)	0.770 (2.26)**
Checks	0.326 (0.46)	2.835 (1.94)*	-1.088 (-0.37)	-18.201 (-3.40)***	0.825 (0.29)	-0.895 (-0.26)
pbcdis	-0.787 (-4.33)***	-0.697 (-6.01)***	0.645 (1.79)*	0.637 (2.51)**	-0.231 (-0.71)	-0.682 (-1.75)*
y(-1)_dis	-0.604 (-5.34)***	-1.041 (-5.30)***	-0.007 (-0.07)	-0.645 (4.14)***	0.093 (0.93)	-0.053 (-0.48)
Constant	-12.783 (-1.65)*	-0.176 (-0.71)	-0.979 (-0.06)	0.096 (0.57)	1.843 (0.13)	0.380 (2.79)***
Method of estimation	Fixed-effects	Arellano-Bond Two-Step	Fixed-effects	Arellano-Bond Two-Step	Fixed-effects	Arellano-Bond Two-Step
R ² within	0.3773	-	0.3006	-	0.2228	-
R ² between	0.3392	-	0.8497	-	0.7606	-
R ² overall	0.3557	-	0.7621	-	0.6867	-
Sargan Test ^b	-	1.000	-	1.000	-	1.0000
2 nd Order Serial Correlation Test ^c	-	0.1107	-	0.9598	-	0.6452
No. countries	64	62	64	62	64	62
No. observations	860	725	868	733	860	725

Notes: For fixed effects estimates, t statistics in parentheses; * significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level. To control for time effects, dummies are included for each five-year period from 1980-84 to 1995-99, while the years 2000-01 are the base level. These coefficients are not reported. For GMM estimates, z statistics in parentheses; * significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level. The instruments used in GMM regressions are two lags of the dependent variable and one lag of covariates. Reported coefficients correspond to the lagged first difference of the dependant variable (second lag not reported) and the first difference of covariates (lagged differences not reported). All instruments are treated as strictly exogenous. (a) Using heteroskedastic-consistent estimator of the variance-covariance matrix of the parameter estimates. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. In one-step estimations p-values come from the one step homoskedastic estimator. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

To track the possible sources of discretionary PBC in the budget surplus, columns (3) and (5) show the results with total expenditure (*texp*) and total revenue and grants (*trg*) as dependant variables. The discretionary PBC cycle is related to a tendency of expenditure to go up, and revenues to go down, in election years (a pattern that is reversed after elections). These effects are not always statistically significant by themselves in the FE estimates. However, it is clear that their combined effect leads to a significant electoral cycle in the budget in column (1). In this sense, the budget surplus is a more sensitive indicator of PBC than its components.

All the regressions so far use country fixed effects. The use of fixed effects estimators in a regression with lagged dependent variables, as in our case with *bal(-1)*, introduces a potential bias. Since the order of the bias is $1/T$, where T is the length of the panel, we expect a small bias.¹² The Arellano-Bond procedure addresses this bias. Nevertheless, it makes use of the lagged values of the explanatory variables as instruments, and this reduces the set of observations. To make sure the estimates are robust to different econometric methodologies, results from the two methods are reported.

Columns (2), (4) and (6) were carried out with the GMM estimator using the Arellano-Bond procedure. We used the one step heteroskedastic-consistent estimator of the variance-covariance matrix of the parameter estimates and the two-step estimator, presenting the best results according to the Sargan test and the second order correlation test. The results from the GMM estimates confirm the results from FE estimates for the relevant variables. Indeed, the GMM estimates of Table 7 show more clearly that effective checks and balances have significant impacts on PBC in the budget surplus, expenditures and revenues. On the other hand, effective checks and balances do not affect the persistence of revenues, only of the budget surplus and expenditures.

D. Rich Established Democracies and Poor Young Democracies

Given the fact that voters in established democracies might behave as fiscal conservatives that punish deficit spending, Brender and Drazen (2004) designed a filter variable *newd* to take into account whether a country is a new democracy or not. We classify a country as *newd* if, according to Brender and Drazen (2004), any of the elections in the sample period belongs to the first four competitive elections.

¹² There are on average between 13 and 14 observations per country in columns (1), (3) and (5).

The fact that new democracies have lower effective checks and balances than old democracies (*checks* of 0.34 vs. 0.55) can help explain the results in Brender and Drazen (2004) on PBC being significant in new democracies. Taking into account *newd*, we present estimates for the two most typical groups: OECD countries that are established democracies (19 out of 23 OECD countries fall into that category) and non-OECD countries that are new democracies (32 out of 44 non-OECD countries).

Table 8. Discretionary PBC in Established OECD Democracies

Dependent variable:	Total	Total	Parliamentary	Proportional	Majoritarian
<i>bal</i>	(1)	(2)	(3)	(4)	(5)
<i>bal(-1)</i>	0.716 (10.75)***	0.774 (5.29)***	0.716 (10.54)***	0.722 (10.14)***	0.767 (1.74)*
<i>lngdp_pc</i>	-0.104 (-0.12)	0.573 (0.17)	-0.447 (-0.48)	-0.437 (-0.43)	-0.044 (-0.02)
<i>gdpr</i>	0.231 (5.12)***	0.235 (3.15)***	0.235 (4.98)***	0.238 (4.66)***	0.228 (2.02)*
<i>trade</i>	-0.002 (-0.16)	-0.020 (-0.83)	-0.007 (-0.48)	-0.017 (-0.90)	-0.011 (-0.18)
<i>pop65</i>	0.517 (2.90)***	0.483 (0.61)	0.566 (3.07)***	0.534 (2.75)***	1.658 (1.48)
<i>pop1564</i>	0.021 (0.14)	0.223 (0.58)	0.092 (0.59)	0.143 (0.80)	-0.642 (-0.75)
<i>ln(1+pi)</i>	4.719 (1.17)	8.600 (0.99)	4.732 (1.14)	4.224 (0.97)	21.956 (0.69)
<i>ln(1+pi)sq</i>	-9.611 (-1.33)	-16.442 (-1.50)	-9.421 (-1.28)	-8.557 (-1.10)	-30.952 (-0.10)
<i>checks</i>	-0.954 (-0.93)	-0.373 (-0.25)	-0.816 (-0.78)	-0.833 (-0.72)	-9.359 (-2.11)**
<i>pbc_dis</i>	-0.817 (-2.89)***	-0.749 (-3.32)***	-0.821 (-2.76)***	-0.802 (-2.41)**	-0.912 (-1.91)*
<i>bal(-1)_dis</i>	0.218 (1.17)	-0.382 (-1.03)	0.208 (1.09)	0.202 (0.97)	-0.387 (-0.37)
<i>Constant</i>	-7.138 (-0.60)	0.022 (0.31)	-8.864 (-0.71)	-11.219 (-0.84)	25.137 (0.38)
Method of estimation	Fixed-effects	Arellano-Bond One-Step ^a	Fixed-effects	Fixed-effects	Fixed-effects
R ² within	0.7844	-	0.7836	0.7797	0.8739
R ² between	0.8790	-	0.8353	0.7996	0.9918
R ² overall	0.8292	-	0.8065	0.7900	0.3091
Sargan Test ^b	-	1.000			
2nd Order Serial Correlation Test ^c	-	0.2276			
No. countries	19	19	18	16	3
No. observations	319	279	302	266	53

Note: For fixed effects estimates and GMM estimates, see notes below Table 4.

Table 9. Discretionary PBC in Young Non-OECD Democracies

Dependent variable:	Total	Total	Presidential	Parliamentary	Proportional	Majoritarian
<i>bal</i>	(1)	(2)	(3)	(4)	(5)	(6)
<i>bal(-1)</i>	0.347 (1.73)*	0.263 (1.07)	0.307 (1.22)	0.415 (1.14)	0.274 (1.21)	1.040 (2.68)**
<i>lngdp_pc</i>	-1.948 (-1.00)	1.490 (0.35)	-3.489 (-1.55)	-1.834 (-0.34)	-1.728 (-0.80)	-12.802 (-1.37)
<i>Gdpr</i>	0.114 (2.88)***	0.097 (1.27)	0.126 (2.77)***	0.038 (0.40)	0.106 (2.41)**	0.106 (1.06)
<i>Trade</i>	-0.005 (-0.26)	0.002 (0.06)	-0.008 (-0.35)	-0.032 (-0.72)	-0.005 (-0.24)	0.009 (0.18)
<i>Pop65</i>	0.081 (0.12)	9.437 (1.84)	0.709 (0.86)	-0.187 (-0.09)	0.154 (0.21)	-1.326 (-0.29)
<i>Pop1564</i>	-0.005 (-0.02)	-1.264 (-1.31)	0.306 (0.13)	0.678 (0.71)	0.008 (0.04)	-1.050 (-0.90)
<i>ln(1+pi)</i>	1.294 (1.25)	0.256 (0.16)	1.330 (1.19)	-14.020 (-1.38)	1.027 (0.93)	12.236 (0.88)
<i>ln(1+pi)sq</i>	-0.391 (-1.46)	-0.214 (-0.87)	-0.407 (-1.42)	10.888 (1.58)	-0.349 (-1.22)	-95.469 (-1.37)
<i>Checks</i>	-2.485 (-1.96)**	-1.952 (-1.49)	-1.981 (-1.27)	-1.170 (-0.48)	-3.693 (-2.52)**	5.427 (1.92)*
<i>Pbc_dis</i>	-0.626 (-2.01)**	-0.755 (-2.91)***	-0.583 (-1.70)*	-0.986 (-1.08)	-0.807 (-2.26)**	0.107 (0.26)
<i>bal(-1)_dis</i>	-0.423 (1.44)	0.327 (1.03)	-0.396 (-1.10)	-0.084 (-1.10)	-0.344 (-1.04)	-1.140 (-2.13)**
<i>Constant</i>	12.398 (0.81)	0.037 (0.27)	18.960 (1.13)	-23.638 (-0.48)	10.192 (0.60)	158.757 (1.41)
Method of estimation	Fixed-effects	Arellano-Bond One-Step ^a	Fixed-effects	Fixed-effects	Fixed-effects	Fixed-effects
R ² within	0.1379	-	0.1410	0.3828	0.1481	0.6023
R ² between	0.0394	-	0.2876	0.9269	0.0263	0.5365
R ² overall	0.0011	-	0.0176	0.8230	0.0249	0.4363
Sargan Test ^b	-	1.000				
2nd Order Serial Correlation Test ^c	-	0.1134				
No. countries	30	28	24	6	26	4
No. observations	294	235	251	43	255	39

Note: For fixed effects estimates and GMM estimates, see notes below Table 4.

Columns (1) and (2) of Table 8 show that PBC are significant even if one restricts the sample to rich established democracies, where Shi and Svensson (2002a) and Brender and Drazen (2004) note that voters are more informed and experienced. The main difference with column (1) of Table 7 is that the degree of discretion does not significantly affect the persistence of the budget deficit. Similar remarks hold for poor new democracies in Table 9.

Persson and Tabellini (2002) focus on the effects of different forms of government (presidential or parliamentary) and electoral rules (proportional or majoritarian) on PBC. The approach followed here in principle attempts to reduce these institutional differences to a common metric of veto players (Tsebelis (2002)).

The finding in Persson and Tabellini (2002) of stronger PBC in countries with presidential systems can be linked in our framework to lower effective checks and

balances compared to countries with parliamentary systems (*checks* of 0.33 vs. 0.60).¹³ We cannot explain differences between countries with proportional or majoritarian electoral rules, because they do not differ much according to checks and balances (*checks* of 0.46 vs. 0.41).

Introducing dummies to condition discretionary PBC for form of government or electoral rules leads to find that cycles do not differ significantly between these groups. Given the large standard errors of coefficients, we do not present the full results, but rather show the sensitivity of the estimates in Tables 8 and 9 when they are restricted to subgroups of countries. The results on discretionary PBC in Table 8 for rich established democracies remain unchanged, except for the subset of presidential countries where there are insufficient degrees of freedom to run a separate regression. As for poor new democracies in Table 9, we are not able to get significant results for parliamentary and majoritarian countries, but we have few observations (and large standard errors).

Our results always show clear evidence of discretionary PBC in countries with proportional electoral rules, but this might be simply due to the fact that this is the group with the most observations (in our complete sample, 55 of the 67 countries have this electoral rule). In poor new democracies, presidential countries have significant discretionary PBC, and in rich established democracies, parliamentary and majoritarian countries have significant discretionary PBC.

VI. Final remarks

Aggregate electoral cycles are more controversial than electoral cycles in the composition of government spending, due to the weak evidence on aggregate PBC in OECD countries. Following the insight in Alesina and Rosenthal (1995) that divided government moderates executive discretion, we use the Henisz political constraints index to derive the presence of a legislative veto player, and combine it with the degree of rule of law to have a measure of the effective checks and balances that the executive faces and isolate the discretionary component of PBC.

Given the literature on how coalition governments and divided government can make it harder to adjust to shocks (Roubini and Sachs (1989), Alt and Lowry (1994)), we also control for the influence of effective checks and balances on the persistence of the budget surplus. Though we find that less discretion indeed increases the persistence

¹³ 18 of the 37 presidential systems are in Latin America, a region characterized by developing countries that are almost all new democracies

of the budget surplus, this result vanishes when we consider OECD and non-OECD countries separately.

Our main finding is that effective checks and balances play a significant role in moderating PBC in OECD countries, reducing the size of cycles. After conditioning for this, discretionary PBC are still present in the countries with the best-informed and experienced voters, namely, OECD countries that are established democracies. This points to a promising path using more detailed measures of veto players and budget institutions.

Our results complement those of Alt and Lassen (2004), who find electoral cycles in fiscal balance in advanced industrialized democracies when there is low transparency, while no such cycles can be observed with high transparency. Together with asymmetric information and learning by voters (and policy players), the message here is that discretionality matters. Incidentally, transparency might not be independent from checks and balances, since the requirement of reliable information on government activities is usually a demand of opposition parties.

Finally, our results can be given an econometric interpretation. There is an errors-in-variables problem in the existing literature if the discretionary component of PBC is the relevant variable. We can also give an omitted variable interpretation, where effective checks and balances is the missing variable. This omission is particularly serious in OECD countries that are established democracies, because they are positively correlated with high checks and balances, biasing the estimate of PBC downwards.

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Chapter 3

Conditional Political Budget Cycles in Argentine Provinces

I. Introduction

In this paper, we investigate the presence of electorally-motivated manipulations in fiscal outcomes in Argentine provinces. We empirically address two main questions:

Are political budget manipulations present in executive election periods in Argentine provinces? And, are there any systematic differences in the size and composition of expenditures and revenues in provinces politically aligned with the federal executive?

The political budget cycles literature relates elections and policy makers' behavior conjecturing that by rising public expenditures in election periods the incumbent can increase his chance of reelection. The first generation studies were based on the idea of adaptive expectations, assuming that in the future the voter will act partly based on what happened in the past. Thus, he or she can systematically be deceived. Later on, following the trend in economic analysis, rational expectations were introduced in formal models. Rational voters builds conjectures related to the competency of the politician and by increasing spending the incumbent can signal his level of competence. Regarding the incumbent motivations, there is a division between opportunist politicians and partisan politicians; namely, those who want to rule for the sake of power itself and those who want to do so in order to put their preferred policies into practice.

In Argentina, the literature has not yet deeply analyzed the phenomenon of budget cycles in election times. Meloni (2001) explores this issue by analyzing the change in provincial current expenditure and its relation with the votes obtained by the governing party. However, this study was not explicitly performed within the framework of the political budget cycle theory.

Considering the federal organization of Argentine provinces, it might seem relevant to sort out evidence about the existence of cycles in fiscal variables that may represent some kind of opportunist behavior on the part of the incumbent looking for reelection. This paper empirically analyzes the evidence of cycles in fiscal balance, expenditures and revenues in 22 Argentine provinces during the period between 1985 and 2001, using econometric methods for panel data. We consider the executive election date as the main explanatory variable for budget cycles. The rest of this paper is structured as follows. The next section briefly summarizes the theoretical framework on

political budget cycles. Section III presents the data set. Section IV describes the empirical specification and econometric techniques. The empirical results and discussion are reported in section V. Finally, Section VI concludes.

II. The Political Budget Cycle

A. Previous Literature

The first models that formalize the political behavior that generates cycles in economic variables in relation to the electoral calendar can be separated into two different views. One of them, called 'Opportunistic Political Business Cycle' makes emphasis on the opportunistic nature of the politician. This means that they have no other preference but to hold office, for which reason they act in such a manner to maximize the chances of being re-elected. The pioneering work of Nordhaus (1975) is within this trend, depicting the politician as a manipulator of pre-electoral outcomes. Lindbeck (1976) also follows this line.

There is another branch of the literature, the one in 'Partisan Political Business Cycle' that stresses the partisan nature – as an ideological aspect- of the incumbent (Hibbs 1977). In this case, cycles in economic variables are originated in the ideological preferences of the politician.

First generation models are based on adaptive expectations. This behavior is sometimes called 'myopic' (or irrational) on the part of the voter since, once the politician adopts expansive policies, the voter does not remember or does not take into account those recessive policies adopted by the incumbent in the past that they are likely to repeat in the future. It is worth noting that these models are based on the assumption of a negative relation between unemployment and inflation; that is, the possibility of exploiting a 'Phillips curve'.

Models based on rational expectations (Kydland and Prescott 1977; and Barro and Gordon 1983) started to appear in the 80s. During the 90s -as well as in the present work- emphasis is made on fiscal policy rather than on monetary policy as the generator of cycles. Among the papers that are representative of this period are those by Rogoff and Sibert (1988) and Rogoff (1990). Here, the opportunist politician manipulates the expenditure policy during electoral years with the sole purpose of showing that they are competent, thus increasing their chances of winning the elections.

Recently, the literature has explored not only the level effect on fiscal variables but also the so-called composition effect; that is, how expenditure components (such as

consumption and investment) change during this cycle of electoral origin. Among the contributions, those of Schucknecht (2000) and Block (2002) can be mentioned. Research into how institutional variables can impose (or relax) constraints on the cycle is not less relevant. That is, how strong institutions should temper the cycle and how, on the other hand, weak institutions would make way for the opportunist politician to distort policies. This kind of study, based on the exploration of institutional variables and their effects on the level and composition of the cycle can be seen in Shi and Svensson (2002) and in Block (2002).

This paper follows the line of models called ‘opportunist –rational’ that, according to the definition by Alesina, Rubini and Cohen (1997), present the following distinctive characteristics:

- (i) Short-term manipulation of fiscal or monetary policy.
- (ii) Strengthening of policies after the elections.
- (iii) Non-systematic effects on unemployment.
- (iv) Politicians struggling for re-election.

B. Theoretical Framework

According to the Rogoff (1990) approach, opportunistic politicians generates cycles in the economic variables while trying to show their competency, which cannot be directly observed by the voter. In this way, they are trying to increase their chances of being re-elected. One of the main issues is asymmetry in information: if the voters were able to directly observe the capacity or ability of the politician, their decision would be obvious.

The fact is that if ability is not easily observable then there is some possibility that the politicians manipulate fiscal policy in such a manner that they might seem to have more competency than the one they really have, thus augmenting the chances of winning the election. Rogoff’s model could be outlined as follows: the politician produces a public good using two inputs, taxes and their competency. The latter can be thought of as a parameter of productivity, since capable incumbents need fewer resources to make more things and vice-versa.

The voters observe taxes and expenditure (which have different grades of visibility) and use that information to make inferences about the politician’s capacity, which is not directly observable since other factors also influence elections vote is probabilistic and, there is some likelihood q that the politicians might be re-elected, and

($1-q$) that might not. This information is known by them, for which reason they are tempted to take political steps so as to augment their chances for re-election q .

Rogoff concludes that given the informational asymmetries regarding the politician's capacity, expenditure will be increased by competent politicians - particularly the most clearly perceived by the voter- so as to pretend to be the most capable politician. In equilibrium, this behavior increases reelection chances of competent politicians.

Some aspects of visible and non-visible expenditures are worth noting. We have so far only referred to biases in expenditure towards its most visible components. Which type of expenditure are the most visible is not a clear-cut classification. Rogoff states that at election times, expenditure biases towards current expenditure, and this means an increase in current expenditure as a percentage of total expenditure. This point of view is not universally shared and empirical evidence appears divided.

Schuknecht (2000) assumes that the bias in expenditure in developing countries is towards capital goods. He states that starting great public works right before the elections, and then bringing them to a halt immediately after seems to be easier than increasing current expenditure, since the latter can entail short and long term commitments.

In the same vein, Krueger and Turán (1993) -when analyzing the electoral process in Turkey- argue that there are pre-electoral increases in both investment and infrastructure programs.

The empirical analysis in Schuknecht's study is performed using relative per capita levels of expenditures. Current expenditure and capital expenditure (as percentage of per capita GDP), both increase before elections. Nevertheless, the composition effect, that is, the current (or capital) expenditure as a percentage of total expenditure is not directly analyzed. Consequently, what is in fact found out that total expenditure increases before elections. However, nothing clear can be stated regarding the bias of the composition effect. Similarly, the work by Krueger and Turán does not test the composition effect either.

On the other hand, Block (2002) follows Rogoff's line of thought and argues that the bias in the composition effect moves towards current expenditure. However, he admits that the evidence accounting for his hypothesis becomes stronger when only the richest countries in the sample – controlling by per capita GDP– are taken into account.

In this paper we follow the approach presented by Block (2002) using current expenditure as percentage of total expenditures in order to test the possible bias in the expenditure composition (“composition effect”) in election periods.

The idea that electoral budget cycles can be found at a sub-national level lies in the federal organization of Argentina. The Argentine Constitution, under sections 122 and 123 states: ‘*The provinces provide their own local institutions and are governed by them. They choose their governors, legislators and the rest of the provincial officers, without intervention of the Federal Government*’ and ‘*each province writes its own constitution....regulating its scope and content in the institutional, political, administrative, economic and financial orders*’. The fiscal autonomy of provinces from the federal government is a factor that allows the potential existence of electoral budget cycles at a local level.

III. Data

We construct a panel data set to test the existence of electoral cycles in provincial fiscal variables. Our data set includes data on provincial government budget balance, spending and revenues, political data on provincial executive election dates and political party in power, per capita Gross Geographic Product (GGP) and GGP growth. Our database has annual observations for 22 provinces for the period between 1985 and 2001, averaging four provincial executive elections.

Two provinces were excluded from the original sample. First, the City of Buenos Aires is excluded from the analysis since it was only in the year 1996 that the elections for Chief of Government (i.e. governor) were held. Up to that moment, the City Mayor was directly appointed by the national executive power. Second, the province of Corrientes is the other exception, because it had to undergo two federal interventions during the 90s. The first one, in 1991, was due to disagreement between the provincial electors; and the one in 1999 was due to serious social disturbances. Both provinces were excluded from the database to perform the econometric estimation.

The source of the fiscal data is the Ministry of Economy (“*Dirección Nacional de Coordinación Fiscal con las Provincias, Secretaría de Hacienda del Ministerio de Economía y Producción de la Nación*”). Geographic Gross Product (GGP) estimates were taken from Mirabella (2002), who approach the provincial GGP using residential electricity consumption. Table 1 presents the variables used for the estimates and Table 2 presents descriptive statistics of the dependent fiscal variables.

