Financial Development and Educational Attainment: A Cross Country Comparison

 $Does\ education\ covary\ with\ financial\ development?\ An\ Instrumental\ variables\\ panel\ regression\ approach$

John Karuitha

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1 Background

In this analysis, I examine the relationship between financial development and education attainment. My premise is that people who attain higher education are better placed to enter the formal labor market and hence demand financial services such as bank accounts. Education also raises awareness even among people in the informal and semi-formal sectors to better manage and access finance from formal financial intermediaries (Allen et al., 2014). I examine financial development as the number of people (aged 15 years and above) that have an account with a financial intermediary. I capture education attainment using secondary school enrollment ratio to primary school enrollment ratio. I also include variables such as institutional quality, region, trade openness as controls as they are well known drivers of financial development (Klapper & Singer, 2015). I use years of compulsory education as the instrumental variable for education.

We estimate an equation of the form;

 $findev_{ij} = \alpha + \beta_1 Education_{ij} + \beta_2 Governance_{ij} + \beta_4 TradeOpenness_{ij} + \epsilon_{ij}$

- Financial Development refers to the extent that people can access and use affordable financial services. In our case, we proxy financial development as the percentage of people with bank accounts in a country.
- We proxy general level of education as high school (secondary school) turnover to primary school turnover.

Prior: We expect a positive relationship between education and financial development.

• Governance captures the quality of institutions in a country and how well these institutions improve the quality of life of the citizens. We use the first principle component of the Worldwide Governance Indicators to proxy governance or institutional quality.

Prior: We expect a positive relationship between governance and financial development.

• Trade Openness: The extent to which a country trades with the outside world. We proxy this using the ratio of imports and exports to GDP.

Prior: We expect a positive relationship between trade openness and financial development.

Instrumental variables: I use years of compulsory education in a country as the instrumental variable for education. Rationale: Years of compulsory education can be a good instrument for the education variable, because it is correlated with education levels but is not directly related to financial development. Similarly, I use the air freight of passengers and press freedom index as instruments for trade openness and governance, respectively.

2 Data

We source the data from the World Bank, World Development Indicators (WDI) and the Worldwide Governance Indicators (WGI) for the years 1998-2002. The data consists of 8875 observations of 11 variables. The data consists of the following variables (see Table 1).

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3 Data Exploration

I explore the data by creating summary tables and visualizations. First, I examine the distribution of the variables (see Figure 1).

Table 1: Variables Definition					
Variable	Definition				
time	Year the data collected, 1998-2022.				
country_name	Name of the country.				
country_code	ISO3c country code				
Continent	Continent of associated country.				
Accounts	% of people aged 15+ with an account				
Governance	KKM Governance Index, defined as the first				
Educ	principal component of the KKM indicators Educational attainment- ratio of high school				
Openess	turnover tp primary school turnover Trade openess, defined as the ratio of imports and				
compulsory_education	exports to GDP Number of years of compulsory education in country.				
Press freedom index	The data is collected through an online questionnaire sent to journalists, media lawyers,				
	researchers and other media specialists selected by Reporters without Borders (RSF) in the 180 countries covered by the Index.				
Air Transport, passengers carried	Passengers carried by air into and out of a country.				

Note:

 $Source: \ The \ World \ Bank, \ https://databank.worldbank.org/source/world-development-indicators$

