

20/6 D1

$$2a) \phi(T) = \begin{cases} 1 & T < C_1(u) \text{ or } T < C_2(u) \\ g_i & T = C_i(u) \quad i=1,2 \\ 0 & \text{else} \end{cases} \quad \text{st } E_{\alpha=0}[\phi(U)] = \alpha$$

$$\& E_{\alpha=0}[U \phi(T|U)] = \alpha E(T|U)$$

where $T = \sum_{i=1}^n x_i y_i$ $U = \sum_{i=1}^n y_i$

$$b) E(T|U) = E\left[\sum_{i=1}^n x_i y_i \mid \sum_{i=1}^n y_i = u\right]$$

$$= \sum_{i=1}^n x_i E(y_i \mid \sum_{j=1}^n y_j = u)$$

$$= \sum_{i=1}^n x_i \Pr(y_i = 1 \mid \sum_{j=1}^n y_j = u)$$

$$\Pr(y_i = 1 \mid \sum_{j=1}^n y_j = u) = \frac{\Pr(y_i = 1, \sum_{j=1}^n y_j = u)}{\Pr(\sum_{j=1}^n y_j = u)}$$

$$\sum_{j=1}^n y_j \sim \text{Bin}(n, \pi_0)$$

$$\Pr(y_i = 1, \sum_{j=1}^n y_j = u) = \Pr(y_i = 1, \sum_{j \neq i} y_j = u-1)$$

$$\sum_{j \neq i} y_j \sim \text{Bin}(n-1, \pi_0)$$

$$= \Pr(y_i = 1) \cdot \Pr(\sum_{j \neq i} y_j = u-1) \quad \text{by ind.}$$

$$= \pi_0 \cdot \binom{n-1}{u-1} \pi_0^{u-1} (1-\pi_0)^{n-u}$$

$$= \binom{n-1}{u-1} \pi_0^u (1-\pi_0)^{n-u}$$

$$\Rightarrow \Pr(y_i = 1 \mid \sum_{j=1}^n y_j = u) = \frac{\binom{n-1}{u-1} \pi_0^u (1-\pi_0)^{n-u}}{\binom{n}{u} \pi_0^u (1-\pi_0)^{n-u}} = \frac{u}{n}$$

$$\Rightarrow E(T|U) = \frac{u}{n} \sum_{i=1}^n x_i$$

under null

$$\pi_0 = \Pr(y_i = 1) \forall i$$

$$= \frac{\exp\{\beta_0\}}{1 + \exp\{\beta_0\}}$$

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b) cont.

$$\begin{aligned} \text{Var}(T|U) &= \text{Var}\left(\sum_{i=1}^n X_i Y_i \mid \sum_{i=1}^n Y_i = u\right) \\ &= \sum_{i=1}^n \left[X_i^2 \text{Var}(Y_i \mid \sum_{k=1}^n Y_k = u) \right] + 2 \sum_{1 \leq i < j \leq n} X_i X_j \text{Cov}(Y_i, Y_j \mid \sum_{k=1}^n Y_k = u) \\ \text{Var}(Y_i \mid \sum_{k=1}^n Y_k = u) &= E(Y_i^2 \mid \sum_{k=1}^n Y_k = u) - \left[E(Y_i \mid \sum_{k=1}^n Y_k = u) \right]^2 \\ &= \frac{u}{n} \left(1 - \frac{u}{n} \right) \end{aligned}$$

$$\text{Cov}(Y_i, Y_j \mid \sum_{k=1}^n Y_k = u) = E(Y_i Y_j \mid \sum_{k=1}^n Y_k = u) - E(Y_i \mid \sum_{k=1}^n Y_k = u) \cdot E(Y_j \mid \sum_{k=1}^n Y_k = u)$$

$$\begin{aligned} E(Y_i Y_j \mid \sum_{k=1}^n Y_k = u) &= \frac{\Pr(Y_i = 1, Y_j = 1 \mid \sum_{k=1}^n Y_k = u)}{\Pr(\sum_{k=1}^n Y_k = u)} \\ &= \frac{\Pr(Y_i = 1, Y_j = 1, \sum_{k=1}^n Y_k = u)}{\Pr(\sum_{k=1}^n Y_k = u)} \\ &= \frac{\pi_0^2 \binom{n-2}{u-2} \pi_0^{u-2} (1-\pi)^{n-u}}{\pi_0^u \binom{n}{u} \pi^u (1-\pi)^{n-u}} \cdot \dots \\ &= \frac{u}{n} \cdot \frac{u-1}{n-1} \end{aligned}$$

$$\begin{aligned} \text{Cov}(\dots) &= \frac{u}{n} \cdot \frac{u-1}{n-1} - \left(\frac{u}{n} \right)^2 \\ \text{Var}(T|U) &= \sum_{i=1}^n X_i^2 \frac{u}{n} \left[1 - \frac{u}{n} \right] + 2 \sum_{1 \leq i < j \leq n} X_i X_j \left[\frac{u(u-1)}{n(n-1)} - \left(\frac{u}{n} \right)^2 \right] \end{aligned}$$

