

Collaborative Research Project - Assignment 3

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```
## [1] "/Users/Gabriel/Desktop/Third-Assignment"

## 'data.frame':   1172 obs. of  10 variables:
## $ X          : int  1 2 3 4 5 6 7 8 9 10 ...
## $ iso2c       : Factor w/ 169 levels "AE","AF","AL",...: 1 1 1 1 1 1 1 2 2 2 ...
## $ year        : int  2005 2006 2007 2008 2009 2010 2011 2005 2006 2007 ...
## $ country     : Factor w/ 170 levels "Afghanistan",...: 161 161 161 161 161 161 161 1 1 1 ...
## $ corrupst    : num  1.069 0.947 1.077 1.131 0.955 ...
## $ gasrents    : num  6.18 5.01 4.42 5.29 2.97 ...
## $ gdppc       : num  40299 38363 34060 30613 25977 ...
## $ oilrents    : num  24.4 24.9 23.3 27.1 17.7 ...
## $ totrents    : num  30.5 29.9 27.8 32.4 20.6 ...
## $ unemp       : num  3.1 3.3 3.4 4 4.2 ...
```

Revisiting the natural resource curse: does the relationship between natural resource rents and corruption persist into the 21st Century?

1. Introduction

Our collaborative research project seeks to provide an answer to the following question: **How significant is the relationship between a country's degree of dependence on natural resource rents and its corruption level?** In the 1990s and early 2000s, several scholars (see among others [@leite1999], [@sala2003], [@pendergast2007] and [@shaxson2007]) identified this relationship and developed theories about the interplay of governance and the fiscal system to account for the "Natural Resource Curse" first identified by Sachs and Werner (-@sachs1995). However, in light of the rapid economic development of many resource-dependent developing countries since the turn of the century (notably in sub-Saharan Africa, which is now the second fastest growing region in the world after Asia-Pacific @africa2014), it is worth revisiting this relationship. Could it be that in the current century the relationship between reliance on natural resource rents and corruption has been broken?. This is, in a nutshell, our motivation for this project.

2. Data preparation

2.1 General plan We seek to collect and analyze data on corruption and natural resource rents for all countries in the world for the last 20 years, because the existing empirical literature covers the previous years. As our research question suggests, we plan to use a *Corruption* indicator as our dependent variable and measures of dependence on natural resource rents as our explanatory variables of interest. We have both *Total Natural Resource Rents* and *Oil Rents*. We want to examine oil separately because there is significant empirical evidence that oil rents have a particularly detrimental effect on governance (see Shaxson -@shaxson and Ross -@ross2012oil). We also want to include control variables such as *GDP/GNI per capita* and *Unemployment*. These variables are included in similar empirical studies. Please see our [Assignment 2](#) for more details on the variables to be included in our analysis, their conceptual definition and their operationalization.

2.1 Data sources The data necessary for our empirical analysis is located in two sources, or rather two different research projects from the World Bank:

- The [World Governance Indicators \(WGI\)](#) of the World Bank, defined as “Aggregate and individual governance indicators for 215 countries and territories over the period 1996–2014, for six dimensions of governance” (@WDI).
- The [World Development Indicators \(WDI\)](#) also from the World Bank. This is “The primary World Bank collection of development indicators, compiled from officially-recognized international sources. It presents the most current and accurate global development data available, and includes national, regional and global estimates” (@WDI).

We need to collect data from these two sources and merge it.

In [Assignment 2](#) we also considered including in our analysis an indicator of prevalence of conflict, using data from the [UCDP/PRIO Armed Conflict Dataset](#). However, upon close inspection of the data we realized that it was not useful since it does not include information on the magnitude of conflicts -it is rather a list of all conflicts, the conflicting parties involved and the duration of the conflict.

2.2 Data gathering

2.2.1 Data from the World Development Indicators We gathered data directly from the **World Development Indicators (WDI)** using the [WDI Package for R](#). This package enables us to automatically download WDI indicators using the World Bank’s APIs. The following five variables were downloaded from the **WDI**:

- **Total natural resources rents (% of GDP)** - This is “the sum of oil rents, natural gas rents, coal rents (hard and soft), mineral rents, and forest rents” (@WGI). World Bank code: NY.GDP.TOTR.RT.ZS
- **Oil rents (% of GDP)** - “Oil rents are the difference between the value of oil production at world prices and total costs of production” (@WGI). World Bank code: NY.GDP.PETR.RT.ZS
- **Natural gas rents (% of GDP)** - “Natural gas rents are the difference between the value of natural gas production at world prices and total costs of production” (@WGI). World Bank code: NY.GDP.NGAS.RT.ZS
- **GDP per capita (constant 2005 US\$)** “GDP per capita is gross domestic product divided by midyear population. GDP is the sum of gross value added by all resident producers in the economy plus any product taxes and minus any subsidies not included in the value of the products. It is calculated without making deductions for depreciation of fabricated assets or for depletion and degradation of natural resources. Data are in constant 2005 U.S. dollars” (@WDI). World Bank code NY.GDP.PCAP.KD.
- **Unemployment, total (% of total labor force) (modeled ILO estimate)** “Unemployment refers to the share of the labor force that is without work but available for and seeking employment” (@WGI). World Bank code: SL.UEM.TOTL.ZS

We want to be create indicators to run the following model:

$$Corruption_{it}^* = \beta_1 NaturalResourceRents_{it}^* + \beta_2 GDPpercapita_{it}^* + \beta_3 Unemployment_{it}^* + \epsilon$$

- For the **World Governance Indicators** there was no API. Hence, the data was automatically downloaded as a .XLSX file and then imported into R.

2.3 Descriptive Statistics

Before conducting our analysis we want to characterize our data, in particular our dependent variable *Corruption* and our interest explanatory variables *Total Natural Resource Rents* and *Oil Rents*.

2.4 Inferencial Statistics

Corruption over time

Regressions

Following the model

$$Corruption_i = \beta_0 + \beta_1 NaturalResourceRents_i + \beta_2 GDPpercapita_i + \beta_3 Unemployment_i + \epsilon_i$$

The model one is a regression with the control variables specified.

