Swiss Dataset exercises

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Introduction

These exercises are based on the swiss dataset available in R. The swiss dataset contains data on fertility and socio-economic indicators for each of 47 French-speaking provinces of Switzerland at about 1888.

Exercises

1. Basic Statistics:

Get the structure of the swiss dataset using the str() function.

```
str(swiss)
```

```
## 'data.frame':
                   47 obs. of 6 variables:
## $ Fertility
                     : num 80.2 83.1 92.5 85.8 76.9 76.1 83.8 92.4 82.4 82.9 ...
## $ Agriculture
                     : num 17 45.1 39.7 36.5 43.5 35.3 70.2 67.8 53.3 45.2 ...
## $ Examination
                            15 6 5 12 17 9 16 14 12 16 ...
                     : int
## $ Education
                     : int
                            12 9 5 7 15 7 7 8 7 13 ...
## $ Catholic
                            9.96 84.84 93.4 33.77 5.16 ...
                     : num
  $ Infant.Mortality: num 22.2 22.2 20.2 20.3 20.6 26.6 23.6 24.9 21 24.4 ...
```

Calculate basic statistics for a variable (mean, median, standard deviation) using summary().

```
mean(swiss$Fertility)

## [1] 70.14255

median(swiss$Fertility)

## [1] 70.4

sd(swiss$Fertility)
```

[1] 12.4917

summary(swiss\$Fertility)

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 35.00 64.70 70.40 70.14 78.45 92.50
```

2. Correlations:

Calculate the correlation between Fertility and Education.

```
cor(swiss$Fertility, swiss$Education)
```

```
## [1] -0.6637889
```

Create a correlation matrix for all numeric variables in the dataset.

```
cor(swiss)
```

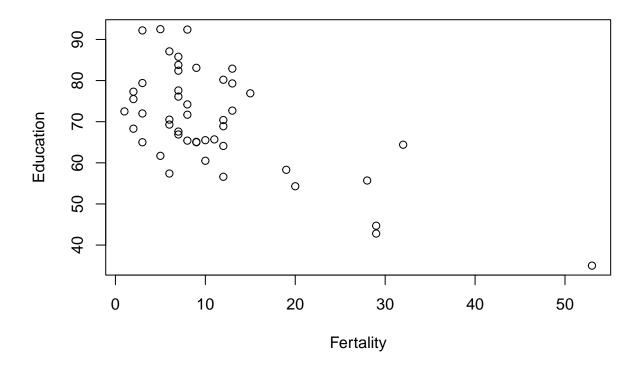
```
##
                     Fertility Agriculture Examination
                                                         Education
                                                                      Catholic
                     1.0000000 0.35307918 -0.6458827 -0.66378886 0.4636847
## Fertility
                     0.3530792 1.00000000 -0.6865422 -0.63952252
## Agriculture
                                                                    0.4010951
## Examination
                    -0.6458827 -0.68654221 1.0000000 0.69841530 -0.5727418
## Education
                    -0.6637889 -0.63952252 0.6984153 1.00000000 -0.1538589
## Catholic
                     0.4636847 \quad 0.40109505 \quad -0.5727418 \quad -0.15385892 \quad 1.0000000
## Infant.Mortality 0.4165560 -0.06085861 -0.1140216 -0.09932185 0.1754959
##
                    Infant.Mortality
## Fertility
                         0.41655603
                         -0.06085861
## Agriculture
## Examination
                         -0.11402160
## Education
                         -0.09932185
## Catholic
                          0.17549591
## Infant.Mortality
                        1.00000000
```

3. Visualization:

Create a scatter plot of Fertility against Education.

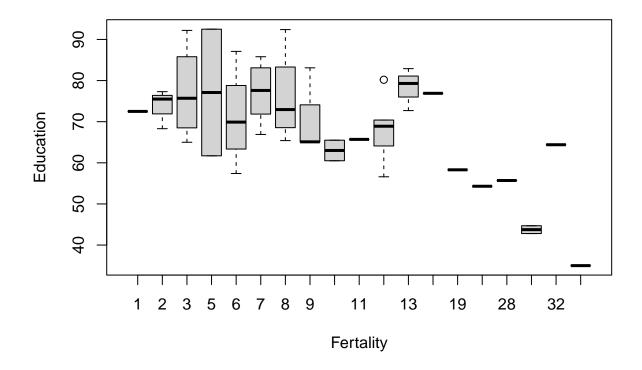
```
plot(swiss$Education, swiss$Fertility,
    ylab = "Education",
    xlab = "Fertality",
    main = "Scatter Plot of Fertility vs Education"
    )
```

Scatter Plot of Fertility vs Education



Create a boxplot of Fertility by Education.

```
boxplot(Fertility ~ Education, data = swiss, xlab = "Fertality", ylab = "Education")
```



4. Linear Regression:

Perform a simple linear regression to predict Fertility based on Education.

```
lm_model <- lm(Fertility ~ Education, data = swiss)
summary(lm_model)</pre>
```

```
##
## lm(formula = Fertility ~ Education, data = swiss)
##
## Residuals:
       Min
                1Q
                   Median
                                3Q
                                      Max
## -17.036 -6.711
                   -1.011
                             9.526
                                   19.689
##
##
  Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 79.6101
                            2.1041
                                   37.836 < 2e-16 ***
                -0.8624
                            0.1448
                                   -5.954 3.66e-07 ***
## Education
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 9.446 on 45 degrees of freedom
## Multiple R-squared: 0.4406, Adjusted R-squared: 0.4282
## F-statistic: 35.45 on 1 and 45 DF, p-value: 3.659e-07
```

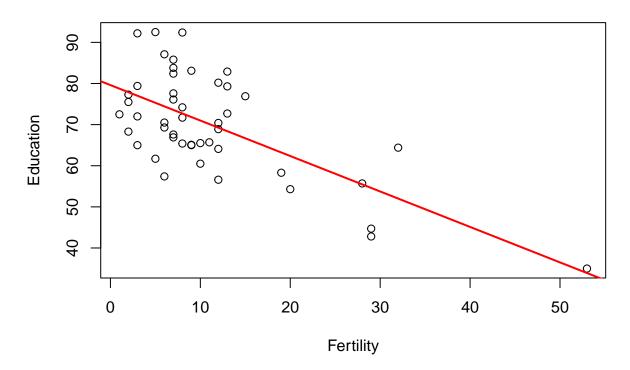
Add the regression line to your scatter plot from the previous exercise.

```
# Create the scatter plot with your specified labels and title
plot(swiss$Education, swiss$Fertility,
    ylab = "Education",
    xlab = "Fertility",
    main = "Scatter Plot of Fertility vs Education"
    )

# Fit the linear regression model
model <- lm(Fertility ~ Education, data = swiss)

# Add the regression line to the existing plot
abline(model, col = "red", lwd = 2)</pre>
```

Scatter Plot of Fertility vs Education



Interpret the summary output of the regression model.

5. Multiple Regression:

Perform a multiple linear regression to predict Fertility based on all other variables. Interpret the summary output of the regression model.

6. Residual Analysis

Create a plot of the residuals from your multiple regression model against the fitted values.

7. Subset Analysis:

Create a subset of the data including only provinces with Agriculture greater than 50. Perform a simple linear regression to predict Fertility based on Education in this subset. Compare the coefficients of this model to the coefficients of the simple linear regression model from exercise 4.