

Chap 9

- For controlled experiments model reduction is not very important. P 347
- For exploratory observational studies, model reduction is important.

Criteria for model selection **p353**

R-Sq criterion

Data : Surgical room data Chap 9

```
#Model selection
surgical=read.table("C:/Users/Mihinda/Desktop/surgical.txt",
header=1) #the data file
surgical
plot(surgical)
cor(surgical)
fit <- lm(Y ~ X1+X2+X3+X4, data=surgical)
summary(fit)
anova(fit)
plot(fitted(fit),residuals(fit))

# Try a log (i.e. ln) transformation on Y
surgical$lnY=log(surgical$Y)
fit2 <- lm(lnY ~ X1+X2+X3+X4, data=surgical)
summary(fit2)
```

```

anova(fit2)
plot(fitted(fit2),residuals(fit2))

#Variable selection
library(leaps)
X <- model.matrix(fit2)
X
X <- model.matrix(fit2)[,-1]
X

# R_sq method
r2.leaps <- leaps(X, surgical$lnY, nbest=3, method='r2')
r2.leaps
plot(r2.leaps$size, r2.leaps$r2)

# Adj R_sq method
adjr2.leaps <- leaps(X, surgical$lnY, nbest=3, method='adjr2')
adjr2.leaps
plot(adjr2.leaps$size, adjr2.leaps$adjr2)

# Cp method
Cp.leaps <- leaps(X, surgical$lnY, nbest=3, method='Cp')
Cp.leaps

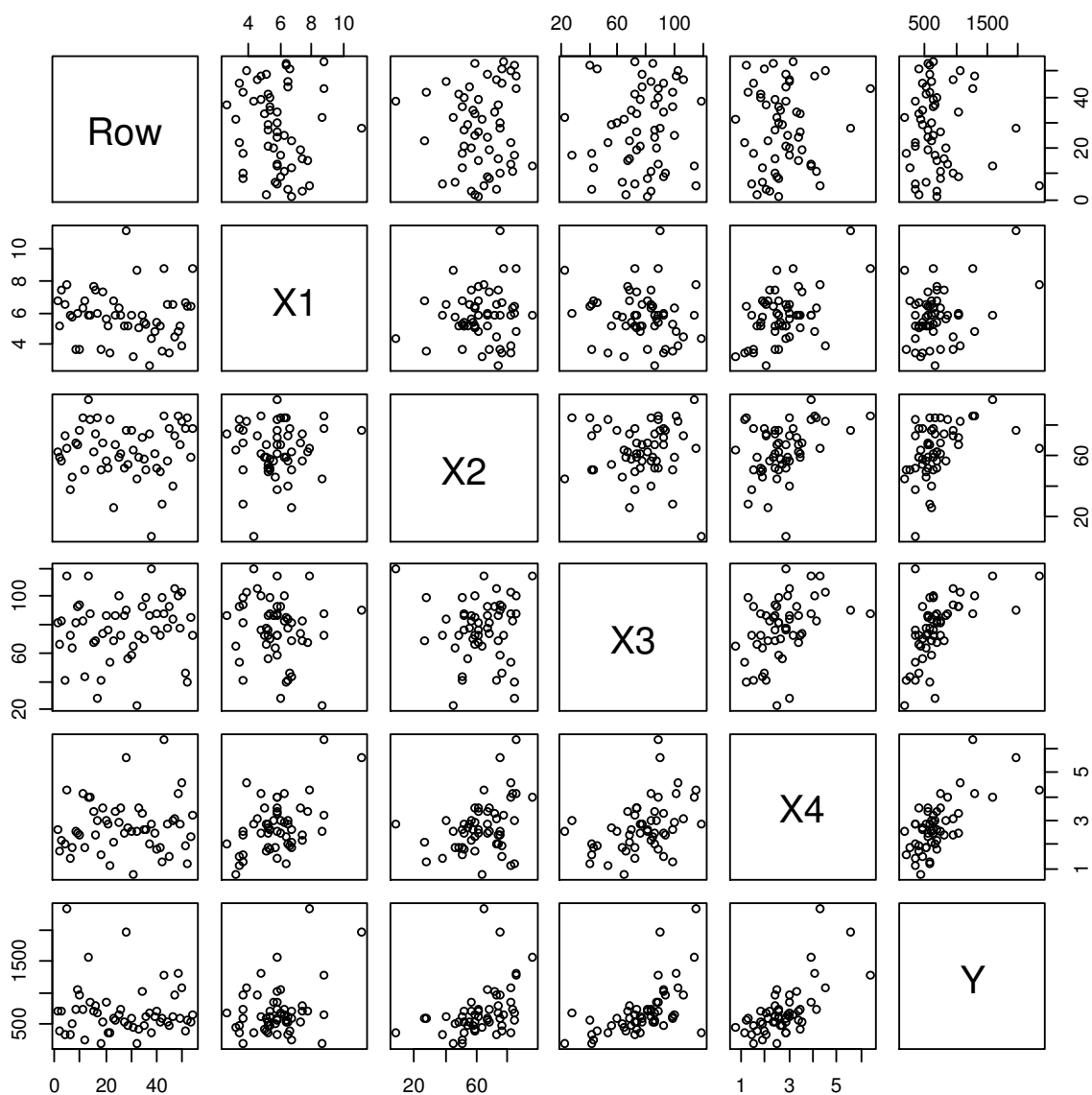
```

R output

```
> #Model selection
> surgical=read.table("C:/Users/Mihinda/Desktop/surgical.txt",
header=1) #the data file
> surgical
```