```
##                2.5 %          97.5 %
## (Intercept) -0.5588637652 -4.295447e-01
## totrents    -0.0152637697 -1.183046e-02
## gdppc       0.0000492183  5.310915e-05
## unemp       0.0027087866  1.272930e-02

##
## Call:
## lm(formula = corrupest ~ totrents + gdppc + unemp, data = merged)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.95468 -0.32190 -0.02724  0.28425  1.76540
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -4.942e-01  3.296e-02 -14.996 < 2e-16 ***
## totrents    -1.355e-02  8.750e-04 -15.483 < 2e-16 ***
## gdppc       5.116e-05  9.916e-07  51.600 < 2e-16 ***
## unemp       7.719e-03  2.554e-03   3.023  0.00256 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5165 on 1168 degrees of freedom
## Multiple R-squared:  0.7367, Adjusted R-squared:  0.736
## F-statistic: 1089 on 3 and 1168 DF, p-value: < 2.2e-16
```

Heteroskedasticity

When testing for heteroskedasticity, ...

```
{r plot(model1, which = 1, main = "Residual vs. Fitted Plot")
```

 The graphical inspection suggests there is heteroskedastocity. For this reason, it is necessary

Pearson's correlation

$$Corruption_i = \beta_0 + \beta_1 NaturalResourceRents_i + \beta_2 GDPpercapita_i + \beta_3 ArmedConflict_i + \beta_4 Unemployment_i + \epsilon_i$$

The next step will be to run a Fixed Effects regression, which will allow us to get rid of the country-specific fixed effects (such that we will only be analyzing the variation in variables within one country across time). To run a Fixed Effects regression, we will subtract from each term its mean. Thus, the regression equation will be:

$$Corruption_{it}^* = \beta_1 NaturalResourceRents_{it}^* + \beta_2 GDPpercapita_{it}^* + \beta_3 ArmedConflict_{it}^* + \beta_4 Unemployment_{it}^* + \epsilon$$

The terms in the equation above are demeaned to eliminate the fixed effects, such that:

$$NaturalResourceRents_{it}^* = NaturalResourceRents_{it} - \overline{NaturalResourceRents_i}$$

$$GDPpercapita_{it}^* = GDPpercapita_{it} - \overline{GDPpercapita_i}$$

$$ArmedConflict_{it}^* = ArmedConflict_{it} - \overline{ArmedConflict_i}$$

$$Unemployment_{it}^* = Unemployment_{it} - \overline{Unemployment_i}$$

