

# Socioeconomic status and Democratization <South Africa, Botswana, and Kenya compared

## Abstract

## Introduction

The dictionary meaning of "Democracy" defined by Encyclopedia Britannica is "literally ruled by the people". Measuring democracy also has been contested that there still have ongoing debates on this subject. The research paper will proceed by presenting background researches about democracy in South Africa, Botswana, and Kenya.

## Background about Democratization and its brief history in Africa

Democratization, according to An Agenda for Democratization by Boutros Boutros-Ghali, the former Secretary-General of the United Nations, is a process of transition from authoritarian rule to democratic rule. From his On Democracy, Robert Dahl suggests three conditions that are essential to attain democratic institutions. Africa has also been the wave of political transitions from various types of dictatorships to more open and democratic systems.

## Research Question and Hypotheses

The current research aims to help our understanding of the democracy with regard to its concept structure and its impact on Sub-Saharan African countries, where continuously have been experiencing democratic progress and setbacks.

1. There is a significant and positive correlation between socioeconomic variables and democracy.
2. Among four socioeconomic variables - GDP, primary education enrollment, gender equality, and child mortality - the educational effect on democracy is stronger than any other variables.
3. The degree of impact of variables on democracy is consistent across selected African countries.

## Literature Review

With regard to measuring democracy, Robert J. Barro, for his paper Determinants of Democracy, used the Polity IV dataset. Further, Barro quotes Lipset's argument based on the Lipset hypothesis, which claims that increased education leads to higher levels of democracy. In the second literature of Democracy and Gender Equality by Caroline Beer, she contrasts the impact of gender equality on democracy. John M. Shandra et al. approaches child mortality from different theoretical perspectives. By taking political and economic factors into account, they find that child mortality is a significant determinant of democracy.

## Data collection and data cleaning

### Data collection

In order to achieve our research purpose, firstly we have to decide what kind of data is necessary. Our research hypotheses are following,

- 1. There is a significant and positive correlation between socioeconomic variables and democracy.

- 2. Among four socioeconomic variables - GDP, primary education enrollment, gender equality, and child mortality - the educational effect on democracy is stronger than any other variables.
- 3. The degree of impact of variables on democracy is consistent across selected African countries.

To test these hypothesis, we need measurement of democracy and socioeconomic variables. We used following variables.

Variable name	Detail	Source
polity4	measurement of democracy	the Center for systemic peace
gdppc	Gross Domestic Production Per Capita	World Bank
pe	Primary ed enrollment	United Nations
mr	Child Mortality under 5	United Nations
gi	gender inequality in labor market	United Nations

- Measurement of democracy We used polity4 as a measurement of democracy. Polity4 represents the degree of democratization. The democratization level being higher, the score also being high. We downloaded from this site. <http://www.systemicpeace.org/inscrdata.html>
- GDP per capita: We bring this data from World Bank database. GDP per capita represents the level of satisfaction in basic needs. The following URLs are WorldBank sites in which we downloaded the time-series data of GDP for each country. <http://data.worldbank.org/indicator/NY.GDP.PCAP.CD>, <http://data.worldbank.org/indicator/NY.GDP.PCAP.CD?locations=BW>, <http://data.worldbank.org/indicator/NY.GDP.PCAP.CD?locations=KE>
- Primary education enrollment: We used Primary education enrollment as one of representative of basic needs. The URL is United Nations site in which we downloaded the data. [http://data.un.org/Data.aspx?q=education&d=UNESCO&f=series%3aE\\_1](http://data.un.org/Data.aspx?q=education&d=UNESCO&f=series%3aE_1)
- Mortality under 5 years old: We used this unit as measurement of health among citizens. <http://data.un.org/Data.aspx?q=mortality&d=PopDiv&f=variableID%3a77>
- Gender Inequality in labor market: We used gender inequality in labor market to test whether there is another important factor other than basic needs which is suggested in modernization theory. We calculated this unit as follows,

