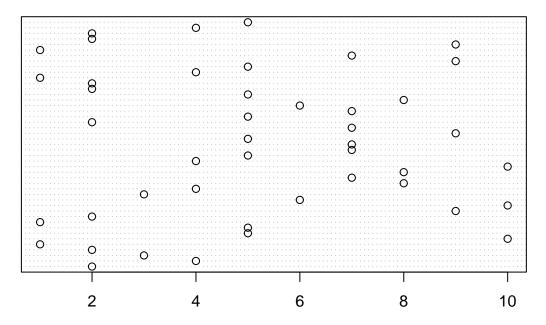
STAT GR5205 - Section 005 HW 3

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```
#1.
#(a)
filename <- "~/Downloads/copiers_full.txt"
copiers_full<- read.table(file=filename, header=T)
x <- copiers_full$copiers
y <- copiers_full$minutes
dotchart(x)</pre>
```

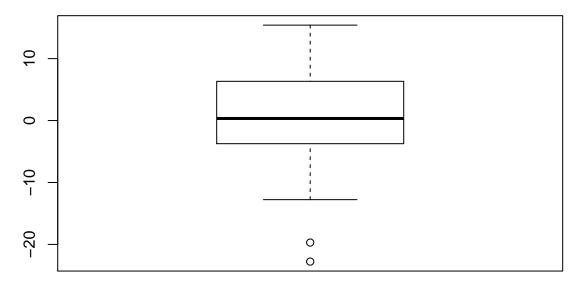


 $\textit{\#The plot is regular, every value is represented. There is no outliers and there is no indication of \textit{temple} is the proof of \textit{temple} is \textit{te$

```
#(b)
fit < -lm(y \sim x)
summary(fit)
##
## Call:
## lm(formula = y \sim x)
##
## Residuals:
        Min
                   1Q
                        Median
## -22.7723 -3.7371
                        0.3334
                                 6.3334 15.4039
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.5802
                             2.8039 -0.207
                                                0.837
## x
                15.0352
                             0.4831 31.123
                                               <2e-16 ***
```

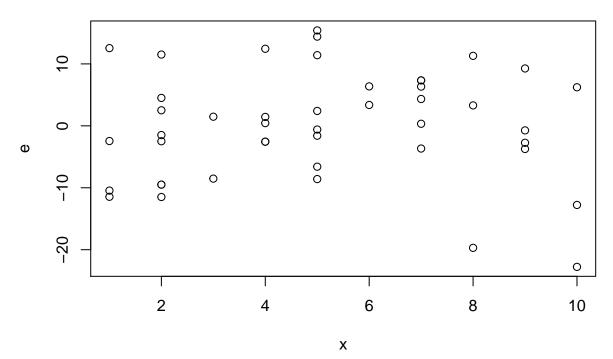
```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.914 on 43 degrees of freedom
## Multiple R-squared: 0.9575, Adjusted R-squared: 0.9565
## F-statistic: 968.7 on 1 and 43 DF, p-value: < 2.2e-16</pre>
```

boxplot(resid(fit))

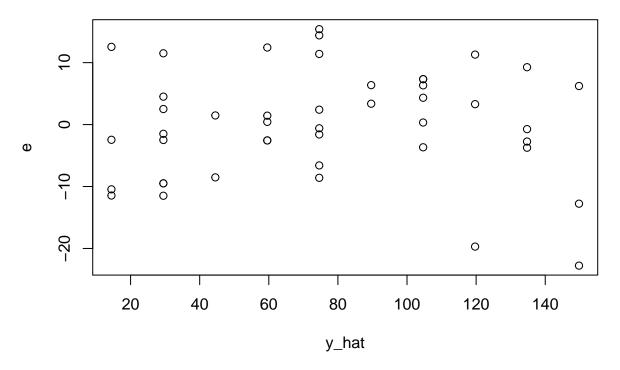


#Not completely symmetric. The distribution of residuals is slightly skewed to the positive side and ha

