

# STAT4355HW8

[Code ▾](#)

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## Problem 1

(a)

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```
#load the data
wind <- read.csv(file = 'windmill.csv')

#Box-Cox transformation
library(car)
boxcox.wind <- boxCox(wind$y ~ wind$x,
                      lambda=seq(-2,2,1/10))
```

[Hide](#)

```
boxcox.wind # y=log-Likelihood, x=lambda
```

\$x

```
[1] -2.00000000 -1.95959596 -1.91919192 -1.87878788 -1.83838384 -1.79797980 -1.75757576
[8] -1.71717172 -1.67676768 -1.63636364 -1.59595960 -1.55555556 -1.51515152 -1.47474747
[15] -1.43434343 -1.39393939 -1.35353535 -1.31313131 -1.27272727 -1.23232323 -1.19191919
[22] -1.15151515 -1.11111111 -1.07070707 -1.03030303 -0.98989899 -0.94949495 -0.90909091
[29] -0.86868687 -0.82828283 -0.78787879 -0.74747475 -0.70707071 -0.66666667 -0.62626263
[36] -0.58585859 -0.54545455 -0.50505051 -0.46464646 -0.42424242 -0.38383838 -0.34343434
[43] -0.30303030 -0.26262626 -0.22222222 -0.18181818 -0.14141414 -0.10101010 -0.06060606
[50] -0.02020202  0.02020202  0.06060606  0.10101010  0.14141414  0.18181818  0.22222222
[57]  0.26262626  0.30303030  0.34343434  0.38383838  0.42424242  0.46464646  0.50505051
[64]  0.54545455  0.58585859  0.62626263  0.66666667  0.70707071  0.74747475  0.78787879
[71]  0.82828283  0.86868687  0.90909091  0.94949495  0.98989899  1.03030303  1.07070707
[78]  1.11111111  1.15151515  1.19191919  1.23232323  1.27272727  1.31313131  1.35353535
[85]  1.39393939  1.43434343  1.47474747  1.51515152  1.55555556  1.59595960  1.63636364
[92]  1.67676768  1.71717172  1.75757576  1.79797980  1.83838384  1.87878788  1.91919192
[99]  1.95959596  2.00000000
```

\$y

```
[1] -109.8312236 -107.8483826 -105.8729994 -103.9052962 -101.9454926 -99.9938055 -98.0504572
[8] -96.1156885 -94.1897450 -92.2728837 -90.3653709 -88.4674789 -86.5794976 -84.7017207
[15] -82.8344574 -80.9780278 -79.1327597 -77.2989995 -75.4770984 -73.6674246 -71.8703576
[22] -70.0862852 -68.3156138 -66.5587542 -64.8161334 -63.0881880 -61.3753615 -59.6781120
[29] -57.9969001 -56.3321966 -54.6844772 -53.0542196 -51.4419049 -49.8480142 -48.2730234
[36] -46.7174065 -45.1816283 -43.6661400 -42.1713869 -40.6977863 -39.2457450 -37.8156434
[43] -36.4078284 -35.0226340 -33.6603397 -32.3212071 -31.0054539 -29.7132478 -28.4447373
[50] -27.1999983 -25.9790850 -24.7819975 -23.6086819 -22.4590622 -21.3329900 -20.2302974
[57] -19.1507665 -18.0941361 -17.0601268 -16.0484072 -15.0586332 -14.0904308 -13.1434029
[64] -12.2171450 -11.3112337 -10.4252422 -9.5587431 -8.7113085 -7.8825174 -7.0719656
[71] -6.2792542 -5.5040146 -4.7458964 -4.0045719 -3.2797598 -2.5711910 -1.8786549
[78] -1.2019709 -0.5409962  0.1043449  0.7341102  1.3482782  1.9467913  2.5295425
[85]  3.0963473  3.6469999  4.1812063  4.6986365  5.1989082  5.6815660  6.1461354
[92]  6.5920658  7.0187801  7.4256639  7.8120570  8.1772911  8.5206716  8.8414961
[99]  9.1390661  9.4126870
```

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```
boxcox.wind$x[which.max(boxcox.wind$y)]
```

```
[1] 2
```

