

ECON-S410: Seminars on econometrics

Capital Cities, Conflict and Misgovernance



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1 Introduction

Filipe Campante, Quoc-Anh Do and Bernardo Guimaraes have conducted a research about the relation between capital cities, conflicts and about misgovernance. Their research was about verifying if proximity to a capital city has an impact on the probability of having a conflict which, in their assumptions, could explain why some capital cities are more remote and also why we can see capital moves. Our goal was to provide alternative regressions and additional research in order to try to bring more accurate calculation or more complete research and conclusions.

2 Overall commentaries on the code (replication)

As a whole, we were able to replicate with little to no problems the codes of the authors and replicate their results. There were some minor differences in terms of standard errors for a few results but these were very few and rather limited in scope and could be related to a typing error.

The main point of contention was with the running of the `psacalc` command. `Psacalc`, as explained by E. Oster in her paper ¹, is a test to compute the treatment effect for observables and unobservables for linear regressions. We first have to set a "delta" which represent a level of selection of the unobservables. For example, if the delta is set to 0,5 the selection of the unobservables is assumed to be half the one of the observables. In our case, the authors set a delta is equal to one, meaning that the effects of both observable and unobservable is supposed to have the same effect on the coefficients. Their results were below 1 in absolute value, and often around 0, meaning that the potential omitted variable bias is not a major threat, and thus the regression can be seen as non-spurious.

The problem with the `psacalc` test was that in `stata16`, the code was changed thereby rendering the `stata14` syntax invalid. The changes to be applied are the following:

Original code: `psacalc avglogdist90adj set, delta(1)`

which has to be changed to the new code supported by `stata16`: `psacalc beta avglogdist90adj, delta(1)`

3 Our contribution

Our goal is to read all the paper's analysis and verify if we could upgrade some of the methods and expand the analysis. It could be changing the regression, adding variables, anything that the econometric field would allow us to do and to improve the analysis.

3.1 Impact of the regime change

One of the regressions, the regression that checks the impact of the distance to the capital on the conflict made on a panel data could have some upgrades. In fact, it is true that, as the paper has verified, the

¹Oster, Emily (2016). "Unobservable Selection and Coefficient Stability: Theory and Validation", Journal of Business Economics and Statistics

distance to the capital has an impact, but the last regime transition wasn't controlled. A grid cell that was recently in a conflict could be less attracted by a new conflict, it could be because of the conflict impacts or in order to see how the new regime would perform. It was thus interesting to conduct a Regression Discontinuity Design verifying if, when the last regime transition was a long time ago, the probability of waking a conflict would rise. Therefore, we decided to perform a regression discontinuity design.

3.1.1 Assumptions

Our first assumption is that, if we are before a conflict, we need to be less than 5 years before it in order to have a true impact. In the same way, if we are after a conflict, we need to be less than 5 years after it.

This is because we assume that it takes less than 5 years to set up a conflict and less than 5 years to verify if a new regime brings results.

As ConfIntra is a binary variable, we could work differently than the paper. We kept our regression as a panel data, but we did not perform a linear regression, we decided to go for a logit, which should fits better than an OLS.

A regression discontinuity design is normally performed on a dataset that include a test, we then divide the group between two groups, the pre-test and the post-test group. The problem in this situation is that we do not have one single test, in fact, every single regime transition may be interpreted as a test. Thus, we needed other assumptions.

First, for every year, we have generated two numerical variables, one to determine how many years passed since the last regime transition, and one other to determine how many years were to pass before the next regime transition.

We created another variable, this one a binary, "pre-test", which was true when we were closer to the next regime change than the last one, or false if we were closer to the last transition. Following our first assumption, we decided to create a value for "pre-test" if, and only if, the next or last transition was distanced by less than 6 years.

3.1.2 Results

Figure 1: Regression discontinuity: coefficients

Estimates	Pre-test group	Post-test group
logdistcap	-0.3036	-0.8638
logbdist2	-0.2199	-0.0353
degtemper	-0.0037	-0.0055
prec	-0.0005	-0.0024

The two first annexes show us the complete information about the regressions on the pre-test and post-test samples. We can see that being after a regime transition decreases the coefficient for proximity to capital cities, in fact, the coefficient (-0.86) is smaller than the coefficient of the pre-test sample (-0.3) for

the logcapdist, which was the coefficient we were looking at. The other variables are control variables, the authors used the same control variable while looking to a potential effect of capital change on the distance to capital cities coefficient.

We can also see that both coefficient have very small value ($< 2e^{16}$) while the degree of significance is of 0.001, which implies that our both regressions bring significant results since the null hypothesis is rejected. Since both samples are pretty similar in the number of observations and knowing the residual deviances (25'000 in the pre-test sample and 24955 in the post-test sample) are pretty similar as well, we can also conclude that both regressions are equally good as it comes to fitting the values.

3.1.3 Interpretation

It would appear that being after a regime transition decreases the coefficients of distances to capital cities decreases. In other words, the effect of proximity on the probability of conflict matters less after a regime change, which would tend to confirm our hypothesis that after a regime change, there is a reduced probability of being in a conflict.

This can be explained by the fact that, once there is a regime transition, it has to be verified, the governance has to prove itself and it generally calms the population. A population would have the willingness to give a chance to the next government. Our time period is between 1946 and 2005. In this period, we could see a communism establishment in many countries of the Eastern Europe, the regime Chinese communistic regime changes Mao Zedong or even the Communists Win in Cuba in 1959. Therefore, we think that the new regime characteristics do not matter as it comes to the regime transition effect. Whatever the new government, we would have a trend to it be in order to prove itself. This can have a true impact and can be explanatory as it comes to why do regime transitions come so easily in certain situations.

3.2 Dummies resampling

3.2.1 Defining new dummy variables

Analysing the results of table 8 in the article, it was decided to push that regression further by adding several dummies to the mix. The regression already accounted for the legal origin of a country, i.e., the type of legal system it follows with the categories being: British, French, Scandinavian and Socialist. The authors also included the polity score index to assess the political constraints.

The dummies added are the following: a dummy for the continent Africa ("Africa") and one for Asia ("Asia"), another dummy for (ex-)communist countries ("prev_com") and a final one about extractive colonies ("extractivecol"). The continental dummies are included as these two continents were the two, last major areas of colonies in the world during the 20th century. Therefore, we wish to better monitor their success at governing themselves well despite such a background. The communist dummy aims at discriminating countries that either belonged to the ex-USSR or that are still in a communist directed economy. It is a well-known fact that the ex-members of the USSR suffered from a rather harsh economic setback after the collapse of their former country. As for the others, their political regime being at odds with the western countries' ways of doing economics could lead to different approach being implemented perhaps

explaining part of their differences in governance with the rest of the world.

Finally, the extractive colonies dummy bases itself on the works of Acemoglu et al. (2003) which describes how countries with a high autochthone population density, abundant resources and high mortality rate for European settlers were turned into extractive colonies. The purpose of these was simply to exploit the resources as much as possible and ship them back to the motherland. The absence of European settlers and the extractive, authoritarian nature of the colonial government meant that strong institutions and property rights were not fostered in these countries leading to possible difficulties in establishing sound governance.

3.2.2 Selecting sample variables on countries governance regression

The first results of the regression of the table 8 in the article with the added dummies are the following.

First, we analyse the main sample, which doesn't imply any assumptions on the polity level. With the "prev_com" dummy, the variable that measures the distance to the capital ("zavlogdist90_adj") is significant and reaches -0.2346. With the African dummy, the distance to capital variable is once again not significant. For the Asian dummy, "zavlogdist90_adj" remains significant and is marginally stronger than in the base model (from -0.16 to -0.18). As for the extractive colonies dummy, the four main variables do not change in significance compared to the base model. "zavlogdist90_adj" coefficient is stronger at -0.21. As can be seen here, only the "Asia" and "extractivecol" dummies get a statistically significant. Interestingly, the "extrativecol" dummy is significant while comprising, amongst others, African countries whereas the "Africa" dummy was not significant. This as well as the higher coefficient than in the base model gives some credence of Acemoglu's claims in this study.

The third column takes only the units where the polity score is ≥ 0 (i.e., autocracies) into account. Here the authors find a significant result for the distance to capital with a coefficient of -0.2670. The "prev_com" dummy is not significant the Africa one. The ones for "Asia" and "extractivecol" are significant and give, for the former, a coefficient of -0.3325 and, for the latter, -0.3764. As such, the distance from the capital has a higher impact on the quality of governance in the countries in these samples as compared with the main one.

As for the fifth column which focuses on established democracies, we cannot analyse the results as too few of the units that remain when we select the sample for any of the 4 dummies are considered as established democracies. This result is in line with the definition of the dummy variables as these are meant to highlight groups of countries that might be more likely to present cases of misgovernance.

3.2.3 Selected sample variables used in the conflict and conflict onset probability table

Taking the dummies previously described, it was decided to apply them as additional dummies in the table 3 of the article which examines the influence of the (average) distance to capital and the (average) distance to the largest non-capital city to determine the likelihood of conflict and of conflict onset.

3.2.3.1 Probability of conflict

When analysing the probability of conflict without posing any assumption on the polity score, none of the extra dummies gives a statistically significant result for the variables for the, average, distance to capital (avg_logcapdist) and, average, distance to the largest non-capital city (avg_logdist_LNC) which mirrors the results of the authors. When we run the regression for the anocracies, the authors get a significant result for the first variable avg_logcapdist (coefficient: -0.0266061). The continental and communist dummies are unable to reproduce such statistically significant result. However, the dummy “extractivecol” gives such a significant result and the coefficient is equal to -0.0387946, with regards to the authors findings, it is an important result. That is an incidence 50% higher than in the authors’ results which seems to give further credence to Acemoglu’s claim of the extractive colonies having had negative impact on the countries future ability to govern themselves, and thus these countries are more sensitive to civils conflicts. As for the cases where the democracies, the results for each subsample do not give statistically significant results. This is not overly surprising as the number observations per subsample.

3.2.3.2 Probability of conflict onset

For this part of the table, only the dummies “Asia” and “extractivecol” bring significant results. In the case of the full sample polity-wise (so no restriction in terms of democracy/autocracy level), “Asia” has a significant value (at 90%) for the distance to the non-capital city with a coefficient of 0.0001991 (stronger than the authors value: 8.81e-05). For “extractivecol”, both variables of interest are significant, albeit one (avg_logcapdist) is so at 90%. The coefficient for this one is -0.0002067 as opposed to -0.000112 in the authors’ sample. As for avg_logdist_LNC , the coefficient reaches (0.0001446) which is quite superior to the base result. When dealing with anocracies, only extractive colonies gives any statistically significant results for the variables average log-distance to capital and average log-distance to largest non-capital city, and only so at the 90% level. The coefficients are, respectively, -0.0003176 (compared to -0.000163) and 0.0001574 (instead of 8.52 e-05). As for the scenario when the polity score is set to be superior to zero (democracies), only the average distance to the largest non-capital for the dummy “Africa” is significant albeit its value 0.0000308 is smaller than the authors’ findings (7.86 e-05)

3.3 Natural resources conflict

The focus of the paper, in part allocated for conflicts, was to test the existence of a possible link between conflicts and the political rulers of the country. Obviously, sources of conflicts are broader than disagreements with political regime. A relevant and current example could be Congo (DRC) where conflicts appeared in 2019. These conflicts are concentrated in North-Kivu and they have been fostered by a minority with the aim to control resources². More generally, concerning intra-territorial conflict, the United Nations Peacekeeping relates that over the last 60 years, 40 percent of them were about natural resources³. These events connect perfectly to the popular idiom: “Money is the crux of the war”. With this in mind, a relevant extension could be the proximity to natural resources. The economic centre may also be interesting to

²<https://news.un.org/fr/story/2020/06/1070292>

³https://www.un.org/en/land-natural-resources-conflict/pdfs/GN_Extractive.pdf

analyse because it may capture something that we did not think before which is linked to the economic activity.

Hopefully, Prio-grid dataset provides us partially with such information: we could approximate the economic centre by richest cell in the country; for natural resources we also have dummies for the existence of gems, diamonds, gold and petroleum in a given cell.

These data were unavailable in the authors Stata's datasets available on AER website. Therefore, we had to collect them on the Prio-Grid website, adapt them to have the same format as our data's and merge them. As the only variables available was dummies, the next step was to create distance variables to stay in line with the idea of proximity of a cell and the probability of conflicts. To do so, we had to leave Stata for a moment and code a script in Python to generate two variables with a loop: the first one was the distance to the closest grid with available ores (all resources mentioned above) and the distance to the closest grid with available petroleum, both variables are computed "within a country". We decided to create another additional variable with petroleum due to its importance for our current world.

We repeated the same process for the richest cell in terms of GDP (in purchasing power parity), where the GDP chosen is the average of the 4 available years (respectively 1990, 1995, 2000 and 2005) in order to avoid one-year fluctuations. We used a linear regression as the authors did. More specifically the function "areg" in Stata which allows us to absorb large numbers of dummies. We also clustered in order to obtain a cluster robust standard errors as each cell is part of a country and may be affected differently, such as different policies or culture. We also decided to reuse the dummies created previously, because by definition, previous extractive colonies are resource rich countries and often less stable countries.

3.3.1 Natural resources

For natural resources (all of the previously mentioned ones, cf. above) in "Confddata", the distance (logdistance) to these points is only significant at 95% for "previous extractive colonies" with a coefficient of -0,02334; which seems to be in line with Acemoglu (2003) and the UN observations. The anocracies sample has a much lower coefficient of -0,0123, and is significant at 90%. The other samples do not produce significant results.

In contrast to ores, the distance to petrol does not appear to matter for conflicts. All the coefficients are small or with p-value too important to be interpretable.

In Onset data, nothing is significant at all, both for ores and petrol. The only coefficient which is important is the one of log-distance to petrol in the extractive colonies (-0,02312) - which is relatively high with regards to what the authors found for the distance to the capital cities- with a p-value of 0,136, too high to be acceptable. However, this result looks counterintuitive for this sample as we find the log-distance significant only for the ores in "Civconf" data, and the opposite for "Onset data".

Despite the unclear coefficients for the extractive colonies, conflicts could be sensitive to distance. We went further and checked the different types of conflicts. In the "Civconf" data, the governmental and

non-intense conflict ⁴ provide significant results. In the “onset” data, nothing provides interpretable results. These results do not allow us to draw robust conclusions about a potential impact of the proximity to natural resources on conflicts.

These weak and insignificant coefficients may arise from different definitions between the Uppsala conflict data and the UN. In fact, while reading both definitions, the one of the UN is broader: they account for both violent and non-violent conflicts; while in the Uppsala data definition, they only take violent conflicts with at least 25 battle-related deaths per year into account. It may also stem from what we use for “natural resources” as the UN data set includes timber, fertile lands, and water as well. Unfortunately, these variables are not available in the Prio-grid data. Another explanation could be that the presence of resources is only accounted by a dummy, thus it does not allow us to approximate conflict incentives due to a large quantity in cell; moreover, it does not tell us if a cell has been drained from its resources or does not provide an average number of years of exploitations. In addition, Civconf has a time span of 20 years (1989 – 2008), while Onset data has 60 years (1946 – 2005).

Again, we followed the same techniques than the authors. In this part, our variable of interest is the log-distance to the most productive cell within a country. As we only use GDP from at oldest 1990, using “onset data” which start from 1946-2005 does not make sense, while “Civconf”, which consider the period 1986-2008, does.

3.3.2 Gross cell’s Products

We first used the same control variables as in the paper. The results are significant for the anocracies and extractive colonies and were even higher than those in the paper. But here was one mistake, we did not control for the distance to the capital which could, according to the authors findings, mitigate the results. Indeed, the results changed drastically: the coefficient for the democracies changes to -0,03476 and became significant at 90%; the extractive colonies coefficient even reaches -0,04869 and it is significant at 99%; and -0,04646 for the Asian with a significance at 90%. To give an order of magnitude, the authors found a coefficient the log-distance to capital cities of -0,024 for the anocracies.

While running the different types of conflicts for extractive colonies and Asia, conflicts are mainly non-intense, and government related. For democracies, coefficients are only negative and significant for non-intense conflicts.

These results are interesting and may capture more precisely existing tensions for the extractive colonies for which we were not able to conclude for natural resources. On the other hand, democracies have more significant coefficients than anocracies, at first, it seems astonishing many countries who experienced civil wars have been taken has democratic countries, such as Chile or Colombia, and thus may explain why we observe such coefficient.

⁴According to Uppsala Conflict Data Program: Non-intense refers to conflict with at most 999 deaths battle-related in a year. While Government conflict refers to conflict to two political institution of a given country which fight against each other (one of the two side needs to be the government).

3.4 alternative econometric models used

As stated in our introductory presentation, the authors make use of linear regression for average conflicts between 1989 and 2008 in a gridded data with log-transformed dependent variables. This has a major advantage: it is straightforward to interpret. The potential issue of this is the possibility to predict values above 1 or below 0.

To remedy the latter, we could replace the linear function by a non-linear one. The first two which came into mind are Probit and Logit regression. But after few unsuccessful trials we realized that we were trying to use binary dependent variable function on fraction dependent variables.

This led us to fractional dependent outcome models. After a research, we found Betaregression and fractional Logit or Probit. Betaregression is more appropriate for values strictly between 0 and 1, due to the shape of its density function ⁵, while the other one works also with the boundaries. With this information in mind, we had to check if the dependant variable (`avg.ConfIntra`) has values equal to 0 or 1. The dependent variable is in a large majority equal to 0 among the observations, and some of them are equal to 1. The limitation here, is that we cannot control for fixed effects (by absorbing large set of dummies) and cluster as the authors did. Despite the impossibility to control for fixed effect and cluster, we tried to version of the fractional Logit and Probit. The first version was the one available in Stata, but we could not have a cluster robust standard error; the second was a version proposed by Richard Williams (University of Notre Dame) which allow us to do a cluster robust standard error ⁶.

Another nonlinear econometric function to use could have been a Poisson. This presents a serious advantage in comparison to an OLS: robustness to heteroskedasticity, and thus an better accuracy while estimating coefficients thanks to the assumption of the variance is equal to the mean. The alternative model could be a negative binomial function, it seems relatively better due to the distribution of the dependent variable. In this case, the variance is larger, and it is represented by the mean to which a function of the squared mean and a dispersion parameter is added (Ford, 2016) ⁷. The higher is the dispersion parameter, the closer it converges to a Poisson distribution. But again, there is any function who could absorb large set of dummies as "areg" is able to do. This could be resolve by a Poisson maximum likelihood for high dimensional fixed effect (PPMLHDFE). According to Correia S. and al. the PpmlHDFE presents a major advantage relevant for our analysis : it works with many zeros. Despite that we could not go further as we had some missing values to compute margins at means discontinuous regions due to missing values. Nevertheless, the results obtained in the regressions keep the same sign as the authors' regressions.

To sum up, the linear is maybe inaccurate, but is able to give us a relatively good general idea in our case and, as stated above, it is easy to interpret. Plus we are in a way stick to this regression as we did not find an alternative model which does not produce error messages due to missing values, or constant dependent variable while clustering which did not occur with "areg" (only zeros in a cluster).

⁵<https://www.stata.com/manuals14/rbetareg.pdf>

⁶<https://www3.nd.edu/~rwilliam/stats3/FractionalResponseModels.pdf>

⁷<https://data.library.virginia.edu/getting-started-with-negative-binomial-regression-modeling/>

4 Conclusion

In order to sum up, we have tried, and mostly succeeded, to give an other point of view for the analysis that has been performed.

The results from our contribution seem to be pretty coherent in the research and give a new way to look at the data. Nevertheless, it is hard to compute regression that would perform better than the ones from the article.

From our results, we can add to the paper's remarks that, first of all, a regime change decreases the probability of having an intra-conflict for the incoming years. The continent seem to impact the probability as well. Moreover, the distance to some natural resources can be explanatory, exactly in the same way as the distance to the capital or to the largest non-capital city.

Although the linear regression is inaccurate, especially in a binary outcome equation, we have not found any better method. Even though we have tried logit, probit, poisson or even other alternative regressions, which made us stick to the linear one.

To conclude, we believe that the paper lacks of diversity and many other factors can be explanatory in the authors targets. We also think that we do not have the econometric knowledge to know which regression to perform or which variables to add in order to perform better than the authors. Yet, we think that we have brought some other results which already can be interesting.

5 Annexes

Annex 1: Regression results of the pre-test sample

```
binomial - logit link

ConfIntra ~ logcapdist + logbdist2 + degtemper + prec | year

Estimates:
      Estimate Std. error z value Pr(> |z|)
logcapdist -3.036e-01  1.993e-02 -15.233  <2e-16 ***
logbdist2   2.199e-01  1.373e-02  16.011  <2e-16 ***
degtemper   -3.744e-03  2.501e-03  -1.497    0.134
prec        -5.823e-04  3.957e-05 -14.715  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

residual deviance= 25113.41,
null deviance= 35368.8,
nT= 38369, N= 23

( 6573 observation(s) deleted due to missingness )
( 77366 observation(s) deleted due to perfect classification )

Number of Fisher Scoring Iterations: 8

Average individual fixed effect= -0.794
```

Annex 2: Regression results of the post-test sample

```
binomial - logit link

ConfIntra ~ logcapdist + logbdist2 + degtemper + prec | year

Estimates:
      Estimate Std. error z value Pr(> |z|)
logcapdist -8.638e-01  2.080e-02 -41.526  < 2e-16 ***
logbdist2  -3.530e-02  1.269e-02  -2.782  0.00541 **
degtemper  -5.594e-03  2.394e-03  -2.337  0.01944 *
prec       -2.417e-03  5.256e-05 -45.993  < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

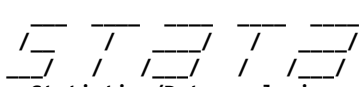
residual deviance= 24955.55,
null deviance= 41773.73,
nT= 58353, N= 27

( 7011 observation(s) deleted due to missingness )
( 66897 observation(s) deleted due to perfect classification )

Number of Fisher Scoring Iterations: 9

Average individual fixed effect= 4.656
```

Annex 3: Regression results on the dummies resampling

 (R)
16.1
Statistics/Data analysis
Special Edition

Copyright 1985-2019 StataCorp LLC
StataCorp
4905 Lakeway Drive
College Station, Texas 77845 USA
800-STATA-PC <https://www.stata.com>
979-696-4600 stata@stata.com
979-696-4601 (fax)

Stata license: Unlimited-user network, expiring 11 Jul 2021
Serial number: 401609323401
Licensed to: Petitjean

Notes:

1. Unicode is supported; see [help unicode advice](#).
2. Maximum number of variables is set to 5,000; see [help set maxvar](#).

```
. do "C:\Users\admin\AppData\Local\Temp\STD49b4_000000.tmp"

. *table 8 twicking misgovernance with the new variables
.
. clear

.
. use "D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\AEJApplied_CrossCountryData2.dta"

.
. gen avglogdist90_adj = 1-pcia_90
(23 missing values generated)

. gen avglogdist90_unadj = 1-pci_90
(23 missing values generated)

. gen largest_avglogdist90_adj = 1-largestgcisc2_90
(10 missing values generated)

. gen other_avglogdist90_adj = 1-othergcisc2_90
(10 missing values generated)

.
. * Generating Autocracy dummy and interaction
. cap drop autocracy

. cap drop tercile*

. pctlte tercile=polity if iso~="ZAF" & iso~="MUS" & iso~="MMR" & iso~="KAZ" & dup<2, nq(3)

. egen tercile1_polity = min(tercile)

. gen autocracy=1 if polity<=tercile1_polity
(129 missing values generated)

. replace autocracy=0 if polity>tercile1_polity & polity!=.
(100 real changes made)

. drop tercile*
```

```

.
. * Generating standardized variables
. egen zavlogdist90_adj=std(avglogdist90_adj)
(23 missing values generated)

. egen zkkm_PolStab_9612=std(kkm_PolStab_9612)
(2 missing values generated)

. gen lavgdays= log(avgdays)
(12 missing values generated)

. egen zlavgdaysletter=std(lavgdays)
(12 missing values generated)

. egen zkkm_pcfirst_9612=std(kkm_pcfirst_9612)
(2 missing values generated)

.

. cap drop zavlogdist90_adjXautocracy

. gen zavlogdist90_adjXautocracy=zavlogdist90_adj*autocracy
(29 missing values generated)

.

. /*several of the variables implemented in the regressions will not be analysed for neither did the authors and in
> /
. * Full sample
. keep if iso~="ZAF" & iso~="MUS" & polity~=. & elf_eth~=. & maj~=. & pres~=. & iso~="MMR" & iso~="KAZ" & dup<2
(44 observations deleted)

. reg zkkm_pcfirst_9612 zavlogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if iso~="ZAF" & iso~="MUS" & p
> ~="KAZ" & dup<2
note: reg_sa omitted because of collinearity
note: leg_socialist omitted because of collinearity