	Row	X1	X2	X3	X4	Y
1	1	6.7	62	81	2.59	695
2	2	5.1	59	66	1.70	403
3	3	7.4	57	83	2.16	710
4	4	6.5	73	41	2.01	349
5	5	7.8	65	115	4.30	2343
6	6	5.8	38	72	1.42	348
7	7	5.7	46	63	1.91	518
8	8	3.7	68	81	2.57	749
9	9	6.0	67	93	2.50	1056
10	10	3.7	76	94	2.40	968
11	11	6.3	84	83	4.13	745
12	12	6.7	51	43	1.86	257
13	13	5.8	96	114	3.95	1573
14	14	5.8	83	88	3.95	858
15	15	7.7	62	67	3.40	702
16	16	7.4	74	68	2.40	809
17	17	6.0	85	28	2.98	682
18	18	3.7	51	41	1.55	205
19	19	7.3	68	74	3.56	550
20	20	5.6	57	87	3.02	838
21	21	5.2	52	76	2.85	359
22	22	3.4	83	53	1.12	353
23	23	6.7	26	68	2.10	599
24	24	5.8	67	86	3.40	562
25	25	6.3	59	100	2.95	651
26	26	5.8	61	73	3.50	751
27	27	5.2	52	86	2.45	545
28	28	11.2	76	90	5.59	1965
29	29	5.2	54	56	2.71	477
30	30	5.8	76	59	2.58	600
31	31	3.2	64	65	0.74	443
32	32	8.7	45	23	2.52	181
33	33	5.0	59	73	3.50	411
34	34	5.8	72	93	3.30	1037
35	35	5.4	58	70	2.64	482
36	36	5.3	51	99	2.60	634
37	37	2.6	74	86	2.05	678
38	38	4.3	8	119	2.85	362
39	39	4.8	61	76	2.45	637
40	40	5.4	52	88	1.81	705

```
41  41  5.2 49  72 1.84  536
42  42  3.6 28  99 1.30  582
43  43  8.8 86  88 6.40 1270
44  44  6.5 56  77 2.85  538
45  45  3.4 77  93 1.48  482
46  46  6.5 40  84 3.00  611
47  47  4.5 73 106 3.05  960
48  48  4.8 86 101 4.10 1300
49  49  5.1 67  77 2.86  581
50  50  3.9 82 103 4.55 1078
51  51  6.6 77  46 1.95  405
52  52  6.4 85  40 1.21  579
53  53  6.4 59  85 2.33  550
54  54  8.8 78  72 3.20  651
> plot(surgical)
```



```
> cor(surgical)
```

```

      Row      X1      X2      X3      X4      Y
Row  1.0000000 -0.09812105  0.02813377  0.11488890  0.06163258 -0.07831648
X1  -0.09812105  1.00000000  0.09011973 -0.14963411  0.50241567  0.34654968
X2   0.02813377  0.09011973  1.00000000 -0.02360544  0.36902563  0.42048097
X3   0.11488890 -0.14963411 -0.02360544  1.00000000  0.41642451  0.57822600
X4   0.06163258  0.50241567  0.36902563  0.41642451  1.00000000  0.67419499
Y   -0.07831648  0.34654968  0.42048097  0.57822600  0.67419499  1.00000000
> fit <- lm(Y ~ X1+X2+X3+X4, data=surgical)
```

```

> summary(fit)

Call:
lm(formula = Y ~ X1 + X2 + X3 + X4, data = surgical)

Residuals:
    Min       1Q   Median       3Q      Max
-391.55 -144.81   -8.34  129.51  970.26

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept) -1279.242    243.808   -5.247 3.30e-06 ***
X1             82.988     26.402    3.143 0.00284 **
X2             8.346      2.120    3.937 0.00026 ***
X3            10.870      1.923    5.652 8.01e-07 ***
X4            49.346     47.126    1.047 0.30018
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

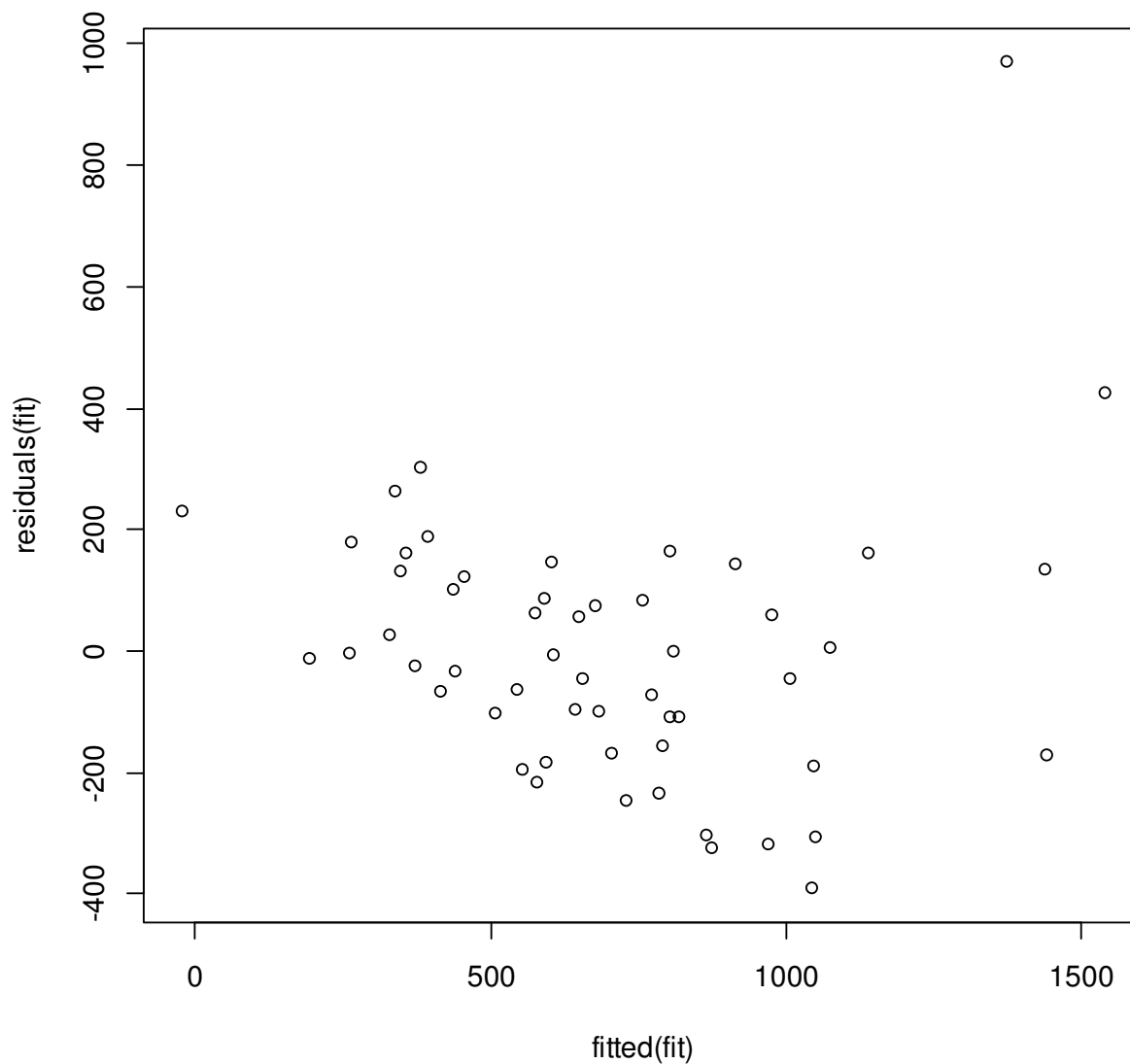
Residual standard error: 229.7 on 49 degrees of freedom
Multiple R-squared: 0.691, Adjusted R-squared: 0.6658
F-statistic: 27.4 on 4 and 49 DF, p-value: 5.704e-12

> anova(fit)
Analysis of Variance Table

Response: Y
      Df Sum Sq Mean Sq F value    Pr(>F)
X1      1 1005152 1005152  19.0470 6.567e-05 ***
X2      1 1278496 1278496  24.2267 1.010e-05 ***
X3      1 3442172 3442172  65.2269 1.461e-10 ***
X4      1   57862   57862   1.0964  0.3002
Residuals 49 2585839   52772
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
plot(fitted(fit), residuals(fit))
```



```
> # Try a log (i.e. ln) transformation on Y
> surgical$lnY=log(surgical$Y)
> fit2 <- lm(lnY ~ X1+X2+X3+X4, data=surgical)
> summary(fit2)
```

```
Call:
lm(formula = lnY ~ X1 + X2 + X3 + X4, data = surgical)
```

```
Residuals:
    Min       1Q   Median       3Q      Max
```

