Assignment3

Research Quetion

Twenty years since the end of Apartheid: Did the collapse of Apartheid play a significant role in reducing racial and social inequality in South Africa? Is post-apartheid South Africa better off or worse off than during the apartheid era?

Definition of Racial and social inequality

Before conducting the data analysis to find the answer for our research question, we begin by clarifying the definitions for racial and social inequality. While social inequality broadly refers to the existence of unequal opportunities for different social status/positions within a society, racial inequality can be seen as one of dimentions of social inequality. It thus indicates the discrimination based on race in access to socioeconomic opportunities or services. In our research study, we will specifically look into racial discrimination in terms of employment, education, and income levels. Becasue these three indicators within the capitalism society can be seen as fundamental yet significant estimators for the quality of human well-being, we decided to include them. In addition, we will try to identify drivers of unequal income distribution by controlling possible factors and variables such as unemployment rate and education level.

Literature review

2

3

Before starting investigation, we need to look around previous researches which has been writen by various ambitious researchers.

According to Leibbrandt, (see Leibbrandt (n.d.)) 1, Since the fall of Apartheid (1993~2008), overall (include all races) income inequality increased. The same is true among four major racial groups. 2, However, the major driver of inequality increase is intra-African inequality in South-Africa.

The reason why we choose this article as the first reference article for this article is that this is the most cited work in the South-African Inequality Study.

Data Gathering by web-scraping

2003

White 14 468 16 581

Race

2012 Increase

According to Leibbrandt, inequality has been increased since the fall of Apartheid. We will test this assumption by using other data which is not used in the article.

We found the data of monthly earnings among races and gender. We tried to scraping the data from the website.

```
URL <- 'http://businesstech.co.za/news/wealth/131524/this-is-the-average-salary-in-south-africa-by-race
RaceEarningsTable <- URL %>% read_html() %>%
                    html_nodes('#container > div.content_holder > div.content > div.post_single > div.p
                    html_table() %>%
                    as.data.frame
RaceEarningsTable
                       Х2
##
                               ХЗ
                                        X4
                                               Х5
                                                       Х6
                                                                Х7
                X 1
## 1
                   Median Median
                                    Median
                                             Mean
                                                     Mean
                                                              Mean
```

2012 Increase

7%

2003

15% 11 249 11 991

```
Asian/Indian 7 825 11 701
                                      50%
                                           5 264
                                                  8 993
                                                              60%
## 5
          Coloured 4 241 7 058
                                           2 437
                                                              60%
                                      66%
                                                  3 897
## 6 Black African 4 059 5 445
                                      34%
                                          2 437
                                                  2 998
                                                              23%
URL <- 'http://businesstech.co.za/news/wealth/131524/this-is-the-average-salary-in-south-africa-by-race
GenderEarningsTable <- URL %>% read_html() %>%
                    html_nodes('#container > div.content_holder > div.content > div.post_single > div.p
                    html table() %>%
                    as.data.frame
GenderEarningsTable
##
         Х1
                X2
                       ХЗ
                                X4
                                      Х5
                                             X6
                                                      Х7
## 1
            Median Median
                            Median
                                    Mean
                                          Mean
                                                    Mean
## 2
              2003
                     2012 Increase
                                    2003
                                          2012 Increase
       Race
```

28%

28%

Data Cleaning and Merging

5 963

4 849

8 299

6 399

3

4 Female

Male

In this section, we will try to clean the data so that they can be statistical analysed.

39% 3 375 4 317

32% 2 435 3 118

Firstly, we use command "summary" to investigate the structure (class of variables, number of vectors) of data frames we got in the previous section.

summary(RaceEarningsTable)

```
ХЗ
##
         Х1
                             X2
##
    Length:6
                        Length:6
                                            Length:6
##
    Class :character
                        Class :character
                                            Class : character
##
   Mode :character
                        Mode :character
                                           Mode :character
##
         Х4
                             Х5
                                                Х6
##
    Length:6
                        Length:6
                                           Length:6
##
    Class : character
                        Class : character
                                           Class : character
##
   Mode :character
                        Mode :character
                                            Mode : character
         X7
##
##
   Length:6
##
    Class : character
   Mode :character
```

summary(GenderEarningsTable)

```
##
         Х1
                             Х2
                                                 ХЗ
##
    Length:4
                        Length:4
                                            Length:4
##
    Class : character
                        Class : character
                                            Class : character
                                            Mode :character
##
    Mode :character
                        Mode :character
##
         Х4
                             Х5
                                                 Х6
                                            Length:4
##
   Length:4
                        Length:4
##
    Class : character
                        Class : character
                                            Class : character
   Mode :character
                        Mode : character
                                            Mode :character
##
##
         X7
##
  Length:4
    Class : character
   Mode :character
```

As shown, every variables has a class of "characters" even though it represents numerical data.