```

Source	SS	df	MS	Number of obs	=	127
Model	98.3594688	15	6.55729792	F(15, 111)	=	34.41
Residual	21.1519956	111	.190558519	Prob > F	=	0.0000
				R-squared	=	0.8230
				Adj R-squared	=	0.7991
Total	119.511464	126	.948503686	Root MSE	=	.43653

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavlogdist90_adj	-.1610203	.0553237	-2.91	0.004	-.2706479	-.0513926
lgdppc	.6295337	.0825262	7.63	0.000	.4660026	.7930649
lpop	-.0624373	.0371558	-1.68	0.096	-.136064	.0111894
SP_URB_TOTL_IN_ZS	.0012624	.0031256	0.40	0.687	-.0049313	.007456
reg_eap	.0496976	.2691681	0.18	0.854	-.4836769	.5830722
reg_eca	-.1872338	.324182	-0.58	0.565	-.8296221	.4551546
reg_mena	-.4329243	.2924664	-1.48	0.142	-1.012466	.1466173
reg_sa	0	(omitted)				
reg_we	.4163342	.3207939	1.30	0.197	-.2193402	1.052009
reg_na	.4282793	.4080429	1.05	0.296	-.380285	1.236844
reg_ssa	.1990277	.2474331	0.80	0.423	-.2912775	.689333
reg_lac	-.3720353	.2770588	-1.34	0.182	-.9210459	.1769753
leg_british	.1575629	.2134953	0.74	0.462	-.2654924	.5806181
leg_french	-.0597966	.2080284	-0.29	0.774	-.4720188	.3524256
leg_socialist	0	(omitted)				
leg_german	.1754474	.1969429	0.89	0.375	-.214808	.5657028
leg_scandinavian	.2877265	.3185723	0.90	0.368	-.3435457	.9189988
_cons	-4.321648	.9384758	-4.60	0.000	-6.1813	-2.461996

```
. psacalc beta zavglogdist90_adj , delta(1)
```

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-0.10897	.00271	
Alt. sol. 1	-6.68764	42.6	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.27304	0.066
Controlled	-0.16102	0.823

Other Inputs	
R_max	1.000
Delta	1.000
Unr. Controls	

```
. local bound = r(output)
```

```
. use workfile, clear
```

```
.
```

```
. *putting an instrumental variable for previously communist countries
```

```
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if iso~="ZAF" & iso~="MUS" & p
```

```
> ~="KAZ" & dup<2 & prev_com==1
```

```
note: reg_eap omitted because of collinearity
```

```
note: reg_mena omitted because of collinearity
```

```
note: reg_sa omitted because of collinearity
```

```
note: reg_we omitted because of collinearity
```

```
note: reg_na omitted because of collinearity
```

```
note: reg_ssa omitted because of collinearity
```

```
note: reg_lac omitted because of collinearity
```

```
note: leg_british omitted because of collinearity
```

```
note: leg_french omitted because of collinearity
```

```
note: leg_socialist omitted because of collinearity
```

```
note: leg_german omitted because of collinearity
```

```
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	24
Model	10.0271739	5	2.00543478	F(5, 18)	=	11.24
Residual	3.21078803	18	.178377113	Prob > F	=	0.0000
				R-squared	=	0.7575
				Adj R-squared	=	0.6901
Total	13.2379619	23	.575563561	Root MSE	=	.42235

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.2346529	.1159017	-2.02	0.058	-.4781534	.0088476
lgdppc	.8714812	.1905161	4.57	0.000	.4712218	1.271741
lpop	-.1194632	.0690472	-1.73	0.101	-.264526	.0255996
SP_URB_TOTL_IN_ZS	-.0009927	.0093749	-0.11	0.917	-.0206887	.0187032
reg_eap	0	(omitted)				
reg_eca	-.4515121	.2698563	-1.67	0.112	-1.018459	.1154351
reg_mena	0	(omitted)				
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	0	(omitted)				
reg_lac	0	(omitted)				
leg_british	0	(omitted)				
leg_french	0	(omitted)				
leg_socialist	0	(omitted)				
leg_german	0	(omitted)				

leg_scandinavian	0 (omitted)					
_cons	-5.013853	1.568782	-3.20	0.005	-8.309742	-1.717964

. psacalc beta zavglogdist90_adj , delta(1)

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-0.29774	.00398	
Alt. sol. 1	3.59825	14.7	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.11882	0.021
Controlled	-0.23465	0.757

Other Inputs	
R_max	1.000
Delta	1.000
Unr. Controls	

. local bound = r(output)

. use workfile, clear

```
.
. *instrumental variable for Asian countries
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if iso~="ZAF" & iso~="MUS" & p
> ~="KAZ" & dup<2 & Asia==1
note: reg_sa omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: reg_ssa omitted because of collinearity
note: reg_lac omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	33
Model	15.9493347	10	1.59493347	F(10, 22)	=	17.58
Residual	1.99574546	22	.090715703	Prob > F	=	0.0000
				R-squared	=	0.8888
				Adj R-squared	=	0.8382
Total	17.9450802	32	.560783756	Root MSE	=	.30119

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.1837178	.0679848	-2.70	0.013	-.3247096	-.042726
lgdppc	.4811139	.1557796	3.09	0.005	.1580468	.804181
lpop	-.0301043	.0470045	-0.64	0.528	-.1275857	.067377
SP_URB_TOTL_IN_ZS	.0078504	.005365	1.46	0.158	-.0032759	.0189767
reg_eap	-.0595591	.2886287	-0.21	0.838	-.6581384	.5390202
reg_eca	-.4644864	.3374746	-1.38	0.183	-1.164366	.235393
reg_mena	-.707498	.363463	-1.95	0.064	-1.461274	.0462781
reg_sa	0 (omitted)					
reg_we	0 (omitted)					
reg_na	0 (omitted)					
reg_ssa	0 (omitted)					
reg_lac	0 (omitted)					
leg_british	.0465899	.2879173	0.16	0.873	-.550514	.6436938
leg_french	-.1671162	.3286013	-0.51	0.616	-.8485936	.5143611
leg_socialist	-.0802124	.3935361	-0.20	0.840	-.8963563	.7359315
leg_german	0 (omitted)					
leg_scandinavian	0 (omitted)					

_cons	-3.815472	1.515834	-2.52	0.020	-6.959119	-.6718251
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. psacalc beta zavglogdist90_adj , delta(1)

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-0.14464	.00153	
Alt. sol. 1	-2.78586	6.77	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.30724	0.190
Controlled	-0.18372	0.889

	Other Inputs	
R_max	1.000	
Delta	1.000	
Unr. Controls		

. local bound = r(output)

. use workfile, clear

```
.
. *instrumental variable for African countries
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if iso~="ZAF" & iso~="MUS" & p
> ~="KAZ" & dup<2 & Africa==1
note: reg_eap omitted because of collinearity
note: reg_eca omitted because of collinearity
note: reg_sa omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: reg_lac omitted because of collinearity
note: leg_socialist omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	36
Model	4.6624869	8	.582810863	F(8, 27)	=	2.26
Residual	6.96341692	27	.257904331	Prob > F	=	0.0541
				R-squared	=	0.4010
				Adj R-squared	=	0.2236
Total	11.6259038	35	.332168681	Root MSE	=	.50784

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.1349056	.1645859	-0.82	0.420	-.4726079	.2027967
lgdppc	.2131935	.1960813	1.09	0.287	-.1891321	.6155191
lpop	-.1432299	.1179299	-1.21	0.235	-.385202	.0987422
SP_URB_TOTL_IN_ZS	-.0068495	.0080393	-0.85	0.402	-.0233448	.0096458
reg_eap	0	(omitted)				
reg_eca	0	(omitted)				
reg_mena	-.4162569	.6189138	-0.67	0.507	-1.686163	.8536494
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	-.8274715	.6932102	-1.19	0.243	-2.249821	.5948784
reg_lac	0	(omitted)				
leg_british	1.485626	.8182972	1.82	0.081	-.193381	3.164633
leg_french	1.132974	.7973457	1.42	0.167	-.5030442	2.768992
leg_socialist	0	(omitted)				
leg_german	0	(omitted)				

leg_scandinavian	0 (omitted)					
_cons	-0.0204797	2.928322	-0.01	0.994	-6.028901	5.987942

. psacalc beta zavglogdist90_adj , delta(1)

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	3.06307	10.2	
Alt. sol. 1	-0.36361	.0523	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.24064	0.125
Controlled	-0.13491	0.401

Other Inputs	
R_max	1.000
Delta	1.000
Unr. Controls	

. local bound = r(output)

. use workfile, clear

```
.
. *instrumental variables for ex- extractive colonies
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if iso~="ZAF" & iso~="MUS" & p
> ~="KAZ" & dup<2 & extractivecol==1
note: reg_eca omitted because of collinearity
note: reg_mena omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: leg_french omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	65
Model	9.79439466	10	.979439466	F(10, 54)	=	4.19
Residual	12.6211851	54	.233725651	Prob > F	=	0.0003
				R-squared	=	0.4369
				Adj R-squared	=	0.3327
Total	22.4155798	64	.350243434	Root MSE	=	.48345

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-0.2106034	.103746	-2.03	0.047	-0.4186016	-0.0026053
lgdppc	.3969797	.1421363	2.79	0.007	.1120136	.6819457
lpop	-0.0103873	.0705965	-0.15	0.884	-0.1519247	.13115
SP_URB_TOTL_IN_ZS	-0.0000435	.0050687	-0.01	0.993	-0.0102056	.0101185
reg_eap	.2563225	.3855553	0.66	0.509	-0.5166694	1.029314
reg_eca	0 (omitted)					
reg_mena	0 (omitted)					
reg_sa	.0948408	.4067681	0.23	0.817	-0.7206801	.9103617
reg_we	0 (omitted)					
reg_na	0 (omitted)					
reg_ssa	.3639661	.3539092	1.03	0.308	-0.3455791	1.073511
reg_lac	.1340536	.2830798	0.47	0.638	-0.4334872	.7015944
leg_british	.3120219	.1709549	1.83	0.074	-0.0307221	.6547659
leg_french	0 (omitted)					
leg_socialist	-0.0255335	.4549999	-0.06	0.955	-0.9377534	.8866864
leg_german	0 (omitted)					
leg_scandinavian	0 (omitted)					

_cons	-3.703399	1.853496	-2.00	0.051	-7.419435	.0126361
-------	-----------	----------	-------	-------	-----------	----------

. psacalc beta zavglogdist90_adj , delta(1)

	—— Treatment Effect Estimate ——		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	3.11352	11	
Alt. sol. 1	-0.38545	.0306	Yes
Alt. sol. 2			

	—— Inputs from Regressions ——	
	Coeff.	R-Squared
Uncontrolled	-0.28165	0.184
Controlled	-0.21060	0.437

	—— Other Inputs ——	
R_max	1.000	
Delta	1.000	
Unr. Controls		

. local bound = r(output)

. use workfile, clear

.
. * Split sample
. * Autocracies
. use workfile, clear

. keep if autocracy==1 & elf_eth~= . & maj~= . & pres~= . & iso~="MMR" & iso~="KAZ" & dup<2
(136 observations deleted)

. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if autocracy==1 & elf_eth~= . & maj~= . & pres~= . & iso~="MMR" & iso~="KAZ" & dup<2
note: reg_sa omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: reg_lac omitted because of collinearity
note: leg_british omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity

Source	SS	df	MS	Number of obs	=	36
Model	12.021023	10	1.2021023	F(10, 25)	=	12.10
Residual	2.48428574	25	.09937143	Prob > F	=	0.0000
				R-squared	=	0.8287
				Adj R-squared	=	0.7602
Total	14.5053088	35	.414437394	Root MSE	=	.31523

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.2670494	.0752617	-3.55	0.002	-.4220538	-.112045
lgdppc	.5193152	.1863474	2.79	0.010	.1355255	.9031049
lpop	-.0440259	.0560537	-0.79	0.440	-.1594706	.0714189
SP_URB_TOTL_IN_ZS	-.0102482	.006031	-1.70	0.102	-.0226692	.0021729
reg_eap	1.358102	.5769535	2.35	0.027	.1698444	2.54636
reg_eca	.8587657	.6616218	1.30	0.206	-.5038699	2.221401
reg_mena	.0026418	.4103919	0.01	0.995	-.8425761	.8478597
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	-.0690739	.3654839	-0.19	0.852	-.8218021	.6836542
reg_lac	0	(omitted)				
leg_british	0	(omitted)				

leg_french	.0144071	.1821363	0.08	0.938	-.3607097	.3895239
leg_socialist	-1.174756	.5048634	-2.33	0.028	-2.214542	-.1349701
leg_german	0	(omitted)				
leg_scandinavian	0	(omitted)				
_cons	-3.510404	1.687342	-2.08	0.048	-6.98555	-.0352587

```
. psacalc beta zavglogdist90_adj , delta(1)
```

	—— Treatment Effect Estimate ——		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-0.18603	.00656	
Alt. sol. 1	-2.01495	3.06	Yes
Alt. sol. 2			

	—— Inputs from Regressions ——	
	Coeff.	R-Squared
Uncontrolled	-0.35713	0.310
Controlled	-0.26705	0.829

	—— Other Inputs ——
R_max	1.000
Delta	1.000
Unr. Controls	

```
. local bound = r(output)
```

```
.
```

```
. *using with prev_com dummy
```

```
. use workfile, clear
```

```
. keep if autocracy==1 & elf_eth~= . & maj~= . & pres~= . & iso~="MMR" & iso~="KAZ" & dup<2  
(136 observations deleted)
```

```
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if autocracy==1 & elf_eth~= . &  
> m==1
```

```
note: reg_eca omitted because of collinearity  
note: reg_mena omitted because of collinearity  
note: reg_sa omitted because of collinearity  
note: reg_we omitted because of collinearity  
note: reg_na omitted because of collinearity  
note: reg_ssa omitted because of collinearity  
note: reg_lac omitted because of collinearity  
note: leg_british omitted because of collinearity  
note: leg_french omitted because of collinearity  
note: leg_socialist omitted because of collinearity  
note: leg_german omitted because of collinearity  
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	10
Model	.691140022	5	.138228004	F(5, 4)	=	2.09
Residual	.264128683	4	.066032171	Prob > F	=	0.2470
				R-squared	=	0.7235
				Adj R-squared	=	0.3779
Total	.955268705	9	.106140967	Root MSE	=	.25697

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.7231122	.636295	-1.14	0.319	-2.48975	1.043526
lgdppc	1.505082	1.63159	0.92	0.408	-3.024937	6.035102
lpop	-.0802214	.1348143	-0.60	0.584	-.4545259	.2940831
SP_URB_TOTL_IN_ZS	-.0313959	.0474165	-0.66	0.544	-.1630452	.1002533
reg_eap	.4213763	.4712157	0.89	0.422	-.8869282	1.729681
reg_eca	0 (omitted)					
reg_mena	0 (omitted)					
reg_sa	0 (omitted)					
reg_we	0 (omitted)					
reg_na	0 (omitted)					
reg_ssa	0 (omitted)					
reg_lac	0 (omitted)					
leg_british	0 (omitted)					
leg_french	0 (omitted)					
leg_socialist	0 (omitted)					
leg_german	0 (omitted)					
leg_scandinavian	0 (omitted)					
_cons	-9.716909	8.475826	-1.15	0.316	-33.24957	13.81576

```
. psacalc beta zavglogdist90_adj , delta(1)
```

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-2.55567	3.36	
Alt. sol. 1	0.16062	.781	Yes
Alt. sol. 2			
	Inputs from Regressions		
	Coeff.	R-Squared	
Uncontrolled	0.10218	0.063	
Controlled	-0.72311	0.724	
	Other Inputs		
R_max	1.000		
Delta	1.000		
Unr. Controls			

```
. local bound = r(output)
```

```
.
```

```
. *using with Asia dummy
```

```
. use workfile, clear
```

```
. keep if autocracy==1 & elf_eth~= . & maj~= . & pres~= . & iso~="MMR" & iso~="KAZ" & dup<2
(136 observations deleted)
```

```
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if autocracy==1 & elf_eth~= . & n
> 1
```

```
note: reg_sa omitted because of collinearity
```

```
note: reg_we omitted because of collinearity
```

```
note: reg_na omitted because of collinearity
```

```
note: reg_ssa omitted because of collinearity
```

```
note: reg_lac omitted because of collinearity
```

```
note: leg_british omitted because of collinearity
```

```
note: leg_german omitted because of collinearity
```

```
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	17
Model	9.69242814	9	1.07693646	F(9, 7)	=	14.86
Residual	.507246298	7	.072463757	Prob > F	=	0.0009
				R-squared	=	0.9503
				Adj R-squared	=	0.8863
Total	10.1996744	16	.637479652	Root MSE	=	.26919

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.3324848	.1156938	-2.87	0.024	-.6060571	-.0589125
lgdppc	.6352885	.3124383	2.03	0.082	-.1035106	1.374088
lpop	-.0363921	.0562848	-0.65	0.539	-.1694846	.0967004
SP_URB_TOTL_IN_ZS	-.0103038	.0150593	-0.68	0.516	-.0459133	.0253057
reg_eap	.9675754	.5768254	1.68	0.137	-.3963998	2.331551
reg_eca	.4625673	.713386	0.65	0.537	-1.224322	2.149457
reg_mena	.2202099	.5430993	0.41	0.697	-1.064016	1.504436
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	0	(omitted)				
reg_lac	0	(omitted)				
leg_british	0	(omitted)				
leg_french	-.4858384	.3798252	-1.28	0.242	-1.383982	.4123054
leg_socialist	-.7871003	.5662808	-1.39	0.207	-2.126142	.5519411
leg_german	0	(omitted)				
leg_scandinavian	0	(omitted)				
_cons	-4.480507	2.262607	-1.98	0.088	-9.830723	.8697095

. psacalc beta zavglogdist90_adj , delta(1)

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-0.33265	2.82e-08	
Alt. sol. 1	557.75119	311457	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.33213	0.273
Controlled	-0.33248	0.950

	Other Inputs	
	R_max	Delta
Unr. Controls	1.000	1.000

. local bound = r(output)

.
. *using with Africa dummy
. use workfile, clear

```
. keep if autocracy==1 & elf_eth~= . & maj~= . & pres~= . & iso~="MMR" & iso~="KAZ" & dup<2
(136 observations deleted)
```

```
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if autocracy==1 & elf_eth~= . & n
> =1
```

```
note: reg_eap omitted because of collinearity
note: reg_eca omitted because of collinearity
note: reg_mena omitted because of collinearity
note: reg_sa omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: reg_lac omitted because of collinearity
note: leg_french omitted because of collinearity
note: leg_socialist omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	18
Model	1.97673603	6	.329456006	F(6, 11)	=	2.45
Residual	1.47871876	11	.134428978	Prob > F	=	0.0938
				R-squared	=	0.5721
				Adj R-squared	=	0.3386
Total	3.45545479	17	.203262047	Root MSE	=	.36665

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.2591155	.2162602	-1.20	0.256	-.7351011	.2168701
lgdppc	.2619812	.3488846	0.75	0.468	-.5059086	1.029871
lpop	-.0465507	.1705577	-0.27	0.790	-.4219456	.3288442
SP_URB_TOTL_IN_ZS	-.0123501	.0085971	-1.44	0.179	-.0312722	.0065721
reg_eap	0	(omitted)				
reg_eca	0	(omitted)				
reg_mena	0	(omitted)				
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	-.3800065	.6765061	-0.56	0.586	-1.868986	1.108973
reg_lac	0	(omitted)				
leg_british	-.1098286	.3009589	-0.36	0.722	-.7722347	.5525776
leg_french	0	(omitted)				
leg_socialist	0	(omitted)				
leg_german	0	(omitted)				
leg_scandinavian	0	(omitted)				
_cons	-1.195108	4.484998	-0.27	0.795	-11.06652	8.676306

```
. psacalc beta zavglogdist90_adj , delta(1)
```

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	2.27057	6.4	
Alt. sol. 1	-0.46248	.0414	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.38505	0.419
Controlled	-0.25912	0.572

	Other Inputs	
	R_max	Delta
Unr. Controls	1.000	1.000

```
. local bound = r(output)
```

```
.
. *using with Extractivecol dummy
. use workfile, clear
```

```
. keep if autocracy==1 & elf_eth~= . & maj~= . & pres~= . & iso~="MMR" & iso~="KAZ" & dup<2
(136 observations deleted)
```

```
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if autocracy==1 & elf_eth~= . & n
> ivecol==1
```

```
note: reg_eca omitted because of collinearity
note: reg_sa omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: reg_lac omitted because of collinearity
note: leg_french omitted because of collinearity
note: leg_socialist omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity
```

Source	SS	df	MS	Number of obs	=	21
Model	2.00975874	8	.251219842	F(8, 12)	=	1.89
Residual	1.5937093	12	.132809108	Prob > F	=	0.1541
				R-squared	=	0.5577
				Adj R-squared	=	0.2629
Total	3.60346804	20	.180173402	Root MSE	=	.36443

zkkm_pcfirst_9612	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.3764323	.1746229	-2.16	0.052	-.7569029	.0040382
lgdppc	.3254471	.3364146	0.97	0.352	-.4075372	1.058432
lpop	.0192121	.1196349	0.16	0.875	-.2414501	.2798742
SP_URB_TOTL_IN_ZS	-.0132741	.0089387	-1.49	0.163	-.0327499	.0062017
reg_eap	.011236	.5486575	0.02	0.984	-1.184186	1.206658
reg_eca	0	(omitted)				
reg_mena	-.0963866	.6059265	-0.16	0.876	-1.416587	1.223814
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	-.1155317	.4645269	-0.25	0.808	-1.127649	.8965856
reg_lac	0	(omitted)				
leg_british	-.1630371	.276546	-0.59	0.566	-.765579	.4395048
leg_french	0	(omitted)				
leg_socialist	0	(omitted)				
leg_german	0	(omitted)				
leg_scandinavian	0	(omitted)				
_cons	-2.887672	3.150862	-0.92	0.377	-9.752811	3.977467