-0.43514 -0.17436 -0.02156 0.18475 0.56054

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	3.851933	0.266263	14.467	< 2e-16	***
X1	0.083739	0.028834	2.904	0.00551	**
X2	0.012671	0.002315	5.474	1.50e-06	***
X3	0.015627	0.002100	7.440	1.38e-09	***
X4	0.032056	0.051466	0.623	0.53627	

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

Residual standard error: 0.2509 on 49 degrees of freedom

Multiple R-squared: 0.7591, Adjusted R-squared: 0.7395

F-statistic: 38.61 on 4 and 49 DF, p-value: 1.398e-14

> anova(fit2)

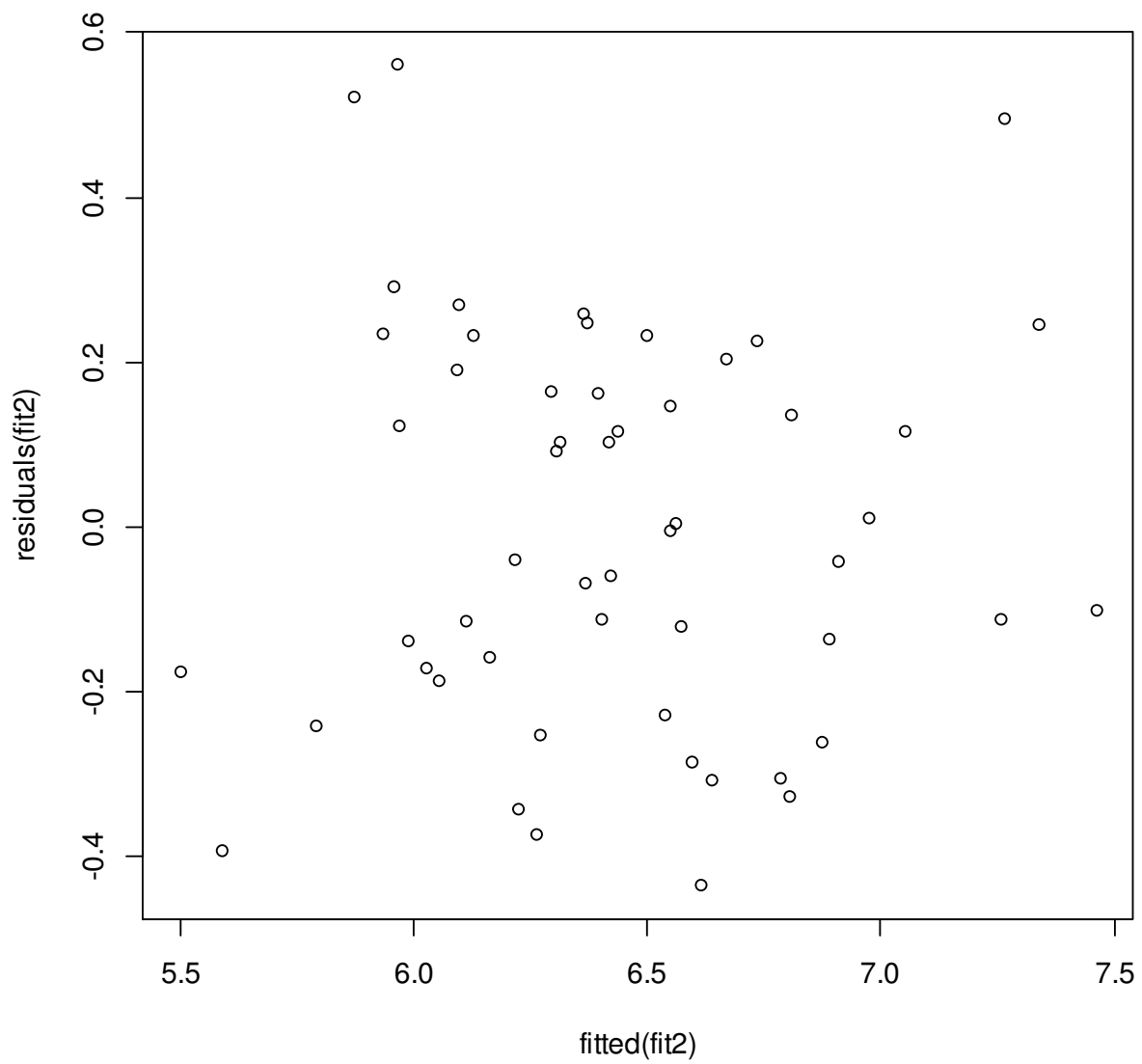
Analysis of Variance Table

Response: lnY

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
X1	1	0.7770	0.7770	12.3443	0.0009618	***
X2	1	2.5904	2.5904	41.1565	5.341e-08	***
X3	1	6.3286	6.3286	100.5490	1.838e-13	***
X4	1	0.0244	0.0244	0.3879	0.5362698	
Residuals	49	3.0841	0.0629			

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

> plot(fitted(fit2),residuals(fit2))



Variable selection methods

```
> #Variable selection
> library(leaps)
> X <- model.matrix(fit2)
> X
```

