

Linear Regression

lm(formula, data, ...)

formula: a symbolic description of the model to be fit. The details of model specification are given below.

data: an optional data frame, list or a data frame.

Example:

```
x=c(75,70,65,60,55,55,50,50,45,40)
```

```
y=c(4,3.9,3.7,3.2,3,2.8,2.2,1.8,1.5,1)
```

```
regmodel=lm(y~x)
```

```
yy=data.frame(x,y)
```

```
pairs(yy)
```

```
regmodel=lm(y~x,yy)
```

```
par(mfrow=c(2,2))
```

```
hist(resid(regmodel))
```

```
> plot(regmodel$fitted.values,resid(regmodel))
```

```
> lines(lowess(regmodel$fitted.values,resid(regmodel)))
```

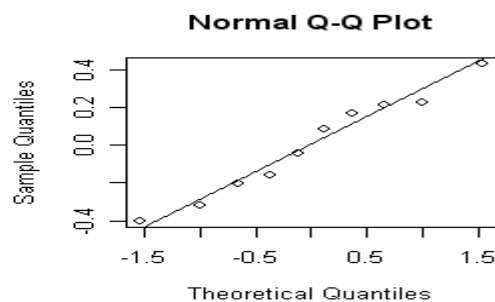
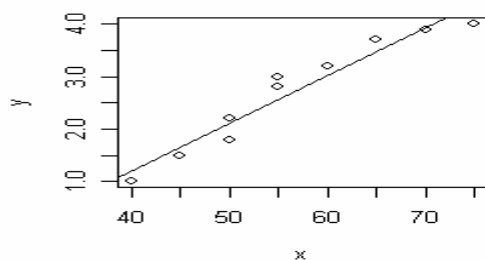
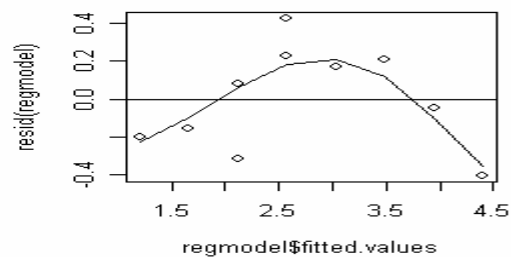
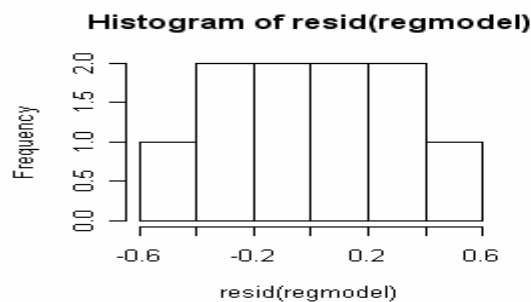
```
> abline(h=0)
```