```
. psacalc beta zavglogdist90_adj , delta(1)
```

	Treatment Effect Estimate		
	Estimate	Sq. difference from controlled beta	Bias changes direction
Beta	-6.82338	41.6	
Alt. sol. 1	-0.31967	.00322	Yes
Alt. sol. 2			

	Inputs from Regressions	
	Coeff.	R-Squared
Uncontrolled	-0.35185	0.373
Controlled	-0.37643	0.558

— Other Inputs —	
R_max	1.000
Delta	1.000
Unr. Controls	

. local bound = r(output)

.
 . * Established democracies
 . use workfile, clear

. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if polity>9 & polity~= . & iso~='
 > MR' & iso~='KAZ' & dup<2, robust
 note: reg_sa omitted because of collinearity
 note: reg_na omitted because of collinearity
 note: reg_ssa omitted because of collinearity
 note: leg_socialist omitted because of collinearity

Linear regression	Number of obs	=	31
	F(13, 17)	=	67.80
	Prob > F	=	0.0000
	R-squared	=	0.8985
	Root MSE	=	.28418

zkkm_pcfirst_9612	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	-.0585187	.1211537	-0.48	0.635	-.3141306	.1970932
lgdppc	.8208785	.1426017	5.76	0.000	.5200152	1.121742
lpop	-.1340008	.0659385	-2.03	0.058	-.273119	.0051174
SP_URB_TOTL_IN_ZS	.0061524	.0037799	1.63	0.122	-.0018224	.0141273
reg_eap	-.1024687	.2422952	-0.42	0.678	-.613667	.4087296
reg_eca	-.6199265	.2187418	-2.83	0.011	-1.081431	-.1584217
reg_mena	-.7141833	.396258	-1.80	0.089	-1.550215	.1218479
reg_sa	0	(omitted)				
reg_we	-.2673766	.2260524	-1.18	0.253	-.7443055	.2095524
reg_na	0	(omitted)				
reg_ssa	0	(omitted)				
reg_lac	-.4940607	.3480559	-1.42	0.174	-1.228395	.240273
leg_british	-.4505996	.195867	-2.30	0.034	-.8638429	-.0373564
leg_french	-.4050737	.1976974	-2.05	0.056	-.8221788	.0120314
leg_socialist	0	(omitted)				
leg_german	-.2362809	.1497753	-1.58	0.133	-.552279	.0797173
leg_scandinavian	-.168308	.1623406	-1.04	0.314	-.5108167	.1742008
_cons	-4.344055	1.164089	-3.73	0.002	-6.800068	-1.888042

. * using the prev_com dummy
 . use workfile, clear

. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if polity>9 & polity~= . & iso~='
 > MR' & iso~='KAZ' & dup<2 & prev_com==1, robust
 note: lgdppc omitted because of collinearity
 note: reg_eap omitted because of collinearity
 note: reg_eca omitted because of collinearity
 note: reg_mena omitted because of collinearity
 note: reg_sa omitted because of collinearity
 note: reg_we omitted because of collinearity
 note: reg_na omitted because of collinearity
 note: reg_ssa omitted because of collinearity
 note: reg_lac omitted because of collinearity
 note: leg_british omitted because of collinearity
 note: leg_french omitted because of collinearity
 note: leg_socialist omitted because of collinearity
 note: leg_german omitted because of collinearity
 note: leg_scandinavian omitted because of collinearity

Linear regression

Number of obs = 4
 $F(0, 0)$ = .
 Prob > F = .
 R-squared = 1.0000
 Root MSE = 0

zkkm_pcfirst_9612	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]
zavglogdist90_adj	-.1855705
lgdppc	0 (omitted)				
lpop	.2467879
SP_URB_TOTL_IN_ZS	.0774887
reg_eap	0 (omitted)				
reg_eca	0 (omitted)				
reg_mena	0 (omitted)				
reg_sa	0 (omitted)				
reg_we	0 (omitted)				
reg_na	0 (omitted)				
reg_ssa	0 (omitted)				
reg_lac	0 (omitted)				
leg_british	0 (omitted)				
leg_french	0 (omitted)				
leg_socialist	0 (omitted)				
leg_german	0 (omitted)				
leg_scandinavian	0 (omitted)				
_cons	-8.073139

. *using the asia dummy
 . use workfile, clear

. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if polity>9 & polity~=. & iso~='
 > MR" & iso~="KAZ" & dup<2 & Asia==1, robust
 note: zavglogdist90_adj omitted because of collinearity
 note: lgdppc omitted because of collinearity
 note: reg_eap omitted because of collinearity
 note: reg_eca omitted because of collinearity
 note: reg_mena omitted because of collinearity
 note: reg_sa omitted because of collinearity
 note: reg_we omitted because of collinearity
 note: reg_na omitted because of collinearity
 note: reg_ssa omitted because of collinearity
 note: reg_lac omitted because of collinearity
 note: leg_british omitted because of collinearity
 note: leg_french omitted because of collinearity
 note: leg_socialist omitted because of collinearity
 note: leg_german omitted because of collinearity
 note: leg_scandinavian omitted because of collinearity

Linear regression

Number of obs = 3
 $F(0, 0)$ = .
 Prob > F = .
 R-squared = 1.0000
 Root MSE = 0

zkkm_pcfirst_9612	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]
zavglogdist90_adj	0 (omitted)				
lgdppc	0 (omitted)				
lpop	.3069635
SP_URB_TOTL_IN_ZS	.0137308
reg_eap	0 (omitted)				
reg_eca	0 (omitted)				
reg_mena	0 (omitted)				
reg_sa	0 (omitted)				
reg_we	0 (omitted)				
reg_na	0 (omitted)				
reg_ssa	0 (omitted)				

reg_lac	0	(omitted)					
leg_british	0	(omitted)					
leg_french	0	(omitted)					
leg_socialist	0	(omitted)					
leg_german	0	(omitted)					
leg_scandinavian	0	(omitted)					
_cons	-5.299221	

```
. *using the Africa dummy
. /*use workfile, clear
> reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if polity>9 & polity~= . & iso~= '
> MR' & iso~="KAZ" & dup<2 & Africa==1, robust*/
.
. *using the extractivecol dummy
. use workfile, clear
```

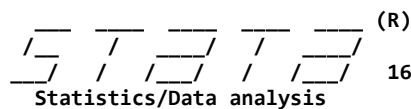
```
. reg zkkm_pcfirst_9612 zavglogdist90_adj lgdppc lpop SP_URB_TOTL_IN_ZS reg_* leg_* if polity>9 & polity~= . & iso~= '
> MR' & iso~="KAZ" & dup<2 & extractivecol==1, robust
note: zavglogdist90_adj omitted because of collinearity
note: lgdppc omitted because of collinearity
note: reg_eap omitted because of collinearity
note: reg_eca omitted because of collinearity
note: reg_mena omitted because of collinearity
note: reg_sa omitted because of collinearity
note: reg_we omitted because of collinearity
note: reg_na omitted because of collinearity
note: reg_ssa omitted because of collinearity
note: reg_lac omitted because of collinearity
note: leg_british omitted because of collinearity
note: leg_french omitted because of collinearity
note: leg_socialist omitted because of collinearity
note: leg_german omitted because of collinearity
note: leg_scandinavian omitted because of collinearity
```

Linear regression	Number of obs	=	3
	$F(0, 0)$	=	.
	Prob > F	=	.
	R-squared	=	1.0000
	Root MSE	=	0

zkkm_pcfirst_9612	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
zavglogdist90_adj	0	(omitted)				
lgdppc	0	(omitted)				
lpop	.1515793
SP_URB_TOTL_IN_ZS	.0046513
reg_eap	0	(omitted)				
reg_eca	0	(omitted)				
reg_mena	0	(omitted)				
reg_sa	0	(omitted)				
reg_we	0	(omitted)				
reg_na	0	(omitted)				
reg_ssa	0	(omitted)				
reg_lac	0	(omitted)				
leg_british	0	(omitted)				
leg_french	0	(omitted)				
leg_socialist	0	(omitted)				
leg_german	0	(omitted)				
leg_scandinavian	0	(omitted)				
_cons	-1.856254

```
.
end of do-file
```

```
.
```



(R)

16.1

Special Edition

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Notes:

1. Unicode is supported; see [help unicode advice](#).
2. Maximum number of variables is set to 5,000; see [help set maxvar](#).

```
. do "C:\Users\admin\AppData\Local\Temp\STDa84_000000.tmp"

. global gdpcontrols "avg_loggcpcpc avg_logpop"

. global controls "avg_loggcpcpc avg_logpop imr logtttime logcellarea avg_logdist_LNC"

. global controls2 "${controls} mountain2000 ycoord avg_degtemper avg_prec" //not including forest2000

.
.
. *part 1 conflict Africa dummy
.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear

.
. *here and many more times afterwards is the code to restrict the dataset to the added dummies to get the results f
> he little (*). if one wishes to use another dummy, clear first then redo the whole procedure
. keep if Africa==1 /*new code*/
(54,337 observations deleted)

.
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the fi
```

Linear regression, absorbing indicators	Number of obs	=	8,982
Absorbed variable: isocode	No. of categories	=	51
	F(11, 50)	=	1.54
	Prob > F	=	0.1477
	R-squared	=	0.8547
	Adj R-squared	=	0.8538
	Root MSE	=	0.0856

(Std. Err. adjusted for 51 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0081142	.0075415	-1.08	0.287	-.0232617	.0070334
avg_loggcpcpc	-.0167877	.0171071	-0.98	0.331	-.0511483	.0175729
avg_logpop	.0027989	.0030764	0.91	0.367	-.0033803	.0089782
imr	.0000452	.0000228	1.98	0.054	-7.32e-07	.000091
logtttime	-.0130223	.0071993	-1.81	0.076	-.0274825	.001438
logcellarea	.0359817	.0296763	1.21	0.231	-.0236249	.0955882
avg_logdist_LNC	.0039189	.0054317	0.72	0.474	-.0069911	.0148288
mountain2000	.0070406	.0129321	0.54	0.589	-.0189342	.0330154
ycoord	-.0055892	.004456	-1.25	0.216	-.0145393	.0033608
avg_degtemper	.0014312	.0023832	0.60	0.551	-.0033555	.006218
avg_prec	.0000501	.0000287	1.75	0.087	-7.52e-06	.0001078
_cons	-.0277527	.1990747	-0.14	0.890	-.4276061	.3721007

```
.
. keep if avg_polity2<=0
(3,216 observations deleted)
```

```
.
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the se
```

Linear regression, absorbing indicators	Number of obs	=	5,972
Absorbed variable: isocode	No. of categories	=	30
	F(11, 29)	=	3.75
	Prob > F	=	0.0021
	R-squared	=	0.8575
	Adj R-squared	=	0.8566
	Root MSE	=	0.0962

(Std. Err. adjusted for 30 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0140265	.013231	-1.06	0.298	-.0410869	.013034
avg_loggcpcpc	-.0278999	.0239242	-1.17	0.253	-.0768304	.0210305
avg_logpop	.0038232	.0057621	0.66	0.512	-.0079617	.0156081
imr	.0000551	.0000328	1.68	0.103	-.0000119	.0001222
logtttime	-.0229125	.0079761	-2.87	0.008	-.0392254	-.0065995
logcellarea	.0700257	.0417452	1.68	0.104	-.0153528	.1554043
avg_logdist_LNC	.0055158	.0080117	0.69	0.497	-.01087	.0219016
mountain2000	.0085867	.017292	0.50	0.623	-.0267795	.0439529
ycoord	-.0072656	.005619	-1.29	0.206	-.0187577	.0042264
avg_degtemper	.0014626	.0034816	0.42	0.678	-.005658	.0085832
avg_prec	.0000847	.0000348	2.43	0.021	.0000135	.0001558
_cons	-.0854597	.3197332	-0.27	0.791	-.7393874	.568468

```
.
. cap drop sample
. gen sample = e(sample)
. global avg_Conf_list = "avg_ConfGov avg_ConfTerr avg_ConfType3 avg_ConfType4 avg_ConfIntense avg_ConfNonIntense av
.
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
. areg avg_ConfIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare
```

Linear regression, absorbing indicators	Number of obs	=	5,972
Absorbed variable: isocode	No. of categories	=	30
	F(10, 29)	=	4.91
	Prob > F	=	0.0004
	R-squared	=	0.8568
	Adj R-squared	=	0.8559
	Root MSE	=	0.0964

(Std. Err. adjusted for 30 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	.0026273	.0097628	0.27	0.790	-.0173399	.0225944
avg_loggcpcpc	-.0264518	.0230994	-1.15	0.262	-.0736954	.0207918
avg_logpop	.0057485	.0058116	0.99	0.331	-.0061377	.0176346
imr	.0000511	.0000314	1.63	0.114	-.000013	.0001153
logtttime	-.0244769	.0081624	-3.00	0.006	-.0411708	-.007783
logcellarea	.0683933	.0420841	1.63	0.115	-.0176783	.1544649
mountain2000	.0088596	.0167884	0.53	0.602	-.0254765	.0431958
ycoord	-.006869	.0052827	-1.30	0.204	-.0176733	.0039353
avg_degtemper	.0015752	.0036208	0.44	0.667	-.0058302	.0089807
avg_prec	.0000877	.0000353	2.48	0.019	.0000154	.0001599
_cons	-.165484	.2991791	-0.55	0.584	-.777374	.4464061

```

. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear

. areg avg_ConfIntra avg_logcapdist ${controls2} if avg_polity2>0 , a(isocode) cluster(iso) /*gives the result to co

```

```

Linear regression, absorbing indicators      Number of obs      =      39,501
Absorbed variable: isocode                 No. of categories =      108
                                           F( 11, 107)        =      1.38
                                           Prob > F           =      0.1938
                                           R-squared          =      0.7933
                                           Adj R-squared      =      0.7927
                                           Root MSE          =      0.0755

```

(Std. Err. adjusted for 108 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	.0094666	.014272	0.66	0.509	-.0188259	.0377592
avg_loggcppc	-.0160397	.0100084	-1.60	0.112	-.0358802	.0038008
avg_logpop	.0013423	.0010094	1.33	0.186	-.0006587	.0033432
imr	.0000621	.0000587	1.06	0.292	-.0000542	.0001784
logtttime	-.0037826	.0041029	-0.92	0.359	-.0119162	.004351
logcellarea	-.0091463	.0061326	-1.49	0.139	-.0213036	.003011
avg_logdist_LNC	-.0049452	.009689	-0.51	0.611	-.0241525	.0142621
mountain2000	.0041888	.0082671	0.51	0.613	-.0121996	.0205773
ycoord	-.0003657	.0004842	-0.76	0.452	-.0013255	.0005941
avg_degtemper	-.0003667	.0005357	-0.68	0.495	-.0014285	.0006952
avg_prec	1.26e-06	8.97e-06	0.14	0.888	-.0000165	.000019
_cons	.237504	.0798987	2.97	0.004	.0791141	.395894

```

.
.
. *part 1 conflict Asia dummy
.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear

.
. keep if Asia==1 /*new code*/
(51,773 observations deleted)

.
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the fi

```

```

Linear regression, absorbing indicators      Number of obs      =      11,762
Absorbed variable: isocode                 No. of categories =      39
                                           F( 11, 38)        =      1.76
                                           Prob > F           =      0.0967
                                           R-squared          =      0.8018
                                           Adj R-squared      =      0.8009
                                           Root MSE          =      0.1291

```

(Std. Err. adjusted for 39 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	.0231485	.0248304	0.93	0.357	-.0271181	.0734151
avg_loggcppc	-.0651971	.0466988	-1.40	0.171	-.1597339	.0293396
avg_logpop	.0094685	.0060889	1.56	0.128	-.0028577	.0217948
imr	.0000484	.0000462	1.05	0.301	-.0000451	.000142
logtttime	-.0171603	.0100593	-1.71	0.096	-.0375243	.0032036
logcellarea	.0265953	.017023	1.56	0.127	-.0078659	.0610565
avg_logdist_LNC	.0047954	.0127832	0.38	0.710	-.0210828	.0306736
mountain2000	-.0193032	.0234192	-0.82	0.415	-.066713	.0281066
ycoord	-.000588	.0014255	-0.41	0.682	-.0034739	.0022978
avg_degtemper	-.0022848	.0014704	-1.55	0.129	-.0052616	.0006919
avg_prec	-7.29e-06	.0000241	-0.30	0.763	-.000056	.0000414
_cons	.306647	.234568	1.31	0.199	-.1682111	.781505

```
.
. keep if avg_polity2<=0
(4,906 observations deleted)
```

```
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the se
```

```
Linear regression, absorbing indicators      Number of obs      =      7,864
Absorbed variable: isocode                 No. of categories =      20
                                           F( 11,      19)    =      1.57
                                           Prob > F           =      0.1876
                                           R-squared          =      0.9024
                                           Adj R-squared      =      0.9020
                                           Root MSE          =      0.0653
```

(Std. Err. adjusted for 20 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0199619	.0138326	-1.44	0.165	-.048914	.0089901
avg_loggcppc	.0082565	.006297	1.31	0.205	-.0049233	.0214364
avg_logpop	.0059543	.004105	1.45	0.163	-.0026375	.0145462
imr	.0000355	.0000318	1.12	0.278	-.000031	.0001019
logtttime	-.0082675	.0064715	-1.28	0.217	-.0218125	.0052774
logcellarea	.0033788	.0072384	0.47	0.646	-.0117713	.018529
avg_logdist_LNC	.0039007	.00691	0.56	0.579	-.0105621	.0183636
mountain2000	.0072323	.0078773	0.92	0.370	-.0092551	.0237197
ycoord	-.0020978	.0024073	-0.87	0.394	-.0071364	.0029408
avg_degtemper	-.0011418	.0007518	-1.52	0.145	-.0027153	.0004317
avg_prec	-.0000592	.0000555	-1.07	0.300	-.0001754	.000057
_cons	.1788416	.1936142	0.92	0.367	-.2263976	.5840809

```
.
. cap drop sample
```

```
. gen sample = e(sample)
```

```
. global avg_Conf_list = "avg_ConfGov avg_ConfTerr avg_ConfType3 avg_ConfType4 avg_ConfIntense avg_ConfNonIntense av
```

```
.
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
```

```
. areg avg_ConfIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare
```

```
Linear regression, absorbing indicators      Number of obs      =      7,864
Absorbed variable: isocode                 No. of categories =      20
                                           F( 10,      19)    =      22.34
                                           Prob > F           =      0.0000
                                           R-squared          =      0.9001
                                           Adj R-squared      =      0.8997
                                           Root MSE          =      0.0661
```

(Std. Err. adjusted for 20 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	.000627	.0067647	0.09	0.927	-.0135317	.0147857
avg_loggcppc	.0090039	.0072806	1.24	0.231	-.0062346	.0242424
avg_logpop	.0083131	.0062058	1.34	0.196	-.0046757	.021302
imr	.0000343	.000031	1.11	0.282	-.0000305	.0000992
logtttime	-.0101575	.0075508	-1.35	0.194	-.0259616	.0056465
logcellarea	.0029445	.0069357	0.42	0.676	-.0115721	.0174612
mountain2000	.0036745	.0077652	0.47	0.641	-.0125782	.0199271
ycoord	-.0016496	.0023478	-0.70	0.491	-.0065636	.0032644
avg_degtemper	-.0015347	.0011023	-1.39	0.180	-.0038419	.0007725
avg_prec	-.0000597	.0000542	-1.10	0.285	-.0001731	.0000538
_cons	.043076	.1687445	0.26	0.801	-.3101102	.3962622

```
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear
```

```
. areg avg_ConfIntra avg_logcapdist ${controls2} if avg_polity2>0 , a(isocode) cluster(iso) /*gives the result to compare with the first
```

```
Linear regression, absorbing indicators      Number of obs      =      39,501
Absorbed variable: isocode                 No. of categories =      108
                                           F( 11, 107)        =      1.38
                                           Prob > F           =      0.1938
                                           R-squared          =      0.7933
                                           Adj R-squared      =      0.7927
                                           Root MSE           =      0.0755
```

(Std. Err. adjusted for 108 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	.0094666	.014272	0.66	0.509	-.0188259	.0377592
avg_loggcpcpc	-.0160397	.0100084	-1.60	0.112	-.0358802	.0038008
avg_logpop	.0013423	.0010094	1.33	0.186	-.0006587	.0033432
imr	.0000621	.0000587	1.06	0.292	-.0000542	.0001784
logtttime	-.0037826	.0041029	-0.92	0.359	-.0119162	.004351
logcellarea	-.0091463	.0061326	-1.49	0.139	-.0213036	.003011
avg_logdist_LNC	-.0049452	.009689	-0.51	0.611	-.0241525	.0142621
mountain2000	.0041888	.0082671	0.51	0.613	-.0121996	.0205773
ycoord	-.0003657	.0004842	-0.76	0.452	-.0013255	.0005941
avg_degtemper	-.0003667	.0005357	-0.68	0.495	-.0014285	.0006952
avg_prec	1.26e-06	8.97e-06	0.14	0.888	-.0000165	.000019
_cons	.237504	.0798987	2.97	0.004	.0791141	.395894

```
.
.
. *part 1 conflict with prev_com dummy
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear
```

```
.
.
. keep if prev_com==1 /*new code*/
(44,230 observations deleted)
```

```
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the first
```

```
Linear regression, absorbing indicators      Number of obs      =      19,426
Absorbed variable: isocode                 No. of categories =      22
                                           F( 11, 21)         =     18795.94
                                           Prob > F           =      0.0000
                                           R-squared          =      0.7778
                                           Adj R-squared      =      0.7774
                                           Root MSE           =      0.0418
```

(Std. Err. adjusted for 22 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0012557	.0021624	-0.58	0.568	-.0057526	.0032412
avg_loggcpcpc	.0023433	.0097422	0.24	0.812	-.0179168	.0226033
avg_logpop	.0014347	.0007631	1.88	0.074	-.0001523	.0030216
imr	-8.77e-06	.0000105	-0.84	0.411	-.0000305	.000013
logtttime	-.0009316	.0025576	-0.36	0.719	-.0062503	.0043871
logcellarea	-.0200148	.008224	-2.43	0.024	-.0371176	-.0029121
avg_logdist_LNC	-.0043071	.0036296	-1.19	0.249	-.0118554	.0032411
mountain2000	.0050896	.0023161	2.20	0.039	.000273	.0099062
ycoord	-.0016263	.0007086	-2.30	0.032	-.0030999	-.0001527
avg_degtemper	-.0001731	.0000944	-1.83	0.081	-.0003695	.0000233
avg_prec	-.0000468	.0000344	-1.36	0.187	-.0001183	.0000246

_cons	.2860204	.0766476	3.73	0.001	.1266231	.4454177
-------	----------	----------	------	-------	----------	----------

```
.
. keep if avg_polity2<=0
(14,045 observations deleted)
```

```
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the se
```

```
Linear regression, absorbing indicators      Number of obs      =      6,335
Absorbed variable: isocode                 No. of categories =      11
                                           F( 10, 10)         =      .
                                           Prob > F           =      .
                                           R-squared          =      0.8503
                                           Adj R-squared      =      0.8498
                                           Root MSE          =      0.0544
```

(Std. Err. adjusted for 11 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0088802	.0121316	-0.73	0.481	-.035911	.0181507
avg_loggcppc	-.0101299	.0142601	-0.71	0.494	-.0419033	.0216436
avg_logpop	.0034941	.0028154	1.24	0.243	-.0027789	.0097671
imr	-3.93e-06	.0000102	-0.39	0.708	-.0000266	.0000188
logtttime	-.0072078	.007456	-0.97	0.356	-.0238208	.0094052
logcellarea	.001534	.0066099	0.23	0.821	-.0131937	.0162618
avg_logdist_LNC	-.0069915	.0060791	-1.15	0.277	-.0205365	.0065536
mountain2000	.0024415	.0077411	0.32	0.759	-.0148067	.0196898
ycoord	-.0031218	.0033729	-0.93	0.376	-.0106371	.0043934
avg_degtemper	-.000897	.0009251	-0.97	0.355	-.0029582	.0011642
avg_prec	-.000072	.0000611	-1.18	0.266	-.0002081	.0000642
_cons	.3816792	.3551023	1.07	0.308	-.4095381	1.172897

```
.
. cap drop sample
. gen sample = e(sample)
. global avg_Conf_list = "avg_ConfGov avg_ConfTerr avg_ConfType3 avg_ConfType4 avg_ConfIntense avg_ConfNonIntense av
.
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
. areg avg_ConfIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare
```

```
Linear regression, absorbing indicators      Number of obs      =      6,335
Absorbed variable: isocode                 No. of categories =      11
                                           F( 10, 10)         =     437.70
                                           Prob > F           =      0.0000
                                           R-squared          =      0.8493
                                           Adj R-squared      =      0.8489
                                           Root MSE          =      0.0546
```

(Std. Err. adjusted for 11 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	-.0096529	.0060926	-1.58	0.144	-.0232281	.0039222
avg_loggcppc	-.0095562	.0130909	-0.73	0.482	-.0387245	.0196121
avg_logpop	.0040359	.0035945	1.12	0.288	-.0039732	.012045
imr	-3.56e-06	9.75e-06	-0.37	0.723	-.0000253	.0000182
logtttime	-.0088868	.0096678	-0.92	0.380	-.0304281	.0126544
logcellarea	.0015839	.0062317	0.25	0.805	-.0123012	.015469
mountain2000	.0017488	.0090052	0.19	0.850	-.018316	.0218137
ycoord	-.0029238	.0031757	-0.92	0.379	-.0099998	.0041521
avg_degtemper	-.0010602	.0011976	-0.89	0.397	-.0037286	.0016081
avg_prec	-.0000725	.0000606	-1.20	0.259	-.0002076	.0000625

_cons	.3335927	.2812038	1.19	0.263	-.2929685	.9601538
-------	----------	----------	------	-------	-----------	----------

```
.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear
.
. areg avg_ConfIntra avg_logcapdist ${controls2} if avg_polity2>0 , a(isocode) cluster(iso) /*gives the result to compare with the fi
```

