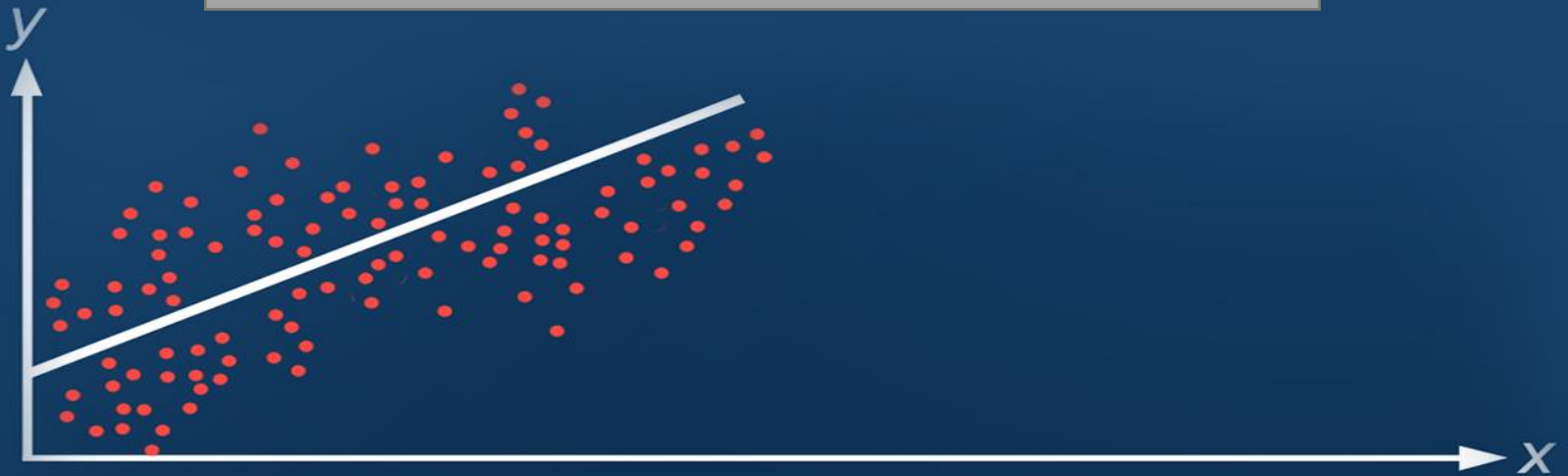




TBA2102 2020/2021 Semester 2 Tutorial 7





STRUCTURE OF TUTORIALS

Duration:

45 mins

Content:

- Tutorial 6 (Qn 2 and 3)

Tutorial 7



DATASET REQUIRED

Tutorial6_WorldBankData.csv

Note: This dataset comes from a publically available dataset from The World Bank.
<https://databank.worldbank.org/source/world-development-indicators>.

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

- **Human.Capital.Index** : Unitless number that goes from 0 to 1.
- **GDP.per.capita.PPP**: In \$. 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 \$.
- **Tertiary.Education.Expenditure.per.student**. In \$.
- **Population**. In people.
- **Life.Expectancy.at.birth**. In years.
- **Diabetes.Prevalence**. In units of % of population ages 20 to 79.
- **Years.of.Compulsory.Education**. In years.

This being a real dataset, there is **lots of missing data**. Be wary of this!



QUESTION 2A

Now let's consider another set of variables in the same dataset:

- Health.Expenditure.per.capita
 - Diabetes.Prevalence, and
 - Life.Expectancy.at.birth.
-
- Design a predictive hypothesis with these three variables.
 - Which would be your dependent variable, and which would be your independent variables? Justify your answer.

$$\text{Life.Expectancy.at.birth.} = b_0 + b_1 \text{Health.Expenditure.per.capita} + b_2 \text{Diabetes.Prevalence}$$

- Value of life expectancy
- Manipulability of Health.Expenditure.per.capita
- Why not Diabetes.Prevalence?



QUESTION 2B

Plot the bivariate relationships between these three variables.

- In other words, **plot x-y scatterplots**. There are 3 variables, so you'll need 3 scatterplots.
- For the **Health.Expenditure.per.capita** variable, please also **apply the same transformation** in (1b) for the scatterplot.
- Comment on the relationship between the variables.