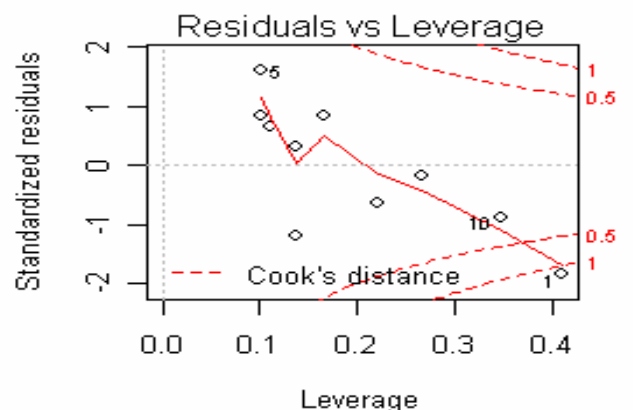
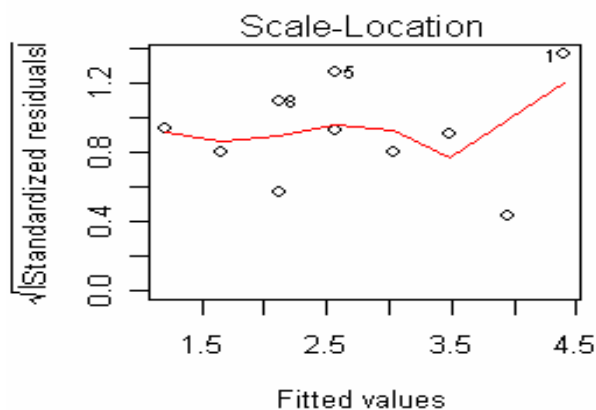
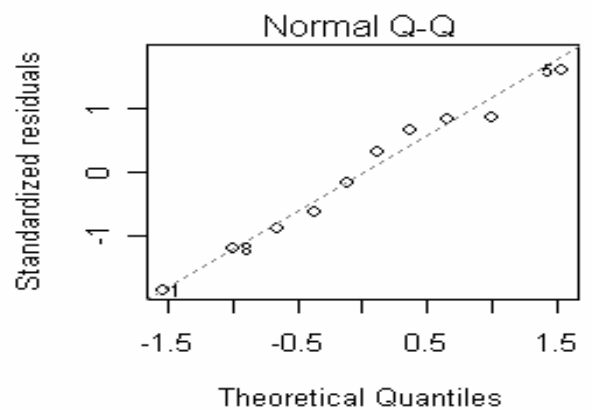
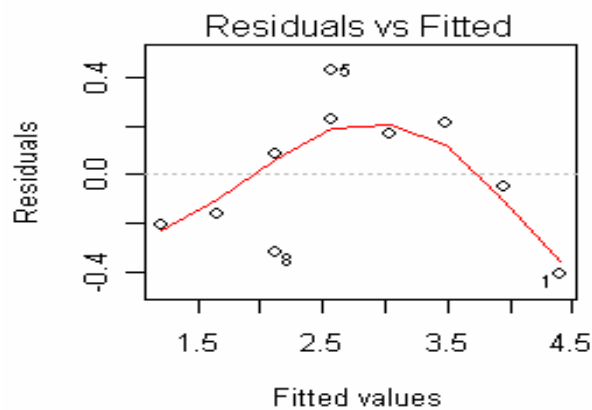
```
> plot(x,y)
```

```
> abline(regmodel)
```

```
> qqnorm(resid(regmodel))
```

```
> qqline(resid(regmodel))
```





plot(regmodel)

summary(regmodel)

Call:

lm(formula = y ~ x)

Residuals:

Min 1Q Median 3Q Max
 -0.40227 -0.19002 0.01984 0.20181 0.42721

Coefficients:

	Estimate	Std. Error	t	value	Pr(> t)
(Intercept)	-2.458277	0.491000	-5.007	0.00104	**
x	0.091474	0.008544	10.706	5.09e-06	***

Residual standard error: 0.2837 on 8 degrees of freedom

Multiple R-Squared: 0.9348, Adjusted R-squared: 0.9266

F-statistic: 114.6 on 1 and 8 DF, p-value: 5.088e-06

cor(yy)

x 1.000000 0.9668298

y 0.9668298 1.000000

> sqrt(0.9348)

