

$$\text{Ex 2g)} \quad X_i = \rho X_{i-1} + \varepsilon_i, \quad \varepsilon_i \perp X_{i-1}, \quad \varepsilon_i \sim N(0, \sigma^2)$$

$i \in \mathbb{Z}$

$, |\rho| < 1$

$\{\varepsilon_i, i \in \mathbb{Z}\}$  i.i.d.

---

$\rightarrow$  WE OBSERVE  $X_1, \dots, X_n$

$$f(x_1, \dots, x_n) = f(x_n | x_{n-1}, x_{n-2}, \dots, x_1) \cdot \\ \cdot f(x_{n-1} | x_{n-2}, \dots) \cdot \dots \cdot f(x_2 | x_1) f(x_1)$$

NOTE THAT,

$$\star \quad \mathcal{L}(X_i | X_{i-1}, \dots, X_1) = \mathcal{L}(X_i | X_{i-1}) = N(\rho X_{i-1}, \sigma^2)$$

$$\star \quad X_1 \sim N(0, \frac{\sigma^2}{1-\rho^2}) \quad \leftarrow \text{AS } X_1 = \sum_{j=0}^{\infty} \varepsilon_{n-j} \rho^j$$

$$\text{var}(X_1) = \sum_{j=0}^{\infty} \sigma^2 \rho^{2j} = \frac{\sigma^2}{1-\rho^2}$$

$$f(x_1, \dots, x_n) \approx f(x_n | x_{n-1}) \cdot f(x_{n-1} | x_{n-2}) \cdot \dots \cdot f(x_1)$$

$$\approx \left( \prod_{i=2}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_i - \rho x_{i-1})^2}{2\sigma^2}} \right) \frac{1}{\sqrt{2\pi \frac{\sigma^2}{1-\rho^2}}} e^{-\frac{x_1^2(1-\rho^2)}{2\sigma^2}}$$

$$\rightarrow \ell_n(\rho, \sigma^2) = \log f(x_1, \dots, x_n)$$

$$= -\frac{n}{2} \log \sigma^2 - \frac{1}{2\sigma^2} \sum_{i=2}^n (x_i - \rho x_{i-1})^2 - \frac{x_1^2(1-\rho^2)}{2\sigma^2} \\ + \frac{1}{2} \log (1-\rho^2) + C$$

$$\text{APPROX. SOLUTIONS : } \hat{\rho}_n = \frac{\sum_{i=2}^n X_i X_{i-1}}{\sum_{i=2}^n X_{i-1}^2}, \hat{\sigma}^2 = \frac{1}{n} \sum_{i=2}^n (X_i - \hat{\rho}_n X_{i-1})^2$$

$$\frac{\partial \ell_n(\rho, \sigma^2)}{\partial \rho} = \frac{1}{\sigma^2} \sum_{i=2}^n (X_i - \rho X_{i-1}) X_{i-1} + \frac{X_1^2 \rho}{\sigma^2} - \frac{\rho}{1-\rho^2} = 0$$

$$\frac{\partial \ell_n(\rho, \sigma^2)}{\partial \sigma^2} = -\frac{n}{2\sigma^4} + \frac{\sum_{i=2}^n (X_i - \rho X_{i-1})^2}{2\sigma^4} + \frac{X_1^2 (1-\rho^2)}{2\sigma^4} = 0$$

$$\frac{\partial^2 \ell_n(\rho, \sigma^2)}{\partial \rho \partial \rho} = -\frac{1}{\sigma^2} \sum_{i=2}^n X_{i-1}^2 - \frac{X_1^2}{\sigma^2} - \frac{\partial}{\partial \rho} \left( \frac{\rho}{1-\rho^2} \right)$$

$$\frac{\partial^2 \ell_n(\rho, \sigma^2)}{\partial \rho \partial \sigma^2} = -\frac{1}{\sigma^4} \sum_{i=2}^n (X_i - \rho X_{i-1}) X_{i-1} - \frac{X_1^2 \rho}{\sigma^4}$$

