

Current terms: (EDUC JANT log[NONW] log[NOX] OVR65 PREC)

| | df | RSS | | k | C_I |
|-------------------|----|---------|--|---|--------|
| Delete: OVR65 | 54 | 59897.9 | | 6 | 6.284 |
| Delete: EDUC | 54 | 66809.3 | | 6 | 12.547 |
| Delete: log[NONW] | 54 | 73178.1 | | 6 | 18.319 |
| Delete: JANT | 54 | 76417.1 | | 6 | 21.255 |
| Delete: PREC | 54 | 83958.1 | | 6 | 28.089 |
| Delete: log[NOX] | 54 | 86823.1 | | 6 | 30.685 |

1) From the above output, which model should be used?

6 terms with constant

(Constant, EDUC, JANT, log(NONW), log(NOX), PREC)

20

Base terms: (x3)

| | df | RSS | | k | C_I |
|---------|-----|---------|--|---|--------|
| Add: x6 | 109 | 43441.6 | | 3 | 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 |

e 2) The above output is for 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. What is the model I_{min} that has the smallest $C_p(I)$?

Constant, x3, x6
 ||
 x1

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3) Suppose $\mathbf{Y}^* \sim N_n(\mathbf{X}\hat{\boldsymbol{\beta}}, \sigma_n^2 \mathbf{I}_n)$. Hence $Y_i^* = \mathbf{x}_i^T \hat{\boldsymbol{\beta}} + \epsilon_i^P$ where $E(\epsilon_i^P) = 0$ and $V(\epsilon_i^P) = \sigma_n^2$. Hence $\mathbf{A}\mathbf{Y}^* \sim N_g(\mathbf{A}\mathbf{X}\hat{\boldsymbol{\beta}}, \sigma_n^2 \mathbf{A}\mathbf{A}^T)$ if \mathbf{A} is a $g \times n$ constant matrix. Recall that \mathbf{X} is an $n \times p$ constant matrix. Simplify quantities when possible.

20 a) What is the distribution of $\hat{\boldsymbol{\beta}}^* = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{Y}^*$?

$$N_p \left((\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{X} \hat{\boldsymbol{\beta}}, \sigma_n^2 (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{X} (\mathbf{X}^T \mathbf{X})^{-1} \right)$$

$$\sim N_p \left[\hat{\boldsymbol{\beta}}, \sigma_n^2 (\mathbf{X}^T \mathbf{X})^{-1} \right]$$

20 b) Using a), what is $E(\hat{\boldsymbol{\beta}}^*)$?

$$\boxed{\hat{\boldsymbol{\beta}}}$$

→ 10 c) Recall that $\mathbf{X}\hat{\boldsymbol{\beta}} = \mathbf{P}\mathbf{Y}$. What is the distribution of $\hat{\boldsymbol{\beta}}_I^* = (\mathbf{X}_I^T \mathbf{X}_I)^{-1} \mathbf{X}_I^T \mathbf{Y}^*$?

$$N_K \left[\underbrace{(\mathbf{X}_I^T \mathbf{X}_I)^{-1} \mathbf{X}_I^T \mathbf{P}\mathbf{Y}}_{\hat{\boldsymbol{\beta}}_I}, \sigma_n^2 (\mathbf{X}_I^T \mathbf{X}_I)^{-1} \mathbf{X}_I^T \mathbf{X}_I (\mathbf{X}_I^T \mathbf{X}_I)^{-1} \right]$$

$$\sim N_K \left[\hat{\boldsymbol{\beta}}_I, \sigma_n^2 (\mathbf{X}_I^T \mathbf{X}_I)^{-1} \right]$$