Table 1: Definition of Variables

Dependent Variables (fiscal variables). (All values expressed in constant 1993 Argentine Pesos deflated by the combined prices index -wholesale-consumer- from INDEC)

DEF_{it}: Fiscal Balance [Deficit (-) Surplus (+)] divided by provincial GGP in province *i* year *t*
Source: MECON

TE_{it}: Total Public Expenditure divided by GGP from province *i*
in year *t*. Source: own elaboration based on Ministry of Economy (MECON)

CE_{it}: Current Expenditure divided by public total expenditure in province *i* in year *t*. Source: MECON

TR_{it}: Total Provincial Revenue divided by GGP in the province *i* in year *t* (includes revenue from provincial taxes, federal revenue sharing –“coparticipación federal”- and other federal transfers –“aportes del tesoro”- Source: MECON

PTR_{it}: Revenue from Provincial Taxes divided by provincial GGP in province *i* in year *t*. Source: MECON

FR_{it}: Provincial revenues from federal revenue sharing ("coparticipation federal") plus transfers from federal government divided by provincial GGP in province *i* in year *t*. Source: MECON

Control Variables

GGP_{it}: Natural log of per capita Geographic Gross Product of province *i* during year *t*
Source: Mirabella (2002) and National Institute of Statistics and Census (INDEC)

GROWTH_{it}: GGP Growth rate in the province *i* between the year *t* and the *t*-1
Source: Mirabella (2002).

Election Variables

ELE_{it}: Election dummy. Binary variable that assumes value 1 if in province *i* elections were held during the year *t* and 0 otherwise.

Source: own elaboration based on “Guia Electoral”.

PBC_{it}: Political Budget Cycle dummy. Variable assuming value 1 if ELE_{it} is equal to 1; -1 if ELE_{it-1} is equal to 1 and 0 otherwise. Source: own elaboration based on “Guia Electoral”.

ELE+1_{it}: Post Election dummy. Binary variable that assumes value 1 if ELE_{it-1} is equal to 1 and 0 otherwise.
Source: own elaboration based on “Guia Electoral”.

ELE_UNAL_{it}: Conditional Election dummy. Binary variable that assumes value 1 if in province *i* elections were held during the year *t* and the provincial and federal executive governments were unaligned (different political party), and 0 otherwise. Source: own elaboration based on “Guia Electoral”.

ELE_AL_{it}: Conditional Election dummy. Binary variable that assumes value 1 if in province *i* elections were held during the year *t* and the provincial and federal executive governments were aligned (same political party), and 0 otherwise. Source: own elaboration based on “Guia Electoral”.

PBC_UNAL_{it}: Conditional Political Budget dummy. Binary variable that assumes value 1 if ELE_UNAL_{it} is equal to 1; -1 if ELE_UNAL_{it-1} is equal to 1 and 0 otherwise. Source: own elaboration based on “Guia Electoral”.

PBC_AL_{it}: Conditional Political Budget dummy. Binary variable that assumes value 1 if ELE_AL_{it} is equal to 1; -1 if ELE_AL_{it-1} is equal to 1 and 0 otherwise. Source: own elaboration based on “Guia Electoral”.

Table 2: Fiscal Variables: Descriptive Statistics

	Mean	Std. Dev.	Min.	Max.	No. Obs
DEF	-0.022	0.031	-0.155	0.058	374
TE	0.237	0.123	0.052	0.812	374
CE	0.807	0.091	0.445	0.952	374
TR	0.215	0.113	0.046	0.825	374
PTR	0.028	0.014	0.004	0.121	374
FR	0.186	0.110	0.024	0.704	374

The electoral budget cycle is analyzed through the variables fiscal balance, total expenditure, expenditure composition¹, total provincial revenue, revenue from provincial taxes and revenue from the federal government². The period of analysis ranges from 1985 to 2001.

IV. Empirical Analysis

The theoretical and empirical literature on political budget cycles suggests that the timing of elections should influence fiscal outcomes. The relationship between a fiscal variable, y_{it} , and the electoral cycle can be stated as follows:

$$y_{i,t} = \alpha + \sum_{j=1}^k \beta_j y_{i,t-j} + \sum_{j=1}^m \gamma_j x_{j,i,t} + \delta_1 e + \eta_i + \varepsilon_{it} \quad (1)$$

for $i = 1..N$, $t = 1..T$, $j = 1 \dots k$, where e is a binary election variable indicating if an election took place in province i during the year t ; x is a vector of control variables that in our estimations include per capita Geographic Gross Product (GGP) and the growth rate of the Geographic Gross Product (GROWTH).

This specification represents a standard dynamic panel, where the dependent variable is a function of its own lagged levels, of set of controls (x_j), of the time when elections take place and of a specific effect per province (η_i). The term ε_{it} is a random error that is assumed iid.

Assuming that the unobserved province-specific effects are identical across provinces, that the error term is not serially correlated, and that the explanatory variables are strictly exogenous then it is possible to estimate this relation consistently

¹ Ratio of current expenditure relative to total public expenditure. The most important component of current expenditure are salaries of provincial public servants. On the other hand, construction is the most important item of capital expenditure.

² Provincial revenues from federal revenue sharing ("coparticipation federal") plus special (discretionary) transfers from federal government ("Aportes del Tesoro Nacional" – ATN).

through OLS. However, these assumptions may not hold in the panel, particularly the assumption of equality of the unobservable effects per province. This being so, then OLS estimates are inconsistent since the lagged dependent variable is correlated to the error term $w_{i,t} = \eta_i + \varepsilon_{it}$.

It is possible to control the specific effects using the panel data Fixed Effects (FE) estimator. However, the transformed error term will still be correlated with the lagged dependent variable. The bias will depend on T (the length of the panel); and provided T tends to infinite, the FE estimator of the coefficients will be consistent.

Considering these problems, the Generalized Method of Moments (GMM) designed for dynamic models by Arellano and Bond (1991) is performed in the estimations. The Arellano-Bond strategy consists in the differentiation of the equations to eliminate the specific effects and solve the inconsistency using the lagged values of the dependent variable as instruments. Assuming the error term is not serially correlated, the dependent variable lagged two periods or more constitute valid instruments for the new dependent variable in differences. Likewise, the same can be said for the control variables.

It will be assumed in our particular case that the vector from variables x_{jit} is slightly exogenous or predetermined; that is to say, it is not correlated with future realizations of the error term. The elections variable will be considered strictly exogenous.

Estimates are performed using three methods: OLS, Fixed Effects and GMM Arellano-Bond for dynamic panel data. The GMM method seems to be preferable due to the characteristics previously mentioned. Nevertheless, since it makes use of the lagged values of the variables as instruments, the set of observations available is smaller. For this reason and for comparative purposes, results from the three methods are reported.

The political cycle is modeled including the binary variable ELE that assumes value 1 in election years, and 0 in the rest of the years.

As usual in the empirical literature, the variable PBC (Political Budget Cycle) is also used, taking value 1 during the election year, -1 in the following year and 0 in the remaining ones. This variable imposes the restriction that the pre-electoral increase in spending or deficit is equivalent in magnitude to the posterior contraction.

Additionally, we also run the non-restricted regressions with the election dummy ELE and the post election dummy ELE+1, and test the validity of the restriction imposed by the use of PBC.

Our analysis includes six fiscal outcomes as dependent variables to test the electoral manipulation, its origins and consequences: ratio of provincial budget balance to GGP (DEF), ratio of total public expenditure to GGP (TE), current public expenditure relative to total public expenditure (CE), total provincial revenue relative to GGP (TR), revenue from provincial taxes relative to GGP (PTR) and provincial revenues from federal revenue sharing plus transfers from federal government relative to GGP (FR). Two basic controls will be included in the regressions: the per capita geographic gross product (GGP) and growth rate of the GGP (GROWTH).

V. Unconditional Budget Cycles

This section presents the empirical analysis of electoral cycles in fiscal variables, focusing on the provincial budget surplus, expenditures and revenues. We first present the unconditional results of elections over the fiscal variables. We then look at the conditional results, controlling for the alignment between the provincial and federal executives.

A. Budget Balance

Table 3 shows the main unconditional results with respect to the provincial budget balance (deficit); that is equation (1) including the election dummies ELE and PBC and using as controls the GGP and the growth of GGP per capita³.

In the columns 1 to 3, with the three different estimation methods, ELE has the expected negative sign, although is not statistically significant in any case.

In columns 4, 5 and 6 the results are obtained making use of the variable PBC as a regressor for the elections. In this case, the coefficients estimated by OLS, Fixed Effects (FE) and GMM are significantly negative, suggesting that the level of electoral cycle – defined as the increase in deficit during the election year and the contraction in the following- is approximately 0.6% of GGP. For GMM estimation the Sargan test is reported, where the null hypothesis is that the instrumental variables are uncorrelated with the residuals. In addition, the serial correlation test is presented, where the null hypothesis is the absence of second order serial correlation in the first-difference residuals. Estimates satisfy both tests (no rejection of null hypothesis).

³ Full econometric estimation results presented in the Chapter 3 Appendix.

Table 3: Elections and Fiscal Balance

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM	4 OLS	5 FIXED EFFECTS	6 GMM
ELE	-0.0037 (-1.17)	-0.0031 (-0.91)	-0.0030 (-0.97)			
PBC				-0.0064 (-3.22)**	-0.0060 (-3.08)**	0.0062 (-3.33)**
F-test ^a		2.07			2.07	
p-value		0.0045			0.0045	
Sargan test ^b			283.73			285.22
p-value			0.9994			0.9993
Serial Corr ^c			-0.04			0.210
p-value			0.9677			0.8359
No.obs.	308	304	302	308	308	308
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.36			0.38		

Notes: Dependent variable DEF is ratio of government surplus to Geographic Gross Product (PBG).

Estimated Regressions:

$$DEF_{it} = \alpha + \beta_1 DEF_{it-1} + \beta_2 DEF_{it-2} + \beta_3 DEF_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$DEF_{it} = \alpha + \beta_1 DEF_{it-1} + \beta_2 DEF_{it-2} + \beta_3 DEF_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

Table 4: Elections and Fiscal Balance

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM
ELE	0.0020 (0.60)	0.0023 (0.66)	0.0016 (0.50)
ELE+1	0.0140 (3.98)***	0.0135 (4.07)***	0.0141 (4.34)***
F-test: <i>ELE = -ELE+1</i>	8.27	7.71	8.78
p-value	0.0043	0.0059	0.0030
F-test ^a		2.11	
p-value		0.0036	
Sargan test ^b			277.02
p-value			0.9998
Serial Corr ^c			1.25
p-value			0.2131
No. obs.	308	308	308
No. provinces	22	22	22
R ² (adj.)	0.40		

Notes: Dependent variable DEF is ratio of government surplus to Geographic Gross Product (PBG).

Estimated Regressions:

$$DEF_{it} = \alpha + \beta_1 DEF_{it-1} + \beta_2 DEF_{it-2} + \beta_3 DEF_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$DEF_{it} = \alpha + \beta_1 DEF_{it-1} + \beta_2 DEF_{it-2} + \beta_3 DEF_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

The PBC variable is meant to capture both pre and post-electoral effects. However as Persson and Tabellini (2002) remark, pre and post electoral effects may differ, so we check if the restriction that the coefficient estimate of ELE is equal to the

coefficient estimate of minus ELE in $t+1$, is not rejected by the data. Estimate results are presented in Table 4.

Table 4 shows that the electoral dummy ELE is non significant and the post electoral dummy ELE+1 is positive and significant. In all estimates, the F test soundly rejects the restriction imposed by the PBC variable, that the post-electoral contraction in the budget surplus as a percentage of GGP is of the same size as the pre-electoral expansion. We can interpret the results as follows: a) there is no evidence of surplus falling in election periods, b) the restriction that surplus falls below its trend, and then jumps above it, is not supported by the data and, c) the significant effect of PBC is driven by the jump of surplus in post electoral periods.

B. Expenditures: Total and Composition

Tables 5 and 6 show the effects of the electoral cycle over total public expenditure in the provinces, measured as a proportion of GGP.

In the OLS and GMM regressions, the coefficients are positive and significant for ELE, with a value indicating that the expenditure over GGP increases approximately one percentage point during the year of elections. We found a short run cycle in spending, approximately of 0.6-0.8 percentage points of GGP, in the OLS, FE and GMM regressions with PBC as explanatory variable.

Results in Table 6 also suggest that the PBC significance is due to reductions in expenditure in the post election years, and the F tests does not reject the null hypothesis of equality between ELE and -ELE+1.

Tables 7 and 8 shows the estimates performed to evaluate the “composition effect” in provincial expenditure around elections. The dependent variable CE represents the expenditure in consumption goods as a proportion of total provincial expenditure. None of the estimates present statistically significant results to provide evidence supporting the hypothesis of a shift in spending towards consumption goods or investment goods in election years. The case of the PBC shows similar results, even though the coefficient estimated by GMM shows some evidence of a slight bias towards capital goods in electoral years and towards consumption goods in post electoral years.

Table 5: Elections and Total Expenditure

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE	0.0115 (2.50)**	0.0057 (1.50)	0.0100 (2.34)***			
PBC				0.0082 (3.15)***	0.0057 (2.55)**	0.0074 (2.83)***
F-test ^a		11.30			11.47	
p-value		0.0000			0.0000	
Sargan test ^b			254.35			250.02
p-value			1.0000			1.0000
Serial Corr. ^c			1.00			1.41
p-value			0.3166			0.1600
No. obs.	308	308	308	308	308	308
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.90			0.90		

Notes: Dependent variable TE is the ratio of total provincial expenditure to Geographic Gross Product (PBG).

Estimated Regressions:

$$TE_{it} = \alpha + \beta_1 TE_{it-1} + \beta_2 TE_{it-2} + \beta_3 TE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$TE_{it} = \alpha + \beta_1 TE_{it-1} + \beta_2 TE_{it-2} + \beta_3 TE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

Table 6: Elections and Total Expenditure

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM
ELE	0.0086 (1.63)	0.0026 (0.64)	0.0074 (1.59)
ELE+1	-0.0078 (-1.44)	-0.0087 (-2.26)**	-0.0073 (-1.61)
F-test: <i>ELE = -ELE+1</i>	0.01	0.88	0.00
p-value	0.9271	0.3484	0.9961
F-test ^a		11.51	
p-value		0.0000	
Sargan test ^b			249.74
p-value			1.0000
Serial Corr ^c			1.42
p-value			0.1568
No. obs.	308	308	308
No. provinces	22	22	22
R ² (adj.)	0.90		

Notes: Dependent variable TE is the ratio of total provincial expenditure to Geographic Gross Product (PBG).

Estimated Regressions:

$$TE_{it} = \alpha + \beta_1 TE_{it-1} + \beta_2 TE_{it-2} + \beta_3 TE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$TE_{it} = \alpha + \beta_1 TE_{it-1} + \beta_2 TE_{it-2} + \beta_3 TE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

Table 7: Elections and Composition Effect

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM	4 OLS	5 FIXED EFFECTS	6 GMM
ELE	0.0000 (0.00)	0.0014 (0.21)	-0.0058 (-0.90)			
PBC				-0.0039 (-0.99)	-0.0025 (-0.65)	-0.0066 (-1.69)*
F-test ^a		2.18			2.16	
p-value		0.0025			0.0028	
Sargan test ^b			244.93			243.05
p-value			1.0000			1.0000
Serial Corr. ^c			-0.89			-0.52
p-value			0.3739			0.6014
No. obs.	308	308	308	308	308	308
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.60			0.60		

Notes: Dependent variable CE is the ratio of current expenditure to total provincial expenditure.

Estimated Regressions:

$$CE_{it} = \alpha + \beta_1 CE_{it-1} + \beta_2 CE_{it-2} + \beta_3 CE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$CE_{it} = \alpha + \beta_1 CE_{it-1} + \beta_2 CE_{it-2} + \beta_3 CE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

Table 8: Elections and Composition Effect

Equation	1	2	3
Estimation Method	OLS	FIXED EFFECTS	GMM
ELE	0.0039 (0.53)	0.0046 (0.66)	-0.0016 (-0.23)
ELE+1	0.0108 (1.64)	0.0090 (1.37)	0.0112 (1.67)*
F-test: $ELE = -ELE + 1$	1.60	1.49	0.76
p-value	0.2073	0.2228	0.3831
F-test ^a		2.15	
p-value		0.0030	
Sargan test ^b			244.37
p-value			1.0000
Serial Corr ^c			-0.36
p-value			0.7182
No. obs.	308	308	308
No. provinces	22	22	22
R ² (adj.)	0.58		

Notes: Dependent variable CE is the ratio of current expenditure to total provincial expenditure.

Estimated Regressions:

$$CE_{it} = \alpha + \beta_1 CE_{it-1} + \beta_2 CE_{it-2} + \beta_3 CE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$CE_{it} = \alpha + \beta_1 CE_{it-1} + \beta_2 CE_{it-2} + \beta_3 CE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals.

(c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

C. Revenues: Total, Federal and Provincial

To track the possible changes in fiscal revenues around elections Tables 9, 10, 11, 12, 13 and 14 present the estimates considering as dependent variables total provincial revenue (TR) and its components: revenue from federal sources (FR), that includes federal tax sharing and other federal transfers (mostly discretionary) and revenue from provincial taxes (PTR).

Tables 9 to 12 show the results with total revenue (TR) and federal revenue (FR) as dependant variables. The electoral years are related to a significant tendency of revenues to go up, explained by the increase in federal revenue, and is important to note that federal revenue is 90% of total provincial revenues.

The PBC variable is non significant in all regressions, so there is no evidence of cycles around elections. The most significant effect is the revenue increase in election years, and the discretionary transfers from the federal government could explain that. The federal tax sharing is mostly determined by fixed coefficients and cannot be easily manipulated.

Results in Tables 13 and 14 show non significant manipulations in provincial taxes; in all regressions revenue from this source is not sensitive to the election and PBC dummy variables. This seems reasonable, because in most provinces local taxes are a very small part of total revenues. Changes (reductions) in this variable may have a non relevant effect over the voter's perceptions about competency of the incumbent, reducing his incentives to engage in electoral manipulations over provincial taxes.

Table 9: Elections and Total Revenue

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM	4 OLS	5 FIXED EFFECTS	6 GMM
ELE	0.0111 (2.62)*	0.0051 (1.59)	0.0095 (2.44)**			
PBC				0.0025 (0.90)	0.0004 (0.21)	0.0019 (0.83)
F-test ^a		17.19			17.41	
p-value		0.0000			0.0000	
Sargan test ^b			270.19			273.64
p-value			0.9999			0.9999
Serial Corr. ^c			1.25			0.98
p-value			0.2096			0.3289
No. obs.	308	308	308	308	308	308
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.89			0.89		

Notes: Dependent variable TR is the ratio of current expenditure to total provincial expenditure.

Estimated Regressions:

$$TR_{it} = \alpha + \beta_1 TR_{it-1} + \beta_2 TR_{it-2} + \beta_3 TR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$TR_{it} = \alpha + \beta_1 TR_{it-1} + \beta_2 TR_{it-2} + \beta_3 TR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

Table 10: Elections and Total Revenue

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM
ELE	0.0138 (2.91)***	0.0070 (2.08)**	0.0125 (3.05)***
ELE+1	0.0075 (1.44)	0.0055 (1.75)*	0.0085 (2.07)**
F-test: $ELE = -ELE+1$	6.30	5.59	9.76
p-value	0.0126	0.0187	0.0018
F-test ^a		17.18	
p-value		0.0000	
Sargan test ^b			274.97
p-value			0.9999
Serial Corr ^c			1.00
p-value			0.3194
No. obs.	305	305	302
No. provinces	22	22	22
R ² (adj.)	0.89		

Notes: Dependent variable TR is the ratio of current expenditure to total provincial expenditure.

Estimated Regressions:

$$TR_{it} = \alpha + \beta_1 TR_{it-1} + \beta_2 TR_{it-2} + \beta_3 TR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$TR_{it} = \alpha + \beta_1 TR_{it-1} + \beta_2 TR_{it-2} + \beta_3 TR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

Table 11: Elections and Revenue from Federal Government⁴

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM	4 OLS	5 FIXED EFFECTS	6 GMM
ELE	0.0114 (2.89)***	0.0053 (1.74)*	0.0097 (2.73)***			
PBC				0.0028 (1.09)	0.0004 (0.23)	-0.0023 (1.06)
F-test ^a		15.51			15.76	
p-value		0.0000			0.0000	
Sargan test ^b			254.86			258.41
p-value			1.0000			1.0000
Serial Corr. ^c			1.49			1.25
p-value			0.1355			0.2115
No. obs.	308	308	308	308	308	308
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.90			0.90		

Notes: Dependent variable FR is the ratio of federal revenues to Gross Geographic Product (PBG).

Estimated Regressions:

$$FR_{it} = \alpha + \beta_1 FR_{it-1} + \beta_2 FR_{it-2} + \beta_3 FR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{FR}$$

$$FR_{it} = \alpha + \beta_1 FR_{it-1} + \beta_2 FR_{it-2} + \beta_3 FR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{FR}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

⁴Provincial revenues from revenue sharing ("coparticipation") plus (discretionary) transfers from federal government (i.e. "Aportes del Tesoro Nacional" – ATN).

Table 12: Elections and Revenue from Federal Government⁵

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM
ELE	0.0138 (3.25)***	0.0073 (2.28)**	0.0124 (3.33)***
ELE+1	0.0071 (1.54)*	0.0057 (1.92)*	0.0078 (2.09)**
F-test: <i>ELE = -ELE+1</i>	8.03	6.69	10.92
p-value	0.0049	0.0102	0.0456
F-test ^a		15.55	
p-value		0.0000	
Sargan test ^b			258.52
p-value			1.0000
Serial Corr ^c			1.30
p-value			0.1944
No. obs.	308	308	308
No. provinces	22	22	22
R ² (adj.)	0.91		

Notes: Dependent variable FR is the ratio of federal revenues to Gross Geographic Product (PBG).

Estimated Regressions:

$$FR_{it} = \alpha + \beta_1 FR_{it-1} + \beta_2 FR_{it-2} + \beta_3 FR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \gamma_4 ELE+1_{it} + \eta_i + \varepsilon_{it}$$

$$FR_{it} = \alpha + \beta_1 FR_{it-1} + \beta_2 FR_{it-2} + \beta_3 FR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

⁵Provincial revenues from revenue sharing ("coparticipation") plus (discretionary) transfers from federal government.