$$\frac{\partial^2 \ell_n(\rho, \sigma^2)}{\partial \sigma^2 \partial \sigma^2} = \frac{n}{2\sigma^4} - \frac{1}{\sigma^6} \sum_{i=2}^n (X_i - \rho X_{i-1})^2 - \frac{X_1^2 (1-\rho^2)}{\sigma^6}$$

$$\bar{I}_{11}(\rho, \sigma^2) = \lim_{n \rightarrow \infty} -\frac{1}{n} \mathbb{E} \left[ -\frac{1}{\sigma^2} \sum_{i=2}^n X_{i-1}^2 - \frac{X_1^2}{n\sigma^2} - \dots \right]$$

$$= \frac{1}{\sigma^2} \mathbb{E} X_i^2 = \frac{\text{var } X_i}{\sigma^2} = \frac{1}{1-\rho^2}$$

$$\bar{I}_{12}(\rho, \sigma^2) = \mathbb{E} [\varepsilon_i X_{i-1}] = \mathbb{E} \varepsilon_i \mathbb{E} X_{i-1} = 0$$

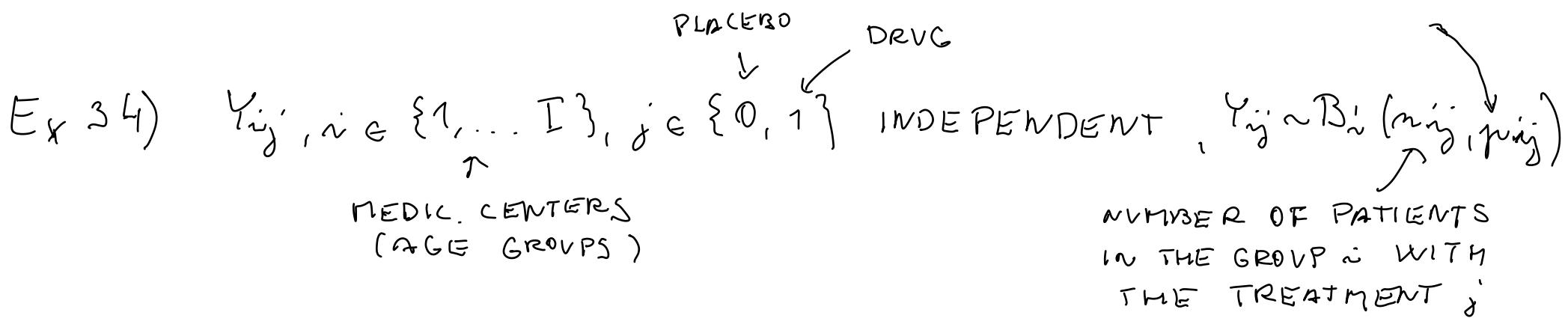
$$\bar{I}_{22}(\rho, \sigma^2) = \frac{1}{2\sigma^4} - \frac{1}{\sigma^6} \mathbb{E} \varepsilon_i^2 = \frac{1}{2\sigma^4} - \frac{\sigma^2}{\sigma^6} = \frac{1}{2\sigma^4}$$

$$\Rightarrow \bar{I}(\rho, \sigma^2) = \begin{pmatrix} \frac{1}{1-\rho^2} & 0 \\ 0 & \frac{1}{2\sigma^4} \end{pmatrix}$$

$$\Rightarrow \sqrt{n} \begin{pmatrix} (\hat{\rho}_n) \\ (\hat{\sigma}^2) \end{pmatrix} - \begin{pmatrix} \rho \\ \sigma^2 \end{pmatrix} \xrightarrow[n \rightarrow \infty]{d} N_2 \left( \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \frac{1-\rho^2}{1-\rho^2} & 0 \\ 0 & \frac{1}{2\sigma^4} \end{pmatrix} \right)$$

$$\sqrt{n} (\hat{\theta}_2 - \theta_X) \xrightarrow[n \rightarrow \infty]{d} N_p (0, \bar{I}^{-1}(\theta_X))$$

PROBS. OF SUCCESS



ASSUMED MODEL:

$$\log\left(\frac{m_{ij}}{1-m_{ij}}\right) = \psi_i + \tau \mathbb{1}_{\{j=1\}}$$