	(Intercept)	X1	X2	X3	X4
1	1	6.7	62	81	2.59
2	1	5.1	59	66	1.70
3	1	7.4	57	83	2.16
4	1	6.5	73	41	2.01
5	1	7.8	65	115	4.30
6	1	5.8	38	72	1.42
7	1	5.7	46	63	1.91
8	1	3.7	68	81	2.57
9	1	6.0	67	93	2.50
10	1	3.7	76	94	2.40
11	1	6.3	84	83	4.13
12	1	6.7	51	43	1.86
13	1	5.8	96	114	3.95
14	1	5.8	83	88	3.95
15	1	7.7	62	67	3.40
16	1	7.4	74	68	2.40
17	1	6.0	85	28	2.98
18	1	3.7	51	41	1.55
19	1	7.3	68	74	3.56
20	1	5.6	57	87	3.02
21	1	5.2	52	76	2.85
22	1	3.4	83	53	1.12
23	1	6.7	26	68	2.10
24	1	5.8	67	86	3.40
25	1	6.3	59	100	2.95
26	1	5.8	61	73	3.50
27	1	5.2	52	86	2.45
28	1	11.2	76	90	5.59
29	1	5.2	54	56	2.71
30	1	5.8	76	59	2.58
31	1	3.2	64	65	0.74
32	1	8.7	45	23	2.52
33	1	5.0	59	73	3.50
34	1	5.8	72	93	3.30
35	1	5.4	58	70	2.64
36	1	5.3	51	99	2.60
37	1	2.6	74	86	2.05
38	1	4.3	8	119	2.85
39	1	4.8	61	76	2.45

```

40          1  5.4 52  88 1.81
41          1  5.2 49  72 1.84
42          1  3.6 28  99 1.30
43          1  8.8 86  88 6.40
44          1  6.5 56  77 2.85
45          1  3.4 77  93 1.48
46          1  6.5 40  84 3.00
47          1  4.5 73 106 3.05
48          1  4.8 86 101 4.10
49          1  5.1 67  77 2.86
50          1  3.9 82 103 4.55
51          1  6.6 77  46 1.95
52          1  6.4 85  40 1.21
53          1  6.4 59  85 2.33
54          1  8.8 78  72 3.20

```

```
attr(,"assign")
```

```
[1] 0 1 2 3 4
```

```
> X <- model.matrix(fit2)[,-1]
```

```
> X
```

```

      X1 X2  X3   X4
1  6.7 62  81 2.59
2  5.1 59  66 1.70
3  7.4 57  83 2.16
4  6.5 73  41 2.01
5  7.8 65 115 4.30
6  5.8 38  72 1.42
7  5.7 46  63 1.91
8  3.7 68  81 2.57
9  6.0 67  93 2.50
10 3.7 76  94 2.40
11 6.3 84  83 4.13
12 6.7 51  43 1.86
13 5.8 96 114 3.95
14 5.8 83  88 3.95
15 7.7 62  67 3.40
16 7.4 74  68 2.40
17 6.0 85  28 2.98
18 3.7 51  41 1.55
19 7.3 68  74 3.56
20 5.6 57  87 3.02
21 5.2 52  76 2.85
22 3.4 83  53 1.12
23 6.7 26  68 2.10
24 5.8 67  86 3.40
25 6.3 59 100 2.95
26 5.8 61  73 3.50
27 5.2 52  86 2.45

```

```

28 11.2 76 90 5.59
29 5.2 54 56 2.71
30 5.8 76 59 2.58
31 3.2 64 65 0.74
32 8.7 45 23 2.52
33 5.0 59 73 3.50
34 5.8 72 93 3.30
35 5.4 58 70 2.64
36 5.3 51 99 2.60
37 2.6 74 86 2.05
38 4.3 8 119 2.85
39 4.8 61 76 2.45
40 5.4 52 88 1.81
41 5.2 49 72 1.84
42 3.6 28 99 1.30
43 8.8 86 88 6.40
44 6.5 56 77 2.85
45 3.4 77 93 1.48
46 6.5 40 84 3.00
47 4.5 73 106 3.05
48 4.8 86 101 4.10
49 5.1 67 77 2.86
50 3.9 82 103 4.55
51 6.6 77 46 1.95
52 6.4 85 40 1.21
53 6.4 59 85 2.33
54 8.8 78 72 3.20
>
> # R_sq method
> r2.leaps <- leaps(X, surgical$lnY, nbest=3, method='r2')
> r2.leaps
$which
      1      2      3      4
1 FALSE FALSE TRUE FALSE
1 FALSE FALSE FALSE TRUE
1 FALSE TRUE FALSE FALSE
2 FALSE TRUE TRUE FALSE
2 FALSE FALSE TRUE TRUE
2 TRUE FALSE TRUE FALSE
3 TRUE TRUE TRUE FALSE
3 FALSE TRUE TRUE TRUE
3 TRUE FALSE TRUE TRUE
4 TRUE TRUE TRUE TRUE

$label
[1] "(Intercept)" "1"      "2"      "3"      "4"

```

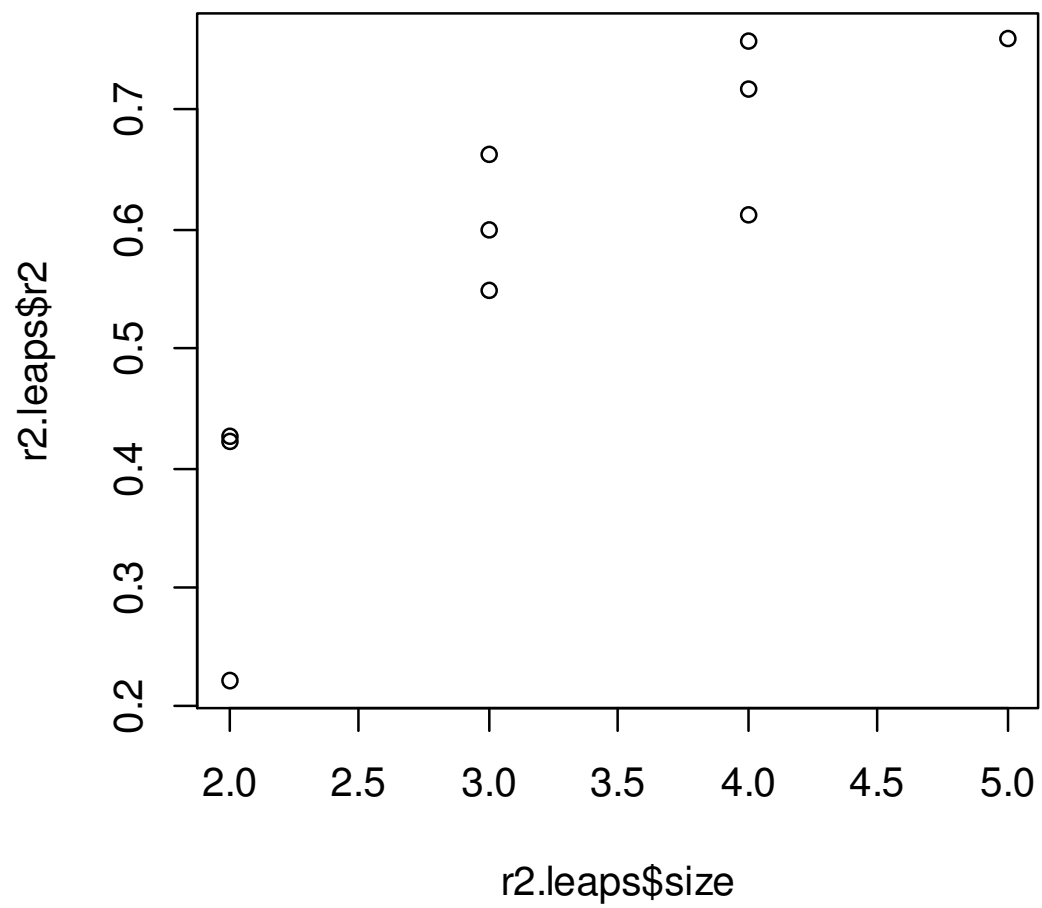
```

$size
[1] 2 2 2 3 3 3 4 4 4 5

$r2
[1] 0.4272545 0.4214649 0.2210432 0.6631757 0.5992084 0.5483980 0.7572331
[8] 0.7176812 0.6118710 0.7591401

> plot(r2.leaps$size, r2.leaps$r2)