Table 13: Elections and Revenue from Provincial Taxes

Equation Estimation Method	1 OLS	2 FIXED EFFECTS	3 GMM	4 OLS	5 FIXED EFFECTS	6 GMM
ELE	0.0002 (0.23)	0.0002 (0.22)	0.0002 (0.23)			
PBC				-0.0002 (-0.55)	-0.0002 (-0.51)	-0.0003 (-0.70)
F-test ^a		3.30			3.30	
p-value		0.0000			0.0000	
Sargan test ^b			338.55			340.42
p-value			0.8362			0.8476
Serial Corr. ^c			0.26			-0.09
p-value			0.7969			0.9280
No. obs.	308	308	308	308	308	308
No. ° provinces	22	22	22	22	22	22
R ² (adj.)	0.84			0.84		

Notes: Dependent variable PTR is the ratio of provincial revenues to Geographic Gross Product (PBG).
Estimated Regressions:

$$PTR_{it} = \alpha + \beta_1 PTR_{it-1} + \beta_2 PTR_{it-2} + \beta_3 PTR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$PTR_{it} = \alpha + \beta_1 PTR_{it-1} + \beta_2 PTR_{it-2} + \beta_3 PTR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

Table 14: Elections and Revenue from Provincial Taxes

Equation	1	2	3
Estimation Method	OLS	FIXED EFFECTS.	GMM
ELE	0.0005 (0.66)	0.0004 (0.56)	0.0005 (0.66)
ELE+1	0.0009 (1.10)	0.0008 (1.12)	0.0011 (1.51)
F-test: <i>ELE = -ELE+1</i>	1.30	1.049	1.78
p-value	0.2559	0.3083	0.1823
F-test ^a		3.28	
p-value		0.0000	
Sargan test ^b			344.69
p-value			0.7708
Serial Corr ^c			-0.13
p-value			0.8958
No. obs.	304	304	302
No. provinces	22	22	22
R ² (adj.)	0.84		

Notes: Dependent variable PTR is the ratio of provincial revenues to Geographic Gross Product (PBG).

Estimated Regressions:

$$PTR_{it} = \alpha + \beta_1 PTR_{it-1} + \beta_2 PTR_{it-2} + \beta_3 PTR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$PTR_{it} = \alpha + \beta_1 PTR_{it-1} + \beta_2 PTR_{it-2} + \beta_3 PTR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as $N(0,1)$ under the null of no serial correlation.

VI. Conditional Findings: Political Alignment Between Provincial and Federal Executives

The results reported in the previous section suggest that there are some systematic increase in expenditures and federal revenues in electoral years, but no electoral or cyclical effects were detected over the budget balance. Decisions over spending are clearly taken at provincial level, but the federal revenues are not a decision variable for the provincial executive. If this is so, how can the provincial executive manipulate at the same time expenditures and federal revenues? What can explain this pattern?

In this section, we will focus on explaining these facts, looking for differences in the behavior of incumbents conditioning for the political alignment between the provincial and federal executive. Our conjecture is that when both executives are members of the same political party (political alignment), the more probable the federal executive increases the discretionary transfers to the province, allowing the provincial executive to increase spending without significant effects over the budget balance.

When both executives (provincial and federal) are not aligned, and with an aligned candidate running for the provincial election, the federal government is not interested in increasing the discretionary transfers to the incumbent. On the contrary, probably the federal government can reduce the transfers, rendering spending manipulations more difficult to the provincial executive and inducing budget deficits.

We then look at the sensitivity of the previous results when conditioned to political alignment between provincial and federal executives. The conditional election variables `ELE_UNAL`, `ELE_AL` and the conditional cycle variables `PBC_UNAL`, `PBC_AL` are now included in the regressions to estimate the differential effect of political alignment.

A. Budget Balance

Table 15 presents the results with the budget balance as the dependent variable. In columns 1 to 3 the coefficients estimates for the conditional election variable are presented. The coefficients associated to the unaligned provinces are all negative and significant at 10% in OLS and GMM regressions and marginally significant (11%) in FE. The election year has no significant effect over fiscal balance in aligned provinces.

The regression results indicates that while the election increases the deficit between 0.8 to 1.0 percentage points in unaligned provinces, the election effect is not relevant in aligned provinces.

Table 15: Elections and Fiscal Balance conditional on alignment of provincial and federal government

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE_UNAL	-0.0106 (-1.93)*	-0.0091 (-1.61)	-0.0084 (-1.80)*			
ELE_AL	-0.0003 (-0.09)	-0.0002 (-0.04)	0.0003 (0.09)			
PBC_UNAL				-0.0085 (-2.34)**	-0.0079 (-2.51)**	-0.0091 (-3.19)***
PBC_AL				-0.0052 (-2.31)**	-0.0048 (-1.97)*	-0.0043 (-1.82)*
F-test ^a		2.03			2.07	
p-value		0.0056			0.0045	
Sargan test ^b			283.40			284.92
p-value			0.9994			0.9993
Serial Corr ^c			-0.19			0.10
p-value			0.8472			0.9165
No. obs.	308	304	302	308	308	308
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.37			0.38		

Notes: Dependent variable DEF is ratio of government surplus to Geographic Gross Product (PBG).
Estimated Regressions:

$$DEF_{it} = \alpha + \beta_1 DEF_{it-1} + \beta_2 DEF_{it-2} + \beta_3 DEF_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$DEF_{it} = \alpha + \beta_1 DEF_{it-1} + \beta_2 DEF_{it-2} + \beta_3 DEF_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

The conditional PBC coefficient estimates, columns 4 to 6, result in all cases negative and significant for aligned and unaligned provinces. These results are driven mostly by an increase in budget surplus in post electoral years in both aligned and unaligned provinces, and are consistent with those obtained in the unconditional regressions.

B. Expenditures: Total and Composition

Table 16 shows the effects of the conditional electoral variables over total public expenditure. Results in columns 1 to 3 show that in electoral years, when the province is politically aligned with federal government spending rises significantly. Depending on estimation method, the increasing in spending ranges between 0.8 to 1.4 percentage points of GGP. For unaligned provinces the estimates are non-significant in all regressions.

The estimated coefficients for the conditional PBC variable are significant, but the effect is driven, as in the unconditional estimates, by the spending contraction in the post electoral period in both, aligned and unaligned provinces. Regarding the composition of expenditures, Table 17, in columns 1 to 3, shows a positive and significant increase in current spending in electoral years for unaligned provinces. However, there are no significant effects over spending composition in aligned provinces. The coefficients associated to the conditional cycle variable, in columns 4 to 6, are non significant in all specifications.

These results suggest that an important shift happens in the expenditure composition towards current goods in unaligned provinces during electoral years. The magnitude of the shift is approximately 1.9 percentage points of total expenditure, reassigned from investment to consumption goods.

Table 16: Elections and Total Expenditure conditional on alignment of provincial and federal government

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE_UNAL	0.0062 (0.70)	0.0005 (0.07)	0.0015 (0.28)			
ELE_AL	0.0140 (3.05)***	0.0082 (1.81)*	0.0101 (2.59)***			
PBC_UNAL				0.0107 (2.10)**	0.0073 (1.99)**	0.0034 (1.05)
PBC_AL				0.0068 (2.31)**	0.0049 (1.74)*	0.0045 (1.87)*
F-test ^a		11.29			11.42	
p-value		0.0000			0.0000	
Sargan test ^b			354.41			350.88
p-value			0.5287			0.5813
Serial Corr ^c			-0.61			-0.59
p-value			0.5406			0.5567
No. obs.	308	308	286	308	308	286
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.90			0.90		

Notes: Dependent variable TE is ratio of total provincial expenditure to Geographic Gross Product (PBG). Estimated Regressions:

$$TE_{it} = \alpha + \beta_1 TE_{it-1} + \beta_2 TE_{it-2} + \beta_3 TE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$TE_{it} = \alpha + \beta_1 TE_{it-1} + \beta_2 TE_{it-2} + \beta_3 TE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged two or more periods are used as instruments. One lag of the dependent variable is included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

Table 17: Elections and Composition Effect conditional on alignment of provincial and federal government

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE_UNAL	0.0190 (1.85)*	0.0195 (1.76)*	0.0198 (1.82)*			
ELE_AL	-0.0092 (-1.16)	-0.0072 (-0.92)	-0.0084 (-1.12)			
PBC_UNAL				0.0012 (0.20)	0.0031 (0.49)	0.0012 (0.20)
PBC_AL				-0.0068 (-1.41)**	-0.0058 (-1.19)	-0.0073 (-1.57)
F-test ^a		2.14			2.17	
p-value		0.0030			0.0026	
Sargan test ^b			235.44			230.71
p-value			1.0000			1.0000
Serial Corr ^c			-0.09			0.01
p-value			0.9250			0.9897
No. obs.	308	308	286	308	308	286
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.60			0.60		

Notes: Dependent variable CE is the ratio of current expenditure to total provincial expenditure.

Estimated Regressions:

$$CE_{it} = \alpha + \beta_1 CE_{it-1} + \beta_2 CE_{it-2} + \beta_3 CE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$CE_{it} = \alpha + \beta_1 CE_{it-1} + \beta_2 CE_{it-2} + \beta_3 CE_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$, t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged two or more periods are used as instruments. One lag of the dependent variable is included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

C. Revenues: Total, Federal and Provincial

Tables 18, 19 and 20 present the conditional estimates considering as dependent variables the total provincial revenue (TR), revenue from federal source (FR) and revenue from provincial taxes (PTR).

There is a strong positive relationship between elections in aligned provinces and federal revenues in the data, independent of the estimation technique. In politically aligned provinces, a positive and significant effect over revenues is present in electoral years, explained by the increase in discretionary federal revenues. The magnitude of the effect is important, from 0.8 to 1.6 percentage points of GGP of increase in federal revenues depending on the estimation technique.

By contrast, for the conditional election dummy in unaligned provinces, the estimated coefficients are non significant in all cases. The conditional cycle dummy is non significant in all regressions, suggesting that there is no evidence of cycles around elections, independently of political alignment.

Results in Table 20 are similar to those obtained in the unconditional regressions. For aligned or unaligned provinces, in all regressions revenue from provincial taxes is not sensitive to the election dummy or the PBC dummy.

To sum up, the findings reported above fit the conjectures about the behavior of federal and provincial governments considering the political alignment. If the provincial executive is aligned with the federal government, the discretionary transfers from this source are bigger in electoral years, and the provincial incumbent is able to increase the total expenditures proportionally, without increasing the fiscal deficit. Our empirical results show that discretionary transfers from the federal governments allows the provincial incumbent to increase the spending in 0.8 – 1.4 percentage points of GGP.

On the other side, if the provincial executive is unaligned, the federal transfers remain approximately constant. With constant revenues from provincial taxes, if the incumbent increases the spending he also increases the fiscal deficit, but in this case he is constrained by the borrowing alternatives. The other alternative action available to the incumbent is to change the expenditure composition, from investment goods to more visible consumption goods. Our empirical results suggest that this last alternative appears to be the more relevant discretionary decision for unaligned provincial executives. Estimates show that the redirection of spending toward consumption goods in electoral years is about 1.9 percentage points of total expenditure for unaligned provinces.

Table 18: Elections and Total Revenue conditional on alignment of provincial and federal government

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE_UNAL	0.0005 (0.08)	-0.0067 (-1.25)	-0.0061 (-1.20)			
ELE_AL	0.0162 (3.37)***	0.0108 (-2.85)***	0.0090 (2.57)*			
PBC_UNAL				-0.00005 (-0.01)	-0.0014 (-0.46)	-0.0057 (-1.96)*
PBC_AL				0.0039 (1.16)	0.0014 (0.62)	-0.0003 (-0.15)
F-test ^a		17.60			17.36	
p-value		0.0000			0.0000	
Sargan test ^b			364.10			373.00
p-value			0.3863			0.2693
Serial Corr ^c			-0.75			-1.09
p-value			0.4523			0.2770
No. obs.	308	308	286	308	308	286
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.89			0.89		

Notes: Dependent variable TR is the ratio of current expenditure to total provincial expenditure.

Estimated Regressions:

$$TR_{it} = \alpha + \beta_1 TR_{it-1} + \beta_2 TR_{it-2} + \beta_3 TR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$TR_{it} = \alpha + \beta_1 TR_{it-1} + \beta_2 TR_{it-2} + \beta_3 TR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$, t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged two or more periods are used as instruments. One lag of the dependent variable is included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

Table 19: Elections and Revenue from Federal Government⁶ conditional on alignment of provincial and federal government

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE_UNAL	0.0007 (0.11)	-0.0066 (-1.30)	-0.0064 (-1.35)			
ELE_AL	0.0166 (3.66)***	0.0110 (3.07)***	0.0082 (2.49)**			
PBC_UNAL				0.0003 (0.08)	-0.0014 (-0.50)	-0.0051 (-1.88)*
PBC_AL				0.0042 (1.30)	0.0015 (0.68)	-0.0004 (-0.21)
F-test ^a		15.94			15.72	
p-value		0.0000			0.0000	
Sargan test ^b			346.63			353.74
p-value			0.6431			0.5387
Serial Corr ^c			-0.36			-0.58
p-value			0.7156			0.5629
No. obs.	308	308	286	308	308	286
No. provinces	22	22	22	22	22	22
R ² (adj.)	0.91			0.90		

Notes: Dependent variable FR is the ratio of federal revenues to Gross Geographic Product (PBG).
Estimated Regressions:

$$FR_{it} = \alpha + \beta_1 FR_{it-1} + \beta_2 FR_{it-2} + \beta_3 FR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \gamma_4 ELE+1_{it} + \eta_i + \varepsilon_{it}$$

$$FR_{it} = \alpha + \beta_1 FR_{it-1} + \beta_2 FR_{it-2} + \beta_3 FR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged two or more periods are used as instruments. One lag of the dependent variable is included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

⁶Provincial revenues from revenue sharing ("coparticipation") plus special (discretionary) transfers from federal government.

Table 20: Elections and Revenue from Provincial Taxes conditional on alignment of provincial and federal government

Equation Estimation Method	1 OLS	2 FIXED EFFECTS.	3 GMM	4 OLS	5 FIXED EFF.	6 GMM
ELE_UNAL	0.0003 (0.29)	0.0001 (0.13)	0.0008 (0.82)			
ELE_AL	0.0001 (0.14)	0.0001 (0.19)	-0.0002 (-0.34)			
PBC_UNAL				0.0000 (0.00)	0.0000 (0.04)	0.0000 (0.15)
PBC_AL				-0.0004 (-0.98)	0.0003 (-0.68)	-0.0005 (-1.05)
F-test ^a		3.29			3.28	
p-value		0.0000			0.0000	
Sargan test ^b			337.35			338.36
p-value			0.8475			0.8380
Serial Corr ^c			0.33			-0.02
p-value			0.7379			0.9852
N° obs.	308	308	308	308	308	308
N° provinces	22	22	22	22	22	22
R ² (adj.)	0.84			0.84		

Notes: Dependent variable PTR is the ratio of provincial revenues to Geographic Gross Product (PBG). Estimated Regressions:

$$PTR_{it} = \alpha + \beta_1 PTR_{it-1} + \beta_2 PTR_{it-2} + \beta_3 PTR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 ELE_{it} + \eta_i + \varepsilon_{it}$$

$$PTR_{it} = \alpha + \beta_1 PTR_{it-1} + \beta_2 PTR_{it-2} + \beta_3 PTR_{it-3} + \gamma_1 PBG_{it} + \gamma_2 CREC_{it} + \gamma_3 PBC_{it} + \eta_i + \varepsilon_{it}$$

The coefficient estimates on the lagged dependent variables add up to a value less than unity. OLS imposes the restriction $\eta_i = \eta \forall i$. t statistics reported in parentheses, calculated using heteroskedastic-consistent standard errors for OLS.

In GMM estimation (Arellano-Bond One Step) z statistics in parentheses. The election dummy variables are treated as strictly exogenous. Variables CREC and PBG are treated as predetermined and levels lagged one or more periods are used as instruments. Two lags of the dependent variable are included.

* significant at the 10% level, ** significant at the 5% level, *** significant at the 1% level

(a) F-test is an F test of the null hypothesis that all province-specific effects in the FE-specification are equal. (b) P-values for rejecting the null hypothesis in test of the over identifying restrictions, asymptotically distributed as a χ^2 under the null hypothesis of instruments uncorrelated with the residuals. (c) P-values for rejecting the null hypothesis in test for second order serial correlation in the first-difference residuals, asymptotically distributed as N(0,1) under the null of no serial correlation.

VII. Conclusions

This paper presents empirical evidence of systematic effects in fiscal balance, public expenditures and revenues in Argentine provinces as a function of elections and political alignment. Our findings are consistent with the predictions of the theoretical literature on rational opportunist political cycles: there are fiscal policy manipulations during elections, and there is a strengthening of the policies after elections.

The data also reveals that there are important systematic differences between provinces in the size and composition of the electoral manipulations, depending on the political alignment with the federal executive. Specifically, the political alignment between provincial and federal executives implies more discretionary transfer of federal revenues and increases the election induced provincial spending without increasing the fiscal deficit. Politically unaligned provinces are constrained by constant federal transfers and fiscal deficits are more frequent in election years. In addition, an important spending switch from capital goods to consumption goods is present in election years for unaligned provinces.

Finally, we believe that our conditional findings fit the predictions of the theoretical models of opportunistic rational behavior and reveals that the institutional and political features are important issues to explain the electoral motivated policy cycles. It is highly likely that further work in the identification of institutional control variables as the effective division of powers and institutional development of the provinces could contribute to study the quantitative effects of electoral cycles more in depth.

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Appendix
Chapter 3
Econometric Estimation Output

TABLES 3 and 4 - Elections and Fiscal Balance

```
. *OLS
. regress def def_1 def_2 def_3 growth lggp ele, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(6, 301) =	21.28
Prob > F =	0.0000
R-squared =	0.3627
Root MSE =	.02471

def	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.4501836	.0612624	7.35	0.000	.3296268	.5707405
def_2	.0273253	.0706419	0.39	0.699	-.1116893	.1663398
def_3	.1371861	.0657659	2.09	0.038	.0077669	.2666053
grow	.0964675	.0203988	4.73	0.000	.056325	.1366099
lggp	.0008223	.0022823	0.36	0.719	-.003669	.0053136
ele	-.0036922	.0031471	-1.17	0.242	-.0098854	.002501
_cons	-.0178905	.0197735	-0.90	0.366	-.0568022	.0210212

```
. regress def def_1 def_2 def_3 grow lggp pbc, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(6, 301) =	25.03
Prob > F =	0.0000
R-squared =	0.3817
Root MSE =	.02434

def	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.4580231	.0589188	7.77	0.000	.3420781	.5739681
def_2	.0400662	.0694408	0.58	0.564	-.0965847	.1767171
def_3	.1196514	.0653006	1.83	0.068	-.0088522	.248155
grow	.0877438	.0202285	4.34	0.000	.0479367	.1275509
lggp	.0008499	.0022743	0.37	0.709	-.0036257	.0053254
pbc	-.0064528	.0020012	-3.22	0.001	-.0103908	-.0025147
_cons	-.0190134	.0196672	-0.97	0.334	-.057716	.0196891

```
. regress def def_1 def_2 def_3 grow lggp ele ele1, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(7, 300) =	23.47
Prob > F =	0.0000
R-squared =	0.3965
Root MSE =	.02408

def	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.4963217	.0581574	8.53	0.000	.3818737	.6107697
def_2	.043854	.0675271	0.65	0.517	-.0890327	.1767407
def_3	.0789868	.0658754	1.20	0.231	-.0506495	.2086232
grow	.0823988	.0196784	4.19	0.000	.0436736	.1211239
lggp	.0009685	.0022395	0.43	0.666	-.0034385	.0053756
ele	.0019633	.0032787	0.60	0.550	-.0044889	.0084155
ele1	.0140562	.0035301	3.98	0.000	.0071093	.0210031
_cons	-.0238065	.0195584	-1.22	0.224	-.0622954	.0146824

```
. test ele=-ele1
```

```

( 1)  ele + ele1 = 0

      F( 1, 300) = 8.27
      Prob > F = 0.0043

. regress def def_1 def_2 def_3 grow lggp ele_1, robust

Regression with robust standard errors                                Number of obs = 308
                                                                    F( 6, 301) = 22.53
                                                                    Prob > F = 0.0000
                                                                    R-squared = 0.3704
                                                                    Root MSE = .02456

-----+-----
      def |          Coef.      Robust Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
      def_1 |    .4554651    .0581458     7.83   0.000     .3410413     .5698889
      def_2 |    .0587623    .0716794     0.82   0.413    - .0822939     .1998184
      def_3 |    .0998277    .0707205     1.41   0.159    - .0393415     .2389969
      grow  |    .0965407    .0200904     4.81   0.000     .0570053     .136076
      lggp  |    .0008047    .0022336     0.36   0.719    - .0035908     .0052002
      ele_1 |   -.0079034    .0031948    -2.47   0.014    - .0141905    -.0016164
      _cons |   -.0168815    .0192939    -0.87   0.382    - .0548496     .0210865
-----+-----

.
.
. *FE
. xtreg def def_1 def_2 def_3 grow lggp ele, fe

Fixed-effects (within) regression                                Number of obs = 308
Group variable (i): id                                           Number of groups = 22

R-sq:  within = 0.2168                                           Obs per group: min = 14
      between = 0.1758                                           avg = 14.0
      overall  = 0.1926                                           max = 14

                                                                    F(6,280) = 12.92
corr(u_i, Xb) = -0.2074                                         Prob > F = 0.0000

-----+-----
      def |          Coef.      Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
      def_1 |    .3177049    .057287     5.55   0.000     .2049371     .4304727
      def_2 |   -.0530212    .0610658    -0.87   0.386    - .1732275     .0671851
      def_3 |    .024791     .0575514     0.43   0.667    - .0884974     .1380794
      grow  |    .0828221    .0199167     4.16   0.000     .0436166     .1220276
      lggp  |    .0237088    .0111388     2.13   0.034     .0017824     .0456352
      ele   |   -.0030769    .003379     -0.91   0.363    - .0097283     .0035745
      _cons |   -.2222733    .0965414    -2.30   0.022    - .4123124    -.0322341
-----+-----
      sigma_u |    .01619047
      sigma_e |    .02383394
      rho     |    .3157495   (fraction of variance due to u_i)
-----+-----

F test that all u_i=0:      F(21, 280) = 2.07      Prob > F = 0.0045

. xtreg def def_1 def_2 def_3 grow lggp pbc, fe

Fixed-effects (within) regression                                Number of obs = 308
Group variable (i): id                                           Number of groups = 22

