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Corruption and Tax Burden: What Is the Joint Effect on Total Factor Productivity?

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Abstract: A common conclusion in the literature is that both corruption and taxation hamper economic growth. It is also plausible that both affect total factor productivity, which, by the famous Solow residual, is a vital driver of economic progress. Moreover, corruption and tax burden are supposed to be intertwined. This paper focuses on the supposedly linked effects of corruption and tax burden on total factor productivity. The empirical study uses panel data from 90 countries for the time span of 1996–2014. The results show that both corruption and tax burden deteriorate total factor productivity, but that an increase in tax burden mitigates the negative effect of corruption.

Keywords: corruption; Solow residual; tax burden; total factor productivity

JEL Classification: D7; O4; H2



Citation: Kéïta, Kouramoudou, and Hannu Laurila. 2021. Corruption and Tax Burden: What Is the Joint Effect on Total Factor Productivity?.

Economies 9: 26. <https://doi.org/10.3390/economies9010026>

Academic Editor: George R. G. Clarke

Received: 27 January 2021

Accepted: 17 February 2021

Published: 1 March 2021

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1. Introduction

Corruption is usually defined as misuse of public office for private gain. The mainstream approach in the literature on the relationship between corruption and economic development is to focus on the link between corruption and investments or output growth. Early assessments along this line include [Shleifer and Vishny \(1993\)](#), [Mauro \(1995\)](#), [Bardhan \(1997\)](#), and [Ades and Di Tella \(1999\)](#). A covering conclusion is that corruption hampers economic performance.

By definition, tax burden is the amount of tax paid by a person, company, or country in a specified period considered as a proportion of total income in that period. It consists of various taxes and tariffs imposed on economic agents and activities. [McBride \(2012\)](#) provides a comprehensive summary of economic studies concerning the effects of taxation. The conclusion is that the overall effect of distortive taxation on economic growth is negative.

The evidence about the specific channels of the effects of corruption on economic progress is substantial but somewhat mixed ([Cieslik and Goczek 2018](#)). [Mauro \(1995\)](#) reports a negative correlation between corruption and growth but also finds that the effect is insignificant when growth is controlled for investments. [Mo \(2001\)](#) adds that the negative effect of corruption on growth disappears when human capital is used as an explanatory variable.

[Knack and Keefer \(1995\)](#) point to the role of institutions as an important factor of economic growth. [Davoodi and Tanzi \(1997\)](#) finds that, under the assumption that effective investments enhance capital productivity, corruption spoils the quality of infrastructures, thus depressing growth. [Gillanders \(2013\)](#) confirms the result. [Pellegrini and Gerlagh \(2004\)](#) find that besides the negative effect on investments, corruption hampers growth through trade, schooling, and political stability. [Reinikka and Svensson \(2005\)](#), [Seka \(2013\)](#),

Dridi (2014), and Bryant and Javalgi (2016) highlight the detrimental impact of corruption on human capital.

In the literature, there are also arguments for possibly positive effects of corruption. For example, Leff (1964), Leys (1965), Huntington (1968), Méon and Sekkat (2005), Méndez and Sepúlveda (2006), and Aidt et al. (2008) propose that corruption may “grease the wheels”, thus enhancing economic growth. This is claimed to be possible if local governance is deficient and institutions are ineffective, whereas corruption is doomed detrimental in well-organized states.

The above remarks pave the path for further explorations of the effects of corruption and taxation on growth. It is well recorded in the previous literature that productivity and economic growth are positively correlated. Productivity explains growth within countries as well as growth disparities between countries. The literature also confirms that the general quality of governance has a role in the evolution of local productivity (Knack and Keefer 1995; Hall and Jones 1999; Nordhaus 2002; Auzina-Emsina 2014).

“Social infrastructure” is an umbrella concept that covers governance, institutions, policy programs, enforcement, and so on. Hall and Jones (1999) state that a favorable economic environment motors the long-run growth of output per capita. Under supportive conditions, economic agents have incentives to invest, innovate, and transfer ideas (Del Mar Salinas-Jiménez and Salinas-Jiménez 2011). To put it more generally, good social infrastructure, provided by the government and financed by taxation, should facilitate the efficient utilization of human and physical capital.

Solow’s growth theory (Solow 1956; Swan 1956) is essentially about the share of physical and human capital in explaining economic growth. Elaborations of the theory show that assuming constant returns to scale and competitive factor markets, such growth accounting is possible (Abramovitz 1956; Baier et al. 2006). Abramovitz (1956) reports that a mere increase in physical and human capital contributed only 10% of output growth per capita in the United States during 1869–1878 and 1944–1953. According to Solow (1957), the respective share was 12% over 1900–1949. The findings suggest that growth is mostly explained by factors other than human and physical capital. The deviation between observed growth and that forecasted by increases in human and physical capital is called total factor productivity (TFP), which is also known as the Solow residual. Then, this a priori unknown productivity should be due to factors such as social infrastructure.

Later studies report somewhat lower but still noteworthy TFP residuals (Kendrick 1961; Jorgenson et al. 1987; Abramovitz and David 2000). Baier et al. (2006) examines 143 countries and finds that the share of TFP is 14% of the growth per worker on average. In Western countries, the share is 34%, and in Southern Europe and in newly industrialized countries, it is 26%, but it is negative in Sub-Saharan Africa and Middle East countries. The study also reveals that variations in output per worker are more sensitive to variations in TFP than to variations in human and physical capital. So, it is reasonable that variables that affect growth should affect also TFP.

Corruption is mainly illegal, secret, and uncertain. It distorts the allocation of economic resources and rots governance (Shleifer and Vishny 1993). By Paldam (1999) and Lambsdorff (1999), integrity is beneficial to economic growth. It should also be obvious that corruption impedes TFP growth, especially when economic growth is very elastic to TFP (Abramovitz 1956; Baier et al. 2006). However, the impact of corruption on productivity has not received much attention in the literature. For example, Méon and Weill (2010) report that corruption may improve efficiency, while Olson et al. (2000) find that among several institutional quality measures, corruption tends to hinder productivity growth.

