

Question 1

- (a) The normal prob plot looks fine; there does not seem to be any problem with the normality assumption.
- (b) The residual vs fitted value looks good; Variance seems to be a constant not depending on \hat{y} value.
- (c) No outliers in plots in (a) and (b). This is also seen from the observation that robust fit (not required) gives the same results as the LSE.

Question 2.

- (a) The normal prob plot looks okay and gives no evidence of a violation of normality assumption.
- (b) The plot of residual vs fitted suggests a nonlinear pattern. Hence the simple linear model is not adequate.
Optional: a quadratic model gives better fit.

Question 3:

- (a) Residual vs fitted value plot indicates a nonlinear pattern. Hence the straight-line model is not adequate.

(b) Use transformed response variable

$$Y = \log(\text{defects})$$

The fitted model is

$$\hat{Y} = 1.714 + 0.171 \times \text{Weeks} \quad (\text{robust fit})$$

$$\text{or } \hat{Y} = 1.716 + 0.174 \times \text{Weeks} \quad (\text{LSE fit})$$

There appears to be one outlier in the transformed data.

Otherwise, the transformed model is adequate.

Question 4 :

$$\begin{aligned} \text{(i)} \quad AX &= V^{-1}X - V^{-1}X(X^T V^{-1}X)^{-1}X^T V^{-1}X \\ &= V^{-1}X - V^{-1}X = 0 \end{aligned}$$

$$\begin{aligned} \text{(ii)} \quad X^T A &= X^T V^{-1} - X^T V^{-1}X(X^T V^{-1}X)^{-1}X^T V^{-1} \\ &= X^T V^{-1} - X^T V^{-1} = 0 \end{aligned}$$

$$\text{(iii)} \quad \underline{Y} = X\underline{\beta} + \underline{\varepsilon}$$

$$A\underline{Y} = A X \underline{\beta} + A \underline{\varepsilon} \stackrel{\text{(i)}}{=} 0 + A \underline{\varepsilon} = A \underline{\varepsilon}$$

$$\underline{Y}^T A = (\underline{\beta}^T X^T + \underline{\varepsilon}^T) A$$

$$= \underline{\beta}^T X^T A + \underline{\varepsilon}^T A \stackrel{\text{(ii)}}{=} 0 + \underline{\varepsilon}^T A = \underline{\varepsilon}^T A$$

$$\therefore \underline{Y}^T A \underline{Y} = \underline{Y}^T A \underline{\varepsilon} = \underline{\varepsilon}^T A \underline{\varepsilon}$$

$$\begin{aligned}
\text{(iv)} \quad E(\underline{\xi}^T A \underline{\xi}) &= E(\text{trace}(\underline{\xi}^T A \underline{\xi})) \\
&= E(\text{trace}(A \underline{\xi} \underline{\xi}^T)) \\
&= \text{trace}(A E(\underline{\xi} \underline{\xi}^T)) \\
&= \text{trace}(A \cdot \sigma^2 V) \\
&= \sigma^2 \text{trace}(I_{n \times n} - V^{-1} X (X^T V^{-1} X)^{-1} X^T) \\
&= \sigma^2 [\text{trace}(I_{n \times n}) - \text{trace}((X^T V^{-1} X)^{-1} X^T V^{-1} X)] \\
&= \sigma^2 [n - \text{trace}(I_{p \times p})] \\
&= \sigma^2 (n - p)
\end{aligned}$$

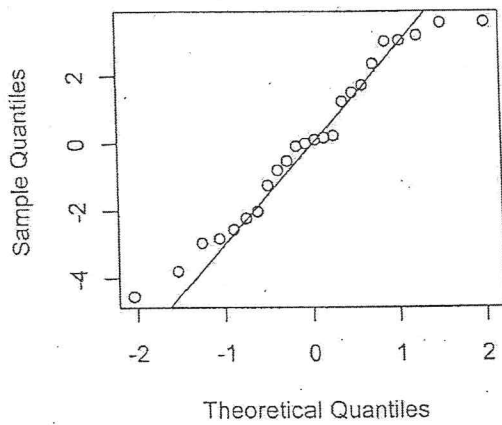
$$\therefore E\left(\frac{\underline{\xi}^T A \underline{\xi}}{n-p}\right) = \sigma^2$$

from (iii), $\underline{Y}^T A \underline{Y} = \underline{\xi}^T A \underline{\xi}$

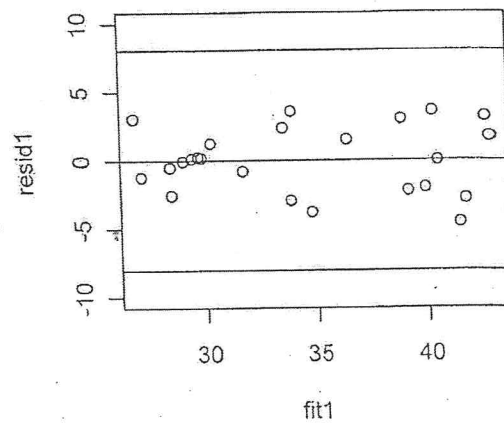
Hence $\underline{Y}^T A \underline{Y} / (n-p)$ is an unbiased estimator for σ^2 .