¹ The last three variables serve as instruments

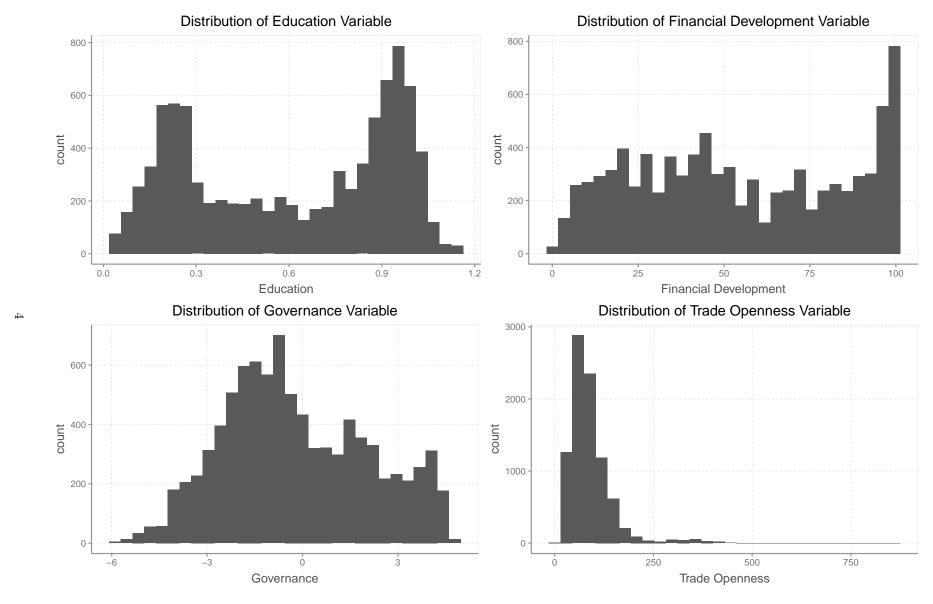


Figure 1: Distribution of the variables

Next, I look at the relationship between education and financial development by continent. We see a consistently strong positive relationship between the two variables across all continents.

Scatterplot of Financial Development against Education

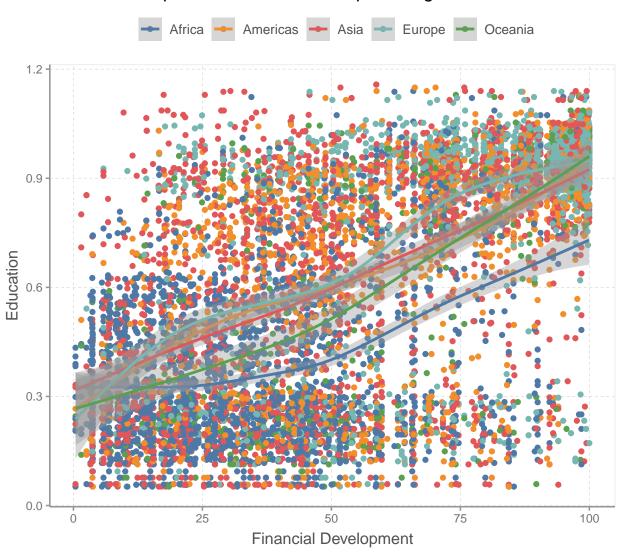


Figure 2: Scatterplot of Financial Development against Education

Education vs Financial Development

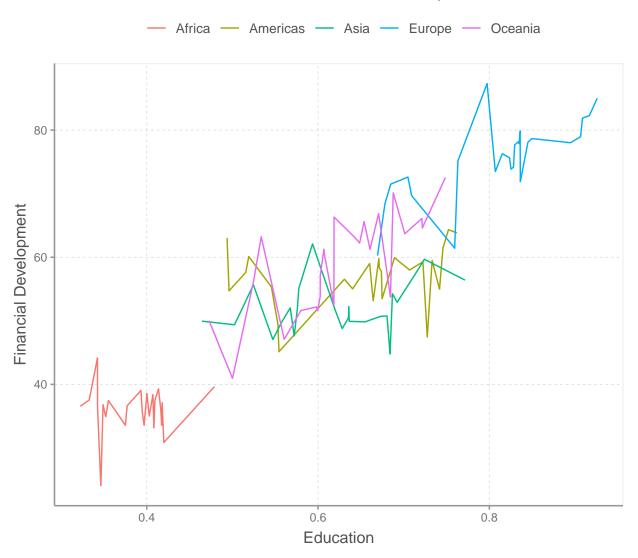


Figure 3: Education vs Financial Development

I then examine the average trends in financial development by continent over time. We see that Africa lies far below the other regions.

