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CA4 Opt RainGrad.csv

#Question 1

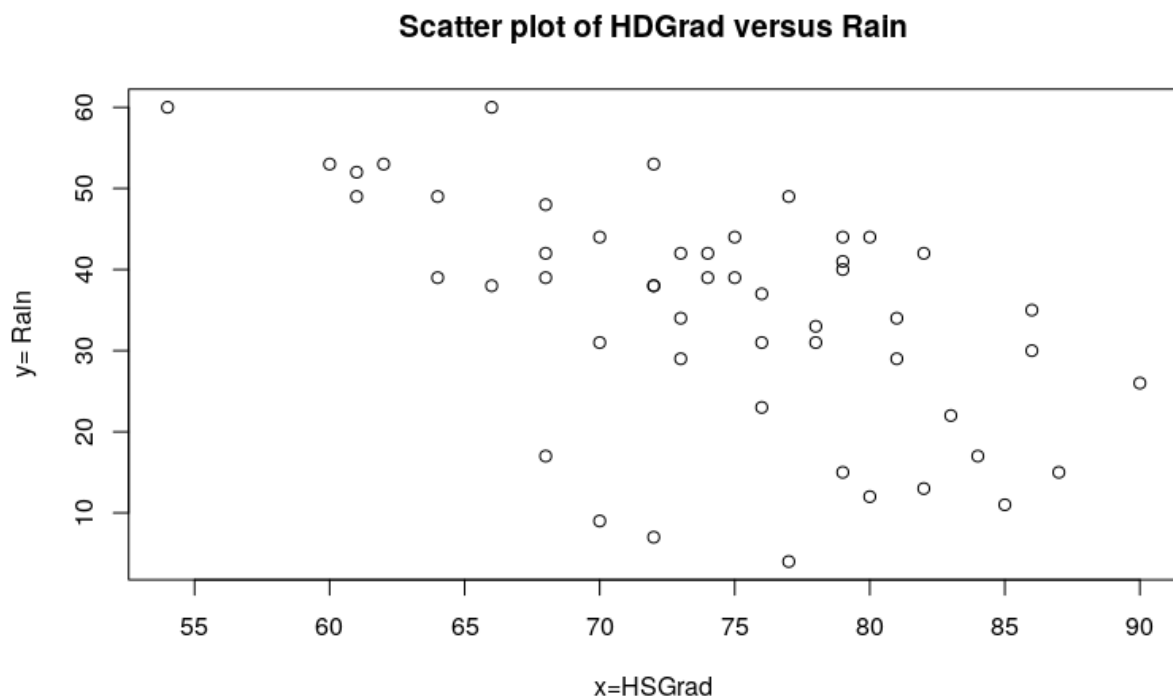
library(readr)

CA4_Opt_RainGrad <- read_csv("SharedFiles/ST204/Data/CA4_Opt_RainGrad.csv")

CA4_Opt_RainGrad

#(a) Generate a scatter plot of HDGrad versus Rain.

plot(CA4_Opt_RainGrad\$HSGRad, CA4_Opt_RainGrad\$Rain, ylab="y= Rain", xlab="x=HSGRad",
main="Scatter plot of HDGrad versus Rain")



#(b) Estimate Pearson's correlation, Spearman's rank correlation, and Kendall's tau for these data

#and perform appropriate two-sided hypothesis tests for each.

#State clearly the null and alternative hypotheses as well as your conclusions in each case.

with(CA4_Opt_RainGrad, cor.test(HSGRad, Rain, method="pearson")) #We reject the hypothesis as
p-value = 2.267e-05 is less than 0.05

with(CA4_Opt_RainGrad, cor.test(HSGRad, Rain, method="spearman")) #We reject the hypothesis as
p-value = 2.863e-05 is less than 0.05

with(CA4_Opt_RainGrad, cor.test(HSGRad, Rain, method="kendall")) #We reject the hypothesis as p-
value = 4.47e-05 is less than 0.05

#(c) Briefly discuss the statement 'correlation does not imply causation' in the context of these data.