```



Adjusted R sq method

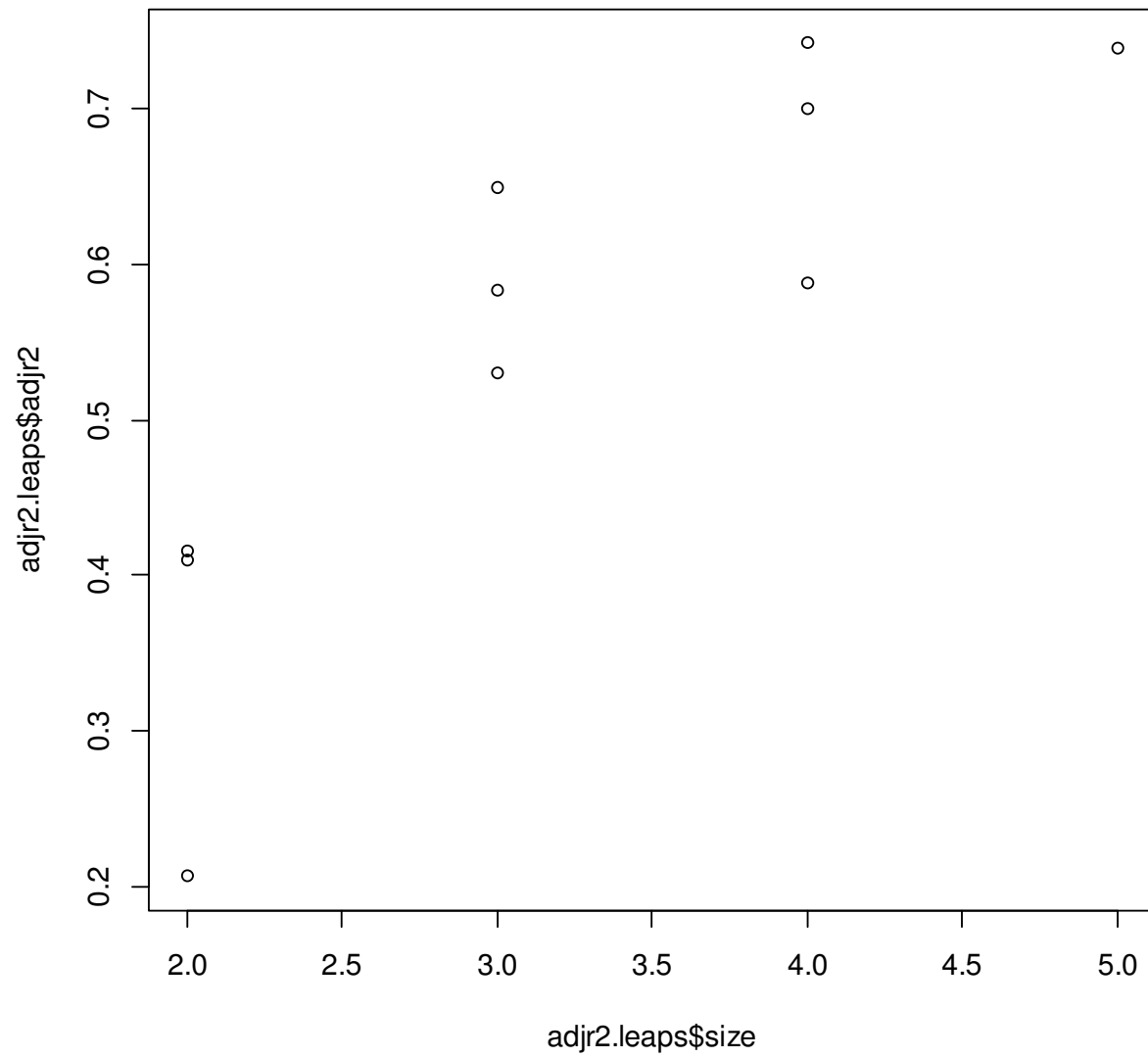
```
> # Adj R_sq method
> adjr2.leaps <- leaps(X, surgical$lnY, nbest=3, method='adjr2')
> adjr2.leaps
$which
      1      2      3      4
1 FALSE FALSE  TRUE FALSE
1 FALSE FALSE FALSE  TRUE
1 FALSE  TRUE FALSE FALSE
2 FALSE  TRUE  TRUE FALSE
2 FALSE FALSE  TRUE  TRUE
2  TRUE FALSE  TRUE FALSE
3  TRUE  TRUE  TRUE FALSE
3 FALSE  TRUE  TRUE  TRUE
3  TRUE FALSE  TRUE  TRUE
4  TRUE  TRUE  TRUE  TRUE

$label
[1] "(Intercept)" "1"      "2"      "3"      "4"

$size
[1] 2 2 2 3 3 3 4 4 4 5

$adjr2
[1] 0.4162401 0.4103392 0.2060632 0.6499669 0.5834910 0.5306881 0.7426671
[8] 0.7007420 0.5885833 0.7394781
```

```
> plot(adjr2.leaps$size, adjr2.leaps$adjr2)
```



Mallows C_p criterion p357

$$C_p = \frac{SSE_p}{MSE_P} - (n - 2p)$$

$P-1$ is the number of potential X variables.

