

Formulas

Stat 151A, Fall 2018

October 6, 2018

1.

$$\frac{RSS}{\sigma^2} \sim \chi_{n-p-1}^2.$$

2. If $\hat{\beta}$ is the ols estimator then (under assumptions that you must know)

$$\frac{\hat{\beta}_j - \beta_j}{s.e(\hat{\beta}_j)} \sim t_{n-p-1}$$

3. For testing a reduced model m against the full model M :

$$T := \frac{(RSS(m) - RSS(M))/(p - q)}{RSS(M)/(n - p - 1)} \sim F_{p-q, n-p-1}.$$

4. In one way ANOVA, if $\mu_1 = \dots = \mu_t$, then

$$T = \frac{\sum_{i=1}^t n_i (\bar{y}_i - \bar{y})^2 / (t - 1)}{\sum_{i=1}^t \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_i)^2 / (n - t)}$$

has F -distribution with $t - 1$ and $n - t$ degrees of freedom.

5.

$$\text{cov}(AZ) = A \text{cov}(Z) A^T$$

6.

$$\begin{aligned} \nabla_x (b^T x) &= b \\ \nabla_x \left(\frac{1}{2} x^T A x \right) &= A x. \end{aligned}$$

7.

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

and

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x},$$

8.

$$x_0^T \hat{\beta} \pm t_{n-p-1}^{(\alpha/2)} \hat{\sigma} \sqrt{1 + x_0^T (X^T X)^{-1} x_0}.$$