

```
achiev <- read.table("achiev.txt", header = T)
```

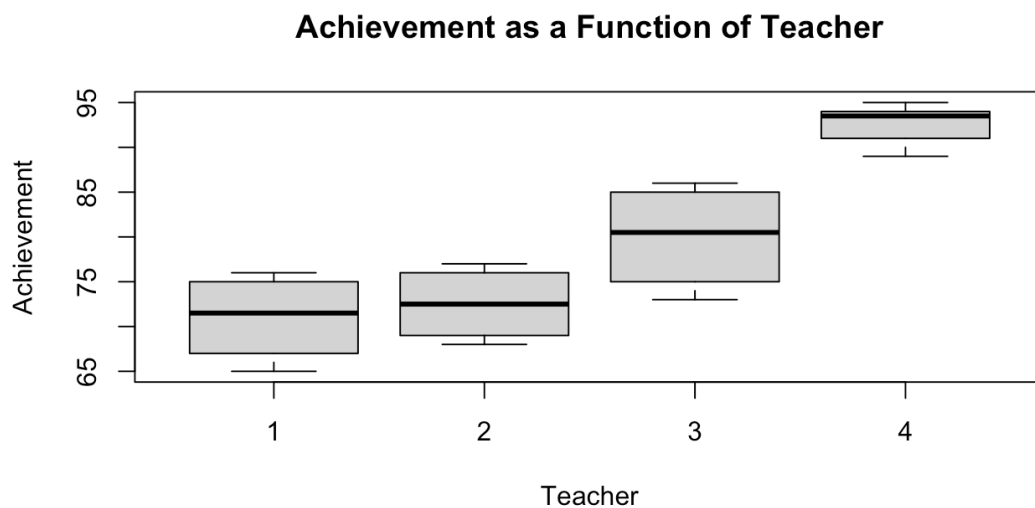
```
achiev
```

```
Ind ac teach text
```

```
1  1 70    1  1
2  2 67    1  1
3  3 65    1  1
```

```
attach(achiev)
```

```
boxplot(ac ~ teach, data=achiev, main="Achievement as a Function of
Teacher", xlab = "Teacher", ylab="Achievement")
```

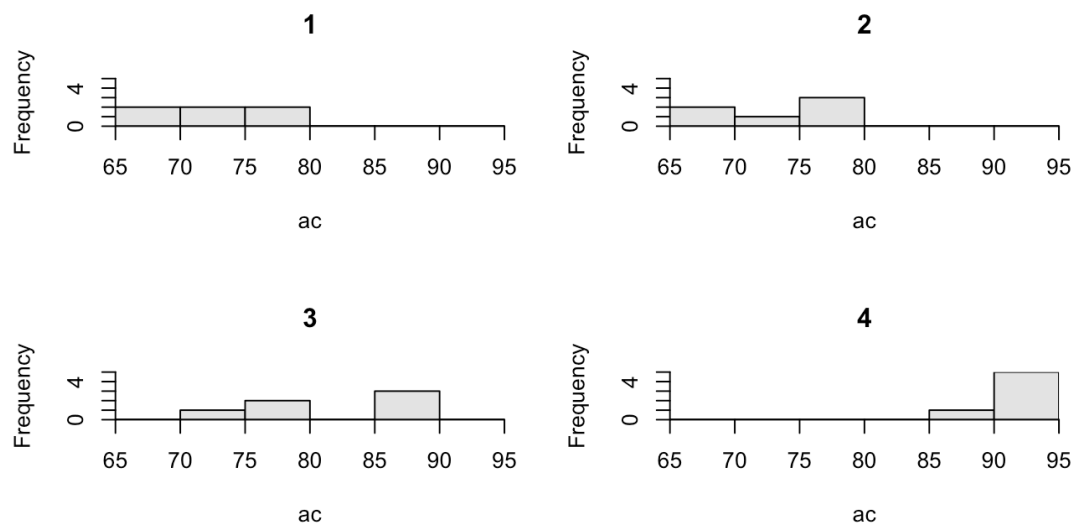


```
install.packages('FSA')
```

```
library(FSA)
```

```
f.teach <- factor(teach)
```

```
hist(ac~f.teach, data = achiev)
```



```
shapiro.test(ac)
```

