

exercise 1

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$$N = 30 \quad r = 3 \quad q = 5 \quad m = 2$$

a) Model V2:

$$H_0: r_{ij} = 0 \quad \forall i, j \quad H_1: r_{ij} \neq 0 \text{ for at least one } (i, j)$$

$$\text{Test statistic} \quad R = \frac{2}{8} \cdot \frac{\hat{D}_r - \hat{D}_{1,2}}{\hat{D}_{1,2}} \sim F_{2, 14, 15}$$

we accept H_0 if $R \leq f$ as f is the $(1-\alpha)$ quantile of the $F_{6, 15}$ distribution.

b) filter data for one seed and make 3 one factor anova tests of the form: (with β_i are the factors due to fertilizers

$$H_0: \beta_1 = \dots = \beta_5 = 0 \quad H_1: \beta_i \neq 0 \text{ for at least one } i \in \{1, \dots, 5\}$$

$$R = \frac{5}{4} \cdot \frac{\hat{D}_\beta}{\hat{D}_\varepsilon}$$

accept H_0 if $R \leq f$ where f is the $(1-\alpha)$ -quantile of the $F_{4, 5}$ distribution

c) make the 2 factor anova test where α_i are the factors due to seeds

$$H_0: \alpha_1 = \alpha_2 = \alpha_3 = 0 \quad H_1: \alpha_i \neq 0 \text{ for at least one } i \in \{1, 2, 3\}$$

$$R = \frac{23}{2} \cdot \frac{\hat{D}_\alpha}{\hat{D}_r}$$

accept H_0 if $R \leq f$ where f is the $(1-\alpha)$ Quantil of the $F_{2, 23}$ distrib.

exercise 4

$$Y_1, \dots, Y_N \text{ ind. } \mathcal{L}(Y_i) = \text{Pois}(\lambda_i) \quad g = \log$$

$$\mathcal{L}(b | Y_1, \dots, Y_N) \stackrel{\substack{\uparrow \\ \text{def of} \\ \text{Pois distrib.}}}{=} \sum_{i=1}^N \log \left(e^{-\lambda_i} \frac{\lambda_i^{Y_i}}{Y_i!} \right) = \sum_{i=1}^N -\lambda_i + Y_i \log(\lambda_i) - \underbrace{\log(Y_i!)}_{\substack{\text{no dependence} \\ \text{on } \lambda_i \Rightarrow \text{can be} \\ \text{omitted for maximizing} \\ \mathcal{L}}}$$

for the saturated model compute b_{\max} :

$$\hat{b}_{\max} = \underset{b}{\operatorname{argmax}} \underbrace{\sum_{i=1}^N (-b_i + Y_i \cdot \log(b_i))}_{=: G(b_i)}$$

$$G' = -1 + Y_i \cdot \frac{1}{b_i} \stackrel{!}{=} 0 \Leftrightarrow b_i = Y_i \Rightarrow \hat{b}_{\max} = \begin{pmatrix} Y_1 \\ \vdots \\ Y_N \end{pmatrix}$$

$$\begin{aligned} \Rightarrow \mathbb{D} &= 2 \cdot (\mathcal{L}(\hat{b}_{\max} | Y_1, \dots, Y_N) - \mathcal{L}(\hat{\lambda} | Y_1, \dots, Y_N)) \\ &= 2 \sum_{i=1}^N -Y_i + Y_i \log(Y_i) + \hat{\lambda}_i - Y_i \log(\hat{\lambda}_i) \\ &= 2 \sum_{i=1}^N \hat{\lambda}_i - Y_i + Y_i \log\left(\frac{Y_i}{\hat{\lambda}_i}\right) = 2 \sum_{i=1}^N \hat{\lambda}_i + Y_i \cdot (\log\left(\frac{Y_i}{\hat{\lambda}_i}\right) - 1) \end{aligned}$$