The data we want to have is the mean of earnings among races and gender in 2003, 2012.

Firstly, we make TimeVector and IndivisualVector to labeling the data.

```
TimeVector <- c(2003,2012) #numerical vector
IndivisualVector <- c("Male", "Female", "White", "Asian/Indian", "Coloured", "BlackAfrican") #character vect</pre>
Then, we try to convert character vector to numerical vector.
male2003 <- as.numeric(gsub("([0-9]+).*$", "\\1", str_replace_all(GenderEarningsTable$X5[3], fixed(" ")
is.numeric(male2003)
## [1] TRUE
male2003
## [1] 3375
As I shown above, the character variable successfully converted to numerical variable. Then, we make function
which conduct this sequence.
Converter <- function(x){</pre>
y <- as.numeric(gsub("([0-9]+).*$", "\\1", str_replace_all(x, fixed(" "), "")))
return(y)
test <- Converter(x = GenderEarningsTable$X5[3])</pre>
is.numeric(test)
## [1] TRUE
test
## [1] 3375
Then, we can apply this function to all data.
#definition of vector
Earnings2003 <- c(0,0,0,0,0,0)
Earnings2012 <- c(0,0,0,0,0,0)
#GenderEarnings
for(i in 3:4){
  Earnings2003[i-2] = Converter(x = GenderEarningsTable$X5[i])
  Earnings2012[i-2] = Converter(x = GenderEarningsTable$X6[i])
#RaceEarnings
for(i in 3:6){
  Earnings2003[i] = Converter(x = RaceEarningsTable$X5[i])
  Earnings2012[i] = Converter(x = RaceEarningsTable$X6[i])
Earnings2003
## [1] 3375 2435 11249 5264 2437 2437
Earnings2012
## [1] 4317 3118 11991 8993 3897 2998
preEarnings <- data.frame(IndivisualVector, Earnings2003, Earnings2012)</pre>
preEarnings
```

IndivisualVector Earnings2003 Earnings2012

##	1	Male	3375	4317
##	2	Female	2435	3118
##	3	White	11249	11991
##	4	Asian/Indian	5264	8993
##	5	Coloured	2437	3897
##	6	${ t BlackAfrican}$	2437	2998

The preEarnings is messy data.

So we are going to transform it into tidy data.

```
library(tidyr)
Earnings <- gather(preEarnings, time, mean, Earnings2003:Earnings2012)
Earnings</pre>
```

```
##
      IndivisualVector
                                time
## 1
                  Male Earnings2003
                                      3375
## 2
                Female Earnings2003
## 3
                 White Earnings2003 11249
## 4
          Asian/Indian Earnings2003
                                      5264
## 5
              Coloured Earnings2003
                                      2437
## 6
          BlackAfrican Earnings2003
                                      2437
## 7
                  Male Earnings2012
                                      4317
## 8
                Female Earnings2012
                                     3118
## 9
                 White Earnings2012 11991
## 10
          Asian/Indian Earnings2012
              Coloured Earnings2012
## 11
                                      3897
## 12
          BlackAfrican Earnings2012
                                      2998
```

We suceeded to make the numerical vector showing the earnings among races and genders.

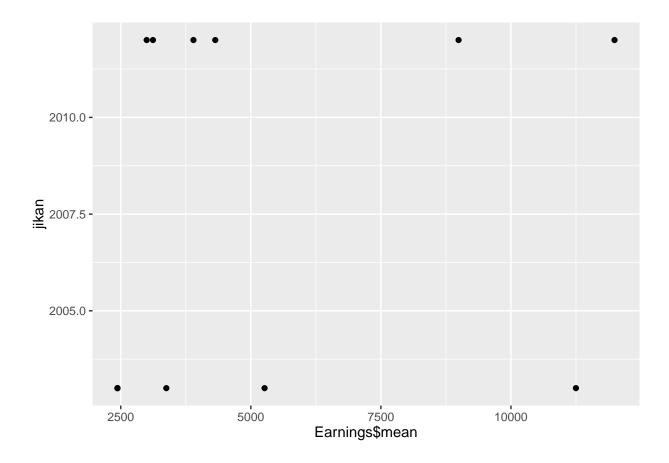
Data Gathering by using Data-API

Then, we try to gather data from WorldBank by using Worldbank Data API.

Conduct basic descriptive statistics

The data we gathered in previous section partialy statistically analysed (mean and median are already calcurated). In this section, we try to figure out the trend of inequality graphycally by using descriptive statistics.

We want to plot the data frame in earning mean vs time among each indivisual.



Briefly discribing

References

Leibbrandt, et al., <!-// //-> M. n.d. "Trends in South African Income Distribution and Poverty Since the Fall of Apartheid." OECD Publishing. doi:http://dx.doi.org/10.1787/5kmms0t7p1ms-en.