The Ranking of Education Systems Based on Economic Factors

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### 1 Introduction

In many conversations today, the educational system of various countries is brought up as a point of pride, strength, and success. This paper will show research of the effect that economic factors have on education as well as a few classroom climate factors and will be a cross section analysis. First, there needs to be way of establish ranks for the countries. The average of the reading, math and science scores on the PISA test, were used. Eight different factors that might affect the rankings from two different years, 2000 and 2010 were included. The data will examine those countries that typically do well. All of the data can be found in the World Data Bank, United Nations Development Programme: Human Development Reports, World Prison Brief, United Nations Educational, Scientific and Cultural Organization, PISA 2000:Overview of the Study Design, Method and Results written by Stanat · Artelt · Baumert · Klieme · Neubrand · Prenzel · Schiefele · Schneider · Schumer · Tillmann · Weiß, and Organisation for Economic Co-operation and Development (OECD).

### 2 Literature Review

This topic is of global interest but in particular education ranking has been a hot topic in the United States in trying to maintain global status. This project will be different from the other literature because it explores ideas that tend to be ignored. Many studies that we found looked at public versus private schools and teacher pay. A country's education system is a useful determinant at showing the potential future progress a country will have and intellectual competitiveness. To measure a student's success or a country's education success, tests are performed to measure their abilities. Government[s], school authorities, and parents are constantly trying to improve student's performance in order to better the student's life and increase economic growth. Many studies have been done on specific factors that schools can control to a degree, such as student to teacher ratio and school expenditures. Student to teacher ratio appears to be an important factor contributing to success or failure of a student's performance. A recent paper helps look into this effect, Effect of Education Policies and Institutions on Student Performance by Jieun Hong. In his paper he is looking to see whether school autonomy is being by the quality of education. Hong, suggests that "the pupil-teacher ratio is significant with negative signs, which indicate that students from a small class size can be educated effectively and efficiently" (Hong 2015). Showing that class size can have either a positive or negative impact on a child's education. Another study, The Tennessee Study of Class Size in the Early School Grades, looks at class sizes in Tennessee over a four year period, where students were comprised of "a small class of 13 and two classes of 22" (Mosteller 1995). The results indicated "students who were originally in smaller classes continued to perform better than the students from regular-sized classes with or without a teacher's aide" (Mosteller 1995).

School expenditure to GDP is another studied variable because questions how been raised whether or not raising education spending increases student performance. The United States spends a smaller portion of its GDP on education but spends more per student than most countries. Many studies have looked into this controversial topic and the overall results show that increasing spending on education does not have a significant impact because it "depends on how the money is spent, not on how much money is spent" (Lips, Watkins, Fleming 2008). In the research paper, Does Spending More on Education Improve Academic Achievement?, the authors, Dan Lips, Shanea J. Watkins, Ph.D., and John Fleming, look into the per-pupil expenditures on public education by each state through grades K-12, as well as, 4th, 8th, and 12th grade NAEP math and reading scores. Their results indicated that increasing education spending does not led to significant refined student performance. Prison population has been a recent variable to study because of the differences in criminal systems among the International community. A recent literature, Mass incarceration and children's outcomes: Criminal justice policy is education policy by Leila Morsy and Richard Rothstein, shows how "children with incarcerated parents are 33 percent more likely to have speech or language problems—like stuttering or stammering—than otherwise similar children whose fathers have not been incarcerated" (Morsy Rothstein 2016). Their results help shed light to a serious issue that the current criminal system can have a negative impact on a child's school performance. Many other variables have shown that there could be links between them and a student's educational performance but more research is required to provide conclusive results.