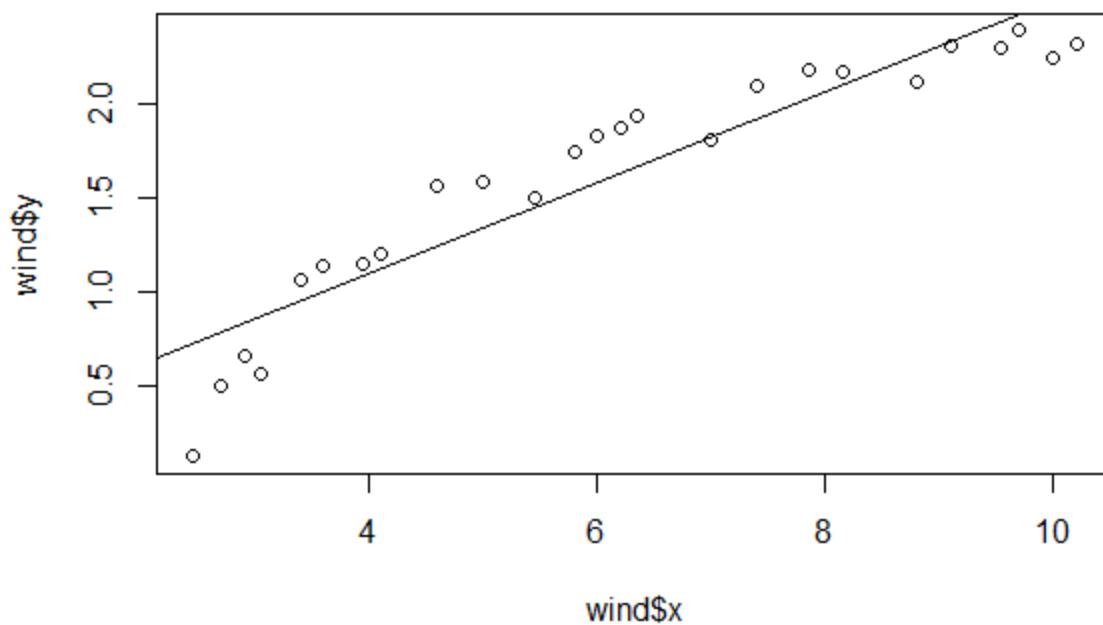
(b)

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```
#build linear model
lm1 <- lm(y~x, wind)

#scatterplot with regression line
plot(wind$x, wind$y) + abline(lm1)
```

```
integer(0)
```



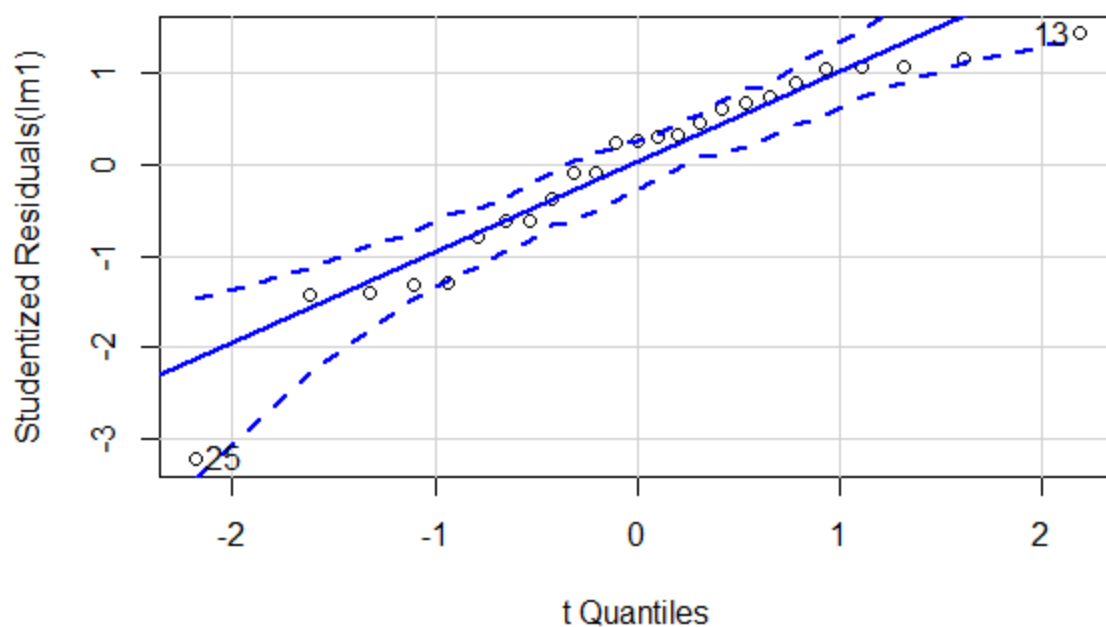
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```
#normal probability plot
qqPlot(lm1)
```

```
[1] 13 25
```