% People 15+ Years Old with Bank Accounts

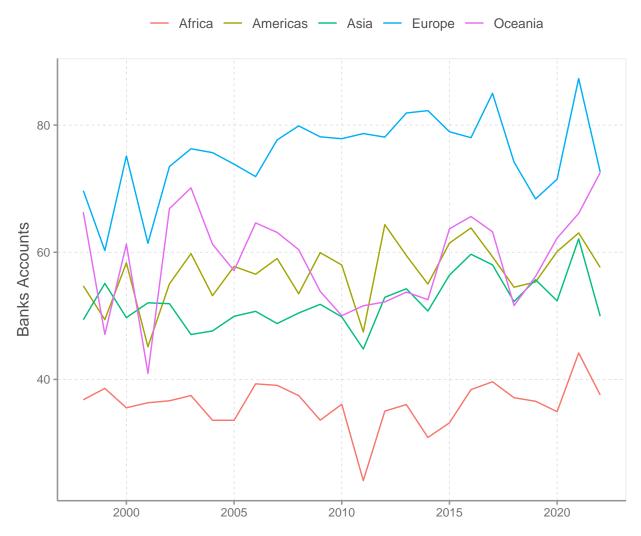


Figure 4: % People 15+ Years Old with Bank Accounts

The figure below reinforces the low level of financial development in Africa, with Europe exhibiting the highest median financial development.

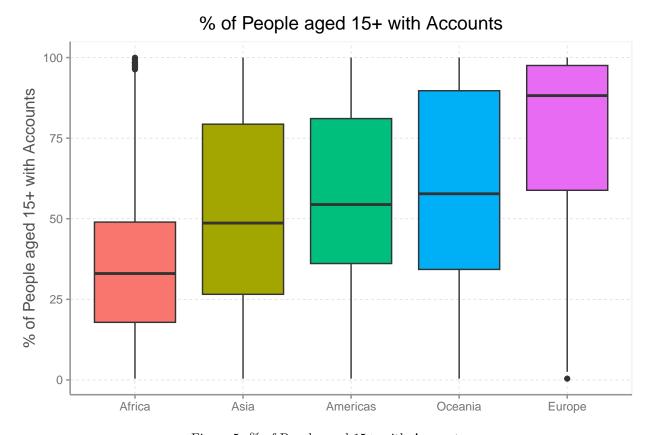


Figure 5: % of People aged 15+ with Accounts

I also look at the correlation between these financial development and the other variables. We see a very high correlation between financial development (accounts) and governance and education. Financial development has moderate relationship with trade openness. The high correlation between education and governance could bring about multicollinearity and cross-sectional dependence problems.

I also summarise the data below

Table 2: Summary Statistics

Table 2. Summary Statistics								
skim_type	Variable	Mean	SD	Min	Q1	Median	Q3	Max
numeric	accounts	54.529	30.195	0.400	28.570	50.760	83.560	100.00
numeric	governance	-0.198	2.288	-5.910	-1.896	-0.576	1.574	4.78
numeric	educ	0.608	0.326	0.052	0.268	0.665	0.921	1.16
numeric	openess	89.432	57.769	2.699	54.632	77.116	106.332	863.20
numeric	$\operatorname{compulsory}_{_}$	_edu &&47 n	2.291	0.000	8.000	9.000	10.000	17.00
numeric	air_freight	13.937	2.432	0.000	12.515	14.044	15.454	20.65
numeric	press_freedo	om 66.228	45.364	-10.000	28.000	63.440	90.000	180.00

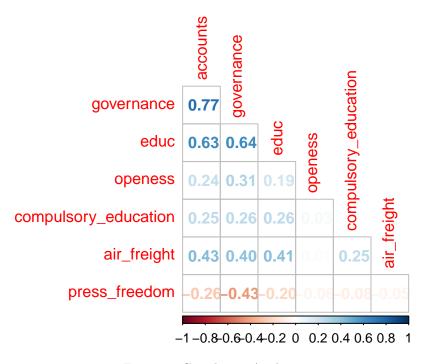


Figure 6: Correlation Analysis

4 Regression Analysis

I run panel regressions as follows;

- Fixed effects model.
- Random effects model.
- Pooled OLS

We estimate an equation of the form;

 $findev_{ij} = \alpha + \beta_1 Education_{ij} + \beta_2 Governance_{ij} + \beta_4 Trade_openness_{ij}$ The individual regression output follows.

