

Assignment 3

Question 1

a) i) $\log(\lambda_{11})$ means $x_1 = 0 \quad x_j = 0$

$$\log(\lambda_{11}) = \gamma_0 + \gamma_1(0) + \gamma_2(0) + \gamma_3(0)(0)$$

$$\log(\lambda_{11}) = \gamma_0$$

ii) $\log(\lambda_{12})$ means $x_1 = 0 \quad x_j = 1$

$$\log(\lambda_{12}) = \gamma_0 + \gamma_1(0) + \gamma_2(1) + \gamma_3(0)(1)$$

$$\log(\lambda_{12}) = \gamma_0 + \gamma_2$$

iii) $\log(\lambda_{21})$ means $x_1 = 1 \quad x_j = 0$

$$\log(\lambda_{21}) = \gamma_0 + \gamma_1(1) + \gamma_2(0) + \gamma_3(1)(0)$$

$$\log(\lambda_{21}) = \gamma_0 + \gamma_1$$

iv) $\log(\lambda_{22})$ means $x_1 = 1 \quad x_j = 1$

$$\log(\lambda_{22}) = \gamma_0 + \gamma_1(1) + \gamma_2(1) + \gamma_3(1)(1)$$

$$\log(\lambda_{22}) = \gamma_0 + \gamma_1 + \gamma_2 + \gamma_3$$

$$b) \quad P(Y_{12} | Y_{11} + Y_{12} = n_i) = \frac{P(Y_{12} | n_i) P(Y_{11} + Y_{12} = n_i)}{P(Y_{11} + Y_{12} = n_i)}$$

$$= \frac{P(Y_{12}) P(Y_{11} = n_i - Y_{12})}{P(Y_{11} + Y_{12} = n_i)} \quad \text{Independence}$$

given that the poisson distribution function (pdf) is $P(Y|\lambda) = \frac{e^{-\lambda} \lambda^Y}{Y!}$

① $P(Y_{12}) P(Y_{11} = n_i - Y_{12})$ substitute the y terms into the poisson pdf.

$$\frac{e^{-\lambda_{12}} \lambda_{12}^{Y_{12}}}{Y_{12}!} \times \frac{e^{-\lambda_{11}} \lambda_{11}^{Y_{11}}}{Y_{11}!} \quad \text{expanding } Y_{11} = n_i - Y_{12} \rightarrow \frac{e^{-\lambda_{12}} \lambda_{12}^{Y_{12}}}{Y_{12}!} \times \frac{e^{-\lambda_{11}} \lambda_{11}^{(n_i - Y_{12})}}{(n_i - Y_{12})!}$$

$$(2) P(Y_{i1} + Y_{i2} = n_i)$$

substitute the y terms into the poisson pdf.

$$\frac{e^{-(\lambda_{i1} + \lambda_{i2})} (\lambda_{i1} + \lambda_{i2})^{(Y_{i1} + Y_{i2})}}{(Y_{i1} + Y_{i2})!}$$

simplify to

$$\frac{e^{-\lambda_{i1} - \lambda_{i2}} \lambda_{i1} + \lambda_{i2}^{n_i}}{n_i!}$$

put them back together =

$$P(Y_{i2} | Y_{i1} + Y_{i2} = n_i) = \frac{\left[\frac{e^{-\lambda_{i2}} \lambda_{i2}^{Y_{i2}}}{Y_{i2}!} \times \frac{e^{-\lambda_{i1}} \lambda_{i1}^{(n_i - Y_{i2})}}{(n_i - Y_{i2})!} \right]}{\left[\frac{e^{-\lambda_{i1} - \lambda_{i2}} (\lambda_{i1} + \lambda_{i2})^{n_i}}{n_i!} \right]} \quad (2)$$

$$= \left[\frac{e^{-\lambda_{i2}} \lambda_{i2}^{Y_{i2}}}{Y_{i2}!} \times \frac{e^{-\lambda_{i1}} \lambda_{i1}^{(n_i - Y_{i2})}}{(n_i - Y_{i2})!} \right] \times \left[\frac{n_i!}{e^{-\lambda_{i1} - \lambda_{i2}} (\lambda_{i1} + \lambda_{i2})^{n_i}} \right]$$

$$= \left[\frac{e^{-\lambda_{i2}} \lambda_{i2}^{Y_{i2}} \times e^{-\lambda_{i1}} \lambda_{i1}^{(n_i - Y_{i2})}}{Y_{i2}! (n_i - Y_{i2})!} \right] \times \left[\frac{n_i!}{e^{-\lambda_{i1} - \lambda_{i2}} (\lambda_{i1} + \lambda_{i2})^{n_i}} \right]$$

$$= \left[\frac{e^{-\lambda_{i2} - \lambda_{i1}} \lambda_{i2}^{Y_{i2}} \lambda_{i1}^{(n_i - Y_{i2})}}{Y_{i2}! (n_i - Y_{i2})!} \right] \times \left[\frac{n_i!}{e^{-\lambda_{i1} - \lambda_{i2}} (\lambda_{i1} + \lambda_{i2})^{n_i}} \right]$$

$$= \frac{\lambda_{i2}^{Y_{i2}} \lambda_{i1}^{(n_i - Y_{i2})} n_i!}{Y_{i2}! (n_i - Y_{i2})! (\lambda_{i1} + \lambda_{i2})^{n_i}}$$

set aside the combination terms \rightarrow $\left[\frac{n_i!}{Y_{i2}! (n_i - Y_{i2})!} \right] \times \frac{\lambda_{i2}^{Y_{i2}} \lambda_{i1}^{(n_i - Y_{i2})}}{(\lambda_{i1} + \lambda_{i2})^{n_i}} \rightarrow$ introduce a 0 $-Y_{i2} + Y_{i2}$

