

1. MCAR

$$X_1, \dots, X_n \text{ i.i.d } N_2 \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \frac{1}{2} \\ \frac{1}{2} & 1 \end{pmatrix} \right)$$

$X_{1,1}, \dots, X_{1000,1}$	NA, ..., NA	$X_{1501,1}, \dots, X_{2000,1}$
$X_{1,2}, \dots, X_{1000,2}$	$X_{1001,2}, \dots, X_{1500,2}$	NA, ..., NA

CCA

CI FOR μ_1 : $\left(\hat{\mu}_1 \pm t_{df} (1-\alpha/2) \sqrt{\frac{\hat{\sigma}_1^2}{n}} \right)$

IMPUTATION

$$\hat{X}_{1000+i,1} = \hat{\alpha}_0 + \hat{\alpha}_1 X_{1000+i,2}$$

↑
ESTIM. FROM X_1, \dots, X_{1000}

MULT. IMPUTATION

$$\hat{X}_{1000+i,1} = \hat{\alpha}_0 + \hat{\alpha}_1 X_{1000+i,2} + \hat{\sigma} \varepsilon_i$$

WHERE $\hat{\sigma}^2 = \frac{1}{1000 \cdot 2} \sum_{i=1}^{1000} (X_{i,2} - \hat{\alpha}_0 - \hat{\alpha}_1 X_{i,2})^2$

$\varepsilon_1, \dots, \varepsilon_{500}$ i.i.d from $N(0,1)$

OBSERVED LIKELIHOOD:

$$l_{obs}(\theta) = \sum_{i=1}^{1000} \log(f_{X_1, X_2}(X_{i,1}, X_{i,2}; \theta)) + \sum_{i=1001}^{1500} \log(f_{X_2}(X_{i,2}; \mu_2, \sigma_2^2)) + \sum_{i=1501}^{2000} \log(f_{X_1}(X_{i,1}; \mu_1, \sigma_1^2))$$

$(\mu_1, \sigma_1^2, \mu_2, \sigma_2^2, \tau_{12})$
 θ

EM-ALGORITHM:

- in \mathcal{L}^C one needs no calculus:

$$E_{\theta^{(k)}} [X_{1000+i,1} | X_{1000+i,2}] \text{ \& \ } E_{\theta^{(k)}} [X_{1000+i,1}^2 | X_{1000+i,2}]$$

$$\hat{\theta}^{(k)} = (\hat{\mu}_1^{(k)}, \hat{\mu}_2^{(k)}, \hat{\sigma}_1^{(k)}, \hat{\sigma}_2^{(k)}, \hat{\tau}_{12}^{(k)})^T$$

2. MAR

$X_{i1}, \dots, X_{i2} \text{ i.i.d } N_2 \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & 1 \end{pmatrix} \right)$

X_{i1} ALWAYS OBSERVED

$$P(X_{i2} \text{ OBSERVED} | X_{i1}) = \frac{e^{4X_{i1}}}{1 + e^{4X_{i1}}}$$

REWEIGHTING

$$\hat{\pi}_{xi} = \frac{e^{\hat{\beta}_0 + \hat{\beta}_1 X_{i1}}}{1 + e^{\hat{\beta}_0 + \hat{\beta}_1 X_{i1}}}$$

WHERE $(\hat{\beta}_0, \hat{\beta}_1)$ COME FROM THE LOGISTIC.

REGRESSION OF "X_{i2} OBSERVED" ~ X_{i1}

3. MNAR $\underline{X}_1, \dots, \underline{X}_n$ i.i.d $N_2\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \rho/2 \\ \rho/2 & \tau \end{pmatrix}\right)$

X_{i1} ALWAYS OBSERVED

X_{i2} OBSERVED IF $\frac{\exp(X_{i1} - X_{i2})}{\tau + \exp(X_{i1} - X_{i2})} \geq \frac{1}{2}$

4. LINEAR REGR. WITH MISS. VALUES IN THE COVAR.

$$\text{MODEL: } Y_i = \beta_0 + \beta_1 X_{i1} + \beta_2 X_{i2} + \varepsilon_i$$

\uparrow
0

\uparrow
0

\uparrow
1

$$\varepsilon_i \perp (X_{i1}, X_{i2})^T \quad \& \quad \varepsilon_i \sim N(0, 1)$$

$$\& \quad \begin{pmatrix} X_{i1} \\ X_{i2} \end{pmatrix} \sim N_2 \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & 1/2 \\ 1/2 & 1 \end{pmatrix} \right)$$

Y_i, X_{i1} ALWAYS OBSERVED

$$X_{i2} \text{ OBSERVED IF } \frac{\exp(X_{i1} + Y_i)}{1 + \exp(X_{i1} + Y_i)} \geq \frac{1}{2}$$

DIRECT LIKELIHOOD

→ NOTE THAT: $\begin{pmatrix} X_{i1} \\ X_{i2} \\ Y_i \end{pmatrix} \sim N_3(\dots, V)$

$$\text{cov} \left(\begin{pmatrix} X_{i1} \\ X_{i2} \end{pmatrix}, Y_i \right) = V[1:2, 3]$$

$$\begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} = \Sigma_X^{-1} \Sigma_{XY}$$

$$\text{var} \left(\begin{pmatrix} X_{i1} \\ X_{i2} \end{pmatrix} \right) = V[1:2, 1:2]$$