4.1 The fixed effects model

4.1.1 Fixed effects without Instruments

Below is a summary of the fixed effects model without instruments.

```
## Twoways effects Within Model
## Call:
## plm(formula = accounts ~ educ + openess + governance, data = my_data,
       effect = "twoway", model = "within", index = c("continent",
##
##
           "time"))
##
## Balanced Panel: n = 5, T = 25, N = 8875
##
## Residuals:
     Min. 1st Qu.
                   Median 3rd Qu.
##
## -58.267 -12.367
                     0.404 11.173 82.002
##
```

```
## Coefficients:
##
        Estimate Std. Error t-value
                                     Pr(>|t|)
## educ
         21.76081 0.81517
                         ## openess
         -0.00196
                  0.00357
                         -0.55
                                        0.58
## governance 8.04936
                 ## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:
                   6210000
## Residual Sum of Squares: 2930000
## R-Squared:
             0.528
## Adj. R-Squared: 0.526
4.1.2 Fixed Effects with Instruments
```

Then the fixed effects model WITH instruments.

```
## Twoways effects Within Model
## Instrumental variable estimation
##
## Call:
## plm(formula = accounts ~ educ + openess + governance | compulsory_education +
      air_freight + press_freedom, data = my_data, effect = "twoway",
##
      model = "within", index = c("continent", "time"))
##
## Balanced Panel: n = 5, T = 25, N = 8875
##
## Residuals:
##
     Min. 1st Qu. Median 3rd Qu.
                  -4.49 14.62 296.60
##
   -76.94 -19.43
##
## Coefficients:
##
            Estimate Std. Error z-value Pr(>|z|)
## educ
             28.683 24.176
                                 1.19 0.23545
## openess
              -0.446
                         0.168
                                -2.66 0.00786 **
             7.167
                         1.883
                                 3.81 0.00014 ***
## governance
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Total Sum of Squares:
                         6210000
## Residual Sum of Squares: 8540000
## R-Squared:
                 0.122
## Adj. R-Squared: 0.118
```

Pooled OLS Model 4.2

4.2.1 Pooled OLS without Instruments

Here, I present the pooled OLS model without instruments

```
## Pooling Model
##
## Call:
## plm(formula = accounts ~ educ + openess + governance, data = my_data,
```

```
##
      effect = "twoway", model = "pooling", index = c("continent",
##
          "time"))
##
## Balanced Panel: n = 5, T = 25, N = 8875
##
## Residuals:
     Min. 1st Qu. Median 3rd Qu.
                                   Max.
## -59.218 -12.368
                  0.606 11.467 81.153
##
## Coefficients:
              Estimate Std. Error t-value
                                                   Pr(>|t|)
## (Intercept) 42.12694
                         0.61939
                                   68.01 < 0.0000000000000000 ***
## educ
              22.77823
                         0.78150
                                 0.00357
## openess
              0.00174
                                  0.49
                                                       0.63
             8.04644
                         0.11469
                                  70.16 < 0.0000000000000000 ***
## governance
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Total Sum of Squares:
                          8090000
## Residual Sum of Squares: 3040000
## R-Squared:
                 0.624
## Adj. R-Squared: 0.624
4.2.2 Pooled OLS with Instruments
Then look at the pooled OLS with instruments
## Pooling Model
## Instrumental variable estimation
##
     (Balestra-Varadharajan-Krishnakumar's transformation)
##
## Call:
## plm(formula = accounts ~ educ + openess + governance | compulsory_education +
      air_freight + press_freedom, data = my_data, effect = "twoway",
##
      model = "pooling", index = c("continent", "time"))
##
## Balanced Panel: n = 5, T = 25, N = 8875
##
## Residuals:
##
     Min. 1st Qu. Median 3rd Qu.
                                   Max.
##
   -91.44 -25.70
                  -6.43 17.65 434.32
##
## Coefficients:
              Estimate Std. Error z-value Pr(>|z|)
##
                          39.832
## (Intercept)
               76.334
                                   1.92
                                          0.0553 .
## educ
                57.623
                          12.494
                                   4.61 0.000004 ***
## openess
               -0.621
                           0.370
                                   -1.68
                                          0.0934 .
                6.293
                           2.099
                                   3.00 0.0027 **
## governance
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:
                          8090000
## Residual Sum of Squares: 14700000
```

R-Squared:

0.129

Combined results 4.3

I combine the regression output into one table below.