#Just because it looks like people did better with less rainfall, doesn't imply that it is because of the rain.

#It could be better schooling, facilities or better textbooks.

#So just because it looks like people did better where there was less rainfall, doesn't mean it is BECAUSE there was less rainfall.

#(d) Let r and r_s denote Pearson's and Spearman's correlation coefficients, respectively. Obtain

#these values in R and use them to verify by hand the values of the associated test statistics t and S . Show your workings.

$S = 34263$, $t = -4.6837$

```
y<- c(CA4_Opt_RainGrad$Rain)
```

```
x<- c(CA4_Opt_RainGrad$HSGrad)
```

```
sort(x)
```

```
sort(y)
```

```
sum(y)
```

```
sum(x)
```

```
mean(x)
```

```
mean(y)
```

```
meany <- 34.63
```

```
meanx <- 74.24
```

```
meany
```

```
meanx
```

```
1:51
```

```
sum((y - meany)^2) * sum((x - meanx)^2)
```

```
sum((x - meanx) * (y - meany))
```

```
sum((x - meanx)^2) * sum((y - meany)^2)
```

```
with(CA4_Opt_RainGrad, cor.test(HSGrad, Rain, method="pearson")) #0.56 = r
```

```
with(CA4_Opt_RainGrad, cor.test(HSGrad, Rain, method="spearman")) #0.55 = rho
```

X=H660 Y=Rain		X=H660 Y=Rain	
1	85 66	33	65 64
2	185 72	34	115 68
3	185 72	35	50 87
4	315 77	36	185 72
5	115 68	37	29 76
6	265 75	38	185 72
7	265 79	39	365 79
8	115 68	40	295 74
9	2 60	41	35 61
10	35 61	42	46 84
11	65 64	43	115 68
12	29 76	44	85 66
13	265 80	45	365 79
14	315 78	46	415 81
15	265 75	47	265 74
16	185 86	48	22 73
17	115 81	49	385 78
18	15 70	50	45 83
19	1 54	51	415 82
20	365 79		
21	22 73		
22	115 80		
23	15 70		
24	51 90		
25	5 62		
26	22 73		
27	47 85		
28	48 86		
29	315 77		
30	29 76		
31	115 82		
32	15 70		

Pearson correlation coefficient (r)

$$r = \frac{\sum (X - \bar{X})(Y - \bar{Y})}{\sqrt{\sum (X - \bar{X})^2 \sum (Y - \bar{Y})^2}}$$

sum of X = 3786

sum of Y = 1766

$$\bar{X} = \frac{3786}{32} = 118.3125$$

$$\bar{Y} = \frac{1766}{32} = 55.1875$$

Spearman

$$r = 1 - \frac{6 \sum d^2}{n^3 - n} = 1 - \frac{6 \sum d^2}{132651 - 51} = 1 - \frac{34263}{132600} = 1 - 0.2584 = 0.7416$$