```
##          exp          wks          bluecol          ind          south
##  Min.      : 1.00    Min.      : 5.00    no :2036    Min.      :0.0000    no :2956
##  1st Qu.:11.00    1st Qu.:46.00    yes:2129    1st Qu.:0.0000    yes:1209
##  Median :18.00    Median :48.00                      Median :0.0000
##  Mean      :19.85    Mean      :46.81                      Mean      :0.3954
##  3rd Qu.:29.00    3rd Qu.:50.00                      3rd Qu.:1.0000
##  Max.      :51.00    Max.      :52.00                      Max.      :1.0000
##  smsa      married      sex          union          ed          black
##  no :1442    no : 773    female: 469    no :2649    Min.      : 4.00    no :3864
##  yes:2723    yes:3392    male :3696    yes:1516    1st Qu.:12.00    yes: 301
##                                     Median :12.00
##                                     Mean      :12.85
##                                     3rd Qu.:16.00
##                                     Max.      :17.00
##          lwage
##  Min.      :4.605
##  1st Qu.:6.395
##  Median :6.685
##  Mean      :6.676
##  3rd Qu.:6.953
##  Max.      :8.537

##
## Call:
## lm(formula = lwage ~ ed, data = Wages)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.92996 -0.26863  0.00931  0.28453  1.83076
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.838779   0.030997  188.37  <2e-16 ***
## ed           0.065204   0.002358   27.65  <2e-16 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4243 on 4163 degrees of freedom
## Multiple R-squared:  0.1552, Adjusted R-squared:  0.155
## F-statistic: 764.5 on 1 and 4163 DF,  p-value: < 2.2e-16

##                2.5 %      97.5 %
## (Intercept) 5.7780088 5.89954938
## ed          0.0605805 0.06982708

##
## <table style="text-align:center"><caption><strong>OLS regression of the Percentage of Wages Variation
## <tr><td colspan="3" style="border-bottom: 1px solid black"></td></tr><tr><td style="text-align:left">
## <tr><td></td><td colspan="2" style="border-bottom: 1px solid black"></td></tr>
## <tr><td style="text-align:left"></td><td colspan="2">lwage</td></tr>
## <tr><td style="text-align:left"></td><td>(1)</td><td>(2)</td></tr>
## <tr><td colspan="3" style="border-bottom: 1px solid black"></td></tr><tr><td style="text-align:left">
## <tr><td style="text-align:left"></td><td>(0.002)</td><td>(0.002)</td></tr>
## <tr><td style="text-align:left"></td><td></td><td></td></tr>
## <tr><td style="text-align:left">exp</td><td></td><td>0.01<sup>***</sup></td></tr>
## <tr><td style="text-align:left"></td><td></td><td>(0.001)</td></tr>
## <tr><td style="text-align:left"></td><td></td><td></td></tr>
## <tr><td style="text-align:left">Constant</td><td>5.84<sup>***</sup></td><td>5.44<sup>***</sup></td></tr>
## <tr><td style="text-align:left"></td><td>(0.03)</td><td>(0.03)</td></tr>
## <tr><td style="text-align:left"></td><td></td><td></td></tr>
## <tr><td colspan="3" style="border-bottom: 1px solid black"></td></tr><tr><td style="text-align:left">
## <tr><td style="text-align:left">R<sup>2</sup></td><td>0.16</td><td>0.25</td></tr>
## <tr><td style="text-align:left">Adjusted R<sup>2</sup></td><td>0.15</td><td>0.25</td></tr>
## <tr><td style="text-align:left">Residual Std. Error</td><td>0.42 (df = 4163)</td><td>0.40 (df = 4162)</td></tr>
## <tr><td style="text-align:left">F Statistic</td><td>764.53<sup>***</sup> (df = 1; 4163)</td><td>681.1</td></tr>
## <tr><td colspan="3" style="border-bottom: 1px solid black"></td></tr><tr><td style="text-align:left">
## </table>
```

```
# Create list of packages
PackagesUsed <- c("knitr", "ggplot2", "repmis")
# Load PackagesUsed and create .bib BibTeX file
# Note must have repmis package installed.
repmis::LoadandCite(PackagesUsed, file = "Packages.bib", install = FALSE)
```