



# Quiz 1

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**9/10** points  
earned (90%)

Quiz passed!



1 / 1  
points

1.

Consider the data set given below

```
1 x <- c(0.18, -1.54, 0.42, 0.95)
```

And weights given by

```
1 w <- c(2, 1, 3, 1)
```

Give the value of  $\mu$  that minimizes the least squares equation

$$\sum_{i=1}^n w_i (x_i - \mu)^2$$



1.077



0.1471



**Correct Response**

```
1 sum(x * w)/sum(w)
```

```
1 ## [1] 0.1471
```



☐ 0.0025

☐ 0.300



1 / 1  
points

2.

Consider the following data set

```
1 x <- c(0.8, 0.47, 0.51, 0.73, 0.36, 0.58, 0.57, 0.85, 0.44, 0.42)
2 y <- c(1.39, 0.72, 1.55, 0.48, 1.19, -1.59, 1.23, -0.65, 1.49, 0.05)
```

Fit the regression through the origin and get the slope treating y

as the outcome and x as the regressor. (Hint, do not center the data since we want regression through the origin, not through the means of the data.)

☒ 0.8263



**Correct Response**

```
1 coef(lm(y ~ x - 1))
```

```
1 ##      x
2 ## 0.8263
```

```
1 sum(y * x)/sum(x^2)
```

```
1 ## [1] 0.8263
```

☐ 0.59915

☐ -0.04462

☐ -1.713

1 / 1  
points

3.

Do `data(mtcars)` from the `datasets` package and fit the regressionmodel with `mpg` as the outcome and `weight` as the predictor. Give

the slope coefficient.

☐ 30.2851☐ -9.559☐ 0.5591☒ -5.344**Correct Response**

```
1 data(mtcars)
2 summary(lm(mpg ~ wt, data = mtcars))
```

```
1 ##
2 ## Call:
3 ## lm(formula = mpg ~ wt, data = mtcars)
4 ##
5 ## Residuals:
6 ##      Min       1Q   Median       3Q      Max
7 ## -4.543 -2.365 -0.125  1.410  6.873
8 ##
9 ## Coefficients:
10 ##              Estimate Std. Error t value Pr(>|t|)
11 ## (Intercept)   37.285      1.878   19.86 < 2e-16 ***
12 ## wt           -5.344      0.559   -9.56 1.3e-10 ***
13 ## ---
14 ## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
15 ##
16 ## Residual standard error: 3.05 on 30 degrees of freedom
17 ## Multiple R-squared:  0.753, Adjusted R-squared:  0.745
18 ## F-statistic: 91.4 on 1 and 30 DF, p-value: 1.29e-10
```

```
1 attach(mtcars)
2 cor(mpg, wt) * sd(mpg)/sd(wt)
```

```
1 ## [1] -5.344
```

```
1 detach(mtcars)
```

0 / 1  
points

4.

Consider data with an outcome ( $Y$ ) and a predictor ( $X$ ). The standard deviation of the predictor is one half that of the outcome. The correlation between the two variables is .5. What value would the slope coefficient for the regression model with  $Y$  as the outcome and  $X$  as the predictor?



1



3



4

**Incorrect Response**

0.25

1 / 1  
points

5.

Students were given two hard tests and scores were normalized to have empirical mean 0 and variance 1. The correlation between the scores on the two tests was 0.4. What would be the expected score on Quiz 2 for a student who had a normalized score of 1.5 on Quiz 1?



0.16



0.6

**Correct Response**

This is the classic regression to the mean problem. We are expecting the score to get multiplied by 0.4. So

```
1 ## [1] 0.0
```

☐ 0.4

☐ 1.0



1 / 1  
points

6.

Consider the data given by the following

```
1 x <- c(8.58, 10.46, 9.01, 9.64, 8.86)
```

What is the value of the first measurement if x were normalized (to have mean 0 and variance 1)?

☐ 8.86

☒ -0.9719

**Correct Response**

```
1 ((x - mean(x))/sd(x))[1]
```

```
1 ## [1] -0.9719
```

☐ 9.31

☐ 8.58



1 / 1  
points

7.

Consider the following data set (used above as well). What is the intercept for

Consider the following data set (used above as well). What is the intercept for fitting the model with x as the predictor and y as the outcome?

```
1 x <- c(0.8, 0.47, 0.51, 0.73, 0.36, 0.58, 0.57, 0.85, 0.44, 0.42)
2 y <- c(1.39, 0.72, 1.55, 0.48, 1.19, -1.59, 1.23, -0.65, 1.49, 0.05)
```

☐ -1.713

☐ 1.252

☐ 2.105

☒ 1.567

**Correct Response**

```
1 coef(lm(y ~ x))[1]
```

```
1 ## (Intercept)
2 ##          1.567
```



1 / 1  
points

8.

You know that both the predictor and response have mean 0. What can be said about the intercept when you fit a linear regression?

☐ It is undefined as you have to divide by zero.

☒ It must be identically 0.

**Correct Response**

The intercept estimate is  $\bar{Y} - \beta_1 \bar{X}$  and so will be zero.

☐ It must be exactly one.

☐ Nothing about the intercept can be said from the information given.

1 / 1  
points

9.

Consider the data given by

```
1 x <- c(0.8, 0.47, 0.51, 0.73, 0.36, 0.58, 0.57, 0.85, 0.44, 0.42)
```

What value minimizes the sum of the squared distances between these points and itself?



0.8



0.573

**Correct Response**

This is the least squares estimate, which works out to be the mean in this case.

```
1 mean(x)
```

```
1 ## [1] 0.573
```



0.44



0.36


1 / 1  
points

10.

Let the slope having fit  $Y$  as the outcome and  $X$  as the predictor be denoted as  $\beta_1$ . Let the slope from fitting  $X$  as the outcome and  $Y$  as the predictor be denoted as  $\gamma_1$ . Suppose that you divide  $\beta_1$  by  $\gamma_1$ ; in other words consider  $\beta_1/\gamma_1$ . What is this ratio always equal to?

 $Cor(Y, X)$ 

1

  $Var(Y)/Var(X)$

**Correct Response**

The  $\beta_1 = Cor(Y, X)SD(Y)/SD(X)$  and  
 $\gamma_1 = Cor(Y, X)SD(X)/SD(Y)$ .

Thus the ratio is then  $Var(Y)/Var(X)$ .

☐  $2SD(Y)/SD(X)$

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