Table 3: Panel Regression Analysis

		Dependen	t variable:					
		Financial Development						
	Fixed Effects	Fixed Effects (Instruments)	Pooled OLS	Pooled OLS (Instruments)				
	(1)	(2)	(3)	(4)				
Education	21.800***	28.700	22.800***	57.600***				
	(0.815)	(24.200)	(0.781)	(12.500)				
Trade Openness	-0.002	-0.446^{***}	0.002	-0.621^{*}				
_	(0.004)	(0.168)	(0.004)	(0.370)				
Governance	8.050***	7.170***	8.050***	6.290***				
	(0.125)	(1.880)	(0.115)	(2.100)				
Constant			42.100***	76.300*				
			(0.619)	(39.800)				
Observations	8,875	8,875	8,875	8,875				
\mathbb{R}^2	0.528	0.122	0.624	0.129				
Adjusted \mathbb{R}^2	0.526	0.118	0.624	0.129				
F Statistic	$3,295.000^{***} (df = 3; 8843)$	989.000***	$4,917.000^{***} (df = 3; 8871)$	1,282.000***				

*p<0.1; **p<0.05; ***p<0.01

The output shows that the three variables have a statistically significant relationship with financial development. Holding governance and trade openness constant, a better education is positively associated with higher average financial development. Governance has a positive relationship with financial development, ceteris paribus. On the contrary, trade openness has a marginal negative relationship with financial development. The relationship is uniform across the models. The models without instruments have a better explanatory power than models with instruments. For instance, the fixed effects model with instruments have an R^2 of 52.6% compared to 11.8% for the model with instruments. Similarly, the pooled OLS without instruments has an adjusted R^2 of 62.4% compared to the pooled OLS with instruments that has an adjusted R^2 of 12.9%. But which of these models is better? Let us examine the AIC.

Table 4: AIC for Panel Models

Model	AIC
Pooled OLS with instruments Fixed effects with instruments	-45482 -43076
Pooled OLS Fixed effects	-38490 -38331

This analysis shows that the pooled OLS with instruments is the better model, followed by the fixed effects with instruments.

The models could be subject to omitted variable bias though it does fairly well with a \mathbb{R}^2 . Hence increased investments in raising institutional quality and education attainment should be encouraged.

Policy recommendation: Given that our analysis is between education and financial development, we recomend substantial investments in education to raise the levels of financial development in the long term. This recommendation is especially pertinent for regions with low levels of financial development like Africa.

5 Conclusion

In this analysis, we have examined the relationship between financial development proxied by the number of bank accounts and three variables that are postulated to drive financial development; governance, education, and trade openness. We find that Africa lags in financial development. All the three measures have a statistically significant relationship with financial development, and the models are all significant going by the F-statistic. Hence increased investments in raising education attainment should be encouraged holding institutional quality trade openness constant in order to raise financial development in the long term.

6 Acknowledgements

We used R version 4.3.1 (R Core Team, 2023) and the following R packages: Amelia v. 1.8.1 (Honaker, King, & Blackwell, 2011), artyfarty v. 0.0.1 (Smeets, 2023), correlationfunnel v. 0.2.0 (Dancho, 2020), corrplot v. 0.92 (Wei & Simko, 2021), countrycode v. 1.5.0 (Arel-Bundock, Enevoldsen, & Yetman, 2018), doParallel v. 1.0.17 (Corporation & Weston, 2022), ggthemes v. 5.0.0 (Arnold, 2023), janitor v. 2.2.0 (Firke, 2023), kableExtra v. 1.3.4.9000 (Zhu, 2023), knitr v. 1.45 (Xie, 2014, 2015, 2023), mice v. 3.16.0 (van Buuren & Groothuis-Oudshoorn, 2011), naniar v. 1.0.0 (Tierney & Cook, 2023), pacman v. 0.5.1 (Rinker & Kurkiewicz, 2018), patchwork v. 1.1.3 (Pedersen, 2023), plm v. 2.6.3 (Croissant & Millo, 2008, 2018; Millo, 2017), remotes v. 2.4.2.1 (Csárdi et al., 2023), rmarkdown v. 2.25 (Allaire et al., 2023; Xie, Allaire, & Grolemund, 2018; Xie, Dervieux, & Riederer, 2020), skimr v. 2.1.5 (Waring et al., 2022), stargazer v. 5.2.3 (Hlavac, 2022), tidyverse v. 2.0.0 (Wickham et al., 2019), wbstats v. 1.0.4 (Piburn, 2020), WDI v. 2.7.8 (Arel-Bundock, 2022).

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