R-sq:  within = 0.2402                                           Obs per group: min = 14
      between = 0.1815                                           avg = 14.0
      overall  = 0.2099                                           max = 14

```

corr(u_i, Xb) = -0.2075 F(6,280) = 14.75
 Prob > F = 0.0000

def	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.3254006	.0558988	5.82	0.000	.2153653	.4354359
def_2	-.039916	.0602769	-0.66	0.508	-.1585694	.0787373
def_3	.0106597	.0563708	0.19	0.850	-.1003047	.1216241
grow	.0739673	.0198411	3.73	0.000	.0349106	.1130241
lggp	.0240468	.0109716	2.19	0.029	.0024495	.0456441
pbc	-.005962	.0019374	-3.08	0.002	-.0097757	-.0021483
_cons	-.2258234	.0950851	-2.37	0.018	-.4129958	-.038651
sigma_u	.01616389					
sigma_e	.02347551					
rho	.32161581	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 2.07 Prob > F = 0.0045

. xtreg def def_1 def_2 def_3 grow lggp ele ele1,fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.2606 Obs per group: min = 14
 between = 0.1755 avg = 14.0
 overall = 0.2205 max = 14

corr(u_i, Xb) = -0.2158 F(7,279) = 14.05
 Prob > F = 0.0000

def	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.3626414	.0568463	6.38	0.000	.2507393	.4745436
def_2	-.0362964	.0595818	-0.61	0.543	-.1535834	.0809906
def_3	-.0304867	.0576452	-0.53	0.597	-.1439614	.0829881
grow	.0681588	.019719	3.46	0.001	.0293419	.1069756
lggp	.0246348	.0108446	2.27	0.024	.0032873	.0459824
ele	.00233	.0035477	0.66	0.512	-.0046536	.0093137
ele1	.0134629	.0033113	4.07	0.000	.0069445	.0199812
_cons	-.2346324	.0940198	-2.50	0.013	-.4197106	-.0495542
sigma_u	.01631482					
sigma_e	.02319926					
rho	.3309053	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 279) = 2.11 Prob > F = 0.0036

. test ele=-ele1

(1) ele + ele1 = 0

F(1, 279) = 7.71
 Prob > F = 0.0059

. xtreg def def_1 def_2 def_3 grow lggp ele_1, fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.2296 Obs per group: min = 14
 between = 0.1707 avg = 14.0
 overall = 0.1995 max = 14

corr(u_i, Xb) = -0.2044 F(6,280) = 13.91
 Prob > F = 0.0000

def	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.3204663	.0563163	5.69	0.000	.2096092	.4313233
def_2	-.0205643	.0623507	-0.33	0.742	-.1433	.1021714
def_3	-.0139053	.0583848	-0.24	0.812	-.1288342	.1010235
grow	.0827858	.0197053	4.20	0.000	.0439964	.1215752
lggp	.0236165	.0110473	2.14	0.033	.0018702	.0453629
ele_1	-.0081503	.0034749	-2.35	0.020	-.0149907	-.00131
_cons	-.2204952	.0957485	-2.30	0.022	-.4089734	-.032017
sigma_u	.01624652					
sigma_e	.02363812					
rho	.32082923	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 2.14 Prob > F = 0.0031

```
.
. *GMM
. xtabond def l(0).ele , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      81.59

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

def	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
def						
LD	.3408471	.0566994	6.01	0.000	.2297183	.4519759
L2D	-.0208584	.0558308	-0.37	0.709	-.1302847	.0885679
grow						
D1	.0475126	.046633	1.02	0.308	-.0438864	.1389115
LD	.016412	.0204578	0.80	0.422	-.0236846	.0565085
lggp						
D1	.04739	.0457473	1.04	0.300	-.0422731	.1370532
LD	-.0469277	.0442286	-1.06	0.289	-.1336142	.0397587
ele						
D1	-.0030125	.0030957	-0.97	0.330	-.00908	.003055
_cons	.0007439	.000439	1.69	0.090	-.0001166	.0016044

Sargan test of over-identifying restrictions:
chi2(365) = 283.73 Prob > chi2 = 0.9994

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -12.96 Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
H0: no autocorrelation z = -0.04 Pr > z = 0.9677

```
. xtabond def l(0).pbc , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      94.84

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

TABLES 5 and 6 - Elections and Total Expenditure

```
. *OLS
. regress te te_1 te_2 te_3 grow lgpp ele, robust
```

Regression with robust standard errors

Number of obs =	308
F(6, 301) =	127.26
Prob > F =	0.0000
R-squared =	0.8975
Root MSE =	.03556

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.8529785	.121196	7.04	0.000	.6144797	1.091477
te_2	-.1169038	.1443267	-0.81	0.419	-.400921	.1671133
te_3	.1654418	.0681497	2.43	0.016	.0313317	.299552
grow	-.1441292	.0274022	-5.26	0.000	-.1980534	-.090205
lgpp	.0015893	.0037383	0.43	0.671	-.0057672	.0089458
ele	.0114646	.00458	2.50	0.013	.0024518	.0204774
_cons	.0078443	.0348913	0.22	0.822	-.0608174	.0765061

```
. regress te te_1 te_2 te_3 grow lgpp pbc, robust
```

Regression with robust standard errors

Number of obs =	308
F(6, 301) =	115.39
Prob > F =	0.0000
R-squared =	0.8984
Root MSE =	.03542

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.8758927	.1205354	7.27	0.000	.638694	1.113091
te_2	-.1268693	.1412937	-0.90	0.370	-.4049177	.1511792
te_3	.1526218	.0657014	2.32	0.021	.0233295	.2819141
grow	-.1394992	.0269941	-5.17	0.000	-.1926203	-.0863781
lgpp	.0014448	.0036981	0.39	0.696	-.0058325	.0087221
pbc	.0082134	.0026115	3.15	0.002	.0030743	.0133525
_cons	.0120095	.0346679	0.35	0.729	-.0562126	.0802317

```
. regress te te_1 te_2 te_3 grow lgpp ele ele1, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	118.04
Prob > F =	0.0000
R-squared =	0.8984
Root MSE =	.03547

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.8744431	.1288108	6.79	0.000	.620956	1.12793
te_2	-.1257908	.1460104	-0.86	0.390	-.4131252	.1615435
te_3	.152989	.0654962	2.34	0.020	.0240987	.2818792
grow	-.1395424	.0269902	-5.17	0.000	-.1926564	-.0864284
lgpp	.0014549	.0036944	0.39	0.694	-.0058152	.0087251
ele	.0086581	.0053059	1.63	0.104	-.0017835	.0190996
ele1	-.0078012	.005399	-1.44	0.150	-.018426	.0028236
_cons	.0117077	.0342289	0.34	0.733	-.0556516	.0790669

```
. test ele=-ele1
```

```

( 1)  ele + ele1 = 0

      F( 1, 300) = 0.01
      Prob > F = 0.9271

. regress te te_1 te_2 te_3 grow lggp ele_1, robust

Regression with robust standard errors
Number of obs = 308
F( 6, 301) = 116.88
Prob > F = 0.0000
R-squared = 0.8958
Root MSE = .03586

```

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.866813	.1196696	7.24	0.000	.631318	1.102308
te_2	-.1362872	.1422076	-0.96	0.339	-.4161343	.1435598
te_3	.1712412	.0734229	2.33	0.020	.026754	.3157285
grow	-.1519528	.0270019	-5.63	0.000	-.2050892	-.0988164
lggp	.0014792	.0037438	0.40	0.693	-.0058881	.0088464
ele_1	.0017953	.0042057	0.43	0.670	-.0064809	.0100716
_cons	.0111131	.0350294	0.32	0.751	-.0578204	.0800465

```

.
.
. *FE
. xtreg te te_1 te_2 te_3 grow lggp ele,fe

```

```

Fixed-effects (within) regression
Group variable (i): id
Number of obs = 308
Number of groups = 22

R-sq:  within = 0.3810
       between = 0.7690
       overall = 0.6013
Obs per group: min = 14
               avg = 14.0
               max = 14

F(6,280) = 28.72
Prob > F = 0.0000
corr(u_i, Xb) = 0.6280

```

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.47562	.0472501	10.07	0.000	.3826095	.5686305
te_2	-.1371475	.0525041	-2.61	0.009	-.2405003	-.0337947
te_3	-.1069057	.0418403	-2.56	0.011	-.1892671	-.0245443
grow	-.1179495	.0225352	-5.23	0.000	-.1623093	-.0735896
lggp	-.0282712	.0127286	-2.22	0.027	-.0533271	-.0032152
ele	.0057427	.0038351	1.50	0.135	-.0018066	.013292
_cons	.4229095	.1145549	3.69	0.000	.1974114	.6484075
sigma_u	.08462865					
sigma_e	.02712363					
rho	.9068474	(fraction of variance due to u_i)				

```

F test that all u_i=0:      F(21, 280) = 11.30      Prob > F = 0.0000

```

```

. xtreg te te_1 te_2 te_3 grow lggp pbc,fe

```

```

Fixed-effects (within) regression
Group variable (i): id
Number of obs = 308
Number of groups = 22

R-sq:  within = 0.3902
       between = 0.7525
       overall = 0.5922
Obs per group: min = 14
               avg = 14.0
               max = 14

```

```
corr(u_i, Xb) = 0.6150
```

F(6,280)	=	29.86
Prob > F	=	0.0000

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.4894211	.0470915	10.39	0.000	.3967228	.5821193
te_2	-.1395606	.0516395	-2.70	0.007	-.2412115	-.0379096
te_3	-.1182188	.0418429	-2.83	0.005	-.2005853	-.0358522
grow	-.1124959	.0225231	-4.99	0.000	-.156832	-.0681599
lggp	-.0295899	.012621	-2.34	0.020	-.0544339	-.0047458
pbcc	.0057465	.0022564	2.55	0.011	.0013048	.0101881
_cons	.4357615	.1134771	3.84	0.000	.212385	.659138

sigma_u	.08470662	
sigma_e	.02692198	
rho	.90825408	(fraction of variance due to u_i)

```
F test that all u_i=0: F(21, 280) = 11.47 Prob > F = 0.0000
```

```
. xtreg te te_1 te_2 te_3 grow lggp ele ele1,fe
```

```
Fixed-effects (within) regression      Number of obs   =      308
Group variable (i): id                 Number of groups =      22

R-sq:  within = 0.3921                  Obs per group:  min =      14
      between = 0.7295                      avg   =     14.0
      overall  = 0.5751                      max   =      14
```

```
corr(u_i, Xb) = 0.5971
```

F(7,279)	=	25.71
Prob > F	=	0.0000

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.497696	.0479183	10.39	0.000	.4033686	.5920234
te_2	-.1474147	.052323	-2.82	0.005	-.2504126	-.0444168
te_3	-.1227199	.0421252	-2.91	0.004	-.2056435	-.0397963
grow	-.1116561	.0225456	-4.95	0.000	-.1560371	-.0672751
lggp	-.0307563	.0126846	-2.42	0.016	-.0557259	-.0057867
ele	.0025798	.0040573	0.64	0.525	-.0054069	.0105666
ele1	-.008665	.0038406	-2.26	0.025	-.0162252	-.0011049
_cons	.4482897	.1142822	3.92	0.000	.223325	.6732545

sigma_u	.08528202	
sigma_e	.02692765	
rho	.90934139	(fraction of variance due to u_i)

```
F test that all u_i=0: F(21, 279) = 11.51 Prob > F = 0.0000
```

```
. test ele=-ele1
```

```
( 1) ele + ele1 = 0
```

```
F( 1, 279) = 0.88
Prob > F = 0.3484
```

```
. xtreg te te_1 te_2 te_3 grow lggp ele_1, fe
```

```
Fixed-effects (within) regression      Number of obs   =      308
Group variable (i): id                 Number of groups =      22

R-sq:  within = 0.3782                  Obs per group:  min =      14
      between = 0.7292                      avg   =     14.0
      overall  = 0.5681                      max   =      14
```

```
corr(u_i, Xb) = 0.5971
```

F(6,280)	=	28.38
Prob > F	=	0.0000

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.4804651	.0474058	10.14	0.000	.3871481	.573782
te_2	-.1400705	.0528345	-2.65	0.008	-.2440737	-.0360672
te_3	-.1180047	.0440876	-2.68	0.008	-.2047899	-.0312195
grow	-.1215008	.0224652	-5.41	0.000	-.165723	-.0772786
lggp	-.0297466	.0127572	-2.33	0.020	-.0548588	-.0046344
ele_1	.0039702	.0040691	0.98	0.330	-.0040398	.0119802
_cons	.438336	.1146815	3.82	0.000	.2125887	.6640833
sigma_u	.08593566					
sigma_e	.02718584					
rho	.90902645	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 11.61 Prob > F = 0.0000

```
.
. *GMM
. xtabond te l(0).ele , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)        =      261.18

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

te	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
te						
LD	.6642248	.0520248	12.77	0.000	.562258	.7661915
L2D	-.2526925	.0519487	-4.86	0.000	-.35451	-.150875
grow						
D1	-.1017775	.0638095	-1.60	0.111	-.2268418	.0232868
LD	.1231811	.0287537	4.28	0.000	.0668249	.1795374
lggp						
D1	-.0872688	.062897	-1.39	0.165	-.2105447	.036007
LD	-.0131114	.0604702	-0.22	0.828	-.1316307	.105408
ele						
D1	.0100806	.0043157	2.34	0.020	.001622	.0185392
_cons	.0022204	.0005992	3.71	0.000	.0010461	.0033947

Sargan test of over-identifying restrictions:
chi2(365) = 254.35 Prob > chi2 = 1.0000

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -9.70 Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
H0: no autocorrelation z = 1.00 Pr > z = 0.3166

```
. xtabond te l(0).pbc , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)        =      262.49

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results


```
. xtabond te l(0).ele_1 , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))
```

Arellano-Bond dynamic panel-data estimation

Number of obs	=	308
Group variable (i): id	Number of groups	= 22
	Wald chi2(7)	= 250.92
Time variable (t): y	Obs per group: min	= 14
	avg	= 14
	max	= 14

te		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
te							
	LD	.6882782	.0531348	12.95	0.000	.5841358	.7924205
	L2D	-.2839229	.0512149	-5.54	0.000	-.3843022	-.1835436
grow							
	D1	-.087486	.064571	-1.35	0.175	-.2140429	.0390709
	LD	.1358809	.0288623	4.71	0.000	.0793119	.1924499
lggp							
	D1	-.1172843	.0627715	-1.87	0.062	-.2403143	.0057457
	LD	.011062	.0605704	0.18	0.855	-.1076537	.1297777
ele_1							
	D1	.0038208	.0046096	0.83	0.407	-.0052139	.0128555
_cons		.0021637	.0006088	3.55	0.000	.0009705	.003357

```

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
      H0: no autocorrelation      z = -10.29   Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
      H0: no autocorrelation      z =   1.72   Pr > z = 0.0848

```

TABLES 7 and 8 - Elections and Composition Effect

```
. *OLS
. regress ce ce_1 ce_2 ce_3 grow lgpp ele, robust
```

Regression with robust standard errors

Number of obs =	308
F(6, 301) =	68.17
Prob > F =	0.0000
R-squared =	0.5964
Root MSE =	.04868

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.597824	.0629698	9.49	0.000	.4739073	.7217407
ce_2	-.0213287	.0697938	-0.31	0.760	-.1586743	.116017
ce_3	.1615112	.0522125	3.09	0.002	.0587634	.264259
grow	.0682283	.0422724	1.61	0.108	-.0149585	.1514152
lgpp	-.0076872	.0052917	-1.45	0.147	-.0181006	.0027262
ele	.0000105	.0068845	0.00	0.999	-.0135374	.0135585
_cons	.2929964	.058347	5.02	0.000	.1781766	.4078161

```
. regress ce ce_1 ce_2 ce_3 grow lgpp pbc, robust
```

Regression with robust standard errors

Number of obs =	308
F(6, 301) =	67.85
Prob > F =	0.0000
R-squared =	0.5976
Root MSE =	.0486

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5959227	.0627891	9.49	0.000	.4723615	.7194839
ce_2	-.0103155	.0694357	-0.15	0.882	-.1469564	.1263255
ce_3	.154159	.0525895	2.93	0.004	.0506694	.2576485
grow	.0619434	.0422938	1.46	0.144	-.0212857	.1451724
lgpp	-.007643	.0052931	-1.44	0.150	-.0180592	.0027731
pbc	-.0038566	.0039057	-0.99	0.324	-.0115425	.0038293
_cons	.2911026	.0585234	4.97	0.000	.1759358	.4062695

```
. regress ce ce_1 ce_2 ce_3 grow lgpp ele ele1,robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	57.92
Prob > F =	0.0000
R-squared =	0.5998
Root MSE =	.04855

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.6108055	.0649001	9.41	0.000	.4830884	.7385227
ce_2	-.0137019	.0694034	-0.20	0.844	-.1502811	.1228773
ce_3	.1427931	.0535513	2.67	0.008	.0374092	.2481769
grow	.0602011	.0417522	1.44	0.150	-.0219632	.1423655
lgpp	-.0074531	.0052384	-1.42	0.156	-.0177616	.0028555
ele	.0039174	.0074204	0.53	0.598	-.0106852	.01852
ele1	.010833	.0066074	1.64	0.102	-.0021697	.0238357
_cons	.2854962	.0585507	4.88	0.000	.1702741	.4007183

```
. test ele=-ele1
```



```
( 1)  ele + ele1 = 0

      F( 1, 300) = 1.60
      Prob > F = 0.2073

. regress ce ce_1 ce_2 ce_3 grow lggp ele_1, robust

Regression with robust standard errors
```

	Number of obs = 308
	F(6, 301) = 74.50
	Prob > F = 0.0000
	R-squared = 0.6071
	Root MSE = .04802

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.6079164	.063583	9.56	0.000	.4827928	.7330399
ce_2	-.0003787	.0705679	-0.01	0.996	-.1392477	.1384903
ce_3	.1317214	.0521353	2.53	0.012	.0291255	.2343173
grow	.0686702	.0403819	1.70	0.090	-.0107963	.1481368
lggp	-.0076549	.0051607	-1.48	0.139	-.0178105	.0025007
ele_1	-.0197734	.0069432	-2.85	0.005	-.0334368	-.0061101
_cons	.2954732	.0567027	5.21	0.000	.1838893	.407057

```
.
. *FE
. xtreg ce ce_1 ce_2 ce_3 grow lggp ele,fe
```

```
Fixed-effects (within) regression      Number of obs   =      308
Group variable (i): id                 Number of groups =      22

R-sq:  within = 0.3691                  Obs per group:  min =      14
      between = 0.9242                      avg   =     14.0
      overall  = 0.5915                      max   =      14

                                     F(6,280)           =     27.30
corr(u_i, Xb) = 0.5274                 Prob > F          =     0.0000
```

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.4984037	.0555412	8.97	0.000	.3890724	.6077351
ce_2	-.0463806	.0640028	-0.72	0.469	-.1723684	.0796072
ce_3	.1227309	.0530716	2.31	0.021	.0182609	.2272008
grow	.0790595	.0386179	2.05	0.042	.0030412	.1550779
lggp	-.0134477	.0224803	-0.60	0.550	-.0576995	.030804
ele	.0013865	.0066562	0.21	0.835	-.011716	.014489
_cons	.4740116	.1804386	2.63	0.009	.1188232	.8291999
sigma_u	.02265106					
sigma_e	.04678539					
rho	.18988946	(fraction of variance due to u_i)				

```
F test that all u_i=0:      F(21, 280) = 2.18      Prob > F = 0.0025
```

```
. xtreg ce ce_1 ce_2 ce_3 grow lggp pbc,fe
```

```
Fixed-effects (within) regression      Number of obs   =      308
Group variable (i): id                 Number of groups =      22

R-sq:  within = 0.3699                  Obs per group:  min =      14
      between = 0.9239                      avg   =     14.0
      overall  = 0.5927                      max   =      14
```

```
corr(u_i, Xb) = 0.5258
```

F(6,280)	=	27.40
Prob > F	=	0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.4960342	.0550021	9.02	0.000	.3877641	.6043044
ce_2	-.0365872	.0640198	-0.57	0.568	-.1626085	.0894341
ce_3	.1183684	.0534819	2.21	0.028	.0130907	.223646
grow	.073767	.0389992	1.89	0.060	-.0030019	.150536
lggp	-.0136598	.0224256	-0.61	0.543	-.0578038	.0304843
pbcc	-.0025314	.003918	-0.65	0.519	-.0102439	.0051811
_cons	.4736657	.180086	2.63	0.009	.1191713	.82816
sigma_u	.02249727					
sigma_e	.04675418					
rho	.18800595	(fraction of variance due to u_i)				

```
F test that all u_i=0: F(21, 280) = 2.16 Prob > F = 0.0028
```

```
. xtreg ce ce_1 ce_2 ce_3 grow lggp ele ele1,fe
```

```
Fixed-effects (within) regression      Number of obs      =      308
Group variable (i): id                 Number of groups   =       22

R-sq:  within = 0.3733                  Obs per group: min =       14
      between = 0.9385                      avg =      14.0
      overall  = 0.5967                      max =       14
```

```
corr(u_i, Xb) = 0.5514
```

F(7,279)	=	23.74
Prob > F	=	0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5094768	.0560442	9.09	0.000	.3991537	.6198
ce_2	-.0405547	.0640459	-0.63	0.527	-.1666293	.0855199
ce_3	.1060174	.0543826	1.95	0.052	-.0010348	.2130696
grow	.0706142	.0390503	1.81	0.072	-.0062564	.1474848
lggp	-.0107449	.0225325	-0.48	0.634	-.0551002	.0336103
ele	.0046506	.0070621	0.66	0.511	-.0092511	.0185522
ele1	.0089791	.0065703	1.37	0.173	-.0039546	.0219128
_cons	.4473069	.1812163	2.47	0.014	.090582	.8040317
sigma_u	.02252725					
sigma_e	.04671307					
rho	.1886821	(fraction of variance due to u_i)				

```
F test that all u_i=0: F(21, 279) = 2.15 Prob > F = 0.0030
```

```
. test ele=-ele1
( 1) ele + ele1 = 0
```

```
F( 1, 279) = 1.49
Prob > F = 0.2228
```

```
. xtreg ce ce_1 ce_2 ce_3 grow lggp ele_1, fe
```

```
Fixed-effects (within) regression      Number of obs      =      308
Group variable (i): id                 Number of groups   =       22

R-sq:  within = 0.3877                  Obs per group: min =       14
      between = 0.9431                      avg =      14.0
      overall  = 0.6036                      max =       14
```

```
corr(u_i, Xb) = 0.5532
```

F(6,280)	=	29.55
Prob > F	=	0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5064715	.0543087	9.33	0.000	.3995664	.6133766
ce_2	-.0247478	.0624277	-0.40	0.692	-.1476349	.0981394
ce_3	.0898578	.0534849	1.68	0.094	-.0154258	.1951415
grow	.0772384	.0378411	2.04	0.042	.0027493	.1517276
lggp	-.0101332	.0221414	-0.46	0.648	-.053718	.0334516
ele_1	-.0194332	.0066483	-2.92	0.004	-.0325203	-.0063462
_cons	.4520359	.1776843	2.54	0.011	.1022692	.8018026
sigma_u	.02273794					
sigma_e	.04609108					
rho	.19573466	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 2.23 Prob > F = 0.0019

```
.
. *GMM
. xtabond ce l(0).ele , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =     142.57

Time variable (t): y                            Obs per group: min =      14
                                                avg =              14
                                                max =              14
```

One-step results

ce	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ce						
LD	.5204356	.0568682	9.15	0.000	.4089759	.6318952
L2D	-.0343043	.0569018	-0.60	0.547	-.1458298	.0772211
grow						
D1	.0892711	.0950972	0.94	0.348	-.097116	.2756583
LD	-.069051	.0411043	-1.68	0.093	-.149614	.0115121
lggp						
D1	.0052189	.0929107	0.06	0.955	-.1768827	.1873206
LD	-.0666234	.0900221	-0.74	0.459	-.2430635	.1098168
ele						
D1	-.0057964	.0064121	-0.90	0.366	-.0183639	.006771
_cons	.0049174	.001014	4.85	0.000	.00293	.0069048

Sargan test of over-identifying restrictions:
chi2(365) = 244.93 Prob > chi2 = 1.0000

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -12.35 Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
H0: no autocorrelation z = -0.89 Pr > z = 0.3739