Shapiro-Wilk normality test

```
data: ac
```

W = 0.90565, p-value = 0.02842

```
fligner.test(ac~f.teach, data = achiev)
```

Fligner-Killeen test of homogeneity of variances

```
data: ac by f.teach
```

Fligner-Killeen:med chi-squared = 10.813, df = 3, p-value = 0.01278

```
aggregate(ac ~ f.teach, FUN = var)
```

	f.teach	ac
1	1	19.600000
2	2	15.500000
3	3	35.200000
4	4	5.066667

```
aggregate(ac ~ f.teach, FUN = mean)
```

	f.teach	ac
--	---------	----

1 1 71.00000

2 2 72.50000

3 3 80.00000

4 4 92.66667

```
anova.fit <- aov(ac ~ f.teach, data = achiev)
```

```
summary(anova.fit)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
f.teach	3	1764.1	588.0	31.21	9.68e-08 ***
Residuals	20	376.8	18.8		

---

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

```
model.tables(anova.fit)
```

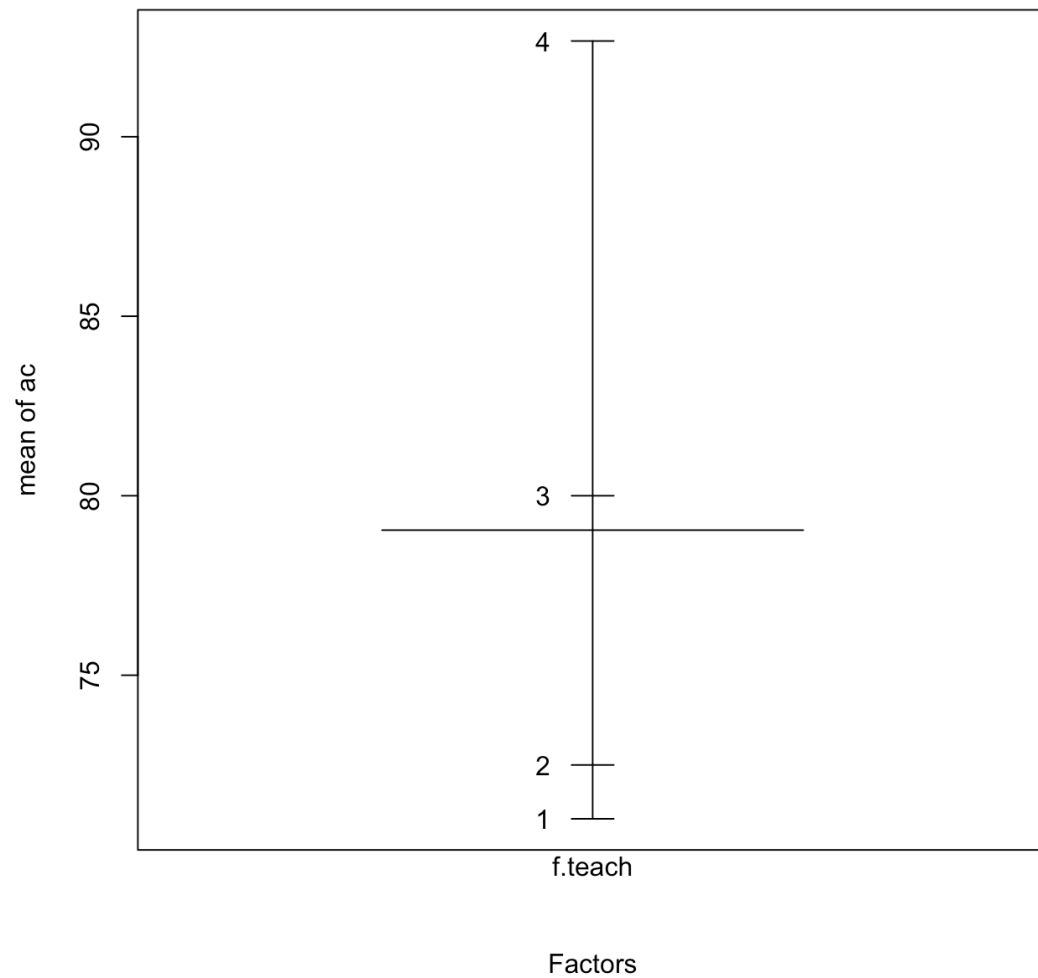
Tables of effects

f.teach

f.teach

	1	2	3	4
	-8.042	-6.542	0.958	13.625