Question 1 Plots (R codes at the back)

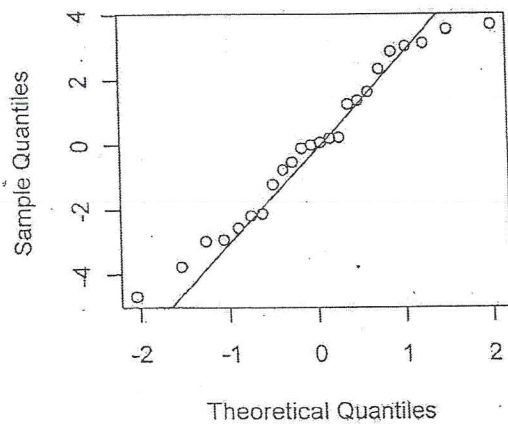
LSE fit



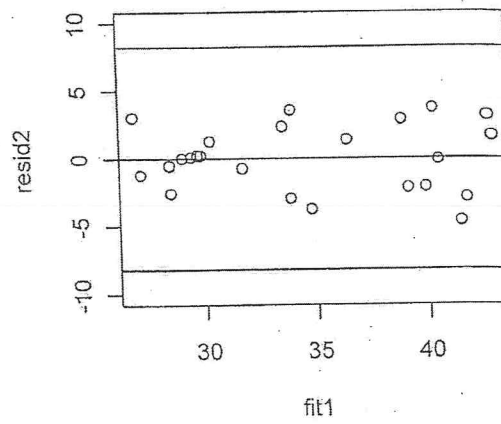
LSE fit



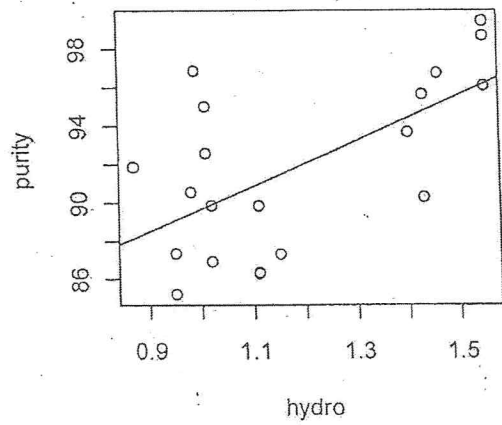
Robust fit



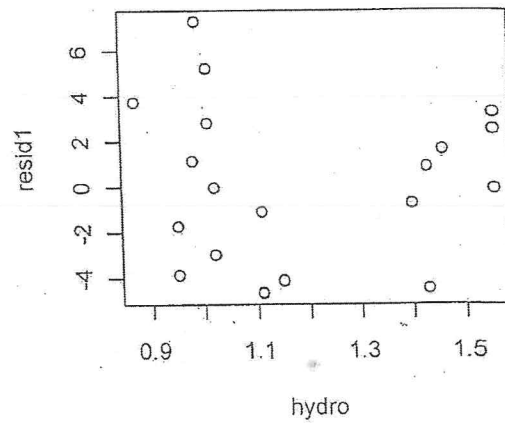
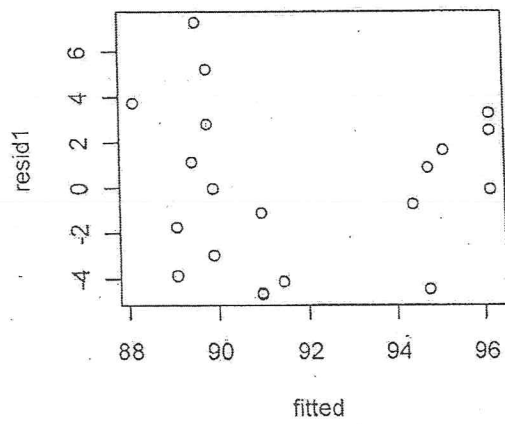
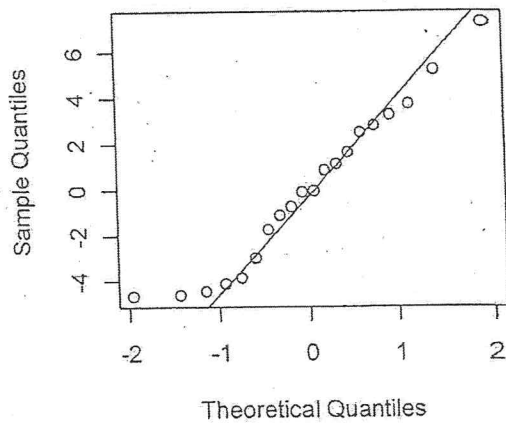
Robust fit



