

# Homework 3

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## 4.6

**a**

$\log \mu_A = \alpha$  and  $\log \mu_B = \alpha + \beta$ , so  $\exp(\beta) = \frac{\mu_B}{\mu_A}$ . Let  $\hat{\alpha}$  and  $\hat{\beta}$  be the estimate, then the interpretation can be that  $\hat{\mu}_A = \exp(\hat{\alpha})$  and  $\hat{\mu}_B = \exp(\hat{\alpha} + \hat{\beta})$

**b**

Let  $x_i$  and  $y_i$  be the samples from  $A$  and  $B$ , we have

$$L(\alpha, \beta) = \sum_{i=1}^{10} (x_i \alpha - e^\alpha - \log x_i!) + \sum_{i=1}^{10} (y_i (\alpha + \beta) - e^{\alpha+\beta} - \log y_i!)$$

The MLE of  $\alpha$  and  $\beta$  are  $\hat{\alpha} = \log \bar{x} = \log 5$  and  $\hat{\beta} = \log \bar{y} - \log \bar{x} = \log 9 - \log 5$ .

$$\mathbb{E}\left(-\frac{\partial^2 L}{\partial \alpha^2}\right) = 10e^\alpha + 10e^{\alpha+\beta} = 10 \times 5 + 10 \times 9 = 140;$$

$$\mathbb{E}\left(-\frac{\partial^2 L}{\partial \alpha \partial \beta}\right) = \mathbb{E}\left((\sum x_i + \sum y_i - 10e^\alpha - 10e^{\alpha+\beta})(\sum y_i - 10e^{\alpha+\beta})\right) = \mathbb{E}\left(\sum x_i \sum y_i + (\sum y_i)^2 - 10e^\alpha \sum y_i - 10e^{\alpha+\beta} \sum y_i\right) = 50 \times 90 + 90^2 + 90 - 10 \times 5 \times 90 - 10 \times 9 \times 90 = 90;$$

$$\mathbb{E}\left(-\frac{\partial^2 L}{\partial \beta^2}\right) = 10e^{\alpha+\beta} = 90$$

$$\text{Information matrix } \mathcal{J} = \begin{pmatrix} \mathbb{E}\left(-\frac{\partial^2 L}{\partial \alpha^2}\right) & \mathbb{E}\left(-\frac{\partial^2 L}{\partial \alpha \partial \beta}\right) \\ \mathbb{E}\left(-\frac{\partial^2 L}{\partial \alpha \partial \beta}\right) & \mathbb{E}\left(-\frac{\partial^2 L}{\partial \beta^2}\right) \end{pmatrix} = \begin{pmatrix} 140 & 90 \\ 90 & 90 \end{pmatrix}; \mathcal{J}^{-1} = \begin{pmatrix} \frac{1}{50} & -\frac{1}{50} \\ -\frac{1}{50} & \frac{7}{225} \end{pmatrix}, \text{ so}$$

$$\widehat{\text{SE}}(\hat{\beta}) = \sqrt{\frac{7}{225}}$$

Wald statistic:  $z_w = \frac{\hat{\beta}}{\widehat{\text{SE}}(\hat{\beta})} = \frac{\log 9 - \log 5}{\sqrt{\frac{7}{225}}} \approx 18.89$ . p-value  $< 0.001$ . So we reject the null hypothesis that  $\mu_A = \mu_B$  and  $\beta = 0$ . That is there is a significant difference between the two treatments.

**c**

Using  $\hat{\beta}$  and  $\widehat{\text{SE}}(\hat{\beta})$ , we can construct a 95% confidence interval for  $\beta$ :  $\hat{\beta} \pm 1.96\widehat{\text{SE}}(\hat{\beta}) = (\log 9 - \log 5) \pm 1.96\sqrt{\frac{7}{225}} = (0.24, 0.93)$ . So the 95% confidence interval for  $\frac{\mu_B}{\mu_A} = \exp(\beta)$  is  $(e^{0.24}, e^{0.93}) = (1.27, 2.54)$

**d**

We know  $\sum x_i \sim \text{Poi}(10\mu_A)$  and  $\sum y_i \sim \text{Poi}(10\mu_B)$ ,  $\sum x_i$  and  $\sum y_i$  are independent, so  $\sum x_i | \sum x_i + \sum y_i \sim \text{Bin}(\sum x_i + \sum y_i, \frac{\mu_A}{\mu_A + \mu_B})$

Under null hypothesis that  $\mu_A = \mu_B$ , we should see  $\pi = \frac{\mu_A}{\mu_A + \mu_B} = \frac{1}{2}$ .