↑  
TREATMENT EFFECT

GOAL: INFERENCE ON  $\tau$

JOINT DENSITY,  $\underline{\psi} = (\psi_1, \dots, \psi_I)^T$

$$\begin{aligned}
 f(\underline{y}; \tau, \underline{\psi}) &= \prod_{i=1}^I \prod_{j=0}^1 \underbrace{\left( \frac{m_{ij}}{1-m_{ij}} \right)^{y_{ij}}}_{\text{CONST}} m_{ij}^{y_{ij}} (1-m_{ij})^{m_{ij}-y_{ij}} \\
 &= \prod_{i=1}^I \prod_{j=0}^1 c_{ij} \exp\left\{ y_{ij} \log m_{ij} + (m_{ij} - y_{ij}) \log(1-m_{ij}) \right\} \\
 &= \exp\left\{ \sum_{i=1}^I \sum_{j=0}^1 \left[ y_{ij} \log\left(\frac{m_{ij}}{1-m_{ij}}\right) + m_{ij} \log(1-m_{ij}) \right] \right\} \cdot c \\
 &= \exp\left\{ \sum_{i=1}^I \sum_{j=0}^1 \left[ y_{ij} (\psi_i + \tau \mathbb{1}_{\{j=1\}}) + m_{ij} \log(1-m_{ij}) \right] \right\} \cdot c \\
 &= \exp\left\{ \sum_{i=1}^I \underbrace{(Y_{i+} + Y_{i-})}_{Y_{i+}} \psi_i + \tau \underbrace{\sum_{i=1}^I Y_{i-}}_{Y_{+1}} + \sum_{i=1}^I \sum_{j=0}^1 m_{ij} \log(1-m_{ij}) \right\} \cdot c
 \end{aligned}$$

$$\rightsquigarrow \underline{S}(\underline{\psi}) = (Y_{i+}, \dots, Y_{I+})^T$$

$$\rightsquigarrow \hat{l}_n^{(c)}(\tau) = \log f(\underline{y}; \tau, \underline{\psi}) - l_{n, \underline{S}(\underline{\psi})}(\tau, \underline{\psi})$$

WE NEED THE DISTRIBUTION OF  $\underline{S}(\underline{\psi}) = (Y_{i+}, \dots, Y_{I+})^T$ .

$$\rightsquigarrow P(Y_{i+} = z) = P(Y_{i0} + Y_{i1} = z) = \sum_{k \in K_i} P(Y_{i0} = k) P(Y_{i1} = z-k)$$

where  $K_i = \{\max(0, z-m_{i0}), \dots, \min(z, m_{i1})\}$

$$\begin{aligned}
 & \text{P}(Y_{i+1} = y) = \sum_{k \in K_i} P(Y_{i+1} = k) P(Y_{i+1} = y - k) \\
 &= \sum_{k \in K_i} \binom{n_{ii}}{k} \left( \frac{p_{ii}}{1-p_{ii}} \right)^k (1-p_{ii})^{n_{ii}-k} \\
 &\quad \left( \frac{n_{io}}{y-k} \right) \left( \frac{p_{io}}{1-p_{io}} \right)^{y-k} (1-p_{io})^{n_{io}-(y-k)} \\
 &= \sum_{k \in K_i} \binom{n_{ii}}{k} \binom{n_{io}}{y-k} e^{(Y_i + \tau)k} e^{Y_i(y-k)} \\
 &\quad \cdot (1-p_{ii})^{n_{ii}} (1-p_{io})^{n_{io}} \\
 &= \sum_{k \in K_i} \binom{n_{ii}}{k} \binom{n_{io}}{y-k} e^{Y_i y} (1-p_{ii})^{n_{ii}} (1-p_{io})^{n_{io}}
 \end{aligned}$$

$$\begin{aligned}
 \rightarrow \ln_{\sim, S(\underline{Y})} (\bar{\tau}, \bar{\psi}) &= \log \left( \prod_{i=1}^I f_i(Y_{i+1}; \bar{\tau}, \bar{\psi}_i) \right) \\
 &= \sum_{i=1}^I \left[ \log \left( A_i(\bar{\tau}, Y_{i+1}) \right) + Y_{i+1} Y_i + n_{ii} \log(1-p_{ii}) \right. \\
 &\quad \left. + n_{io} \log(1-p_{io}) \right]
 \end{aligned}$$