[1] 0.9668506

```
regmodel=lm(y~log(x))
plot(regmodel)
summary(regmodel)
```

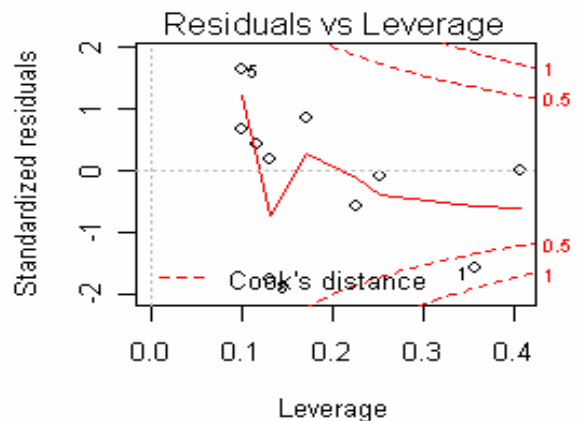
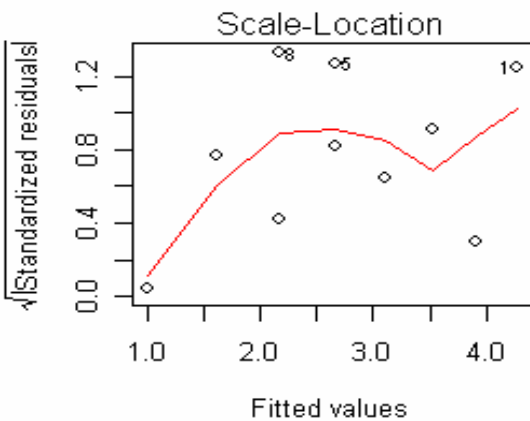
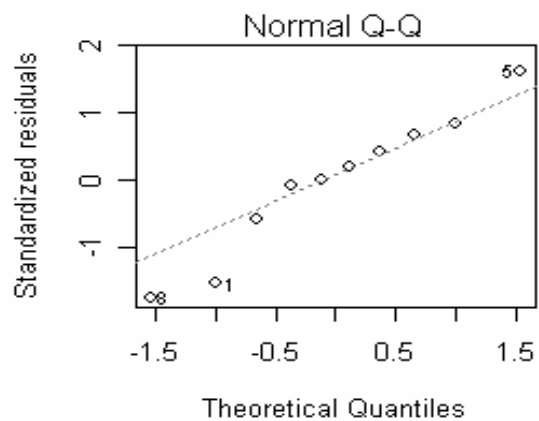
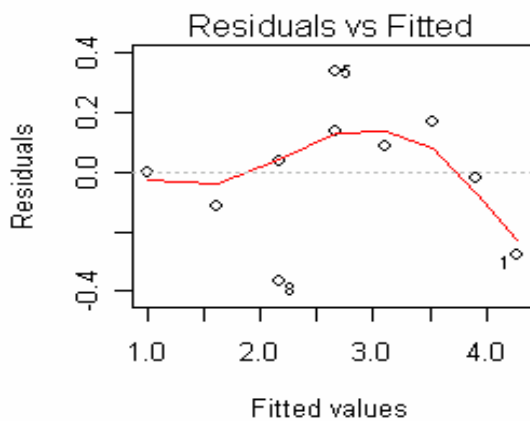
Residuals:

Min	1Q	Median	3Q	Max
-0.36339	-0.08996	0.01813	0.12647	0.33985

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-18.226	1.512	-12.06	2.07e-06 ***
log(x)	5.212	0.376	13.86	7.09e-07 ***

Residual standard error: 0.222 on 8 degrees of freedom
 Multiple R-Squared: 0.96, Adjusted R-squared: 0.955
 F-statistic: 192.2 on 1 and 8 DF, p-value: 7.09e-07