```
plot.design(ac~f.teach)
```



TukeyHSD(anova.fit)

Tukey multiple comparisons of means

95% family-wise confidence level

Fit: aov(formula = ac ~ f.teach, data = achiev)

\$f.teach

	diff	lwr	upr	p adj
2-1	1.50000	-5.5144241	8.514424	0.9313130

```
3-1  9.00000  1.9855759 16.014424 0.0090868
```

```
4-1 21.66667 14.6522425 28.681091 0.0000002
```

```
3-2  7.50000  0.4855759 14.514424 0.0334428
```

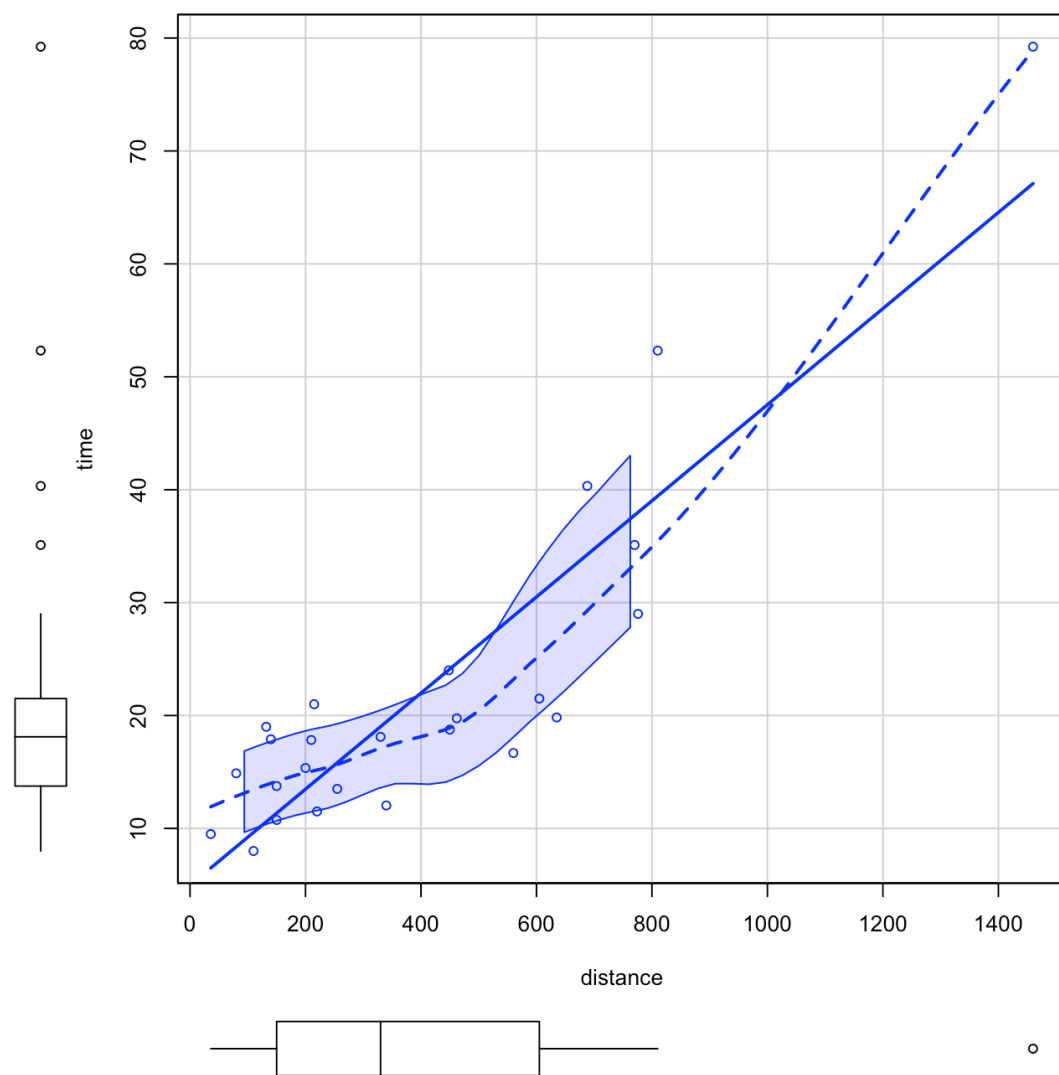
```
delivery <- read.csv('delivery.csv', header = T)
```