Some call corruption “informal taxation”, which does not involve the coffers of the state (Sanyal et al. 2000). This suggests that taxation and corruption are, to some extent, alternatives to each other. Wei (2000) finds that formal taxation is more efficient than the informal corrupt system by showing empirically that bribery has a much stronger negative impact on foreign direct investments than taxation. Lambsdorff (1999) finds that corruption reduces the gross domestic product (GDP) to capital stock ratio because of its correlation

with poor bureaucracy. [Fisman and Svensson \(2007\)](#) investigate the growth of Ugandan firms and show that bribery has a significantly stronger negative effect on the firms' growth than taxation.

Inspired by the findings reviewed above, this paper tests empirically the hypothesis that both corruption and tax burden affect TFP negatively. Moreover, the paper studies the implicit link between tax burden and corruption in that context. The spin-off hypothesis is that variations in the tax burden affect the negative impact of corruption. The reasoning rests on the assumption that an increase in tax burden motivates tax evasion and bribery ([Treismann 2000](#); [Goerke 2008](#); [Dzhumashev 2014](#); [Alm et al. 2016](#)). [Cieřlik and Goczek \(2018\)](#) argue that bribery, unlike taxation, causes unpredictable distortions in governance.

Speculations concerning the link between tax burden and corruption are manifold: under deficient governance, corruption is common and feasible. Then, a heavier tax burden may increase tax evasion and bribery, thus enforcing the effect of corruption on TFP. On the other hand, under good governance, corruption is costly. Then, grown tax revenues could facilitate investments in social infrastructure as well as in anti-corruption activities, which should dampen the negative effect of corruption on TFP. Moreover, [Wu and Schneider \(2019\)](#) postulate a U-shape evolution of the informal sector against economic growth: First, the informal sector tends to decrease but may even start to revive at some level of development. This suggests that the informal sector can generate some positive effects on the formal sector.

Our panel analysis shows that tax burden and corruption are positively correlated, but a high tax burden makes corruption less harmful. In a worldwide context, hard taxation is usually attached to developed countries with established institutions, effective social and technical infrastructures, and a small black market. On the other hand, in developing and poor countries, one may find high taxes but also a distorted allocation of resources, messy institutions, scrappy infrastructures, and a flourishing shadow economy.

The paper is organized as follows: Section 2 presents the data, the baseline model, and the applied econometrical methods. Section 3 presents and discusses the empirical findings, and Section 4 includes conclusions, policy recommendations, and prospects for future studies.

2. Data, Modeling, and Methods

The data cover 90 countries worldwide over the time span 1996–2014. The selection of the sample was determined by the availability of full data. The data sources are reported in Table 1 below, and the considered countries are listed in Appendix A, Table A1.

Table 1. Descriptive statistics and data sources.

Variables	Mean	Standard Deviation	Minimum	Maximum	Sources
<i>TFP</i> (level at current PPPs, in million 2011 US dollars)	0.680	0.324	0.105	2.492	Feenstra et al. (2015) , Penn World Table 9.0
<i>Inflation</i> (price level of household consumption, US GDP in 2011 = 1)	0.615	0.311	0.143	1.713	Feenstra et al. (2015) , Penn World Table 9.0
<i>Openness</i> (sum of shares of merchandise exports and imports in GDP, at current)	−0.042	0.162	−0.846	0.588	Feenstra et al. (2015) , Penn World Table 9.0
<i>Corruption</i> (Corruption Perception Index)	5.633	2.373	1.000	9.600	Transparency International
<i>Property rights/100</i> (overall quality of legal framework)	0.561	0.242	0.100	0.950	The Heritage Foundation
<i>Tax burden/100</i> (% of tax revenue to GDP)	0.703	0.145	0.298	0.999	The Heritage Foundation

Table 1. Cont.

Variables	Mean	Standard Deviation	Minimum	Maximum	Sources
<i>Government spending</i> (government consumption and all transfer)/100	0.663	0.219	0.100	0.993	The Heritage Foundation
<i>Improved sanitation</i> /100, rural (facilities, % of rural population with access)	0.683	0.339	0.021	1.000	World Development Indicators, World Bank
<i>Energy use</i> /10,000 (kg of oil equivalent per capita)	0.297	0.338	0.0009	2.276	World Development Indicators, World Bank
<i>Transport service</i> (% of commercial services exports/100)	0.236	0.149	0.0007	0.880	World Development Indicators, World Bank
<i>Electricity</i> (Electric power consumption (in kWh per capita)/10,000)	1.030	3.786	1.030	55.578	World Development Indicators, World Bank
<i>Military expenditure</i> (% GDP)	2.612	3.943	0.000	63.100	World Development Indicators, World Bank

Following the discussions in the previous section, the empirical estimations are based on the following baseline model:

$$TFP_{i,t} = \beta Corruption_{i,t} + \gamma TaxBurden_{i,t} + \lambda Z_{i,t} + \omega_i + \eta_t + \varepsilon_{i,t}. \quad (1)$$

On the left-hand side of Equation (1), the dependent variable $TFP_{i,t}$ measures the level of total factor productivity for country i in period t at constant purchasing power parity (PPP) rates. On the right-hand side, the main explanatory variables are $Corruption_{i,t}$ and $TaxBurden_{i,t}$. We concentrate on these two variables because of their documented negative effects on growth, e.g., (Ades and Di Tella 1999; McBride 2012), and because of their intrinsic interrelation postulated in the previous literature e.g., (Dzhumashev 2014).

Transparency International's Corruption Perception Index (CPI) is used as a proxy variable to capture the perceived prevalence of corruption in the sample countries. We are aware of the fact that the CPI values are comparable in time only from 2012 on (Fisman and Golden 2017; Gründler and Potrafke 2019), but we expect that the commonly used index still manages to reflect any trends over 1996–2014. The original CPI ranges from 0 to 10, where 0 stands for the highest possible level of corruption and 10 indicates full integrity. To ease the interpretation of the estimation results, the index values have been rescaled so that $Corruption_{i,t} = 11 - CPI_{i,t}$, meaning that a high variable value indicates extensive corruption and vice versa.

The second explanatory variable $TaxBurden_{i,t}$ measures the taxpayers' fiscal strain in the countries. It includes both marginal tax rates and the overall level of taxation (direct and indirect taxes) imposed by central and local governments as a percentage of gross domestic product (GDP). Thus, it is a rough measure of the formal public economy.