Multiple Regressions

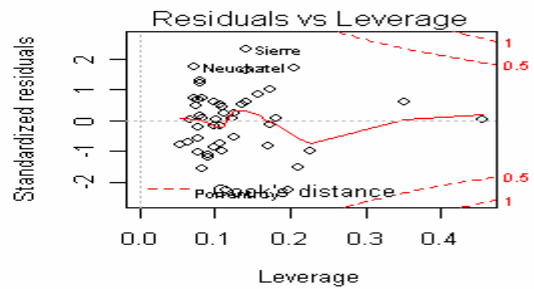
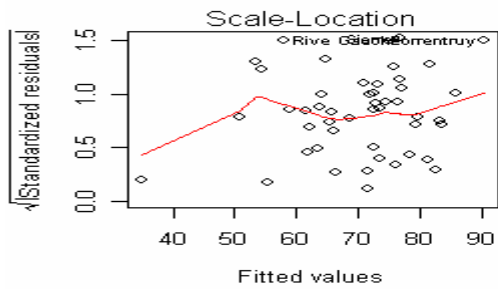
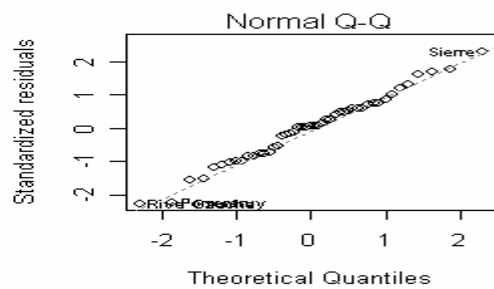
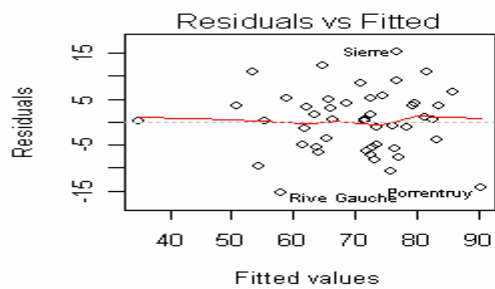
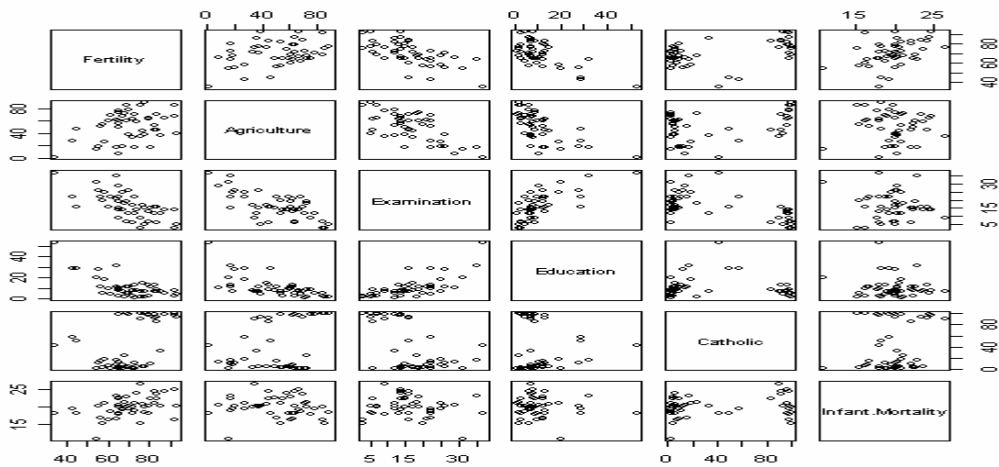
Example with swiss data

> pairs(swiss)

> cor(swiss)

Fertility

Fertility 1.0000000
Agriculture 0.3530792
Examination -0.6458827
Education -0.6637889
Catholic 0.4636847
Infant.Mortality 0.4165560



Multiple Regressions

```
summary(lm1 <- lm(Fertility ~ ., data = swiss))  
summary(lm1 <- lm(Fertility ~ Agriculture+ Examination+  
Education+ Catholic+ Infant.Mortality, data = swiss))
```

Call:

```
lm(formula = Fertility ~ ., data = swiss)
```

Residuals:

```
   Min     1Q  Median     3Q    Max  
-15.2743 -5.2617  0.5032  4.1198 15.3213
```

Coefficients:

```
           Estimate Std. Error t value Pr(>|t|)  
(Intercept)  66.91518  10.70604   6.250 1.91e-07 ***  
Agriculture   -0.17211   0.07030  -2.448 0.01873 *  
Examination   -0.25801   0.25388  -1.016 0.31546  
Education     -0.87094   0.18303  -4.758 2.43e-05 ***  
Catholic       0.10412   0.03526   2.953 0.00519 **  
Infant.Mortality 1.07705   0.38172   2.822 0.00734 **  
Residual standard error: 7.165 on 41 degrees of freedom  
Multiple R-Squared: 0.7067,    Adjusted R-squared: 0.671  
F-statistic: 19.76 on 5 and 41 DF,  p-value: 5.594e-10
```

AIC: the (generalized) Akaike Information Criterion for 'fit'.

