



1) Use the above plot to give the transformations, if any, that you would use to remove strong nonlinearities from variables.

$\log(x_1), \log(x_2), \log(x_3), \log(y)$

ask which ones need log trans or log rule

	L1	L2	L3	L4
# of predictors	6	4	3	2
# with $0.01 \leq p\text{-value} \leq 0.05$	0	0	0	0
# with $p\text{-value} > 0.05$	3	2	0	0
R^2	0.913	0.913	0.913	0.900
$\text{corr}(\hat{Y}, \hat{Y}_I)$	1.0	0.9999	0.9996	0.9926
$C_p(I)$	6.0	2.190	0.793	14.389
\sqrt{MSE}	20.17	20.00	19.96	21.28
p-value for partial F test	1.0	0.910	0.851	0.004

$\frac{\max}{\min} > 10$

2) The above table gives summary statistics for 4 MLR models considered as final submodels after performing variable selection. The data set had 112 cases and the response and residual plots for the full model L1 was good. Which model should be used as the final submodel? Explain briefly why each of the other 3 submodels should not be used.

L3 should be used $L_3 = I_{\min} = II$

L1 both have too many predictors
L2

L4 has C_p too high ($C_p > 2k$) and partial F test pval too low

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Backward

Current terms: (x3 x4 x6)

	df	RSS	k	C _I
Delete: x4	109	43441.6	<u>3</u>	0.793
Delete: x6	109	48569.8	3	13.399
Delete: x3	109	194726.	3	372.694

I_{min}

constant
x3, x6

Current terms: (x3 x6)

	df	RSS	k	C _I
Delete: x6	110	49786.2	2	14.389
Delete: x3	110	196210.	2	374.343

x4 is
deleted

Forward

Base terms: Intercept

	df	RSS	k	C _I
Add: x3	110	49786.2	2	14.389
Add: x6	110	196210.	2	374.343
Add: x2	110	205931.	2	398.241
Add: x4	110	431391.	2	952.489
Add: x5	110	461290.	2	1025.988

Base terms: (x3)

	df	RSS	k	C _I
Add: x6	109	43441.6	<u>3</u>	0.793
Add: x4	109	48569.8	3	13.399
Add: x5	109	49618.9	3	15.978
Add: x2	109	49772.8	3	16.356

I_{min}

constant, x3
x6 is added

see 204
@5014

3) The above output is for backward elimination and forward selection. The full model has a constant x_1 and predictors $x_2 = \text{height while sitting}$, $x_3 = \text{height while kneeling}$, $x_4 = \text{head length}$, $x_5 = \text{nasal breadth}$, and $x_6 = \text{arm span}$. Hence the full model has 6 terms.

a) What is the model I_{min} that has the smallest $C_p(I)$?

constant, x3, x6

Note $k=3$

b) Suppose the backward elimination or forward selection model with $C_p(I) = 14.398$ had $C_p(I) = 1.440$. Would this model then be I_I ? Explain briefly.

yes since $k=2 < 3$ and

$$C_p(I) = 1.44 < C_p(I_{min}) + 1 = 1.793$$