QUESTION 2B: DF\$LOGHEALTHEXPENDITURE, DF\$DIABETES.PREVALENCE

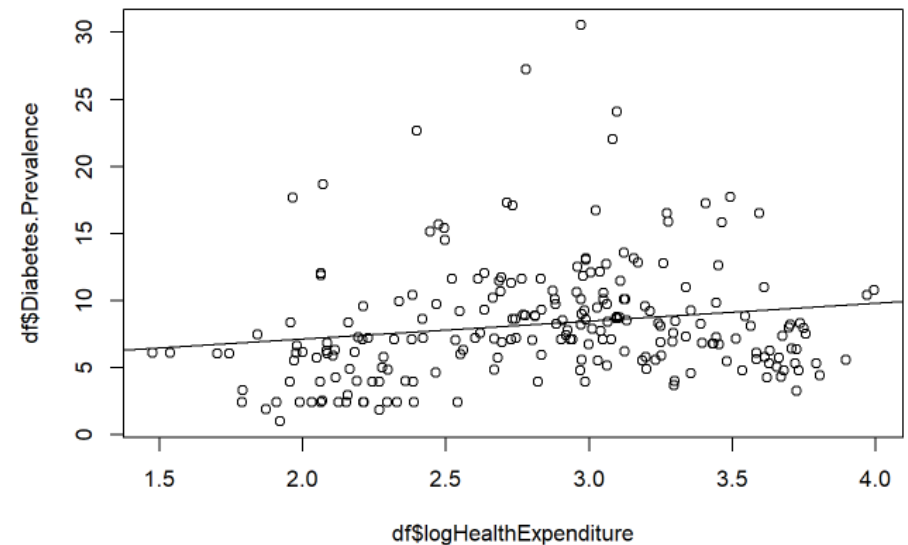
```
# log transform Health.Expenditure.per.capita
df$logHealthExpenditure<-log10(df$Health.Expenditure.per.capita)

? (df$logHealthExpenditure, df$Diabetes.Prevalence)
```

- What is the nature of this relationship?



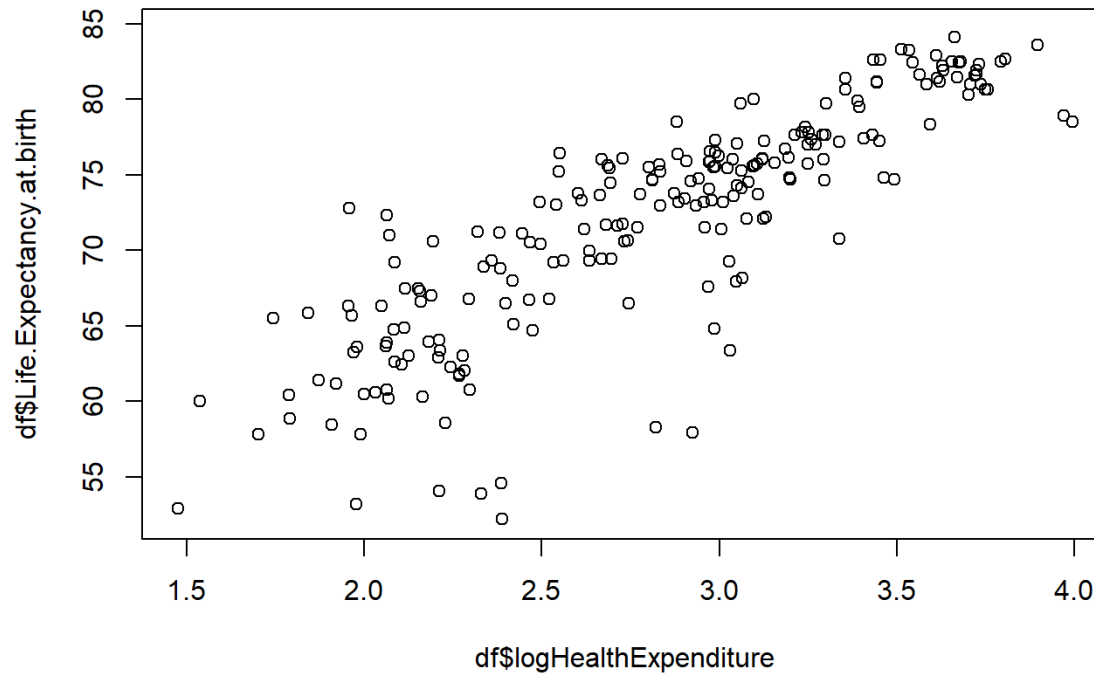
- We can draw best fit line with the ? function.



```
abline(lm(Diabetes.Prevalence ~ logHealthExpenditure, data=df))
```