$AIC = -2 * \log\text{-likelihood} + k * npar,$

where $npar$ represents the number of parameters in the fitted model, and $k = 2$ for the usual AIC, or $k = \log(n)$ (n the number of observations)

When comparing fitted objects, the smaller the AIC, the better the fit.

Single term additions

```
lm1 <- lm(Fertility ~ 1, data = swiss)  
add1(lm1, ~Agriculture+ Examination+ Education+ Catholic+  
Infant.Mortality)
```

Model:

```
Fertility ~ 1
```

```
           Df Sum of Sq  RSS  AIC  
<none>             7178.0 238.3  
Agriculture      1    894.8 6283.1 234.1  
Examination      1   2994.4 4183.6 215.0  
Education        1   3162.7 4015.2 213.0  
Catholic         1   1543.3 5634.7 229.0  
Infant.Mortality 1   1245.5 5932.4 231.4
```

sml1 <- step(lm1)

Start: AIC= 190.69

Fertility ~ Agriculture + Examination + Education + Catholic +
Infant.Mortality

	Df	Sum of Sq	RSS	AIC
- Examination	1	53.0	2158.1	189.9
<none>		2105.0	190.7	
- Agriculture	1	307.7	2412.8	195.1
- Infant.Mortality	1	408.8	2513.8	197.0
- Catholic	1	447.7	2552.8	197.8
- Education	1	1162.6	3267.6	209.4

Step: AIC= 189.86

Fertility ~ Agriculture + Education + Catholic + Infant.Mortality

	Df	Sum of Sq	RSS	AIC
<none>		2158.1	189.9	
- Agriculture	1	264.2	2422.2	193.3
- Infant.Mortality	1	409.8	2567.9	196.0
- Catholic	1	956.6	3114.6	205.1
- Education	1	2250.0	4408.0	221.4

Single term deletions

drop1(lm1)

Model:

Fertility ~ Agriculture + Examination + Education + Catholic +
Infant.Mortality

	Df	Sum of Sq	RSS	AIC
<none>		2105.0	190.7	
Agriculture	1	307.7	2412.8	195.1
Examination	1	53.0	2158.1	189.9
Education	1	1162.6	3267.6	209.4
Catholic	1	447.7	2552.8	197.8
Infant.Mortality	1	408.8	2513.8	197.0

```
lm4 <- lm(Fertility ~ Agriculture+ Education+ Catholic+
Infant.Mortality, data = swiss)
> lm1 <- lm(Fertility ~ Agriculture+ Examination+ Education+
Catholic+ Infant.Mortality, data = swiss)
```

```
> anova(lm1,lm4)
Analysis of Variance Table
```

```
Model 1: Fertility ~ Agriculture + Examination + Education + Catholic +
Infant.Mortality
```

```
Model 2: Fertility ~ Agriculture + Education + Catholic +
Infant.Mortality
```

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	41	2105.04				
2	42	2158.07	-1	-53.03	1.0328	0.3155

```
> lm3 <- lm(Fertility ~ Education+ Catholic+ Infant.Mortality, data
= swiss)
```

```
> anova(lm1,lm3)
Analysis of Variance Table
```

```
Model 1: Fertility ~ Agriculture + Examination + Education + Catholic +
Infant.Mortality
```

```
Model 2: Fertility ~ Education + Catholic + Infant.Mortality
```

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	41	2105.0				
2	43	2422.3	-2	-317.2	3.0891	0.05628 .

```
---
```

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