Equation (1) also includes the vector of instrument variables $Z_{i,t}$ with λ as the respective vector of coefficients. The instruments are as follows: *Inflation*—commonly understood to affect economic performance negatively (Romer 1993; Lane 1997); *Openness*—captures “creation-transmission and absorption of knowledge”, “integration”, and “competition” as facilitators of TFP (Maddison 1997, 1999; Frankel and Romer 1999); *Property rights*—represents the quality of legal frameworks with supposedly positive correlation with TFP (Ulubasoglu and Doucouliagos 2004; Isaksson 2007); *Government spending*—the size of public budgets that should link to TFP through social infrastructure (Del Mar Salinas-Jiménez and Salinas-Jiménez 2011); *Improved sanitation*—a proxy for health with assumedly positive connection to TFP; *Energy use*, *Electricity*, and *Transport service*—reflect technical infrastructure that should facilitate TFP (Aschauer 1989); *Military expenditure*—generates technological progress on one hand and links to corruption on the other hand (Delavallade 2006).

Lastly, in Equation (1), ω_i and η_t are dummies that capture country-specific effects and relevant time effects, respectively, and $\varepsilon_{i,t}$ represents the random error term including the influence of omitted variables.

Based on previous literature, the estimated coefficients β and γ of $Corruption_{i,t}$ and $TaxBurden_{i,t}$, respectively, are expected to be negative ($\beta < 0$, $\gamma < 0$). Figures 1 and 2 below illustrate the nexus between the main explanatory variables and TFP. Figure 1 shows the relationship between corruption and TFP in the whole sample of 90 countries over 1996–2014.

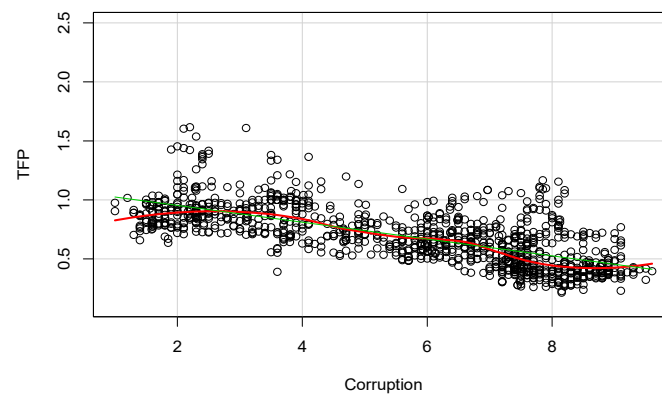


Figure 1. Total factor productivity (TFP) with respect to corruption.

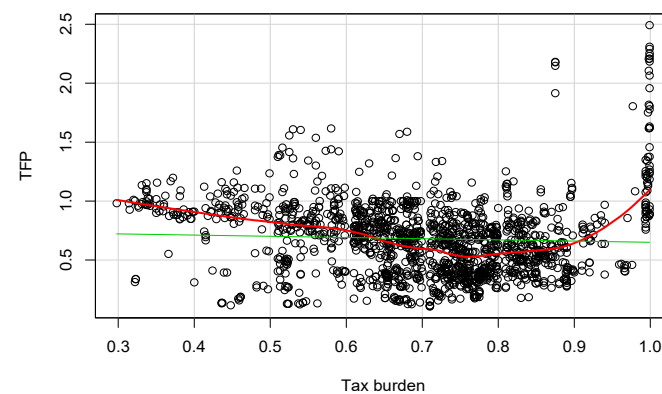


Figure 2. Total factor productivity (TFP) with respect to tax burden.

Figure 1 portrays a clearly negative relationship between corruption and TFP. The regional relationships presented in Appendix A, Figure A1 display the same expected pattern. Figure 2 illustrates the relationship between tax burden and TFP.

Figure 2 shows a negative relationship between tax burden and TFP. However, the green regression line representing fitted values is nearly horizontal, which indicates only a weak correlation between the variables. This is mainly due to the outlying observations along the tax burden value 1.0, which suggests data problems possibly due to incomparability between countries with differing political regimes cf. (Baier et al. 2006).

As proposed, there should also be a correlation between corruption and tax burden. Figure 3 illustrates the relationship in the study sample.

Figure 3 plots a scattered image but a clearly positive linkage between corruption and tax burden. One interpretation is that in certain countries, excessive taxation makes market agents avoid taxes by corrupt means (Alm et al. 2016). In turn, the expansion of the informal economy at the expense of the formal one should lead to a decline in tax revenue. Another explanation is that after some stage of economic development, the informal sector starts to generate positive effects on the formal sector (Wu and Schneider 2019).

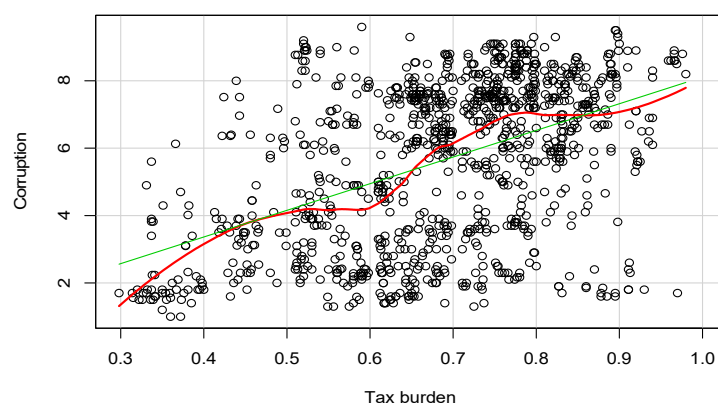


Figure 3. Corruption with respect to tax burden.

Descriptive statistics and data sources of our variables are reported in Table 1.

As a result of the time dimension, we first ensured that the time series are stationary, meaning that their distribution neither follows any trend nor changes over time. The Fisher type unit-root test was applied, and Appendix A, Table A2 reports the chi-squares and p -values associated with both Augmented Dickey–Fuller (AD–F) and Phillips–Perron (P–P) methods. In both cases, the non-stationarity hypothesis was systematically rejected.

To assess the validity of the econometrical method, estimations of Equation (1) were conducted sequentially by using Fixed Effects estimation, Two-Stage Least Squares (2SLS) method, and Two-Stage System Generalized Method of Moments (System GMM). To determine which estimator fits best to our data, standard specification tests (or homogeneity tests) for panel data were made.