$$\begin{aligned}
 \ln_{\sim} (\bar{\tau}, \bar{\psi}) &= \log f(\underline{Y}; \bar{\tau}, \bar{\psi}) = \\
 &= \sum_{i=1}^I \left[ Y_{i+1} Y_i + \tau Y_{i+1} + n_{ii} \log(1-p_{ii}) \right. \\
 &\quad \left. + n_{io} \log(1-p_{io}) \right]
 \end{aligned}$$

$$\begin{aligned}
 \ln_{\sim}^{(c)} (\tau) &= \ln_{\sim} (\bar{\tau}, \bar{\psi}) - \ln_{\sim, S(\underline{Y})} (\bar{\tau}, \bar{\psi}) \\
 &= \sum_{i=1}^I \left[ \bar{Y}_{i+1} - \log \left( A_i(\bar{\tau}, Y_{i+1}) \right) \right] \\
 &= \tau \boxed{\sum_{i=1}^I \bar{Y}_{i+1}} - \sum_{i=1}^I \log \left( A_i(\bar{\tau}, Y_{i+1}) \right)
 \end{aligned}$$

$$\hat{l}_n^{(c)}(\tau) = \tau Y_{+1} - \sum_{i=1}^T \log(A_i(\tau, Y_{i+}))$$

→ MAXIMUM CONDIT. LIKEL. ESTIM:

$$\hat{\tau}_n = \underset{\tau > 0}{\operatorname{argmax}} \hat{l}_n^{(c)}(\tau)$$

RAD SCORE TEST (based on COND. LIK.  $\hat{l}_n^{(c)}(\tau)$ )

$$H_0 : \tau = \tau_0 \quad H_1 : \tau \neq \tau_0$$

$$R_n^{(c)} = \frac{1}{n} \left[ U_n^{(c)}(\tau_0) \right]^2$$

$$A_i(\tau, Y_i) = \sum_{k \in K_i} \binom{n_{i1}}{k} \binom{n_{i0}}{Y_{i+}-k} e^{\tau k}$$

where

$$U_n^{(c)}(\tau) = \frac{\partial \hat{l}_n^{(c)}(\tau)}{\partial \tau} = Y_{+1} - \sum_{i=1}^T \frac{A_i'(\tau, Y_{i+})}{A_i(\tau, Y_{i+})}$$

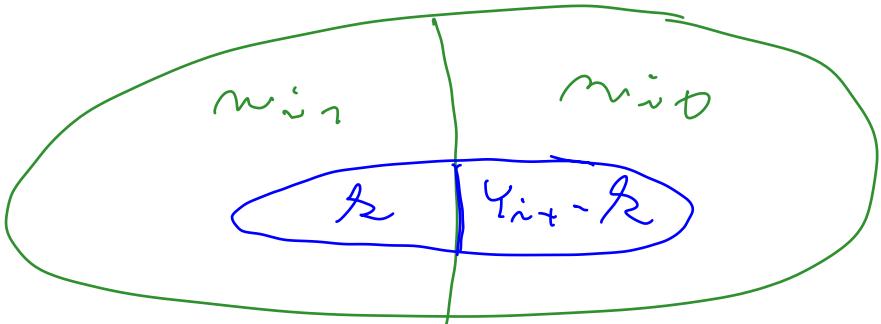
$$= Y_{+1} - \sum_{i=1}^T \frac{\sum_{k \in K_i} k \binom{n_{i1}}{k} \binom{n_{i0}}{Y_{i+}-k} e^{\tau k}}{\sum_{k \in K_i} \binom{n_{i1}}{k} \binom{n_{i0}}{Y_{i+}-k} e^{\tau k}}$$

→ for  $\tau_0 = 0$ :

$$U_n^{(c)}(0) = Y_{+1} - \sum_{i=1}^T \frac{\sum_{k \in K_i} k \binom{n_{i1}}{k} \binom{n_{i0}}{Y_{i+}-k}}{\sum_{k \in K_i} \binom{n_{i1}}{k} \binom{n_{i0}}{Y_{i+}-k}}$$

THE WAY  
HOW  $Y_{i+}$  (PATIENTS)  
CAN BE  
CHOSEN FROM  
TWO GROUPS  
OR SIZES  
 $n_{i1}$  AND  $n_{i0}$ .