$$d = \text{sum of } R(X) - R(Y) = 10.245 - 0.55 = 9.695$$

Number beside number = Rank

	$(x-\bar{x})$	$(y-\bar{y})$	$(x-\bar{x})^2$	$(y-\bar{y})^2$	$(x-\bar{x})(y-\bar{y})$
1	-8.24	25.37	-209.05	6790	693.69
2	-2.24	18.37	-41.15	802	337.46
3	-2.24	-27.63	-61.94	502	763.92
4	2.76	14.37	39.66	762	206.56
5	-6.24	-17.63	110.01	3894	310.82
6	0.76	9.37	7.12	058	87.80
7	4.76	6.37	30.32	22.66	40.58
8	-6.24	4.37	-27.27	38.94	19.10
9	-14.24	18.37	-261.59	202.78	337.46
10	-13.24	14.37	-190.26	175.30	206.50
11	-10.24	14.37	-147.15	104.86	135.26
12	1.76	-11.63	-20.47	3.10	82.13526
13	5.76	-22.63	-130.35	33.18	512.12
14	3.76	-1.63	-6.13	14.14	2.66
15	0.76	4.37	3.32	0.58	19.10
16	11.76	0.37	4.35	138.30	0.14
17	6.76	-5.63	-38.06	45.70	31.70
18	-4.24	9.37	-34.73	17.98	87.80
19	-10.24	25.37	-513.49	409.66	673.69
20	4.76	9.37	44.60	22.66	87.80
21	-1.24	7.37	-9.14	1.54	54.32
22	5.76	9.37	53.97	33.18	87.80
23	-4.24	-3.63	15.39	17.98	13.18
24	15.76	-8.63	-136.01	248.38	74.48
25	-12.24	18.37	-224.85	149.82	337.46
26	-1.24	-0.63	0.78	1.54	0.40
27	10.76	-23.63	-254.26	115.78	552.38
28	11.76	-4.63	-59.45	138.30	21.44
29	2.76	-20.61	-84.54	7.62	917.820
30	1.76	2.37	4.17	3.10	5.62
31	7.76	7.37	57.19	60.22	54.32
32	-4.24	-25.63	108.67	17.98	656.90

33
 34
 35
 36
 37
 38
 39
 40
 41
 42
 43
 44
 45
 46
 47
 48
 49
 50
 51

	$(X - \bar{X})$	$(Y - \bar{Y})$	$(X - \bar{X}) * (Y - \bar{Y})$	$(X - \bar{X})^2$	$(Y - \bar{Y})^2$
33	-10.24	4.37	-44.75	104.86	19.10
34	-6.24	7.37	-45.99	38.94	54.32
35	12.76	-19.63	-250.48	162.82	385.34
36	-2.24	3.37	-7.55	5.02	11.36
37	1.76	-3.63	-6.39	3.10	13.18
38	-2.24	3.37	-7.55	5.02	11.36
39	4.76	5.37	25.56	22.66	28.84
40	-0.24	7.37	-1.77	0.06	54.32
41	-13.24	17.37	-229.98	175.30	301.72
42	9.76	-17.63	-172.07	95.26	310.82
43	-6.24	13.37	-83.43	38.94	178.76
44	-8.24	3.37	-27.77	67.90	11.36
45	4.76	-19.63	-93.44	22.66	385.34
46	6.76	-0.63	-4.26	45.70	0.40
47	-0.24	4.37	-1.05	0.06	19.10
48	-1.24	-5.63	6.98	1.54	31.70
49	3.76	-3.63	-13.65	14.14	13.18
50	8.76	-12.63	-110.64	76.74	159.52
51	7.76	-21.63	-167.85	60.22	467.86

$$\Sigma = -3082.53 \quad \Sigma = 3091.18 \quad \Sigma = 9939.92$$

$$B \quad r = \frac{\Sigma((X - \bar{X})(Y - \bar{Y}))}{\sqrt{\Sigma(X - \bar{X})^2 \Sigma(Y - \bar{Y})^2}} \quad r = \frac{-3082.53}{\sqrt{3091.18 \times 9939.92}} = \frac{-3082.53}{\sqrt{30726000}} = -0.556$$

	d	d ²
1	-42	1764
2	-29.5	870.25
3	16.5	272.25
4	-12.5	156.25
5	2	4
6	-13	169
7	3.5	12.25
8	-18	324
9	-46	2116
10	-40.5	1640.25
11	-37.5	1406.25
12	17	289
13	34.5	1190.25
14	13.5	182.25
15	-3	9
16	25.5	650.25
17	30	900
18	-24.5	600.25
19	-44.5	1980.25
20	-3	9
21	-13.5	182.25
22	0	0
23	-3	9
24	38	1444
25	-43	1849
26	0.5	.25
27	43	1849
28	32.5	1056.25
29	30.5	930.25
30	5	25

	d	d ²
31	8	64
32	12	144
33	-23	529
34	42.5	1806.25
35	-7.5	56.25
36	11	121
37	-7.5	56.25
38	4.5	20.25
39	-11	121
40	-11	121
41	36.5	1332.25
42	-30.5	930.25
43	-17.5	306.25
44	29	841
45	20	400
46	-5	25
47	7.5	56.25
48	15.5	240.25
49	34	1156
50	37.5	1406.25
51	-24	576

