### Format:

In order to receive full credit for the practice problems, you need to submit an R script file. All the non-R codes need to be placed after the # signs. You also need to write your name on top of the page. Please make sure to include your name as part of the file name.

To complete this assignment, you need to use what you've learned from this course.

You can only ask questions for clarification purposes on the Discussion Board. That is to say, you can't ask questions about how to answer a specific question or help you to debug your program.

# Problems 1

The following code runs regression by using mpg as the outcome and wt as the X variable for each type of cylinder (cyl).

```
> library(tidyverse)
> models <- mtcars %>%
      split(.$cyl) %>%
      map(function(df) lm(mpg ~ wt, data = df))
> models
$`4`
Call:
lm(formula = mpg ~ wt, data = df)
Coefficients:
(Intercept)
                      wt
     39.571
                -5.647
$`6`
Call:
lm(formula = mpg ~ wt, data = df)
Coefficients:
(Intercept)
                      wt
      28.41
                   -2.78
$`8`
lm(formula = mpg ~ wt, data = df)
```

```
Coefficients:
```

```
(Intercept) wt
23.868 -2.192
```

Based on the result above, extracts the coefficients of each model by using a series of map functions. Here's the expected result:

```
4 6 8
-5.647025 -2.780106 -2.192438
```

Hint:

```
> models [[1]]
```

#### Call:

lm(formula = mpg ~ wt, data = df)

### Coefficients:

```
(Intercept) wt 39.571 -5.647
```

> summary(models [[1]])

#### Call:

lm(formula = mpg ~ wt, data = df)

#### Residuals:

```
Min 1Q Median 3Q Max -4.1513 -1.9795 -0.6272 1.9299 5.2523
```

#### Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 39.571 4.347 9.104 7.77e-06 ***
wt -5.647 1.850 -3.052 0.0137 *
```

---

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```
Residual standard error: 3.332 on 9 degrees of freedom \,
```

Multiple R-squared: 0.5086, Adjusted R-squared: 0.454

F-statistic: 9.316 on 1 and 9 DF, p-value: 0.01374

### > summary(models [[1]])\$coefficients

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 39.571196 4.346582 9.103980 7.771511e-06
wt -5.647025 1.850119 -3.052251 1.374278e-02
```

> summary(models [[1]])\$coefficients[2,1]

# [1] -5.647025

# Problems 2

Create a function to calculate the skewness of a given vector based on the formula below.

Skew(x) = 
$$\frac{\frac{1}{n-2} \left( \sum_{i=1}^{n} (x_i - \bar{x})^3 \right)}{\text{Var}(x)^{3/2}}$$
.

Here's the expected output:

> skewness(c(1, 2, 5, 100))

[1] 1.494554

# Problems 3

Create a function, called miss\_count, which counts the number of missing and non-missing values of each variable from a given data. The result should be a matrix or a data frame with first columns contains the number of missing values, and the second column contains number of non-missing values. You will use patient\_tib to test this function. Here's the expected result.

- > load("patient\_tib.RData")
- > patient\_tib

#	A tibbl	le: 10	x 8					
	ID	GLUC	TGL	HDL	LDL	HRT	MMAM	${\tt SMOKE}$
	<fct></fct>	<int></int>	<int></int>	<int></int>	<int></int>	<fct></fct>	<fct></fct>	<fct></fct>
1	Α	88	NA	32	99	Y	<na></na>	ever
2	В	NA	150	60	NA	<na></na>	no	never
3	C	110	NA	NA	120	N	<na></na>	<na></na>
4	D	NA	200	65	165	<na></na>	yes	never
5	E	90	210	NA	150	Y	<na></na>	never
6	F	88	NA	32	210	<na></na>	yes	ever
7	G	120	164	NA	NA	Y	yes	<na></na>
8	H	110	170	70	188	<na></na>	<na></na>	ever
9	I	NA	190	NA	190	N	no	<na></na>
10	J	90	NA	75	NA	<na></na>	yes	never

> miss\_count(patient\_tib)

	$N_{miss}$	N_non_missing
ID	0	10
GLUC	3	7
TGL	4	6
HDL	4	6
LDL	3	7
HRT	5	5
MMAM	4	6
SMOKE	3	7

# Problems 4

For this problem, you need to create a function, called **continuous\_stats**. This function takes only two arguments:

...: one or more variable names in dat (the second argument of this function) that you would like to compute descriptive statistics

dat: the name of the data frame

The function returns a matrix that contains the descriptive statistics, including mean, median, standard deviation (SD), number of non missing values (N\_non\_missing), and number of missing values (N\_missing), for each of the variable. Here are the expected results by calling this function.

> continuous\_stats("LDL", dat=patient\_tib)

```
LDL
Mean 160.28571
Median 165.00000
SD 40.06126
N_non_missing 7.00000
N_missing 3.00000
```

> continuous\_stats("TGL", "LDL", dat=patient\_tib)

```
TGL LDL
Mean 180.6667 160.28571
Median 180.0000 165.00000
SD 23.0362 40.06126
N_non_missing 6.0000 7.00000
N_missing 4.0000 3.00000
```

> continuous\_stats("TGL", "HDL", "LDL", dat=patient\_tib)

```
TGL HDL LDL
Mean 180.6667 55.66667 160.28571
Median 180.0000 62.50000 165.00000
SD 23.0362 19.00175 40.06126
N_non_missing 6.0000 6.00000 7.00000
N_missing 4.0000 4.00000 3.00000
```