First, a Fisher-type test for poolability was used. The null hypothesis argues for a pool data structure against the existence of fixed effects. Since the observed p -value appeared smaller than the standard threshold of 5%, the null hypothesis was rejected, pointing to the existence of fixed effects. This implies that countries in the sample exhibit distinctive features. The OLS (Ordinary Least Squares) estimators are not appropriate, as we can assume neither identical TFP functions for all the countries in the sample nor identical elasticities for corruption and tax burden.

Second, the Hausman test was performed to determine whether individual effects are random or fixed. The p -values were again smaller than 5%, so the Fixed Effects method is acceptable as it provides consistent estimators. Yet, some doubts may remain, because corruption is possibly an endogenous variable, which would make the Fixed Effects estimators inconsistent (Baltagi 1995). Due to the connection between TFP and social infrastructure (Méon and Sekkat 2005), the endogeneity problem is relevant, suggesting the use of Instrumental Variable (IV-2SLS) and System GMM methods. In particular, the System GMM proposed by Blundell and Bond (1998) provides consistent estimators that are likely to fulfill the orthogonality conditions while allowing rigorous control over instruments.

Since the dependent variable (TFP) and the endogenous regressor (corruption) influence each other, it is important to guarantee strict exogeneity and good quality of the instruments. According to Dzhumashev (2014), corruption is prevalent in public expenditures, and by Mauro (1998) and Delavallade (2006), corruption is positively associated with military expenditures. Therefore, these instruments are likely to be correlated with corruption, which is a key requirement for a proper instrument.

Lagged corruption should also be correlated, since corruption is supposedly endogenous. Therefore, the variables used to instrument corruption are *Government spending* _{t} , *Military expenditure* _{t} , *Corruption*₁₉₉₆ (initial corruption), and *Corruption* _{$t-1$} and *Corruption* _{$t-2$} (lagged corruption). The OLS regressions in Appendix A, Table A3 show that all the estimated coefficients of the instruments are statistically highly significant at the 0.1% level. The correlations remain robust with Fixed Effects estimators also when the determinant variables are included among explanatory variables. Therefore, the instruments are strongly

correlated with the endogenous variable. The Fisher test also corroborated this finding, since the p -values are systematically below the 5% level.

To determine whether there exists endogeneity, the Wu–Hausman test was conducted. Residuals in the first-stage regressions were introduced as explanatory variables in the basic model. The OLS estimations results are reported in Appendix A, Table A4. The estimated coefficients associated with the residuals are all statistically significant at the 5% level, meaning that the null hypothesis of exogeneity is rejected. Hence, the endogeneity problem exists. This aligns with the statistical significance of the estimated coefficients of residuals that indicate OLS to be biased toward the IV method.

As the last diagnostic on the instruments, the Sargan test was used. The test is applicable only if there are more instruments than endogenous variables. For IV-2SLS, *Corruption*₁₉₉₆, *Government spending*, and *Military expenditure* were used to instrument *Corruption* _{t,t} . Since the p -value appeared lower than the 5% level, the null hypothesis that instruments are exogenous is rejected.

The conclusion at this point is that the System GMM technique with the full list of instruments alongside those generated by default would yield the most reliable results.

3. Results

Table 2 below includes preliminary estimations relative to the Fixed Effects method, IV-2SLS, and System GMM (Models 1–6). In Models 1 and 2, TFP is estimated solely with respect to corruption, Model 3 includes the lagged dependent variable TFP_{t-1} , and Models 4–6 include select instrument variables.

Table 2. Preliminary estimations of TFP with respect to corruption.

	Dependent Variable: Total Factor Productivity (TFP)					
	Fixed Effects	IV-2SLS		System GMM		
	1	2	3	4	5	6
<i>Corruption</i>	−0.024 *** (0.005)	−0.264 *** (0.063)	−0.041 *** (0.024)	−0.319 *** (0.073)	−0.463 *** (0.091)	−0.383 *** (0.073)
<i>TFP</i> _{$t-1$}			1.094 *** (0.027)	1.305 *** (0.092)	1.760 *** (0.104)	1.514 *** (0.261)
<i>Openness</i>				0.174 ** (0.184)	0.106 ** (0.317)	0.384 ** (0.258)
<i>Property rights</i>				−0.273 *** (0.056)	−0.153 ** (0.042)	−0.559 ** (0.069)
<i>Improved sanitation</i>				0.092 ** (0.195)	0.219 ** (0.395)	0.407 ** (0.201)
<i>Energy use</i>				−0.031 * (0.039)		
<i>Electricity</i>					0.025 (0.107)	
<i>Transport service</i>						0.001 (0.038)
<i>Intercept</i>		2.434 *** (0.069)				
R–Squared	0.018	0.031				
Sargan test (p -value)			0.268	0.325	0.195	0.271
AR (1) (p -value)			0.0048	0.001	0.000	0.003
AR (2) (p -value)			0.170	0.371	0.309	0.103
Wald test, coefficients (p -value)		0.000	0.000	0.000	0.000	0.000
Wald test, dummies (p -value)			0.000	0.000	0.000	0.000
Number of observations	1088	1088	1984	1877	1606	1828

Notes. Robust standard errors of System GMM estimator are in parentheses. The superscripts ***, ** and * represent statistical significance at 0.1%, 1% and 5% levels, respectively.

The preliminary findings from Table 2 reveal that corruption has a negative influence on TFP. The estimated coefficients of *Corruption* are negative and statistically significant with all estimation methods. This confirms the trend shown by Figure 1 that corruption is associated with a decline in productivity. The estimation of Model 1 (Fixed Effects estimators) suggests that a one-unit increase in corruption would reduce TFP by 0.024%, *ceteris paribus*, while Model 2 (IV–2SLS) yields a considerably stronger reduction of 0.264%. In both cases, the R^2 measure is very low, implying that important determinants of TFP are not accounted in the estimations.

Model 3 with System GMM predicts a smaller negative effect compared to Model 2, namely 0.041%. The Sargan test shows that the instruments are valid (p -value > 0.05). The Arrelano–Bond tests AR (1) and AR (2) with p -values $0.0048 < 0.05$ and $0.17 > 0.05$ and the Wald test (p -value < 0.05) confirm the conclusiveness of the results.

