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The Transition of Corruption: From Poverty to Honesty

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Abstract

Measures of corruption and income are highly correlated across countries. We use prehistoric measures of biogeography as instruments for modern income levels to identify an exogenous long-run income effect. We find that our corruption-free incomes explain the cross-country pattern of corruption just as well as actual incomes. This result suggests that the long-run causality is exclusively from income to corruption. As countries get rich, corruption vanishes and there is a transition of corruption from poverty to honesty.

Keywords Long-run growth, corruption, biogeography

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1. Introduction

Measures of corruption and income are highly correlated across countries, as shown by Table 1. No agreement has been reached about the main direction of causality between the two variables. Some (as Lambsdorff 2007) claim that corruption is a vice causing low growth, so that causality is mainly from corruption to long-run income levels. Others (as Treisman 2000 and Paldam 2001 and 2002) claim that corruption is a poverty driven disease that vanishes when countries develop, so that causality is mainly from the level of income to corruption.

The empirical problem is that corruption data series contain much autocorrelation, so the existing data series are too short to permit formal time series tests of causality. We have summarized the findings till now in Paldam and Gundlach (2008), which also provides an informal causality analysis. This informal test suggests that the relation between income and corruption is just another *transition* that occurs when countries grow from poor to rich.³ The purpose of the present paper is to provide a formal test of the long-run direction of causality between income and corruption. Our test is based on the present cross-country pattern (*c-cp*) of corruption and income.

Given that all countries experienced fairly similar average income levels 200-500 years ago, the modern income levels reveal cross-country differences in the long-run growth rate. So regressions of measures of corruption on levels of income will identify the long-run income effect if the possible reverse causality from corruption to income can be controlled for. We explain the present *c-cp* of income by a set of extreme instruments (listed in the Appendix) that are fully independent of the present *c-cp* of corruption to obtain a corruption-free *c-cp* of incomes. We then show that these incomes explain the *c-cp* of corruption just as well as the actual *c-cp* of incomes and conclude from this result that the long-run causality is entirely from income to corruption: hence the transition of corruption.

The paper proceeds as follows: Section 2 explains the logic of the test and presents the main result. Section 3 presents some robustness test, and Section 4 concludes. We are brief since a parallel paper analyzing the relation between democracy and income (Gundlach and Paldam 2008) provides a more detailed discussion of our empirical model and of the literature that appears to support our empirical strategy. All variables used are defined in the Appendix.

3. We use the term *Grand Transition* for the set of transitions (the demographic, the urban, the democratic, etc.), which together constitute development, see Paldam and Gundlach (2008).

Table 1. Correlations between corruption and income for all available years

Year covered ^{a)}	1994	1995	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006
<i>N</i> of TI-index ^{b)}	41	53	52	85	99	90	91	102	133	146	159	163	176
<i>N</i> in correlation ^{c)}	40	52	51	84	98	89	89	101	131	141	149	153	153
Correlation	0.76	0.81	0.79	0.77	0.77	0.78	0.83	0.81	0.81	0.80	0.79	0.79	0.78

Note: Variables are defined in Appendix. (a) Year of underlying corruption poll, income is for previous year. (b) Number of countries with corruption data. (c) Number of these countries for which an income observation is also available. The average correlation is 0.79, with SD 0.02. The correlation is trendless.

2. The logic of the test, the instruments, and the key result

The core of this paper is a test comparing two regressions. One is an OLS-regression explaining the c-cp of corruption by the c-cp of incomes with a lag. The other is a TSIV-regression between the same two variables where the income variable is instrumented by extreme variables chosen to be fully independent of the current pattern of corruption.

The instruments used are biogeographic variables, which are all listed in the Appendix. Most are taken from Hibbs and Olsson (2004, 2005). They are exogenous to present levels of income in the perspective of at least the last millennium, and it is inconceivable that the present c-cp of corruption can have affected any of these variables.

