

Problem 1.

$$F_{\alpha\beta} = \frac{1}{2} \text{tr} \left(C^{-1} \frac{\partial C}{\partial \theta^\alpha} C^{-1} \frac{\partial C}{\partial \theta^\beta} + 2C^{-1} \frac{\partial \mu}{\partial \theta^\alpha} \frac{\partial \mu^T}{\partial \theta^\beta} \right)$$

$$\frac{\partial C^{-1}}{\partial \theta^\alpha} = -C^{-1} \frac{\partial C}{\partial \theta^\alpha} C^{-1}$$

The Fisher Matrix is given by :

$$F_{\alpha\beta} = - \left\langle \frac{\partial \ln L(D|\theta)}{\partial \theta^\alpha} \frac{\partial \ln L(D|\theta)}{\partial \theta^\beta} \right\rangle$$

The negative of the log-likelihood is given by :

$$- \ln L = \frac{1}{2} \text{tr} (\ln C + C^{-1} D) + \text{const}$$

Takin 1st derivative: $\frac{\partial}{\partial \alpha} = \frac{\partial}{\partial \theta^\alpha}$

$$- \frac{\partial \ln L}{\partial \theta^\alpha} = \frac{1}{2} \text{tr} \left(C^{-1} \frac{\partial C}{\partial \alpha} + \frac{\partial C^{-1}}{\partial \alpha} D + C^{-1} \frac{\partial D}{\partial \alpha} \right)$$

$$\begin{aligned} - \frac{\partial^2 \ln L}{\partial \theta^\alpha \partial \theta^\beta} &= \frac{1}{2} \text{tr} \left(\frac{\partial C^{-1}}{\partial \beta} \frac{\partial C}{\partial \alpha} + \underbrace{C^{-1} \frac{\partial^2 C}{\partial \beta \partial \alpha}}_{\text{sym}} + \frac{\partial^2 C^{-1}}{\partial \beta \partial \alpha} D + \frac{\partial C^{-1}}{\partial \alpha} \frac{\partial D}{\partial \beta} + \frac{\partial C^{-1}}{\partial \beta} \frac{\partial D}{\partial \alpha} + \underbrace{C^{-1} \frac{\partial^2 D}{\partial \beta \partial \alpha}}_{\text{sym}} \right) \\ &= \frac{1}{2} \text{tr} \left(\frac{\partial C^{-1}}{\partial \beta} \left(\frac{\partial C}{\partial \alpha} + \frac{\partial D}{\partial \alpha} \right) + C^{-1} \left(\frac{\partial^2 C}{\partial \beta \partial \alpha} + \frac{\partial^2 D}{\partial \beta \partial \alpha} \right) + \frac{\partial^2 C^{-1}}{\partial \beta \partial \alpha} D + \frac{\partial C^{-1}}{\partial \alpha} \frac{\partial D}{\partial \beta} \right) \end{aligned}$$

Lets compute the expectation values for:

$$\left\langle \frac{\partial D}{\partial \alpha} \right\rangle = \frac{\partial}{\partial \alpha} \langle D \rangle = \frac{\partial C}{\partial \alpha} = \left\langle \frac{\partial}{\partial \alpha} ((x-\mu)(x-\mu)^T) \right\rangle = \left\langle -\frac{\partial \mu}{\partial \alpha} (x-\mu)^T - (x-\mu) \left(\frac{\partial \mu}{\partial \alpha} \right)^T \right\rangle$$

$$\left\langle \frac{\partial^2 D}{\partial \alpha \partial \beta} \right\rangle = \frac{\partial^2 C}{\partial \alpha \partial \beta} = \left\langle \frac{\partial^2 \mu}{\partial \alpha \partial \beta} (x-\mu)^T + \frac{\partial \mu}{\partial \beta} \frac{\partial \mu^T}{\partial \alpha} + \frac{\partial \mu}{\partial \alpha} \frac{\partial \mu^T}{\partial \beta} - \frac{\partial^2 \mu^T}{\partial \alpha \partial \beta} (x-\mu) \right\rangle = 0 \quad \text{since } \langle x-\mu \rangle = 0 \text{ and } \frac{\partial \mu}{\partial \alpha} = 0$$

$$\frac{\partial C^{-1}}{\partial \beta \partial \alpha} = \frac{\partial}{\partial \beta} \left(C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} \right) = C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} - \underbrace{C^{-1} \frac{\partial^2 C}{\partial \beta \partial \alpha} C^{-1}}_{\text{sym}} + C^{-1} \frac{\partial C}{\partial \alpha} \left(C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \right) = 2C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial C}{\partial \alpha} C^{-1}$$