```
Linear regression, absorbing indicators      Number of obs      =      39,501
Absorbed variable: isocode                 No. of categories =      108
                                           F( 11, 107)        =      1.38
                                           Prob > F           =      0.1938
                                           R-squared          =      0.7933
                                           Adj R-squared      =      0.7927
                                           Root MSE          =      0.0755
```

(Std. Err. adjusted for 108 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	.0094666	.014272	0.66	0.509	-.0188259	.0377592
avg_loggcppc	-.0160397	.0100084	-1.60	0.112	-.0358802	.0038008
avg_logpop	.0013423	.0010094	1.33	0.186	-.0006587	.0033432
imr	.0000621	.0000587	1.06	0.292	-.0000542	.0001784
logtttime	-.0037826	.0041029	-0.92	0.359	-.0119162	.004351
logcellarea	-.0091463	.0061326	-1.49	0.139	-.0213036	.003011
avg_logdist_LNC	-.0049452	.009689	-0.51	0.611	-.0241525	.0142621
mountain2000	.0041888	.0082671	0.51	0.613	-.0121996	.0205773
ycoord	-.0003657	.0004842	-0.76	0.452	-.0013255	.0005941
avg_degtemper	-.0003667	.0005357	-0.68	0.495	-.0014285	.0006952
avg_prec	1.26e-06	8.97e-06	0.14	0.888	-.0000165	.000019
_cons	.237504	.0798987	2.97	0.004	.0791141	.395894

```
.
.
. *part 1 conflict with the extractivecol dummy
.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear
```

```
.
. keep if extractivecol==1 /*new code*/
(44,062 observations deleted)
```

```
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the fi
```

```
Linear regression, absorbing indicators      Number of obs      =      18,031
Absorbed variable: isocode                 No. of categories =      91
                                           F( 11, 90)         =      1.48
                                           Prob > F           =      0.1531
                                           R-squared          =      0.8077
                                           Adj R-squared      =      0.8066
                                           Root MSE          =      0.1209
```

(Std. Err. adjusted for 91 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0068088	.016812	-0.40	0.686	-.0402087	.0265911
avg_loggcppc	-.0316456	.0212004	-1.49	0.139	-.0737638	.0104727
avg_logpop	.0095055	.0044249	2.15	0.034	.0007146	.0182963
imr	.0000744	.0000406	1.83	0.070	-6.25e-06	.0001551
logtttime	-.0068736	.0079399	-0.87	0.389	-.0226476	.0089003
logcellarea	.0421727	.0269981	1.56	0.122	-.0114638	.0958092
avg_logdist_LNC	.0016829	.0067254	0.25	0.803	-.0116784	.0150441
mountain2000	.0330855	.0150856	2.19	0.031	.0031154	.0630556
ycoord	-.0042592	.0024601	-1.73	0.087	-.0091466	.0006282
avg_degtemper	.0030165	.0013973	2.16	0.034	.0002406	.0057924

avg_prec	.0000184	.0000171	1.07	0.286	-.0000157	.0000525
_cons	-.078982	.2604917	-0.30	0.762	-.5964942	.4385303

```
.
. keep if avg_polity2<=0
(11,874 observations deleted)
```

```
.
. areg avg_ConfIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the se
```

Linear regression, absorbing indicators	Number of obs	=	7,445
Absorbed variable: isocode	No. of categories	=	37
	F(11, 36)	=	4.73
	Prob > F	=	0.0002
	R-squared	=	0.8301
	Adj R-squared	=	0.8290
	Root MSE	=	0.1077

(Std. Err. adjusted for 37 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0387946	.0169666	-2.29	0.028	-.0732044	-.0043847
avg_loggcppc	-.0263285	.0267127	-0.99	0.331	-.0805043	.0278474
avg_logpop	.0076568	.0055982	1.37	0.180	-.0036969	.0190106
imr	.00008	.0000271	2.95	0.006	.000025	.000135
logtttime	-.0230315	.0075278	-3.06	0.004	-.0382986	-.0077643
logcellarea	.0556039	.0284531	1.95	0.058	-.0021017	.1133094
avg_logdist_LNC	.0120344	.0090443	1.33	0.192	-.0063082	.030377
mountain2000	-.0011111	.0177531	-0.06	0.950	-.0371161	.034894
ycoord	-.0090992	.0056687	-1.61	0.117	-.0205959	.0023975
avg_degtemper	-.0004814	.002376	-0.20	0.841	-.0053002	.0043374
avg_prec	4.77e-06	.0000602	0.08	0.937	-.0001173	.0001268
_cons	.2486716	.3251443	0.76	0.449	-.4107516	.9080947

```
.
. cap drop sample

. gen sample = e(sample)

. global avg_Conf_list = "avg_ConfGov avg_ConfTerr avg_ConfType3 avg_ConfType4 avg_ConfIntense avg_ConfNonIntense av

.
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)

. areg avg_ConfIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare
```

Linear regression, absorbing indicators	Number of obs	=	7,445
Absorbed variable: isocode	No. of categories	=	37
	F(10, 36)	=	5.11
	Prob > F	=	0.0001
	R-squared	=	0.8242
	Adj R-squared	=	0.8231
	Root MSE	=	0.1096

(Std. Err. adjusted for 37 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	.0098159	.0104254	0.94	0.353	-.0113278	.0309596
avg_loggcppc	-.0208851	.0246369	-0.85	0.402	-.070851	.0290808
avg_logpop	.0137317	.0063323	2.17	0.037	.0008892	.0265741
imr	.0000674	.0000248	2.72	0.010	.0000171	.0001177
logtttime	-.0282149	.0081606	-3.46	0.001	-.0447654	-.0116644
logcellarea	.0487676	.0297841	1.64	0.110	-.0116375	.1091726
mountain2000	-.0053751	.0181333	-0.30	0.769	-.0421511	.0314008
ycoord	-.0081396	.0052078	-1.56	0.127	-.0187014	.0024223
avg_degtemper	-.0006299	.0026042	-0.24	0.810	-.0059114	.0046516

avg_prec	.0000119	.0000597	0.20	0.843	-.0001091	.000133
_cons	.0074469	.2705038	0.03	0.978	-.5411602	.5560541

```
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetconfdata.dta", clear
```

```
. areg avg_ConfIntra avg_logcapdist ${controls2} if avg_polity2>0 , a(isocode) cluster(iso) /*gives the result to co
```

Linear regression, absorbing indicators	Number of obs	=	39,501
Absorbed variable: isocode	No. of categories	=	108
	F(11, 107)	=	1.38
	Prob > F	=	0.1938
	R-squared	=	0.7933
	Adj R-squared	=	0.7927
	Root MSE	=	0.0755

(Std. Err. adjusted for **108** clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	.0094666	.014272	0.66	0.509	-.0188259	.0377592
avg_loggcppc	-.0160397	.0100084	-1.60	0.112	-.0358802	.0038008
avg_logpop	.0013423	.0010094	1.33	0.186	-.0006587	.0033432
imr	.0000621	.0000587	1.06	0.292	-.0000542	.0001784
logtttime	-.0037826	.0041029	-0.92	0.359	-.0119162	.004351
logcellarea	-.0091463	.0061326	-1.49	0.139	-.0213036	.003011
avg_logdist_LNC	-.0049452	.009689	-0.51	0.611	-.0241525	.0142621
mountain2000	.0041888	.0082671	0.51	0.613	-.0121996	.0205773
ycoord	-.0003657	.0004842	-0.76	0.452	-.0013255	.0005941
avg_degtemper	-.0003667	.0005357	-0.68	0.495	-.0014285	.0006952
avg_prec	1.26e-06	8.97e-06	0.14	0.888	-.0000165	.000019
_cons	.237504	.0798987	2.97	0.004	.0791141	.395894

```
.
.
.
. clear
.
.
.
. *"Table 3 part 2 Onset"
.
. //partie Onset dummy Africa
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta"

. keep if Africa==1 /*new code*/
(54,949 observations deleted)
```

```
. areg avg_onsetIntra avg_logcapdist ${controls2}, a(isocode) cluster(iso) /*gives the result to compare with the 5t
```

Linear regression, absorbing indicators	Number of obs	=	8,483
Absorbed variable: isocode	No. of categories	=	50
	F(11, 49)	=	1.33
	Prob > F	=	0.2353
	R-squared	=	0.0552
	Adj R-squared	=	0.0485
	Root MSE	=	0.0021

(Std. Err. adjusted for 50 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.000176	.000152	-1.16	0.252	-.0004815	.0001294
avg_loggcppc	.0001095	.0000684	1.60	0.116	-.000028	.000247
avg_logpop	.0000605	.0000256	2.37	0.022	9.14e-06	.000112
imr	1.06e-07	2.49e-07	0.43	0.673	-3.94e-07	6.06e-07
logtttime	.0000137	.0000426	0.32	0.750	-.0000719	.0000992
logcellarea	-2.16e-06	.0001998	-0.01	0.991	-.0004037	.0003993
avg_logdist_LNC	.0000479	.00008	0.60	0.552	-.0001128	.0002087
mountain2000	-.0001818	.0001319	-1.38	0.174	-.0004469	.0000832
ycoord	-6.75e-06	8.27e-06	-0.82	0.418	-.0000234	9.86e-06
avg_degtemper	-7.25e-06	.0000103	-0.70	0.485	-.0000279	.0000134
avg_prec	-1.17e-07	2.00e-07	-0.59	0.561	-5.20e-07	2.85e-07
_cons	-.0003054	.0019146	-0.16	0.874	-.0041529	.0035421

```
.
. keep if avg_polity2<=0
(1,021 observations deleted)
```

```
.
. areg avg_onsetIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the 6
```

Linear regression, absorbing indicators	Number of obs	=	7,555
Absorbed variable: isocode	No. of categories	=	42
	F(11, 41)	=	1.45
	Prob > F	=	0.1870
	R-squared	=	0.0560
	Adj R-squared	=	0.0494
	Root MSE	=	0.0022

(Std. Err. adjusted for 42 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.000193	.0001649	-1.17	0.249	-.0005261	.0001401
avg_loggcppc	.0001647	.0000885	1.86	0.070	-.0000141	.0003434
avg_logpop	.0000634	.0000292	2.17	0.036	4.38e-06	.0001223
imr	1.17e-07	2.91e-07	0.40	0.691	-4.72e-07	7.05e-07
logtttime	.0000105	.0000455	0.23	0.818	-.0000814	.0001024
logcellarea	.000047	.0002269	0.21	0.837	-.0004113	.0005053
avg_logdist_LNC	.0000467	.0000809	0.58	0.567	-.0001167	.00021
mountain2000	-.0002272	.0001656	-1.37	0.178	-.0005615	.0001072
ycoord	-6.65e-06	9.00e-06	-0.74	0.464	-.0000248	.0000115
avg_degtemper	-9.14e-06	.0000117	-0.78	0.437	-.0000327	.0000144
avg_prec	-1.22e-07	2.10e-07	-0.58	0.563	-5.46e-07	3.01e-07
_cons	-.000891	.0021629	-0.41	0.683	-.0052591	.003477

```
.
. cap drop sample
. gen sample = e(sample)
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
```

```
. areg avg_onsetIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare
```

```
Linear regression, absorbing indicators      Number of obs      =      7,555
Absorbed variable: isocode                  No. of categories  =       42
                                           F(   10,    41)    =       1.26
                                           Prob > F          =      0.2833
                                           R-squared         =      0.0543
                                           Adj R-squared     =      0.0479
                                           Root MSE         =      0.0022
```

(Std. Err. adjusted for 42 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	-5.34e-07	.0000602	-0.01	0.993	-.0001221	.000121
avg_loggcppc	.0001954	.0001047	1.87	0.069	-.000016	.0004068
avg_logpop	.0000907	.0000428	2.12	0.040	4.18e-06	.0001772
imr	-1.65e-08	2.10e-07	-0.08	0.938	-4.40e-07	4.07e-07
logtttime	-8.25e-06	.0000412	-0.20	0.842	-.0000914	.0000749
logcellarea	-9.53e-06	.0002244	-0.04	0.966	-.0004626	.0004436
mountain2000	-.0002247	.0001604	-1.40	0.169	-.0005486	.0000991
ycoord	-4.59e-06	8.46e-06	-0.54	0.590	-.0000217	.0000125
avg_degtemper	-6.78e-06	.0000121	-0.56	0.578	-.0000312	.0000176
avg_prec	-7.98e-08	1.90e-07	-0.42	0.677	-4.64e-07	3.04e-07
_cons	-.0016782	.0022498	-0.75	0.460	-.0062219	.0028654

```
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta", clear
```

```
. areg avg_onsetIntra avg_logcapdist ${controls2} if avg_polity2>0, a(isocode) cluster(iso) /*gives the result to co
```

```
Linear regression, absorbing indicators      Number of obs      =     21,203
Absorbed variable: isocode                  No. of categories  =      67
                                           F(   11,    66)    =       1.40
                                           Prob > F          =      0.1961
                                           R-squared         =      0.0458
                                           Adj R-squared     =      0.0423
                                           Root MSE         =      0.0012
```

(Std. Err. adjusted for 67 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0000545	.000037	-1.47	0.145	-.0001284	.0000193
avg_loggcppc	-.0000987	.0000686	-1.44	0.155	-.0002357	.0000384
avg_logpop	.0000111	8.41e-06	1.33	0.190	-5.65e-06	.0000279
imr	-8.35e-07	5.17e-07	-1.61	0.111	-1.87e-06	1.97e-07
logtttime	-.0000309	.0000306	-1.01	0.317	-.0000919	.0000302
logcellarea	.0000571	.000071	0.80	0.424	-.0000847	.000199
avg_logdist_LNC	.0000786	.0000442	1.78	0.080	-9.70e-06	.0001669
mountain2000	.0000273	.0000286	0.95	0.344	-.0000298	.0000843
ycoord	3.55e-06	1.62e-06	2.20	0.031	3.26e-07	6.78e-06
avg_degtemper	-7.57e-06	5.60e-06	-1.35	0.181	-.0000188	3.62e-06
avg_prec	7.70e-08	6.79e-08	1.13	0.261	-5.86e-08	2.13e-07
_cons	.0005976	.0004202	1.42	0.160	-.0002413	.0014364

```

.
. //Partie Onset dummy Asia
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta",clear

. keep if Asia==1 /*new code*/
(52,131 observations deleted)

.
. areg avg_onsetIntra avg_logcapdist ${controls2}, a(isocode) cluster(iso) /*gives the result to compare with the 5t

```

```

Linear regression, absorbing indicators      Number of obs      =      11,461
Absorbed variable: isocode                 No. of categories =       38
                                           F( 11, 37)         =       1.14
                                           Prob > F           =      0.3585
                                           R-squared          =      0.0390
                                           Adj R-squared      =      0.0349
                                           Root MSE          =      0.0022

```

(Std. Err. adjusted for 38 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0001063	.0001513	-0.70	0.487	-.0004128	.0002003
avg_loggcppc	-.0000216	.0000895	-0.24	0.811	-.000203	.0001598
avg_logpop	.0000165	.0000179	0.92	0.361	-.0000197	.0000528
imr	-1.75e-07	2.95e-07	-0.59	0.557	-7.72e-07	4.22e-07
logtttime	-.0000888	.0000568	-1.56	0.127	-.000204	.0000263
logcellarea	.0000572	.0000963	0.59	0.556	-.000138	.0002524
avg_logdist_LNC	.0001991	.0001154	1.73	0.093	-.0000347	.0004329
mountain2000	.0001171	.0000951	1.23	0.226	-.0000755	.0003098
ycoord	-6.14e-06	.0000111	-0.55	0.585	-.0000287	.0000164
avg_degtemper	-.0000119	5.45e-06	-2.19	0.035	-.000023	-9.10e-07
avg_prec	1.80e-07	1.04e-07	1.73	0.092	-3.07e-08	3.92e-07
_cons	-.0001389	.0015175	-0.09	0.928	-.0032137	.0029359

```

.
.
. keep if avg_polity2<=0
(2,771 observations deleted)

.
. areg avg_onsetIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the 6

```

```

Linear regression, absorbing indicators      Number of obs      =      9,232
Absorbed variable: isocode                 No. of categories =       27
                                           F( 11, 26)         =       0.74
                                           Prob > F           =      0.6918
                                           R-squared          =      0.0375
                                           Adj R-squared      =      0.0336
                                           Root MSE          =      0.0018

```

(Std. Err. adjusted for 27 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0001948	.0001739	-1.12	0.273	-.0005523	.0001626
avg_loggcppc	.0001216	.0000739	1.64	0.112	-.0000304	.0002735
avg_logpop	.0000242	.0000179	1.36	0.186	-.0000125	.000061
imr	1.77e-07	1.92e-07	0.92	0.367	-2.19e-07	5.72e-07
logtttime	-.0000649	.0000524	-1.24	0.226	-.0001726	.0000427
logcellarea	2.44e-06	.0000961	0.03	0.980	-.0001952	.0002001
avg_logdist_LNC	.0001977	.0001309	1.51	0.143	-.0000714	.0004669
mountain2000	.0001276	.000088	1.45	0.159	-.0000533	.0003086
ycoord	-5.29e-06	.0000105	-0.50	0.620	-.0000269	.0000164
avg_degtemper	-4.73e-06	4.99e-06	-0.95	0.352	-.000015	5.53e-06
avg_prec	1.83e-08	6.86e-08	0.27	0.792	-1.23e-07	1.59e-07
_cons	-.0006859	.0015769	-0.43	0.667	-.0039273	.0025554

```

.
. cap drop sample
. gen sample = e(sample)
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
. areg avg_onsetIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare

```

```

Linear regression, absorbing indicators      Number of obs      =      9,232
Absorbed variable: isocode                No. of categories =      27
                                           F( 10, 26)         =      0.81
                                           Prob > F           =      0.6194
                                           R-squared          =      0.0348
                                           Adj R-squared       =      0.0310
                                           Root MSE           =      0.0018

```

(Std. Err. adjusted for 27 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	.0001479	.000105	1.41	0.171	-.0000679	.0003637
avg_loggcppc	.0001239	.0000776	1.60	0.123	-.0000356	.0002834
avg_logpop	.0000421	.0000222	1.89	0.070	-3.64e-06	.0000877
imr	1.59e-07	1.71e-07	0.93	0.361	-1.93e-07	5.11e-07
logtttime	-.0000865	.0000596	-1.45	0.158	-.000209	.0000359
logcellarea	4.92e-06	.0001009	0.05	0.961	-.0002025	.0002124
mountain2000	.0000879	.0000733	1.20	0.241	-.0000627	.0002386
ycoord	-1.84e-06	8.55e-06	-0.22	0.831	-.0000194	.0000157
avg_degtemper	-8.17e-06	5.94e-06	-1.38	0.181	-.0000204	4.04e-06
avg_prec	2.48e-08	8.77e-08	0.28	0.780	-1.56e-07	2.05e-07
_cons	-.0018082	.0012841	-1.41	0.171	-.0044477	.0008313

```

.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta", clear

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2} if avg_polity2>0, a(isocode) cluster(iso) /*gives the result to compare

```

```

Linear regression, absorbing indicators      Number of obs      =      21,203
Absorbed variable: isocode                No. of categories =      67
                                           F( 11, 66)         =      1.40
                                           Prob > F           =      0.1961
                                           R-squared          =      0.0458
                                           Adj R-squared       =      0.0423
                                           Root MSE           =      0.0012

```

(Std. Err. adjusted for 67 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0000545	.000037	-1.47	0.145	-.0001284	.0000193
avg_loggcppc	-.0000987	.0000686	-1.44	0.155	-.0002357	.0000384
avg_logpop	.0000111	8.41e-06	1.33	0.190	-5.65e-06	.0000279
imr	-8.35e-07	5.17e-07	-1.61	0.111	-1.87e-06	1.97e-07
logtttime	-.0000309	.0000306	-1.01	0.317	-.0000919	.0000302
logcellarea	.0000571	.000071	0.80	0.424	-.0000847	.000199
avg_logdist_LNC	.0000786	.0000442	1.78	0.080	-9.70e-06	.0001669
mountain2000	.0000273	.0000286	0.95	0.344	-.0000298	.0000843
ycoord	3.55e-06	1.62e-06	2.20	0.031	3.26e-07	6.78e-06
avg_degtemper	-7.57e-06	5.60e-06	-1.35	0.181	-.0000188	3.62e-06
avg_prec	7.70e-08	6.79e-08	1.13	0.261	-5.86e-08	2.13e-07
_cons	.0005976	.0004202	1.42	0.160	-.0002413	.0014364


```

. //Partie Onset dummy prev_com
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta",clear

. keep if prev_com==1 /*new code*/
(44,373 observations deleted)

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2}, a(isocode) cluster(iso) /*gives the result to compare with the 5t

```

```

Linear regression, absorbing indicators      Number of obs      =      19,271
Absorbed variable: isocode                No. of categories =       22
                                           F( 11,      21)    =      205.32
                                           Prob > F           =      0.0000
                                           R-squared          =      0.0450
                                           Adj R-squared      =      0.0434
                                           Root MSE          =      0.0011

```

(Std. Err. adjusted for 22 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0001995	.0001915	-1.04	0.309	-.0005978	.0001988
avg_loggcppc	-8.99e-06	.000043	-0.21	0.836	-.0000983	.0000803
avg_logpop	9.66e-06	4.68e-06	2.06	0.052	-7.56e-08	.0000194
imr	6.36e-09	5.44e-08	0.12	0.908	-1.07e-07	1.20e-07
logtttime	.0000219	9.43e-06	2.32	0.030	2.28e-06	.0000415
logcellarea	-.0001087	.0001034	-1.05	0.305	-.0003238	.0001064
avg_logdist_LNC	.0001897	.0001731	1.10	0.285	-.0001702	.0005496
mountain2000	.0000514	.0000365	1.41	0.174	-.0000246	.0001273
ycoord	-2.54e-06	5.28e-06	-0.48	0.636	-.0000135	8.45e-06
avg_degtemper	-5.33e-08	1.30e-06	-0.04	0.968	-2.75e-06	2.65e-06
avg_prec	9.89e-08	8.96e-08	1.10	0.282	-8.75e-08	2.85e-07
_cons	.0008399	.0007787	1.08	0.293	-.0007794	.0024593

```

. keep if avg_polity2<=0
(0 observations deleted)

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the 6

```

```

Linear regression, absorbing indicators      Number of obs      =      19,271
Absorbed variable: isocode                No. of categories =       22
                                           F( 11,      21)    =      205.32
                                           Prob > F           =      0.0000
                                           R-squared          =      0.0450
                                           Adj R-squared      =      0.0434
                                           Root MSE          =      0.0011

```

(Std. Err. adjusted for 22 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0001995	.0001915	-1.04	0.309	-.0005978	.0001988
avg_loggcppc	-8.99e-06	.000043	-0.21	0.836	-.0000983	.0000803
avg_logpop	9.66e-06	4.68e-06	2.06	0.052	-7.56e-08	.0000194
imr	6.36e-09	5.44e-08	0.12	0.908	-1.07e-07	1.20e-07
logtttime	.0000219	9.43e-06	2.32	0.030	2.28e-06	.0000415
logcellarea	-.0001087	.0001034	-1.05	0.305	-.0003238	.0001064
avg_logdist_LNC	.0001897	.0001731	1.10	0.285	-.0001702	.0005496
mountain2000	.0000514	.0000365	1.41	0.174	-.0000246	.0001273
ycoord	-2.54e-06	5.28e-06	-0.48	0.636	-.0000135	8.45e-06
avg_degtemper	-5.33e-08	1.30e-06	-0.04	0.968	-2.75e-06	2.65e-06
avg_prec	9.89e-08	8.96e-08	1.10	0.282	-8.75e-08	2.85e-07
_cons	.0008399	.0007787	1.08	0.293	-.0007794	.0024593

```

.
. cap drop sample
. gen sample = e(sample)
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
. areg avg_onsetIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare

```

```

Linear regression, absorbing indicators      Number of obs      =      19,271
Absorbed variable: isocode                  No. of categories =       22
                                           F( 10,      21) =     549.73
                                           Prob > F          =     0.0000
                                           R-squared         =     0.0414
                                           Adj R-squared     =     0.0399
                                           Root MSE         =     0.0011

```

(Std. Err. adjusted for 22 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	.0000433	.0000466	0.93	0.363	-.0000535	.0001402
avg_loggcppc	-.0000377	.0000175	-2.16	0.043	-.0000741	-1.32e-06
avg_logpop	.0000159	4.16e-06	3.83	0.001	7.27e-06	.0000246
imr	2.55e-09	3.62e-08	0.07	0.945	-7.28e-08	7.79e-08
logtttime	1.10e-06	.0000155	0.07	0.944	-.0000311	.0000333
logcellarea	-.0000777	.0000815	-0.95	0.351	-.0002473	.0000918
mountain2000	.0000257	.0000271	0.95	0.352	-.0000306	.000082
ycoord	5.65e-07	3.08e-06	0.18	0.856	-5.84e-06	6.97e-06
avg_degtemper	2.33e-07	1.66e-06	0.14	0.889	-3.21e-06	3.68e-06
avg_prec	5.05e-08	9.41e-08	0.54	0.597	-1.45e-07	2.46e-07
_cons	.0004299	.0004529	0.95	0.353	-.0005119	.0013716