Our justification for using these variables as instruments is taken from various sources. Diamond (1997) argues that the diversity of prehistoric biological and geographic conditions has been crucial for the diversity of early economic development in different regions of the world, which may explain the different sizes of regional populations at average income levels not far above the subsistence level throughout human history until about 500 years ago. Galor (2005) develops a unified growth theory that captures in one model the Malthusian era of constant per capita income and the era of modern growth with persistently rising per capita income. His theory is supported, mainly for England, by historical facts about slow but steady changes in the structure of the population during the Malthusian era that would eventually allow for a take off to modern growth (see Clark 2007).

This combined empirical-theoretical argument suggests to us that in some sense the diversity of present levels of development may at least partly originate from prehistoric biological and geographic conditions, which Hibbs and Olsson (2004, 2005) have collected and rescaled to be applicable for present day countries. Their work shows that measures of biogeography do explain modern levels of economic development surprisingly well. Our

conclusion from this literature is that measures of prehistoric biogeography appear to be strong instruments for modern levels of income. Our basic estimation equation is given by

$$\kappa_i = \alpha + \beta y_i + X_i' \gamma + \varepsilon_i, \quad (1)$$

where κ_i is the degree of corruption in country i , y is the natural logarithm of GDP per capita in constant international dollars, X_i' is a matrix of other covariates that may be included, α is a regression constant, ε is an error term, and β is the coefficient of interest that measures the long-run effect of income on corruption once income is appropriately instrumented.

Table 2. The estimated effect of income on corruption: TSIV and OLS regressions

	Dependent variable: κ (average TI index for 1995-2006)				
	(1)	(2)	(3)	(4)	(5)
No. of obs. (countries)	98	98	103	98	141
y (IV) for 2003	1.49 (0.16)	1.27 (0.15)	1.57 (0.15)	1.52 (0.16)	1.29 (0.13)
Instruments	<i>animals,</i> <i>plants</i>	<i>axis, size,</i> <i>climate</i>	<i>bioavg,</i> <i>geoav</i>	<i>biofpc,</i> <i>geofpc</i>	<i>coast, frost,</i> <i>maleco</i>
First stage partial R^2	0.43	0.53	0.51	0.43	0.52
CD F-statistic	35.86	35.17	52.50	36.09	49.60
CD critical value (size)	19.93	22.30	19.93	19.93	22.30
Sargan test (p-value)	1.45 (0.23)	4.37 (0.11)	0.93 (0.34)	0.02 (0.90)	7.54 (0.02)
y (OLS) for 2003	1.36 (0.11)	1.36 (0.11)	1.47 (0.11)	1.36 (0.11)	1.41 (0.09)
Adjusted R^2	0.62	0.62	0.65	0.62	0.62

Notes: Variables are defined in Appendix. The selected combination of variables and years maximizes the number of available observations. The results are unchanged (available upon request) if both the income variable and TI index refer to 2003, or if the average κ -index is explained with the 1995 income levels. Standard errors in parentheses. All specifications include a constant term (not reported). A Cragg-Donald (CD) F-statistic below the critical value (10 percent maximal size) indicates weak instruments. The Sargan test for overidentification tests the joint null hypothesis that the instruments are valid and correctly excluded from the estimated equation.

Our results for specifications of equation (1) with alternative sets of instrumental variables are reported in Table 2. Note first that the selected measures of biogeography pass the Cragg-Donald test for weak instruments and in all but one case the Sargan test for overidentification. Also, all ten estimates of the income coefficient are statistically highly significant. The average point estimate of the long-run income effect on corruption is about 1.4. Hence the difference between the 10th percentile (6.61) and the 90th percentile (9.93) of the (log) income measure predicts a $3.3 \cdot 1.4 \approx 4.65$ corruption-point difference in our sample of countries. The actual difference between two sample countries that are close to the 10th percentile and the

90th percentile, Haiti and Finland, is 7.85 corruption points, so our estimate explains about 60 percent of the observed difference in the corruption index of the two countries. Yet the key result from Table 2 is that the TSIV-results do not differ statistically significantly from the corresponding OLS-results in any of the five specifications. This suggests that the OLS estimates are not upwardly biased due to reverse causality, and hence all long-run causality appears to run from income to corruption.