Models 4–6 in Table 2 pay more attention to System GMM. In line with Isaksson (2007), additional control variables are used to assess the robustness of the relationship between *Corruption* and *TFP*. The extra variables make the negative effect of corruption and the positive effect of lagged TFP stronger with high statistical significance. So, the results are acceptably robust. The estimates of Models 4–6 also show that *Openness* and *Improved sanitation* are clearly positively correlated with TFP as expected. However, *Property rights* is negatively correlated contrary to expectations, as well as *Energy use* in Model 4. We come back to this later.

In Table 3, Models 7–12 incorporate *Government spending* and *Inflation* as additional explanatory variables to the System GMM estimations.

Table 3. System GMM estimations of TFP including more control variables.

	Dependent Variable: Total Factor Productivity (TFP)					
	System GMM					
	7	8	9	10	11	12
<i>Corruption</i>	−0.043 *** (0.012)	−0.074 *** (0.024)	−0.067 *** (0.021)	−0.068 *** (0.034)	−0.084 *** (0.017)	−0.055 *** (0.022)
<i>TFP</i> _{<i>t</i>−1}	0.946 *** (0.032)	0.984 *** (0.020)	0.961 *** (0.034)	0.904 *** (0.031)	0.917 *** (0.024)	0.984 *** (0.038)
<i>Openness</i>	0.071 *** (0.031)	0.077 *** (0.035)	0.068 *** (0.027)	0.069 *** (0.031)	0.072 ** (0.030)	0.081 ** (0.032)
<i>Property rights</i>	−0.061 *** (0.023)	−0.057 ** (0.024)	−0.071 ** (0.038)	−0.038 ** (0.044)	−0.064 *** (0.028)	−0.051 ** (0.032)
<i>Improved sanitation</i>	0.071 ** (0.047)	0.073 ** (0.041)	0.061 *** (0.034)	0.063** (0.051)	0.071 *** (0.063)	0.083 *** (0.074)
<i>Energy use</i>	0.017 (0.005)	0.008 * (0.005)				
<i>Government spending</i>		−0.012 * (0.025)		−0.043 * (0.031)		−0.014 * (0.045)
<i>Inflation</i>	−0.016 ** (0.036)	−0.027 *** (0.024)	−0.034 *** (0.031)	−0.039 *** (0.042)	−0.041 *** (0.031)	−0.051 *** (0.042)
<i>Electricity</i>			0.015 (0.027)	0.027 (0.061)		
<i>Transport service</i>					0.053 ** (0.041)	0.049 * (0.038)
Sargan test (p -value)	0.732	0.537	0.342	0.479	0.529	0.268
AR (1) (p -value)	0.003	0.001	0.028	0.019	0.010	0.008
AR (2) (p -value)	0.695	0.720	0.648	0.746	0.681	0.539
Wald test, coeffic. (p -value)	0.000	0.000	0.000	0.000	0.000	0.000
Wald test, dumm. (p -value)	0.000	0.000	0.000	0.000	0.000	0.000
Number of observations	1877	1846	1606	1586	1828	1803

Notes. Robust standard errors of System GMM estimator are in parentheses. The superscripts ***, **, and * represent statistical significance at 0.1%, 1%, and 5% levels, respectively.

The findings from both Tables 2 and 3 confirm that corruption seriously impedes TFP. The estimated coefficients of *Corruption* are overall negative with the best statistical significance standard. More specifically, over the nine System GMM regressions (Models 4–12), the predicted drop in TFP is about 0.17% when corruption increases by one unit (calculated as the mean of estimated coefficients of corruption over the estimations). Thus, the preliminary estimations are corroborated. All the diagnostic tests are also approvable.

In both Tables 2 and 3, the estimated coefficients of the variable *Openness* (compounding of “creation—transmission and absorption of knowledge”, “integration”, and “competition”) are clearly positive and statistically significant as expected.

Our expectation on *Property rights* was that it should be a positive institutional factor of TFP. Knack and Keefer (1995) find a positive relationship between property rights and GDP, and Ulubasoglu and Doucouliagos (2004) find a positive relation between democratic institutions and TFP (see also Przeworski and Limongi 1993; Isaksson 2007). However, both Tables 2 and 3 show a clearly negative and significant correlation between *Property rights* and TFP in our sample. Aidt (2016) and Treismann (2000) illuminate the issue by suggesting that looser regulation on starting and engaging business may reduce corruption, thus favoring the formal economy. Ulubasoglu and Doucouliagos (2004) also report a negative relationship between property rights and the growth of capital and labor inputs.

The variable *Improved sanitation* measures access to those facilities as a percentage of rural population. Taking it as a proxy for health, the results clearly confirm the alleged positive relationship between health and TFP. In addition, the GDP share of health expenditures was tried in the regressions as another explanatory variable. The estimated coefficients of the two health variables were then positive, but their statistical significance was poor. A possible explanation for the ambiguity is that in many developing countries, malaria and other waterborne diseases cause high health service expenditures but make the productivity of labor erode (Cole and Neumayer 2006).

In both Tables 2 and 3, the variables *Energy use*, *Electricity*, and *Transport service* are proxy variables for technical infrastructure. Quite surprisingly, they do not seem to be strong explanators of TFP. The estimated coefficients are overall positive but seldom statistically significant. According to Aschauer (1989), physical infrastructures clearly foster the productivity of capital, but Lachler and Aschauer (1998) point out that among developing countries, the positive effect arises only if the infrastructures are financed by means other than public debt. This may explain the weakness of our results, recalling that over 2/3 of our sample countries are developing ones.

The findings in Table 3 also confirm the common reasoning regarding the pernicious effect of inflation on economic performance (Romer 1993; Lane 1997). The estimated coefficients of *Inflation* are constantly negative and highly statistically significant, which unequivocally points to a decline in TFP. Note that corruption has also been identified as one link in the causal chain that leads to high inflation (Al-Marhubi 2000).

Government spending is also found to reduce TFP with 5% statistical significance. This is reasonable, since public consumption and transfer payments included in the variable are partly unproductive and may even cause detrimental budget deficits and public debt. The penetration of corruption into public budgets is also noteworthy. Mauro (1998) finds that corruption distorts the allocation of resources between sectoral budgets, and Delavallade (2006) claims that corruption favors sectors such as military, energy, and order at the expense of health, education, and social protection. Dzhumashev (2014) reports a positive correlation between corruption and social security expenditures.