```

.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta", clear

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2} if avg_polity2>0, a(isocode) cluster(iso) /*gives the result to compare

```

```

Linear regression, absorbing indicators      Number of obs      =      21,203
Absorbed variable: isocode                  No. of categories =       67
                                           F( 11,      66) =       1.40
                                           Prob > F          =     0.1961
                                           R-squared         =     0.0458
                                           Adj R-squared     =     0.0423
                                           Root MSE         =     0.0012

```

(Std. Err. adjusted for 67 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0000545	.000037	-1.47	0.145	-.0001284	.0000193
avg_loggcppc	-.0000987	.0000686	-1.44	0.155	-.0002357	.0000384
avg_logpop	.0000111	8.41e-06	1.33	0.190	-5.65e-06	.0000279
imr	-8.35e-07	5.17e-07	-1.61	0.111	-1.87e-06	1.97e-07
logtttime	-.0000309	.0000306	-1.01	0.317	-.0000919	.0000302
logcellarea	.0000571	.000071	0.80	0.424	-.0000847	.000199
avg_logdist_LNC	.0000786	.0000442	1.78	0.080	-9.70e-06	.0001669
mountain2000	.0000273	.0000286	0.95	0.344	-.0000298	.0000843
ycoord	3.55e-06	1.62e-06	2.20	0.031	3.26e-07	6.78e-06
avg_degtemper	-7.57e-06	5.60e-06	-1.35	0.181	-.0000188	3.62e-06
avg_prec	7.70e-08	6.79e-08	1.13	0.261	-5.86e-08	2.13e-07
_cons	.0005976	.0004202	1.42	0.160	-.0002413	.0014364

```

. //Partie Onset dummy extractivecol
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta",clear

. keep if extractivecol==1 /*new code*/
(45,000 observations deleted)

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2}, a(isocode) cluster(iso) /*gives the result to compare with the 5t

```

```

Linear regression, absorbing indicators      Number of obs      =      17,237
Absorbed variable: isocode                No. of categories =       89
                                           F(   11,      88)  =       3.02
                                           Prob > F           =      0.0019
                                           R-squared          =      0.0419
                                           Adj R-squared      =      0.0363
                                           Root MSE          =      0.0022

```

(Std. Err. adjusted for 89 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0002067	.0001226	-1.69	0.096	-.0004504	.0000371
avg_loggcppc	.000036	.0000581	0.62	0.537	-.0000794	.0001515
avg_logpop	.0000542	.0000182	2.97	0.004	.000018	.0000904
imr	-2.60e-07	3.67e-07	-0.71	0.481	-9.88e-07	4.69e-07
logtttime	-.0000183	.0000494	-0.37	0.712	-.0001166	.00008
logcellarea	.0000539	.0000885	0.61	0.544	-.000122	.0002298
avg_logdist_LNC	.0001446	.0000688	2.10	0.038	7.85e-06	.0002813
mountain2000	.0000413	.0001243	0.33	0.741	-.0002057	.0002883
ycoord	1.19e-06	4.30e-06	0.28	0.783	-7.36e-06	9.75e-06
avg_degtemper	-6.71e-06	6.67e-06	-1.01	0.317	-.00002	6.54e-06
avg_prec	1.42e-07	7.86e-08	1.81	0.074	-1.40e-08	2.98e-07
_cons	-.0004001	.0008588	-0.47	0.642	-.0021068	.0013066

```

. keep if avg_polity2<=0
(8,113 observations deleted)

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2} , a(isocode) cluster(iso) /*gives the result to compare with the 6

```

```

Linear regression, absorbing indicators      Number of obs      =      10,162
Absorbed variable: isocode                No. of categories =       55
                                           F(   11,      54)  =       1.98
                                           Prob > F           =      0.0486
                                           R-squared          =      0.0513
                                           Adj R-squared      =      0.0452
                                           Root MSE          =      0.0024

```

(Std. Err. adjusted for 55 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0003176	.0001678	-1.89	0.064	-.0006541	.0000189
avg_loggcppc	.0002104	.0000923	2.28	0.027	.0000254	.0003954
avg_logpop	.0000533	.0000263	2.03	0.048	5.69e-07	.0001061
imr	2.51e-07	3.09e-07	0.81	0.419	-3.67e-07	8.70e-07
logtttime	-.0000579	.000061	-0.95	0.347	-.0001802	.0000645
logcellarea	.0000531	.0001498	0.35	0.725	-.0002472	.0003534
avg_logdist_LNC	.0001574	.0000858	1.84	0.072	-.0000146	.0003293
mountain2000	-.00003	.0001564	-0.19	0.849	-.0003435	.0002835
ycoord	-1.49e-06	8.04e-06	-0.19	0.854	-.0000176	.0000146
avg_degtemper	-3.23e-06	9.31e-06	-0.35	0.730	-.0000219	.0000154
avg_prec	-5.95e-08	1.63e-07	-0.37	0.717	-3.86e-07	2.67e-07
_cons	-.0010557	.0016745	-0.63	0.531	-.0044128	.0023015

```

.
. cap drop sample
. gen sample = e(sample)
. global controls2x = substr("$controls2","avg_logdist_LNC","",.)
. areg avg_onsetIntra avg_logdist_LNC ${controls2x} if sample, a(isocode) cluster(iso) /*gives the result to compare

```

```

Linear regression, absorbing indicators      Number of obs      =      10,162
Absorbed variable: isocode                 No. of categories =       55
                                           F( 10,      54) =      1.90
                                           Prob > F          =      0.0657
                                           R-squared         =      0.0473
                                           Adj R-squared     =      0.0413
                                           Root MSE         =      0.0024

```

(Std. Err. adjusted for 55 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logdist_LNC	.0001039	.0000712	1.46	0.150	-.0000389	.0002467
avg_loggcppc	.0002635	.0001157	2.28	0.027	.0000315	.0004955
avg_logpop	.0000974	.0000329	2.96	0.005	.0000314	.0001633
imr	3.00e-08	2.36e-07	0.13	0.899	-4.44e-07	5.04e-07
logttime	-.0001025	.0000594	-1.73	0.090	-.0002215	.0000165
logcellarea	.0000128	.0001586	0.08	0.936	-.0003051	.0003307
mountain2000	-.0000853	.0001284	-0.66	0.509	-.0003427	.000172
ycoord	7.10e-07	8.17e-06	0.09	0.931	-.0000157	.0000171
avg_degtemper	-3.28e-06	.0000103	-0.32	0.752	-.0000239	.0000174
avg_prec	2.54e-08	1.62e-07	0.16	0.876	-3.00e-07	3.50e-07
_cons	-.0027956	.0014117	-1.98	0.053	-.0056258	.0000346

```

.
. use"D:\documents\unif\Master1\term 2\Econometrics seminar\data\data\newbygid_datasetonset.dta", clear

```

```

. areg avg_onsetIntra avg_logcapdist ${controls2} if avg_polity2>0, a(isocode) cluster(iso) /*gives the result to compare

```

```

Linear regression, absorbing indicators      Number of obs      =      21,203
Absorbed variable: isocode                 No. of categories =       67
                                           F( 11,      66) =      1.40
                                           Prob > F          =      0.1961
                                           R-squared         =      0.0458
                                           Adj R-squared     =      0.0423
                                           Root MSE         =      0.0012

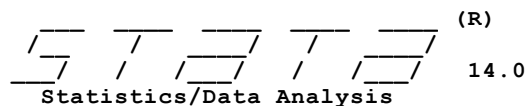
```

(Std. Err. adjusted for 67 clusters in isocode)

avg_onsetIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
avg_logcapdist	-.0000545	.000037	-1.47	0.145	-.0001284	.0000193
avg_loggcppc	-.0000987	.0000686	-1.44	0.155	-.0002357	.0000384
avg_logpop	.0000111	8.41e-06	1.33	0.190	-5.65e-06	.0000279
imr	-8.35e-07	5.17e-07	-1.61	0.111	-1.87e-06	1.97e-07
logttime	-.0000309	.0000306	-1.01	0.317	-.0000919	.0000302
logcellarea	.0000571	.000071	0.80	0.424	-.0000847	.000199
avg_logdist_LNC	.0000786	.0000442	1.78	0.080	-9.70e-06	.0001669
mountain2000	.0000273	.0000286	0.95	0.344	-.0000298	.0000843
ycoord	3.55e-06	1.62e-06	2.20	0.031	3.26e-07	6.78e-06
avg_degtemper	-7.57e-06	5.60e-06	-1.35	0.181	-.0000188	3.62e-06
avg_prec	7.70e-08	6.79e-08	1.13	0.261	-5.86e-08	2.13e-07
_cons	.0005976	.0004202	1.42	0.160	-.0002413	.0014364

```
.  
end of do-file
```

```
.
```



(R)

14.0

MP - Parallel Edition

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Notes:

1. Unicode is supported; see [help unicode advice](#).
2. Maximum number of variables is set to 5000; see [help set maxvar](#).

```
. doedit "C:\Users\bozet\OneDrive\Bureau\data - Copie\conflict part\stata natural ressources.do"

. do "C:\Users\bozet\AppData\Local\Temp\STD00000000.tmp"

. //Natural resources and conflicts part
. // Let's reuse and adapt sets of control variables
. use newbygid_confdata, clear

. //what they use :
. global gdpcontrols "avg_loggcppc avg_logpop"

. global controls "avg_loggcppc avg_logpop imr logtttime logcellarea avg_logdist_LNC"

. global controls2 "${controls} mountain2000 ycoord avg_degtemper avg_prec"

. // we can now proceed to regressions (by adapting them)
. *** Ore (all resources) ***
. **part 1 : full sample
. use newbygid_confdata, clear

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	49,909
	F(11, 125)	=	1.51
	Prob > F	=	0.1371
	R-squared	=	0.8249
	Adj R-squared	=	0.8245
	Root MSE	=	0.0810

(Std. Err. adjusted for 126 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0074003	.0041342	-1.79	0.076	-.0155823	.0007817
avg_loggcppc	-.0165973	.0082951	-2.00	0.048	-.0330145	-.0001802
avg_logpop	.0021043	.0011326	1.86	0.066	-.0001373	.0043459
imr	.0000439	.0000025	1.76	0.081	-5.51e-06	.0000933
logtttime	-.0053837	.0028074	-1.92	0.057	-.0109399	.0001725
logcellarea	-.0141959	.0078686	-1.80	0.074	-.0297689	.0013771
avg_logdist_LNC	.0036278	.0037972	0.96	0.341	-.0038873	.011143
mountain2000	-.0006365	.0063345	-0.10	0.920	-.0131732	.0119002
ycoord	-.0004234	.0004694	-0.90	0.369	-.0013523	.0005055
avg_degtemper	-.0010651	.0006054	-1.76	0.081	-.0022633	.000133
avg_prec	8.68e-06	.00001	0.87	0.388	-.0000112	.0000285
_cons	.3521579	.1218282	2.89	0.005	.1110449	.593271
isocode	absorbed		(126 categories)			

```

.
. **part 2 : anocracies
. use newbygid_confdata, clear

. keep if avg_polity2<=0
(48,680 observations deleted)

```

```

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      13,742
                                           F(   11,          47) =      2.05
                                           Prob > F            =      0.0443
                                           R-squared           =      0.8666
                                           Adj R-squared       =      0.8661
                                           Root MSE            =      0.0885

```

(Std. Err. adjusted for 48 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0122683	.0064375	-1.91	0.063	-.0252189	.0006824
avg_loggcppc	-.0167066	.0178596	-0.94	0.354	-.0526355	.0192222
avg_logpop	.007301	.0035782	2.04	0.047	.0001026	.0144993
imr	.0000383	.0000237	1.61	0.113	-.9.41e-06	.000086
logtttime	-.0143461	.0072806	-1.97	0.055	-.0289927	.0003005
logcellarea	.0032797	.0095314	0.34	0.732	-.0158949	.0224543
avg_logdist_LNC	.008885	.0075443	1.18	0.245	-.0062922	.0240622
mountain2000	-.0255869	.0196572	-1.30	0.199	-.065132	.0139583
ycoord	-.0038074	.0032049	-1.19	0.241	-.0102549	.00264
avg_degtemper	-.0039561	.0019234	-2.06	0.045	-.0078254	-.0000868
avg_prec	-4.99e-06	.0000318	-0.16	0.876	-.0000689	.000059
_cons	.3994148	.280952	1.42	0.162	-.1657877	.9646172

isocode absorbed (48 categories)

```

.
. **part 3 : democracies
. use newbygid_confdata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

```

```

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      36,167
                                           F(   11,          78) =      4.21
                                           Prob > F            =      0.0001
                                           R-squared           =      0.7804
                                           Adj R-squared       =      0.7799
                                           Root MSE            =      0.0772

```

(Std. Err. adjusted for 79 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0056643	.0047444	-1.19	0.236	-.0151098	.0037811
avg_loggcppc	-.0173279	.0103476	-1.67	0.098	-.0379285	.0032726
avg_logpop	.0008097	.0010581	0.77	0.446	-.0012968	.0029161
imr	.0000627	.0000567	1.11	0.272	-.0000501	.0001755
logtttime	-.0026742	.0028389	-0.94	0.349	-.008326	.0029775
logcellarea	-.0173555	.0094022	-1.85	0.069	-.036074	.0013629
avg_logdist_LNC	.0020089	.0037798	0.53	0.597	-.005516	.0095338
mountain2000	.0041485	.0062342	0.67	0.508	-.0082628	.0165597
ycoord	-.0004473	.0005645	-0.79	0.431	-.0015711	.0006766
avg_degtemper	-.0002453	.0005339	-0.46	0.647	-.0013082	.0008176
avg_prec	2.93e-06	.0000101	0.29	0.774	-.0000173	.0000231
_cons	.3551922	.1461264	2.43	0.017	.064277	.6461075

```

isocode | absorbed (79 categories)

.
. **part 4 : extractive colony (weak political institution and ressource rich countries)
.
. use newbygid_confddata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)

Linear regression, absorbing indicators      Number of obs      =      15,904
                                           F( 11,      70)    =      1.47
                                           Prob > F           =      0.1643
                                           R-squared          =      0.8058
                                           Adj R-squared      =      0.8049
                                           Root MSE          =      0.1256

```

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0233478	.0098397	-2.37	0.020	-.0429724	-.0037232
avg_loggcppc	-.0351178	.0241317	-1.46	0.150	-.083247	.0130113
avg_logpop	.0104709	.0040321	2.60	0.011	.0024292	.0185127
imr	.000077	.0000415	1.86	0.067	-5.66e-06	.0001597
logtttime	-.0040285	.0059902	-0.67	0.503	-.0159755	.0079184
logcellarea	.0551399	.0338292	1.63	0.108	-.0123304	.1226102
avg_logdist_LNC	.0029113	.0064324	0.45	0.652	-.0099176	.0157402
mountain2000	.0239357	.013859	1.73	0.089	-.0037052	.0515765
ycoord	-.0051207	.0028046	-1.83	0.072	-.0107144	.0004729
avg_degtemper	.0033587	.0015283	2.20	0.031	.0003106	.0064068
avg_prec	.0000225	.0000189	1.19	0.237	-.0000151	.0000602
_cons	-.1125055	.255329	-0.44	0.661	-.6217431	.3967321

isocode | absorbed (71 categories)

```

.
. **part 5 : Asia
. use newbygid_confddata, clear

. keep if Asia == 1
(51,846 observations deleted)

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      10,840
                                           F( 11,      34)    =      1.33
                                           Prob > F           =      0.2491
                                           R-squared          =      0.8012
                                           Adj R-squared      =      0.8003
                                           Root MSE          =      0.1317

```

(Std. Err. adjusted for 35 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0192308	.0144883	-1.33	0.193	-.0486746	.010213
avg_loggcppc	-.0727156	.052349	-1.39	0.174	-.1791015	.0336704
avg_logpop	.0059787	.0042252	1.42	0.166	-.0026079	.0145652
imr	.0000423	.0000404	1.05	0.302	-.0000398	.0001244
logtttime	-.0130102	.0072186	-1.80	0.080	-.0276803	.0016598
logcellarea	.0168188	.012481	1.35	0.187	-.0085457	.0421832
avg_logdist_LNC	.0146808	.0119346	1.23	0.227	-.0095733	.0389349
mountain2000	-.0238994	.0237119	-1.01	0.321	-.0720877	.0242889
ycoord	-.0026172	.0025166	-1.04	0.306	-.0077315	.0024971
avg_degtemper	-.0023028	.0012843	-1.79	0.082	-.0049128	.0003072

avg_prec	1.47e-07	.0000252	0.01	0.995	-.000051	.0000513
_cons	.7048185	.4659203	1.51	0.140	-.2420455	1.651683

isocode	absorbed	(35 categories)
---------	----------	-----------------

```
.
. **part 6 : Africa
.
. use newbygid_confdata, clear

. keep if Africa == 1
(54,641 observations deleted)

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      8,498
                                           F( 11,      38)    =      1.52
                                           Prob > F           =      0.1630
                                           R-squared          =      0.8510
                                           Adj R-squared      =      0.8501
                                           Root MSE          =      0.0897
```

(Std. Err. adjusted for 39 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0093962	.0060184	-1.56	0.127	-.0215798	.0027875
avg_loggcppc	-.0166764	.0176738	-0.94	0.351	-.0524551	.0191024
avg_logpop	.0033212	.00297	1.12	0.270	-.0026912	.0093336
imr	.0000507	.0000251	2.02	0.050	-1.09e-07	.0001015
logttime	-.011845	.0081103	-1.46	0.152	-.0282635	.0045734
logcellarea	.0375485	.0375027	1.00	0.323	-.0383717	.1134688
avg_logdist_LNC	.002358	.0079247	0.30	0.768	-.0136848	.0184008
mountain2000	.0040545	.0120805	0.34	0.739	-.0204013	.0285102
ycoord	-.0057331	.004492	-1.28	0.210	-.0148268	.0033606
avg_degtemper	.0012215	.0024801	0.49	0.625	-.0037992	.0062423
avg_prec	.000055	.0000295	1.86	0.070	-4.71e-06	.0001148
_cons	-.034783	.2408788	-0.14	0.886	-.5224165	.4528506

isocode	absorbed	(39 categories)
---------	----------	-----------------

```
.
.
.
. *** petroleum only ***
. **part 1 : full sample
. use newbygid_confdata, clear

.
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =     43,120
                                           F( 11,     61)    =      1.98
                                           Prob > F           =      0.0457
                                           R-squared          =      0.8321
                                           Adj R-squared      =      0.8318
                                           Root MSE          =      0.0781
```

(Std. Err. adjusted for 62 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0044393	.0031849	-1.39	0.168	-.0108078	.0019292
avg_loggcppc	-.0141145	.0078454	-1.80	0.077	-.0298024	.0015735
avg_logpop	.0020508	.0011859	1.73	0.089	-.0003205	.004422
imr	.0000383	.0000308	1.24	0.219	-.0000234	.0000999
logttime	-.0050104	.0034796	-1.44	0.155	-.0119682	.0019475
logcellarea	-.0088394	.0060432	-1.46	0.149	-.0209236	.0032447
avg_logdist_LNC	.002723	.0041826	0.65	0.517	-.0056405	.0110866

mountain2000	.0069215	.0068848	1.01	0.319	-.0068454	.0206884
ycoord	.0000917	.0003843	0.24	0.812	-.0006767	.0008601
avg_degtemper	-.0006499	.0004804	-1.35	0.181	-.0016105	.0003106
avg_prec	1.26e-06	.0000101	0.12	0.901	-.000019	.0000215
_cons	.2655539	.0891954	2.98	0.004	.0871967	.443911
isocode	absorbed	(62 categories)				

```
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)
```

(Std. Err. adjusted for **23** clusters in isocode)

```
. **part 3 : democracies
. use newbygid confdata, clear
```

```
. areg avg ConfIntra logdist petrol ${controls2}, a(isocode) cluster(iso)
```

(Std. Err. adjusted for **40** clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.00381	.0042358	-0.90	0.374	-.0123776	.0047576
avg_loggcppc	-.0172558	.0108943	-1.58	0.121	-.0392916	.0047799
avg_logpop	.0008997	.0012583	0.72	0.479	-.0016454	.0034448
imr	.0000903	.0000846	1.07	0.292	-.0000807	.0002613
logttime	-.0031277	.0033512	-0.93	0.356	-.0099062	.0036508
logcellarea	-.0155856	.008646	-1.80	0.079	-.0330739	.0019026
avg_logdist_LNC	.001839	.0042175	0.44	0.665	-.0066917	.0103698
mountain2000	.0071558	.0079219	0.90	0.372	-.0088678	.0231793
ycoord	-.0004496	.00067	-0.67	0.506	-.0018048	.0009057
avg_degtemper	-.0004627	.0006097	-0.76	0.452	-.0016959	.0007704
avg_prec	4.11e-06	.0000107	0.38	0.703	-.0000176	.0000258
_cons	.3423347	.1337098	2.56	0.014	.0718811	.6127882
isocode	absorbed		(40 categories)			

```
.
. **part 4 : extractive colony (weak political institution and ressource rich countries)
. use newbygid_confdata, clear
```

```
. keep if extractivecol == 1
(45,014 observations deleted)
```

```
.
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	9,786
	F(11 , 21)	=	7.92
	Prob > F	=	0.0000
	R-squared	=	0.8017
	Adj R-squared	=	0.8010
	Root MSE	=	0.1413

(Std. Err. adjusted for **22** clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0153936	.0107782	-1.43	0.168	-.037808	.0070208
avg_loggcppc	-.0241782	.0272904	-0.89	0.386	-.0809317	.0325753
avg_logpop	.0113138	.0066141	1.71	0.102	-.0024409	.0250684
imr	.0000825	.0000853	0.97	0.344	-.0000949	.0002599
logttime	-.0049833	.0079636	-0.63	0.538	-.0215446	.011578
logcellarea	.0478141	.0448861	1.07	0.299	-.0455317	.1411599
avg_logdist_LNC	-.0033333	.007587	-0.44	0.665	-.0191113	.0124447
mountain2000	.0468541	.0248552	1.89	0.073	-.0048351	.0985432
ycoord	-.0033326	.0034476	-0.97	0.345	-.0105023	.0038371
avg_degtemper	.0034243	.0018741	1.83	0.082	-.0004731	.0073218
avg_prec	.0000148	.0000235	0.63	0.535	-.000034	.0000636
_cons	-.0847657	.3519227	-0.24	0.812	-.816629	.6470976
isocode	absorbed		(22 categories)			

```
.
. **part 5 : Asia
. use newbygid_confdata, clear
```

```
. keep if Asia == 1
(51,846 observations deleted)
```

```
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      10,734
                                           F(   11,      25)   =        2.00
                                           Prob > F           =      0.0741
                                           R-squared          =      0.8029
                                           Adj R-squared      =      0.8023
                                           Root MSE           =      0.1351
```

(Std. Err. adjusted for 26 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0102996	.0077708	-1.34	0.194	-.0261744	.0055753
avg_loggcppc	-.0757095	.0547643	-1.38	0.179	-.1884986	.0370797
avg_logpop	.0064168	.0044383	1.45	0.161	-.002724	.0155576
imr	.0000503	.0000425	1.18	0.248	-.0000373	.0001379
logttime	-.0150178	.0086829	-1.73	0.096	-.0329005	.0028649
logcellarea	.0222271	.0152563	1.46	0.158	-.0091938	.053648
avg_logdist_LNC	.0110581	.0107848	1.03	0.315	-.0111535	.0332698
mountain2000	-.0233093	.0237122	-0.98	0.335	-.0721454	.0255268
ycoord	-.0020502	.0021851	-0.94	0.357	-.0065505	.0024501
avg_degtemper	-.0023612	.0013786	-1.71	0.099	-.0052004	.000478
avg_prec	3.05e-06	.0000249	0.12	0.904	-.0000483	.0000544
_cons	.6640803	.4292111	1.55	0.134	-.2198966	1.548057
isocode	absorbed		(26 categories)			

