Human Capital Index Hypotheses Testing

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```
# load required packages
library(dplyr)
library(tidyr)
library(car)
library(ggplot2)
```

Dataset required: WorldBankData.csv

(Note: This dataset comes from a publically available dataset from The World Bank.

https://databank.worldbank.org/source/world-development-indicators

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There are 8 variables in this real-world dataset, from 258 countries in 2016/2017:

- Country.Name.Name: name of country
- Country. Code: code given to country
- Human.Capital.Index: unitless number that goes from 0 to 1.
- GDP.per.capita.PPP in US Dollar. This is GDP per capita, but taking into account the purchasing power of the local currency, by comparing how much it costs to buy a basket of goods (e.g. food) compared to the reference currency (USD). (PPP stands for Purchasing Power Parity)
- Health.Expenditure.per.capita in US Dollar.
- Tertiary.Education.Expenditure.per.student in US Dollar.
- Population . Life Expectancy at birth in years.
- Diabetes. Prevalence in units of % of population ages 20 to 79.
- Years.of.Compulsory.Education in years.

Being a data set in real world, there are lots of missing data. Be wary of this!

```
dta_wb <- read.csv("/Users/minhchau/Downloads/WorldBankData.csv")</pre>
```

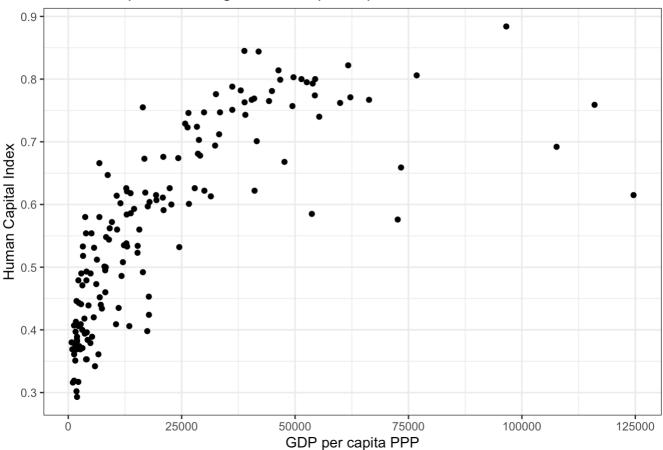
First, let's investigate Human.Capital.Index. As noted by Prime Minister Lee in his 2019 National Day Rally, Singapore topped the world on this Human Capital Index in 2018. Let's try to see what are some of the possible variables that correlate with this.

(1a) Start off by plotting Human.Capital.Index (on the y-axis) versus GDP.per.capita.PPP on the x-axis.