Finally, Table 4 presents the System GMM results with *Tax burden* as a main explanatory variable. The related variable *Government spending* is skipped because of statistical insignificance, which may be due to effective spending in developed countries. Now, the spin-off hypothesis stated in Section 1 is tested by including the interaction term that indicates the interrelation between corruption and tax burden, namely *Corruption* × *Tax burden* in Models 13, 15–17, and 19. For a simple robustness test, also the interaction

between corruption and inflation, namely *Corruption* \times *Inflation* is included in Models 16 and 19.

Table 4. System GMM estimations of TFP including the influences of tax burden.

	Dependent Variable: Total Factor Productivity (TFP)						
	System GMM						
	13	14	15	16	17	18	19
<i>Corruption</i>	−0.023 *** (0.004)	−0.011 *** (0.002)	−0.015 *** (0.002)	−0.022 *** (0.007)	−0.021 *** (0.006)	−0.017 *** (0.002)	−0.023 *** (0.006)
<i>TFP</i> _{<i>t</i>−1}	0.945 *** (0.012)	0.956 *** (0.013)	0.964 *** (0.014)	0.971 *** (0.011)	0.977 *** (0.013)	0.963 *** (0.016)	0.960 *** (0.014)
<i>Openness</i>		0.063 *** (0.019)	0.050 *** (0.017)	0.089 *** (0.024)	0.081 *** (0.025)	0.062 ** (0.025)	0.071 ** (0.021)
<i>Property rights</i>		−0.081 *** (0.015)	−0.094 ** (0.018)	−0.068 ** (0.017)	−0.047 ** (0.020)	−0.084 *** (0.019)	−0.082 ** (0.017)
<i>Improved sanitation</i>		0.052 ** (0.014)	0.041 ** (0.016)	0.037 * (0.018)	0.038 ** (0.016)	0.039 ** (0.016)	0.040 ** (0.017)
<i>Energy use</i>		0.011 * (0.012)	0.012 (0.013)				
<i>Inflation</i>		−0.061 *** (0.033)	−0.042 *** (0.032)	−0.051 ** (0.031)	−0.063 *** (0.043)	−0.075 ** (0.057)	−0.082 ** (0.061)
<i>Tax burden</i>	−0.033 *** (0.063)	−0.031 *** (0.06)	−0.018 *** (0.066)	−0.024 *** (0.072)	−0.023 *** (0.081)	−0.037 *** (0.059)	−0.022 *** (0.061)
<i>Corruption</i> \times <i>Tax burden</i>	0.042 *** (0.017)		0.023 *** (0.006)	0.028 *** (0.007)	0.025 *** (0.006)		0.027 ** (0.007)
<i>Corruption</i> \times <i>Inflation</i>				0.006 ** (0.005)			0.006 ** (0.004)
<i>Electricity</i>				0.007 (0.006)	0.006 * (0.003)		
<i>Transport service</i>						0.047 * (0.019)	0.033 ** (0.020)
Sargan test (<i>p</i> -value)	0.250	0.536	0.438	0.594	0.749	0.631	0.537
AR (1) (<i>p</i> -value)	0.002	0.005	0.005	0.018	0.019	0.010	0.001
AR (2) (<i>p</i> -value)	0.427	0.841	0.980	0.958	0.974	0.850	0.964
Wald test, coeff. (<i>p</i> -value)	0.000	0.000	0.000	0.000	0.000	0.000	0.000
Wald test, dumm. (<i>p</i> -value)	0.000	0.000	0.000	0.000	0.000	0.000	0.000
Number of observations	1980	1877	1873	1602	1602	1828	1824

Notes. Robust standard errors of System GMM estimator are in parentheses. The superscripts ***, **, and * represent statistical significance at 0.1%, 1%, and 5% levels, respectively.

In Table 4, the basic estimation (Model 13) means that TFP is controlled for *TFP*_{*t*−1}, *Corruption*, *Tax burden*, and *Corruption* \times *Tax burden*. The estimation results indicate that the main effects of *Corruption* and *Tax burden* are clearly negative and statistically highly significant. So, both corruption and tax burden deteriorate total factor productivity as expected. The effect described by the interaction term *Corruption* \times *Tax burden* is clearly positive and statistically significant at the 0.1% level, too.

The combination of the negative main effect of corruption on TFP and the positive effect from the interaction of corruption and tax burden implies that the negative effect of corruption on TFP gets smaller when the tax burden is increased (and vice versa). In other words, an increase in tax burden alleviates the detrimental impact of corruption on total factor productivity.

A comparison of Model 3 in Table 2 and Model 13 in Table 4 supports the above argument. The difference between them is that Model 3 ignores the influence of tax burden on TFP, whereas Model 13 takes it into account. In Model 13, the combined effect of corruption (the net sum of the estimated coefficient −0.042 of *Corruption* and that of *Corruption* \times *Tax burden* 0.023) on TFP is positive: 0.019. Thus, the negative effect of corruption from Model 13 is considerably different from that in Model 3.

In Table 4, Models 14–19 include several alternative determinants of TFP among the regressors. The interaction term *Corruption* \times *Inflation* is only slightly positive and statistically significant at the 1% level. So, it can be ignored, and a closer look on the estimated coefficients of the interaction term *Corruption* \times *Tax burden* is more relevant. Without the term (Models 14 and 18), the effects of both corruption and tax burden on productivity are negative and statistically significant. When the interaction term is included (Models 15–17 and 19), the signs and significances of the main effects stay unchanged, and the coefficient of *Corruption* \times *Tax burden* is positive and statistically significant at the 0.1% level. The combined effect of corruption is clearly positive in all estimations.

To sum up, a comparison of the model-wise results in Table 4 confirm the robustness of the finding that both corruption and tax burden reduce total factor productivity, and that an increase in tax burden mitigates the negative impact of corruption. The summary aligns with Olson et al. (2000), who report that corruption tends to hinder productivity growth. In fact, the results are not too far from those of Méon and Weill (2010), which suggests that corruption can improve efficiency: Reading the effect of the interaction term the other way round says that an increase in corruption mitigates the negative effect of tax burden on total factor productivity. Broadly interpreted, this also matches with the argument by Wu and Schneider (2019) about positive impacts from the informal sector to the formal sector.

4. Conclusions

The paper presented an empirical investigation of the effects of corruption and tax burden on total factor productivity, which is also known as the Solow residual. Panel data from 90 countries worldwide over 1996–2014 were used, and the System GMM estimation method was applied to derive the main findings.