~~180~~ ~~34263~~
~~185~~ ~~34263~~

$85 - 50.5 = -42$
 $18.5 - 48 = -29.5$
 $18.5 - 2 = 16.5$
 $31.5 - 44 = -12.5$
 $11.5 - 9.5 = 2$
 $26.5 - 39.5 = -13$
 $36.5 - 33 = 3.5$
 $11.5 - 29.5 = -18$
 $2 - 48 = -46$
 $3.5 - 44 = -40.5$
 $6.5 - 47 = -37.5$
 $29 - 12 = 17$
 $39.5 - 5 = 34.5$
 $33.5 - 20 = 13.5$
 $26.5 - 29.5 = -3$
 $48.5 - 23 = 25.5$
 $44.5 - 14.5 = 30$
 $15 - 34.5 = -19.5$
 $1 - 50.5 = -49.5$
 $36.5 - 39.5 = -3$
 $22 - 35.5 = -13.5$
 $39.5 - 39.5 = 0$
 $15 - 18 = -3$
 $51 - 13 = 38$
 $5 - 48 = -43$
 $22 - 21.5 = 0.5$
 $47 - 4 = 43$
 $48.5 - 16 = 32.5$
 $31.5 - 1 = 30.5$
 $29 - 24 = 5$
 $43.5 - 35.5 = 8$
 $15 - 3 = 12$

$6.5 - 29.5 = -23$
 $11.5 - 35.5 = -24$
 $50 - 7.5 = 42.5$
 $18.5 - 26 = -7.5$
 $29 - 18 = 11$
 $18.5 - 26 = -7.5$
 $36.5 - 32 = 4.5$
 $29.5 - 35.5 = -6$
 $35 - 46 = -11$
 $46 - 9.5 = 36.5$
 $11.5 - 42 = -30.5$
 $8.5 - 26 = -17.5$
 $36.5 - 7.5 = 29$
 $41.5 - 21.5 = 20$
 $24.5 - 29.5 = -5$
 $22 - 14.5 = 7.5$
 $37.5 - 18 = 19.5$
 $45 - 11 = 34$
 $43.5 - 6 = 37.5$

~~34263~~
~~34263~~
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$25 \times 25 = 625$

#Question 2: The following data set was taken from a bivariate (X, Y) population.