Question 2 plots (R code at the back)

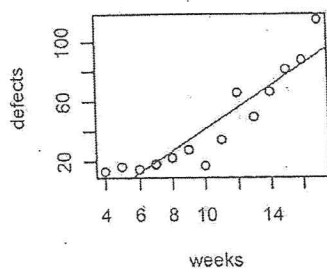


Question 4.6

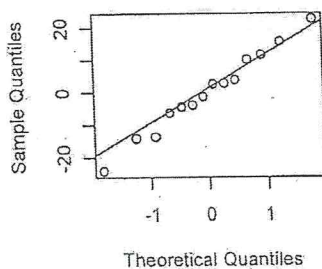


Question 3 plots (R code at the back)

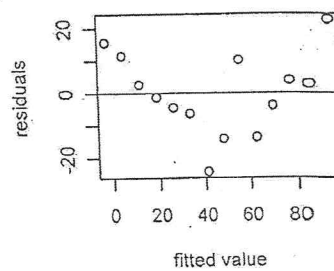
Question 5.5



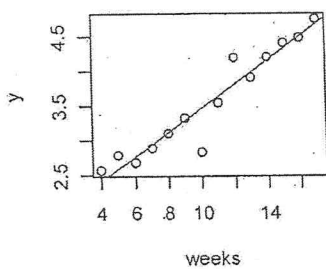
Normal Q-Q Plot



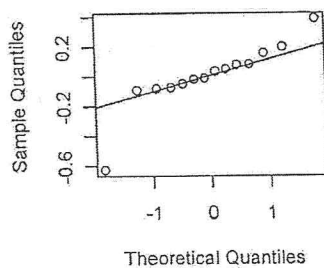
(a)



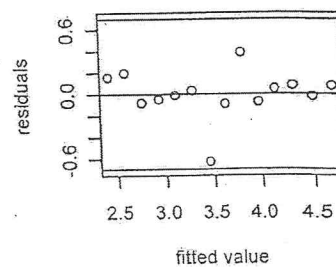
transformed data



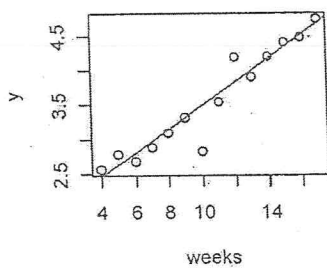
Normal Q-Q Plot



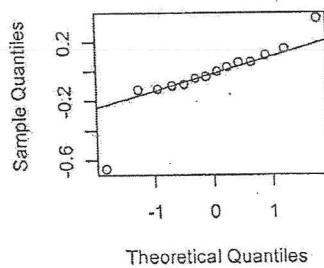
LSE fit



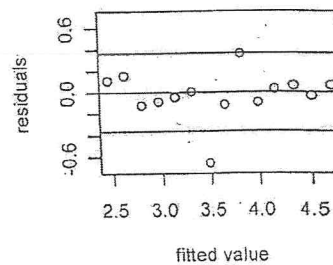
robust fit



Normal Q-Q Plot



Robust fit




```

#Question 4.5

data3=read.table(file="TableB4.prn", header=TRUE)
attach(data3)

print(data3)

#LSE fit
l1=lm(y~x1+x2+x5+x7,data=data3)
summary(l1)

resid1=l1$residuals
fit1=l1$fitted.values
s1=summary.lm(l1)$sigma

library(car)
vif(l1)

# Robust fit
library(robustbase) #load robustbase
l2=lmrob(y~x1+x2+x5+x7, data=data3)
summary(l2)
s2=l2$scale

resid2=l2$residuals
fit2=y-resid2

par(mfrow=c(2,2))
qqnorm(resid1, main="LSE fit")
qqline(resid1)

plot(fit1, resid1, main="LSE fit",ylim=c(-10,10))
abline(h=0)
abline(h=3*s1,col="blue")
abline(h=-3*s1, col="blue")

qqnorm(resid2, main="Robust fit")
qqline(resid2)

plot(fit1, resid2, main="Robust fit",ylim=c(-10,10))
abline(h=0)
abline(h=3*s2,col="blue")
abline(h=-3*s2, col="blue")

shapiro.test(resid1)
shapiro.test(resid2)

#There are no outliers in the data.
#There seems to be no problem with normality.
#It seems to have constant variance for the errors.

Call:
lm(formula = y ~ x1 + x2 + x5 + x7, data = data3)

Residuals:
    Min       1Q   Median       3Q      Max
-4.5510 -2.0807  0.0391  1.8805  3.5912

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  13.5091     3.6331   3.718 0.001458 **
x1           2.4192     0.5169   4.680 0.000163 ***
x2           8.4802     3.2943   2.574 0.018577 *
x5           2.0006     1.2104   1.653 0.114793
x7          -2.1823     1.2762  -1.710 0.103557

