# H and J matrix under multivariate setting

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### H matrix

$$\hat{H}_{\mu\mu} = \begin{bmatrix} -\frac{\partial^2 \ell}{\partial (\mu_1^{BA})^2} & 0 & 0 & 0 & 0 & 0 \\ 0 & -\frac{\partial^2 \ell}{\partial (\mu_1^{CA})^2} & 0 & 0 & 0 & 0 \\ 0 & 0 & -\frac{\partial^2 \ell}{\partial (\mu_1^{BC})^2} & 0 & 0 & 0 \\ 0 & 0 & 0 & -\frac{\partial^2 \ell}{\partial (\mu_2^{BA})^2} & 0 & 0 \\ 0 & 0 & 0 & 0 & -\frac{\partial^2 \ell}{\partial (\mu_2^{CA})^2} & 0 \\ 0 & 0 & 0 & 0 & 0 & -\frac{\partial^2 \ell}{\partial (\mu_2^{CA})^2} & 0 \\ 0 & 0 & 0 & 0 & 0 & -\frac{\partial^2 \ell}{\partial (\mu_2^{CA})^2} \end{bmatrix}$$

where 
$$\frac{\partial^2 \ell}{\partial \left(\mu_2^{XZ}\right)^2} = -\sum_{i \in N_{XZ}^k} \frac{1}{s_{k,i}^{XZ^2} + \tau_k^{XZ^2}}$$

### J matrix

Denote

$$\hat{S}_{i}^{XY} = \frac{\partial \left[ \log \left( s_{k,i}^{XY}, \tau_{k}^{XY^2} \right) + \frac{\left( y_{k,i}^{XY} - \mu_{k}^{XY} \right)^2}{s_{k,i}^{XY} + \tau_{k}^{XY^2}} \right]}{\partial \eta}$$

where  $\hat{S}_i^{XY}$  should be a matrix with dimension  $1 \times 6$ J matrix would be

$$\hat{J}_{\mu\mu} = \frac{1}{4} \sum_{k=1}^{2} \left\{ \sum_{XY \in \{AB, BC, AC\}} \sum_{i \in \mathcal{N}_{XY}^{k}} S_{i}^{XY} S_{i}^{XY^{T}} \right\}$$

$$= \text{Diag} \left( \left[ \sum_{i \in \mathcal{N}_{XY}^{k}} \frac{\left(y_{k,i}^{BA} - \mu_{k}^{BA}\right)^{2}}{\left(s_{k,i}^{XY} + \tau_{k}^{XY^{2}}\right)^{2}} \right] \right)_{6 \times 6}$$

where  $\eta = \left[\mu_k^{XY}\right]_{1\times 6}$  for outcome k=1,2 and treatment  $XY \in \{BA,CA,BC\}$ 

## Covariance Matrix

$$\hat{V}(\hat{\eta}) = \hat{H}_{\mu\mu}^{-1} \hat{J}_{\mu\mu} \hat{H}_{\mu\mu}^{-1}$$

#### Reference

The appendix in Duan2020