

Homework 6

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1.

a.

```
> BFHS<-read.table("/Users/Constance/Desktop/BFHS.dat",header=TRUE,skip=6)
```

```
> BFHS
```

	Town	Intervention	ExternalComparison
1	Portsmouth	5.574	5.742
2	Darlington	5.812	5.831
3	Gloucester	5.651	5.714
4	Carlisle	5.438	6.067
5	Burton	5.595	5.702
6	Lincoln	5.530	5.717
7	Dunfermline	5.786	5.621
8	Bridgend	5.553	5.415
9	Bury	5.533	5.624
10	Huddersfield	5.594	5.752
11	Ipswich	5.391	5.585
12	Newport	5.629	5.619
13	Poole	5.370	5.515

```
> summary(BFHS)
```

	Town	Intervention	ExternalComparison
Bridgend	:1	Min. :5.370	Min. :5.415
Burton	:1	1st Qu.:5.530	1st Qu.:5.619
Bury	:1	Median :5.574	Median :5.702
Carlisle	:1	Mean :5.574	Mean :5.685
Darlington	:1	3rd Qu.:5.629	3rd Qu.:5.742
Dunfermline	:1	Max. :5.812	Max. :6.067
(Other)	:7		

b.

```
> t.test(BFHS$Intervention-BFHS$ExternalComparison,var.equal=TRUE)
```

One Sample t-test

```
data: BFHS$Intervention - BFHS$ExternalComparison
t = -2.0702, df = 12, p-value = 0.06067
alternative hypothesis: true mean is not equal to 0
95 percent confidence interval:
 -0.228613888  0.005844657
sample estimates:
mean of x
-0.1113846
```

The p-value of t-test is 0.06067, which is larger than 0.05. So we conclude that there is no significant mean difference in the cholesterol levels of the intervention and comparison groups.

c.

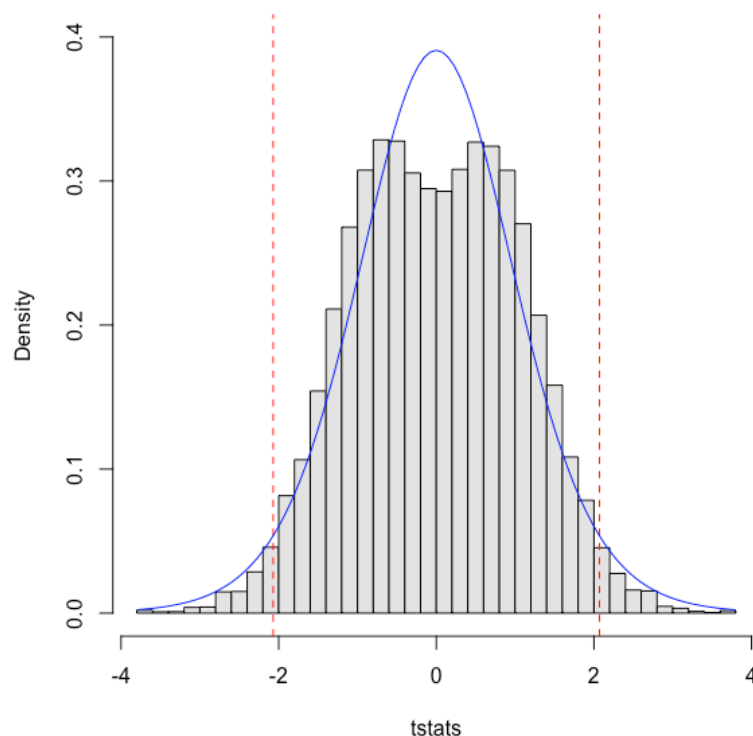
```
> tstats<-replicate(100000,t.test((BFHS$Intervention-BFHS$ExternalComparison)*sample(c(-1,1),
13,replace=TRUE))$statistic)
> t.observed<-t.test(BFHS$Intervention-BFHS$ExternalComparison)$statistic
> approx.pval<-mean(abs(tstats)>=abs(t.observed))
> approx.pval
[1] 0.03922
```