```
. xtabond ce l(0).pbc , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =     145.08

Time variable (t): y                            Obs per group: min =      14
                                                avg =              14
                                                max =              14
```

One-step results


```
. xtabond ce l(0).ele_1 , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))
```

Arellano-Bond dynamic panel-data estimation

Number of obs	=	308
Group variable (i): id	Number of groups	= 22
	Wald chi2(7)	= 148.01
Time variable (t): y	Obs per group: min	= 14
	avg	= 14
	max	= 14

ce		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ce							
	LD	.5460397	.0559632	9.76	0.000	.4363538	.6557256
	L2D	-.0449919	.0550667	-0.82	0.414	-.1529206	.0629368
grow							
	D1	.0807695	.0943044	0.86	0.392	-.1040637	.2656027
	LD	-.0788889	.0406327	-1.94	0.052	-.1585274	.0007497
lggp							
	D1	.0262696	.0910579	0.29	0.773	-.1522006	.2047398
	LD	-.0790056	.0883866	-0.89	0.371	-.2522403	.094229
ele_1							
	D1	-.0148937	.0067283	-2.21	0.027	-.028081	-.0017064
_cons		.0046604	.0010169	4.58	0.000	.0026673	.0066536

```
chi2(365) = 242.60    Prob > chi2 = 1.0000
```

H0: no autocorrelation $z = -12.32$ $\Pr > z = 0.0000$

H0: no autocorrelation $z = -0.50$ $\Pr > z = 0.6165$

TABLES 9 and 10 - Elections and Total Revenue

```
. *OLS
. regress tr tr_1 tr_2 tr_3 grow lgpp ele, robust
```

Regression with robust standard errors

Number of obs =	308
F(6, 301) =	72.19
Prob > F =	0.0000
R-squared =	0.8920
Root MSE =	.0332

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	1.007357	.1669972	6.03	0.000	.678727	1.335987
tr_2	-.280931	.2239186	-1.25	0.211	-.7215751	.1597131
tr_3	.158464	.0864167	1.83	0.068	-.0115934	.3285214
grow	-.051079	.0274964	-1.86	0.064	-.1051884	.0030304
lgpp	.0007143	.0031351	0.23	0.820	-.0054552	.0068839
ele	.0111455	.004261	2.62	0.009	.0027604	.0195306
_cons	.0141335	.0306862	0.46	0.645	-.0462533	.0745202

```
. regress tr tr_1 tr_2 tr_3 grow lgpp pbc, robust
```

Regression with robust standard errors

Number of obs =	308
F(6, 301) =	65.81
Prob > F =	0.0000
R-squared =	0.8903
Root MSE =	.03347

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	1.011438	.1685758	6.00	0.000	.679702	1.343175
tr_2	-.299546	.2214875	-1.35	0.177	-.735406	.136314
tr_3	.1719821	.0842619	2.04	0.042	.0061651	.3377991
grow	-.053444	.0276693	-1.93	0.054	-.1078938	.0010058
lgpp	.0006089	.0031729	0.19	0.848	-.0056351	.0068528
pbc	.0024608	.002744	0.90	0.371	-.002939	.0078606
_cons	.0179009	.0312325	0.57	0.567	-.0435607	.0793626

```
. regress tr_1 tr_2 tr_3 grow lgpp ele ele1, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	77.98
Prob > F =	0.0000
R-squared =	0.8930
Root MSE =	.0331

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.9937757	.1704347	5.83	0.000	.6583768	1.329175
tr_2	-.2660834	.2282896	-1.17	0.245	-.7153353	.1831684
tr_3	.1580741	.0865007	1.83	0.069	-.0121509	.3282991
grow	-.0556565	.0270965	-2.05	0.041	-.1089799	-.0023332
lgpp	.0008512	.003169	0.27	0.788	-.005385	.0070873
ele	.0137865	.0047429	2.91	0.004	.0044529	.0231202
ele1	.0075717	.005269	1.44	0.152	-.0027971	.0179405
_cons	.0101636	.0305855	0.33	0.740	-.0500258	.070353

```
. test ele=-ele1
```

```

( 1) ele + ele1 = 0

      F( 1, 300) = 6.30
      Prob > F = 0.0126

. regress trtr_1 tr_2 tr_3 grow lggp ele_1, robust

Regression with robust standard errors                                     Number of obs = 308
                                                                           F( 6, 301) = 65.24
                                                                           Prob > F = 0.0000
                                                                           R-squared = 0.8901
                                                                           Root MSE = .03349

-----+-----
      tr_1 |      Coef.   Robust Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
      tr_1 |    1.001558   .1695455     5.91   0.000    .6679135    1.335203
      tr_2 |   -.2998994   .2223525    -1.35   0.178   -.7374615    .1376628
      tr_3 |    .1814686   .0844263     2.15   0.032    .0153281    .3476092
      grow |   -.05567    .0278415    -2.00   0.046   -.1104586   -.0008815
      lggp |    .0006317   .003206     0.20   0.844   -.0056773    .0069406
      ele_1 |   -.0029734   .0039099    -0.76   0.448   -.0106675    .0047207
      _cons |    .0183559   .0314147     0.58   0.559   -.0434643    .0801762
-----+-----

.
.
. *FE
. xtreg tr tr_1 tr_2 tr_3 grow lggp ele,fe

Fixed-effects (within) regression                                     Number of obs = 308
Group variable (i): id                                             Number of groups = 22

R-sq:  within = 0.4882                                           Obs per group: min = 14
      between = 0.9475                                           avg = 14.0
      overall  = 0.5713                                           max = 14

corr(u_i, Xb) = 0.6137                                           F(6,280) = 44.51
                                                                           Prob > F = 0.0000

-----+-----
      tr_1 |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
      tr_1 |    .5988355   .0416372    14.38   0.000    .5168737    .6807972
      tr_2 |   -.2727789   .0490768    -5.56   0.000   -.3693852   -.1761727
      tr_3 |   -.1350257   .0384525    -3.51   0.001   -.2107184   -.059333
      grow |   -.0559339   .0186706    -3.00   0.003   -.0926865   -.0191813
      lggp |   -.0076158   .0100646    -0.76   0.450   -.0274277   .0121962
      ele  |    .0051387   .003222     1.59   0.112   -.0012036    .0114811
      _cons |    .2346181   .0887453     2.64   0.009    .0599255    .4093107
-----+-----
      sigma_u |    .07997137
      sigma_e |    .02275159
      rho     |    .92512213   (fraction of variance due to u_i)
-----+-----

F test that all u_i=0:      F(21, 280) = 17.19      Prob > F = 0.0000

. xtreg tr tr_1 tr_2 tr_3 grow lggp pbc,fe

Fixed-effects (within) regression                                     Number of obs = 308
Group variable (i): id                                             Number of groups = 22

R-sq:  within = 0.4836                                           Obs per group: min = 14
      between = 0.9417                                           avg = 14.0
      overall  = 0.5554                                           max = 14

```

corr(u_i, Xb) = 0.6010 F(6,280) = 43.70
 Prob > F = 0.0000

tr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.5962759	.0420166	14.19	0.000	.5135674	.6789843
tr_2	-.2815121	.0489819	-5.75	0.000	-.3779317	-.1850926
tr_3	-.129584	.0385897	-3.36	0.001	-.2055468	-.0536211
grow	-.0580038	.0188756	-3.07	0.002	-.0951599	-.0208476
lggp	-.0078138	.0101177	-0.77	0.441	-.0277301	.0121025
pbc	.0004041	.0019049	0.21	0.832	-.0033457	.0041539
_cons	.2387625	.0891719	2.68	0.008	.0632301	.4142948
sigma_u	.08059363					
sigma_e	.02285287					
rho	.92557934	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 17.41 Prob > F = 0.0000

. xtreg trtr_1 tr_2 tr_3 grow lggp ele ele1,fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22
 R-sq: within = 0.4938 Obs per group: min = 14
 between = 0.9606 avg = 14.0
 overall = 0.5816 max = 14

corr(u_i, Xb) = 0.6222 F(7,279) = 38.87
 Prob > F = 0.0000

tr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.5906884	.0417431	14.15	0.000	.508517	.6728599
tr_2	-.2623382	.0492571	-5.33	0.000	-.3593009	-.1653756
tr_3	-.1337519	.0383177	-3.49	0.001	-.2091804	-.0583234
grow	-.0599213	.0187403	-3.20	0.002	-.0968115	-.023031
lggp	-.0062672	.010057	-0.62	0.534	-.0260644	.01353
ele	.0070712	.003394	2.08	0.038	.0003902	.0137522
ele1	.0054775	.0031235	1.75	0.081	-.0006712	.0116262
_cons	.220364	.088791	2.48	0.014	.0455785	.3951494
sigma_u	.07961596					
sigma_e	.02266774					
rho	.92501649	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 279) = 17.18 Prob > F = 0.0000

. test ele=-ele1

(1) ele + ele1 = 0

F(1, 279) = 5.59
 Prob > F = 0.0187

. xtregtrtr_1 tr_2 tr_3 grow lggp ele_1, fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22
 R-sq: within = 0.4846 Obs per group: min = 14
 between = 0.9470 avg = 14.0
 overall = 0.5572 max = 14

corr(u_i, Xb) = 0.6028 F(6,280) = 43.87
 Prob > F = 0.0000

it	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.5911466	.0420922	14.04	0.000	.5082893	.6740038
tr_2	-.2812704	.048933	-5.75	0.000	-.3775937	-.1849472
tr_3	-.1250099	.038729	-3.23	0.001	-.2012468	-.048773
grow	-.0579308	.0186826	-3.10	0.002	-.0947069	-.0211547
lggp	-.0073739	.0101109	-0.73	0.466	-.0272769	.0125291
ele_1	-.0024626	.0032909	-0.75	0.455	-.0089406	.0040154
_cons	.235504	.0890983	2.64	0.009	.0601165	.4108916
sigma_u	.08055516					
sigma_e	.02283188					
rho	.92564008	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 17.50 Prob > F = 0.0000

```
.
. *GMM
. xtabond it l(0).ele , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      268.76

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

it	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
it						
LD	.7139933	.0506127	14.11	0.000	.6147942	.8131923
L2D	-.2993771	.0483703	-6.19	0.000	-.3941812	-.204573
grow						
D1	-.0576599	.0589354	-0.98	0.328	-.1731712	.0578514
LD	.111362	.025856	4.31	0.000	.0606853	.1620388
lggp						
D1	-.0320426	.0577771	-0.55	0.579	-.1452835	.0811984
LD	-.053781	.0558419	-0.96	0.336	-.163229	.0556671
ele						
D1	.0095186	.003906	2.44	0.015	.001863	.0171741
_cons	.002413	.0005494	4.39	0.000	.0013363	.0034898

Sargan test of over-identifying restrictions:
chi2(365) = 270.19 Prob > chi2 = 0.9999

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -8.33 Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
H0: no autocorrelation z = 1.25 Pr > z = 0.2096

```
. xtabond it l(0).pbc , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      261.15

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

TABLES 11 and 12 - Elections and Revenue from Federal Government

```
. *OLS
. regress fr fr_1 fr_2 fr_3 grow lgpp ele, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(6, 301) =	131.61
Prob > F =	0.0000
R-squared =	0.9049
Root MSE =	.03052

fr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	1.004842	.1443267	6.96	0.000	.7208247	1.288859
fr_2	-.2122897	.1985607	-1.07	0.286	-.6030325	.1784532
fr_3	.1084967	.0781301	1.39	0.166	-.0452537	.2622471
grow	-.0498083	.025191	-1.98	0.049	-.0993811	-.0002355
lgpp	-.0004904	.003127	-0.16	0.875	-.006644	.0056632
ele	.0113934	.0039385	2.89	0.004	.0036429	.0191439
_cons	.0182307	.0305493	0.60	0.551	-.0418866	.0783479

```
. regress fr fr_1 fr_2 fr_3 grow lgpp pbc, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(6, 301) =	119.05
Prob > F =	0.0000
R-squared =	0.9030
Root MSE =	.03081

fr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	1.008407	.1462416	6.90	0.000	.7206219	1.296193
fr_2	-.2309958	.1985459	-1.16	0.246	-.6217095	.159718
fr_3	.1225613	.0766267	1.60	0.111	-.0282305	.2733532
grow	-.0518731	.0256435	-2.02	0.044	-.1023363	-.0014098
lgpp	-.0006274	.0031662	-0.20	0.843	-.0068581	.0056033
pbc	.0027726	.0025483	1.09	0.277	-.0022422	.0077874
_cons	.0223104	.0311374	0.72	0.474	-.0389641	.0835849

```
. regress fr fr_1 fr_2 fr_3 grow lgpp ele ele1,robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(7, 300) =	122.71
Prob > F =	0.0000
R-squared =	0.9058
Root MSE =	.03042

fr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	.9891585	.1484102	6.67	0.000	.6971016	1.281215
fr_2	-.1942327	.2037218	-0.95	0.341	-.5951374	.2066719
fr_3	.1068462	.0783029	1.36	0.173	-.0472463	.2609387
grow	-.0543071	.0250902	-2.16	0.031	-.1036823	-.0049319
lgpp	-.0003557	.0031491	-0.11	0.910	-.0065528	.0058414
ele	.0138578	.0042657	3.25	0.001	.0054633	.0222522
ele1	.0071179	.0046235	1.54	0.125	-.0019807	.0162165
_cons	.0144918	.0303717	0.48	0.634	-.0452767	.0742604

```
. test ele=-ele1
```

```

( 1)  ele + ele1 = 0

      F( 1, 300) = 8.03
      Prob > F = 0.0049

. regress fr fr_1 fr_2 fr_3 grow lggp ele_1, robust

Regression with robust standard errors                                     Number of obs = 308
                                                                              F( 6, 301) = 117.48
                                                                              Prob > F = 0.0000
                                                                              R-squared = 0.9029
                                                                              Root MSE = .03083

-----+-----
      fr |          Coef.   Robust Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
    fr_1 |    .9940371    .1472625     6.75   0.000    .7042428    1.283832
    fr_2 |   -.2296705    .2000504    -1.15   0.252   -.623345    .1640039
    fr_3 |    .1345786    .0779217     1.73   0.085   -.0187618    .287919
    grow |   -.054518    .0256805    -2.12   0.035   -.1050541   -.0039819
    lggp |   -.000635    .003203     -0.20   0.843   -.0069381    .0056681
    ele_1 |   -.0038346    .0037698    -1.02   0.310   -.011253    .0035838
    _cons |    .0232096    .0312747     0.74   0.459   -.0383351    .0847544
-----+-----

.
.
. *FE
. xtreg fr fr_1 fr_2 fr_3 grow lggp ele,fe

Fixed-effects (within) regression                                     Number of obs = 308
Group variable (i): id                                             Number of groups = 22

R-sq:  within = 0.4508                                           Obs per group: min = 14
      between = 0.8928                                           avg = 14.0
      overall = 0.6720                                           max = 14

corr(u_i, Xb) = 0.6993                                           F(6,280) = 38.30
                                                                  Prob > F = 0.0000

-----+-----
      fr |          Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
    fr_1 |    .5847009    .0443512    13.18   0.000    .4973967    .672005
    fr_2 |   -.2077448    .0521033    -3.99   0.000   -.3103086   -.105181
    fr_3 |   -.1474463    .039764    -3.71   0.000   -.2257207   -.0691719
    grow |   -.0452483    .0175997    -2.57   0.011   -.0798927   -.0106038
    lggp |   -.0154633    .0096178    -1.61   0.109   -.0343955    .003469
    ele   |    .0052995    .0030445     1.74   0.083   -.0006936    .0112925
    _cons |    .2719578    .0849833     3.20   0.002    .1046705    .4392452
-----+-----
    sigma_u |    .07463975
    sigma_e |    .02151267
    rho     |    .9233007   (fraction of variance due to u_i)
-----+-----

F test that all u_i=0:      F(21, 280) = 15.51      Prob > F = 0.0000

. xtreg fr fr_1 fr_2 fr_3 grow lggp pbc,fe

Fixed-effects (within) regression                                     Number of obs = 308
Group variable (i): id                                             Number of groups = 22

R-sq:  within = 0.4449                                           Obs per group: min = 14
      between = 0.8817                                           avg = 14.0
      overall = 0.6561                                           max = 14

```

corr(u_i, Xb) = 0.6870 F(6,280) = 37.41
 Prob > F = 0.0000

fr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	.5799352	.0448884	12.92	0.000	.4915736	.6682968
fr_2	-.2158632	.0521866	-4.14	0.000	-.3185911	-.1131353
fr_3	-.1418172	.0399137	-3.55	0.000	-.2203862	-.0632482
grow	-.0472567	.0178163	-2.65	0.008	-.0823276	-.0121857
lggp	-.0158394	.0096719	-1.64	0.103	-.0348782	.0031995
pbcc	.0004165	.0018057	0.23	0.818	-.003138	.003971
_cons	.2777569	.0854076	3.25	0.001	.1096343	.4458794
sigma_u	.0753887					
sigma_e	.0216267					
rho	.92396341	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 15.76 Prob > F = 0.0000

. xtreg fr fr_1 fr_2 fr_3 grow lggp ele ele1,fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22
 R-sq: within = 0.4579 Obs per group: min = 14
 between = 0.9099 avg = 14.0
 overall = 0.6823 max = 14

corr(u_i, Xb) = 0.7090 F(7,279) = 33.67
 Prob > F = 0.0000

fr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	.5738351	.0445019	12.89	0.000	.486233	.6614373
fr_2	-.1935724	.0523787	-3.70	0.000	-.2966799	-.0904649
fr_3	-.1474734	.0395749	-3.73	0.000	-.2253766	-.0695701
grow	-.0495832	.017661	-2.81	0.005	-.084349	-.0148175
lggp	-.0140664	.0095996	-1.47	0.144	-.0329633	.0048305
ele	.0072865	.0032021	2.28	0.024	.0009832	.0135897
ele1	.0056745	.0029569	1.92	0.056	-.0001462	.0114953
_cons	.2573784	.0849196	3.03	0.003	.0902138	.4245429
sigma_u	.07433849					
sigma_e	.02141034					
rho	.9234031	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 279) = 15.55 Prob > F = 0.0000

. test ele=-ele1

(1) ele + ele1 = 0

F(1, 279) = 6.69
 Prob > F = 0.0102

. xtreg fr fr_1 fr_2 fr_3 grow lggp ele_1, fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22
 R-sq: within = 0.4474 Obs per group: min = 14
 between = 0.8895 avg = 14.0
 overall = 0.6592 max = 14

corr(u_i, Xb) = 0.6903 F(6,280) = 37.79
 Prob > F = 0.0000

fr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	.5722638	.0446705	12.81	0.000	.4843313	.6601964
fr_2	-.215896	.0520624	-4.15	0.000	-.3183795	-.1134126
fr_3	-.1346562	.0401207	-3.36	0.001	-.2136326	-.0556798
grow	-.0471067	.0176013	-2.68	0.008	-.0817543	-.012459
lggp	-.0152872	.0096545	-1.58	0.114	-.0342919	.0037174
ele_1	-.0035662	.0031072	-1.15	0.252	-.0096826	.0025502
_cons	.273778	.0852407	3.21	0.001	.105984	.441572
sigma_u	.07535224					
sigma_e	.02157805					
rho	.92421147	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 15.93 Prob > F = 0.0000

```
.
. *GMM
. xtabond fr l(0).ele , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      267.16

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

fr	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
fr						
LD	.7235389	.052213	13.86	0.000	.6212032	.8258746
L2D	-.2624027	.0488318	-5.37	0.000	-.3581113	-.1666942
grow						
D1	-.0632153	.0539575	-1.17	0.241	-.1689701	.0425394
LD	.085278	.0236323	3.61	0.000	.0389594	.1315965
lggp						
D1	-.0199589	.0528216	-0.38	0.706	-.1234874	.0835696
LD	-.0494736	.0511049	-0.97	0.333	-.1496373	.0506901
ele						
D1	.0097197	.0035651	2.73	0.006	.0027323	.0167071
_cons	.0017769	.0005019	3.54	0.000	.0007931	.0027607

Sargan test of over-identifying restrictions:
chi2(365) = 254.86 Prob > chi2 = 1.0000

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -9.18 Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
H0: no autocorrelation z = 1.49 Pr > z = 0.1355

```
. xtabond fr l(0).pbc , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      257.36

Time variable (t): y                            Obs per group: min =      14
                                                avg =      14
                                                max =      14
```

One-step results

TABLES 13 and 14 - Elections and Revenue from Provincial Taxes

```
. *OLS
. regress ptr ptr_1 ptr_2 ptr_3 grow lggp ele, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(6, 301) =	132.20
Prob > F =	0.0000
R-squared =	0.8422
Root MSE =	.00561

ptr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.942762	.144488	6.52	0.000	.6584275	1.227096
ptr_2	-.3730599	.1568082	-2.38	0.018	-.6816391	-.0644808
ptr_3	.2536957	.0794738	3.19	0.002	.0973011	.4100903
grow	-.0033285	.0048578	-0.69	0.494	-.012888	.006231
lggp	.0026659	.0009725	2.74	0.006	.0007522	.0045796
ele	.0001689	.0007483	0.23	0.822	-.0013037	.0016416
_cons	-.0178248	.0074551	-2.39	0.017	-.0324954	-.0031541

```
. regress ptr ptr_1 ptr_2 ptr_3 grow lggp pbc, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(6, 301) =	128.07
Prob > F =	0.0000
R-squared =	0.8424
Root MSE =	.00561

ptr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.9453385	.144014	6.56	0.000	.6619368	1.22874
ptr_2	-.3809934	.1547589	-2.46	0.014	-.6855398	-.076447
ptr_3	.2592307	.0804054	3.22	0.001	.1010028	.4174587
grow	-.003823	.0047544	-0.80	0.422	-.0131792	.0055331
lggp	.0026665	.0009774	2.73	0.007	.000743	.0045899
pbc	-.0002816	.0005121	-0.55	0.583	-.0012894	.0007261
_cons	-.0177999	.007467	-2.38	0.018	-.0324942	-.0031057

```
. regress ptr ptr_1 ptr_2 ptr_3 grow lggp ele ele1, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(7, 300) =	114.91
Prob > F =	0.0000
R-squared =	0.8430
Root MSE =	.0056

ptr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.9415359	.1426356	6.60	0.000	.6608429	1.222229
ptr_2	-.3741504	.1539818	-2.43	0.016	-.6771716	-.0711292
ptr_3	.2574332	.0800395	3.22	0.001	.0999233	.4149431
grow	-.0039038	.0047272	-0.83	0.410	-.0132066	.0053989
lggp	.0026621	.0009674	2.75	0.006	.0007583	.0045659
ele	.0004969	.0007576	0.66	0.512	-.0009939	.0019878
ele1	.0009575	.0008671	1.10	0.270	-.0007489	.0026639
_cons	-.0181551	.0074988	-2.42	0.016	-.032912	-.0033983