```
#(c)
e<-resid(fit)
plot(x,e)</pre>
```



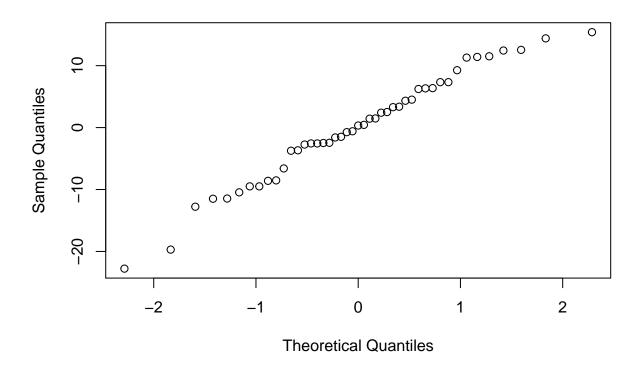
```
y_hat<-fitted(fit)
plot(y_hat,e)</pre>
```



#These two plots provide the same information. Because the fitted values and predictor variable #are linearly related. These plots indicate no severe departure from the assumption of constant variance

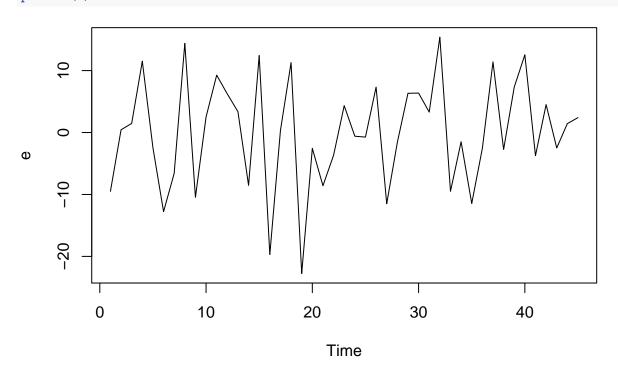
#(d)
qqnorm(e)

Normal Q-Q Plot

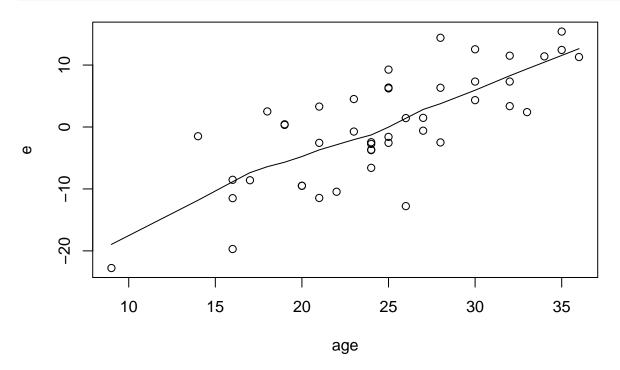


#The plot seems a straight line if we don't consider that two points.

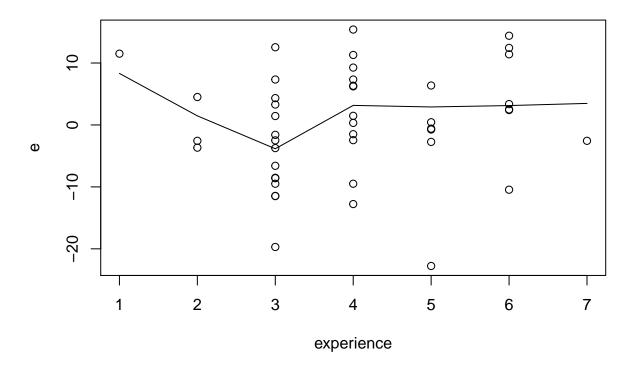
#(e)
plot.ts(e)



```
#(f)
plot(e ~ age, data=copiers_full)
lines(lowess(e ~ copiers_full$age))
```



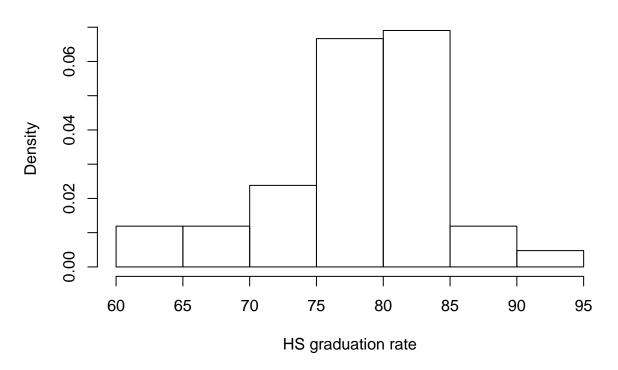
plot(e ~ experience, data=copiers_full)
lines(lowess(e ~ copiers_full\$experience))



#The age of the copiers gets larger, the person need more time to fix copiers.
#The experience does not have any additional effect for predicting service time.

```
#2
#(a)
filename <- "~/Downloads/crime_rates.txt"
crime_rates <- read.table(file=filename, header=T)
hist(crime_rates$x,freq=F,xlab="HS graduation rate",main = "Density hist of HS graduation rate")</pre>
```

Density hist of HS graduation rate



#Distribution is skewed to the left. The HS graduation rate of all counties is between 60 and 95. #And most of counties has HS graduation rate between 75 and 85.

