

## Quiz 6

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### R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <http://rmarkdown.rstudio.com>.

- Physicians depend on laboratory test results when managing medical problems such as diabetes or epilepsy. In a uniformity test for glucose tolerance, three different laboratories were each sent  $n = 5$  identical blood samples from a person who had drunk 50 milligrams (mg) of glucose dissolved in water. The laboratory results (in mg/dl) are listed here:

For each  $i = 1, 2, 3$ , let  $\mu_i$  be the mean of test results (in mg/dl) from Lab  $i$ .

The null hypothesis is  $H_0: \mu_1 = \mu_2 = \mu_3$ .

The alternative hypothesis is  $H_1$ : At least one  $\mu_i$  is different from others.

```
Lab1 <- c(121.3, 111.9, 110.1, 105.4, 101.6)
Lab2 <- c(99.5, 113.2, 108.9, 109.1, 100.4)
Lab3 <- c(104.2, 109.7, 102.3, 111.2, 106.6)
DAT2 <- cbind.data.frame("Lab1"=Lab1, "Lab2"=Lab2, "Lab3"=Lab3)
DAT2
```

```
##   Lab1  Lab2  Lab3
## 1 121.3  99.5 104.2
## 2 111.9 113.2 109.7
## 3 110.1 108.9 102.3
## 4 105.4 109.1 111.2
## 5 101.6 100.4 106.6
```

```
DAT2stacked=stack(DAT2)
DAT2stacked
```

```
##   values  ind
## 1  121.3 Lab1
## 2  111.9 Lab1
## 3  110.1 Lab1
## 4  105.4 Lab1
## 5  101.6 Lab1
## 6   99.5 Lab2
## 7  113.2 Lab2
## 8  108.9 Lab2
## 9  109.1 Lab2
```

```
## 10 100.4 Lab2
## 11 104.2 Lab3
## 12 109.7 Lab3
## 13 102.3 Lab3
## 14 111.2 Lab3
## 15 106.6 Lab3

mod <- aov(values ~ as.factor(ind), data=DAT2stacked)
summary(mod)

##              Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(ind)  2  42.8    21.42    0.611  0.559
## Residuals      12 421.1    35.09
```

From the anova table, p-value is 0.559 which is greater than 0.05. Then we do not reject the null hypothesis  $H_0$ . Then there is NO difference in average readings for the three laboratories.

3. A professor of economics wants to study the relationship between income (y in \$1000s) and education (x in years). A random sample of eight individuals is taken and the results are shown below.

```
Education <- c(16,11,15,8,12,10,13,14)
Income <- c(58,40,55,35,43,41,52,49)
mod3 <- lm(Income ~ Education)
summary(mod3)

##
## Call:
## lm(formula = Income ~ Education)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.624 -2.398  0.782  1.150  3.556
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  10.6165     4.3539   2.438 0.050583 .
## Education     2.9098     0.3449   8.437 0.000151 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.436 on 6 degrees of freedom
## Multiple R-squared:  0.9223, Adjusted R-squared:  0.9093
## F-statistic: 71.18 on 1 and 6 DF, p-value: 0.0001513
```

Determine the coefficient of determination.  $R^2 = 0.9223$ .

Discuss what its value tells you about the two variables.

92.03 % of the variation in income is explained by the variation in years of education.

Calculate the Pearson correlation coefficient.

```
sqrt(0.9223)
## [1] 0.9603645
```

$r = 0.9603645$ .

Why does it have the sign it has?

It has a positive sign since the slope of the regression line ( $= 2.9098$ ) is positive.

What is the equation of the estimated regression line?

$Income = 10.6165 + 2.9098 \times Education$

Is there a linear relationship between education and income? Explain?

Yes, since  $R^2$  close to 1.

4. A study was conducted to determine the effect of extra help sessions attended on students ability to avoid mistakes on a 20- multiple choice test. The data shown below represent the number of extra help sessions attended ( $x$ ) and the average number of mistakes ( $y$ ) recorded.

```
x <- 1:6
y <- c(6.1,5.1,5.0,4.2,3.7,3.2)
mod4 <- lm(y ~ x)
summary(mod4)

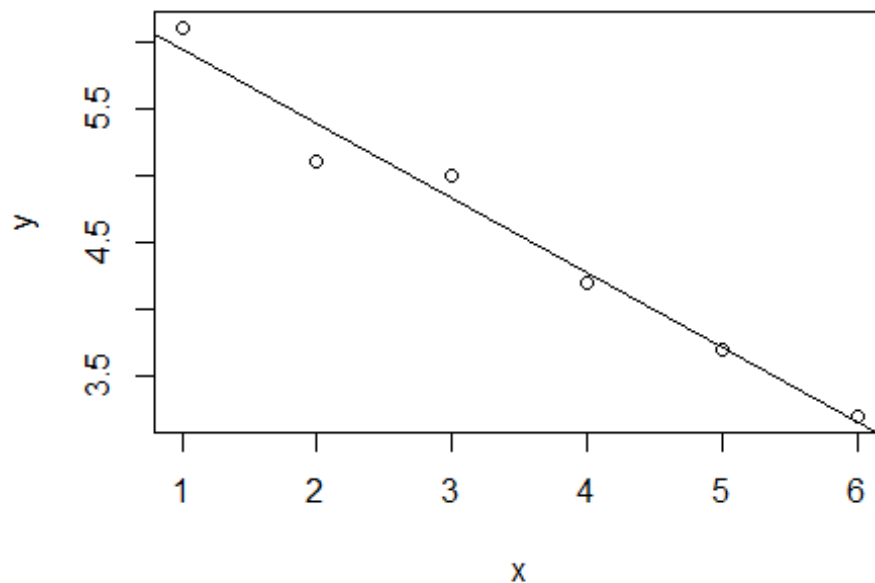
##
## Call:
## lm(formula = y ~ x)
##
## Residuals:
##      1      2      3      4      5      6
## 0.15714 -0.28571  0.17143 -0.07143 -0.01429  0.04286
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.50000    0.17593   36.95  3.2e-06 ***
## x           -0.55714    0.04518  -12.33 0.000248 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.189 on 4 degrees of freedom
## Multiple R-squared:  0.9744, Adjusted R-squared:  0.968
## F-statistic: 152.1 on 1 and 4 DF,  p-value: 0.0002484
```

Use the regression formulas to find the least-squares line for the data.

$$\hat{y} = 6.50000 - 0.55714 \times x$$

Plot the six points and graph the line.

```
plot( y ~ x)
abline( a=6.50000,b=-0.55714)
```



Does the line appear to provide a good fit to the data points?

Yes.

Use the least-Squares line to predict the value of y when x = 3.5.

```
6.50000-0.55714*3.5
```

```
## [1] 4.55001
```

Do the data provide sufficient evidence to indicate that y and x are linearly related at the 1% level of significance?

Yes, since  $p - value = 0.000248 ***$ .

The Multiple R-squared is 0.9744. Then 97.44 % of the variation of y can be explained by the variation in the x - value.