```
.
. **part 6 : Africa
.
. use newbygid_confdata, clear
```

```
. keep if Africa == 1
(54,641 observations deleted)
```

```
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =        2,519
                                           F(    6,      6)   =          .
                                           Prob > F           =          .
                                           R-squared          =      0.9346
                                           Adj R-squared      =      0.9342
                                           Root MSE           =      0.0508
```

(Std. Err. adjusted for 7 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0024788	.0088574	-0.28	0.789	-.0241521	.0191945
avg_loggcppc	.0024649	.0077388	0.32	0.761	-.0164712	.021401
avg_logpop	.0021199	.001588	1.34	0.230	-.0017657	.0060055
imr	2.39e-06	.0000184	0.13	0.901	-.0000427	.0000475
logttime	-.0119371	.0085762	-1.39	0.213	-.0329224	.0090482
logcellarea	.0141852	.0336594	0.42	0.688	-.0681763	.0965467
avg_logdist_LNC	-.002265	.0062687	-0.36	0.730	-.0176039	.013074
mountain2000	.0250597	.031367	0.80	0.455	-.0516925	.101812
ycoord	.0005312	.0026108	0.20	0.846	-.0058573	.0069196
avg_degtemper	.0000973	.0035195	0.03	0.979	-.0085147	.0087093
avg_prec	.0000405	.0000183	2.21	0.069	-4.31e-06	.0000853
_cons	.1192615	.2711179	0.44	0.675	-.5441401	.7826631
isocode	absorbed		(7 categories)			

```

.
.
.
. ***onsetdata
. **ores
. //part 1 : full sample
. use newbygid_onsetdata, clear

```

```

. areg avg_ConfIntra logdist_ore ${controls2} , a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      49,143
                                           F(   11,    126)    =       1.14
                                           Prob > F            =      0.3369
                                           R-squared           =      0.8513
                                           Adj R-squared       =      0.8509
                                           Root MSE            =      0.0543

```

(Std. Err. adjusted for 127 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0007712	.0030095	-0.26	0.798	-.0067268	.0051844
avg_loggcppc	-.0071673	.0039192	-1.83	0.070	-.0149233	.0005887
avg_logpop	.0010672	.0009256	1.15	0.251	-.0007646	.0028989
imr	-7.19e-06	.00002	-0.36	0.720	-.0000467	.0000324
logttime	-.0020735	.0018557	-1.12	0.266	-.0057458	.0015988
logcellarea	.0056548	.006348	0.89	0.375	-.0069077	.0182173
avg_logdist_LNC	-.0017547	.0038986	-0.45	0.653	-.00947	.0059606
mountain2000	.0059378	.0054939	1.08	0.282	-.0049343	.01681
ycoord	.0000984	.0004413	0.22	0.824	-.000775	.0009718
avg_degtemper	-.0008307	.0005434	-1.53	0.129	-.001906	.0002447
avg_prec	7.26e-06	6.24e-06	1.16	0.247	-5.09e-06	.0000196
_cons	.0937991	.0753076	1.25	0.215	-.0552324	.2428305
isocode	absorbed		(127 categories)			

```

.
. //part 2 : anocracies
. use newbygid_onsetdata, clear

```

```

. keep if avg_polity2<=0
(28,904 observations deleted)

```

```

. areg avg_ConfIntra logdist_ore ${controls2} , a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      30,006
                                           F(   11,    79)    =       1.87
                                           Prob > F            =      0.0562
                                           R-squared           =      0.8709
                                           Adj R-squared       =      0.8705
                                           Root MSE            =      0.0551

```

(Std. Err. adjusted for 80 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	.0028297	.0039897	0.71	0.480	-.0051115	.0107709
avg_loggcppc	-.0008387	.0037127	-0.23	0.822	-.0082287	.0065513
avg_logpop	.0025156	.002036	1.24	0.220	-.0015371	.0065682
imr	.0000199	.0000143	1.39	0.168	-8.57e-06	.0000484
logttime	-.0031031	.0028216	-1.10	0.275	-.0087193	.002513
logcellarea	.0051978	.0101084	0.51	0.609	-.0149224	.0253181
avg_logdist_LNC	-.0021441	.0054197	-0.40	0.693	-.0129317	.0086435
mountain2000	.0030592	.0085768	0.36	0.722	-.0140125	.0201309
ycoord	.0002937	.0009288	0.32	0.753	-.001555	.0021423
avg_degtemper	-.0007804	.0007945	-0.98	0.329	-.0023617	.000801
avg_prec	.0000135	.0000132	1.02	0.309	-.0000127	.0000398
_cons	.0063968	.1212687	0.05	0.958	-.2349824	.2477761

isocode	absorbed	(80 categories)
---------	----------	-----------------

```
.
. //part 3 : autocracies
. use newbygid_confdata, clear

. keep if avg_polity2>=0
(15,882 observations deleted)

.
. areg avg_ConfIntra logdist_ore ${controls2} , a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      36,762
                                           F( 11,      80)    =      4.78
                                           Prob > F           =      0.0000
                                           R-squared          =      0.8116
                                           Adj R-squared      =      0.8112
                                           Root MSE          =      0.0797
```

(Std. Err. adjusted for 81 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0066271	.0049954	-1.33	0.188	-.0165683	.003314
avg_loggcppc	-.0168167	.0102008	-1.65	0.103	-.0371169	.0034835
avg_logpop	.0008725	.0010738	0.81	0.419	-.0012644	.0030094
imr	.0000778	.0000553	1.41	0.163	-.0000322	.0001878
logtttime	-.0023381	.0027414	-0.85	0.396	-.0077938	.0031175
logcellarea	-.0176449	.0094818	-1.86	0.066	-.0365143	.0012244
avg_logdist_LNC	.0005968	.0039979	0.15	0.882	-.0073592	.0085529
mountain2000	.0048347	.0062877	0.77	0.444	-.0076782	.0173476
ycoord	-.0005166	.0005646	-0.92	0.363	-.0016402	.0006069
avg_degtemper	-.0004216	.0005713	-0.74	0.463	-.0015585	.0007153
avg_prec	5.90e-06	.0000101	0.58	0.562	-.0000143	.0000261
_cons	.3693068	.1472112	2.51	0.014	.0763471	.6622665

isocode	absorbed	(81 categories)
---------	----------	-----------------

```
.
. //part 4 : extractive colony (weak political institution and ressource rich countries)
. use newbygid_onsetdata, clear

. keep if extractivecol == 1
(45,116 observations deleted)

.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      15,206
                                           F( 11,      70)    =      1.74
                                           Prob > F           =      0.0815
                                           R-squared          =      0.8318
                                           Adj R-squared      =      0.8309
                                           Root MSE          =      0.0910
```

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0019388	.0115313	-0.17	0.867	-.0249373	.0210597
avg_loggcppc	-.0155071	.0094358	-1.64	0.105	-.0343263	.003312
avg_logpop	.0038985	.0030658	1.27	0.208	-.0022161	.0100131
imr	-.0000384	.0000433	-0.89	0.379	-.0001248	.000048
logtttime	-.0063695	.004644	-1.37	0.175	-.0156317	.0028926
logcellarea	.0381202	.0304518	1.25	0.215	-.0226139	.0988544
avg_logdist_LNC	.0018468	.0116434	0.16	0.874	-.0213752	.0250689
mountain2000	.0152699	.0210471	0.73	0.471	-.0267072	.0572469
ycoord	.0002997	.0014619	0.21	0.838	-.0026159	.0032153
avg_degtemper	-.0019214	.0015309	-1.26	0.214	-.0049747	.0011318

avg_prec	.0000127	9.49e-06	1.34	0.185	-6.23e-06	.0000316
_cons	-.0041351	.2515566	-0.02	0.987	-.5058489	.4975788

isocode	absorbed	(71 categories)
---------	----------	-----------------

```
.
. //part 5 : Asia
. use newbygid_onsetdata, clear
```

```
. keep if Asia == 1
(51,951 observations deleted)
```

```
.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	10,565
	F(11, 34)	=	10.46
	Prob > F	=	0.0000
	R-squared	=	0.7755
	Adj R-squared	=	0.7746
	Root MSE	=	0.0829

(Std. Err. adjusted for 35 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.014022	.0111231	-1.26	0.216	-.0366269	.0085829
avg_loggcppc	-.0065214	.0090621	-0.72	0.477	-.0249378	.011895
avg_logpop	.0035472	.0047267	0.75	0.458	-.0060586	.0131529
imr	-.0000239	.0000419	-0.57	0.571	-.000109	.0000611
logttime	-.0035075	.0036459	-0.96	0.343	-.0109169	.003902
logcellarea	.0176621	.018383	0.96	0.343	-.0196966	.0550209
avg_logdist_LNC	.0067881	.0161402	0.42	0.677	-.0260129	.039589
mountain2000	.0202273	.0151995	1.33	0.192	-.0106619	.0511164
ycoord	-.0006423	.0008157	-0.79	0.436	-.0023	.0010153
avg_degtemper	-.0018775	.0009706	-1.93	0.061	-.0038499	.0000949
avg_prec	.000016	.0000113	1.42	0.165	-6.90e-06	.0000389
_cons	.0497915	.1683121	0.30	0.769	-.2922599	.3918428

isocode	absorbed	(35 categories)
---------	----------	-----------------

```
.
. //part 6 : Africa
. use newbygid_onsetdata, clear
```

```
. keep if Africa == 1
(54,743 observations deleted)
```

```
.
. areg avg_ConfIntra logdist_ore ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	8,058
	F(11, 38)	=	2.70
	Prob > F	=	0.0112
	R-squared	=	0.8609
	Adj R-squared	=	0.8600
	Root MSE	=	0.0807

(Std. Err. adjusted for 39 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	.0216432	.0134477	1.61	0.116	-.0055803	.0488667
avg_loggcppc	-.0040438	.0063833	-0.63	0.530	-.0169661	.0088784
avg_logpop	.0022425	.003095	0.72	0.473	-.0040229	.0085079
imr	.000012	.0000136	0.88	0.382	-.0000155	.0000395
logttime	-.0150465	.0074764	-2.01	0.051	-.0301816	.0000886
logcellarea	.067077	.0368097	1.82	0.076	-.0074403	.1415944
avg_logdist_LNC	-.002043	.0102582	-0.20	0.843	-.0228095	.0187236
mountain2000	.0023892	.0305394	0.08	0.938	-.0594345	.0642129

ycoord	-.002653	.0039201	-0.68	0.503	-.0105888	.0052828
avg_degtemper	.0029493	.0013989	2.11	0.042	.0001173	.0057812
avg_prec	.0000148	.0000269	0.55	0.585	-.0000396	.0000692
_cons	-.4651848	.3030238	-1.54	0.133	-1.078624	.1482548
isocode	absorbed (39 categories)					

```
.
. **petrol
. //part 1 : full sample
. use newbygid_onsetdata, clear
```

```
.
. areg avg_ConfIntra logdist_petrol ${controls2} , a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      42,736
                                             F( 11,      62)    =      1.88
                                             Prob > F           =      0.0589
                                             R-squared         =      0.8676
                                             Adj R-squared     =      0.8674
                                             Root MSE         =      0.0492
```

(Std. Err. adjusted for 63 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0056701	.0034646	-1.64	0.107	-.0125957	.0012555
avg_loggcppc	-.0067846	.0042073	-1.61	0.112	-.015195	.0016257
avg_logpop	.0012874	.001011	1.27	0.208	-.0007335	.0033083
imr	-8.83e-06	.0000254	-0.35	0.729	-.0000595	.0000419
logttime	-.0008699	.0016904	-0.51	0.609	-.004249	.0025091
logcellarea	.005015	.0061971	0.81	0.421	-.0073728	.0174029
avg_logdist_LNC	-.0041158	.004349	-0.95	0.348	-.0128093	.0045778
mountain2000	.0106709	.0050916	2.10	0.040	.0004929	.0208488
ycoord	.0002665	.0003437	0.78	0.441	-.0004206	.0009536
avg_degtemper	-.0009644	.0005804	-1.66	0.102	-.0021245	.0001958
avg_prec	7.82e-06	6.18e-06	1.27	0.210	-4.53e-06	.0000202
_cons	.1176857	.0819634	1.44	0.156	-.0461569	.2815283
isocode	absorbed (63 categories)					

```
.
. //part 2 : anocracies
. use newbygid_onsetdata, clear
```

```
. keep if avg_polity2<=0
(28,904 observations deleted)
```

```
.
. areg avg_ConfIntra logdist_petrol ${controls2} , a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      25,067
                                             F( 11,      36)    =      9.72
                                             Prob > F           =      0.0000
                                             R-squared         =      0.9018
                                             Adj R-squared     =      0.9016
                                             Root MSE         =      0.0454
```

(Std. Err. adjusted for 37 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0033986	.0016917	-2.01	0.052	-.0068296	.0000324
avg_loggcppc	.0012884	.0039131	0.33	0.744	-.0066478	.0092246
avg_logpop	.0034474	.0023528	1.47	0.152	-.0013242	.008219
imr	.0000301	.0000193	1.56	0.127	-8.95e-06	.0000692
logttime	-.0013225	.0023224	-0.57	0.573	-.0060325	.0033875
logcellarea	.0055873	.008889	0.63	0.534	-.0124404	.023615
avg_logdist_LNC	-.0067528	.0066432	-1.02	0.316	-.0202258	.0067202
mountain2000	.0066565	.0031037	2.14	0.039	.0003619	.0129511

ycoord	.0005404	.0006897	0.78	0.438	-.0008584	.0019392
avg_degtemper	-.0009305	.0006305	-1.48	0.149	-.0022091	.0003482
avg_prec	.0000139	.0000135	1.03	0.309	-.0000134	.0000413
_cons	.007731	.1010391	0.08	0.939	-.1971859	.2126478
isocode	absorbed (37 categories)					

```
. //part 3 : autocracies
. use newbygid_confdata, clear

. keep if avg_polity2>=0
(15,882 observations deleted)
```

```
. areg avg ConfIntra logdist petrol ${controls2} , a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	33,168
	F(11, 39)	=	0.97
	Prob > F	=	0.4910
	R-squared	=	0.7918
	Adj R-squared	=	0.7915
	Root MSE	=	0.0810

(Std. Err. adjusted for **40** clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.00381	.0042358	-0.90	0.374	-.0123776	.0047576
avg_loggcppc	-.0172558	.0108943	-1.58	0.121	-.0392916	.0047799
avg_logpop	.0008997	.0012583	0.72	0.479	-.0016454	.0034448
imr	.0000903	.0000846	1.07	0.292	-.0000807	.0002613
logtttime	-.0031277	.0033512	-0.93	0.356	-.0099062	.0036508
logcellarea	-.0155856	.008646	-1.80	0.079	-.0330739	.0019026
avg_logdist_LNC	.001839	.0042175	0.44	0.665	-.0066917	.0103698
mountain2000	.0071558	.0079219	0.90	0.372	-.0088678	.0231793
ycoord	-.0004496	.00067	-0.67	0.506	-.0018048	.0009057
avg_degtemper	-.0004627	.0006097	-0.76	0.452	-.0016959	.0007704
avg_prec	4.11e-06	.0000107	0.38	0.703	-.0000176	.0000258
_cons	.3423347	.1337098	2.56	0.014	.0718811	.6127882
isocode	absorbed (40 categories)					

```
. //part 4 : extractive colony (weak political institution and ressource rich countries)
. use newbygid_onsetdata, clear

. keep if extractivecol == 1
(45,116 observations deleted)
```

```
. areg avg ConfIntra logdist petrol ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	9,425
	F(11, 21)	=	2.72
	Prob > F	=	0.0235
	R-squared	=	0.8621
	Adj R-squared	=	0.8616
	Root MSE	=	0.0923

(Std. Err. adjusted for 22 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0231295	.0149155	-1.55	0.136	-.054148	.007889
avg_loggcppc	-.0117924	.0089893	-1.31	0.204	-.0304866	.0069018
avg_logpop	.0048354	.004247	1.14	0.268	-.0039967	.0136676
imr	-.0000729	.0000702	-1.04	0.311	-.0002189	.0000732
logttime	-.0046558	.0045763	-1.02	0.321	-.0141727	.004861
logcellarea	.0304608	.0304631	1.00	0.329	-.0328907	.0938123
avg_logdist_LNC	-.001401	.0157895	-0.09	0.930	-.034237	.0314351
mountain2000	.0324699	.0245017	1.33	0.199	-.0184842	.083424
ycoord	.0013677	.0009509	1.44	0.165	-.0006098	.0033452
avg_degtemper	-.0026931	.0014609	-1.84	0.079	-.0057312	.000345
avg_prec	.0000122	9.10e-06	1.34	0.195	-6.74e-06	.0000311
_cons	.1846205	.2611923	0.71	0.487	-.3585586	.7277995
isocode	absorbed		(22 categories)			

```

.
. //part 5 : Asia
. use newbygid_onsetdata, clear

. keep if Asia == 1
(51,951 observations deleted)

```

```

.
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)

```

Linear regression, absorbing indicators	Number of obs	=	10,475
	F(11, 25)	=	24.52
	Prob > F	=	0.0000
	R-squared	=	0.7836
	Adj R-squared	=	0.7828
	Root MSE	=	0.0839

(Std. Err. adjusted for 26 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0201096	.0117071	-1.72	0.098	-.0442207	.0040016
avg_loggcppc	-.0043306	.0083878	-0.52	0.610	-.0216056	.0129444
avg_logpop	.0039937	.0052834	0.76	0.457	-.0068877	.0148751
imr	-.0000217	.0000411	-0.53	0.602	-.0001063	.0000629
logttime	-.0044153	.0036552	-1.21	0.238	-.0119433	.0031127
logcellarea	.0142728	.0174103	0.82	0.420	-.0215845	.05013
avg_logdist_LNC	.0043918	.0156713	0.28	0.782	-.0278838	.0366673
mountain2000	.0208282	.0159511	1.31	0.204	-.0120237	.0536801
ycoord	-.0011235	.0008661	-1.30	0.206	-.0029073	.0006603
avg_degtemper	-.00257	.0011085	-2.32	0.029	-.004853	-.0002869
avg_prec	.0000195	.0000118	1.66	0.110	-4.72e-06	.0000438
_cons	.140538	.1718286	0.82	0.421	-.2133496	.4944257
isocode	absorbed		(26 categories)			

```

.
. //part 6 : Africa
. use newbygid_onsetdata, clear

```

```
. keep if Africa == 1
(54,743 observations deleted)
```

```
.
. areg avg_ConfIntra logdist_petrol ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      2,383
                                           F(      6,      6)  =      .
                                           Prob > F            =      .
                                           R-squared           =      0.9487
                                           Adj R-squared       =      0.9484
                                           Root MSE            =      0.0612
```

(Std. Err. adjusted for 7 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_petrol	-.0110011	.012279	-0.90	0.405	-.0410467	.0190444
avg_loggcppc	.0013095	.0045346	0.29	0.782	-.0097862	.0124051
avg_logpop	.0037091	.0031642	1.17	0.286	-.0040335	.0114516
imr	.0000432	.0000168	2.57	0.042	2.09e-06	.0000843
logttime	-.0053376	.0063056	-0.85	0.430	-.020767	.0100917
logcellarea	.1008171	.0646462	1.56	0.170	-.0573665	.2590006
avg_logdist_LNC	-.0308333	.0103919	-2.97	0.025	-.0562613	-.0054053
mountain2000	.0431507	.0160812	2.68	0.036	.0038013	.0825001
ycoord	-.0041124	.0021477	-1.91	0.104	-.0093677	.0011428
avg_degtemper	.0018335	.0014956	1.23	0.266	-.001826	.0054931
avg_prec	.0001115	.0000634	1.76	0.129	-.0000436	.0002665
_cons	-.4788893	.565738	-0.85	0.430	-1.8632	.9054218

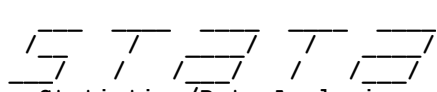
isocode absorbed

(7 categories)

```
.
end of do-file
```

```
.
```

Annex 4: Regression results on the natural resources distance

 **(R)**
Statistics/Data Analysis **14.0** Copyright 1985-2015 StataCorp LP
MP - Parallel Edition StataCorp
 4905 Lakeway Drive
 College Station, Texas 77845 USA
 800-STATATA-PC <http://www.stata.com>
 979-696-4600 stata@stata.com
 979-696-4601 (fax)

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Notes:

1. Unicode is supported; see [help unicode advice](#).
2. Maximum number of variables is set to 5000; see [help set maxvar](#).

```

. doedit "C:\Users\bozet\OneDrive\Bureau\data - Copie\conflict part\stata natural ressources.do"

. do "C:\Users\bozet\AppData\Local\Temp\STD000000000.tmp"

. global gdpcontrols "avg_loggcppc avg_logpop"

. global controls "avg_loggcppc avg_logpop imr logtttime logcellarea avg_logdist_LNC"

. global controls2 "${controls} mountain2000 ycoord avg_degtemper avg_prec"

.
end of do-file

. do "C:\Users\bozet\AppData\Local\Temp\STD000000000.tmp"

. ///Other conflicts for extractive colonies
> ///in Civconf
> use newbygid_Confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfGov logdist_ore ${controls2}, a(isocode) cluster(iso)

```

Linear regression, absorbing indicators	Number of obs	=	15,904
	F(11, 70)	=	2.14
	Prob > F	=	0.0279
	R-squared	=	0.7377
	Adj R-squared	=	0.7363
	Root MSE	=	0.1244

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfGov	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0135682	.0069723	-1.95	0.056	-.0274741	.0003376
avg_loggcppc	-.0298269	.023107	-1.29	0.201	-.0759123	.0162585
avg_logpop	.0114026	.0040621	2.81	0.006	.0033009	.0195043
imr	.0000921	.0000635	1.45	0.152	-.0000346	.0002188
logtttime	-.0040493	.0076301	-0.53	0.597	-.019267	.0111684
logcellarea	.0640771	.0452066	1.42	0.161	-.0260846	.1542388
avg_logdist_LNC	-.0052989	.0106504	-0.50	0.620	-.0265405	.0159428
mountain2000	.0025418	.0228713	0.11	0.912	-.0430734	.0481571
ycoord	-.0056504	.0039175	-1.44	0.154	-.0134636	.0021628
avg_degtemper	.0053384	.0030411	1.76	0.084	-.0007268	.0114036
avg_prec	2.60e-06	.0000153	0.17	0.865	-.0000278	.000033
_cons	-.3045286	.329466	-0.92	0.359	-.9616277	.3525706
isocode	absorbed		(71 categories)			


```

isocode | absorbed (71 categories)

.
.
. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfTerr logdist_ore ${controls2}, a(isocode) cluster(iso)

Linear regression, absorbing indicators      Number of obs      =      15,904
                                           F( 11,          70) =      0.81
                                           Prob > F           =      0.6258
                                           R-squared          =      0.7386
                                           Adj R-squared      =      0.7373
                                           Root MSE          =      0.0925

```

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfTerr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0084102	.0074555	-1.13	0.263	-.0232797	.0064594
avg_loggcppc	-.004037	.0035476	-1.14	0.259	-.0111124	.0030384
avg_logpop	-.0001932	.0022535	-0.09	0.932	-.0046876	.0043012
imr	-.0000102	.0000328	-0.31	0.757	-.0000757	.0000553
logttime	-.0020367	.0033191	-0.61	0.541	-.0086565	.0045831
logcellarea	-.0005913	.016407	-0.04	0.971	-.033314	.0321314
avg_logdist_LNC	.0104142	.0128135	0.81	0.419	-.0151416	.0359699
mountain2000	.0273042	.0198272	1.38	0.173	-.0122398	.0668482
ycoord	.0005493	.001473	0.37	0.710	-.0023884	.0034871
avg_degtemper	-.0018706	.0019029	-0.98	0.329	-.0056659	.0019246
avg_prec	.0000238	.0000188	1.26	0.211	-.0000138	.0000613
_cons	.0956355	.1693839	0.56	0.574	-.24219	.4334611
isocode	absorbed	(71 categories)				