```
. test ele=-ele1
```

```
( 1)  ele + ele1 = 0

      F( 1, 300) = 1.30
      Prob > F = 0.2559

. regress ptr ptr_1 ptr_2 ptr_3 grow lggp ele_1, robust

Regression with robust standard errors
```

	Number of obs = 308
	F(6, 301) = 125.68
	Prob > F = 0.0000
	R-squared = 0.8435
	Root MSE = .00559

ptr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.9546462	.1469553	6.50	0.000	.6654564	1.243836
ptr_2	-.3821857	.1534544	-2.49	0.013	-.6841649	-.0802064
ptr_3	.2529327	.0776832	3.26	0.001	.1000618	.4058037
grow	-.0037868	.0048006	-0.79	0.431	-.0132338	.0056603
lggp	.0026233	.0009849	2.66	0.008	.0006851	.0045615
ele_1	.001259	.0007054	1.78	0.075	-.0001291	.0026471
_cons	-.0177284	.0074827	-2.37	0.018	-.0324535	-.0030033

```
.
.
. *FE
. xtreg ptr ptr_1 ptr_2 ptr_3 grow lggp ele,fe
```

```
Fixed-effects (within) regression      Number of obs   =      308
Group variable (i): id                 Number of groups =      22

R-sq:  within = 0.5246                  Obs per group:  min =      14
      between = 0.9070                      avg   =     14.0
      overall  = 0.7956                      max   =      14

                                     F(6,280)          =     51.50
corr(u_i, Xb) = 0.5065                 Prob > F          =     0.0000
```

ptr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.7615754	.0479806	15.87	0.000	.667127	.8560239
ptr_2	-.3799457	.0562935	-6.75	0.000	-.490758	-.2691335
ptr_3	.1011362	.0482906	2.09	0.037	.0060776	.1961949
grow	-.006	.0044446	-1.35	0.178	-.0147491	.0027491
lggp	.0077	.0023265	3.31	0.001	.0031205	.0122796
ele	.0001654	.000737	0.22	0.823	-.0012853	.0016161
_cons	-.0515511	.0197267	-2.61	0.009	-.0903826	-.0127196
sigma_u	.00462312					
sigma_e	.00520647					
rho	.44086209	(fraction of variance due to u_i)				

```
F test that all u_i=0:      F(21, 280) = 3.30      Prob > F = 0.0000
```

```
. xtreg ptr ptr_1 ptr_2 ptr_3 grow lggp pbc,fe
```

```
Fixed-effects (within) regression      Number of obs   =      308
Group variable (i): id                 Number of groups =      22

R-sq:  within = 0.5250                  Obs per group:  min =      14
      between = 0.9060                      avg   =     14.0
      overall  = 0.7953                      max   =      14
```

corr(u_i, Xb) = 0.5007 F(6,280) = 51.58
 Prob > F = 0.0000

ptr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.763964	.0478951	15.95	0.000	.6696838	.8582441
ptr_2	-.3867665	.0560174	-6.90	0.000	-.4970352	-.2764978
ptr_3	.1059282	.0484006	2.19	0.029	.010653	.2012035
grow	-.006467	.0044717	-1.45	0.149	-.0152695	.0023355
lggp	.0077662	.002328	3.34	0.001	.0031836	.0123488
pbcc	-.0002204	.0004341	-0.51	0.612	-.0010749	.0006342
_cons	-.052095	.0197446	-2.64	0.009	-.0909617	-.0132284
sigma_u	.00461705					
sigma_e	.00520454					
rho	.44039739	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 3.30 Prob > F = 0.0000

. xtreg ptr ptr_1 ptr_2 ptr_3 grow lggp ele ele1,fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.5268 Obs per group: min = 14
 between = 0.9050 avg = 14.0
 overall = 0.7955 max = 14

corr(u_i, Xb) = 0.4922 F(7,279) = 44.36
 Prob > F = 0.0000

ptr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.7613549	.0479597	15.87	0.000	.6669461	.8557636
ptr_2	-.3811856	.0562794	-6.77	0.000	-.4919719	-.2703994
ptr_3	.1051559	.0484029	2.17	0.031	.0098746	.2004372
grow	-.006579	.0044727	-1.47	0.142	-.0153837	.0022256
lggp	.0078497	.0023293	3.37	0.001	.0032645	.0124349
ele	.0004351	.0007752	0.56	0.575	-.0010908	.001961
ele1	.0007915	.0007082	1.12	0.265	-.0006026	.0021857
_cons	-.0531774	.0197716	-2.69	0.008	-.0920978	-.014257
sigma_u	.00459919					
sigma_e	.00520415					
rho	.4385246	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 279) = 3.28 Prob > F = 0.0000

. test ele=-ele1

(1) ele + ele1 = 0

F(1, 279) = 1.04
 Prob > F = 0.3083

. xtreg ptr ptr_1 ptr_2 ptr_3 grow lggp ele_1, fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.5281 Obs per group: min = 14
 between = 0.9093 avg = 14.0
 overall = 0.7978 max = 14

corr(u_i, Xb) = 0.5150 F(6,280) = 52.21
 Prob > F = 0.0000

ptr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.7724179	.0481472	16.04	0.000	.6776415	.8671943
ptr_2	-.3878745	.0552707	-7.02	0.000	-.4966733	-.2790756
ptr_3	.1005916	.047777	2.11	0.036	.0065439	.1946393
grow	-.0064194	.004407	-1.46	0.146	-.0150944	.0022556
lggp	.0075677	.0023198	3.26	0.001	.0030012	.0121342
ele_1	.0010658	.0007395	1.44	0.151	-.0003899	.0025215
_cons	-.050653	.0196658	-2.58	0.011	-.0893645	-.0119414
sigma_u	.00461188					
sigma_e	.00518773					
rho	.44143963	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 280) = 3.28 Prob > F = 0.0000

```
.
. *GMM
. xtabond ptr l(0).ele , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      235.51

Time variable (t): y                            Obs per group: min =      14
                                                avg =              14
                                                max =              14
```

One-step results

ptr	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ptr						
LD	.6402529	.0477316	13.41	0.000	.5467006	.7338051
L2D	-.3534784	.04877	-7.25	0.000	-.4490658	-.257891
grow						
D1	.008156	.0106523	0.77	0.444	-.012722	.0290341
LD	.0218891	.0045774	4.78	0.000	.0129175	.0308607
lggp						
D1	-.0114323	.0105563	-1.08	0.279	-.0321221	.0092576
LD	-.0002762	.0102463	-0.03	0.978	-.0203585	.0198062
ele						
D1	.0001608	.0007098	0.23	0.821	-.0012303	.0015519
_cons	.0006998	.0001023	6.84	0.000	.0004992	.0009003

Sargan test of over-identifying restrictions:
chi2(365) = 338.55 Prob > chi2 = 0.8362

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -7.33 Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
H0: no autocorrelation z = 0.26 Pr > z = 0.7969

```
. xtabond ptr l(0).pbc , lags(2) pre(grow, lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(7)       =      238.17

Time variable (t): y                            Obs per group: min =      14
                                                avg =              14
                                                max =              14
```

One-step results

TABLE 15 - Elections and Fiscal Balance conditional on alignment of provincial and federal government

```
.
. *OLS
. regress def def_1 def_2 def_3 grow lgpp ele_nal ele_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	18.62
Prob > F =	0.0000
R-squared =	0.3681
Root MSE =	.02464

def	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.4480437	.0617306	7.26	0.000	.326564	.5695235
def_2	.0265503	.0704657	0.38	0.707	-.1121194	.16522
def_3	.1342168	.0660242	2.03	0.043	.0042876	.2641459
grow	.0958032	.0203864	4.70	0.000	.0556848	.1359216
lgpp	.0009753	.0023021	0.42	0.672	-.0035551	.0055057
ele_nal	-.0105915	.0054873	-1.93	0.055	-.02139	.000207
ele_al	-.0003116	.003355	-0.09	0.926	-.0069139	.0062907
_cons	-.0193228	.0199505	-0.97	0.334	-.0585834	.0199377

```
. test ele_nal=ele_al
( 1) ele_nal - ele_al = 0
F( 1, 300) = 2.95
Prob > F = 0.0869
```

```
.
. regress def def_1 def_2 def_3 grow lgpp pbc_nal pbc_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	21.55
Prob > F =	0.0000
R-squared =	0.3830
Root MSE =	.02435

def	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.4627852	.0584999	7.91	0.000	.3476631	.5779073
def_2	.0383323	.0692571	0.55	0.580	-.0979589	.1746235
def_3	.1185021	.0654454	1.81	0.071	-.0102881	.2472924
grow	.0884154	.0201296	4.39	0.000	.0488023	.1280285
lgpp	.0008753	.0022786	0.38	0.701	-.0036087	.0053594
pbc_nal	-.0085363	.0036535	-2.34	0.020	-.0157261	-.0013465
pbc_al	-.0052224	.0022566	-2.31	0.021	-.0096632	-.0007817
_cons	-.0192706	.0197101	-0.98	0.329	-.0580582	.0195169

```
. test pbc_al=pbc_nal
( 1) - pbc_nal + pbc_al = 0
F( 1, 300) = 0.61
Prob > F = 0.4344
```

```
.
. regress def def_1 def_2 def_3 grow lgpp ele_nal ele_al ele1_nal
ele1_al,robust
```


Regression with robust standard errors

Number of obs = 308
 F(9, 298) = 18.31
 Prob > F = 0.0000
 R-squared = 0.4024
 Root MSE = .02405

def	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.4906409	.058432	8.40	0.000	.3756493	.6056326
def_2	.0441476	.0669999	0.66	0.510	-.0877053	.1760005
def_3	.0758215	.0661612	1.15	0.253	-.0543809	.2060238
grow	.0810632	.0194898	4.16	0.000	.0427081	.1194183
lggp	.0011404	.0022679	0.50	0.615	-.0033228	.0056036
ele_nal	-.0050489	.0054465	-0.93	0.355	-.0157674	.0056695
ele_al	.0053321	.0035467	1.50	0.134	-.0016477	.0123118
elel_nal	.0126689	.0054326	2.33	0.020	.0019777	.0233601
elel_al	.0149723	.0038884	3.85	0.000	.0073201	.0226245
_cons	-.0254291	.0198264	-1.28	0.201	-.0644465	.0135883

```
.
.
. *FE
. xtreg def def_1 def_2 def_3 grow lggp ele_nal ele_al,fe
```

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.2217 Obs per group: min = 14
 between = 0.1807 avg = 14.0
 overall = 0.1970 max = 14

corr(u_i, Xb) = -0.2134 F(7,279) = 11.35
 Prob > F = 0.0000

def	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.3167075	.0572156	5.54	0.000	.2040785	.4293366
def_2	-.0525461	.0609854	-0.86	0.390	-.1725962	.0675039
def_3	.0233805	.0574846	0.41	0.685	-.0897782	.1365391
grow	.081999	.0198999	4.12	0.000	.042826	.121172
lggp	.0241361	.0111286	2.17	0.031	.0022293	.0460428
ele_nal	-.0091026	.0056708	-1.61	0.110	-.0202657	.0020604
ele_al	-.0001709	.0040272	-0.04	0.966	-.0080984	.0077566
_cons	-.2259702	.0964533	-2.34	0.020	-.4158388	-.0361017

sigma_u	.01618276	
sigma_e	.02380216	
rho	.31612005	(fraction of variance due to u_i)

F test that all u_i=0: F(21, 279) = 2.03 Prob > F = 0.0056

```
. test ele_nal=ele_al
```

```
( 1) ele_nal - ele_al = 0
```

F(1, 279) = 1.75
 Prob > F = 0.1872

```
.
. xtreg def def_1 def_2 def_3 grow lggp pbc_nal pbc_al,fe
```

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.2419 Obs per group: min = 14
 between = 0.1776 avg = 14.0
 overall = 0.2087 max = 14

corr(u_i, Xb) = -0.2181 F(7,279) = 12.72
 Prob > F = 0.0000

def	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.3298263	.0562128	5.87	0.000	.2191713	.4404813
def_2	-.0418431	.0603655	-0.69	0.489	-.1606727	.0769866
def_3	.0096215	.0564232	0.17	0.865	-.1014477	.1206908
grow	.0744957	.0198654	3.75	0.000	.0353906	.1136008
lggp	.0244439	.0109902	2.22	0.027	.0028096	.0460782
pbc_nal	-.0079502	.0031667	-2.51	0.013	-.014184	-.0017165
pbc_al	-.0047893	.0024372	-1.97	0.050	-.0095868	8.28e-06
_cons	-.2292877	.0952479	-2.41	0.017	-.4167835	-.0417919
sigma_u	.01627861					
sigma_e	.02349101					
rho	.32442006	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 279) = 2.07 Prob > F = 0.0046

. test pbc_nal=pbc_al

(1) pbc_nal - pbc_al = 0

F(1, 279) = 0.63
 Prob > F = 0.4279

. xtreg def def_1 def_2 def_3 grow lggp ele_nal ele_al elel_nal elel_al,fe

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.2661 Obs per group: min = 14
 between = 0.1835 avg = 14.0
 overall = 0.2266 max = 14

corr(u_i, Xb) = -0.2185 F(9,277) = 11.16
 Prob > F = 0.0000

def	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
def_1	.359114	.057382	6.26	0.000	.2461538	.4720742
def_2	-.0344669	.0596956	-0.58	0.564	-.1519815	.0830478
def_3	-.0320873	.0576502	-0.56	0.578	-.1455755	.0814008
grow	.0667094	.0197935	3.37	0.001	.0277446	.1056743
lggp	.0249521	.0108543	2.30	0.022	.0035848	.0463194
ele_nal	-.0041232	.0057258	-0.72	0.472	-.0153948	.0071483
ele_al	.0054027	.0041545	1.30	0.195	-.0027756	.0135811
elel_nal	.0123814	.004699	2.63	0.009	.0031312	.0216316
elel_al	.0142442	.0039573	3.60	0.000	.0064539	.0220345
_cons	-.2373958	.0941035	-2.52	0.012	-.4226447	-.0521468
sigma_u	.01625373					
sigma_e	.02319634					
rho	.32930193	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 277) = 2.06 Prob > F = 0.0048

.
 .
 . *GMM

lggp	LD	.0291702	.0203258	1.44	0.151	-.0106677	.0690081
	D1	.0237321	.0449345	0.53	0.597	-.064338	.1118021
pbc_nal	LD	-.0261715	.0433608	-0.60	0.546	-.1111572	.0588142
	D1	-.0090788	.0028501	-3.19	0.001	-.0146649	-.0034926
pbc_al	D1	-.0042633	.0023424	-1.82	0.069	-.0088544	.0003277
	D1	.0007186	.0004287	1.68	0.094	-.0001216	.0015588

Sargan test of over-identifying restrictions:

chi2(365) = 284.92 Prob > chi2 = 0.9993

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -11.96 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = 0.10 Pr > z = 0.9165

. test D1.pbc_nal=D1.pbc_al

(1) D.pbc_nal - D.pbc_al = 0

chi2(1) = 1.79
Prob > chi2 = 0.1810

.

. xtabond def l(0).ele_nal ele_al ele1_nal ele1_al, lags(2) pre(grow,
lag(1,.)) pre(lggp, lag(1,.))

Arellano-Bond dynamic panel-data estimation
Group variable (i): id

Number of obs = 308
Number of groups = 22

Wald chi2(10) = 106.17

Time variable (t): y

Obs per group: min = 14
avg = 14
max = 14

One-step results

def		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
def	LD	.3792731	.0566973	6.69	0.000	.2681484	.4903977
	L2D	-.0182932	.0549103	-0.33	0.739	-.1259153	.0893289
grow	D1	.0516656	.0459368	1.12	0.261	-.0383688	.1417
	LD	.0337473	.0204248	1.65	0.098	-.0062846	.0737793
lggp	D1	.0305162	.045106	0.68	0.499	-.0578899	.1189223
	LD	-.0331017	.0435865	-0.76	0.448	-.1185297	.0523264
ele_nal	D1	-.0034857	.0047616	-0.73	0.464	-.0128184	.0058469
	D1	.0048919	.003899	1.25	0.210	-.00275	.0125339
ele1_nal	D1	.0149939	.0047189	3.18	0.001	.0057451	.0242428
	D1	.0135665	.0039398	3.44	0.001	.0058446	.0212884
ele1_al	D1	.0007661	.0004355	1.76	0.079	-.0000874	.0016196

Sargan test of over-identifying restrictions:

chi2(365) = 276.69 Prob > chi2 = 0.9998

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -12.24 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = 1.08 Pr > z = 0.2792

TABLE 16 - Elections and Total Expenditure conditional on alignment of provincial and federal government.

```
. *OLS
. regress te te_1 te_2 te_3 grow lggp ele_nal ele_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	111.40
Prob > F =	0.0000
R-squared =	0.8978
Root MSE =	.03557

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.8539374	.1215124	7.03	0.000	.6148128	1.093062
te_2	-.119193	.1445241	-0.82	0.410	-.4036024	.1652165
te_3	.1676242	.0684653	2.45	0.015	.0328912	.3023573
grow	-.144725	.0272205	-5.32	0.000	-.1982923	-.0911577
lggp	.0016833	.0037043	0.45	0.650	-.0056065	.008973
ele_nal	.006153	.0088279	0.70	0.486	-.0112195	.0235254
ele_al	.0140205	.0045969	3.05	0.002	.0049742	.0230668
_cons	.0068581	.0346684	0.20	0.843	-.0613659	.0750822

```
.
. regress te te_1 te_2 te_3 grow lggp pbc_nal pbc_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	101.40
Prob > F =	0.0000
R-squared =	0.8985
Root MSE =	.03545

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.877509	.1210146	7.25	0.000	.639364	1.115654
te_2	-.1254512	.1414914	-0.89	0.376	-.4038924	.1529901
te_3	.1502828	.0648306	2.32	0.021	.0227026	.2778631
grow	-.1405096	.0272827	-5.15	0.000	-.1941994	-.0868198
lggp	.0014054	.0037177	0.38	0.706	-.0059106	.0087214
pbc_nal	.0107453	.0051104	2.10	0.036	.0006886	.020802
pbc_al	.0067668	.0029336	2.31	0.022	.0009938	.0125398
_cons	.0122931	.034825	0.35	0.724	-.0562391	.0808253

```
.
. regress te te_1 te_2 te_3 grow lggp ele_nal ele_al ele1_nal ele1_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(9, 298) =	96.27
Prob > F =	0.0000
R-squared =	0.8995
Root MSE =	.0354

te	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.8804366	.1304036	6.75	0.000	.623808	1.137065
te_2	-.1261447	.1471834	-0.86	0.392	-.4157952	.1635058
te_3	.1517482	.0650154	2.33	0.020	.0238007	.2796956
grow	-.14437	.0269612	-5.35	0.000	-.1974284	-.0913116
lggp	.0015851	.0036576	0.43	0.665	-.0056129	.0087831

ele_nal		.0033287	.0093622	0.36	0.722	-.0150957	.021753
ele_al		.0111605	.0052609	2.12	0.035	.0008072	.0215138
elel_nal		-.0153207	.0079202	-1.93	0.054	-.0309073	.0002659
elel_al		-.0027616	.0056323	-0.49	0.624	-.0138456	.0083225
_cons		.0097218	.0340566	0.29	0.775	-.0573001	.0767437

```

.
.
.
. *FE
. xtreg te te_1 te_2 te_3 grow lgpp ele_nal ele_al,fe

```

Fixed-effects (within) regression	Number of obs	=	308
Group variable (i): id	Number of groups	=	22

R-sq: within = 0.3833	Obs per group: min =	14
between = 0.7705	avg =	14.0
overall = 0.6013	max =	14

	F(7,279)	=	24.77
corr(u_i, Xb) = 0.6281	Prob > F	=	0.0000

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
te_1	.4770493	.0472696	10.09	0.000	.3839988 .5700997
te_2	-.1393148	.0525461	-2.65	0.008	-.2427519 -.0358777
te_3	-.1056939	.041856	-2.53	0.012	-.1880876 -.0233002
grow	-.1187113	.022547	-5.27	0.000	-.1630951 -.0743274
lgpp	-.0279396	.0127324	-2.19	0.029	-.0530034 -.0028758
ele_nal	.0004721	.0064768	0.07	0.942	-.0122776 .0132218
ele_al	.0082504	.0045688	1.81	0.072	-.0007434 .0172442
_cons	.4199735	.1145877	3.67	0.000	.1944072 .6455398
sigma_u	.08465536				
sigma_e	.02712268				
rho	.90690664	(fraction of variance due to u_i)			

F test that all u_i=0:	F(21, 279) =	11.29	Prob > F = 0.0000
------------------------	--------------	-------	-------------------

```

.
. xtreg te te_1 te_2 te_3 grow lgpp pbc_nal pbc_al,fe

```

Fixed-effects (within) regression	Number of obs	=	308
Group variable (i): id	Number of groups	=	22

R-sq: within = 0.3908	Obs per group: min =	14
between = 0.7483	avg =	14.0
overall = 0.5904	max =	14

	F(7,279)	=	25.57
corr(u_i, Xb) = 0.6123	Prob > F	=	0.0000

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
te_1	.4905815	.0472002	10.39	0.000	.3976677 .5834953
te_2	-.1387299	.0517279	-2.68	0.008	-.2405564 -.0369033
te_3	-.1196374	.0419784	-2.85	0.005	-.2022719 -.0370028
grow	-.1129704	.0225688	-5.01	0.000	-.1573971 -.0685437
lgpp	-.0299444	.0126541	-2.37	0.019	-.054854 -.0050349
pbc_nal	.0073168	.0036847	1.99	0.048	.0000635 .0145702
pbc_al	.0048557	.0027983	1.74	0.084	-.0006527 .0103642
_cons	.4387467	.1137557	3.86	0.000	.2148183 .6626752
sigma_u	.08470989				
sigma_e	.02695613				

```

rho | .90804906 (fraction of variance due to u_i)
-----
F test that all u_i=0:      F(21, 279) =      11.42      Prob > F = 0.0000
.
. xtreg te te_1 te_2 te_3 grow lggp ele_nal ele_al elel_nal elel_al,fe

Fixed-effects (within) regression      Number of obs      =      308
Group variable (i): id                  Number of groups    =      22

R-sq:  within = 0.4004                  Obs per group: min =      14
      between = 0.7163                      avg =      14.0
      overall  = 0.5684                      max =      14

corr(u_i, Xb) = 0.5865                  F(9,277)           =      20.56
                                          Prob > F           =      0.0000

```

te	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
te_1	.5054411	.0479296	10.55	0.000	.4110887	.5997936
te_2	-.1485521	.0521958	-2.85	0.005	-.251303	-.0458013
te_3	-.1251661	.0420609	-2.98	0.003	-.2079658	-.0423665
grow	-.1158163	.0225742	-5.13	0.000	-.160255	-.0713775
lggp	-.031298	.0126577	-2.47	0.014	-.0562155	-.0063804
ele_nal	-.0039277	.0065996	-0.60	0.552	-.0169194	.0090641
ele_al	.0055675	.0047452	1.17	0.242	-.0037738	.0149087
elel_nal	-.015121	.0054274	-2.79	0.006	-.0258051	-.0044369
elel_al	-.0044441	.0045695	-0.97	0.332	-.0134395	.0045513
_cons	.4521684	.1140153	3.97	0.000	.2277219	.6766149
sigma_u	.08531184					
sigma_e	.02683879					
rho	.90994223					(fraction of variance due to u_i)

```

F test that all u_i=0:      F(21, 277) =      11.50      Prob > F = 0.0000
.
.
. *GMM
. xtabond te l(0).ele_nal ele_al , lags(1) pre(grow, lag(2,.)) pre(lggp,
lag(2,.))