$(n_{i1} + n_{i0})$   
 $Y_{i+}$



$$\sim U_n^{(c)}(0) = Y_{i+1} - \sum_{i=1}^I \sum_{k \in K_i} \frac{\binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k}}{\binom{m_{i1}+m_{i0}}{Y_{i+1}}}$$

$$P(Z_i = k)$$

$$P_{H_0}^{(c)}(Y_{i+1} = k | Y_{i+1})$$

$$\sim U_n^{(c)}(0) = \underbrace{Y_{i+1}}_{\text{OBSERVED}} - \underbrace{\sum_{i=1}^I \mathbb{E} Z_i}_{\text{EXPECTED UNDER } H_0}$$

$$(\mathbb{E}_{H_0}[Y_{i+1} | Y_{i+1}, \dots, Y_I])$$

$$\sim U_n^{(c)}(0) = \sum_{i=1}^I \left[ Y_{i+1} - Y_{i+1} \frac{m_{i1}}{m_{i1}+m_{i0}} \right]$$

$$U_n^{(c)}(t) = Y_{i+1} - \sum_{i=1}^I \frac{\sum_{k \in K_i} k \binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k} t^k}{\sum_{k \in K_i} \binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k} t^k}$$

FURTHER WE NEED:

$$I_n^{(c)}(t) = \frac{1}{n} \frac{\partial^2 U_n^{(c)}(t)}{\partial t^2} = -\frac{1}{n} \frac{\partial U_n^{(c)}(t)}{\partial t}$$

$$\sim I_n^{(c)}(0) = \frac{1}{n} \sum_{i=1}^I \left[ \frac{\sum_{k \in K_i} k^2 \binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k}}{\sum_{k \in K_i} \binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k}} - \left\{ \frac{\sum_{k \in K_i} k \binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k}}{\sum_{k \in K_i} \binom{m_{i1}}{k} \binom{m_{i0}}{Y_{i+1}-k}} \right\}^2 \right]$$

$$= \frac{1}{n} \sum_{i=1}^I \left[ \mathbb{E} Z_i^2 - \{ \mathbb{E} Z_i \}^2 \right]$$

$$= \frac{1}{n} \sum_{i=1}^I \text{var}(Z_i) \leftarrow \text{var}_{H_0}(Y_{i+1} | Y_{i+1})$$

$$= \frac{1}{n} \sum_{i=1}^I Y_{i+1} \frac{m_{i1} m_{i0}}{(m_{i1} + m_{i0})^2}$$

ANALOGOUS  
TO BINOMIAL  
DISTRIBUTION

$$\frac{m_{i1} + m_{i0} - Y_{i+1}}{m_{i1} + m_{i0} - 1}$$

FINITE SAMPLE

CORRECTION

$$R_n^{(c)} = \frac{1}{n} \frac{\left[ U_n^{(c)}(0) \right]^2}{I_n^{(c)}(0)} = \frac{\cancel{\frac{1}{n}}}{\cancel{\sum_{i=1}^I \sum_{j=1}^{m_i} \text{var}_{H_0}(Y_{ij}|Y_{i+})}} \frac{\left( \sum_{i=1}^I Y_{i+} - \mathbb{E}_{H_0}(Y_{i+}|Y_{i+}) \right)^2}{\sum_{i=1}^I \text{var}_{H_0}(Y_{i+}|Y_{i+})}$$

$$= \frac{\left( \sum_{i=1}^I Y_{i+} - Y_{i+} \frac{m_{i+}}{m_{i+} + m_{i0}} \right)^2}{\sum_{i=1}^I Y_{i+} \frac{m_{i+} m_{i0}}{(m_{i+} + m_{i0})^2} \frac{m_{i+} m_{i0} - Y_{i+}}{m_{i+} + m_{i0} - 1}}$$

$R_n^{(c)}$   $\xrightarrow[H_0: T=0]{\sim} \chi_1^2$   $\leftarrow$  COCHRAN-MANTEL  
- HAENSZEL TEST