$$= \binom{n_i}{Y_{i2}} \times \frac{\lambda_{i2}^{Y_{i2}} \lambda_{i1}^{(n_i - Y_{i2})}}{(\lambda_{i1} + \lambda_{i2})^{n_i - Y_{i2} + Y_{i2}}}$$

$$= \binom{n_i}{Y_{i2}} \frac{\lambda_{i2}^{Y_{i2}} \lambda_{i1}^{(n_i - Y_{i2})}}{(\lambda_{i1} + \lambda_{i2})^{Y_{i2}} (\lambda_{i1} + \lambda_{i2})^{n_i - Y_{i2}}}$$

$$= \binom{n_i}{Y_{i2}} \frac{\lambda_{i2}^{Y_{i2}}}{(\lambda_{i1} + \lambda_{i2})^{Y_{i2}}} \times \frac{\lambda_{i1}^{n_i - Y_{i2}}}{(\lambda_{i1} + \lambda_{i2})^{n_i - Y_{i2}}}$$

factor out the common exponent.

$$= \binom{n_i}{Y_{i2}} \left(\frac{\lambda_{i2}}{\lambda_{i1} + \lambda_{i2}} \right)^{Y_{i2}} \times \left(\frac{\lambda_{i1}}{\lambda_{i1} + \lambda_{i2}} \right)^{n_i - Y_{i2}}$$

$$\downarrow$$

$$\binom{n_i}{y_{i2}} \left(\frac{x_{i2}}{x_{i1} + x_{i2}} \right)^{y_{i2}} \left(\frac{x_{i1}}{x_{i1} + x_{i2}} \right)^{n - y_{i2}}$$

equivalent to the Binomial distribution formula

$$= \binom{n}{x} p^x q^{n-x} \sim B(n, p)$$

$$Y \sim B\left(n, \frac{x_{i2}}{x_{i1} + x_{i2}}\right) \#$$

c) $\log\left(\frac{\pi_i}{1 - \pi_i}\right) = \beta_0 + \beta_1 x_i$ given that from b) $\pi_i = \frac{x_{i2}}{x_{i1} + x_{i2}}$

$$\log\left[\frac{\left(\frac{x_{i2}}{x_{i1} + x_{i2}}\right)}{\left(1 - \frac{x_{i2}}{x_{i1} + x_{i2}}\right)}\right] = \beta_0 + \beta_1 x_i$$

$$\log\left[\frac{\left(\frac{x_{i2}}{x_{i1} + x_{i2}}\right)}{\frac{(x_{i1} + x_{i2}) - x_{i2}}{(x_{i1} + x_{i2})}}\right] = \log\left[\frac{x_{i2}}{x_{i1} + x_{i2}} \times \frac{(x_{i1} + x_{i2})}{x_{i1} + x_{i2} - x_{i2}}\right]$$

$$= \log\left(\frac{x_{i2}}{x_{i1}}\right)$$

$$= \log x_{i2} - \log x_{i1}$$

$$\log x_{i2} - \log x_{i1} = \beta_0 + \beta_1 x_i$$

If $i=1$, both x_{12} and x_{11} refers to treatment A \hookrightarrow This implies that $x_i = 0$

$$\beta_0 + \beta_1 x_i = \log x_{i2} - \log x_{i1} \hookrightarrow \text{from the result in a)}$$

$$\beta_0 + \beta_1(0) = \gamma_0 + \gamma_2 - \gamma_0$$

$$\beta_0 = \gamma_2 \#$$

If $i=2$, both x_{22} and x_{21} refers to treatment B \hookrightarrow This implies that $x_i = 1$

$$\beta_0 + \beta_1 x_i = \log x_{i2} - \log x_{i1} \hookrightarrow \text{from the result in b)}$$

$$\beta_0 + \beta_1(1) = \gamma_0 + \gamma_1 + \gamma_2 + \gamma_3 - (\gamma_0 + \gamma_1)$$

$$\beta_0 + \beta_1 = \gamma_2 + \gamma_3 \quad \text{substitute } \beta_0 = \gamma_2$$

$$\gamma_2 + \beta_1 = \gamma_2 + \gamma_3$$

$$\beta_1 = \gamma_2 - \gamma_2 + \gamma_3$$

$$\beta_1 = \gamma_3 \neq$$

- d) The idea behind testing treatment has no effect on the probability of being a case means that the predictor term x_i for treatment has 0 coefficient.

in the case of

$$\log\left(\frac{\pi_i}{1-\pi_i}\right) = \beta_0 + \beta_1 x_i \quad \checkmark \beta_1 = 0$$

$$\log\left(\frac{\pi_i}{1-\pi_i}\right) = \beta_0$$

from (c) we know that $\beta_1 = \gamma_3$, this means $\beta_1 = 0 = \gamma_3$

substituting $\gamma_3 = 0$ into the poisson model

$$\log(x_{ij}) = \gamma_0 + \gamma_1 x_i + \gamma_2 x_j + \gamma_3 x_i x_j$$

$$\log(x_{ij}) = \gamma_0 + \gamma_1 x_i + \gamma_2 x_j + \underbrace{(0) x_i x_j}$$

$$\log(x_{ij}) = \gamma_0 + \gamma_1 x_i + \gamma_2 x_j$$

as you can see the interaction terms in the poisson model disappear.

Hence, showing that testing that treatment has no effect on the probability of being a case, is equivalent for test for no interaction in the poisson model.