$$GenderInequality = \frac{EmploymentRateAmongWomen}{EmploymentRateAmongMen}$$

You can find the data from ILO official site. [http://www.ilo.org/ilostat/faces/help\\_home/data\\_by\\_subject/subject-details/indicator-details-by-subject?subject=EMP&indicator=EMP\\_2EMP\\_SEX\\_\\_AGE\\_NB&datasetCode=YI&collectionCode=ILOEST&\\_afLoop=201714902217437#!%40%40%3Findicator%3DEMP\\_2EMP\\_SEX\\_\\_AGE\\_NB%26subject%3DEMP%26\\_afLoop%3D201714902217437%26datasetCode%3DYI%26collectionCode%3DIOEST%26\\_adf.ctrl-state%3D102r3mzd68\\_271](http://www.ilo.org/ilostat/faces/help_home/data_by_subject/subject-details/indicator-details-by-subject?subject=EMP&indicator=EMP_2EMP_SEX__AGE_NB&datasetCode=YI&collectionCode=ILOEST&_afLoop=201714902217437#!%40%40%3Findicator%3DEMP_2EMP_SEX__AGE_NB%26subject%3DEMP%26_afLoop%3D201714902217437%26datasetCode%3DYI%26collectionCode%3DIOEST%26_adf.ctrl-state%3D102r3mzd68_271)

## Data cleaning and importing into R

We downloaded each data as excel and csv file. Then, we substracted time series data for each variables and countries and put together into three csv files which represents each countries so that we can easily import data into R. So we have three csv files each represents SouthAfrica, Botswana, and Kenya. We will import it into R.

```
dfsa <- read.csv("SA.csv", header = TRUE, sep = ",")
dfbo <- read.csv("BTW.csv", header = TRUE, sep = ",")
dfkn <- read.csv("KNY.csv", header = TRUE, sep = ",")
dfpanel <- read.csv("panel.csv", header = TRUE, sep = ",")
```

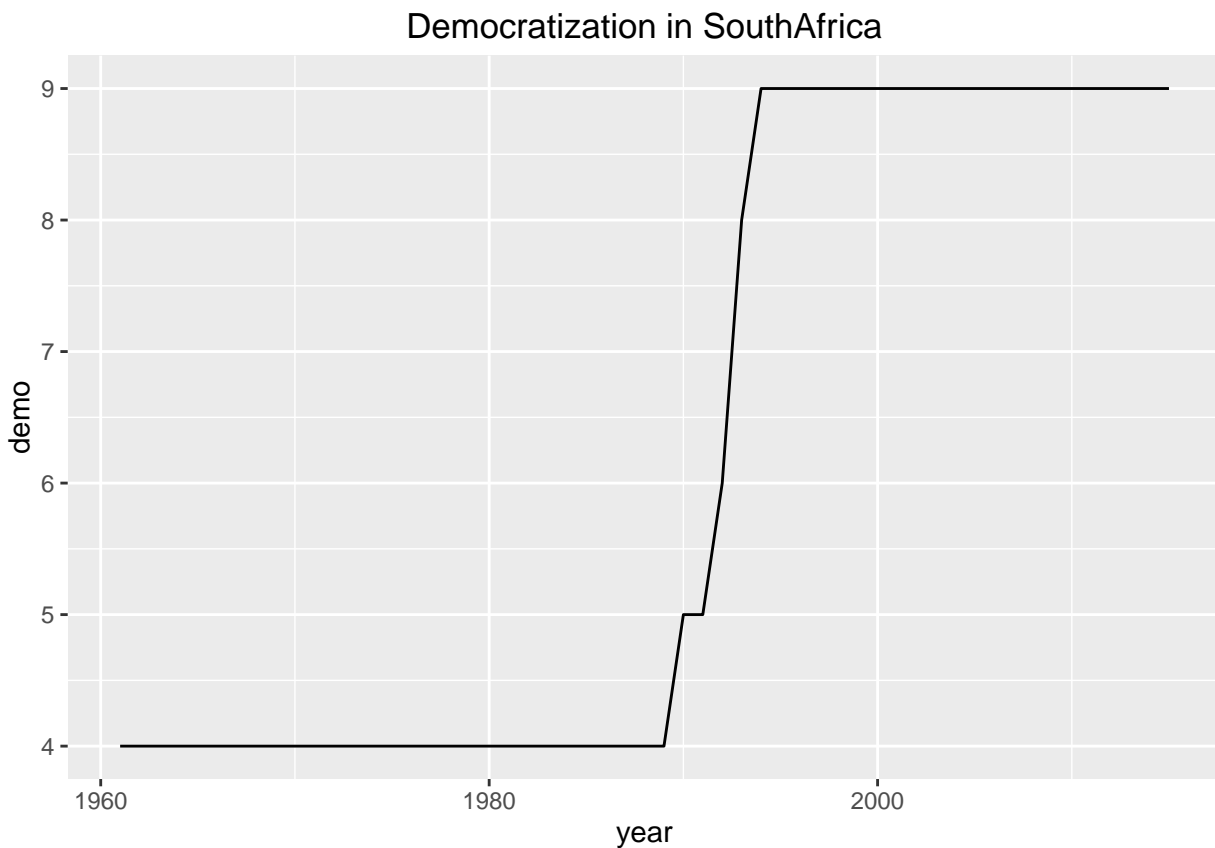
Now we succeeded to import all files.