```

```

Arellano-Bond dynamic panel-data estimation      Number of obs      =      286
Group variable (i): id                  Number of groups    =      22

Wald chi2(9)           =      218.00

Time variable (t): y                  Obs per group: min =      13
                                          avg =      13
                                          max =      13

```

One-step results

te	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
te						
LD	.3542436	.0342517	10.34	0.000	.2871114	.4213757
grow						
D1	-.0610862	.0510332	-1.20	0.231	-.1611095	.0389371
LD	-.0134358	.0516954	-0.26	0.795	-.114757	.0878853
L2D	.0458159	.0205939	2.22	0.026	.0054526	.0861792
lggp						
D1	-.0744527	.0501499	-1.48	0.138	-.1727447	.0238394
LD	.0327572	.052699	0.62	0.534	-.070531	.1360454
L2D	-.0438775	.04678	-0.94	0.348	-.1355646	.0478095
ele_nal						
D1	.0015802	.005632	0.28	0.779	-.0094584	.0126187

te		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
te							
grow	LD	.3642079	.0349848	10.41	0.000	.2956389	.4327769
	D1	-.0624094	.0513818	-1.21	0.225	-.1631159	.0382971
	LD	-.007328	.0521566	-0.14	0.888	-.109553	.094897
	L2D	.0470332	.0206934	2.27	0.023	.0064748	.0875916
lggp							
	D1	-.0741531	.0503219	-1.47	0.141	-.1727822	.0244761
	LD	.0255778	.0534345	0.48	0.632	-.0791519	.1303074
	L2D	-.034831	.0478941	-0.73	0.467	-.1287016	.0590397
ele_nal							
	D1	-.0000804	.0058109	-0.01	0.989	-.0114695	.0113087
ele_al							
	D1	.010124	.0040844	2.48	0.013	.0021187	.0181294
elel_nal							
	D1	-.006566	.0048799	-1.35	0.178	-.0161304	.0029984
elel_al							
	D1	.0014815	.0039938	0.37	0.711	-.0063462	.0093092
_cons		.0031005	.0005069	6.12	0.000	.002107	.0040941

Sargan test of over-identifying restrictions:

chi2(357) = 352.06 Prob > chi2 = 0.5639

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -5.70 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -0.70 Pr > z = 0.4836

.

.

TABLE 17 - Elections and Composition Effect conditional on alignment of provincial and federal government

```
.
. *OLS
. regress ce ce_1 ce_2 ce_3 grow lggp ele_nal ele_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	61.36
Prob > F =	0.0000
R-squared =	0.6031
Root MSE =	.04835

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5995961	.061782	9.71	0.000	.478015	.7211771
ce_2	-.0213828	.0680687	-0.31	0.754	-.1553353	.1125698
ce_3	.1608774	.0507902	3.17	0.002	.0609272	.2608277
grow	.0693663	.0414834	1.67	0.096	-.012269	.1510017
lggp	-.0080006	.0052204	-1.53	0.126	-.0182739	.0022727
ele_nal	.0190341	.0103003	1.85	0.066	-.0012359	.0393041
ele_al	-.0092474	.0079625	-1.16	0.246	-.0249168	.006422
_cons	.2947636	.0573407	5.14	0.000	.1819227	.4076045

```
.
. regress ce ce_1 ce_2 ce_3 grow lggp pbc_nal pbc_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	59.98
Prob > F =	0.0000
R-squared =	0.5989
Root MSE =	.0486

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5976495	.0624706	9.57	0.000	.4747134	.7205857
ce_2	-.0133118	.0690227	-0.19	0.847	-.1491417	.1225181
ce_3	.1532459	.0522022	2.94	0.004	.0505171	.2559747
grow	.0601375	.0420054	1.43	0.153	-.0225249	.1428
lggp	-.0077647	.0052582	-1.48	0.141	-.0181124	.002583
pbc_nal	.001249	.0062511	0.20	0.842	-.0110525	.0135506
pbc_al	-.0068275	.0048365	-1.41	0.159	-.0163452	.0026903
_cons	.2940792	.0581322	5.06	0.000	.1796808	.4084777

```
. regress ce ce_1 ce_2 ce_3 grow lggp ele_nal ele_al ele1_nal ele1_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(9, 298) =	46.85
Prob > F =	0.0000
R-squared =	0.6068
Root MSE =	.04829

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.6113	.0640376	9.55	0.000	.4852768	.7373232
ce_2	-.0117046	.0684177	-0.17	0.864	-.1463477	.1229385
ce_3	.142987	.0521572	2.74	0.006	.0403439	.2456301
grow	.062752	.0411132	1.53	0.128	-.0181569	.1436609
lggp	-.0077523	.0051948	-1.49	0.137	-.0179754	.0024707
ele_nal	.0229098	.010766	2.13	0.034	.0017229	.0440968

ele_al	-	.0054081	.0084171	-0.64	0.521	-.0219725	.0111564
ele1_nal	-	.0138903	.008527	1.63	0.104	-.0028904	.0306711
ele1_al	-	.0087037	.0078404	1.11	0.268	-.006726	.0241333
_cons	-	.285859	.0582352	4.91	0.000	.1712548	.4004633

```

.
.
. regress ce ce_1 ce_2 ce_3 grow lgpp ele_1_nal ele_1_al, robust

```

Regression with robust standard errors	Number of obs =	307
	F(7, 299) =	63.23
	Prob > F =	0.0000
	R-squared =	0.6059
	Root MSE =	.04819

ce	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]
ce_1	.6061938	.0635952	9.53	0.000	.4810429 .7313447
ce_2	-.0017324	.0705638	-0.02	0.980	-.140597 .1371323
ce_3	.1347668	.052169	2.58	0.010	.0321018 .2374317
grow	.0683599	.0405587	1.69	0.093	-.0114568 .1481766
lgpp	-.0073873	.0050994	-1.45	0.148	-.0174227 .002648
ele_1_nal	-.0220308	.0109319	-2.02	0.045	-.043544 -.0005175
ele_1_al	-.017825	.007972	-2.24	0.026	-.0335135 -.0021366
_cons	.2932242	.0562071	5.22	0.000	.1826126 .4038359

```

.
.
. *FE
. xtreg ce ce_1 ce_2 ce_3 grow lgpp ele_nal ele_al,fe

```

Fixed-effects (within) regression	Number of obs =	308
Group variable (i): id	Number of groups =	22
R-sq: within = 0.3783	Obs per group: min =	14
between = 0.9245	avg =	14.0
overall = 0.5972	max =	14

corr(u_i, Xb) = 0.5238	F(7,279) =	24.25
	Prob > F =	0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ce_1	.5030042	.0552797	9.10	0.000	.3941859 .6118225
ce_2	-.0473224	.0636497	-0.74	0.458	-.172617 .0779723
ce_3	.1199319	.0527954	2.27	0.024	.016004 .2238597
grow	.0812057	.0384184	2.11	0.035	.005579 .1568324
lgpp	-.0143168	.0223597	-0.64	0.523	-.058332 .0296984
ele_nal	.0195942	.0111393	1.76	0.080	-.0023335 .041522
ele_al	-.0072665	.0078704	-0.92	0.357	-.0227595 .0082264
_cons	.4806433	.179468	2.68	0.008	.1273599 .8339267

sigma_u	.02246686
sigma_e	.04652605
rho	.1890887 (fraction of variance due to u_i)

F test that all u_i=0:	F(21, 279) =	2.14	Prob > F = 0.0030
------------------------	--------------	------	-------------------

```

. xtreg ce ce_1 ce_2 ce_3 grow lgpp pbc_nal pbc_al,fe

```

Fixed-effects (within) regression	Number of obs =	308
Group variable (i): id	Number of groups =	22

R-sq: within = 0.3728 Obs per group: min = 14
 between = 0.9206 avg = 14.0
 overall = 0.5932 max = 14

corr(u_i, Xb) = 0.5188 F(7,279) = 23.69
 Prob > F = 0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.4976502	.0549923	9.05	0.000	.3893977	.6059026
ce_2	-.0399575	.0640558	-0.62	0.533	-.1660516	.0861366
ce_3	.1173338	.0534622	2.19	0.029	.0120933	.2225743
grow	.0719694	.0390114	1.84	0.066	-.0048246	.1487634
lggp	-.0144201	.022424	-0.64	0.521	-.0585619	.0297217
pbclnal	.0031389	.0063482	0.49	0.621	-.0093575	.0156353
pbclal	-.0058203	.0048717	-1.19	0.233	-.0154103	.0037697
_cons	.4826527	.1801676	2.68	0.008	.1279922	.8373132
sigma_u	.02257763					
sigma_e	.04673016					
rho	.18925469	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 279) = 2.17 Prob > F = 0.0026

```
.
. xtreg ce ce_1 ce_2 ce_3 grow lggp ele_nal ele_al elel_nal elel_al,fe
```

Fixed-effects (within) regression Number of obs = 308
 Group variable (i): id Number of groups = 22

R-sq: within = 0.3827 Obs per group: min = 14
 between = 0.9392 avg = 14.0
 overall = 0.6028 max = 14

corr(u_i, Xb) = 0.5487 F(9,277) = 19.08
 Prob > F = 0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5139267	.0558756	9.20	0.000	.4039319	.6239214
ce_2	-.0404036	.0639201	-0.63	0.528	-.1662344	.0854273
ce_3	.1033572	.0541931	1.91	0.058	-.0033254	.2100397
grow	.0736409	.039063	1.89	0.060	-.003257	.1505389
lggp	-.0115245	.0224492	-0.51	0.608	-.0557172	.0326683
ele_nal	.0232328	.0114591	2.03	0.044	.0006748	.0457908
ele_al	-.004182	.0082449	-0.51	0.612	-.0204127	.0120487
elel_nal	.0107419	.0091966	1.17	0.244	-.0073623	.028846
elel_al	.007824	.0079533	0.98	0.326	-.0078327	.0234806
_cons	.4522695	.180622	2.50	0.013	.0967033	.8078357
sigma_u	.02229704					
sigma_e	.04652836					
rho	.1867579	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 277) = 2.09 Prob > F = 0.0040

```
.
. xtreg ce ce_1 ce_2 ce_3 grow lggp ele_l_nal ele_l_al,fe
```

Fixed-effects (within) regression Number of obs = 307
 Group variable (i): id Number of groups = 22

R-sq: within = 0.3874 Obs per group: min = 13
 between = 0.9402 avg = 14.0
 overall = 0.6023 max = 14

	F(7,278)	=	25.12
corr(u_i, Xb) = 0.5479	Prob > F	=	0.0000

ce	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
ce_1	.5036629	.0545102	9.24	0.000	.3963578	.610968
ce_2	-.0270155	.0627164	-0.43	0.667	-.1504749	.0964439
ce_3	.0968443	.0536343	1.81	0.072	-.0087365	.2024252
grow	.0764725	.0380743	2.01	0.046	.001522	.151423
lgpp	-.0093118	.022347	-0.42	0.677	-.0533026	.0346789
ele_1_nal	-.0253947	.0111714	-2.27	0.024	-.047386	-.0034034
ele_1_al	-.0151406	.0078785	-1.92	0.056	-.0306498	.0003686
_cons	.4433675	.1794741	2.47	0.014	.0900666	.7966684
sigma_u	.02271984					
sigma_e	.04624685					
rho	.19442504	(fraction of variance due to u_i)				

F test that all $u_i=0$: $F(21, 278) = 2.22$ Prob > F = 0.0019

```
.
. *GMM
. xtabond ce l(0).ele_nal ele_al , lags(1) pre(grow, lag(2,.)) pre(lggp,
lag(2,.))
```

```
Arellano-Bond dynamic panel-data estimation      Number of obs   =      286
Group variable (i): id                          Number of groups =       22
```

Wald chi2(9) = 116.02

```
Time variable (t): y                      Obs per group: min =    13
                                           avg =    13
                                           max =    13
```

One-step results

ce		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ce							
grow	LD	.4244163	.0481967	8.81	0.000	.3299525	.5188802
	D1	.0437274	.0978521	0.45	0.655	-.1480591	.2355139
	LD	-.0248147	.0993933	-0.25	0.803	-.219622	.1699925
lggp	L2D	-.1247953	.0389633	-3.20	0.001	-.2011619	-.0484286
	D1	.0241884	.0960115	0.25	0.801	-.1639907	.2123675
	LD	-.0579368	.1011588	-0.57	0.567	-.2562044	.1403308
ele_nal	L2D	-.0017364	.0898229	-0.02	0.985	-.177786	.1743132
	D1	.0198176	.010916	1.82	0.069	-.0015773	.0412126
	ele_al						
_cons	D1	-.0084013	.0074824	-1.12	0.262	-.0230664	.0062639
		.0035316	.0010719	3.29	0.001	.0014307	.0056326

Sargan test of over-identifying restrictions:
chi2(357) = 235.44 Prob > chi2 = 1.0000

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -8.11 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -0.09 Pr > z = 0.9250

```
. xtabond ce l(0).pbc_nal pbc_al , lags(1) pre(grow, lag(2,.)) pre(lggp, lag(2,.))
```

```

Arellano-Bond dynamic panel-data estimation      Number of obs      =      286
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(9)       =      112.68

Time variable (t): y                            Obs per group: min =      13
                                                avg =              13
                                                max =              13

```

One-step results

ce		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ce	LD	.41893	.0483384	8.67	0.000	.3241886	.5136715
grow	D1	.0658286	.0973504	0.68	0.499	-.1249746	.2566318
	LD	-.0115323	.0994981	-0.12	0.908	-.206545	.1834804
	L2D	-.1214897	.0391498	-3.10	0.002	-.1982219	-.0447574
lggp	D1	-.0120289	.0951552	-0.13	0.899	-.1985297	.1744719
	LD	-.0310876	.1026825	-0.30	0.762	-.2323417	.1701664
	L2D	.0020614	.0902111	0.02	0.982	-.1747491	.1788719
pbc_nal	D1	.0012497	.0062639	0.20	0.842	-.0110272	.0135267
pbc_al	D1	-.0073465	.0046828	-1.57	0.117	-.0165246	.0018316
_cons		.0036662	.0010789	3.40	0.001	.0015517	.0057808

```

Sargan test of over-identifying restrictions:
      chi2(357) =   230.71      Prob > chi2 = 1.0000

```

```

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
      H0: no autocorrelation      z =  -8.15      Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
      H0: no autocorrelation      z =   0.01      Pr > z = 0.9897

```

```

.
.
. xtabond ce l(0).ele_1_nal ele_1_al , lags(1) pre(grow, lag(2,.)) pre(lggp,
lag(2,.))

```

```

Arellano-Bond dynamic panel-data estimation      Number of obs      =      285
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(9)       =      114.89

Time variable (t): y                            Obs per group: min =      12
                                                avg =      12.95455
                                                max =      13

```

One-step results

ce		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ce	LD	.428159	.0488848	8.76	0.000	.3323466	.5239713
grow	D1	.0587193	.096841	0.61	0.544	-.1310856	.2485241
	LD	-.0271393	.0998154	-0.27	0.786	-.2227739	.1684954
	L2D	-.1077007	.0399224	-2.70	0.007	-.1859472	-.0294542
lggp	D1	.0098829	.0941948	0.10	0.916	-.1747356	.1945015
	LD	-.0340245	.1004902	-0.34	0.735	-.2309816	.1629326
	L2D	-.0076788	.0901485	-0.09	0.932	-.1843667	.169009
ele_1_nal	D1	-.0214231	.0109311	-1.96	0.050	-.0428476	1.45e-06
ele_1_al							

	D1		-.0089575	.0077136	-1.16	0.246	-.0240759	.0061609
_cons			.0034386	.0010808	3.18	0.001	.0013202	.005557

Sargan test of over-identifying restrictions:

chi2(357) = 233.23 Prob > chi2 = 1.0000

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -8.04 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -0.01 Pr > z = 0.9901

.

.

TABLE 18 - Elections and Total Revenue conditional on alignment of provincial and federal government

```
.
. *OLS
. regress tr tr_1 tr_2 tr_3 grow lgpp ele_nal ele_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	62.69
Prob > F =	0.0000
R-squared =	0.8932
Root MSE =	.03307

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	1.009453	.1677577	6.02	0.000	.6793225	1.339584
tr_2	-.2862726	.2245212	-1.28	0.203	-.7281084	.1555632
tr_3	.1631853	.0863944	1.89	0.060	-.0068305	.3332011
grow	-.0519202	.0272518	-1.91	0.058	-.1055491	.0017087
lgpp	.0008989	.0030511	0.29	0.768	-.0051053	.0069032
ele_nal	.0004781	.0063636	0.08	0.940	-.0120449	.0130011
ele_al	.0162604	.0048283	3.37	0.001	.0067587	.0257621
_cons	.0122643	.0301458	0.41	0.684	-.0470598	.0715883

```
.
. regress tr tr_1 tr_2 tr_3 grow lgpp pbc_nal pbc_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(7, 300) =	58.50
Prob > F =	0.0000
R-squared =	0.8905
Root MSE =	.0335

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	1.012568	.1688137	6.00	0.000	.6803589	1.344777
tr_2	-.3035379	.2220007	-1.37	0.173	-.7404138	.1333379
tr_3	.1742855	.0845343	2.06	0.040	.0079302	.3406408
grow	-.0524522	.0278074	-1.89	0.060	-.1071745	.0022701
lgpp	.0006539	.0031627	0.21	0.836	-.00557	.0068779
pbc_nal	-.0000464	.0047776	-0.01	0.992	-.0094483	.0093554
pbc_al	.0039302	.0033777	1.16	0.246	-.0027167	.0105772
_cons	.0175292	.0311409	0.56	0.574	-.043753	.0788114

```
.
. regress tr tr_1 tr_2 tr_3 grow lgpp ele_nal ele_al ele1_nal ele1_al, robust
```

Regression with robust standard errors

Number of obs =	308
F(9, 298) =	62.76
Prob > F =	0.0000
R-squared =	0.8945
Root MSE =	.03298

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.9943475	.1721787	5.78	0.000	.6555074	1.333188
tr_2	-.2661339	.2315013	-1.15	0.251	-.7217184	.1894505
tr_3	.1605942	.0873215	1.84	0.067	-.0112508	.3324392
grow	-.0585203	.0272922	-2.14	0.033	-.11223	-.0048105
lgpp	.0010462	.0030872	0.34	0.735	-.0050293	.0071216

ele_nal		.0031448	.0068274	0.46	0.645	-.0102913	.0165809
ele_al		.0189662	.0052129	3.64	0.000	.0087075	.0292249
elel_nal		.0038419	.0078429	0.49	0.625	-.0115927	.0192765
elel_al		.0102612	.0058747	1.75	0.082	-.0012999	.0218224
_cons		.0078894	.0300726	0.26	0.793	-.0512923	.067071

```

.
.
. *FE
. xtreg tr tr_1 tr_2 tr_3 grow lgpp ele_nal ele_al,fe

```

Fixed-effects (within) regression	Number of obs	=	308
Group variable (i): id	Number of groups	=	22

R-sq: within = 0.5016	Obs per group: min =	14
between = 0.9384	avg =	14.0
overall = 0.5627	max =	14

	F(7,279)	=	40.12
corr(u_i, Xb) = 0.6027	Prob > F	=	0.0000

tr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.6013849	.0411699	14.61	0.000	.5203418	.6824281
tr_2	-.2775477	.0485447	-5.72	0.000	-.3731081	-.1819873
tr_3	-.1328645	.0380194	-3.49	0.001	-.2077059	-.0580231
grow	-.0574305	.0184644	-3.11	0.002	-.0937778	-.0210832
lgpp	-.0069166	.0099524	-0.69	0.488	-.0265079	.0126748
ele_nal	-.0067297	.0053694	-1.25	0.211	-.0172995	.00384
ele_al	.0108026	.0037947	2.85	0.005	.0033327	.0182725
_cons	.2286671	.0877537	2.61	0.010	.0559237	.4014106

sigma_u	.08015312
sigma_e	.02249053
rho	.92701321 (fraction of variance due to u_i)

F test that all u_i=0:	F(21, 279) =	17.60	Prob > F = 0.0000
------------------------	--------------	-------	-------------------

```

. xtreg tr tr_1 tr_2 tr_3 grow lgpp pbc_nal pbc_al,fe

```

Fixed-effects (within) regression	Number of obs	=	308
Group variable (i): id	Number of groups	=	22

R-sq: within = 0.4847	Obs per group: min =	14
between = 0.9476	avg =	14.0
overall = 0.5585	max =	14

	F(7,279)	=	37.48
corr(u_i, Xb) = 0.6039	Prob > F	=	0.0000

tr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.5974332	.0420769	14.20	0.000	.5146046	.6802617
tr_2	-.2845449	.0491842	-5.79	0.000	-.3813642	-.1877256
tr_3	-.1274304	.0387248	-3.29	0.001	-.2036604	-.0512004
grow	-.0575113	.0189014	-3.04	0.003	-.0947189	-.0203038
lgpp	-.0072853	.0101497	-0.72	0.473	-.0272649	.0126944
pbc_nal	-.0014398	.0030999	-0.46	0.643	-.0075419	.0046623
pbc_al	.0014785	.0023797	0.62	0.535	-.0032059	.0061629
_cons	.2340775	.0894564	2.62	0.009	.0579824	.4101727

sigma_u	.08050632
sigma_e	.02287047
rho	.92532352 (fraction of variance due to u_i)

F test that all u i=0: $F(21, 279) = 17.36$ Prob > F = 0.0000

```
. regress tr tr_1 tr_2 tr_3 grow lgpp ele_nal ele_al ele1_nal ele1_al,robust
```

```

Regression with robust standard errors
Number of obs =      308
F(  9,   298) =    62.76
Prob > F      =    0.0000
R-squared     =    0.8945
Root MSE     =    .03298

```

tr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Intervall
tr_1	.9943475	.1721787	5.78	0.000	.6555074 1.333188
tr_2	-.2661339	.2315013	-1.15	0.251	-.7217184 .1894505
tr_3	.1605942	.0873215	1.84	0.067	-.0112508 .3324392
grow	-.0585203	.0272922	-2.14	0.033	-.11223 -.0048105
lggp	.0010462	.0030872	0.34	0.735	-.0050293 .0071216
ele1_nal	.0031448	.0068274	0.46	0.645	-.0102913 .0165809
ele1_al	.0189662	.0052129	3.64	0.000	.0087075 .0292249
ele1_nal	.0038419	.0078429	0.49	0.625	-.0115927 .0192765
ele1_al	.0102612	.0058747	1.75	0.082	-.0012999 .0218224
_cons	.0078894	.0300726	0.26	0.793	-.0512923 .067071

•

.

* GMM