3. Robustness of the main result

One major objection to the main result in Table 2 is that our estimates are biased due to omitted variables. We check the robustness of our estimate of the income effect on corruption by adding ten control variables one by one to avoid multicollinearity. Our control variables are either socio-political (Table 3) or ethno-cultural (Table 4). They are chosen to have an effect on corruption which may either be independent of the income effect or may even dominate the presumed income effect.⁴ All results presented in Tables 3 and 4 are based on a specification with our preferred instrumental variables, namely the first principal components of the measures of biogeography (see column (4) of Table 2).

Table 3. The effect of selected socio-political variables on corruption

	Dependent variable: κ (average TI index for 1995-2006)			
	(1)	(2)	(3)	(4)
No. of obs. (countries)	90	71	61	39
y (IV) in 2003	1.66 (0.15)	1.61 (0.23)	1.67 (0.27)	2.03 (0.41)
<i>mining</i>	2.25 (1.83)			
<i>gini</i>		-0.00 (0.02)		
<i>homicavg</i>			-0.01 (0.01)	
<i>suicide</i>				0.06 (0.02)
First stage partial R^2	0.47	0.33	0.39	0.39
CD F-statistic	38.74	16.66	18.47	11.23
CD critical value (size)	19.93	19.93	19.93	19.93
Sargan test (p-value)	0.03 (0.86)	0.01 (0.94)	0.01 (0.94)	0.52 (0.47)
y (OLS)	1.46 (0.10)	1.48 (0.13)	1.69 (0.17)	2.09 (0.26)
Adjusted R^2	0.68	0.67	0.64	0.74

Notes: Variables are defined in Appendix. See Table 2. In the IV regressions, *biofpc* and *geofpc* are used as instruments, as in column (4) of Table 2. OLS results are conditional on the inclusion of the control variable.

4. For our purpose, it is less important whether the additional control variables are actually exogenous. We are mainly interested in the robustness of our estimated income coefficient.

Table 4. The effect of selected ethno-cultural variables on corruption

	Dependent variable: κ (average TI index for 1995-2006)					
	(1)	(2)	(3)	(4)	(5)	(6)
No. of obs. (countries)	94	98	98	98	98	98
y (IV) in 2003	1.83 (0.25)	1.42 (0.16)	1.59 (0.17)	1.42 (0.14)	1.46 (0.15)	1.52 (0.16)
<i>ethnoel</i>	0.93 (0.71)					
<i>lofre</i>		-0.57 (0.25)				
<i>loeng</i>			0.66 (0.30)			
<i>prot</i>				3.18 (0.54)		
<i>romcat</i>					-0.83 (0.35)	
<i>muslim</i>						0.15 (0.45)
First stage partial R^2	0.28	0.42	0.41	0.42	0.50	0.50
CD F-statistic	17.57	34.69	33.27	34.54	46.29	47.69
CD critical value (size)	19.93	19.93	19.93	19.93	19.93	19.93
Sargan test (p-value)	0.00 (0.99)	0.09 (0.77)	0.01 (0.92)	0.92 (0.34)	0.00 (0.97)	0.00 (0.96)
y (OLS)	1.39 (0.13)	1.33 (0.11)	1.41 (0.11)	1.26 (0.09)	1.41 (0.11)	1.36 (0.11)
Adjusted R^2	0.64	0.64	0.63	0.73	0.64	0.63

Notes: Variables are defined in Appendix. See Table 3. In the IV regressions, *biofpc* and *geofpc* are used as instruments, as in column (4) of Table 2. OLS results are conditional on the inclusion of the control variable.

Table 3 considers four variables characterized as socio-political controls. Here we encounter some problems with the Cragg-Donald test for weak instruments, so the results are not fully reliable. However, the first stage partial R^2 is still high and the estimated income effects do look very much as before, whereas none of the included controls appears to have an independent statistically significant effect on corruption.

Table 4 considers six ethno-cultural variables. In line with previous studies, we find that a high share of the population with protestant religious belief (*prot*) increases the degree of honesty for a given level of income (Paldam 2001). For all other included control variables, we do not find statistically significant effects on corruption. Our preferred instruments are rejected as weak in only one case by the Cragg-Donald tests (column (1)). Our estimates of the income effect on corruption again remain the same as before, with all OLS-estimates within the 95 percent confidence interval of the TSIV-estimates. More generally, we find with only one (narrow) exception that a point estimate of 1.4 is within the 95 percent confidence interval of all income effects reported in Tables 2-4.