#### 3 Data

The dependent variable will be an average of the reading, math and science scores on the PISA test. We will include observations from 2000 and from 2010 from the PISA 2000:Overview of the Study Design, Method and Results written by Stanat · Artelt · Baumert · Klieme · Neubrand · Prenzel · Schiefele · Schneider · Schumer · Tillmann · Weiß, and Organization for Economic Co-operation and Development (OECD). This paper will include 8 independent variables and each will include observations from 2000 and 2010. Included: enrollment rate of both sexes as a percentage from the Data World Bank, the Education Index which is a number between 0 and 1 which accounts for the mean years of schooling in a country as percentage of the expected years of schooling for an individual from Human Development Report, School Expenditure as percentage of GDP from the Data World Bank, and the unemployment Rate as a percentage from the Data World Bank. Also included: health expenditure per capita in USD from the Data World Bank, the ratio of pupils to teacher from United Nations Educational, Scientific and Cultural Organization, the prison population rate in per hundred thousand of the national population from World Prison Brief, and gross domestic saving as a percentage of GDP from the Data World Bank.

The variable names are labeled differently for computational analysis. For Year 2000, x is the average PISA scores, x1 is enrollment rate, x2 is education Index, x3 is School Expenditure as a percentage of GDP, x4 is unemployment rate, x5 is Health Expenditure per Capita, x6 is Pupil to Teacher Ratio, x7

Prison Population Rate, x8 is Gross Domestic Savings. For Year 2010, x0 is the average PISA scores, x11 is enrollment rate, x21 is education Index, x31 is School Expenditure as a percentage of GDP, x41 is unemployment rate, x51 is Health Expenditure per Capita, x61 is Pupil to Teacher Ratio, x71 Prison Population Rate, x81 is Gross Domestic Savings.

Limitations to the project include that the data was not entirely available for a select number of observations for the years 2000 and 2010. In order to have a complete data set to work with, information had to be looked at from the years immediately before or after 2000 and 2010. This project is also limited by my own personal opinions since variables were chosen that were thought to have a possible affect on the average PISA score.

The predictions were as follows; Leaving all else constant for all variables: it is predicted that the enrollment rate of both sexes will have a positive effect on the rank of education because we believe having a higher enrollment rate of both sexes will result in better overall test scores. The education index will have a positive effect on the rank of education because having high test scores is a good indicator schools are retaining students. School expenditure will have a negative effect on the rank of education because increasing educational spending does not necessarily produce positive results. Unemployment rate will have a negative effect on the rank of education because we believe that the more people that are unemployed would likely be tied to a less educated

population. Health expenditure per capita will have a positive effect on the rank of education because if health expenditure is rising then the population is less likely to attend school and is as sign that people are becoming sick more often. We believe that if a student is frequently missing school that they would not perform well on standard tests. The ratio of pupils to teacher will have a positive effect on the rank of education because having a low student to teacher ratio indicates there is more personal involvement between students and teachers. Prison population rate will have a negative effect on the rank of education because of the negative psychological effects that having a parent in prison as well as the likelihood of a child going to prison if their parent has been there. Gross domestic savings will have a positive effect on the rank of education because a positive gross domestic savings rate would indicate that a country has money to invest in education.

### 4 Methods

The method to calculate results for this interesting topic is not complex. The methods for this project consisted of using a Cross-Sectional analysis following an Ordinary Least Squares regression on the Years 2000 and 2010, with all the work being done in R Studio.

The primary empirical model for Year 2000 can be depicted in the following equation:

$$y = 0 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8$$

While the primary empirical model for Year 2010 can be depicted in the following equation:

$$y = 0 + x11 + x21 + x31 + x41 + x51 + x61 + x71 + x81$$

where y, labeled x for 2000 and x0 for 2010 in the data set, is average PISA score that dictates a country's education rank. While the x's are the variables of interest that impact y. A linear regression was used to determine the strength that the x's have on y in order to see which variables have more or less impact on PISA scores. Individual Linear Regressions were performed on both Years 2000 and 2010. Additionally, a regression for the differences between Years 2000 and 2010 was performed as well, which can be depicted below: Lkasjflkjfldsjfljdafljdsflkjflk;dsjflkjdsfl;dasfj;ldkjfl;kdsjfl;kdsjfl;kdsjfl;kdsjfl;kdsjfl;kdsjfl;kdsjfl

Once the individual regression were performed, correlation, heterosked asticity, and multicollinearity test were performed on both the Years 2000 and 2010 to address any possible issues with the data.