```
delivery <- read.table('delivery.csv', sep=',', header = T)
```

```
attach(delivery)
```

```
library(car)
```

```
scatterplot(time ~ distance)
```



```
scatterplot(time ~ distance)
```

```
fit <- lm(time ~ distance)
```

```
fit
```

```
Call:
```

```
lm(formula = time ~ distance)
```

```
Coefficients:
```

(Intercept)	distance
4.96116	0.04257

```
coef(fit)
```

(Intercept)	distance
4.96115886	0.04256949

```
cor(fitted(fit),time)^2
```

```
[1] 0.7950755
```

```
summary(fit)
```

```
Call:
```

```
lm(formula = time ~ distance)
```

```
Residuals:
```

Min	1Q	Median	3Q	Max
-12.1628	-4.8783	-0.5966	6.0810	12.8776

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.961159	2.337360	2.123	0.0448 *
distance	0.042569	0.004506	9.447	2.21e-09 ***

```
---
```

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

Residual standard error: 7.179 on 23 degrees of freedom

Multiple R-squared: 0.7951, Adjusted R-squared: 0.7862

F-statistic: 89.24 on 1 and 23 DF, p-value: 2.214e-09

anova(fit)

Analysis of Variance Table

Response: time

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
distance	1	4599.1	4599.1	89.237	2.214e-09 ***
Residuals	23	1185.4	51.5		

---

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

new <- data.frame(distance=c(100, 500, 1000))

predict(fit, new)

	1	2	3
	9.218108	26.245904	47.530650

fit.mr <- lm(time~distance+case)

summary(fit.mr)

Call:

lm(formula = time ~ distance + case)

Residuals:

	Min	1Q	Median	3Q	Max
	-5.7880	-0.6629	0.4364	1.1566	7.4197

anova(fit.mr)

Analysis of Variance Table

Response: time

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
distance	1	4599.1	4599.1	432.895	5.825e-16 ***
case	1	951.7	951.7	89.575	3.255e-09 ***
Residuals	22	233.7	10.6		

---

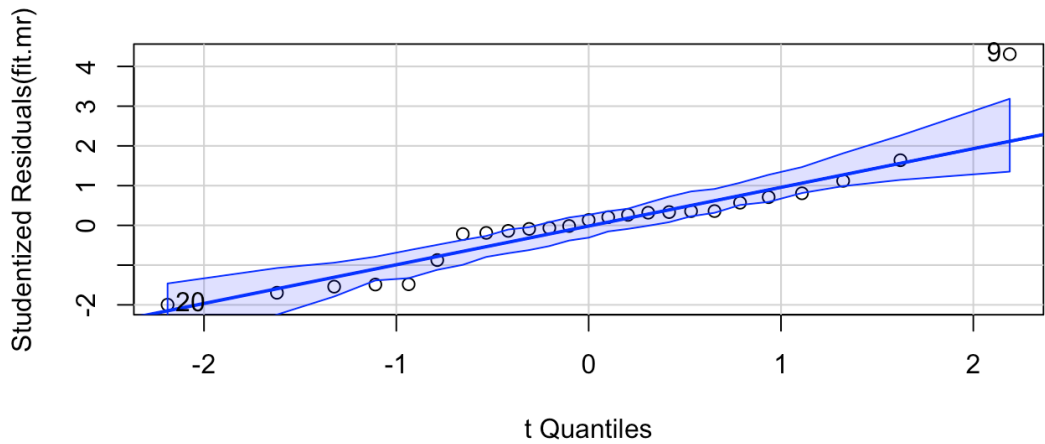
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

`cor(fitted(fit),time)^2`

[1] 0.7950755

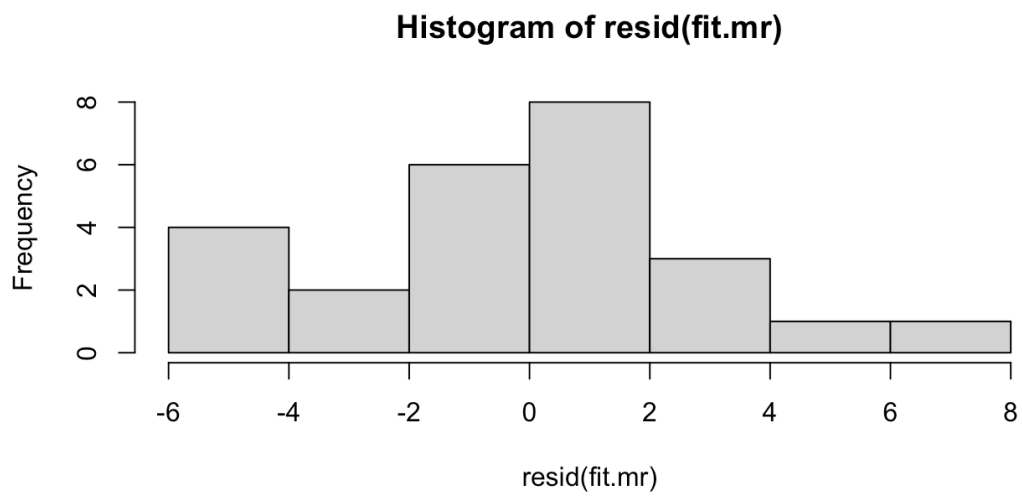
`qqPlot(fit.mr)`

[1] 9 20

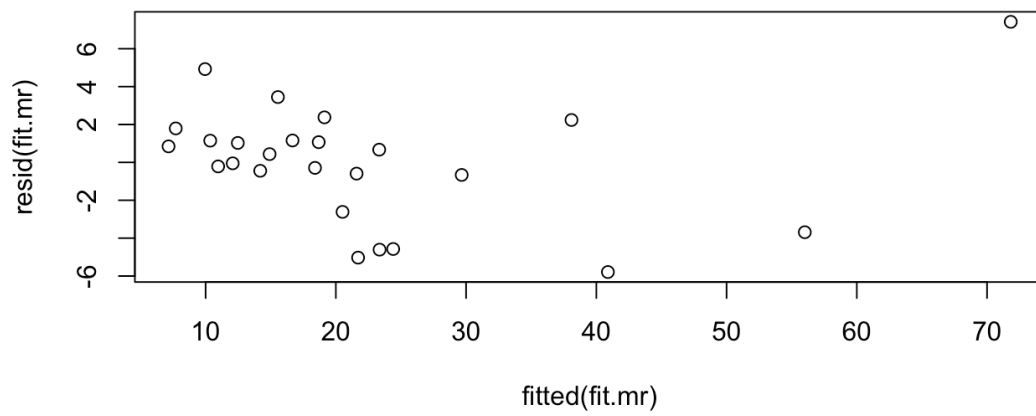


`hist(resid(fit.mr))`

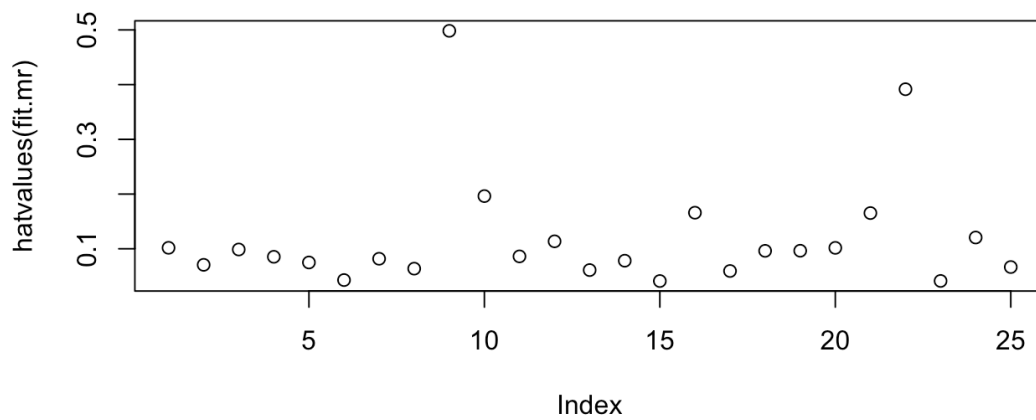




`plot(fitted(fit.mr), resid(fit.mr))`



`plot(hatvalues(fit.mr))`