4. Conclusion: the transition of corruption

Our paper uses the cross-country levels of income and corruption to identify the long-run effect of income on corruption. We handle the problem of reverse causality by a unique set of prehistoric measures of biogeography, which pass statistical tests for weak instruments. We follow Simon Kuznets (see his essays since 1954 in Kuznets 1965) and many later researchers by arguing that current cross-country levels of income provide the best information about cross-country differences in long-run growth. Once this argument is taken for granted, our results show:

The observed cross-country pattern of corruption can be fully explained by the cross-country differences in the long-run development of income. To the extent that there is short-run interaction between corruption and income – and there may very well be one – it is irrelevant for the long-run effect. Consequently, the long-run causality is exclusively from income to corruption. As countries get rich, corruption vanishes by the transition of corruption from poverty to honesty.

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Appendix Table: Definitions and sources of variables used in tables

Dependent variable and main explanatory variable. Used in Tables 1 to 4.

κ	Corruption/honesty index from Transparency International, scale from 0 (honesty) to 10 (corruption).
y	Natural logarithm of GDP per capita in 1990 PPP-\$. Source: Maddison homepage.

Biogeographical instrumental variables. All are used in Tables 2. Table 3 and 4 use only *biofpc* and *geofpc*

<i>animals</i>	Number of domesticable big mammals in prehistory. Source: Olsson and Hibbs (2005).
<i>axis</i>	Captures the rate of East-West orientation of country. Source: Olsson and Hibbs (2005).
<i>bioavg</i>	Average of plants and animals. Source: Hibbs and Olsson (2004).
<i>biofpc</i>	The first principal component of plants and animals. Source: Olsson and Hibbs (2005).
<i>climate</i>	Ranking of climates by how favorable they are to agriculture. Source: Olsson and Hibbs (2005).
<i>coast</i>	Proportion of land area within 100 km of the sea coast. Source: McArthur and Sachs (2001).
<i>frost</i>	Proportion of a country with five or more frost days. Source: Masters and McMillan (2001).
<i>geoavg</i>	Average of climate, lat, and axis. Source: Hibbs and Olsson (2004).
<i>geofpc</i>	The first principal component of climate, lat, axis and size. Source: Olsson and Hibbs (2005).
<i>lat</i>	Distance from the equator. Source: Hall and Jones (1999).
<i>maleco</i>	Measure of malaria ecology, www.earth.columbia.edu/about/director/malaria/index.html#datasets .
<i>plants</i>	Number of domesticable wild grasses in prehistory. Source: Olsson and Hibbs (2005).
<i>size</i>	The size of the landmass to which the country belongs. Source: Olsson and Hibbs (2005).

Socio-political control variables. Used in Table 3

<i>mining</i>	Share of GDP in the mining and quarrying sector, approx. 1988. Source: Hall and Jones (1999).
<i>gini</i>	Gini coefficient, approx. 1990. Source: Deininger and Squire (1996).
<i>homicavg</i>	Total completed homicides per 100,000 population, average for 1990-2000. Source: UNODC (2005).
<i>suicide</i>	Total number of suicides per 100,000 population, estimates for early 1990s. Source: Parker (1997).

Ethno-cultural control variables. Used in Table 4

<i>ethnoel</i>	Average value of five indices of ethnolinguistic fractionalization. Source: La Porta et al. (1998).
<i>loeng</i>	Dummy for English origin of Commercial Law. Source: La Porta et al. 1998
<i>lofre</i>	Dummy for French origin of Commercial Law. Source: La Porta et al. 1998
<i>muslim</i>	Share of the population with Muslim religious belief. Source: La Porta et al. (1998).
<i>prot</i>	Share of the population with protestant religious belief. Source: La Porta et al. (1998).
<i>romcat</i>	Share of the population with roman-catholic religious belief. Source: La Porta et al. (1998).

Note: See Gundlach and Paldam (2008) and sources.