```
ggplot(dta_wb, aes(x=GDP.per.capita.PPP, y=Human.Capital.Index)) + ggtitle("Human Cap
ital Index against GDP per capita PPP") + xlab("GDP per capita PPP") +ylab("Human Cap
ital Index") +geom_point()+theme_bw()
```

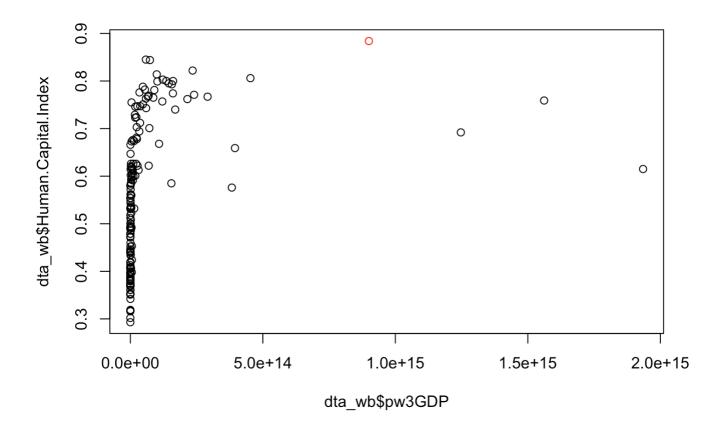
```
## Warning: Removed 101 rows containing missing values or values outside the scale ra
nge
## (`geom_point()`).
```

Human Capital Index against GDP per capita PPP

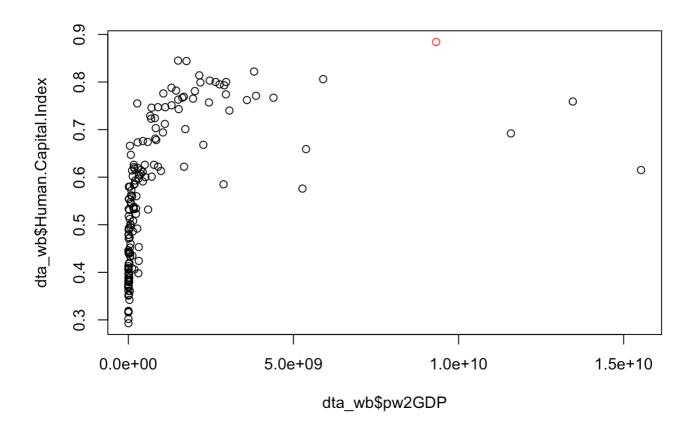


(1b) What type of transformation could you apply? Try a few functions that were shown in class: x^2 , x^3 , ..., exp(x), log(x). Make a plot that shows a linear relationship, and describe what you did. Highlight the dot that represents Singapore.

dta_wb\$pw3GDP = (dta_wb\$GDP.per.capita.PPP)^3
plot(dta_wb\$pw3GDP, dta_wb\$Human.Capital.Index, col=ifelse(dta_wb\$Country.Name == "Si
ngapore", 'red', 'black'))

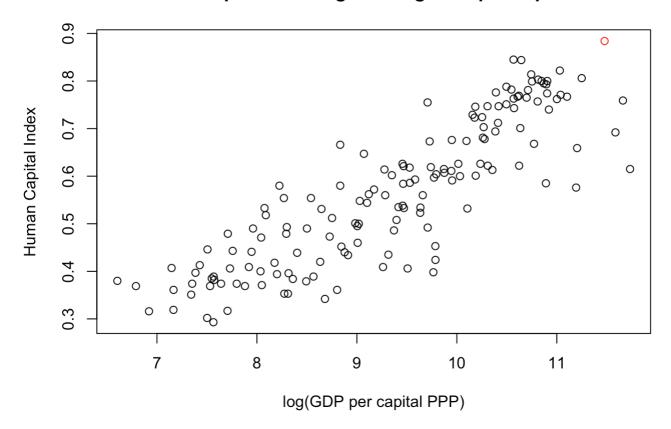


dta_wb\$pw2GDP = (dta_wb\$GDP.per.capita.PPP)^2
plot(dta_wb\$pw2GDP, dta_wb\$Human.Capital.Index, col=ifelse(dta_wb\$Country.Name == "Si
ngapore", 'red', 'black'))



dta_wb\$logGDP = log(dta_wb\$GDP.per.capita.PPP)
plot(dta_wb\$logGDP, dta_wb\$Human.Capital.Index, col = ifelse(dta_wb\$Country.Name ==
"Singapore", 'red', 'black'), main="Human Capital Index against log GDP per capital P
PP", xlab = "log(GDP per capital PPP)", ylab = "Human Capital Index")

Human Capital Index against log GDP per capital PPP



(1c) Now that you have a plot of a linear relationship, run a linear regression using lm(), predicting Human Capital Index. Run summary() on the lm object to produce an output table. Interpret the output of the lm(). What do the b's regression coefficients estimates mean? (Interpret them and try to make sense of the numbers. How many countries made it into this regression? (What happened to the rest?) Comment on the goodness-of-fit statistics.

(1d) Do you think that log(GDP) a significant predictor in our linear regression model?

summary(lm(Human.Capital.Index ~ logGDP, dta_wb))

```
##
## Call:
## lm(formula = Human.Capital.Index ~ logGDP, data = dta_wb)
##
## Residuals:
       Min
##
                  10
                      Median
                                    30
                                            Max
## -0.21270 -0.04959
                     0.01103 0.06164
                                       0.15487
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
                           0.047260 -9.155 3.03e-16 ***
## (Intercept) -0.432638
## logGDP
                0.106843
                           0.005008 21.335 < 2e-16 ***
## ---
## Signif. codes:
                  0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.07666 on 155 degrees of freedom
     (101 observations deleted due to missingness)
##
## Multiple R-squared: 0.746, Adjusted R-squared:
## F-statistic: 455.2 on 1 and 155 DF, p-value: < 2.2e-16
```

We state our hypotheses H0: $\beta\beta\log(\text{GDP})=0$ H1: $\beta\beta\log(\text{GDP})\neq0$ where $\beta\beta\log(\text{GDP})$ represents the population slope parameter of the natural log- transformed GDP variable Given that p-value = 2e-16 < 0.05 at 5% level of significance We can reject the null hypothesis H0: $\beta\beta\log(\text{GDP})=0$, meaning that the population slope parameter of log(GDP) is statistically significantly different from zero. Thus we conclude that the log-transformed GDP variable (log(GDP)) is a significant predictor in our linear regression model.