The p-value is 0.03922, which is less than 0.05. So we conclude that there is significant mean difference in the cholesterol levels of the intervention and comparison groups. The conclusion change.

d.

```
> hist(tstats, breaks=50, freq=FALSE, col="grey90", ylim=c(0,0.4),main="Randomization Distrib
. (Sim.), with t.observed and t(12) density")
> abline(v=c(t.observed,-t.observed), lty=2, col="red")
> curve(dt(x, df=12), add=TRUE, col="blue")
```

Randomization Distrib. (Sim.), with t.observed and t(12) density



2.

a.

```
> Barley<-read.csv("/Users/Constance/Desktop/Barley1928.csv",header=TRUE)
> table(Barley$Block)
```

```
A B C D E F
5 5 5 5 5 5
```

There are six blocks: A, B, C, D, E, F, and there are 30 experimental units in total.
b.

```
> class(Barley$Block)
[1] "factor"
> class(Barley$Treatment)
[1] "integer"
```

The treatment is an integer variable, so we use factor(Treatment) when fitting the linear model.

```
> Barleymod<-lm(Yield~Block+factor(Treatment), data=Barley)
> summary(Barleymod)
```

Call:

```
lm(formula = Yield ~ Block + factor(Treatment), data = Barley)
```

Residuals:

Min	1Q	Median	3Q	Max
-42.57	-12.37	-4.10	15.58	36.63

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	222.77	14.80	15.051	2.26e-12	***
BlockB	0.40	16.21	0.025	0.980562	
BlockC	11.40	16.21	0.703	0.490078	
BlockD	79.20	16.21	4.885	8.96e-05	***
BlockE	57.80	16.21	3.565	0.001940	**
BlockF	64.60	16.21	3.984	0.000730	***
factor(Treatment)2	61.67	14.80	4.167	0.000477	***
factor(Treatment)3	92.67	14.80	6.261	4.10e-06	***
factor(Treatment)4	115.83	14.80	7.826	1.63e-07	***
factor(Treatment)5	128.00	14.80	8.648	3.43e-08	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 25.64 on 20 degrees of freedom

Multiple R-squared: 0.8779, Adjusted R-squared: 0.823

F-statistic: 15.98 on 9 and 20 DF, p-value: 2.662e-07

c.

```
> anova(Barleymod)
Analysis of Variance Table
```

Response: Yield

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
Block	5	31634	6326.8	9.6274	8.458e-05	***
factor(Treatment)	4	62903	15725.9	23.9298	2.205e-07	***
Residuals	20	13143	657.2			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

The p-value of treatment is 2.205e-07, which is less than 0.05, so treatment effects are significant.

d.

```
> TukeyHSD(aov(Yield~Block+factor(Treatment), data=Barley))$"factor(Treatment)"
```

	diff	lwr	upr	p adj
2-1	61.66667	17.377951	105.95538	3.874425e-03
3-1	92.66667	48.377951	136.95538	3.673806e-05
4-1	115.83333	71.544617	160.12205	1.501867e-06
5-1	128.00000	83.711284	172.28872	3.175531e-07
3-2	31.00000	-13.288716	75.28872	2.607787e-01
4-2	54.16667	9.877951	98.45538	1.201739e-02
5-2	66.33333	22.044617	110.62205	1.900159e-03
4-3	23.16667	-21.122049	67.45538	5.349191e-01
5-3	35.33333	-8.955383	79.62205	1.597099e-01
5-4	12.16667	-32.122049	56.45538	9.206894e-01

e. According to the Tukey intervals above, pairs (2,1), (3,1), (4,1), (5,1), (4, 2) and (5, 2) have significantly different means.