The estimation results unambiguously showed that both corruption and tax burden have a negative effect on total factor productivity (TFP). However, when both the direct influence of corruption on TFP and its interaction with tax burden were considered, the negative effect of corruption on TFP was found to be smoothed by the positive interaction effect. The conclusion was that an increase in tax burden alleviates the detrimental impacts of corruption on productivity. The analyses included a wide set of control variables and estimation settings to validate the robustness of the main results.

The general message of the paper is that since total factor productivity is an important element in economic growth and both corruption and tax burden hamper productivity, it is important to understand the proposed link between corruption and tax burden. They are positively correlated, but when high tax burden and corruption appear together, corruption gets less harmful. The intuition is that in a worldwide panel context, a high tax burden is usually attached to developed countries with established institutions, effective social and technical infrastructures, and a small black market. It should also imply active fight against the meanest forms of corruption.

The paper yielded short-term results, but they also have some long-term implications. The policy recommendation for developing countries seems straightforward: To alleviate the negative effects of persistent corruption on productivity and consequently on economic progress, increase the (balanced) public budget and use the tax revenues productively. However, this is a “double-edged sword” in the hands of developing countries, where corruption often nests in bad governance and shadow economy. Optimal taxation and the proper use of tax revenues constitute an effective instrument in building social infrastructure to improve productivity and defeat corruption, but any malpractice is likely to exacerbate corruption and endanger economic development.

In terms of global strategies in the fight against corruption, the findings of the paper suggest that upgrading the efficiency of the formal sector in developing countries, including both their economy and governance, is possibly the best way to tackle corruption. Defeating corruption as such is an endless mole-whacking game unless more profound reforms are made. Practical experience tells that interventions by international organizations are most successful in this respect.

As usual, there are some caveats in the study. First, the time span 1996–2014 is quite short as a time series dimension. Second, the Corruption Perceptions Index provided by Transparency International is truly comparable over time only from 2012. The use of other measures of corruption such as the World Bank Control of Corruption data might solve both issues. We leave that to future studies.

For other future studies, long-term country-wise investigations of the influences of corruption and taxation on physical and human capital and the Solow residual could reveal the dynamic consequences of corruption. Another path to explore would be to dig deeper into possible channels from corruption to economic performance and development. These enquiries should give important information also for anti-corruption efforts in developing countries.

Author Contributions: Conceptualization, K.K. and H.L.; methodology, K.K.; validation, H.L.; formal analysis, K.K.; investigation, K.K. and H.L.; data duration, K.K.; writing, K.K. and H.L. All authors have read and agreed to the published version of the manuscript.

Funding: The research received no external funding.

Institutional Review Board Statement: Not applicable.

Informed Consent Statement: Not applicable.

Data Availability Statement: Not applicable.

Conflicts of Interest: The authors declare no conflict of interest.

Appendix A

Table A1. List of countries in the sample.

Argentina, Australia, Austria, Bahrain, Barbados, Belgium, Benin, Bolivia, Botswana, Brazil, Bulgaria, Burundi, Cameroon, Canada, Central African Republic, Chile, China, Colombia, Costa Rica, Cote d'Ivoire, Cyprus, Denmark, Dominican Republic, Egypt, Finland, France, Gabon, Germany, Greece, Guatemala, Honduras, Hungary, Iceland, India, Indonesia, Iran, Ireland, Israel, Italy, Jamaica, Japan, Jordan, Kenya, Kuwait, Laos, Lesotho, Luxembourg, Malaysia, Malta, Mauritania, Mauritius, Mexico, Mongolia, Morocco, Mozambique, Namibia, Netherlands, New Zealand, Niger, Norway, Panama, Paraguay, Peru, Philippines, Poland, Portugal, Qatar, Republic of Korea, Romania, Rwanda, Saudi Arabia, Senegal, Sierra Leone, Singapore, South Africa, Spain, Sri Lanka, Swaziland, Sweden, Switzerland, Tanzania, Thailand, Togo, Trinidad and Tobago, Tunisia, Turkey, United Kingdom, United States, Uruguay, Venezuela.

Table A2. Unit-root test, Augmented Dickey–Fuller (AD–F) and Phillips–Perron (P–P) methods.

Variable	Augmented Dickey–Fuller		Phillips–Perron	
	DF Statistics	<i>p</i> -Value	DF Statistics	<i>p</i> -Value
<i>TFP</i>	−9.846	<0.01	−31.332	0.01
<i>Corruption</i>	−10.778	<0.01	−32.583	0.01
<i>Openness</i>	−10.035	<0.01	−31.959	0.01
<i>Property rights</i>	−9.953	<0.01	−31.645	0.01
<i>Inflation</i>	−10.821	<0.01	−31.664	0.01
<i>Tax burden</i>	−10.104	<0.01	−30.934	0.01
<i>Government spending</i>	−9.720	<0.01	−29.000	0.01
<i>Health expenditure</i>	−10.977	<0.01	−32.182	0.01
<i>Improved sanitation</i>	−10.886	<0.01	−31.853	0.01
<i>Electricity</i>	−9.306	<0.01	−30.809	0.01
<i>Transport service</i>	−10.045	<0.01	−32.198	0.01
<i>Energy use</i>	−9.249	<0.01	−30.904	0.01

Notes. For both methods, the null hypothesis is that the series contain a unit root, against an alternative hypothesis that at least one panel is stationary. The null hypothesis is rejected under the condition that the *p*-value associated with DF statistics is smaller than the critical value 0.05. In both cases, all series are rigorously stationary.

Table A3. First-stage regressions, $Corruption_{i,t}$ is the dependent variable.