```

.
.
. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfInter logdist_ore ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      15,904
                                           F( 11,          70) =      1.66
                                           Prob > F           =      0.1016
                                           R-squared          =      0.2127
                                           Adj R-squared      =      0.2087
                                           Root MSE          =      0.0296

```

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfInter	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	.0009887	.0021176	0.47	0.642	-.0032347	.0052121
avg_loggcppc	.0002686	.0016334	0.16	0.870	-.002989	.0035263
avg_logpop	.0003188	.0004826	0.66	0.511	-.0006437	.0012813
imr	-4.96e-07	1.53e-06	-0.32	0.746	-3.54e-06	2.55e-06
logttime	.0007681	.0017544	0.44	0.663	-.0027309	.0042671
logcellarea	-.0098616	.0063768	-1.55	0.126	-.0225797	.0028566
avg_logdist_LNC	.0008008	.0011027	0.73	0.470	-.0013985	.003
mountain2000	.0056594	.0056981	0.99	0.324	-.0057051	.0170239
ycoord	.0007718	.0004488	1.72	0.090	-.0001233	.0016668
avg_degtemper	-.0011185	.0007975	-1.40	0.165	-.002709	.000472
avg_prec	1.27e-07	1.67e-06	0.08	0.940	-3.21e-06	3.46e-06

_cons	.0851418	.0714083	1.19	0.237	-.0572775	.2275612
isocode	absorbed					(71 categories)

```
.
.
.
. ///in Onset data
> use newbygid_onsetdata, clear
```

```
. keep if extractivecol == 1
(45,116 observations deleted)
```

```
.
. areg avg_ConfGov logdist_ore ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      15,206
                                             F( 11,      70)    =      1.77
                                             Prob > F           =      0.0770
                                             R-squared          =      0.8666
                                             Adj R-squared      =      0.8659
                                             Root MSE          =      0.0744
```

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfGov	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0042143	.0063248	-0.67	0.507	-.0168287	.0084002
avg_loggcppc	-.0088651	.0078018	-1.14	0.260	-.0244252	.0066951
avg_logpop	.0058372	.0028391	2.06	0.044	.0001747	.0114997
imr	-.0000162	.0000157	-1.03	0.307	-.0000475	.0000152
logttime	-.0001558	.004463	-0.03	0.972	-.0090571	.0087455
logcellarea	.0422038	.0292049	1.45	0.153	-.0160436	.1004511
avg_logdist_LNC	-.0049029	.0068254	-0.72	0.475	-.0185157	.0087099
mountain2000	.0092065	.0133666	0.69	0.493	-.0174523	.0358653
ycoord	-.0009352	.0013367	-0.70	0.486	-.0036012	.0017308
avg_degtemper	.0009634	.0013142	0.73	0.466	-.0016576	.0035845
avg_prec	-4.12e-06	7.81e-06	-0.53	0.600	-.0000197	.0000115
_cons	-.1830962	.2664412	-0.69	0.494	-.7144964	.3483041
isocode	absorbed					(71 categories)

```
.
.
. use newbygid_onsetdata, clear

. keep if extractivecol == 1
(45,116 observations deleted)
```

```
.
. areg avg_ConfNonIntense logdist_ore ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      15,206
                                             F( 11,      70)    =      1.77
                                             Prob > F           =      0.0755
                                             R-squared          =      0.7959
                                             Adj R-squared      =      0.7948
                                             Root MSE          =      0.0711
```

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfNonIn~e	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0030287	.0080902	-0.37	0.709	-.0191641	.0131067
avg_loggcppc	-.0080985	.0068976	-1.17	0.244	-.0218554	.0056584
avg_logpop	.0030404	.0026541	1.15	0.256	-.002253	.0083338
imr	-.00002	.0000302	-0.66	0.509	-.0000802	.0000401
logttime	-.0064396	.0029319	-2.20	0.031	-.012287	-.0005921
logcellarea	.0346893	.0249609	1.39	0.169	-.0150936	.0844722
avg_logdist_LNC	.006809	.0104463	0.65	0.517	-.0140254	.0276434

mountain2000	.0014549	.0162375	0.09	0.929	-.0309298	.0338395
ycoord	.0003816	.0010116	0.38	0.707	-.001636	.0023991
avg_degtemper	-.0011283	.0013002	-0.87	0.388	-.0037215	.0014648
avg_prec	.0000104	6.64e-06	1.57	0.122	-2.85e-06	.0000236
_cons	-.1242367	.2210771	-0.56	0.576	-.565161	.3166876

isocode	absorbed	(71 categories)
---------	----------	-----------------

```
.
.
. use newbygid_onsetdata, clear
```

```
. keep if extractivecol == 1
(45,116 observations deleted)
```

```
. areg avg_ConfIntense logdist ore ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	15,206
	F(11, 70)	=	1.28
	Prob > F	=	0.2551
	R-squared	=	0.8520
	Adj R-squared	=	0.8512
	Root MSE	=	0.0502

(Std. Err. adjusted for **71** clusters in isocode)

avg_ConfIntense	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	.003922	.0057351	0.68	0.496	-.0075163	.0153603
avg_loggcppc	-.0049072	.0058911	-0.83	0.408	-.0166566	.0068421
avg_logpop	.0022438	.0014268	1.57	0.120	-.0006019	.0050894
imr	-.0000168	.0000163	-1.03	0.306	-.0000492	.0000156
logtttime	-.0014018	.0038561	-0.36	0.717	-.0090925	.0062889
logcellarea	.0162711	.0172811	0.94	0.350	-.0181949	.0507371
avg_logdist_LNC	-.0018891	.0036044	-0.52	0.602	-.0090779	.0052997
mountain2000	.0141407	.0134804	1.05	0.298	-.0127452	.0410266
ycoord	-.0001835	.0010271	-0.18	0.859	-.0022321	.001865
avg_degtemper	-.0007408	.0006796	-1.09	0.279	-.0020962	.0006145
avg_prec	3.60e-06	6.49e-06	0.55	0.582	-9.36e-06	.0000165
_cons	-.0373712	.1341752	-0.28	0.781	-.3049752	.2302328

isocode	absorbed	(71 categories)
---------	----------	-----------------

```
.
.
. use newbygid onsetdata, clear
```

```
. keep if extractivecol == 1
(45,116 observations deleted)
```

```
. areg avg ConfTerr logdist ore ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	15,206
	F(11, 70)	=	1.48
	Prob > F	=	0.1593
	R-squared	=	0.6344
	Adj R-squared	=	0.6324
	Root MSE	=	0.0860

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfTerr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	.0075639	.0089744	0.84	0.402	-.0103351	.0254629
avg_loggcppc	-.0021505	.0060901	-0.35	0.725	-.0142968	.0099957
avg_logpop	.000904	.0024553	0.37	0.714	-.0039929	.0058009
imr	-.0000212	.0000306	-0.69	0.492	-.0000823	.0000399
logttime	-.0099841	.0044254	-2.26	0.027	-.0188103	-.0011579
logcellarea	.0251271	.0273463	0.92	0.361	-.0294134	.0796677
avg_logdist_LNC	.0150535	.0107818	1.40	0.167	-.0064502	.0365572
mountain2000	.0076748	.0200797	0.38	0.703	-.0323728	.0477225
ycoord	.0009753	.0016672	0.58	0.560	-.0023499	.0043004
avg_degtemper	-.0029781	.0017777	-1.68	0.098	-.0065236	.0005674
avg_prec	.0000192	9.03e-06	2.12	0.037	1.16e-06	.0000372
_cons	-.1611972	.2504689	-0.64	0.522	-.6607417	.3383474
isocode	absorbed		(71 categories)			

```
.  
.   
. use newbygid_onsetdata, clear  
  
. keep if extractivecol == 1  
(45,116 observations deleted)  
  
.   
. areg avg_ConfInter logdist_ore ${controls2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	15,206
	F(11, 70)	=	2.50
	Prob > F	=	0.0104
	R-squared	=	0.4487
	Adj R-squared	=	0.4458
	Root MSE	=	0.0325

(Std. Err. adjusted for 71 clusters in isocode)

avg_ConfInter	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_ore	-.0031157	.0017282	-1.80	0.076	-.0065625	.0003311
avg_loggcppc	.003262	.0018568	1.76	0.083	-.0004412	.0069652
avg_logpop	.0016851	.0010514	1.60	0.113	-.0004117	.003782
imr	1.93e-06	4.70e-06	0.41	0.682	-7.44e-06	.0000113
logttime	-.0000907	.0019569	-0.05	0.963	-.0039936	.0038122
logcellarea	-.0031333	.0052229	-0.60	0.550	-.0135502	.0072835
avg_logdist_LNC	.0057378	.0039149	1.47	0.147	-.0020701	.0135458
mountain2000	-.0026888	.0036662	-0.73	0.466	-.0100007	.0046231
ycoord	.0009707	.0005452	1.78	0.079	-.0001167	.0020582
avg_degtemper	-.0016477	.0009512	-1.73	0.088	-.0035448	.0002494
avg_prec	1.63e-06	2.62e-06	0.62	0.536	-3.60e-06	6.86e-06
_cons	.005608	.0801919	0.07	0.944	-.1543297	.1655457
isocode	absorbed		(71 categories)			

```
.  
.   
.   
.   
. 
```

```

. ***GCP PPP and GCP PC***
. *** GCP PPP ***
. // Civconf
. **part 1 : full sample
. use newbygid_confdata, clear

```

```

.
. areg avg_ConfIntra logdist_gcp_ppp ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      53,996
                                             F(   11,    154)    =       1.40
                                             Prob > F            =      0.1758
                                             R-squared           =      0.8314
                                             Adj R-squared       =      0.8309
                                             Root MSE            =      0.0808

```

(Std. Err. adjusted for 155 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0147698	.0075635	-1.95	0.053	-.0297114	.0001719
avg_loggcppc	-.015992	.0081352	-1.97	0.051	-.0320631	.000079
avg_logpop	.0018486	.001009	1.83	0.069	-.0001448	.0038419
imr	.0000467	.0000267	1.75	0.082	-5.97e-06	.0000993
logtttime	-.004166	.0035392	-1.18	0.241	-.0111578	.0028257
logcellarea	-.0089406	.0063649	-1.40	0.162	-.0215143	.0036331
avg_logdist_LNC	.0116982	.0055648	2.10	0.037	.000705	.0226915
mountain2000	.0044105	.0067643	0.65	0.515	-.0089524	.0177733
ycoord	-.0003332	.0004616	-0.72	0.472	-.0012451	.0005787
avg_degtemper	-.0008935	.0005715	-1.56	0.120	-.0020226	.0002355
avg_prec	5.23e-06	.0000103	0.51	0.612	-.0000151	.0000256
_cons	.3068341	.0989264	3.10	0.002	.1114061	.5022621
isocode	absorbed		(155 categories)			

```

.
. **part 2 : anocracies
. use newbygid_confdata, clear

```

```

. keep if avg_polity2<=0
(48,680 observations deleted)

```

```

.
. areg avg_ConfIntra logdist_gcp_ppp ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      14,655
                                             F(   11,     54)    =       2.40
                                             Prob > F            =      0.0165
                                             R-squared           =      0.8696
                                             Adj R-squared       =      0.8690
                                             Root MSE            =      0.0869

```

(Std. Err. adjusted for 55 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0291939	.0122488	-2.38	0.021	-.0537512	-.0046366
avg_loggcppc	-.0203418	.0193402	-1.05	0.298	-.0591164	.0184329
avg_logpop	.005405	.0030831	1.75	0.085	-.0007762	.0115862
imr	.0000365	.0000234	1.56	0.125	-.0000104	.0000835
logtttime	-.0139736	.0070016	-2.00	0.051	-.0280111	.0000638
logcellarea	.0084675	.0112858	0.75	0.456	-.0141592	.0310941
avg_logdist_LNC	.0142698	.0084978	1.68	0.099	-.0027673	.031307
mountain2000	-.0175426	.0188744	-0.93	0.357	-.0553835	.0202982
ycoord	-.0034435	.0030074	-1.15	0.257	-.0094729	.002586
avg_degtemper	-.0036094	.0018778	-1.92	0.060	-.0073742	.0001554
avg_prec	-.0000195	.0000347	-0.56	0.576	-.0000891	.00005
_cons	.4872371	.2766503	1.76	0.084	-.0674133	1.041888
isocode	absorbed		(55 categories)			

```

.
. **part 3 : democracies
. use newbygid_confdata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

```

```

.
. areg avg_ConfIntra logdist_gcp_ppp ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      39,341
                                           F(   11,      100)  =       1.51
                                           Prob > F           =      0.1400
                                           R-squared          =      0.7969
                                           Adj R-squared      =      0.7963
                                           Root MSE           =      0.0775

```

(Std. Err. adjusted for 101 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0083925	.0096387	-0.87	0.386	-.0275154	.0107303
avg_loggcppc	-.0154808	.0100442	-1.54	0.126	-.0354083	.0044467
avg_logpop	.0008349	.0010228	0.82	0.416	-.0011943	.0028641
imr	.0000677	.0000588	1.15	0.252	-.0000489	.0001843
logttime	-.0024613	.003659	-0.67	0.503	-.0097207	.004798
logcellarea	-.0114641	.0066529	-1.72	0.088	-.0246633	.0017351
avg_logdist_LNC	.0079647	.0066257	1.20	0.232	-.0051804	.0211099
mountain2000	.0069462	.0077069	0.90	0.370	-.008344	.0222364
ycoord	-.0003855	.0005609	-0.69	0.493	-.0014984	.0007273
avg_degtemper	-.0002809	.0004853	-0.58	0.564	-.0012438	.000682
avg_prec	2.85e-06	.0000102	0.28	0.780	-.0000174	.0000231
_cons	.2795237	.106623	2.62	0.010	.0679867	.4910607
isocode	absorbed		(101 categories)			

```

.
. **part 4 : extractive colony (weak political institution and ressource rich countries)
. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

```

```

.
. areg avg_ConfIntra logdist_gcp_ppp ${controls2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      17,194
                                           F(   11,      76)  =       1.66
                                           Prob > F           =      0.0999
                                           R-squared          =      0.8136
                                           Adj R-squared      =      0.8127
                                           Root MSE           =      0.1257

```

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0345092	.0161084	-2.14	0.035	-.066592	-.0024265
avg_loggcppc	-.036798	.0244641	-1.50	0.137	-.0855224	.0119264
avg_logpop	.0079319	.0038021	2.09	0.040	.0003594	.0155045
imr	.0000886	.0000408	2.17	0.033	7.39e-06	.0001698
logttime	-.0044496	.0072672	-0.61	0.542	-.0189235	.0100243
logcellarea	.059164	.0342571	1.73	0.088	-.0090649	.1273929
avg_logdist_LNC	.0137322	.0093153	1.47	0.145	-.0048209	.0322854
mountain2000	.0316411	.0153518	2.06	0.043	.0010653	.0622169
ycoord	-.0042926	.0027366	-1.57	0.121	-.009743	.0011579
avg_degtemper	.00284	.0013851	2.05	0.044	.0000813	.0055986
avg_prec	.0000194	.000019	1.02	0.312	-.0000186	.0000573
_cons	-.0653709	.2790068	-0.23	0.815	-.6210612	.4903194

isocode	absorbed	(77 categories)
---------	----------	-----------------

```
.
. **part 5 : Asia
. use newbygid_confddata, clear

. keep if Asia == 1
(51,846 observations deleted)

.
. areg avg_ConfIntra logdist_gcp_ppp ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      11,852
                                           F(   11,        37) =        1.52
                                           Prob > F           =        0.1660
                                           R-squared          =        0.8053
                                           Adj R-squared      =        0.8045
                                           Root MSE          =        0.1326
```

(Std. Err. adjusted for 38 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0115841	.014115	-0.82	0.417	-.0401837	.0170156
avg_loggcppc	-.0826004	.0582174	-1.42	0.164	-.20056	.0353593
avg_logpop	.0056024	.0037646	1.49	0.145	-.0020254	.0132302
imr	.0000401	.0000366	1.10	0.281	-.0000341	.0001142
logtttime	-.0143816	.008368	-1.72	0.094	-.0313368	.0025735
logcellarea	.0260765	.0157156	1.66	0.106	-.0057664	.0579194
avg_logdist_LNC	.0187671	.0106115	1.77	0.085	-.0027338	.0402681
mountain2000	-.0192253	.024361	-0.79	0.435	-.0685853	.0301347
ycoord	-.0014124	.0020009	-0.71	0.485	-.0054665	.0026417
avg_degtemper	-.0018386	.0012416	-1.48	0.147	-.0043543	.0006772
avg_prec	7.48e-07	.0000268	0.03	0.978	-.0000535	.000055
_cons	.6360791	.4105911	1.55	0.130	-.1958575	1.468016

isocode	absorbed	(38 categories)
---------	----------	-----------------

```
.
. **part 6 : Africa
. use newbygid_confddata, clear

. keep if Africa == 1
(54,641 observations deleted)

.
. areg avg_ConfIntra logdist_gcp_ppp ${controls2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =        8,910
                                           F(   11,        47) =        1.43
                                           Prob > F           =        0.1919
                                           R-squared          =        0.8535
                                           Adj R-squared      =        0.8526
                                           Root MSE          =        0.0883
```

(Std. Err. adjusted for 48 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0141715	.0100383	-1.41	0.165	-.0343661	.006023
avg_loggcppc	-.0177924	.018489	-0.96	0.341	-.0549875	.0194027
avg_logpop	.0025611	.0028371	0.90	0.371	-.0031464	.0082687
imr	.0000498	.0000259	1.92	0.061	-2.34e-06	.0001019
logtttime	-.0114345	.0078105	-1.46	0.150	-.0271472	.0042781
logcellarea	.0424139	.0354258	1.20	0.237	-.0288535	.1136814
avg_logdist_LNC	.0055692	.0064105	0.87	0.389	-.0073271	.0184655
mountain2000	.0064349	.0121854	0.53	0.600	-.0180791	.0309488
ycoord	-.0057947	.0044556	-1.30	0.200	-.0147582	.0031688
avg_degtemper	.0012083	.0023911	0.51	0.616	-.0036019	.0060185

avg_prec	.0000502	.0000282	1.78	0.082	-6.53e-06	.000107
_cons	-.0446796	.2224091	-0.20	0.842	-.492109	.4027498
isocode	absorbed (48 categories)					

Linear regression, absorbing indicators	Number of obs	=	53,996
	F(12, 154)	=	2.52
	Prob > F	=	0.0047
	R-squared	=	0.8320
	Adj R-squared	=	0.8314
	Root MSE	=	0.0807

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0245258	.0134967	-1.82	0.071	-.0511883	.0021367
avg_logcapdist	.0129935	.0153915	0.84	0.400	-.0174123	.0433993
avg_loggcppc	-.0186652	.0101504	-1.84	0.068	-.0387171	.0013867
avg_logpop	.002191	.0010816	2.03	0.045	.0000542	.0043277
imr	.0000482	.0000291	1.65	0.100	-9.35e-06	.0001058
logtttime	-.0045693	.0041272	-1.11	0.270	-.0127226	.0035839
logcellarea	-.0073693	.0065718	-1.12	0.264	-.0203518	.0056133
avg_logdist_LNC	.0101797	.0061526	1.65	0.100	-.0019748	.0223341
mountain2000	.0032521	.0075095	0.43	0.666	-.0115829	.0180872
ycoord	-.0002422	.000442	-0.55	0.585	-.0011153	.0006309
avg_degtemper	-.0009323	.0005852	-1.59	0.113	-.0020884	.0002239
avg_prec	4.26e-06	9.65e-06	0.44	0.660	-.0000148	.0000233
_cons	.3050896	.0942552	3.24	0.001	.1188896	.4912896
isocode	absorbed (155 categories)					

Linear regression, absorbing indicators	Number of obs	=	14,655
	F(12, 54)	=	2.93
	Prob > F	=	0.0034
	R-squared	=	0.8705
	Adj R-squared	=	0.8699
	Root MSE	=	0.0866

(Std. Err. adjusted for 55 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0143665	.0109142	-1.32	0.194	-.0362482	.0075152
avg_logcapdist	-.0202171	.0102807	-1.97	0.054	-.0408286	.0003945
avg_loggcppc	-.0207377	.0196953	-1.05	0.297	-.0602244	.0187489
avg_logpop	.0043681	.0027971	1.56	0.124	-.0012397	.009976
imr	.0000368	.000024	1.53	0.131	-.0000113	.0000848
logtttime	-.0122578	.0066518	-1.84	0.071	-.025594	.0010783
logcellarea	.0083087	.0116267	0.71	0.478	-.0150014	.0316189
avg_logdist_LNC	.0129926	.0082686	1.57	0.122	-.003585	.0295701
mountain2000	-.0166124	.0185404	-0.90	0.374	-.0537837	.0205589
ycoord	-.0038815	.0031411	-1.24	0.222	-.0101789	.002416
avg_degtemper	-.0032897	.0017777	-1.85	0.070	-.0068537	.0002743
avg_prec	-.0000158	.000035	-0.45	0.653	-.000086	.0000544
_cons	.5352943	.2944003	1.82	0.075	-.0549428	1.125531
isocode	absorbed		(55 categories)			

```
.
.
. **part 3 : democracies
. use newbygid_confdata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

.
. areg avg_ConfIntra logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	39,341
	F(12, 100)	=	1.34
	Prob > F	=	0.2066
	R-squared	=	0.8001
	Adj R-squared	=	0.7996
	Root MSE	=	0.0769

(Std. Err. adjusted for 101 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0347695	.018707	-1.86	0.066	-.0718836	.0023447
avg_logcapdist	.0304549	.0224538	1.36	0.178	-.0140928	.0750027
avg_loggcppc	-.019101	.0117929	-1.62	0.108	-.0424977	.0042957
avg_logpop	.0014727	.0011274	1.31	0.194	-.0007641	.0037095
imr	.000071	.0000623	1.14	0.257	-.0000525	.0001946
logtttime	-.0032231	.0042886	-0.75	0.454	-.0117315	.0052854
logcellarea	-.0083322	.0058304	-1.43	0.156	-.0198996	.0032352
avg_logdist_LNC	.0068644	.0070815	0.97	0.335	-.0071851	.0209139
mountain2000	.005173	.008332	0.62	0.536	-.0113574	.0217035
ycoord	-.0001835	.0004125	-0.44	0.657	-.001002	.0006349
avg_degtemper	-.0003733	.0005339	-0.70	0.486	-.0014324	.0006859
avg_prec	3.41e-06	.0000103	0.33	0.740	-.0000169	.0000238
_cons	.2631437	.0903434	2.91	0.004	.083905	.4423824
isocode	absorbed		(101 categories)			

```

.
.
. **part 4 : extractive colony (weak political institution and ressource rich countries)
. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

```

```

. areg avg_ConfIntra logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      17,194
                                           F( 12,      76)    =      1.74
                                           Prob > F           =      0.0738
                                           R-squared          =      0.8143
                                           Adj R-squared      =      0.8133
                                           Root MSE          =      0.1254

```

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0486924	.0181647	-2.68	0.009	-.0848705	-.0125143
avg_logcapdist	.0186178	.0224011	0.83	0.409	-.0259978	.0632335
avg_loggcppc	-.0367326	.0241778	-1.52	0.133	-.0848868	.0114216
avg_logpop	.0086326	.0043256	2.00	0.050	.0000175	.0172477
imr	.0000882	.0000414	2.13	0.036	5.69e-06	.0001708
logtttime	-.0060196	.0086356	-0.70	0.488	-.0232189	.0111796
logcellarea	.0588961	.033374	1.76	0.082	-.0075739	.1253662
avg_logdist_LNC	.016043	.0094979	1.69	0.095	-.0028736	.0349597
mountain2000	.0299175	.0150005	1.99	0.050	.0000413	.0597937
ycoord	-.0037873	.0023293	-1.63	0.108	-.0084264	.0008519
avg_degtemper	.0024626	.0012327	2.00	0.049	7.49e-06	.0049176
avg_prec	.0000186	.0000179	1.04	0.303	-.0000171	.0000543
_cons	-.0943667	.2921073	-0.32	0.748	-.6761488	.4874153
isocode	absorbed		(77 categories)			

```

.
. **part 5 : Asia
. use newbygid_confdata, clear

```

```

. keep if Asia == 1
(51,846 observations deleted)

```

```

. areg avg_ConfIntra logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      11,852
                                           F( 12,      37)    =      3.12
                                           Prob > F           =      0.0038
                                           R-squared          =      0.8101
                                           Adj R-squared      =      0.8093
                                           Root MSE          =      0.1309

```

(Std. Err. adjusted for 38 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0464642	.0238071	-1.95	0.059	-.0947019	.0017735
avg_logcapdist	.0490849	.035394	1.39	0.174	-.0226302	.1208
avg_loggcppc	-.0844811	.0587302	-1.44	0.159	-.2034798	.0345176
avg_logpop	.0093528	.0062579	1.49	0.144	-.0033269	.0220326
imr	.0000506	.0000462	1.09	0.281	-.000043	.0001442
logtttime	-.0174746	.0095707	-1.83	0.076	-.0368667	.0019175
logcellarea	.0280326	.0169296	1.66	0.106	-.0062699	.0623352
avg_logdist_LNC	.025175	.0106618	2.36	0.024	.0035721	.0467779
mountain2000	-.0169497	.0210558	-0.80	0.426	-.0596129	.0257134
ycoord	.0012244	.0016498	0.74	0.463	-.0021184	.0045672
avg_degtemper	-.002074	.0011979	-1.73	0.092	-.0045011	.0003532

avg_prec	-2.91e-06	.0000245	-0.12	0.906	-.0000525	.0000467
_cons	.3974721	.2606147	1.53	0.136	-.1305834	.9255277

isocode	absorbed	(38 categories)
---------	----------	-----------------