```
which.max(hatvalues(fit.mr))
```

```
9
```

```
9
```

```
cooks.distance(fit.mr)
```

	1	2	3	4	5
6	7				
1.000921e-01	3.375704e-03	9.455785e-06	7.764718e-02	5.432217e-04	
1.231067e-04	2.171604e-03				
	8	9	10	11	12
13	14				
3.051135e-03	3.419318e+00	5.384516e-02	1.619975e-02	1.596392e-03	
2.294737e-03	3.292786e-03				
	15	16	17	18	19
20	21				
6.319880e-04	3.289086e-03	4.013419e-04	4.397807e-02	1.191868e-02	
1.324449e-01	5.086063e-02				
	22	23	24	25	
4.510455e-01	2.989892e-02	1.023224e-01	1.084694e-04		

```
p1 <- c(2, 4, 6, 8, 10)
```

```
p2 <- c(4, 8, 12, 16, 20)
```

```
y <- c(12, 15, 8, 4, 1)
```

```
model <- lm(y ~ p1 + p2)
```

```
summary(model)
```

```
Call:
```

```
lm(formula = y ~ p1 + p2)
```

Residuals:

	1	2	3	4	5
	-2.600e+00	3.700e+00	6.661e-16	-7.000e-01	-4.000e-01

Coefficients: (1 not defined because of singularities)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	17.9000	2.7815	6.435	0.00761 **
p1	-1.6500	0.4193	-3.935	0.02924 *
p2	NA	NA	NA	NA

---

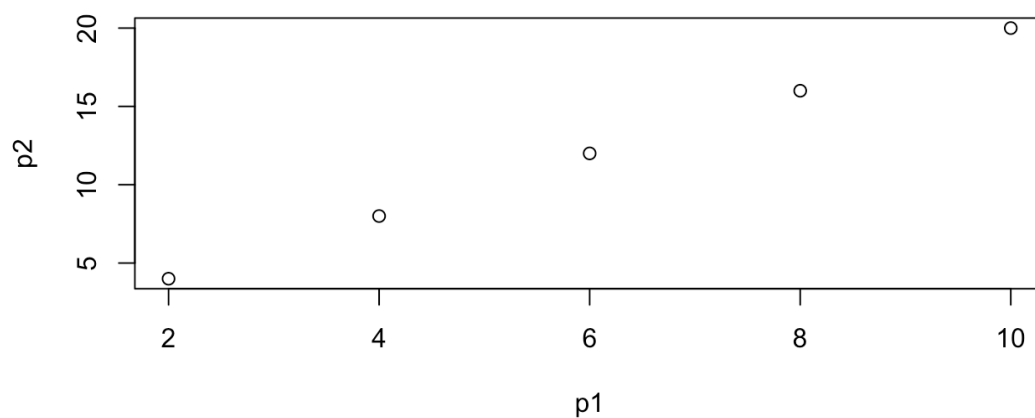
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.652 on 3 degrees of freedom

Multiple R-squared: 0.8377, Adjusted R-squared: 0.7836

F-statistic: 15.48 on 1 and 3 DF, p-value: 0.02924

plot(p1,p2)



```
model.1 <- lm(y ~ p1)
```

```
summary(model.1)
```

Call:

```
lm(formula = y ~ p1)
```

Residuals:

	1	2	3	4	5
	-2.600e+00	3.700e+00	6.661e-16	-7.000e-01	-4.000e-01

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	17.9000	2.7815	6.435	0.00761 **
p1	-1.6500	0.4193	-3.935	0.02924 *

---

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

Residual standard error: 2.652 on 3 degrees of freedom

Multiple R-squared: 0.8377, Adjusted R-squared: 0.7836

F-statistic: 15.48 on 1 and 3 DF, p-value: 0.02924