## Descriptive Analyses

In this section, we will provide descriptive statistics of our variables.

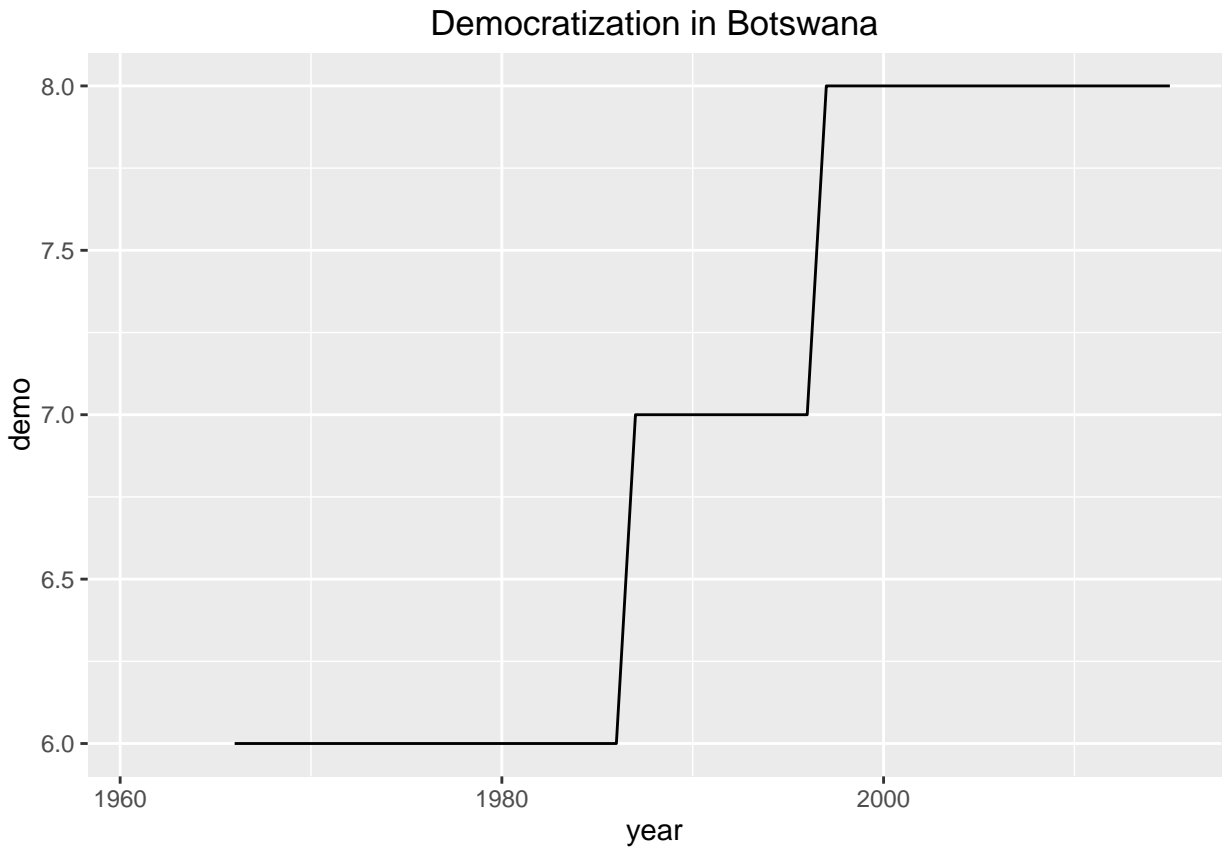
### Democratization

Following graphs are trend of democratization for each countries.

