Quiz 1

March 18th 2020

1 Lecture 5

$$\log \frac{Pr(G = 1|X = x)}{1 - Pr(G = 1|X = x)} = \beta_0 + x^T \beta$$

$$\frac{Pr(G = 1|X = x)}{1 - Pr(G = 1|X = x)} = \exp(\beta_0 + x^T \beta)$$

$$Pr(G = 1|X = x) = \frac{\exp(\beta_0 + x^T \beta)}{1 + \exp(\beta_0 + x^T \beta)}$$

$$Pr(G = 2|X = x) = 1 - Pr(G = 1|X = x) = \frac{1}{1 + \exp(\beta_0 + x^T \beta)}$$

2 Lecture 6

$$\begin{split} \hat{\Sigma}* &= \frac{\sum_{k=1}^{K} \sum_{g_i=k} (x_i^* - \hat{\mu}_k^*) (x_i^* - \hat{\mu}_k^*)^T}{N - K} \\ &= \frac{\sum_{k=1}^{K} \sum_{g_i=k} (\hat{\Sigma}^{-\frac{1}{2}} x_i - \hat{\Sigma}^{-\frac{1}{2}} \hat{\mu}_k) (\hat{\Sigma}^{-\frac{1}{2}} x_i - \hat{\Sigma}^{-\frac{1}{2}} \hat{\mu}_k)^T}{N - K} \\ &= \frac{\sum_{k=1}^{K} \sum_{g_i=k} \hat{\Sigma}^{-\frac{1}{2}} (x_i - \hat{\mu}_k) (x_i - \hat{\mu}_k)^T \hat{\Sigma}^{-\frac{1}{2}}}{N - K} \\ &= \hat{\Sigma}^{-\frac{1}{2}} \frac{\sum_{k=1}^{K} \sum_{g_i=k} (\hat{x}_i - \hat{\mu}_k) (x_i - \hat{\mu}_k)^T}{N - K} \hat{\Sigma}^{-\frac{1}{2}} \\ &= \hat{\Sigma}^{-\frac{1}{2}} \hat{\Sigma} \hat{\Sigma}^{-\frac{1}{2}} \\ &= I \end{split}$$