```
.
. **part 6 : Africa
. use newbygid_confdata, clear

. keep if Africa == 1
(54,641 observations deleted)

.
. areg avg_ConfIntra logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	8,910
	F(12, 47)	=	1.32
	Prob > F	=	0.2369
	R-squared	=	0.8535
	Adj R-squared	=	0.8526
	Root MSE	=	0.0883

(Std. Err. adjusted for 48 clusters in isocode)

avg_ConfIntra	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0153681	.0124016	-1.24	0.221	-.0403169	.0095808
avg_logcapdist	.0014699	.0120512	0.12	0.903	-.0227741	.0257139
avg_loggcppc	-.017764	.0183593	-0.97	0.338	-.0546981	.0191702
avg_logpop	.0026299	.0032077	0.82	0.416	-.0038231	.0090829
imr	.0000495	.0000272	1.82	0.075	-5.25e-06	.0001042
logttime	-.0114974	.0080387	-1.43	0.159	-.0276692	.0046744
logcellarea	.0427988	.0371407	1.15	0.255	-.0319187	.1175163
avg_logdist_LNC	.0055911	.0065001	0.86	0.394	-.0074855	.0186677
mountain2000	.0064645	.0120646	0.54	0.595	-.0178063	.0307353
ycoord	-.0057857	.0044156	-1.31	0.196	-.0146688	.0030975
avg_degtemper	.0012105	.0024103	0.50	0.618	-.0036385	.0060594
avg_prec	.0000502	.0000281	1.79	0.080	-6.30e-06	.0001067
_cons	-.0498003	.2505864	-0.20	0.843	-.553915	.4543145

isocode	absorbed	(48 categories)
---------	----------	-----------------

```
.
.
.
. //conflict type for Extractive colonies
. use newbygid_Confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfGov logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	17,194
	F(12, 76)	=	1.17
	Prob > F	=	0.3170
	R-squared	=	0.7524
	Adj R-squared	=	0.7511
	Root MSE	=	0.1242

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfGov	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0579887	.0298548	-1.94	0.056	-.1174498	.0014723
avg_logcapdist	.0240099	.0271709	0.88	0.380	-.0301056	.0781255
avg_loggcppc	-.0329883	.0242843	-1.36	0.178	-.0813547	.0153781
avg_logpop	.0095284	.0042131	2.26	0.027	.0011371	.0179196
imr	.000107	.0000677	1.58	0.118	-.0000277	.0002418
logttime	-.0024394	.0082083	-0.30	0.767	-.0187876	.0139087
logcellarea	.0679954	.0448811	1.52	0.134	-.021393	.1573838
avg_logdist_LNC	.0108654	.0095713	1.14	0.260	-.0081974	.0299282
mountain2000	.0003163	.0285935	0.01	0.991	-.0566326	.0572651
ycoord	-.0043742	.0032516	-1.35	0.183	-.0108503	.0021018
avg_degtemper	.0042604	.0023337	1.83	0.072	-.0003876	.0089085
avg_prec	-9.13e-06	.0000165	-0.55	0.582	-.000042	.0000238
_cons	-.2298453	.338787	-0.68	0.500	-.9045981	.4449075
isocode	absorbed		(77 categories)			

```
.
.
. use newbygid_confddata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfNonIntense logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	17,194
	F(12, 76)	=	2.05
	Prob > F	=	0.0310
	R-squared	=	0.7918
	Adj R-squared	=	0.7908
	Root MSE	=	0.1166

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfNonIn~e	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0549529	.0214513	-2.56	0.012	-.0976768	-.0122289
avg_logcapdist	.0343279	.0289612	1.19	0.240	-.0233533	.0920091
avg_loggcppc	-.0264003	.0222995	-1.18	0.240	-.0708136	.018013
avg_logpop	.0062466	.0037211	1.68	0.097	-.0011647	.0136579
imr	.0000795	.0000481	1.65	0.102	-.0000163	.0001753
logttime	-.0073392	.0093241	-0.79	0.434	-.0259098	.0112314
logcellarea	.0561369	.0376939	1.49	0.141	-.0189371	.1312108
avg_logdist_LNC	.0134695	.0089619	1.50	0.137	-.0043797	.0313188
mountain2000	.021702	.0127236	1.71	0.092	-.0036392	.0470432
ycoord	-.0027589	.0020851	-1.32	0.190	-.0069118	.0013939
avg_degtemper	.0025901	.0014755	1.76	0.083	-.0003485	.0055287
avg_prec	.0000114	.0000174	0.65	0.515	-.0000233	.000046
_cons	-.1979534	.3158704	-0.63	0.533	-.8270637	.431157
isocode	absorbed		(77 categories)			

.

```

. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfIntense logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      17,194
                                             F(   12,         76) =        1.28
                                             Prob > F           =        0.2480
                                             R-squared          =        0.7838
                                             Adj R-squared      =        0.7827
                                             Root MSE          =        0.0521

```

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfIntense	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0036187	.006631	0.55	0.587	-.009588	.0168254
avg_logcapdist	-.0156824	.0080308	-1.95	0.055	-.0316771	.0003122
avg_loggcppc	-.0093795	.0121276	-0.77	0.442	-.0335337	.0147747
avg_logpop	.0026729	.0019577	1.37	0.176	-.0012263	.0065721
imr	.0000128	.0000133	0.96	0.339	-.0000137	.0000393
logtttime	.001227	.0022741	0.54	0.591	-.0033022	.0057562
logcellarea	.0031559	.0143924	0.22	0.827	-.0255091	.031821
avg_logdist_LNC	.0043946	.0031978	1.37	0.173	-.0019744	.0107637
mountain2000	.0117312	.0061395	1.91	0.060	-.0004967	.0239591
ycoord	-.0009149	.0015859	-0.58	0.566	-.0040736	.0022437
avg_degtemper	-.000072	.0007335	-0.10	0.922	-.0015329	.0013888
avg_prec	8.37e-06	7.62e-06	1.10	0.276	-6.81e-06	.0000236
_cons	.0919113	.0843975	1.09	0.280	-.076181	.2600036
isocode	absorbed		(77 categories)			

```

.
.
. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

.
. areg avg_ConfTerr logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      17,194
                                             F(   12,         76) =        0.98
                                             Prob > F           =        0.4729
                                             R-squared          =        0.7212
                                             Adj R-squared      =        0.7198
                                             Root MSE          =        0.0962

```

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfTerr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0085342	.0175959	0.49	0.629	-.0265111	.0435795
avg_logcapdist	-.00416	.0115507	-0.36	0.720	-.0271653	.0188452
avg_loggcppc	-.0022808	.003554	-0.64	0.523	-.0093593	.0047976
avg_logpop	-.0000568	.0021477	-0.03	0.979	-.0043342	.0042206
imr	-.0000137	.0000357	-0.38	0.703	-.0000848	.0000575
logtttime	-.005565	.0056272	-0.99	0.326	-.0167726	.0056425
logcellarea	-.0008576	.0166762	-0.05	0.959	-.0340712	.0323559
avg_logdist_LNC	.0083494	.007168	1.16	0.248	-.0059269	.0226257
mountain2000	.0351616	.0240125	1.46	0.147	-.0126635	.0829866
ycoord	.0005724	.0013965	0.41	0.683	-.0022091	.0033538
avg_degtemper	-.0016577	.0018226	-0.91	0.366	-.0052877	.0019724
avg_prec	.0000315	.0000253	1.24	0.217	-.0000189	.0000819
_cons	.034109	.1371949	0.25	0.804	-.2391384	.3073564

```

isocode | absorbed (77 categories)

.
.
. use newbygid_confdata, clear

. keep if extractivecol == 1
(45,014 observations deleted)

```

```

. areg avg_ConfInter logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

Linear regression, absorbing indicators      Number of obs   =    17,194
F(   12,      76)   =      1.58
Prob > F           =      0.1148
R-squared          =      0.2145
Adj R-squared      =      0.2105
Root MSE          =      0.0304

```

(Std. Err. adjusted for 77 clusters in isocode)

avg_ConfInter	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0064169	.0036819	1.74	0.085	-.0009162	.0137499
avg_logcapdist	-.0074709	.0040656	-1.84	0.070	-.0155683	.0006266
avg_loggcppc	.0009497	.0012733	0.75	0.458	-.0015864	.0034857
avg_logpop	.000246	.0005091	0.48	0.630	-.0007679	.0012599
imr	-2.19e-06	3.15e-06	-0.69	0.489	-8.47e-06	4.09e-06
logtttime	.0018035	.0019138	0.94	0.349	-.0020082	.0056151
logcellarea	-.0107909	.0064948	-1.66	0.101	-.0237264	.0021447
avg_logdist_LNC	-.0001254	.0006729	-0.19	0.853	-.0014655	.0012147
mountain2000	.0068864	.0049436	1.39	0.168	-.0029597	.0167325
ycoord	.0005945	.0003981	1.49	0.139	-.0001984	.0013875
avg_degtemper	-.0008981	.0006077	-1.48	0.144	-.0021086	.0003123
avg_prec	1.07e-06	1.29e-06	0.83	0.407	-1.49e-06	3.64e-06
_cons	.0944143	.0682274	1.38	0.170	-.0414724	.230301

isocode | absorbed (77 categories)

```

.
.
.
. //conflict type for Asia
. use newbygid_Confdata, clear

. keep if Asia == 1
(51,846 observations deleted)

.
. areg avg_ConfGov logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs   =    11,852
F(   12,      37)   =      11.76
Prob > F           =      0.0000
R-squared          =      0.7614
Adj R-squared      =      0.7604
Root MSE          =      0.1254

```

(Std. Err. adjusted for 38 clusters in isocode)

avg_ConfGov	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0513863	.0301397	-1.70	0.097	-.1124551	.0096824
avg_logcapdist	.0366353	.0328982	1.11	0.273	-.0300227	.1032933
avg_loggcppc	-.0235042	.0380182	-0.62	0.540	-.1005364	.0535279
avg_logpop	.0069635	.0058364	1.19	0.240	-.0048621	.0187891
imr	.0000818	.00007	1.17	0.250	-.00006	.0002235
logtttime	-.0184976	.0142923	-1.29	0.204	-.0474566	.0104614
logcellarea	.015858	.0165945	0.96	0.345	-.0177656	.0494817
avg_logdist_LNC	.0061557	.0109536	0.56	0.578	-.0160385	.0283499

mountain2000	-0.0452466	.0496811	-0.91	0.368	-0.1459101	.0554169
ycoord	-0.0030654	.0028539	-1.07	0.290	-0.008848	.0027173
avg_degtemper	-0.0025708	.0020002	-1.29	0.207	-0.0066235	.001482
avg_prec	-0.000054	.000025	-2.16	0.037	-0.0001047	-3.29e-06
_cons	.4177914	.3719111	1.12	0.269	-0.3357722	1.171355
isocode	absorbed	(38 categories)				

```
. areg avg ConfNonIntense logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

(Std. Err. adjusted for **38** clusters in isocode)

```
. use newbygid_confdata, clear

. keep if Asia == 1
(51,846 observations deleted)

. areg avg ConfIntense logdist gcp ppp ${newcontrols2}, a(isocode) cluster(iso)
```

(Std. Err. adjusted for 38 clusters in isocode)

avg_ConfIntense	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0015492	.0080429	0.19	0.848	-.0147473	.0178457
avg_logcapdist	-.0123402	.0060001	-2.06	0.047	-.0244976	-.0001828
avg_loggcppc	-.0063057	.0139871	-0.45	0.655	-.0346463	.0220349
avg_logpop	.0024851	.0015111	1.64	0.109	-.0005768	.0055469
imr	.000013	.0000159	0.82	0.420	-.0000192	.0000451
logttime	.001679	.0034199	0.49	0.626	-.0052504	.0086084
logcellarea	-.0004177	.0059734	-0.07	0.945	-.012521	.0116857
avg_logdist_LNC	.0098198	.0054748	1.79	0.081	-.0012733	.0209129
mountain2000	.0130939	.0118473	1.11	0.276	-.010911	.0370988
ycoord	.0007764	.0007086	1.10	0.280	-.0006595	.0022122
avg_degtemper	-.0002394	.0005716	-0.42	0.678	-.0013976	.0009189
avg_prec	6.37e-06	1.81e-06	3.53	0.001	2.71e-06	.00001
_cons	.0172005	.0857458	0.20	0.842	-.1565369	.190938
isocode	absorbed		(38 categories)			

```
.
.
. use newbygid_confdata, clear

. keep if Asia == 1
(51,846 observations deleted)

.
. areg avg_ConfTerr logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	11,852
	F(12, 37)	=	5.58
	Prob > F	=	0.0000
	R-squared	=	0.5729
	Adj R-squared	=	0.5712
	Root MSE	=	0.1195

(Std. Err. adjusted for 38 clusters in isocode)

avg_ConfTerr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0072651	.0150207	0.48	0.631	-.0231698	.0376999
avg_logcapdist	.0114872	.0157514	0.73	0.470	-.0204281	.0434025
avg_loggcppc	-.0583106	.0427149	-1.37	0.180	-.1448592	.028238
avg_logpop	.0028286	.0037023	0.76	0.450	-.004673	.0103303
imr	-.0000291	.0000273	-1.07	0.293	-.0000844	.0000262
logttime	-.0007847	.0093495	-0.08	0.934	-.0197286	.0181591
logcellarea	.0190303	.0124188	1.53	0.134	-.0061326	.0441931
avg_logdist_LNC	.0217076	.0128331	1.69	0.099	-.0042948	.0477099
mountain2000	.0266529	.0312123	0.85	0.399	-.0365892	.0898951
ycoord	.0039742	.0038489	1.03	0.309	-.0038244	.0117728
avg_degtemper	.000256	.00126	0.20	0.840	-.002297	.002809
avg_prec	.000055	.0000467	1.18	0.247	-.0000397	.0001496
_cons	-.0988543	.3553177	-0.28	0.782	-.8187963	.6210877
isocode	absorbed		(38 categories)			

.

```

.
. use newbygid_confdata, clear

. keep if Asia == 1
(51,846 observations deleted)

```

```

. areg avg_ConfInter logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      11,852
                                           F(   12,        37) =      31.72
                                           Prob > F           =      0.0000
                                           R-squared          =      0.1646
                                           Adj R-squared      =      0.1611
                                           Root MSE          =      0.0344

```

(Std. Err. adjusted for 38 clusters in isocode)

avg_ConfInter	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0069701	.0043837	1.59	0.120	-.0019122	.0158523
avg_logcapdist	-.0094766	.0050373	-1.88	0.068	-.0196832	.00073
avg_loggcppc	-.0022562	.0029681	-0.76	0.452	-.00827	.0037577
avg_logpop	.0008093	.0005838	1.39	0.174	-.0003735	.0019922
imr	-4.38e-06	4.73e-06	-0.93	0.360	-.000014	5.20e-06
logtttime	.0049566	.0032869	1.51	0.140	-.0017033	.0116165
logcellarea	-.0033177	.0029857	-1.11	0.274	-.0093673	.0027319
avg_logdist_LNC	-.0005144	.0012031	-0.43	0.671	-.0029521	.0019233
mountain2000	.0065801	.0060347	1.09	0.283	-.0056474	.0188077
ycoord	.0004746	.0005712	0.83	0.411	-.0006828	.0016319
avg_degtemper	-.0000504	.0002659	-0.19	0.851	-.0005892	.0004883
avg_prec	6.44e-06	2.64e-06	2.44	0.020	1.09e-06	.0000118
_cons	.0090349	.0307177	0.29	0.770	-.0532051	.0712749

isocode absorbed (38 categories)

```

.
.
.
. //conflict type for Extractive colonies
. use newbygid_Confdata, clear

```

```

. keep if avg_polity2>0
(16,515 observations deleted)

```

```

. areg avg_ConfGov logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)

```

```

Linear regression, absorbing indicators      Number of obs      =      39,341
                                           F(   12,       100) =      1.11
                                           Prob > F           =      0.3583
                                           R-squared          =      0.7565
                                           Adj R-squared      =      0.7558
                                           Root MSE          =      0.0751

```

(Std. Err. adjusted for 101 clusters in isocode)

avg_ConfGov	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0453588	.0311876	-1.45	0.149	-.1072341	.0165166
avg_logcapdist	.0367968	.03428	1.07	0.286	-.0312138	.1048074
avg_loggcppc	.0007932	.0056528	0.14	0.889	-.0104219	.0120082
avg_logpop	.001188	.0009073	1.31	0.193	-.000612	.002988
imr	.0001123	.0000965	1.16	0.248	-.0000793	.0003038
logtttime	.000823	.0029839	0.28	0.783	-.005097	.006743
logcellarea	-.0106539	.0089395	-1.19	0.236	-.0283896	.0070818
avg_logdist_LNC	-.0001234	.0088711	-0.01	0.989	-.0177234	.0174767
mountain2000	-.0005602	.0122405	-0.05	0.964	-.024845	.0237245
ycoord	-.000289	.0006415	-0.45	0.653	-.0015618	.0009837
avg_degtemper	.0007157	.0007094	1.01	0.315	-.0006918	.0021232

avg_prec	-.000018	.0000112	-1.60	0.113	-.0000403	4.32e-06
_cons	.1428681	.1196228	1.19	0.235	-.0944601	.3801964

isocode	absorbed	(101 categories)
---------	----------	------------------

```
.
.
. use newbygid_confddata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

.
. areg avg_ConfNonIntense logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      39,341
                                             F( 12, 100)        =      1.51
                                             Prob > F           =      0.1320
                                             R-squared          =      0.7675
                                             Adj R-squared      =      0.7669
                                             Root MSE          =      0.0719
```

(Std. Err. adjusted for 101 clusters in isocode)

avg_ConfNonIn~e	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	-.0444974	.0228874	-1.94	0.055	-.0899053	.0009105
avg_logcapdist	.042492	.0287022	1.48	0.142	-.0144524	.0994364
avg_loggcppc	-.0164525	.010723	-1.53	0.128	-.0377265	.0048216
avg_logpop	.0009257	.0009771	0.95	0.346	-.0010129	.0028642
imr	.0000759	.0000688	1.10	0.273	-.0000607	.0002124
logttime	-.0030007	.0041096	-0.73	0.467	-.0111541	.0051526
logcellarea	-.0091802	.0053685	-1.71	0.090	-.0198311	.0014707
avg_logdist_LNC	.004501	.0067824	0.66	0.508	-.0089552	.0179572
mountain2000	.0011692	.0087711	0.13	0.894	-.0162325	.0185709
ycoord	-.0002982	.0003987	-0.75	0.456	-.0010892	.0004928
avg_degtemper	-.0000982	.0003708	-0.26	0.792	-.000834	.0006375
avg_prec	3.94e-06	9.69e-06	0.41	0.685	-.0000153	.0000232
_cons	.2409759	.0920351	2.62	0.010	.0583808	.4235709

isocode	absorbed	(101 categories)
---------	----------	------------------

```
.
. use newbygid_confddata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

.
. areg avg_ConfIntense logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

```
Linear regression, absorbing indicators      Number of obs      =      39,341
                                             F( 12, 100)        =      2.17
                                             Prob > F           =      0.0189
                                             R-squared          =      0.7741
                                             Adj R-squared      =      0.7734
                                             Root MSE          =      0.0229
```

(Std. Err. adjusted for 101 clusters in isocode)

avg_ConfIntense	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0099572	.0057571	1.73	0.087	-.0014648	.0213791
avg_logcapdist	-.0121785	.0065806	-1.85	0.067	-.0252343	.0008773
avg_loggcppc	-.0027125	.0026894	-1.01	0.316	-.0080482	.0026232
avg_logpop	.0005719	.0003176	1.80	0.075	-.0000583	.0012021
imr	-4.33e-06	9.29e-06	-0.47	0.642	-.0000228	.0000141
logttime	-.0002226	.0007908	-0.28	0.779	-.0017915	.0013462
logcellarea	.0009397	.0021104	0.45	0.657	-.0032474	.0051267
avg_logdist_LNC	.0024536	.0012611	1.95	0.055	-.0000484	.0049556

mountain2000	.0041038	.0029897	1.37	0.173	-.0018278	.0100353
ycoord	.0001062	.0001886	0.56	0.575	-.000268	.0004804
avg_degtemper	-.0002816	.0002224	-1.27	0.208	-.0007229	.0001596
avg_prec	-3.67e-07	1.05e-06	-0.35	0.727	-2.45e-06	1.71e-06
_cons	.0207839	.0271632	0.77	0.446	-.0331071	.074675

isocode	absorbed	(101 categories)
---------	----------	------------------

```
.
.
. use newbygid_confdata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

.
. areg avg_ConfTerr logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	39,341
	F(12, 100)	=	2.39
	Prob > F	=	0.0094
	R-squared	=	0.4224
	Adj R-squared	=	0.4208
	Root MSE	=	0.0653

(Std. Err. adjusted for 101 clusters in isocode)

avg_ConfTerr	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0118066	.015606	0.76	0.451	-.0191553	.0427685
avg_logcapdist	-.0065542	.0139661	-0.47	0.640	-.0342625	.0211542
avg_loggcppc	-.0196383	.0099072	-1.98	0.050	-.0392939	.0000173
avg_logpop	.0002798	.0007043	0.40	0.692	-.0011175	.0016772
imr	-.0000395	.0000406	-0.97	0.333	-.0001202	.0000411
logttime	-.0045415	.0035441	-1.28	0.203	-.0115729	.0024898
logcellarea	.0031838	.0099615	0.32	0.750	-.0165795	.0229471
avg_logdist_LNC	.0073141	.0043145	1.70	0.093	-.0012458	.0158741
mountain2000	.0054474	.0071933	0.76	0.451	-.008824	.0197188
ycoord	.0000548	.0004699	0.12	0.907	-.0008774	.0009869
avg_degtemper	-.0011541	.0010874	-1.06	0.291	-.0033115	.0010034
avg_prec	.0000237	.0000207	1.14	0.257	-.0000175	.0000648
_cons	.1060088	.0747683	1.42	0.159	-.0423294	.254347

isocode	absorbed	(101 categories)
---------	----------	------------------

```
.
.
. use newbygid_confdata, clear

. keep if avg_polity2>0
(16,515 observations deleted)

.
. areg avg_ConfInter logdist_gcp_ppp ${newcontrols2}, a(isocode) cluster(iso)
```

Linear regression, absorbing indicators	Number of obs	=	39,341
	F(12, 100)	=	1.21
	Prob > F	=	0.2860
	R-squared	=	0.0663
	Adj R-squared	=	0.0636
	Root MSE	=	0.0191

(Std. Err. adjusted for **101** clusters in isocode)

avg_ConfInter	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
logdist_gcp_ppp	.0070721	.0035514	1.99	0.049	.0000261	.014118
avg_logcapdist	-.0085403	.0043054	-1.98	0.050	-.017082	1.53e-06
avg_loggcppc	-.0009272	.0009583	-0.97	0.336	-.0028284	.000974
avg_logpop	.0003781	.0002438	1.55	0.124	-.0001056	.0008619
imr	-6.63e-06	4.98e-06	-1.33	0.186	-.0000165	3.26e-06
logttime	.0008413	.0008428	1.00	0.321	-.0008309	.0025135
logcellarea	.0017054	.0014797	1.15	0.252	-.0012303	.004641
avg_logdist_LNC	.0010105	.0005362	1.88	0.062	-.0000532	.0020742
mountain2000	.0015601	.0014781	1.06	0.294	-.0013723	.0044926
ycoord	.0000583	.0000777	0.75	0.455	-.000096	.0002125
avg_degtemper	-.0003154	.0002156	-1.46	0.147	-.0007431	.0001124
avg_prec	1.54e-06	8.74e-07	1.77	0.081	-1.91e-07	3.28e-06
_cons	-.0078313	.0082636	-0.95	0.346	-.024226	.0085635
isocode	absorbed		(101 categories)			

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end of do-file

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