3.

a.

```
> Spelling<-read.csv("/Users/Constance/Desktop/Spelling1941.csv",header=TRUE)
> Spelling
```

	List	Group	Testing	Number
1	A	1	MC	81
2	A	2	SD	41
3	A	3	WS	44
4	A	4	SW	53
5	B	1	SW	38
6	B	2	MC	97
7	B	3	SD	42
8	B	4	WS	49
9	C	1	WS	31
10	C	2	SW	43
11	C	3	MC	67
12	C	4	SD	36
13	D	1	SD	57
14	D	2	WS	33
15	D	3	SW	43
16	D	4	MC	81

The two blocking factors: List and Group.

b.

```
> matrix(Spelling$Testing,4,4)
      [,1] [,2] [,3] [,4]
[1,] "MC"  "SW"  "WS"  "SD"
[2,] "SD"  "MC"  "SW"  "WS"
[3,] "WS"  "SD"  "MC"  "SW"
[4,] "SW"  "WS"  "SD"  "MC"
```

List corresponds to row, and Group corresponds to column.

c.

```
> class(Spelling$List)
[1] "factor"
> class(Spelling$Group)
[1] "integer"
> class(Spelling$Testing)
[1] "factor"
```

The Group is an integer variable, so we use factor(Group) when fitting the linear model.

```
> Spellingmod<-lm(Number~List+factor(Group)+Testing, data=Spelling)
> summary(Spellingmod)
```

Call:

```
lm(formula = Number ~ List + factor(Group) + Testing, data = Spelling)
```

Residuals:

Min	1Q	Median	3Q	Max
-10.000	-3.500	-1.125	4.188	12.250

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	83.500	7.948	10.505	4.37e-05 ***
ListB	1.750	7.109	0.246	0.81376
ListC	-10.500	7.109	-1.477	0.19016
ListD	-1.250	7.109	-0.176	0.86621
factor(Group)2	1.750	7.109	0.246	0.81376
factor(Group)3	-2.750	7.109	-0.387	0.71224
factor(Group)4	3.000	7.109	0.422	0.68774
TestingSD	-37.500	7.109	-5.275	0.00187 **
TestingSW	-37.250	7.109	-5.240	0.00194 **
TestingWS	-42.250	7.109	-5.943	0.00101 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 10.05 on 6 degrees of freedom

Multiple R-squared: 0.893, Adjusted R-squared: 0.7324

F-statistic: 5.563 on 9 and 6 DF, p-value: 0.02458

d.

```
> anova(Spellingmod)
```

Analysis of Variance Table

Response: Number

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
List	3	359.5	119.83	1.1855	0.391361
factor(Group)	3	74.5	24.83	0.2457	0.861666
Testing	3	4626.5	1542.17	15.2564	0.003254 **
Residuals	6	606.5	101.08		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

The p-value of testing is 0.003254, which is less than 0.05, so treatment effects are statistically significant.

e.

```
> TukeyHSD(aov(Number~List+factor(Group)+Testing, data=Spelling))$Testing
```

	diff	lwr	upr	p adj
SD-MC	-37.50	-62.11023	-12.88977	0.007497477
SW-MC	-37.25	-61.86023	-12.63977	0.007749362
WS-MC	-42.25	-66.86023	-17.63977	0.004107668
SW-SD	0.25	-24.36023	24.86023	0.999982530
WS-SD	-4.75	-29.36023	19.86023	0.905385377
WS-SW	-5.00	-29.61023	19.61023	0.892265654

f. According to the Tukey intervals above, pairs (SD, MC), (SW, MC), and (WS, MC) have significantly different means.

4.

$$(a) P(I \text{ chosen}) = P(II \text{ chosen}) = \frac{1}{\frac{16!}{(4!)^4}} = \frac{1}{63063000}$$

$$(b) P(I \text{ chosen}) = \frac{1}{(4!)^4} = \frac{1}{331776}, P(II \text{ chosen}) = 0$$

$$(c) P(I \text{ chosen}) = P(II \text{ chosen}) = \frac{1}{(4!)^4} = \frac{1}{331776}$$

(d) Assignment I is a Latin square design, but Assignment II is not.

$$P(I \text{ chosen}) = \frac{1}{576}, P(II \text{ chosen}) = 0$$