```
#(b)
fit <- lm(y ~ x, data=crime_rates)
summary(fit)

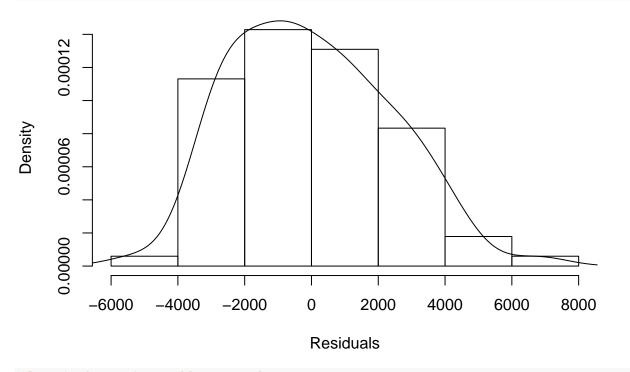
##
## Call:</pre>
```

```
##
## Residuals:
## Min 1Q Median 3Q Max
## -5278.3 -1757.5 -210.5 1575.3 6803.3
##
## Coefficients:
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 20517.60 3277.64 6.260 1.67e-08 ***
```

lm(formula = y ~ x, data = crime_rates)

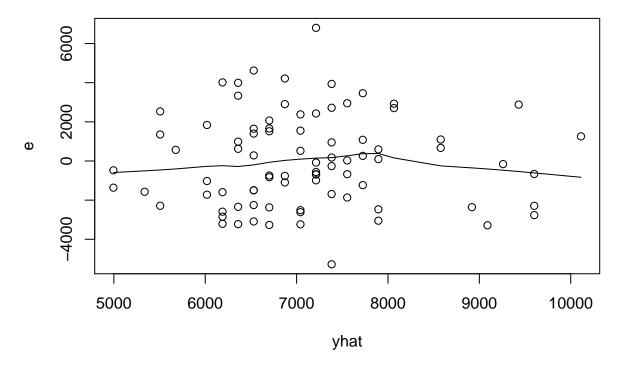
```
## x -170.58   41.57 -4.103 9.57e-05 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2356 on 82 degrees of freedom
## Multiple R-squared: 0.1703, Adjusted R-squared: 0.1602
## F-statistic: 16.83 on 1 and 82 DF, p-value: 9.571e-05
```

```
hist(resid(fit), freq=F,xlab="Residuals", main="")
lines(density(resid(fit)))
```



#Density does not resemble a normal curve.

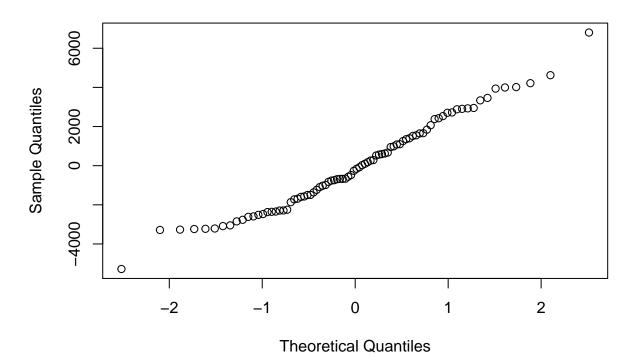
```
#(c)
e <- resid(fit)
yhat <- fitted(fit)
plot(e ~ yhat)
lines(lowess(e ~ yhat))</pre>
```



#Constant variance assumption seems fair.

#(d)
qqnorm(e)

Normal Q-Q Plot



#3.

```
#(a)
filename <- "~/Downloads/SENIC.txt"
SENIC <- read.table(file=filename, header=T)</pre>
fit1 <- lm(Stay ~ Risk, data=SENIC)</pre>
fit2 <- lm(Stay ~ AFS, data=SENIC)</pre>
fit3 <- lm(Stay ~ Xray, data=SENIC)</pre>
par(mfrow=c(2,3))
plot(resid(fit1) ~ fitted(fit1))
plot(resid(fit2) ~ fitted(fit2))
plot(resid(fit3) ~ fitted(fit3))
qqnorm(resid(fit1))
qqnorm(resid(fit2))
qqnorm(resid(fit3))
                                            10
     ω
                                                                                    ω
                                                                                                     0
     9
                                                                                    9
                                            9
resid(fit1)
                                       resid(fit2)
                                                                               resid(fit3)
    \alpha
                                            \alpha
    7
                                            7
                                                                                    7
            8
                     10
                          11
                                                8.0
                                                       9.0
                                                              10.0
                                                                    11.0
                                                                                       8.0
                                                                                              9.0
                                                                                                   10.0
                                                                                                          11.0
                 fitted(fit1)
                                                         fitted(fit2)
                                                                                                fitted(fit3)
           Normal Q-Q Plot
                                                   Normal Q-Q Plot
                                                                                          Normal Q-Q Plot
                                            10
                                                                                    ω
                                       Sample Quantiles
Sample Quantiles
                                                                               Sample Quantiles
     ဖ
                                                                                    9
                                            9
     4
                                                                                    4
    0
                                                                                    \alpha
                                            \alpha
                                            7
                                                                                    ^{2}
                    0
                                                          0
                                                                                                    0
```

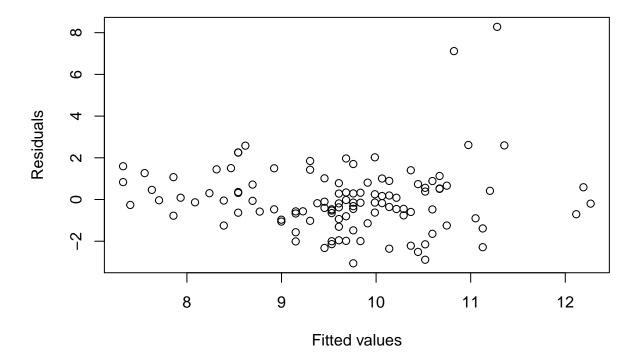
#These three plots show that the distribution of residuals are quite similar, and #all of them have two outliers.

Theoretical Quantiles