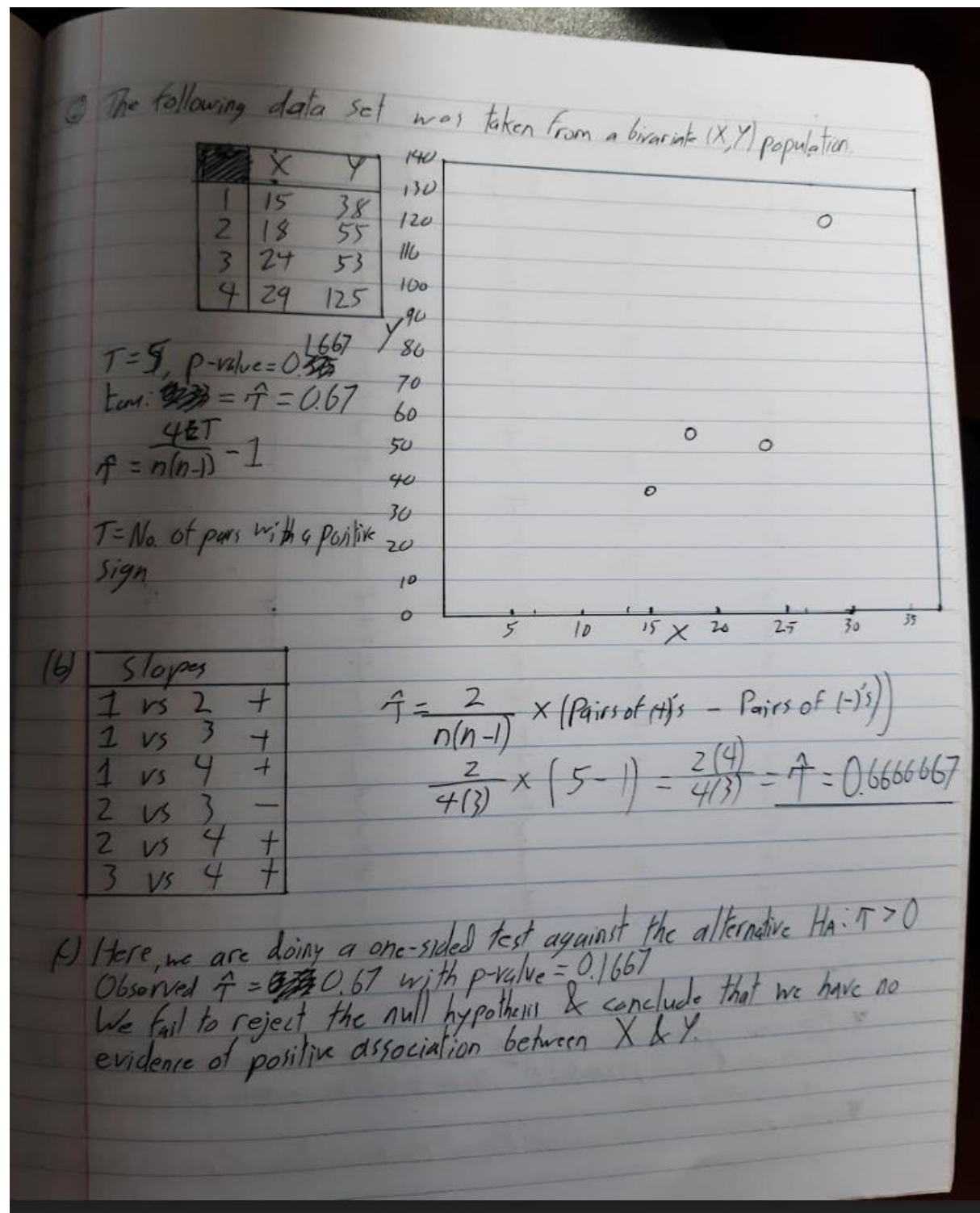
A1 <- c(15,18,24,29)

A2 <- c(38,55,53,125)

#For observed data, $r^* = 0.88129394$, 24 permutations.

#(a) Sketch, by hand, a scatter plot of Y versus X.

#(b) Estimate Kendall's tau for these data by hand.



→ (15, 18, 29, 29)

$R(X_1)$	$R(X_2)$	$R(X_3)$	$R(X_4)$	Association \hat{r}^*
1 (15)	2 (18)	3 (29)	4 (29)	Estimates
1 (38)	2 (53)	3 (55)	4 (125)	* 0.88
1 (38)	2 (53)	4 (125)	3 (55)	0.40
1 (38)	3 (55)	2 (53)	4 (125)	* 0.86
1 (38)	3 (55)	4 (125)	2 (53)	0.37
1 (38)	4 (125)	2 (53)	3 (55)	-0.19
1 (38)	4 (125)	3 (55)	2 (53)	-0.20
2 (53)	1 (38)	3 (55)	4 (125)	* 0.82
2 (53)	1 (38)	4 (125)	3 (55)	0.34
2 (53)	3 (55)	1 (38)	4 (125)	-0.66
2 (53)	3 (55)	4 (125)	1 (38)	0.08
2 (53)	4 (125)	1 (38)	3 (55)	-0.37
2 (53)	4 (125)	3 (55)	1 (38)	-0.44
3 (55)	1 (38)	2 (53)	4 (125)	* 0.79
3 (55)	1 (38)	4 (125)	2 (53)	0.30
3 (55)	2 (53)	1 (38)	4 (125)	-0.66
3 (55)	2 (53)	4 (125)	1 (38)	0.07
3 (55)	4 (125)	1 (38)	2 (53)	-0.41
3 (55)	4 (125)	2 (53)	1 (38)	-0.51
4 (125)	1 (38)	2 (53)	3 (55)	-0.54
4 (125)	1 (38)	3 (55)	2 (53)	-0.56
4 (125)	2 (53)	1 (38)	3 (55)	-0.67
4 (125)	2 (53)	3 (55)	1 (38)	-0.78
4 (125)	3 (55)	1 (38)	2 (53)	-0.70
4 (125)	3 (55)	2 (53)	1 (38)	-0.81