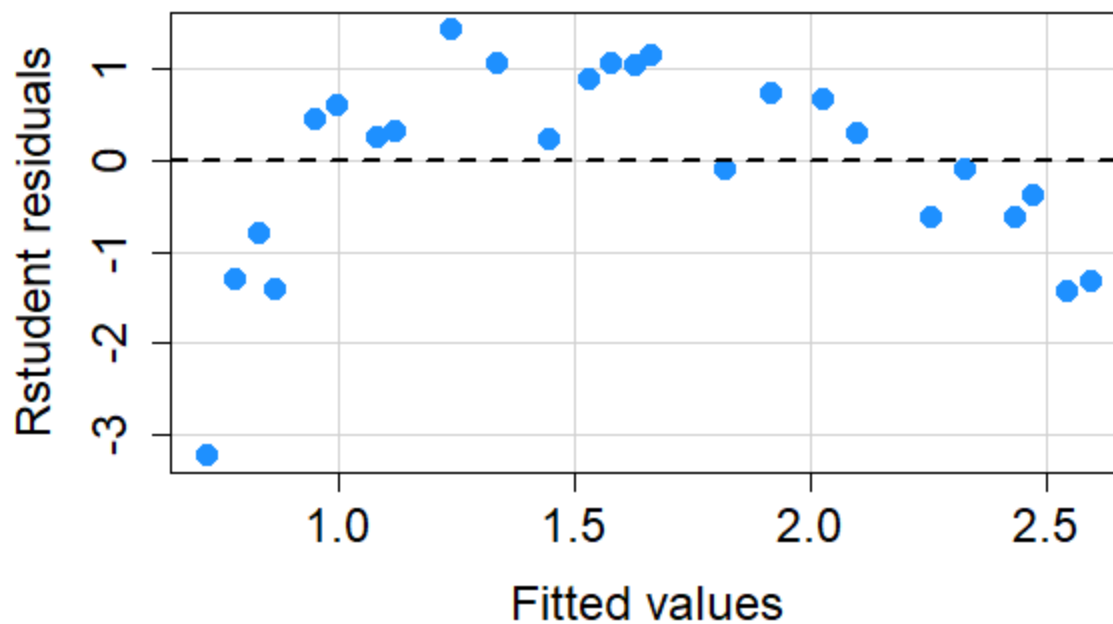
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```
#studentized residuals versus the fitted values plot
par(mfrow=c(1,1))
```



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```
residualPlot(lm1, type="rstudent", quadratic=F, col = "dodgerblue",
             pch=16, cex=1.5, cex.axis=1.5, cex.lab=1.5)
```