```
vif(model)
```

Error in vif.default(model) : there are aliased coefficients in the model

```
p2 <- c(4, 8, 12, 16, 21)
```

```
model <- lm(y ~ p1 + p2)
```

```
summary(model)
```

Call:

```
lm(formula = y ~ p1 + p2)
```

Residuals:

	1	2	3	4	5
	-2.400e+00	3.700e+00	-2.000e-01	-1.100e+00	1.965e-14

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	17.500	3.940	4.441	0.0471 *
p1	0.450	10.694	0.042	0.9703
p2	-1.000	5.087	-0.197	0.8623

---

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

Residual standard error: 3.217 on 2 degrees of freedom

Multiple R-squared: 0.8408, Adjusted R-squared: 0.6815

F-statistic: 5.28 on 2 and 2 DF, p-value: 0.1592

```
gaf.data <- read.table("gaf.txt", header = T)
```

```
gaf.data
```

```
ID gaf age pretherapy n_therapy
```

1	1	25	21	52	6
2	2	25	19	61	17
3	3	0	18	50	5

```
attach(gaf.data)
```

```
gaf.fit <- lm(gaf ~ age)
```

```
summary(gaf.fit)
```

Call:

```
lm(formula = gaf ~ age)
```

Residuals:

Min	1Q	Median	3Q	Max
-13.657	-9.196	4.359	7.617	9.784

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-15.6806	12.1427	-1.291	0.23264
age	1.6299	0.4369	3.731	0.00578 **

---

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

Residual standard error: 10.19 on 8 degrees of freedom

Multiple R-squared: 0.635, Adjusted R-squared: 0.5894

F-statistic: 13.92 on 1 and 8 DF, p-value: 0.00578

```
gaf.fit <- lm(gaf ~ age + pretherapy)
```

```
summary(gaf.fit)
```

Call:

```
lm(formula = gaf ~ age + pretherapy)
```

Residuals:

Min	1Q	Median	3Q	Max
-8.8399	-7.2939	-0.0415	6.6845	9.2966

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-102.7837	39.6255	-2.594	0.0357 *
age	1.2671	0.3888	3.259	0.0139 *
pretherapy	1.7669	0.7786	2.269	0.0575 .

---

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

Residual standard error: 8.265 on 7 degrees of freedom

Multiple R-squared: 0.7897, Adjusted R-squared: 0.7297

F-statistic: 13.15 on 2 and 7 DF, p-value: 0.004263

```
mod.1 <- lm(gaf ~ age)
```

```
summary(mod.1)
```

Call:

```
lm(formula = gaf ~ age)
```

Residuals:

Min	1Q	Median	3Q	Max
-13.657	-9.196	4.359	7.617	9.784

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-15.6806	12.1427	-1.291	0.23264
age	1.6299	0.4369	3.731	0.00578 **

---

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

Residual standard error: 10.19 on 8 degrees of freedom

Multiple R-squared: 0.635, Adjusted R-squared: 0.5894

F-statistic: 13.92 on 1 and 8 DF, p-value: 0.00578

```
mod.2 <- lm(gaf ~ age + pretherapy)
```

```
summary(mod.2)
```

Call:

```
lm(formula = gaf ~ age + pretherapy)
```

Residuals:

Min	1Q	Median	3Q	Max
-----	----	--------	----	-----

-8.8399 -7.2939 -0.0415 6.6845 9.2966

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-102.7837	39.6255	-2.594	0.0357 *
age	1.2671	0.3888	3.259	0.0139 *
pretherapy	1.7669	0.7786	2.269	0.0575 .

---

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

Residual standard error: 8.265 on 7 degrees of freedom

Multiple R-squared: 0.7897, Adjusted R-squared: 0.7297

F-statistic: 13.15 on 2 and 7 DF, p-value: 0.004263

library(MASS)

gaf.fit <- lm(time ~ distance + case, data = delivery)

step <- stepAIC(gaf.fit, direction = "both")

Start: AIC=61.88

time ~ distance + case

	Df	Sum of Sq	RSS	AIC
<none>			233.73	61.882
- distance	1	168.40	402.13	73.448
- case	1	951.66	1185.39	100.474