## Formulas

## Stat 151A, Fall 2017

## December 6, 2017

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:

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

q is the number of variables in the reduced model without counting the intercept.

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

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

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

5. Let  $L \in \mathbb{R}^{q \times (p+1)}$  be a full rank matrix with rank  $q \leq p+1$ . If  $L\beta = c$ , for  $c \in \mathbb{R}^q$ , then

$$\frac{(L\hat{\beta} - c)^T [L(X^T X)^{-1} L^T]^{-1} (L\hat{\beta} - c)}{q \,\hat{\sigma}^2} \sim F_{q, n-p-1}.$$

6.

$$RSS_{[i]} = RSS - \frac{\hat{e}_i^2}{1 - h_i}.$$

7. Suppose Z is a random vector with mean  $\mu$  and covariance matrix  $\Sigma$ . Then

$$\mathbb{E}(Z^T A Z) = tr(A\Sigma) + \mu^T A \mu. \tag{1}$$

8. 
$$\hat{\beta}_{[i]} = \hat{\beta} - \frac{\hat{e}_i}{1 - h_i} (X^T X)^{-1} x_i. \tag{2}$$

9. 
$$\hat{e}_{[i]} = \frac{\hat{e}_i}{1 - h_i}. \tag{3}$$

10. 
$$t_i = \frac{\hat{e}_{[i]}\sqrt{1 - h_i}}{\sqrt{RSS_{[i]}/(n - p - 2)}} = \frac{\hat{e}_i}{\sqrt{RSS_{[i]}/(n - p - 2)}\sqrt{1 - h_i}}.$$

11. 
$$r_i = \frac{\hat{e}_i}{\sqrt{RSS/(n-p-1)}\sqrt{1-h_i}}.$$

12. 
$$C_i = r_i^2 \frac{h_i}{(1 - h_i)(p+1)} = \frac{(\hat{\beta} - \hat{\beta}_{[i]})^T X^T X (\hat{\beta} - \hat{\beta}_{[i]})}{(p+1)\hat{\sigma}^2}.$$

13. 
$$cov(AZ) = A cov(Z) A^{T}$$

14.  $f(x; \theta_i, \phi_i) := h(x, \phi_i) \exp\left(\frac{x\theta_i - b(\theta_i)}{a(\phi_i)}\right).$ 

15. 
$$var(y_i) = b''(\theta_i)a(\phi_i).$$

16. 
$$\beta^{(m+1)} = \beta^{(m)} - (H\ell(\beta^{(m)}))^{-1} \nabla \ell(\beta^{(m)})$$

17. Logistic

$$\nabla \ell(\beta) = \sum_{i=1}^{n} (y_i - p_i)(1, x_{i1}, \dots, x_{ip})^T H \ell(\beta) = -\sum_{i=1}^{n} p_i (1 - p_i)(1, x_{i1}, \dots, x_{ip})^T (1, x_{i1}, \dots, x_{ip}).$$

18.  $\beta^{(m+1)} = (X^T W X)^{-1} X^T W Z$  $Z = X \beta^{(m)} + W^{-1} (Y - p).$ 

19. 
$$RSS(j,c) = n_1\bar{p}_1(1-\bar{p}_1) + n_2\bar{p}_2(1-\bar{p}_2),$$
 Cross–entropy or Deviance : 
$$-2n_1(\bar{p}_1\log\bar{p}_1 + (1-\bar{p}_1)\log(1-\bar{p}_1))$$
 Misclassification Error : 
$$n_1\min(\bar{p}_1,1-\bar{p}_1).$$

20. 
$$\nabla_x(b^T x) = b \\
\nabla_x(\frac{1}{2}x^T A x) = A x.$$