# 5 Findings

The following three regression models are need with their statistics can be seen in

Figures 1-3. In Figure 1, the Year 2000 had an adjusted  $R^2 of 0.5135$ , while in Figure 2, the Year 2010 had and adjusted  $R^2 of 0.5135$ , while in Figure 2.

values less than a 5% significant level. However, the rewas only one significant variable in both the models, Education labeled x 2 for 2000 and x 21 for 2010. Education Index was the only significant variable at the 5% significant level. As Figures 1-2. In Figures 4-5, a matrix graph shows the individual linear models for both Years. In both models, the 7.

After the original regression were performed, reduced models were made for both years with the significant variables that passed the 5% criteria level. The reduced equations for each year with their statistics are below: Year 2000 x=0+2x2 Year 2010 x0=0+21x21

In the reduced equations, it is noticed that there was still no presence of multicollinearity due to the fact that there is only one independent variable in each equation. In these models, there is no presence of heteroskedasticity and our adjusted R2 value is lower than in our original model for the year 2000. There is no heteroskedasticity in 2010 for the adjusted model and only a slightly higher adjusted R2 value. For these reasons, the original two models are the ones that fit the data the best.

Since this is cross sectional data, testing was done on the original model for heteroskedasticity and noted that there was none as seen in Figures 8-9. If there had been heteroskedasticity, it would have to be concerned that even though the R2 were larger, that these models would not be efficient. Furthermore, it would imply that the best linear unbiased estimators were not correctly selected. Since the models do not have this issue, we know that we can rely on the inferential procedures. However, the limitations must still be aware that the

majority of our variables were insignificant.

## 6 Conclusion

Though there were imperfections with the model and data, it is believed that the research that had been done is a good stepping-stone for future studies on this topic. There are other variables that could be considered and this topic lends itself to a time series studies of the data as well. Education is a subject that has many facets to consider. There are also many other factors that cannot be measured or are extremely difficult to. Such effects could be psychological trauma to a child; such trauma can be hard to assess into a model. Though there were not many significant variables, some of them can impact a child's learning and furthermore impact a country's education rank.

## References

???? ???? ???? ???? ???? ???? Hong, Jieun. 2015. "Effects of Education Policies and Institutions on Student Performance.". Lips, Dan, Shanea J Watkins, and John Fleming. 2008. "Does Spending More on Education Improve Academic Achievement? Backgrounder. No. 2179." Heritage Foundation . Morsy, Leila and Richard Rothstein. 2016. "Mass incarceration and children's outcomes: Criminal justice policy is education policy." Washington, DC: Economic Policy Institute . Mosteller, Frederick. 1995. "The Tennessee study of class size in the early school

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 $\label{thm:constraint} Uis.~\ref{eq:constraint} \ URL~\ http://data.uis.unesco.org/?queryid=180.$ 