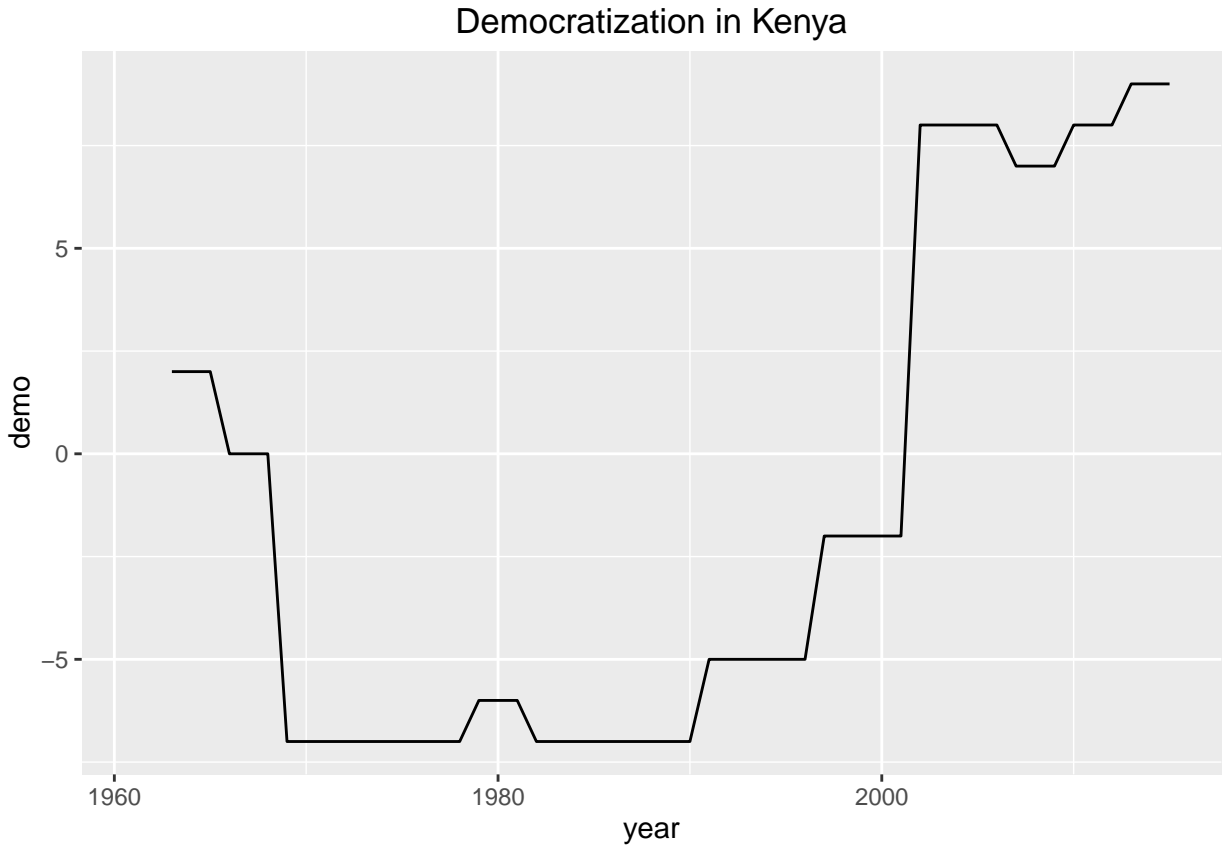


This is the trend of democratization in South Africa. As you can easily see, the level of democratization dramatically increased during 1990~1995. The contributor of this trend is the abolishment of Apartheid.

```
## Warning: Removed 5 rows containing missing values (geom_path).
```



## Warning: Removed 6 rows containing missing values (geom\_path).



As graphs showing, the level of democratization has been fluctuated. Sometimes democracy advances, and sometimes it setbacks. We will investigate the data to clarify what is the driver of these fluctuation.

## Multivariate Analyses

### OLS results

% Table created by stargazer v.5.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu  
 % Date and time: Thu, Dec 08, 2016 - 1:34:24 PM