QUESTION 2B: LOGHEALTHEXPENDITURE, DF\$LIFE.EXPECTANCY.AT.BIRTH

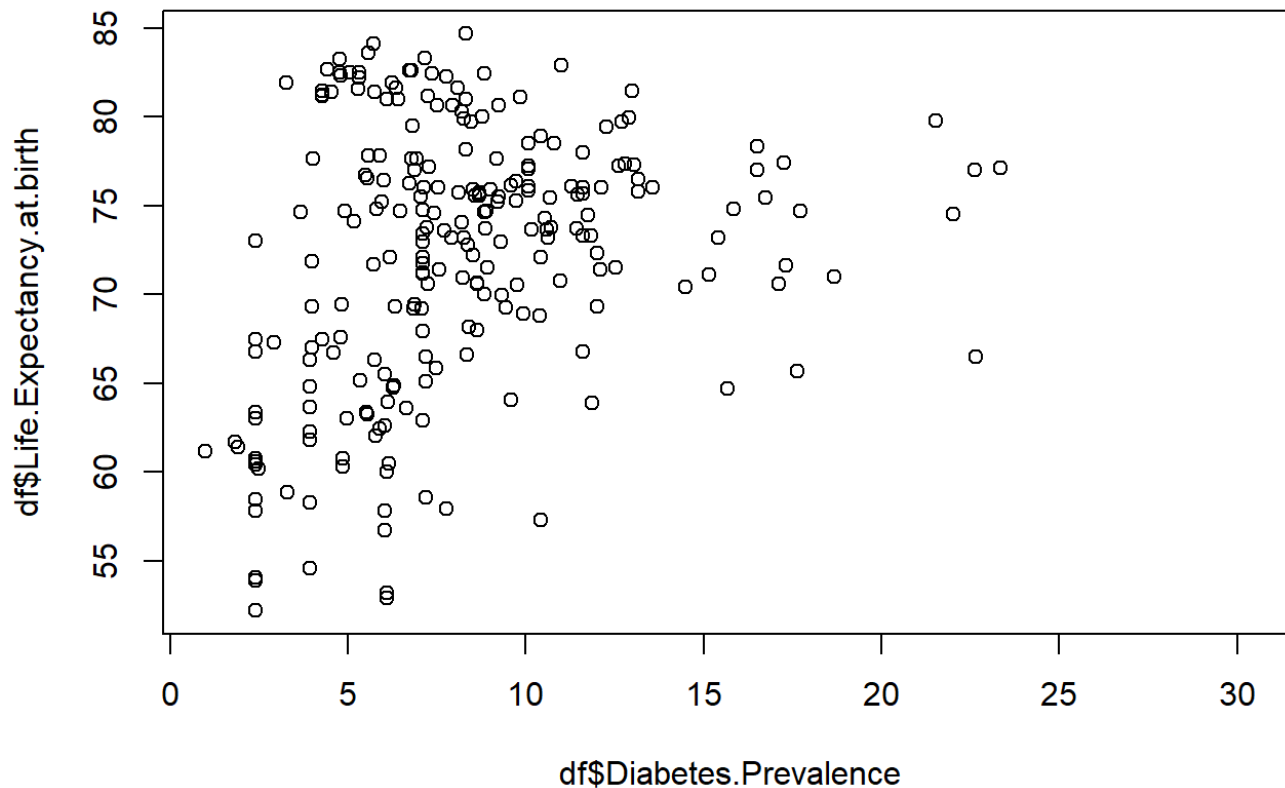
```
plot(df$logHealthExpenditure, df$Life.Expectancy.at.birth)
```



- What is the nature of this relationship?

QUESTION 2B: DF\$DIABETES.PREVALENCE, DF\$LIFE.EXPECTANCY.AT.BIRTH

```
plot(df$Diabetes.Prevalence, df$Life.Expectancy.at.birth)
```



- What is the nature of this relationship?



QUESTION 2C

- Run a multiple regression [predicting Life.Expectancy.at.birth](#) using the other 2 variables.
- [Interpret the coefficients](#), spelling out what the numbers mean.
- Comment on your answers.

QUESTION 2C

```
fit2<- ? (Life.Expectancy.at.birth ~ log10(Health.Expenditure.per.capita) + Diabetes.Prevalence, data=df)
? (fit2)
```

```
##
## Call:
## lm(formula = Life.Expectancy.at.birth ~ log10(Health.Expenditure.per.capita) +
##     Diabetes.Prevalence, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -15.0787  -1.4875   0.6018   2.0976  10.0565
##
## Coefficients:
##              Estimate Std. Error    ? Pr(>|t|)
## (Intercept)      39.61736      1.33051  29.78 < 2e-16 ***
## log10(Health.Expenditure.per.capita) 10.77368      0.45941  23.45 < 2e-16 ***
## Diabetes.Prevalence      0.24448      0.06847   3.57 0.000438 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.796 on 218 degrees of freedom
## (37 observations deleted due to missingness)
## Multiple R-squared:  0.7388, Adjusted R-squared:  0.7364
## ?
```

t-distribution or z-distribution?