# 7 Tables

Table 1: Year 2000 Regression

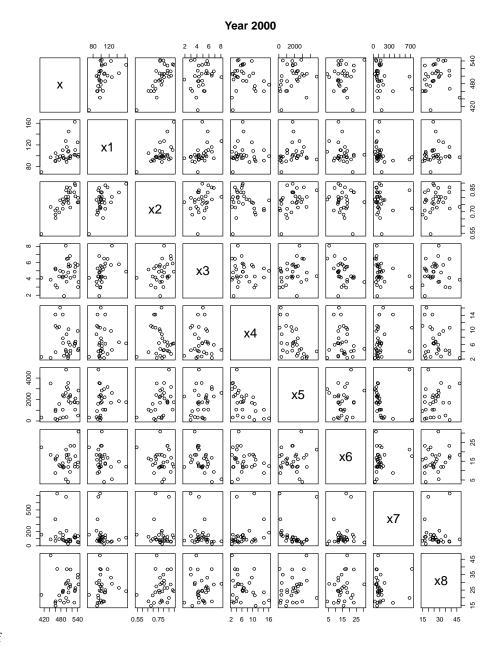
	$Dependent\ variable:$
_	Dependent variable.
	x
x1	0.165 (0.301)
x2	321.682*** (71.789)
x3	-0.484 (3.965)
x4	0.016 (1.370)
x5	$-0.002\ (0.005)$
x6	1.181 (0.874)
x7	$-0.043 \; (0.027)$
x8	$-0.258 \; (0.557)$
Constant	232.685*** (50.959)
Observations	30
$\mathbb{R}^2$	0.648
Adjusted $\mathbb{R}^2$	0.513
Residual Std. Error	21.719 (df = 21)
F Statistic	$4.826^{***} (df = 8; 21)$

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

Table 2: Year 2010 Regression

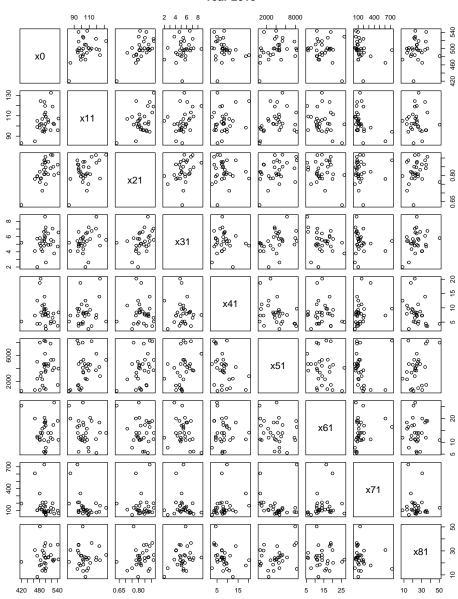
_	Dependent variable:			
	x0			
x11	-0.044 (0.494)			
x21	261.602*** (75.110)			
x31	-1.453 (3.230)			
x41	-0.656 (1.198)			
x51	0.0004 (0.002)			
x61	0.234 (0.778)			
x71	$-0.053 \ (0.031)$			
x81	$-0.222 \ (0.500)$			
Constant	309.655*** (59.026)			
Observations	30			
$\mathbb{R}^2$	0.511			
Adjusted $R^2$	0.325			
Residual Std. Error	19.888 (df = 21)			
F Statistic	$2.747^{**} (df = 8; 21)$			

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



correlation.pdf

### Year 2010



 $2010\ correlation.pdf$ 

Table 3: Reduced Year 2000 Regression

_	Dependent variable:			
	X			
x2	312.511*** (52.279)			
Constant	255.274*** (40.657)			
Observations	30			
$\mathbb{R}^2$	0.561			
Adjusted $\mathbb{R}^2$	0.545			
Residual Std. Error	21.004 (df = 28)			
F Statistic	$35.734^{***} (df = 1; 28)$			

Note:

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

Table 4: Reduced Year 2010 Regression

	Dependent variable:				
	x0				
x21	239.774*** (57.750)				
Constant	300.760*** (47.701)				
Observations	30				
$\mathbb{R}^2$	0.381				
Adjusted R <sup>2</sup>	0.359				
Residual Std. Error	19.384 (df = 28)				
F Statistic	$17.239^{***} (df = 1; 28)$				
Note:	*p<0.1; **p<0.05; ***p<0.01				

Table 5: Year 2000 VIF

x1	x2	x3	x4	x5	x6	x7	x8
1.694	1.764	1.631	1.680	1.791	1.437	1.201	1.182

Table 6: Year 2010 VIF

x11	x21	x31	x41	x51	x61	x71	x81
2.224	1.607	1.360	1.729	1.719	1.205	1.659	1.304