$p-1$ is the X variables in a subset.

When $p = P$, $C_p = P$

The basic idea is to find the smallest value of p such that $C_p \approx p$.

```

> # Cp method
> Cp.leaps <- leaps(X, surgical$lnY, method='Cp')
> Cp.leaps
$which
      1      2      3      4
1 FALSE FALSE  TRUE FALSE
1 FALSE FALSE FALSE  TRUE
1 FALSE  TRUE FALSE FALSE
1  TRUE FALSE FALSE FALSE
2 FALSE  TRUE  TRUE FALSE
2 FALSE FALSE  TRUE  TRUE
2  TRUE FALSE  TRUE FALSE
2 FALSE  TRUE FALSE  TRUE
2  TRUE FALSE FALSE  TRUE
2  TRUE  TRUE FALSE FALSE
3  TRUE  TRUE  TRUE FALSE
3 FALSE  TRUE  TRUE  TRUE
3  TRUE FALSE  TRUE  TRUE
3  TRUE  TRUE FALSE  TRUE
4  TRUE  TRUE  TRUE  TRUE

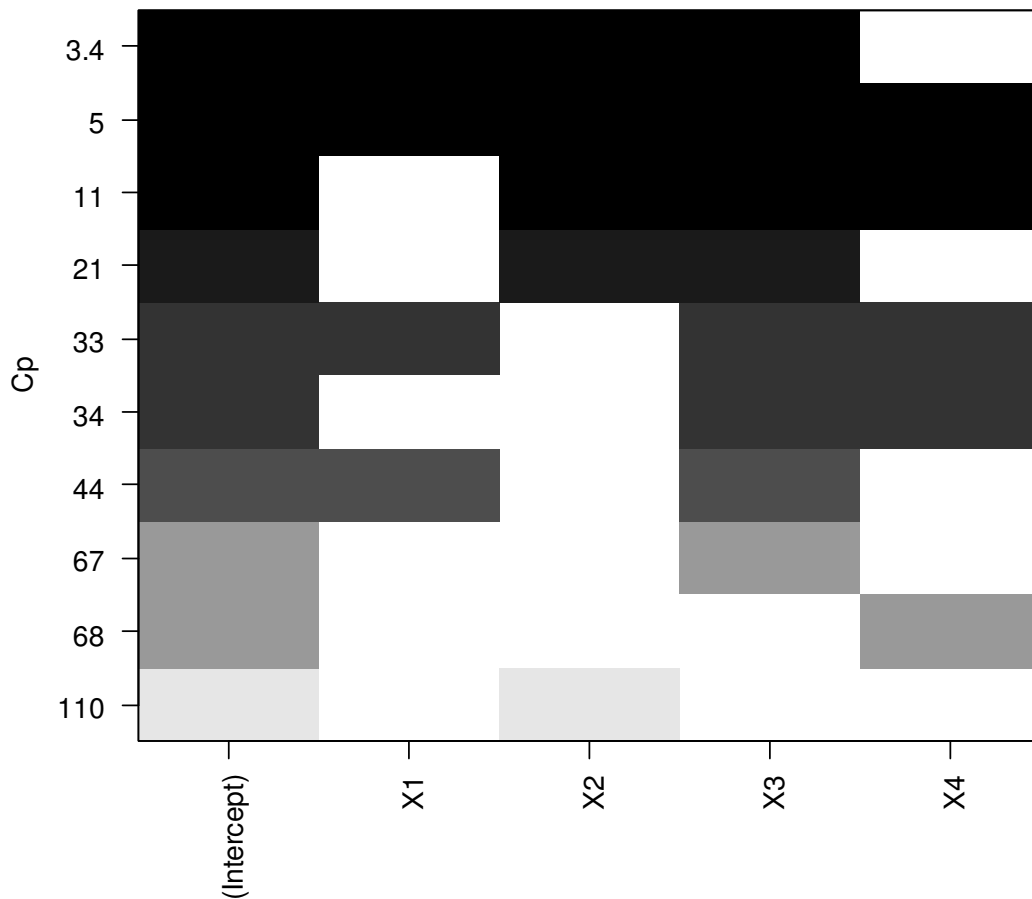
$label
[1] "(Intercept)" "1"      "2"      "3"      "4"

$size
[1] 2 2 2 2 3 3 3 3 3 3 4 4 4 4 5

$Cp
[1] 66.518070 67.695892 108.469236 141.093441 20.522784 33.536154
[7] 43.872905 57.174572 67.961243 101.936972 3.387945 11.434313
[13] 32.960084 58.357601 5.000000

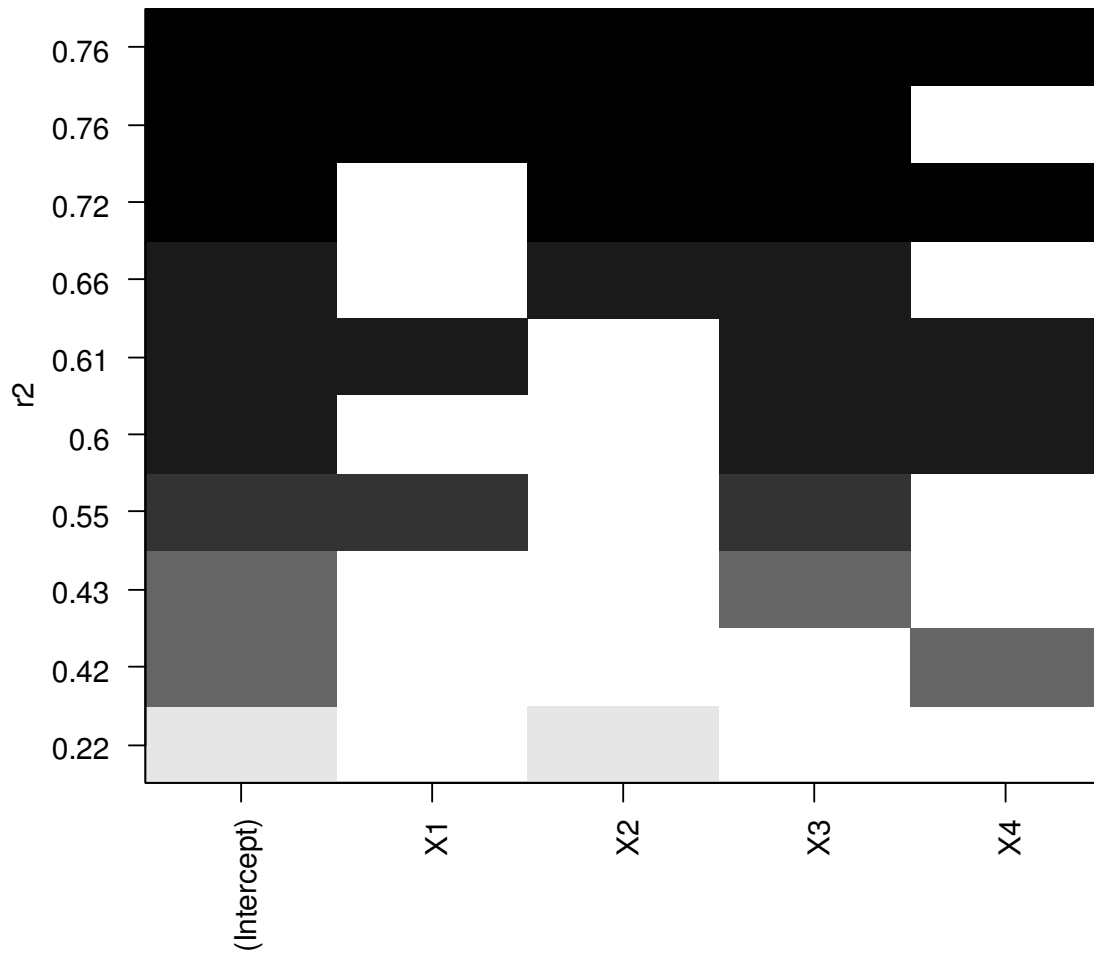
```

```
plot(leaps, scale="Cp")
```