### Pooled OLS

% Table created by stargazer v.5.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu  
 % Date and time: Thu, Dec 08, 2016 - 1:34:24 PM

Table 2: Regression results for each country

	<i>Dependent variable:</i>		
	SouthAfrica	demo Botswana	Kenya
	(1)	(2)	(3)
log(gdppc)	−2.78* (1.43)	1.86*** (0.56)	−6.37 (3.79)
log(pe)	0.81 (2.83)	−3.29 (2.31)	9.05 (8.68)
log(mr)	−11.16 (7.05)	2.84*** (0.84)	−34.05*** (9.39)
log(gi)	−43.66*** (13.70)	4.61 (2.85)	−147.18*** (38.41)
Constant	44.40 (73.50)	23.80 (22.73)	21.71 (157.22)
Observations	17	21	19
R <sup>2</sup>	0.93	0.85	0.92
Adjusted R <sup>2</sup>	0.90	0.82	0.90
Residual Std. Error	0.72 (df = 12)	0.28 (df = 16)	1.94 (df = 14)
F Statistic	38.99*** (df = 4; 12)	23.04*** (df = 4; 16)	42.66*** (df = 4; 14)

*Note:*

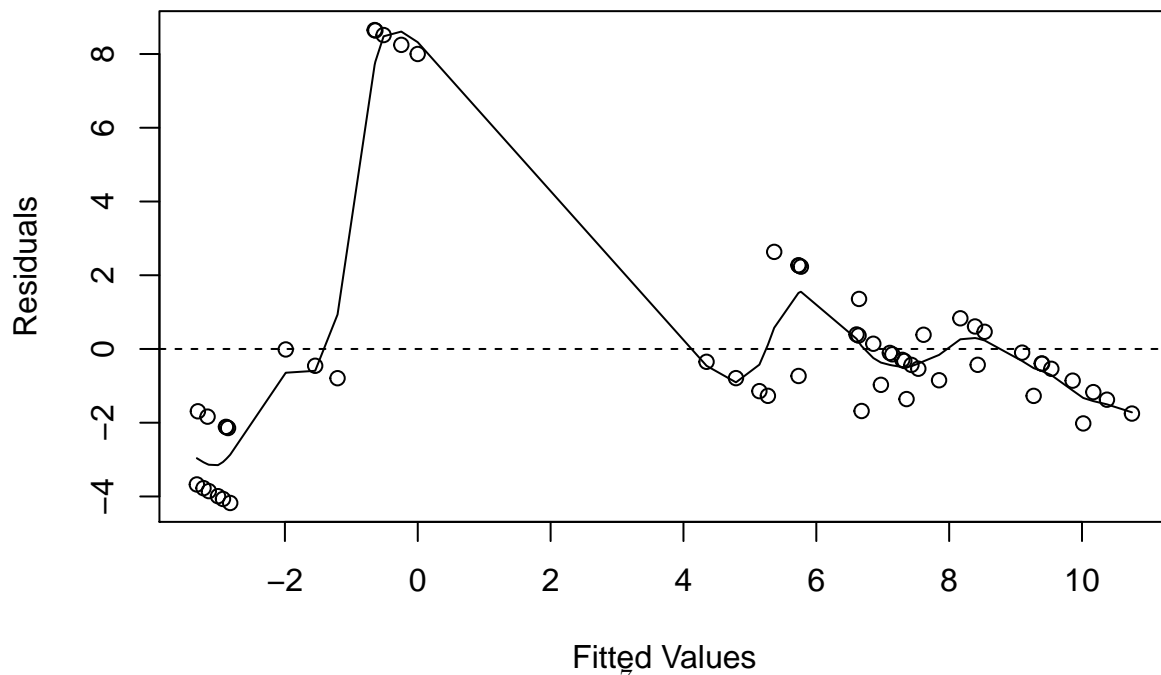
\*p&lt;0.1; \*\*p&lt;0.05; \*\*\*p&lt;0.01