All crossed terms are zero, it follows

$$- \ln L = \frac{1}{2} \text{tr} \left(C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial C}{\partial \alpha} - C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} \frac{\partial D}{\partial \beta} + 2C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} \frac{\partial C^{-1}}{\partial \beta} D - \underbrace{C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} \frac{\partial D}{\partial \beta}}_{\text{sym}} \right)$$

$$= \frac{1}{2} \text{tr} \left(-C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial C}{\partial \alpha} - 2C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial D}{\partial \alpha} + 2C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} \frac{\partial C^{-1}}{\partial \beta} D \right)$$

$$= \frac{1}{2} \text{tr} \left(-C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial C}{\partial \alpha} - 2C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial D}{\partial \alpha} - 2C^{-1} \frac{\partial C}{\partial \alpha} C^{-1} \left(C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \right) D \right)$$

$$= \frac{1}{2} \text{tr} \left(-C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \frac{\partial C}{\partial \alpha} - 2C^{-1} \frac{\partial C}{\partial \beta} C^{-1} \left(\frac{\partial D}{\partial \alpha} + \frac{\partial C}{\partial \alpha} C^{-1} D \right) \right)$$

Where the 2nd term

$$\left\langle \frac{\partial D}{\partial \alpha} \right\rangle = \frac{\partial C}{\partial \alpha} \Rightarrow \underbrace{\left\langle -\frac{\partial \mu}{\partial \alpha} (x-\mu)^T - (x-\mu) \frac{\partial \mu^T}{\partial \alpha} \right\rangle C^{-1} \left(-\frac{\partial \mu}{\partial \alpha} (x-\mu)^T - (x-\mu) \frac{\partial \mu^T}{\partial \alpha} - \frac{\partial \mu}{\partial \alpha} (x-\mu)^T C^{-1} D - (x-\mu) \frac{\partial \mu^T}{\partial \alpha} C^{-1} D \right)}_{\text{should be equal}} \Rightarrow \frac{\partial \mu}{\partial \alpha} \frac{\partial \mu^T}{\partial \beta}$$

$$2. \vec{y} = A\vec{x} \Rightarrow C' = \text{cov}(\vec{y}) = \langle (y - E(y))(y - E(y))^T \rangle = \langle A(\vec{x} - \vec{\mu}) \cdot [A(\vec{x} - \vec{\mu})]^T \rangle = A \cdot C \cdot A^T$$

$$C'^{-1} = (ACA^T)^{-1} = A^{-T} C^{-1} A^{-1}$$

$$\mu' = A\mu, \quad \mu'^T = (A\mu)^T = \mu^T A^T$$

$$\text{Then } C'^{-1} \frac{\partial C'}{\partial \theta^\alpha} C' = A^{-T} C^{-1} A^{-1} \cdot \frac{\partial (ACA^T)}{\partial \theta^\alpha} \cdot (A^{-T} C^{-1} A^{-1}) \frac{\partial (ACA^T)}{\partial \theta^\beta} A^T$$

$$= A^{-T} C^{-1} (A^{-1} A) \frac{\partial C}{\partial \theta^\alpha} (A^T A^{-T}) C^{-1} (A^{-1} A) \frac{\partial C}{\partial \theta^\beta} A^T$$

$$= A^{-T} C^{-1} \frac{\partial C}{\partial \theta^\alpha} C^{-1} \frac{\partial C}{\partial \theta^\beta} A^T = C^{-1} \frac{\partial C}{\partial \theta^\alpha} C^{-1} \frac{\partial C}{\partial \theta^\beta}$$

$$2C^{-1} \frac{\partial \mu'}{\partial \theta^\alpha} \frac{\partial \mu'^T}{\partial \theta^\beta} = 2 \cdot A^{-T} C^{-1} A^{-1} \cdot \frac{\partial (A\mu)}{\partial \theta^\alpha} \cdot \frac{\partial (\mu^T A^T)}{\partial \theta^\beta}$$

$$= 2 \cdot A^{-T} C^{-1} (A^{-1} A) \frac{\partial \mu}{\partial \theta^\alpha} \cdot \frac{\partial \mu^T}{\partial \theta^\beta} A^T$$

$$= 2 \cdot A^{-T} \cdot C^{-1} \frac{\partial \mu}{\partial \theta^\alpha} \frac{\partial \mu^T}{\partial \theta^\beta} A^T$$

$$= 2C^{-1} \frac{\partial \mu}{\partial \theta^\alpha} \frac{\partial \mu^T}{\partial \theta^\beta}$$

Therefore, any linear transformation of data leaves Fisher matrix unchanged.