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```
#influential analysis
myInf <- influence.measures(lm1)
myInf
```

Influence measures of  
lm(formula = y ~ x, data = wind) :

	dfb.1_ <dbl>	dfb.x <dbl>	dffit <dbl>	cov.r <dbl>	cook.d <dbl>	hat inf <dbl> <chr>
1	0.1749174670	-0.100094360	0.24091867	1.0378975	0.0288421683	0.04834508
2	0.0916665937	-0.011512178	0.21645112	1.0309042	0.0233028359	0.04011347
3	0.1355473962	-0.106977460	0.14443645	1.1761799	0.0107997414	0.08860703
4	-0.4550287954	0.379908437	-0.46860871	1.0696651	0.1067252379	0.11670652
5	0.3299085156	-0.480183601	-0.57029443	1.0610827	0.1555747283	0.13743403
6	0.0764483382	-0.114639354	-0.13958124	1.2305302	0.0101202968	0.12290628
7	0.1195456645	-0.182390775	-0.22529127	1.1941854	0.0260735760	0.11608199
8	-0.4538043800	0.369408522	-0.47402926	1.0241531	0.1077543940	0.10185902
9	-0.0235933316	0.050527326	0.08002279	1.1613545	0.0033337440	0.06652040
10	0.0741159018	0.005822197	0.21227680	1.0344342	0.0224520142	0.04003011

1-10 of 25 rows

Previous 1 2 3 Next

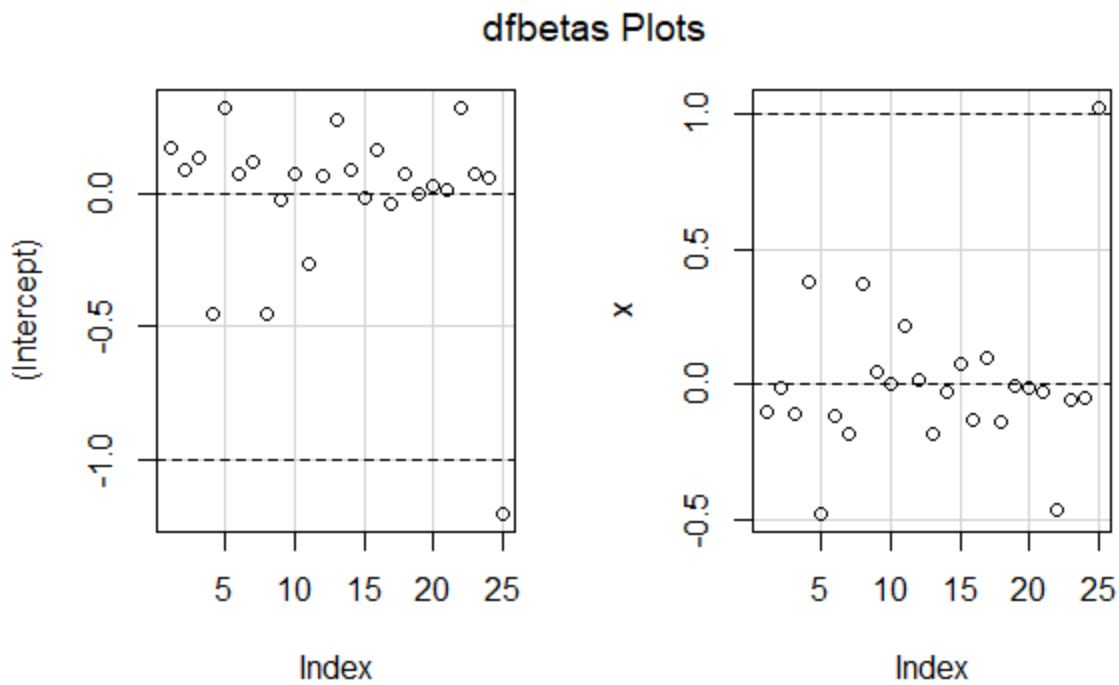
```
summary(myInf)
```

Potentially influential observations of

```
lm(formula = y ~ x, data = wind) :
```

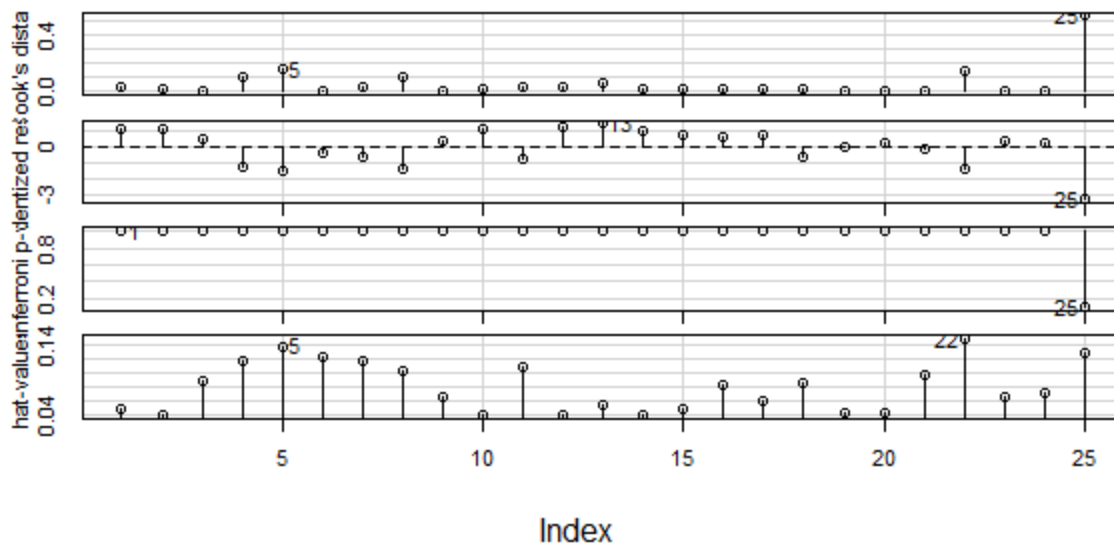
	dfb.1_	dfb.x	dffit	cov.r	cook.d	hat
25	-1.21_*	1.03_*	-1.24_*	0.58_*	0.54	0.13