Table 3: Pooled OLS

	<i>Dependent variable:</i>
	demo
$\log(\text{gdppc})$	-0.64 (1.39)
$\log(\text{pe})$	-1.10** (0.46)
$\log(\text{mr})$	-6.89** (3.06)
$\log(\text{gi})$	-39.08*** (10.91)
Constant	41.87* (21.14)
Observations	57
$R^2$	0.72
Adjusted $R^2$	0.70
Residual Std. Error	3.16 (df = 52)
F Statistic	33.01*** (df = 4; 52)

Note: \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

residual vs fitted value



## Breusch-Pagan test

```
##
## studentized Breusch-Pagan test
##
## data: L4
## BP = 12.672, df = 4, p-value = 0.01299
```

## Fixed-Effect model

% Table created by stargazer v.5.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu  
 % Date and time: Thu, Dec 08, 2016 - 1:34:25 PM

Table 4: pooled OLS and fixed effects OLS

	<i>Dependent variable:</i>	
	demo	
	PooledOLS	FixedOLS
	(1)	(2)
log(gdppc)	−0.644 (1.393)	−1.496 (1.424)
log(pe)	−1.103** (0.456)	15.793*** (4.012)
log(mr)	−6.886** (3.062)	−12.087*** (3.176)
log(gi)	−39.075*** (10.914)	−28.940** (11.221)
Constant	41.873* (21.138)	
Observations	57	57
R <sup>2</sup>	0.717	0.610
Adjusted R <sup>2</sup>	0.696	0.563
F Statistic	33.013*** (df = 4; 52)	19.532*** (df = 4; 50)
<i>Note:</i>		
*p<0.1; **p<0.05; ***p<0.01		

constants:

```
fixef(fixed)
```

```
## Botswana Kenya South Africa
## -139.0599 -186.3148 -191.4030
```



## Do panel specific effects exist?

```
pFtest(fixed,pooled)
```

```
##  
## F test for individual effects  
##  
## data: demo ~ log(gdppc) + log(pe) + log(mr) + log(gi)  
## F = 17.075, df1 = 2, df2 = 50, p-value = 2.228e-06  
## alternative hypothesis: significant effects
```

## Breush-Pagan test

```
##  
## Lagrange Multiplier Test - (Breusch-Pagan) for unbalanced panels  
##  
## data: demo ~ log(gdppc) + log(pe) + log(mr) + log(gi)  
## chisq = 0.67568, df = 1, p-value = 0.4111  
## alternative hypothesis: significant effects
```

we cannot reject the null hypothesis. (residuals doesn't correlated with independent variables)

## Random-Effects OLS

```
% Table created by stargazer v.5.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu  
% Date and time: Thu, Dec 08, 2016 - 1:34:26 PM
```

