Course 2 Section 3.6 - OPTIMISATION

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```
#load library
library(tidyverse)
library(broom)
```

Give it a go

Continue to develop your understanding of the OLS estimator, optimal fitted line and residuals by making your way through this exercise. You'll need to use a pen or pencil and paper to produce your work, which you can then share with other learners.

Below is a data set with 5 observations for the two variables, x and y:

```
df <- tibble(x=c(1.5, 2, 3.1, 3.9, 5), y=c(3.7, 4.6, 5.5, 5.7, 7.2)) df
```

```
## # A tibble: 5 x 2
##
         х
               У
     <dbl> <dbl>
##
       1.5
             3.7
## 1
## 2
       2
             4.6
## 3
       3.1
             5.5
## 4
       3.9
             5.7
       5
             7.2
## 5
```

Training a linear model of y using this data set produces the following equation:

```
fit_model <- lm(y ~ x, df)
tidy(fit_model)</pre>
```

```
## # A tibble: 2 x 5
##
     term
                  estimate std.error statistic p.value
     <chr>
                     <dbl>
                                <dbl>
                                           <dbl>
                                                   <dbl>
                                            7.01 0.00595
## 1 (Intercept)
                     2.53
                                0.361
                     0.905
                                0.108
                                            8.39 0.00355
\hat{y} = 2.5338 + 0.9052x
```

Q1.compute the fitted values of y, i.e., the prediction of y for each value of x.

```
df <- df %>%
 mutate(y_hat = predict(fit_model))
## # A tibble: 5 x 3
        X
              y y_hat
     <dbl> <dbl> <dbl>
##
           3.7 3.89
## 1
      1.5
## 2
      2
            4.6 4.34
## 3 3.1
           5.5 5.34
           5.7 6.06
## 4 3.9
## 5
            7.2 7.06
Q2.use the predictions to compute the residuals and then the SSR
df <- df %>%
 mutate(residual = residuals(fit_model))
df
## # A tibble: 5 x 4
##
        x y y_hat residual
     <dbl> <dbl> <dbl>
           3.7 3.89
## 1
      1.5
                       -0.192
## 2
      2
            4.6 4.34
                         0.256
## 3
     3.1 5.5 5.34
                       0.16
## 4
     3.9 5.7 6.06
                       -0.364
            7.2 7.06
## 5
      5
                         0.140
# SSR
sum(df$residual ^ 2)
## [1] 0.27998
Repeat the exercise above using the following less optimal fitted model:
\hat{y} = 2.5 + 1.1x
Q3.compute the fitted values of y
df2 \leftarrow tibble(x=c(1.5, 2, 3.1, 3.9, 5), y=c(3.7, 4.6, 5.5, 5.7, 7.2)) %%
 mutate(y_hat = 2.5+1.1*x)
df2
```

A tibble: 5 x 3

X

y y_hat

##

```
## 2 2 4.6 4.7
## 3 3.1 5.5 5.91
## 4 3.9 5.7 6.79
## 5 5 7.2 8
```

Q4.use the predictions to compute the residuals and then the SSR

```
df2 <- df2 %>%
 mutate(residual = y - y_hat)
df2
## # A tibble: 5 x 4
##
             y y_hat residual
       x
##
   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 1.5 3.7 4.15 -0.45
           4.6 4.7
## 2 2
                      -0.1
## 3 3.1 5.5 5.91
                     -0.41
## 4 3.9 5.7 6.79 -1.09
## 5
           7.2 8
     5
                      -0.800
# SSR
sum(df2$residual ^ 2)
```