```
. xtabond tr l(0).ele_nal ele_al , lags(1) pre(grow, lag(2,.)) pre(lggp,
lag(2,.))
```

```
Arellano-Bond dynamic panel-data estimation      Number of obs   =      286
Group variable (i): id                          Number of groups =       22
```

Wald chi2(9) = 183.08

```
Time variable (t): y          Obs per group: min =    13
                               avg =    13
                               max =    13
```

One-step results

tr		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
tr						
grow	LD	.3500565	.0315188	11.11	0.000	.2882808 .4118322
	D1	.0255954	.0458895	0.56	0.577	-.0643463 .1155372
	LD	.0594925	.0465128	1.28	0.201	-.0316708 .1506559
lggp	L2D	.0162432	.0187645	0.87	0.387	-.0205345 .0530209
	D1	-.056169	.0450179	-1.25	0.212	-.1444025 .0320646
	LD	-.0451783	.0473647	-0.95	0.340	-.1380113 .0476548
	L2D	.0365996	.0421273	0.87	0.385	-.0459684 .1191676
ele_nal						
	D1	-.0061026	.0051061	-1.20	0.232	-.0161104 .0039052
ele_al						
	D1	.0090788	.0035288	2.57	0.010	.0021624 .0159951
_cons		.0033951	.000453	7.49	0.000	.0025072 .004283

Sargan test of over-identifying restrictions:
chi2(357) = 364.10 Prob > chi2 = 0.3863

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
H0: no autocorrelation z = -4.47 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -0.75 Pr > z = 0.4523

```
. xtabond tr l(0).pbc_nal pbc_al , lags(1) pre(grow, lag(2,.)) pre(lggp, lag(2,.))
```

```
Arellano-Bond dynamic panel-data estimation      Number of obs      =      286
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(9)       =     179.14

Time variable (t): y                            Obs per group: min =      13
                                                avg =             13
                                                max =             13
```

One-step results

tr		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
tr	LD	.3357761	.0313136	10.72	0.000	.2744025	.3971497
grow	D1	.032752	.045178	0.72	0.468	-.0557953	.1212992
	LD	.0583417	.0461052	1.27	0.206	-.0320229	.1487062
	L2D	.012797	.0186396	0.69	0.492	-.023736	.04933
lggp	D1	-.0671218	.0441787	-1.52	0.129	-.1537104	.0194668
	LD	-.0295629	.0476002	-0.62	0.535	-.1228577	.0637318
	L2D	.0305214	.0419206	0.73	0.467	-.0516415	.1126843
pbc_nal	D1	-.0057203	.002923	-1.96	0.050	-.0114493	8.76e-06
pbc_al	D1	-.0003357	.0021771	-0.15	0.877	-.0046029	.0039314
_cons		.0034825	.0004526	7.69	0.000	.0025953	.0043696

```
Sargan test of over-identifying restrictions:
      chi2(357) =    373.00      Prob > chi2 = 0.2693
```

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -4.21 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -1.09 Pr > z = 0.2770

```
.
. xtabond tr l(0).ele_nal ele_al elel_nal elel_al, lags(1) pre(grow, lag(2,.)) pre(lggp, lag(2,.))
```

```
Arellano-Bond dynamic panel-data estimation      Number of obs      =      286
Group variable (i): id                          Number of groups   =      22

                                                Wald chi2(11)     =     212.53

Time variable (t): y                            Obs per group: min =      13
                                                avg =             13
                                                max =             13
```

One-step results

tr		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
tr	LD	.3475647	.0306374	11.34	0.000	.2875165	.4076128
grow	D1	.009285	.0446843	0.21	0.835	-.0782947	.0968646
	LD	.0445231	.0453884	0.98	0.327	-.0444364	.1334827
	L2D	.0122236	.0182512	0.67	0.503	-.0235482	.0479954
lggp	D1	-.0521419	.0436858	-1.19	0.233	-.1377644	.0334807
	LD	-.0207505	.0464785	-0.45	0.655	-.1118467	.0703457

	L2D		.0066651	.0417441	0.16	0.873	-.0751518	.088482
ele_nal								
	D1		-.0024982	.0050867	-0.49	0.623	-.0124679	.0074715
ele_al								
	D1		.01353	.0035753	3.78	0.000	.0065226	.0205374
ele1_nal								
	D1		.0089831	.0041688	2.15	0.031	.0008125	.0171537
ele1_al								
	D1		.013825	.0034746	3.98	0.000	.007015	.020635
_cons			.0034533	.0004423	7.81	0.000	.0025864	.0043202

Sargan test of over-identifying restrictions:

chi2(357) = 369.15 Prob > chi2 = 0.3175

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -4.46 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -0.58 Pr > z = 0.5640

.

.

TABLE 19 - Elections and Revenue from Federal Government conditional on alignment of provincial and federal government

```
. *OLS
. regress fr fr_1 fr_2 fr_3 grow lggp ele_nal ele_al, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(7, 300) =	115.80
Prob > F	= 0.0000
R-squared	= 0.9061
Root MSE	= .03036

fr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	1.00824	.1446013	6.97	0.000	.7236783	1.292801
fr_2	-.2187173	.1986039	-1.10	0.272	-.6095505	.172116
fr_3	.1128744	.0775378	1.46	0.147	-.0397124	.2654612
grow	-.0507363	.0249693	-2.03	0.043	-.0998733	-.0015992
lggp	-.0002838	.0030505	-0.09	0.926	-.0062868	.0057192
ele_nal	.0006585	.0058137	0.11	0.910	-.0107823	.0120993
ele_al	.0165713	.0045323	3.66	0.000	.0076522	.0254904
_cons	.016237	.0300383	0.54	0.589	-.0428754	.0753494

```
.
. regress fr fr_1 fr_2 fr_3 grow lggp pbc_nal pbc_al, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(7, 300) =	103.34
Prob > F	= 0.0000
R-squared	= 0.9032
Root MSE	= .03083

fr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	1.010338	.1464464	6.90	0.000	.722146	1.298531
fr_2	-.2361745	.1991115	-1.19	0.237	-.6280066	.1556575
fr_3	.1252573	.0767807	1.63	0.104	-.0258397	.2763543
grow	-.0509198	.025696	-1.98	0.048	-.101487	-.0003526
lggp	-.0005945	.0031477	-0.19	0.850	-.0067889	.0055998
pbc_nal	.0003441	.0041343	0.08	0.934	-.0077919	.00848
pbc_al	.0042087	.0032478	1.30	0.196	-.0021827	.0106002
_cons	.0220303	.0309912	0.71	0.478	-.0389573	.083018

```
. regress fr fr_1 fr_2 fr_3 grow lggp ele_nal ele_al ele1_nal ele1_al, robust
```

```
Regression with robust standard errors
```

Number of obs =	308
F(9, 298) =	99.72
Prob > F	= 0.0000
R-squared	= 0.9074
Root MSE	= .03027

fr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
fr_1	.9900392	.149949	6.60	0.000	.6949461	1.285132
fr_2	-.1932028	.2066576	-0.93	0.351	-.5998959	.2134902
fr_3	.1078675	.0783245	1.38	0.169	-.0462716	.2620066
grow	-.0573516	.0251816	-2.28	0.023	-.1069079	-.0077953
lggp	-.0001107	.0030843	-0.04	0.971	-.0061803	.005959
ele_nal	.0031823	.0061115	0.52	0.603	-.0088448	.0152094

ele_al		.0190761	.0048024	3.97	0.000	.0096253	.028527
elel_nal		.0032	.0066132	0.48	0.629	-.0098146	.0162145
elel_al		.009921	.00539	1.84	0.067	-.0006863	.0205283
_cons		.011899	.0300007	0.40	0.692	-.0471411	.0709391

```

.
.
.
. *FE
. xtreg fr fr_1 fr_2 fr_3 grow lgpp ele_nal ele_al,fe

```

Fixed-effects (within) regression	Number of obs	=	308
Group variable (i): id	Number of groups	=	22

R-sq: within	=	0.4671	Obs per group: min	=	14
between	=	0.8886	avg	=	14.0
overall	=	0.6649	max	=	14

	F(7,279)	=	34.93
corr(u_i, Xb) = 0.6902	Prob > F	=	0.0000

fr		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
fr_1		.5892793	.043795	13.46	0.000	.5030687 .6754899
fr_2		-.213446	.0514539	-4.15	0.000	-.3147332 -.1121589
fr_3		-.1464361	.0392417	-3.73	0.000	-.2236835 -.0691887
grow		-.0468293	.0173763	-2.70	0.007	-.0810345 -.0126241
lgpp		-.0148224	.0094936	-1.56	0.120	-.0335106 .0038657
ele_nal		-.0065903	.0050604	-1.30	0.194	-.0165518 .0033711
ele_al		.0110163	.003586	3.07	0.002	.0039571 .0180754
_cons		.26652	.0838845	3.18	0.002	.1013932 .4316469
sigma_u		.07481693				
sigma_e		.02122926				
rho		.92548566	(fraction of variance due to u_i)			

F test that all u_i=0:	F(21, 279) =	15.94	Prob > F =	0.0000
------------------------	--------------	-------	------------	--------

```

.
. xtreg fr fr_1 fr_2 fr_3 grow lgpp pbc_nal pbc_al,fe

```

Fixed-effects (within) regression	Number of obs	=	308
Group variable (i): id	Number of groups	=	22

R-sq: within	=	0.4463	Obs per group: min	=	14
between	=	0.8889	avg	=	14.0
overall	=	0.6600	max	=	14

	F(7,279)	=	32.12
corr(u_i, Xb) = 0.6910	Prob > F	=	0.0000

fr		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
fr_1		.5818032	.0449729	12.94	0.000	.4932739 .6703324
fr_2		-.2199583	.0524566	-4.19	0.000	-.3232193 -.1166973
fr_3		-.1392907	.0400564	-3.48	0.001	-.2181419 -.0604395
grow		-.0467731	.0178366	-2.62	0.009	-.0818846 -.0116617
lgpp		-.015308	.0096994	-1.58	0.116	-.0344012 .0037852
pbc_nal		-.001471	.0029299	-0.50	0.616	-.0072385 .0042964
pbc_al		.0015274	.0022599	0.68	0.500	-.0029212 .0059759
_cons		.2730504	.0856513	3.19	0.002	.1044455 .4416554
sigma_u		.07530624				
sigma_e		.02163946				
rho		.92372636	(fraction of variance due to u_i)			

F test that all u_i=0: F(21, 279) = 15.72 Prob > F = 0.0000

```
.
. xtreg tr tr_1 tr_2 tr_3 grow lggp ele_nal ele_al ele1_nal ele1_al,fe
```

Fixed-effects (within) regression Number of obs = 308
Group variable (i): id Number of groups = 22

R-sq: within = 0.5132 Obs per group: min = 14
 between = 0.9240 avg = 14.0
 overall = 0.5583 max = 14

corr(u_i, Xb) = 0.5952 F(9,277) = 32.45
 Prob > F = 0.0000

tr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
tr_1	.5905748	.0411159	14.36	0.000	.5096354	.6715142
tr_2	-.2586676	.0487209	-5.31	0.000	-.3545778	-.1627575
tr_3	-.1376176	.0378614	-3.63	0.000	-.2121502	-.063085
grow	-.0641876	.0185117	-3.47	0.001	-.1006291	-.0277461
lggp	-.0067447	.0099218	-0.68	0.497	-.0262765	.0127871
ele_nal	-.0058484	.0054614	-1.07	0.285	-.0165995	.0049027
ele_al	.0133248	.0039312	3.39	0.001	.005586	.0210636
ele1_nal	-.0001588	.0043569	-0.04	0.971	-.0087357	.008418
ele1_al	.0095074	.0037683	2.52	0.012	.0020894	.0169255
_cons	.2246446	.0875908	2.56	0.011	.0522164	.3970729
sigma_u	.08014266					
sigma_e	.02230825					
rho	.92808936	(fraction of variance due to u_i)				

F test that all u_i=0: F(21, 277) = 17.83 Prob > F = 0.0000

```
.
. *GMM
. xtabond fr l(0).ele_nal ele_al, lags(1) pre(grow, lag(2,.)) pre(lggp, lag(2,.))
```

Arellano-Bond dynamic panel-data estimation Number of obs = 286
Group variable (i): id Number of groups = 22

Wald chi2(9) = 185.37

Time variable (t): y Obs per group: min = 13
 avg = 13
 max = 13

One-step results

fr	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
fr						
LD	.3738862	.0329054	11.36	0.000	.3093929	.4383795
grow						
D1	.0229733	.0429132	0.54	0.592	-.061135	.1070815
LD	.0590242	.0435073	1.36	0.175	-.0262485	.144297
L2D	.0055715	.0174559	0.32	0.750	-.0286414	.0397844
lggp						
D1	-.0627462	.0421046	-1.49	0.136	-.1452697	.0197773
LD	-.0372138	.0443214	-0.84	0.401	-.1240821	.0496545
L2D	.0462688	.0394034	1.17	0.240	-.0309605	.123498
ele_nal						
D1	-.0064311	.0047781	-1.35	0.178	-.015796	.0029338
ele_al						
D1	.0082487	.0033091	2.49	0.013	.001763	.0147344

```

_cons      |      .002559   .0004199    6.09   0.000    .001736   .0033819
-----+-----
Sargan test of over-identifying restrictions:
      chi2(357) =   346.63    Prob > chi2 = 0.6431

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
      H0: no autocorrelation    z =  -5.19   Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
      H0: no autocorrelation    z =  -0.36   Pr > z = 0.7156

.
.
. xtabond fr l(0).pbc_nal pbc_al, lags(1) pre(grow, lag(2,.)) pre(lggp,
lag(2,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =       286
Group variable (i): id                          Number of groups   =       22

                                                Wald chi2(9)       =      179.93

Time variable (t): y                            Obs per group: min =       13
                                                avg =              13
                                                max =              13

One-step results
-----+-----
fr      |      Coef.   Std. Err.      z    P>|z|    [95% Conf. Interval]
-----+-----
fr      |
LD      |      .3576535   .0327499    10.92   0.000    .2934649   .4218421
grow    |
D1      |      .0281981   .0423676     0.67   0.506   -.0548409   .1112371
LD      |      .0574934   .0432459     1.33   0.184   -.0272671   .1422538
L2D     |      .0023443   .0173902     0.13   0.893   -.0317399   .0364285
lggp    |
D1      |     -.0715109   .0414332    -1.73   0.084   -.1527186   .0096967
LD      |     -.0241315   .0446573    -0.54   0.589   -.1116582   .0633951
L2D     |      .0403942   .0393127     1.03   0.304   -.0366572   .1174456
pbc_nal |
D1      |     -.0051577   .0027407    -1.88   0.060   -.0105293   .0002139
pbc_al  |
D1      |     -.0004197   .0020449    -0.21   0.837   -.0044275   .0035881
_cons   |      .0026318   .0004205     6.26   0.000    .0018076   .003456
-----+-----
Sargan test of over-identifying restrictions:
      chi2(357) =   353.74    Prob > chi2 = 0.5387

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:
      H0: no autocorrelation    z =  -5.04   Pr > z = 0.0000
Arellano-Bond test that average autocovariance in residuals of order 2 is 0:
      H0: no autocorrelation    z =  -0.58   Pr > z = 0.5629

.
. xtabond fr l(0).ele_nal ele_al ele1_nal ele1_al , lags(1) pre(grow,
lag(2,.)) pre(lggp, lag(2,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =       286
Group variable (i): id                          Number of groups   =       22

                                                Wald chi2(11)      =      211.37

Time variable (t): y                            Obs per group: min =       13
                                                avg =              13
                                                max =              13

One-step results
-----+-----
fr      |      Coef.   Std. Err.      z    P>|z|    [95% Conf. Interval]
-----+-----

```


fr							
	LD	.3721047	.0321465	11.58	0.000	.3090987	.4351108
grow							
	D1	.0082227	.0420188	0.20	0.845	-.0741327	.090578
	LD	.0460954	.042689	1.08	0.280	-.0375735	.1297644
	L2D	.0019717	.0170738	0.12	0.908	-.0314923	.0354358
lggp							
	D1	-.0592305	.0410833	-1.44	0.149	-.1397522	.0212912
	LD	-.0157837	.0437323	-0.36	0.718	-.1014974	.0699301
	L2D	.0202422	.039256	0.52	0.606	-.0566981	.0971825
ele_nal							
	D1	-.0033791	.0047862	-0.71	0.480	-.0127598	.0060017
ele_al							
	D1	.0122515	.0033712	3.63	0.000	.0056441	.0188589
ele1_nal							
	D1	.007472	.0039281	1.90	0.057	-.0002269	.015171
ele1_al							
	D1	.0126459	.0032706	3.87	0.000	.0062357	.0190561
_cons		.0025978	.0004122	6.30	0.000	.00179	.0034056

Sargan test of over-identifying restrictions:

chi2(357) = 347.76 Prob > chi2 = 0.6270

Arellano-Bond test that average autocovariance in residuals of order 1 is 0:

H0: no autocorrelation z = -5.30 Pr > z = 0.0000

Arellano-Bond test that average autocovariance in residuals of order 2 is 0:

H0: no autocorrelation z = -0.04 Pr > z = 0.9697

TABLE 20 - Elections and Revenue from Provincial Taxes conditional on alignment of provincial and federal government

```
.
. *OLS
. regress ptr ptr_1 ptr_2 ptr_3 grow lggp ele_nal ele_al, robust
```

Regression with robust standard errors

Number of obs	=	308
F(7, 300)	=	112.99
Prob > F	=	0.0000
R-squared	=	0.8422
Root MSE	=	.00562

ptr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.9428694	.1447413	6.51	0.000	.6580326	1.227706
ptr_2	-.3730106	.157066	-2.37	0.018	-.6821013	-.06392
ptr_3	.2532924	.0795903	3.18	0.002	.0966665	.4099184
grow	-.0033364	.0048706	-0.69	0.494	-.0129213	.0062485
lggp	.002668	.0009754	2.74	0.007	.0007486	.0045874
ele_nal	.0002664	.0009324	0.29	0.775	-.0015684	.0021012
ele_al	.0001219	.0008605	0.14	0.887	-.0015715	.0018152
_cons	-.0178357	.0074739	-2.39	0.018	-.0325436	-.0031278

```
. regress ptr ptr_1 ptr_2 ptr_3 grow lggp pbc_nal pbc_al, robust
```

Regression with robust standard errors

Number of obs	=	308
F(7, 300)	=	113.08
Prob > F	=	0.0000
R-squared	=	0.8425
Root MSE	=	.00561

ptr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ptr_1	.9464471	.1443851	6.56	0.000	.6623113	1.230583
ptr_2	-.3815714	.1550447	-2.46	0.014	-.6866842	-.0764585
ptr_3	.258749	.0803934	3.22	0.001	.1005425	.4169555
grow	-.0039578	.0048465	-0.82	0.415	-.0134952	.0055795
lggp	.0026596	.0009787	2.72	0.007	.0007337	.0045855
pbc_nal	-1.40e-06	.0010759	-0.00	0.999	-.0021186	.0021158
pbc_al	-.0004472	.0004581	-0.98	0.330	-.0013488	.0004544
_cons	-.0177302	.0074854	-2.37	0.018	-.0324607	-.0029997

```
.
.
.
. *FE
. xtreg ptr ptr_1 ptr_2 ptr_3 grow lggp ele_nal ele_al, fe
```

Fixed-effects (within) regression

Number of obs	=	308
Group variable (i): id		
Number of groups	=	22

R-sq: within = 0.5246 Obs per group: min = 14
 between = 0.9070 avg = 14.0
 overall = 0.7956 max = 14

F(7,279)	=	43.99
Prob > F	=	0.0000

corr(u_i, Xb) = 0.5065

ptr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----	-------	-----------	---	------	----------------------	--

```

      ptr_1 | .7615601 .0481035 15.83 0.000 .6668682 .8562519
      ptr_2 | -.3799494 .0563961 -6.74 0.000 -.4909653 -.2689334
      ptr_3 | .1011669 .0485239 2.08 0.038 .0056476 .1966863
      grow | -.006 .0044526 -1.35 0.179 -.0147649 .002765
      lgpp | .0077006 .0023316 3.30 0.001 .0031108 .0122904
     ele_nal | .0001572 .0012493 0.13 0.900 -.002302 .0026164
     ele_al | .0001693 .0008805 0.19 0.848 -.001564 .0019026
      _cons | -.0515562 .0197718 -2.61 0.010 -.090477 -.0126353
-----+-----
      sigma_u | .0046232
      sigma_e | .00521579
      rho | .43998939 (fraction of variance due to u_i)
-----+-----
F test that all u_i=0:      F(21, 279) =      3.29      Prob > F = 0.0000

. xtreg ptr ptr_1 ptr_2 ptr_3 grow lgpp pbc_nal pbc_al,fe

Fixed-effects (within) regression      Number of obs      =      308
Group variable (i): id      Number of groups      =      22

R-sq:  within = 0.5253      Obs per group: min =      14
      between = 0.9071      avg =      14.0
      overall = 0.7959      max =      14

F(7,279) =      44.11
corr(u_i, Xb) = 0.5064      Prob > F =      0.0000

-----+-----
      ptr |      Coef.      Std. Err.      t      P>|t|      [95% Conf. Interval]
-----+-----
      ptr_1 | .7649158 .0480094 15.93 0.000 .6704092 .8594224
      ptr_2 | -.3871655 .0561041 -6.90 0.000 -.4976065 -.2767245
      ptr_3 | .1051567 .0484995 2.17 0.031 .0096853 .2006281
      grow | -.0065774 .0044847 -1.47 0.144 -.0154056 .0022508
      lgpp | .0076987 .0023361 3.30 0.001 .0031001 .0122973
     pbc_nal | .0000305 .0007047 0.04 0.965 -.0013567 .0014178
     pbc_al | -.0003677 .0005432 -0.68 0.499 -.0014369 .0007016
      _cons | -.0514966 .0198169 -2.60 0.010 -.0905062 -.012487
-----+-----
      sigma_u | .00462108
      sigma_e | .00521195
      rho | .44012564 (fraction of variance due to u_i)
-----+-----
F test that all u_i=0:      F(21, 279) =      3.28      Prob > F = 0.0000

.
. *GMM
.
. xtabond ptr l(0).ele_nal ele_al , lags(2) pre(grow, lag(1,.)) pre(lgpp,
lag(1,.))

Arellano-Bond dynamic panel-data estimation      Number of obs      =      308
Group variable (i): id      Number of groups      =      22

Wald chi2(8) =      235.98

Time variable (t): y      Obs per group: min =      14
      avg =      14
      max =      14

One-step results
-----+-----
ptr |      Coef.      Std. Err.      z      P>|z|      [95% Conf. Interval]
-----+-----
ptr |
      LD | .6402771 .0477628 13.41 0.000 .5466637 .7338905
      L2D | -.3542546 .0488073 -7.26 0.000 -.4499151 -.2585941
grow |

```