	OLS 1	OLS 2	OLS 3	OLS 4	Fixed Effects 1	Fixed Effects 2
Constant	3.054 *** (0.047)	4.162 *** (0.021)	2.742 *** (0.053)	2.732 *** (0.014)		
$Corruption_{1996}$	-0.141 *** (0.036)			-0.049 *** (0.028)		
$Government\ spending_t$		0.218 *** (0.091)		0.205 *** (0.133)	0.073 *** (0.025)	0.061 *** (0.032)
$Military\ expenditure$			0.002 *** (0.073)	0.001 ** (0.033)	0.003 ** (0.042)	0.001 ** (0.040)
$Corruption_{t-1}$					0.012 ** (0.037)	0.0014 ** (0.021)
$Corruption_{t-2}$					-0.004 * (0.001)	-0.0034 * (0.001)
$Inflation_t$						0.013 *** (0.035)
$Property\ right_t$						0.044 ** (0.018)
$Health\ expenditure$						0.014 * (0.364)
$Improved\ sanitation_t$						-0.053 (0.016)
Adjusted-R ²	0.0381	0.1271	0.023	0.1506	0.1073	0.2454
Number of obs.	1088	1088	1088	1088	1088	1088
Fisher test (p-value)	8.44×10^{-8} ***	$<2.2 \times 10^{-16}$ ***	$<2.2 \times 10^{-16}$ ***	$<2.2 \times 10^{-16}$ ***		

Notes. Regressions are based on OLS and Fixed Effects estimators, and ***, **, * symbolize statistical significance at 0.1%, 1% and 5% levels, respectively.

Table A4. Exogenous test regressions, $TFP_{i,t}$ as the dependent variable.

	OLS 5	OLS 6	OLS 7	OLS 8	Fixed Effects 3
Constant	2.547 *** (0.028)	2.830 *** (0.037)	3.135 *** (0.015)	2.652 *** (0.027)	
Corruption	-0.392 *** (0.026)	-0.149 *** (0.046)	-0.272 *** (0.183)	-0.248 *** (0.082)	-0.149 *** (0.053)
Tax Burden	-0.164 *** (0.015)	-0.103 *** (0.014)	-0.117 *** (0.011)	-0.209 *** (0.013)	-0.018 *** (0.035)
Residual (OLS 1)	0.043 *** (0.061)				
Residual (OLS 2)		0.162 *** (0.038)			
Residual (OLS 3)			0.217 *** (0.051)		
Residual (OLS 4)				0.251 *** (0.028)	
Residual (Fixed effects 1)					0.169 ** (0.009)
Adjusted-R ²	0.3312	0.4607	0.5028	0.6376	0.3401
Number of obs.	1088	1088	1088	1088	1088
Wu-Hausman (p-value)	2.16×10^{-16} ***	2.0×10^{-12}	$<2.2 \times 10^{-16}$ ***	$<2.2 \times 10^{-16}$ ***	$<2.2 \times 10^{-16}$ ***
Conclusion	IV	IV	IV	IV	IV

Notes. Regressions are based on OLS and Fixed Effects estimators, and ***, **, and * symbolize statistical significance at 0.1%, 1%, and 5% levels, respectively.

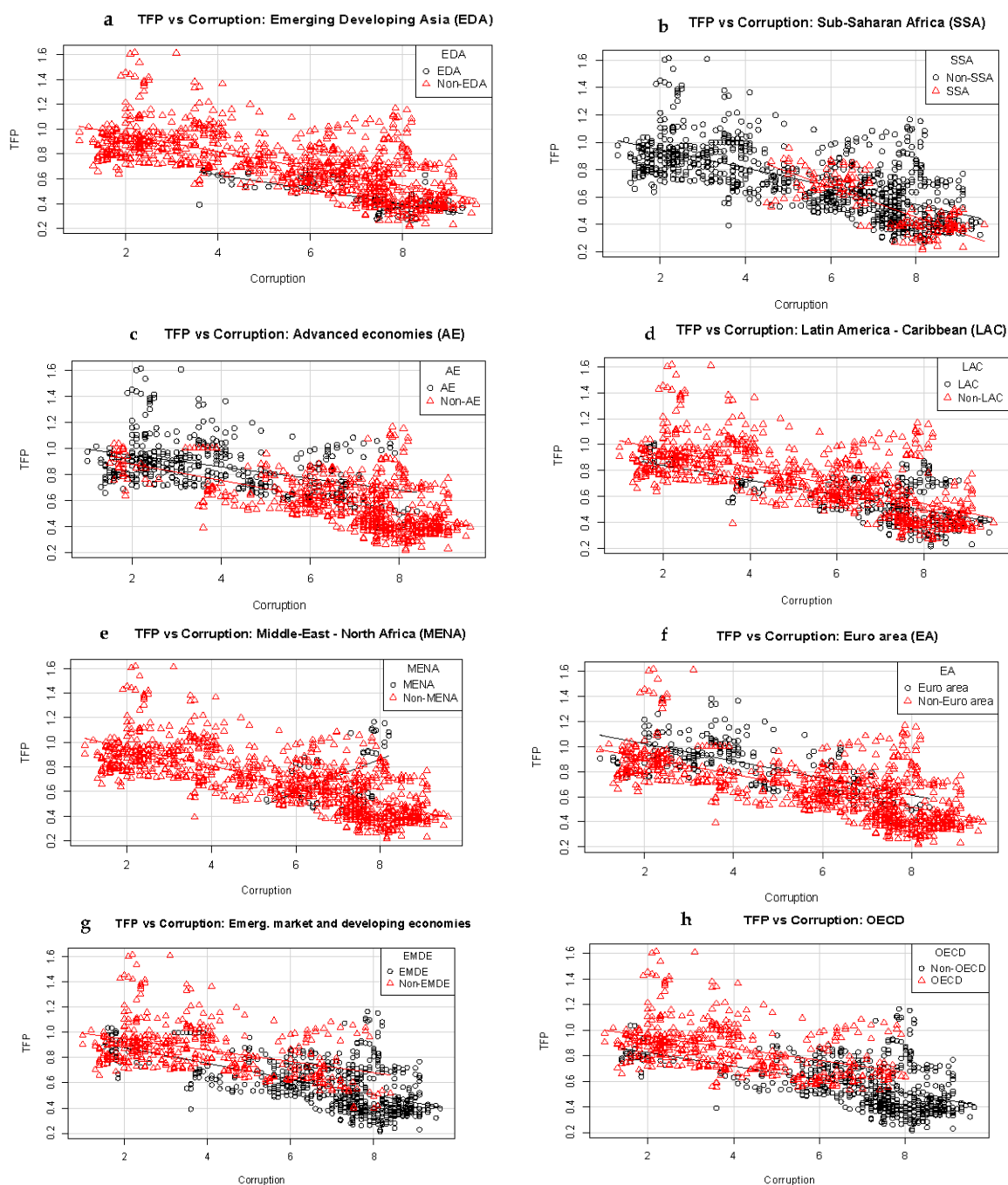


Figure A1. Regional relationships between TFP and corruption.

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