1, 2, 3, 4 = (38, 53, 55, 128)

For this upper-tailed test, count amount of permuted correlations $\geq \hat{r}^* = 0.88$. There are 4 of 24 permuted \hat{r}^* values as extreme or more extreme than our observed values.

Thus, p-value = $4/24 = 0.16$. Reject H_0 , no evidence of positive association.

#(c) Perform an appropriate hypothesis test. You may use the permutations function in R's gtools library.

```
library(gtools)
```

```
perms <- permutations(4, 4, A2)
```

```
N <- nrow(perms)
```

```
estimates <- numeric(N)
```

```
for(i in 1:N) {
```

```
  estimates[i] <- cor(A1, perms[i,], method="pearson")
```

```
}
```

```
output <- cbind(perms, estimates)
```

```
output
```

```
robs <- cor(A2, A1)
```

```
pval <- sum(estimates >= robs)/N # greater than
```

```
pval # fail to reject
```

```
robs
```

```
sum(estimates <= robs)/N # more than
```

```
sum(abs(estimates) >= abs(robs))/N # two-sided
```

#(d) Kendall's tau is an appropriate measure of association for these data as opposed to Pearson's Correlation because:

#With Pearson's Correlation Coefficient it goes with the assumptions that:

#Each observation should have a pair of values,

#Each variable should be continuous,

#It shouldn't have outliers

#Kendall correlation is best used when there are small samples or some outliers. Which this data is.

#3. The diameter and height of individuals of a particular type of plant were recorded in CA4 Opt Plant.csv. You may use R to aid doing this question.

```
CA4_Opt_Plant <- read_csv("SharedFiles/ST204/Data/CA4_Opt_Plant.csv")
```

```
CA4_Opt_Plant
```