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$$2c) \left(\begin{array}{c} \frac{d}{d\beta_0} l \\ \frac{d}{d\beta_1} l \end{array} \right) \bigg|_{\beta_1=0, \tilde{\beta}_0} = \begin{pmatrix} \sum y_i - n \tilde{\pi}_0 \\ \sum x_i (y_i - \tilde{\pi}_0) \end{pmatrix} \xrightarrow{\beta_0 \rightarrow 0}$$

$$\tilde{\pi}_0 = \frac{e^{\tilde{\beta}_0}}{1 + e^{\tilde{\beta}_0}}$$

where $\tilde{\beta}_0$ is mle under null

$$\left(\begin{array}{cc} \frac{d^2}{d\beta_0^2} l & \frac{d^2}{d\beta_0 d\beta_1} l \\ \frac{d}{d\beta_0 d\beta_1} l & \frac{d^2}{d\beta_1^2} l \end{array} \right) \bigg|_{\beta_1=0, \tilde{\beta}_0} = \begin{pmatrix} n \tilde{\pi}_0 (1 - \tilde{\pi}_0) & \tilde{\pi}_0 (1 - \tilde{\pi}_0) \sum x_i \\ \tilde{\pi}_0 (1 - \tilde{\pi}_0) \sum x_i & \tilde{\pi}_0 (1 - \tilde{\pi}_0) \sum x_i^2 \end{pmatrix}$$

$$\tilde{\beta}_0 = \log\left(\frac{\bar{y}}{1 - \bar{y}}\right)$$

$$\Rightarrow \tilde{\pi}_0 = \frac{\frac{\bar{y}}{1 - \bar{y}}}{1 + \frac{\bar{y}}{1 - \bar{y}}} = \frac{\frac{\bar{y}}{1 - \bar{y}}}{\left(\frac{1 - \bar{y} + \bar{y}}{1 - \bar{y}}\right)} = \bar{y}$$

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d) $Y_1, \dots, Y_n \stackrel{i.i.d.}{\sim} \text{Bern}(\pi_i)$

$$\pi_i = \frac{\exp\{x_i' \beta\}}{1 + \exp\{x_i' \beta\}}$$

$$\beta \in \mathbb{R}^p$$

Test

$$H_0: \lambda' \beta = 0 \quad \lambda = (\lambda_1, \dots, \lambda_p)$$

$$H_1: \lambda' \beta \neq 0$$

$$f_X(x | \beta) = \prod_{i=1}^n \frac{\exp\{y_i x_i' \beta\}}{1 + \exp\{x_i' \beta\}}$$

$$= \frac{\exp\{\sum_{i=1}^n y_i x_i' \beta\}}{\prod_{i=1}^n (1 + \exp\{x_i' \beta\})}$$

$$= \exp\left\{ \beta_1 \sum_{i=1}^n x_{i1} y_i + \dots + \beta_p \sum_{i=1}^n x_{ip} y_i - \sum_{i=1}^n \ln(1 + \exp\{x_i' \beta\}) \right\}$$

$$= \exp\left\{ \beta_1 \sum_{i=1}^n x_{i1} y_i + \dots + \beta_p \sum_{i=1}^n x_{ip} y_i + (\lambda_1 \beta_1 + \dots + \lambda_p \beta_p) \left(\sum_{i=1}^n x_{i1} y_i + \dots + \sum_{i=1}^n x_{ip} y_i \right) - \sum_{i=1}^n \ln(1 + \exp\{x_i' \beta\}) \right\} \quad c(\beta) = \sum_{i=1}^n \ln(\dots)$$

$$T(x) = \sum_{i=1}^n x_{i1} y_i + \dots + \sum_{i=1}^n x_{ip} y_i$$

$$= \exp\left\{ (\lambda' \beta) T(x) + \beta_1 \left(\sum_{i=1}^n x_{i1} y_i - \lambda_1 T(x) \right) + \dots + \beta_p \left(\sum_{i=1}^n x_{ip} y_i - \lambda_p T(x) \right) - c(\beta) \right\}$$