```

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

```
Residual standard error: 2.681 on 19 degrees of freedom
Multiple R-squared:  0.8352,    Adjusted R-squared:  0.8005
F-statistic: 24.08 on 4 and 19 DF,  p-value: 3.249e-07
```

```
#Question 4.6
```

```
data2=read.table(file="data-prob-2-7.prn", header=TRUE)
attach(data2)
print(data2)
```

```
par(mfrow=c(2,2))
plot(hydro,purity)
```

```
fit1=lm(purity~hydro)
resid1=fit1$residual
fitted=purity-resid1
```

```
abline(fit1)
```

```
qqnorm(resid1, main="Question 4.6")
qqline(resid1)
shapiro.test(resid1)
```

```
plot(fitted, resid1)
```

```
plot(hydro, resid1)
```

```
#There is a nonlinear pattern in the plot of residuals vs fitted values.
```

```
#Question 5.5
```

```
Data3=read.table(file="data-prob-5-5.prn", header=TRUE)
attach(Data3)
print(Data3)
```

```
par(mfrow=c(3,3))
plot(weeks, defects, main="Question 5.5")
```

```
#Fit the simple linear regression model
l1=lm(defects~weeks, data=Data3)
abline(l1, col="red")
summary(l1)
```

```
resid1=l1$residuals
fit1=l1$fitted.values
```

```
qqnorm(resid1)
qqline(resid1)
shapiro.test(resid1)
```

```
plot(fit1,resid1,xlab="fitted value", ylab="residuals", main="(a)")
abline(h=0)
```

```
#The simple linear regression model is not adequate.
```

```
#Make transformation
```

```
y=log(defects)
l2=lm(y~weeks)
plot(weeks, y, main="transformed data")
abline(l2, col="red")
```



```

summary(l2)

resid2=l2$residuals
fit2=l2$fitted.values
qqnorm(resid2)
qqline(resid2)
shapiro.test(resid2)
s2=summary.lm(l2)$sigma
plot(fit2,resid2,ylim=c(-0.7,0.7), main="LSE fit",xlab="fitted value",
ylab="residuals")
abline(h=0)
abline(h=-3*s2, col="blue")
abline(h=3*s2, col="blue")

#Robust fit
l3=lmrob(y~weeks)
plot(weeks, y, main="robust fit")
abline(l3, col="red")
summary(l3)
resid3=l3$residuals
fit3=l3$fitted.values
qqnorm(resid3)
qqline(resid3)
shapiro.test(resid3)
s3=l3$scale
plot(fit3,resid3,ylim=c(-0.7,0.7), main="Robust fit",
xlab="fitted value", ylab="residuals" )
abline(h=0)
abline(h=-3*s3, col="blue")
abline(h=3*s3, col="blue")

#There is one outlier in the data.
#Robust fit for the transformed data looks good.

Call:
lmrob(formula = y ~ weeks)

Weighted Residuals:
      Min       1Q   Median       3Q      Max
-0.66240 -0.09849 -0.02065  0.05968  0.35789

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  1.774244    0.098976   17.93 4.99e-10 ***
weeks        0.170954    0.006582   25.97 6.48e-12 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Robust residual standard error: 0.1203
Convergence in 9 IRWLS iterations

Robustness weights:
observation 7 is an outlier with |weight| = 0 (< 0.0071);
one weight is ~1. The remaining 12 ones are summarized as
      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.3562  0.9065   0.9424  0.8976  0.9800  0.9960
Algorithmic parameters:
tuning.chi      bb tuning.psi refine.tol    rel.tol
1.5476400  0.5000000  4.6850610  0.0000001  0.0000001
nResample      max.it   groups  n.group best.r.s      k.fast.s      k.max
      500           50       5       400       2         1         200
trace.lev compute.rd
      0           0
seed : int(0)
> summary(l2)

Call:

```

```
lm(formula = y ~ weeks)
```

```
Residuals:
```

	Min	1Q	Median	3Q	Max
	-0.62990	-0.06982	0.00977	0.07727	0.38529

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.71622	0.17311	9.914	3.93e-07	***
weeks	0.17351	0.01539	11.273	9.68e-08	***

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

Residual standard error: 0.2322 on 12 degrees of freedom
Multiple R-squared: 0.9137, Adjusted R-squared: 0.9065
F-statistic: 127.1 on 1 and 12 DF, p-value: 9.676e-08