$\hat{\pi} = \frac{\sum x_i}{\sum x_i + \sum y_i} = \frac{5}{14}$ ,  $\widehat{\text{SE}}(\hat{\pi}) = \sqrt{\frac{\hat{\pi}(1-\hat{\pi})}{\sum x_i + \sum y_i}} = \sqrt{\frac{\frac{5}{14}(1-\frac{5}{14})}{140}} \approx 0.04$ . So the wald statistic is  $\frac{\hat{\pi} - \pi}{\widehat{\text{SE}}(\pi)} = \frac{\frac{5}{14} - \frac{1}{2}}{0.04} \approx -3.53$ . p-value  $< 0.001$ . So we reject the null hypothesis that  $\mu_A = \mu_B$ .

## 4.13

**a**

Total number made is 145 and total attempts is 296. So  $\hat{\alpha} = \frac{145}{296} \approx 0.49$  and  $\widehat{SE}(\hat{\alpha}) = \sqrt{\frac{\hat{\alpha}(1-\hat{\alpha})}{296}} \approx 0.029$ . The interpretation is that the estimated percentage that O'Neal's free throw is made is 49% and the standard error is 2.9%.

Let  $x_i$  be the number made and  $n_i$  be the number of attempts in each game.  $X^2 = \sum_{i=1}^{23} \frac{(x_i - n_i \hat{\alpha})^2}{n_i \hat{\alpha}} + \frac{(n_i - x_i - n_i(1-\hat{\alpha}))^2}{n_i(1-\hat{\alpha})} = 35.5$  with  $df = 22$ . p-value is 0.03. So the model does not appear to fit adequately.

**b**

$\sqrt{\frac{X^2}{22}} \approx 1.27$ , so we adjust SE by a factor of 1.27. The adjusted SE is 0.037. The original confidence interval of  $\alpha$  is  $0.49 \pm 1.96 \times 0.029 = (0.40, 0.51)$ . The adjusted confidence interval is  $0.49 \pm 1.96 \times 0.037 = (0.38, 0.53)$ . We can see that the adjusted confidence interval is wider, reflecting overdispersion.

## 4.29

**a**

Since the PDF is symmetric around 0, we have  $\Phi(0) = 0.5 \Rightarrow \alpha + \beta x = 0 \Rightarrow x = -\alpha/\beta$

**b**

$$\frac{d\pi(x)}{dx} = \frac{d}{dx} \Phi(\alpha + \beta x) = \phi(\alpha + \beta x)\beta, \text{ when } \pi(x) = 0.5, x = -\alpha/\beta, \frac{d\pi(x)}{dx} = \beta\phi(0)$$

For logit link,  $\phi(x) = \frac{e^{-x}}{(1+e^{-x})^2}$ ,  $\beta\phi(0) = 0.25\beta$ ; For probit link,  $\phi(x) = \frac{1}{\sqrt{2\pi}}e^{-\frac{x^2}{2}}$ ,  $\beta\phi(0) = \frac{\beta}{\sqrt{2\pi}}$

**c**

$\pi(x)$  is from the class of normal CDFs, suppose its mean and variance is  $\mu$  and  $\sigma^2$ , we have  $\pi(\mu) = 0.5 \Rightarrow \mu = -\alpha/\beta$

$$\frac{d}{dx} \pi(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \text{ when } x = \mu, \frac{d}{dx} \pi(x) = \frac{1}{\sqrt{2\pi\sigma^2}} = \frac{\beta}{\sqrt{2\pi}} \Rightarrow \sigma = \frac{1}{|\beta|}$$