Based on results above? What is the result of the F-statistic?

- Interpret the intercept.
- Interpret the coefficient before $\log_{10}(\text{Health.Expenditure.per.capita})$.
- What is the nature of the relationship between Diabetes.Prevalence & Life.Expectancy.at.birth?



QUESTION 3

- Let's again return to **Human.Capital.Index** as our outcome of interest.
- According to the World Bank (see footnote), this **index measures the amount of capital that a child born today can expect to attain by age 18**, and is influenced by education and healthcare.

$$\text{Human.Capital.Index} = \text{?} + \beta_1 \text{Tertiary.Education.Expenditure.per.student} + \beta_2 \text{Healthcare}$$



QUESTION 3A

A fellow student comes up with a hypothesis that quality of education should affect Human.Capital.Index.

- But something tells you that this is not so straightforward. What is the danger of putting Tertiary.Education.Expenditure.per.student into a linear model as a regressor?
- (Hint: Check its distribution by plotting and/or using the summary() function. Is there anything worth noting about this variable?) [2 marks]

$$\text{Human.Capital.Index} = \beta_0 + \beta_1 \text{Tertiary.Education.Expenditure.per.student}$$

```
summary(df$Tertiary.Education.Expenditure.per.student)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.      NA's  
##    3.121  16.412  27.276  43.677  38.411 334.000     208
```

```
sum(!is.na(df$Tertiary.Education.Expenditure.per.student))
```

```
## [1] 50
```

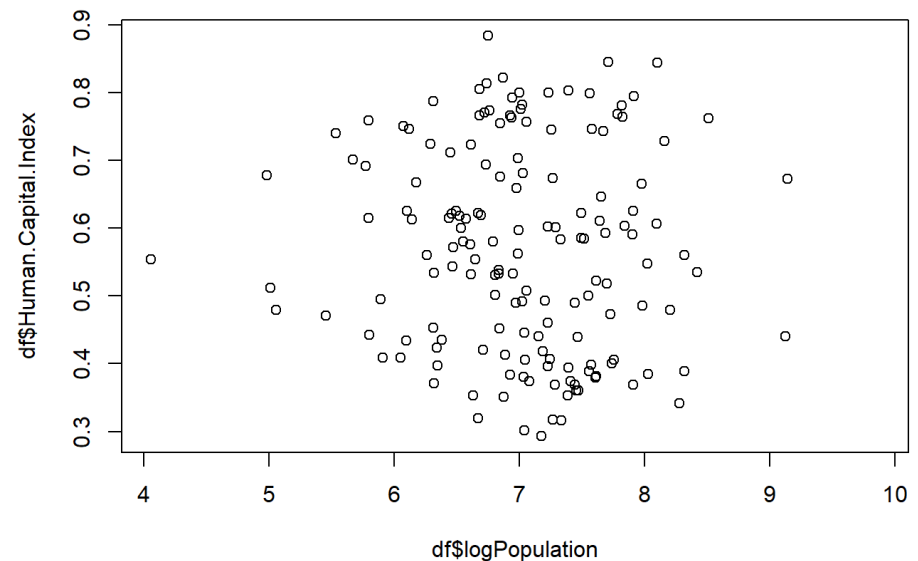
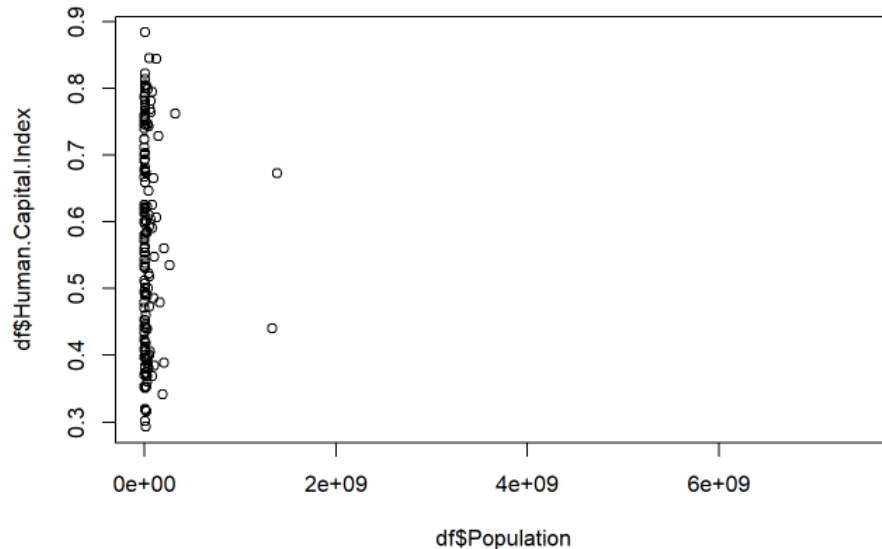
- Would you include this variable [Tertiary.Education.Expenditure.per.student] in the model?