Black indicates that a variable is included in the model, while white indicates that they are not.

```
> plot(leaps, scale="r2")
```



Step wise selection with AIC_p criterion
p359

$$AIC_p = n \ln SSE_p - n \ln n + 2p$$

We look for models with smaller AIC_p

Note: SSE_p decreases as p increases, the second term is fixed and the third term (i.e. penalty) increases as p increases. The models with smaller SSE_p will do well by this criterion as long as the penalties ($2p$) is not too large.

```

> #Stepwise regression
> # Remove the unnecessary columns i.e. row and Y
> # from the data set to avoid them going to candidate set.
> surgical=surgical[-1]
> surgical

```

	X1	X2	X3	X4	Y	lnY
1	6.7	62	81	2.59	695	6.543912
2	5.1	59	66	1.70	403	5.998937
3	7.4	57	83	2.16	710	6.565265
4	6.5	73	41	2.01	349	5.855072
5	7.8	65	115	4.30	2343	7.759187
6	5.8	38	72	1.42	348	5.852202
7	5.7	46	63	1.91	518	6.249975
8	3.7	68	81	2.57	749	6.618739
9	6.0	67	93	2.50	1056	6.962243
10	3.7	76	94	2.40	968	6.875232
11	6.3	84	83	4.13	745	6.613384
12	6.7	51	43	1.86	257	5.549076
13	5.8	96	114	3.95	1573	7.360740
14	5.8	83	88	3.95	858	6.754604
15	7.7	62	67	3.40	702	6.553933
16	7.4	74	68	2.40	809	6.695799
17	6.0	85	28	2.98	682	6.525030
18	3.7	51	41	1.55	205	5.323010
19	7.3	68	74	3.56	550	6.309918
20	5.6	57	87	3.02	838	6.731018
21	5.2	52	76	2.85	359	5.883322
22	3.4	83	53	1.12	353	5.866468
23	6.7	26	68	2.10	599	6.395262
24	5.8	67	86	3.40	562	6.331502
25	6.3	59	100	2.95	651	6.478510
26	5.8	61	73	3.50	751	6.621406
27	5.2	52	86	2.45	545	6.300786
28	11.2	76	90	5.59	1965	7.583248
29	5.2	54	56	2.71	477	6.167516
30	5.8	76	59	2.58	600	6.396930
31	3.2	64	65	0.74	443	6.093570
32	8.7	45	23	2.52	181	5.198497
33	5.0	59	73	3.50	411	6.018593
34	5.8	72	93	3.30	1037	6.944087
35	5.4	58	70	2.64	482	6.177944
36	5.3	51	99	2.60	634	6.452049
37	2.6	74	86	2.05	678	6.519147
38	4.3	8	119	2.85	362	5.891644
39	4.8	61	76	2.45	637	6.456770
40	5.4	52	88	1.81	705	6.558198
41	5.2	49	72	1.84	536	6.284134
42	3.6	28	99	1.30	582	6.366470
43	8.8	86	88	6.40	1270	7.146772
44	6.5	56	77	2.85	538	6.287859
45	3.4	77	93	1.48	482	6.177944
46	6.5	40	84	3.00	611	6.415097
47	4.5	73	106	3.05	960	6.866933
48	4.8	86	101	4.10	1300	7.170120
49	5.1	67	77	2.86	581	6.364751
50	3.9	82	103	4.55	1078	6.982863
51	6.6	77	46	1.95	405	6.003887
52	6.4	85	40	1.21	579	6.361302
53	6.4	59	85	2.33	550	6.309918
54	8.8	78	72	3.20	651	6.478510

```

> surgical=surgical[-5]
> surgical
      X1 X2  X3   X4      lnY
1  6.7 62  81 2.59 6.543912
2  5.1 59  66 1.70 5.998937
3  7.4 57  83 2.16 6.565265
4  6.5 73  41 2.01 5.855072
5  7.8 65 115 4.30 7.759187
6  5.8 38  72 1.42 5.852202
7  5.7 46  63 1.91 6.249975
8  3.7 68  81 2.57 6.618739
9  6.0 67  93 2.50 6.962243
10 3.7 76  94 2.40 6.875232
11 6.3 84  83 4.13 6.613384
12 6.7 51  43 1.86 5.549076
13 5.8 96 114 3.95 7.360740
14 5.8 83  88 3.95 6.754604
15 7.7 62  67 3.40 6.553933
16 7.4 74  68 2.40 6.695799
17 6.0 85  28 2.98 6.525030
18 3.7 51  41 1.55 5.323010
19 7.3 68  74 3.56 6.309918
20 5.6 57  87 3.02 6.731018
21 5.2 52  76 2.85 5.883322
22 3.4 83  53 1.12 5.866468
23 6.7 26  68 2.10 6.395262
24 5.8 67  86 3.40 6.331502
25 6.3 59 100 2.95 6.478510
26 5.8 61  73 3.50 6.621406
27 5.2 52  86 2.45 6.300786
28 11.2 76  90 5.59 7.583248
29 5.2 54  56 2.71 6.167516
30 5.8 76  59 2.58 6.396930
31 3.2 64  65 0.74 6.093570
32 8.7 45  23 2.52 5.198497
33 5.0 59  73 3.50 6.018593
34 5.8 72  93 3.30 6.944087
35 5.4 58  70 2.64 6.177944
36 5.3 51  99 2.60 6.452049
37 2.6 74  86 2.05 6.519147
38 4.3  8 119 2.85 5.891644
39 4.8 61  76 2.45 6.456770
40 5.4 52  88 1.81 6.558198
41 5.2 49  72 1.84 6.284134
42 3.6 28  99 1.30 6.366470
43 8.8 86  88 6.40 7.146772
44 6.5 56  77 2.85 6.287859
45 3.4 77  93 1.48 6.177944
46 6.5 40  84 3.00 6.415097
47 4.5 73 106 3.05 6.866933
48 4.8 86 101 4.10 7.170120
49 5.1 67  77 2.86 6.364751
50 3.9 82 103 4.55 6.982863
51 6.6 77  46 1.95 6.003887
52 6.4 85  40 1.21 6.361302
53 6.4 59  85 2.33 6.309918
54 8.8 78  72 3.20 6.478510
>