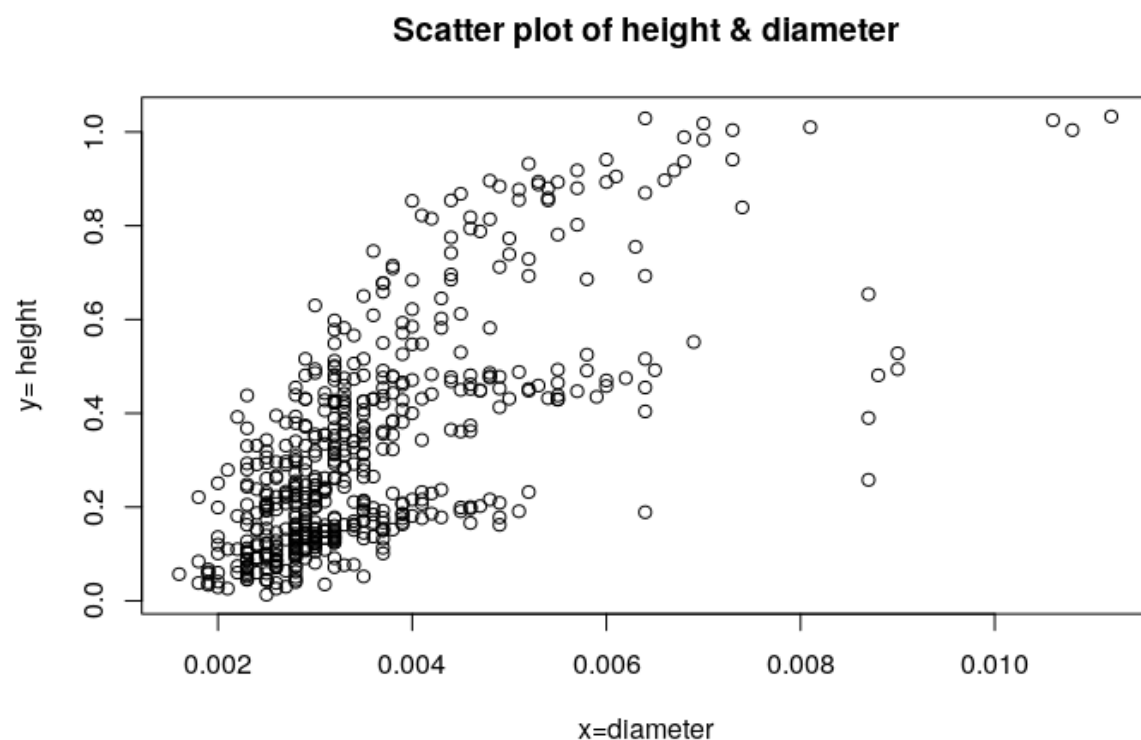
#(a) Create a scatter plot to illustrate the relationship between height & diameter, and estimate Pearson's correlation, Spearman's rank correlation, & Kendall's tau between the two variables.

```
plot(CA4_Opt_Plant$diameter, CA4_Opt_Plant$height, ylab="y= height", xlab="x=diameter", main="Scatter plot of Height versus Diameter")
```

```
with(CA4_Opt_Plant, cor.test(diameter,height,method = "pearson")) #0.69
```

```
with(CA4_Opt_Plant, cor.test(diameter,height,method= "spearman")) #0.69
```

```
with(CA4_Opt_Plant, cor.test(diameter,height,method= "kendall")) #0.51
```



#(b) Construct 95% confidence intervals for each type of association using a bootstrap approach,
 #with 5000 bootstrap samples.

```
with(CA4_Opt_Plant, cor.boot.ci(diameter,height, method="pearson", conf=0.95, nbs=5000)) #(0.64 - 0.74)
```

```
with(CA4_Opt_Plant, cor.boot.ci(diameter,height, method="spearman",conf=0.95, nbs=5000))  
#(0.64 - 0.73)
```

```
with(CA4_Opt_Plant, cor.boot.ci(diameter,height, method="kendall",conf=0.95, nbs=5000)) #(0.47 - 0.55)
```


#(c) Need to Interpret?(Explain) The CIs.

#The Pearson & Spearman test are very similar in confidence intervals.

#The Kendall's Confidence interval considerable lower than the other two.

#Computing the Kendall association between R(A1) and R(A2) is exactly equivalent to computing the Kendall association between A1 and A2 as it implicitly uses ranks anyway

#(d) repeat (b), change .057 to 57.

#Which measure is the most sensitive to the influence of the outlier.

```
CA4_Opt_Plant[1,2] <- 57
```

```
with(CA4_Opt_Plant, cor.boot.ci(diameter,height, method="pearson", conf=0.95, nbs=5000)) #(-0.08 - 0.73)
```

```
with(CA4_Opt_Plant, cor.boot.ci(diameter,height, method="spearman",conf=0.95, nbs=5000))  
#(0.62 - 0.72)
```

```
with(CA4_Opt_Plant, cor.boot.ci(diameter,height, method="kendall",conf=0.95, nbs=5000)) #(0.46 - 0.55)
```

#The one most sensitive to the influence of the outlier is the pearson test. Going from (0.64 - 0.74) to (-0.08 - 0.73), #A difference of (.72,.01)