

STAT537: Statistics for Research I: HW#7

Due on Sep. 20, 2016

Dr. Schmidhammer TR 11:10am – 12:25pm

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Problem 1

Homework on Correlation Coefficients

Solution. 1. Generate a scatterplot for these data, with grain size on the horizontal axis, and yield on the vertical axis.

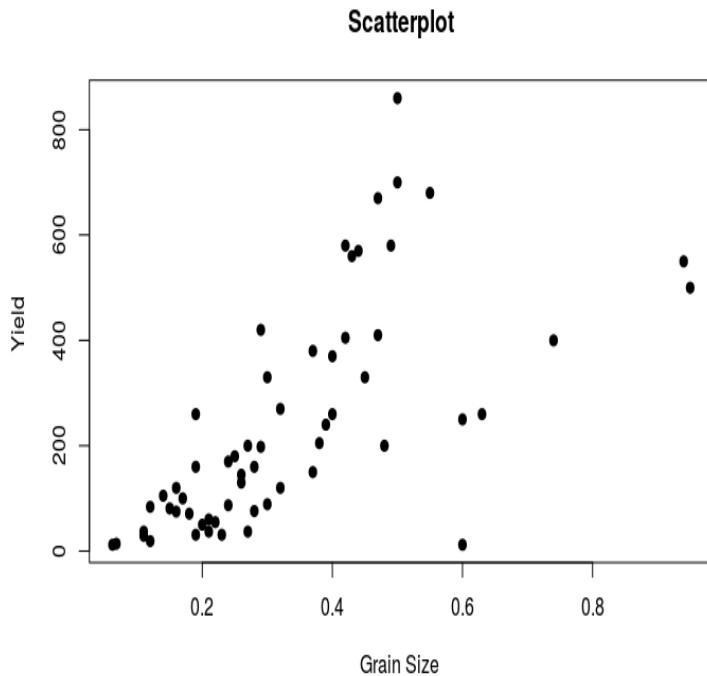


Figure 1: Scatterplot for these data, with grain size on the horizontal axis, and yield on the vertical axis.

2. Compute the Pearson Product Moment correlation coefficient r and Spearmans rho

(a) Pearson Product Moment correlation coefficient $r = 0.667871$

```
> cor(GrainSize, Yield, method="pearson")
[1] 0.667871
```

(b) Spearman's rho $\rho = 0.7634203$

```
> cor(GrainSize, Yield, method="spearman")
[1] 0.7634203
```

3. Test the hypothesis:

(a) Pearson Product Moment correlation coefficient r : Since the p-value = $7.543e - 09 < 0.05$, so reject H_0 . Hence we can say that we have enough evidence to believe H_1 , i.e. we have enough evidence to believe that the Grain Size values are correlated with the Yield value at 95% level.

```
> out
```

Pearson's product-moment correlation

```
data: GrainSize and Yield
t = 6.7748, df = 57, p-value = 7.543e-09
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
0.4967473 0.7890091
sample estimates:
cor
0.667871
```

- (b) Spearmans rho: Since the p-value = $2.059e-12 < 0.05$, so reject H_0 . Hence we can say that we have enough evidence to believe H_1 , i.e. we have enough evidence to believe that the Grain Size values are correlated with the Yield value at 95% level.

Spearman's rank correlation rho

```
data: GrainSize and Yield
S = 8095.8, p-value = 2.059e-12
alternative hypothesis: true rho is not equal to 0
sample estimates:
rho
0.7634203
```

4. Construct a 95% Confidence Interval for ρ .

- (a) 95 percent confidence interval: [0.630624, 0.8527845]

```
> CIrho(out2$estimate, N)
      rho    2.5 %    97.5 %
[1,] 0.7634203 0.630624 0.8527845
```

□

Problem 2

Homework on Simple Linear Regression

Solution. 1. Generate a scatterplot for these data, with grain size on the horizontal axis, and yield on the vertical axis.

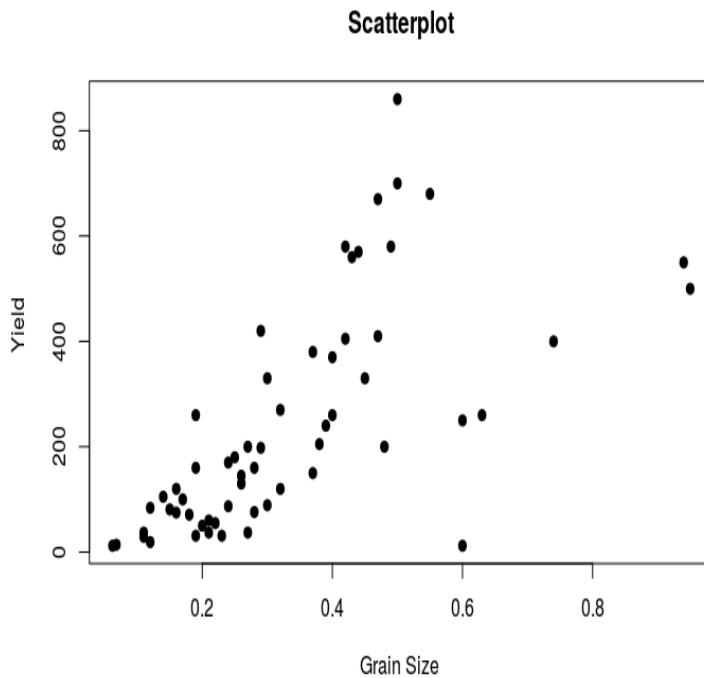


Figure 2: Scatterplot for these data, with grain size on the horizontal axis, and yield on the vertical axis.