```
#(b)
x1<-fitted(fit1)
e<-resid(fit1)
plot(x1, e,xlab="Fitted values", ylab="Residuals")
identify(x1, e, n=2)</pre>
```

Theoretical Quantiles

Theoretical Quantiles



integer(0)

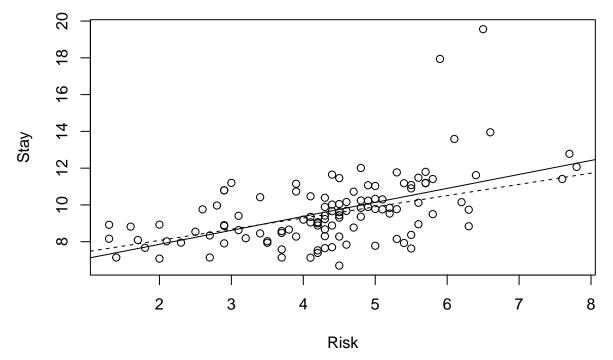
Call:

```
#two outliers are 47 and 112.
summary(fit1)
```

```
##
## Call:
## lm(formula = Stay ~ Risk, data = SENIC)
## Residuals:
##
                1Q Median
                                       Max
## -3.0587 -0.7776 -0.1487 0.7159 8.2805
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                            0.5213 12.156 < 2e-16 ***
## (Intercept)
                 6.3368
## Risk
                 0.7604
                            0.1144
                                     6.645 1.18e-09 ***
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.624 on 111 degrees of freedom
## Multiple R-squared: 0.2846, Adjusted R-squared: 0.2781
## F-statistic: 44.15 on 1 and 111 DF, p-value: 1.177e-09
refit <- update(fit1, subset=-c(47,112))</pre>
summary(refit)
```

$lm(formula = Stay \sim Risk, data = SENIC, subset = -c(47, 112))$

```
##
## Residuals:
                  1Q Median
##
       Min
## -2.89309 -0.67980 -0.08822 0.87180 3.07644
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                           0.40137
                                  17.065 < 2e-16 ***
## (Intercept) 6.84922
## Risk
                0.60975
                           0.08881
                                     6.866 4.23e-10 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.238 on 109 degrees of freedom
## Multiple R-squared: 0.3019, Adjusted R-squared: 0.2955
## F-statistic: 47.14 on 1 and 109 DF, p-value: 4.233e-10
plot(Stay ~ Risk, data=SENIC)
abline(fit1, lty=1)
abline(refit, lty=2)
```



 $\hbox{\it \#The distribution of residual is much normal after we remove that two outliers.}$

112 112 17.94 56.2 5.9 26.4 91.8 835 1

```
#(c)
SENIC[c(47,112),]

## ID Stay Age Risk Cult Xray Beds MS Reg Cen Nurses AFS
## 47 47 19.56 59.9 6.5 17.2 113.7 306 2 1 273 172 51.4
```

1 791

predict(refit, SENIC[c(47,112),], interval="prediction")

```
## 47 10.81259 8.318631 13.30654
## 112 10.44674 7.966822 12.92665
```

#For hospital 47, the 95% PI is (8.3,13.3), the average length of stay in hospital 47 is 19.56, #it is not in the prediction interval, hence it is a outlier.
#For hospital 112, the 95% PI is (7.9,12.9), the average length of stay in hospital 112 is 17.94, #it is not in the prediction interval, hence it is a outlier.