```

```
> null=lm(lnY~1, data=surgical)
> null
```

```
Call:
lm(formula = lnY ~ 1, data = surgical)
```

```
Coefficients:
(Intercept)
      6.431
```

```
> full=lm(lnY~., data=surgical)
> full
```

```
Call:
lm(formula = lnY ~ ., data = surgical)
```

```
Coefficients:
(Intercept)          X1          X2          X3          X4
   3.85193    0.08374    0.01267    0.01563    0.03206
```

Forward selection

```
> step(null, scope=list(lower=null, upper=full), direction="forward")
```

```
Start:  AIC=-75.72
```

```
lnY ~ 1
```

		Df	Sum of Sq	RSS	AIC
+	X3	1	5.4708	7.3337	-103.811
+	X4	1	5.3967	7.4079	-103.268
+	X2	1	2.8303	9.9742	-87.205
+	X1	1	0.7770	12.0275	-77.096
<none>				12.8045	-75.716

```
Step:  AIC=-103.81
```

```
lnY ~ X3
```

		Df	Sum of Sq	RSS	AIC
+	X2	1	3.0209	4.3129	-130.48
+	X4	1	2.2018	5.1319	-121.09
+	X1	1	1.5512	5.7825	-114.64
<none>				7.3337	-103.81

```
Step:  AIC=-130.48
```

```
lnY ~ X3 + X2
```

	Df	Sum of Sq	RSS	AIC
+ X1	1	1.20436	3.1085	-146.16
+ X4	1	0.69792	3.6150	-138.01
<none>			4.3129	-130.48

Step: AIC=-146.16
lnY ~ X3 + X2 + X1

	Df	Sum of Sq	RSS	AIC
<none>			3.1085	-146.16
+ X4	1	0.024418	3.0841	-144.59

Call:

```
lm(formula = lnY ~ X3 + X2 + X1, data = surgical)
```

Coefficients:

(Intercept)	X3	X2	X1
3.76644	0.01644	0.01334	0.09547

```
> fitx3 <- lm(lnY ~ X3, data=surgical)
```

```
> anova(fitx3)
```

Analysis of Variance Table

Response: lnY

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
X3	1	5.4708	5.4708	38.791	8.381e-08

Residuals	52	7.3337	0.1410		

n=54

p=2

SSE = 7.3337

AIC = n*ln(SSE)-n*ln(n)+2*p = -103.8112097

```
> fitx1x2x3 <- lm(lnY ~ X1+X2+X3, data=surgical)
> summary(fitx1x2x3)
```

Call:

```
lm(formula = lnY ~ X1 + X2 + X3, data = surgical)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	3.766441	0.226757	16.610	< 2e-16	***
X1	0.095475	0.021692	4.401	5.66e-05	***
X2	0.013344	0.002035	6.558	2.95e-08	***
X3	0.016444	0.001630	10.089	1.19e-13	***

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

Residual standard error: 0.2493 on 50 degrees of freedom

Multiple R-squared: 0.7572, Adjusted R-squared: 0.7427

F-statistic: 51.99 on 3 and 50 DF, p-value: 2.137e-15

```
> step(full, scope=list(lower=null, upper=full), direction="backward")
Start:  AIC=-144.59
lnY ~ X1 + X2 + X3 + X4
```

	Df	Sum of Sq	RSS	AIC
- X4	1	0.0244	3.1085	-146.16
<none>			3.0841	-144.59
- X1	1	0.5309	3.6150	-138.01
- X2	1	1.8857	4.9698	-120.82
- X3	1	3.4842	6.5683	-105.76

```
Step:  AIC=-146.16
lnY ~ X1 + X2 + X3
```

	Df	Sum of Sq	RSS	AIC
<none>			3.1085	-146.161
- X1	1	1.2044	4.3129	-130.479
- X2	1	2.6740	5.7825	-114.644
- X3	1	6.3286	9.4371	-88.194

```
Call:
lm(formula = lnY ~ X1 + X2 + X3, data = surgical)
```

```
Coefficients:
(Intercept)          X1          X2          X3
   3.76644      0.09547      0.01334      0.01644
```

```
> step(full, scope=list(lower=null, upper=full), direction="both")
Start:  AIC=-144.59
lnY ~ X1 + X2 + X3 + X4
```

	Df	Sum of Sq	RSS	AIC
- X4	1	0.0244	3.1085	-146.16
<none>			3.0841	-144.59
- X1	1	0.5309	3.6150	-138.01
- X2	1	1.8857	4.9698	-120.82
- X3	1	3.4842	6.5683	-105.76

```
Step:  AIC=-146.16
lnY ~ X1 + X2 + X3
```

	Df	Sum of Sq	RSS	AIC
<none>			3.1085	-146.161
+ X4	1	0.0244	3.0841	-144.587
- X1	1	1.2044	4.3129	-130.479
- X2	1	2.6740	5.7825	-114.644
- X3	1	6.3286	9.4371	-88.194

```
Call:
lm(formula = lnY ~ X1 + X2 + X3, data = surgical)
```

```
Coefficients:
(Intercept)          X1          X2          X3
   3.76644    0.09547    0.01334    0.01644
```

```
>
```