## Solution Intro-3

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05/07/2019

## Exercise 1:

a) consider the following data - similar to the example used for practical Intro-1 - to generate a data frame named DT:

id	ht	wt	gender
1	155	80	m
2	152	85	$\mathbf{m}$
3	164	45	f
4	175	69	$\mathbf{m}$
5	193	86	f
6	203	110	f
7	190	106	f
8	183	96	$\mathbf{m}$
9	155	90	f
10	169	89	m

```
DT = data.frame(id = 1:10,
                ht=c(155, 152, 164, 175, 193, 203, 190, 183, 155, 169),
                wt=c(80, 85, 45, 69, 86, 110, 106, 96, 90, 89),
                gender=c("m", "m", "f", "m", "f", "f", "f", "m", "f", "m"))
DT
##
      id ht wt gender
       1 155
## 1
              80
## 2
       2 152
              85
       3 164
## 3
              45
       4 175
              69
## 5
       5 193 86
                      f
## 6
       6 203 110
                      f
## 7
       7 190 106
## 8
       8 183 96
                      m
## 9
       9 155
              90
                      f
## 10 10 169 89
```

- b) Add a new categorical variable bmi.grp to the data frame which is defined as follows:
- bmi  $\leq 18.5 \rightarrow$  underweight
- $18.5 < \text{bmi} \le 25 \rightarrow \text{normal}$
- $25 < \text{bmi} \le 30 \rightarrow \text{overweight}$
- $30 < \text{bmi} \rightarrow \text{obesity}$

```
DT$bmi = DT$wt/(DT$ht/100)^2
DT$bmi.grp = ifelse(DT$bmi <= 18.5, "underweight", "obesity")
DT$bmi.grp = ifelse(DT$bmi > 18.5 & DT$bmi <= 25, "normal", DT$bmi.grp)
DT$bmi.grp = ifelse(DT$bmi > 25 & DT$bmi <= 30, "overweight", DT$bmi.grp)</pre>
```

c) Generate a vector **z** consisting of the first 30 elements of the Fibonacci series. By definition, the first two elements of this series are 0 and 1. All further elements are calculated as sum of the preceding two elements, so  $z = 0, 1, 1, 2, 3, 5, \ldots$ 

```
# Manually set the first two elements
z = c(0.1)
for(i in 2:29){
  z[i+1] = z[i] + z[i-1]
}
z
    [1]
              0
                      1
                                     2
                                             3
                                                                   13
                                                                          21
                                                                                  34
                             1
             55
                                   233
                                                                        2584
                     89
                           144
                                          377
                                                  610
                                                          987
                                                                 1597
## [11]
                                                                                4181
## [21]
           6765
                 10946
                         17711
                                 28657
                                        46368
                                                75025 121393 196418 317811 514229
```

## Exercise 2:

a) load the internal R data set mtcars and view its help page to find out abouts its variable description.

```
str(mtcars)
##
   'data.frame':
                    32 obs. of 11 variables:
                 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
   $ mpg : num
   $ cyl : num
                 6 6 4 6 8 6 8 4 4 6 ...
##
##
   $ disp: num
                 160 160 108 258 360 ...
##
   $ hp
         : num
                 110 110 93 110 175 105 245 62 95 123 ...
   $ drat: num
                 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
                 2.62 2.88 2.32 3.21 3.44 ...
##
           num
##
                 16.5 17 18.6 19.4 17 ...
   $ qsec: num
##
   $ vs
         : num
                 0 0 1 1 0 1 0 1 1 1 ...
                 1 1 1 0 0 0 0 0 0 0 ...
##
   $ am : num
```

?mtcars

\$ gear: num

b) Fit a regression model of "Number of car cylinders" on "Miles per gallon" using the function lm(mpg ~ cyl, data=mtcars) and assign the Model for that object. Then show summary of that object.

```
Model = lm(mpg ~ cyl, data=mtcars)
summary(Model)
```

```
##
## Call:
## lm(formula = mpg ~ cyl, data = mtcars)
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
  -4.9814 -2.1185
                   0.2217
                           1.0717
                                    7.5186
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                            2.0738
## (Intercept)
               37.8846
                                     18.27 < 2e-16 ***
## cyl
                                     -8.92 6.11e-10 ***
                -2.8758
                            0.3224
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
```

4 4 4 3 3 3 3 4 4 4 ...

\$ carb: num 4 4 1 1 2 1 4 2 2 4 ...

```
## Residual standard error: 3.206 on 30 degrees of freedom
## Multiple R-squared: 0.7262, Adjusted R-squared: 0.7171
## F-statistic: 79.56 on 1 and 30 DF, p-value: 6.113e-10
```

c) Write a function to extract the effect estimate, standard error and the p-value from a linear regression model. Call the function beta\_se. Hint: use the regression model object as the input and extract summary(model)\$coefficient.

```
# to explore what names can be extracted from the summary object.
names(summary(Model))
                                                          "coefficients"
##
    [1] "call"
                         "terms"
                                         "residuals"
##
    [5] "aliased"
                         "sigma"
                                                          "r.squared"
    [9] "adj.r.squared" "fstatistic"
                                         "cov.unscaled"
# The function
beta_se = function(Simple.Reg.model){
  Summary.model = summary(Simple.Reg.model)
  Out = Summary.model$coefficients[2,c("Estimate", "Std. Error", "Pr(>|t|)")]
 return(Out)
# Implement the `beta_se` function
beta_se(Model)
```

```
## Estimate Std. Error Pr(>|t|)
## -2.875790e+00 3.224089e-01 6.112687e-10
```

d) Use the beta\_se function to extract the effect estimates, standard errors and p-values of the regression models of (1) "Displacement" on "Miles per gallon" and (2) "Rear axle ratio" on "Miles per gallon".

```
Model1 = lm(mpg ~ disp, data = mtcars)
beta_se(Model1)

## Estimate Std. Error Pr(>|t|)
## -4.121512e-02 4.711833e-03 9.380327e-10

Model2 = lm(mpg ~ drat, data = mtcars)
beta_se(Model2)

## Estimate Std. Error Pr(>|t|)
## 7.6782326020 1.5067051076 0.0000177624
```