

1st U7

3) W_1, \dots, W_K ... unbiased estimators of θ with $\text{Var} = \sigma_i^2$ and $\text{Cov}(W_i, W_j) = 0$ if $i \neq j$

show $a_i \dots$ constant and $E_\theta(\sum a_i W_i) = \theta$

of all estimators $\sum a_i W_i$ $W^* = \frac{\sum W_i / \sigma_i^2}{\sum (1/\sigma_i^2)}$ has minimum variance

$$\theta = E(\sum a_i W_i) = \sum a_i E(W_i) = \sum a_i \theta = \theta (\sum a_i) \Rightarrow \sum a_i = 1$$

$$\text{Var}(\sum a_i W_i) = \sum a_i^2 \text{Var}(W_i) = \sum a_i^2 \sigma_i^2$$

our goal is to minimize $\sum a_i^2 \sigma_i^2$ with the condition $\sum a_i = 1$

$$L(a, \lambda) = \sum a_i^2 \sigma_i^2 - \lambda (\sum a_i - 1)$$

$$\frac{\partial L}{\partial a_i} = 2a_i \sigma_i^2 - \lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = -(\sum a_i - 1) = 0$$

$$\Rightarrow a_i = \frac{\lambda}{2\sigma_i^2}$$

$$\Rightarrow \sum a_i = \sum \frac{\lambda}{2\sigma_i^2} = \frac{\lambda}{2} \sum \frac{1}{\sigma_i^2}$$

$$\Rightarrow \lambda = 2(\sum \frac{1}{\sigma_i^2})^{-1} \Rightarrow a_i = \frac{\frac{1}{\sigma_i^2}}{\sum \frac{1}{\sigma_j^2}}$$

$$\Rightarrow \sum a_i W_i = \frac{\sum \frac{W_i}{\sigma_i^2}}{\sum \frac{1}{\sigma_i^2}} = W^*$$

$$\text{show } \text{Var } W^* = \frac{1}{\sum \frac{1}{\sigma_i^2}}$$

$$\begin{aligned} \text{Var}(W^*) &= \sum a_i^2 \sigma_i^2 = \left(\sum_i \left(\frac{\frac{1}{\sigma_i^2}}{\sum_j \frac{1}{\sigma_j^2}} \right) \right)^2 \sigma_i^2 = \sum_i \frac{\frac{1}{\sigma_i^2}}{\sum_j \frac{1}{\sigma_j^2}^2} = \frac{1}{(\sum_j \frac{1}{\sigma_j^2})^2} \sum_i \frac{1}{\sigma_i^2} \\ &= \frac{1}{\sum \frac{1}{\sigma_i^2}} \end{aligned}$$