2. Find the least squares estimates of β_0 and β_1 in the model: $\beta_0 = -9.294$ and $\beta_1 = 744.979$.

```
> fit
```

```
Call:
lm(formula = Yield ~ Grainsize)

Coefficients:
(Intercept)      Grainsize
-9.294          744.979
```

3. Test the hypothesis: From the following summary, we can see that the p-value for β_1 is $7.543e-09 < 0.05$, Hence reject H_0 . Therefore we can say that we have enough evidence to believe H_1 , i.e. $\beta_1 \neq 0$.

```
> summary(fit)
```

```
Call:
lm(formula = Yield ~ Grainsize)

Residuals:
    Min      1Q  Median      3Q     Max 
-425.69 -100.43 -28.70  55.03 496.80 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) -9.294     1.000  -9.294 7.543e-09 ***
Grainsize    744.979   10.000  74.498 <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Intercept) -9.294      42.255   -0.220     0.827
GrainSize    744.979     109.964    6.775  7.54e-09 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 159.4 on 57 degrees of freedom
 Multiple R-squared: 0.4461, Adjusted R-squared: 0.4363
 F-statistic: 45.9 on 1 and 57 DF, p-value: 7.543e-09

4. Compute the residuals for these data. Do any residuals exceed $\pm 3s_\epsilon$?

- (a) Residual:

```
> res = fit$residuals
> res
      1          2          3          4          5          6
-24.8948554 -27.3647300 -43.6538518 -35.6538518 -61.1036427  3.8963573
      7          8          9         10         11         12
  9.9967755 -21.4530154 -34.9028063  10.0971937 -17.3525972 -53.8023882
      13         14         15         16         17         18
-101.2521791  27.7478209 127.7478209 -89.7019700 -110.1517609 -87.1517609
      19         20         21         22         23         24
-99.6015518 -131.0513427 -82.5011336   0.4988664   3.0490755 -54.4007154
      25         26         27         28         29         30
-39.4007154 -154.8505063   8.1494937 -123.3002972 -39.3002972 -8.7500881
      31         32         33         34         35         36
-125.1998790  213.2499119 115.8001210 -109.0994609  40.9005391 -116.3484154
      37         38         39         40         41         42
  113.6515846 -68.7982063 -41.2479972 -28.6977881  81.3022119 101.4026301
      43         44         45         46         47         48
  276.4026301  248.9528391 251.5030482   4.0532573  69.1536755 329.1536755
      49         50         51         52         53         54
-148.2961154  224.2540937 336.8043028  496.8043028 279.5553483 -425.6936063
      55         56         57         58         59
-187.6936063 -200.0429790 -141.9906790 -140.9864972 -198.4362881
```

- (b) Since

$$s_\epsilon^2 = \frac{\sum_1^n (y_i - \hat{y}_i)^2}{n - 2}$$

hence

$$s_\epsilon = \sqrt{\frac{\sum_1^n (y_i - \hat{y}_i)^2}{n - 2}} = \sqrt{\frac{\sum_1^n \text{residual}^2}{n - 2}} = 159.3766.$$

Therefore $\pm 3s_\epsilon = [-478.1297, 478.1297]$.

- (c) Check: From the following check table, we can see that the 52-th residual (496.8043028) is not in the range.

```
> check = checkRange(res, -3*s_eps, 3*s_eps)
[1] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
[8] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
[15] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
```

```
[22] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
[29] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
[36] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
[43] TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE  TRUE
[50] TRUE  TRUE FALSE TRUE  TRUE  TRUE  TRUE  TRUE
[57] TRUE  TRUE  TRUE
```

□

Appendix

R code for HW#7

Listing 1: Source code for problem 1

```
rm(list = ls())
# set the path or environment
setwd("/home/feng/Dropbox/UTK_Course/Stat537/HW#7/HW#7/code")

#install.packages("readxl") # CRAN version
library(readxl)
#install.packages("moments")
library(moments)
rawdata = read_excel("Data.xlsx", sheet = 1)
attach(rawdata)

plot(GrainSize, Yield, main="Scatterplot",
      xlab="Grain Size ", ylab="Yield ", pch=19)

#correlation
cor(GrainSize, Yield, method="pearson")
cor(GrainSize, Yield, method="spearman")

#install.packages("Hmisc")
library(Hmisc)
out1<-cor.test(GrainSize,Yield,method ="pearson",conf.level=0.95)
out1
out2<-cor.test(GrainSize,Yield,method ="spearman",conf.level=0.95)
out2
#install.packages("mada")
library(mada)
N = dim(rawdata)[1]
CIrho(out2$estimate,N)

# regression

fit <- lm(Yield ~ GrainSize)
fit
summary(fit)

res= fit$residuals
s_eps = sqrt(sum(res^2)/(length(res)-2))
```

```
40 s_eps
range =c (-3*s_eps, 3*s_eps)
range
i=length (res)

45 checkRange <- function (data, lower, upper) {
  n = length (data)
  result = logical (length = n)
  for (i in 1:n){
    result[i]= data[i] >= lower && data[i] <= upper
  }
  print (result)
}

50 check = checkRange(res,-3*s_eps,3*s_eps)

55 a= -3*s_eps <= res
b= res <= 3*s_eps
a&&b
```