QUESTION 3B

- Is there a relationship between country population and its Human Capital Index?
- Is there a need to transform variables?
- Show this visually and using a linear model.

```
df$logPopulation=log10(df$Population)
plot(x=df$Population, y=df$Human.Capital.Index)
```

```
plot(x=df$logPopulation, y=df$Human.Capital.Index)
```



QUESTION 3B

- Is there a relationship between country population and its Human Capital Index? Is there a need to transform variables? Show this visually and using a linear model.

```
fit3<-? (Human.Capital.Index ~ logPopulation, df)
summary( ? )
```

```
##
## Call:
## lm(formula = Human.Capital.Index ~ logPopulation, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.272186 -0.144511 -0.001899  0.123382  0.313550
##
## Coefficients:
##              Estimate Std. Error    ? Pr(>|t|)
## (Intercept)   0.65356    0.11074   5.902 2.19e-08 ***
## logPopulation -0.01231    0.01569  -0.785   0.434
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1518 on 155 degrees of freedom
## (101 observations deleted due to missingness)
## Multiple R-squared:  0.003959,    Adjusted R-squared:  -0.002468
## F-statistic: 0.616 on 1 and 155 DF,  p-value: 0.4337
```

t-distribution or z-distribution?

- Interpret the intercept.
- Interpret the coefficient before logPopulation



QUESTION 3C

Your manager is interested in factors that **predict whether a country is above average on Human.Capital.Index or below average.**

- First, **do a median-split on Human.Capital.Index**. Specifically, create a variable that is 1 (or TRUE) if the country's Human.Capital.Index is greater than or equal to the MEDIAN of all countries, and 0 (or FALSE) otherwise.
- Next, **run a generalized linear model to predict** whether a country will be above the median on Human.Capital.Index, using the following variables: GDP.per.capita.PPP, Health.Expenditure.per.capita, Life.Expectancy.at.birth and Diabetes.Prevalence.
- Apply transformations using your best judgment. [2 marks] Interpret the output of the model, and discuss the meaning of each of the coefficients. [2 marks]

QUESTION 3C

```
df$HCI.aboveMedian <- (df$Human.Capital.Index >= median(df$Human.Capital.Index, na.rm=T))
```

```
summary( ? (HCI.aboveMedian ~ log10(GDP.per.capita.PPP) + log10(Health.Expenditure.per.capita) + Life.Expectancy.  
at.birth + Diabetes.Prevalence, df, ? ))
```

```
##  
## Call:  
## glm(formula = HCI.aboveMedian ~ log10(GDP.per.capita.PPP) + log10(Health.Expenditure.per.capita) +  
##   Life.Expectancy.at.birth + Diabetes.Prevalence, family = "binomial",  
##   data = df)  
##  
## Deviance Residuals:  
##      Min       1Q   Median       3Q      Max   
## -2.4368  -0.1834   0.0396   0.2748   2.5107   
##  
## Coefficients:  
##              Estimate Std. Error ? Pr(>|z|)        
## (Intercept)      -42.12004     9.68770  -4.348 1.38e-05 ***  
## log10(GDP.per.capita.PPP)      6.13016     2.61398   2.345 0.01902 *   
## log10(Health.Expenditure.per.capita) -0.95825     2.24115  -0.428 0.66896   
## Life.Expectancy.at.birth      0.28392     0.10454   2.716 0.00661 **  
## Diabetes.Prevalence      -0.11529     0.09372  -1.230 0.21862   
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## (Dispersion parameter for binomial family taken to be 1)  
##  
##    Null deviance: 206.390  on 148  degrees of freedom  
## Residual deviance:  67.554  on 144  degrees of freedom  
## (109 observations deleted due to missingness)  
## AIC: 77.554  
##  
## Number of Fisher Scoring iterations: 7
```

t-distribution or z-distribution?

- Interpret the intercept.
- Interpret the coefficient before all the predictors



THANK YOU.

SEE YOU NEXT WEEK.