```
dfbetasPlots(lm1,intercept=T)
```



```
influenceIndexPlot(lm1)
```

## Diagnostic Plots



## Problem 2

(a)

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```
#load the data
patient <- read.table('patient.txt', header=T)
x1 <- patient[,2] # age
x2 <- patient[,3] # severity
x3 <- patient[,4] # surgical-medical
x4 <- patient[,5] # anxiety
y <- patient[,6] # satisfaction

x11=x12=x13=x14=array(0,length(y)) # n by 1 vector of 0's
x11[x1<30] <- 1 # if age group 1
x12[30<=x1 & x1<40] <- 1 # if age group 2
x13[40<=x1 & x1<50] <- 1 # if age group 3
x14[50<=x1 & x1<60] <- 1 # if age group 4

fit <- lm(y~x11+x12+x13+x14+x2+x3+x4)

#test for a general linear hypothesis
library(multcomp)
D <- matrix(0,3,8)
D[1,2]=D[2,3]=D[3,4]=1
D
```

```
      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
[1,]    0    1    0    0    0    0    0    0
[2,]    0    0    1    0    0    0    0    0
[3,]    0    0    0    1    0    0    0    0
```

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```
d <- c(0,0,0)
mytest <- glht(fit, linfct=D, rhs=d) # from page 15
summary(mytest,test=Ftest())
```

### General Linear Hypotheses

Linear Hypotheses:

	Estimate
1 == 0	48.71
2 == 0	49.14
3 == 0	42.14

Global Test:

	<b>F</b> <dbl>	<b>DF1</b> <int>	<b>DF2</b> <int>	<b>Pr(&gt;F)</b> <dbl>
	11.78488	3	17	0.0002035494

1 row

i.

H0:  $\beta_{11} = \beta_{12} = \beta_{13} = 0$

H1:  $\beta_{11}$  or  $\beta_{12}$  or  $\beta_{13} \neq 0$

ii.

D = 01000000 00100000 00010000

d = 000

iii.

As  $F = 11.78488$  and  $p = 0.0002035494 < 0.05$ , we reject the null hypothesis and confirm either  $\beta_{11}$  or  $\beta_{12}$  or  $\beta_{13}$  does not equal zero.

(b)

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```
x11=x12=array(0,length(y))
x11[x1<50] <- 1 # if age group 1
x12[50<=x1 & x1<60] <- 1 # if age group 2

fit <- lm(y~x11+x12+x2+x3+x4)
summary(fit)
```

```
Call:
lm(formula = y ~ x11 + x12 + x2 + x3 + x4)
```

Residuals:

Min	1Q	Median	3Q	Max
-13.157	-7.407	-1.525	6.006	16.002

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	38.8757	15.5052	2.507	0.021410	*
x11	48.8871	8.1199	6.021	8.59e-06	***
x12	32.6479	8.3719	3.900	0.000963	***
x2	-0.4358	0.2006	-2.172	0.042695	*
x3	-1.7310	4.0850	-0.424	0.676501	
x4	4.9538	2.0958	2.364	0.028907	*

---

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

Residual standard error: 9.902 on 19 degrees of freedom

Multiple R-squared: 0.828, Adjusted R-squared: 0.7827

F-statistic: 18.29 on 5 and 19 DF, p-value: 1.123e-06

x11: With all other variables remaining the same, if a patient is in the age group under 50, satisfaction is associated with a 48.8871 increase on average.

x12: With all other variables remaining the same, if a patient is in the age group between 50 and 60, satisfaction is associated with a 32.6479 increase on average.

x2: With all other variables remaining the same, for a one unit increase in severity, satisfaction is associated with a 0.4358 decrease on average.

x3: With all other variables remaining the same, if a patient is a medical patient, satisfaction is associated with a 1.7310 decrease on average.

x4: With all other variables remaining the same, for a one unit increase in anxiety, satisfaction is associated with a 4.9538 increase on average.

## R code is within this notebook