## Hausman test

Hausman Test

```
data: demo ~ log(gdppc) + log(pe) + log(mr) + log(gi) chisq = 4.6666e-19, df = 4, p-value = 1 alternative  
hypothesis: one model is inconsistent
```

## results

```
% Table created by stargazer v.5.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu  
% Date and time: Thu, Dec 08, 2016 - 1:34:27 PM
```

Oneway (individual) effect Within Model

```
Call: plm(formula = demo ~ log(gdppc) + log(pe) + log(mr) + log(gi), data = dfpanel, model = "within",  
index = c("country", "year"))
```

Unbalanced Panel: n=3, T=17-21, N=57

Residuals : Min. -3.160 1st Qu. -1.560 Median -0.584 3rd Qu. 0.730 Max. 8.200

Table 5: random effects OLS

	<i>Dependent variable:</i>		
	demo		
	(1)	(2)	(3)
log(gdppc)	−0.644 (1.393)	−1.496 (1.424)	−1.496 (1.396)
log(pe)	−1.103** (0.456)	15.793*** (4.012)	15.793*** (3.934)
log(mr)	−6.886** (3.062)	−12.087*** (3.176)	−12.087*** (3.114)
log(gi)	−39.075*** (10.914)	−28.940** (11.221)	−28.940** (11.003)
Constant	41.873* (21.138)		−172.259 (997,305.100)
Observations	57	57	57
R <sup>2</sup>	0.717	0.610	0.610
Adjusted R <sup>2</sup>	0.696	0.563	0.580
F Statistic	33.013*** (df = 4; 52)	19.532*** (df = 4; 50)	20.314*** (df = 4; 52)

*Note:*

\*p&lt;0.1; \*\*p&lt;0.05; \*\*\*p&lt;0.01

Table 6: Regression results

	<i>Dependent variable:</i>		
	demo		
	(1)	(2)	(3)
log(gdppc)	−0.644 (1.393)	−1.496 (1.424)	−1.496 (1.396)
log(pe)	−1.103** (0.456)	15.793*** (4.012)	15.793*** (3.934)
log(mr)	−6.886** (3.062)	−12.087*** (3.176)	−12.087*** (3.114)
log(gi)	−39.075*** (10.914)	−28.940** (11.221)	−28.940** (11.003)
Constant	41.873* (21.138)		−172.259 (997,305.100)
Observations	57	57	57
R <sup>2</sup>	0.717	0.610	0.610
Adjusted R <sup>2</sup>	0.696	0.563	0.580
F Statistic	33.013*** (df = 4; 52)	19.532*** (df = 4; 50)	20.314*** (df = 4; 52)

*Note:*

\*p&lt;0.1; \*\*p&lt;0.05; \*\*\*p&lt;0.01

Coefficients : Estimate log(gdppc) -1.4964 log(pe) 15.7935 log(mr) -12.0870 log(gi) -28.9397 Std. Error log(gdppc) 1.4239 log(pe) 4.0120 log(mr) 3.1755 log(gi) 11.2214 t-value log(gdppc) -1.0509 log(pe) 3.9366 log(mr) -3.8063 log(gi) -2.5790 Pr(>|t|) log(gdppc) 0.2983644 log(pe) 0.0002560 log(mr) 0.0003859 log(gi) 0.0128989

log(gdppc)  
log(pe) **log(mr)** log(gi) \*

— Signif. codes:

0 ‘ ‘ **0.001** ’ ’ 0.01 ’ ’ 0.05 ‘ ‘ 0.1 ‘ ‘ 1

Total Sum of Squares: 788.23 Residual Sum of Squares: 307.59 R-Squared: 0.60977 Adj. R-Squared: 0.56294  
F-statistic: 19.5324 on 4 and 50 DF, p-value: 9.8562e-10

$$democratization = -1.496\log(gdppc) + 15.793\log(pe) - 12.087\log(mr) - 28.940\log(gi) + \alpha_i$$

where  $\alpha_i$  represents panel specific effects

## Conclusion

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