

# Some of my white board writing in week 8 (1)

1) I discussed the exponential distribution example on p. 46-47 of the notes. Looking at the amount of material provided in the notes, I believe that you should be able to reconstruct the details of this example by yourself.

2) Proof of the Neyman-Pearson Lemma. Again, I did complete derivation on the white board but looking at the content that is put on p. 50-51 of the notes, I believe that you should be able to reconstruct the details of the proof yourself.

3.) Example about uniformly most powerful (UMP)  $\chi$ -test for the normal distribution:

$X = (X_1, X_2, \dots, X_n)$  i.i.d.  $N(\theta, 1)$ . Consider  $H_0: \theta = \theta_0 \in \mathbb{R}^1$  versus a composite  $H_1: \theta > \theta_0$ . are all these alternative

Looking for UMP  $\chi$ -test  $\varphi^*$  which

means: if we take any competitor

$\varphi \in \Phi_{\chi} = \{ \text{set of all tests } \varphi \text{ such that } E_{\theta_0} \varphi \leq \alpha \}$

then we claim that  $E_{\theta} \varphi^* \geq E_{\theta} \varphi$  for all  $\theta > \theta_0$ .

We first simplify the problem by considering testing simple  $H_0: \theta = \theta_0$  versus simple  $H_1: \theta = \theta_1$  for a fixed  $\theta_1 > \theta_0$ . Because this is a Neyman-Pearson lemma-type problem, for it we have the most powerful  $\chi$  test and it is given by

$$\varphi^* = \begin{cases} 1 & \text{if } L(X, \theta_1) / L(X, \theta_0) > C \\ 0 & \text{if } L(X, \theta_1) / L(X, \theta_0) \leq C \end{cases}$$

(2)

Notice that

$$\frac{L(X, \theta_1)}{L(X, \theta_0)} = \exp\left((\theta_1 - \theta_0) \sum_{i=1}^n X_i + \frac{n}{2}(\theta_0^2 - \theta_1^2)\right)$$

Since  $\theta_1 - \theta_0 > 0$ ,  $\frac{L(X, \theta_1)}{L(X, \theta_0)}$  is monotonically increasing in  $T = \sum_{i=1}^n X_i$  and  $\frac{L(X, \theta_1)}{L(X, \theta_0)} > C$  is equivalent to

$$\sum_{i=1}^n X_i > C_1 \quad \text{or, by renaming constants, to } \bar{X} > \bar{C}.$$

To find  $\bar{C}$ , we must "exhaust" the given level  $\alpha$  which means  $E_{\theta_0} \varphi^* = 1 * P_{\theta_0}(\bar{X} > \bar{C}) = \alpha$  must hold (see the statement of the NP Lemma).

$$\begin{aligned} \text{But } E_{\theta_0} \varphi^* &= P_{\theta_0}(\bar{X} > \bar{C}) = P_{\theta_0}\left(\frac{\sqrt{n}(\bar{X} - \theta_0)}{1} > \frac{\sqrt{n}(\bar{C} - \theta_0)}{1}\right) \\ &= P(Z > \frac{\sqrt{n}(\bar{C} - \theta_0)}{1}) = \alpha \quad \text{where } Z \sim N(0, 1). \end{aligned}$$

This implies that  $\frac{\sqrt{n}(\bar{C} - \theta_0)}{1} = z_\alpha$  must hold where

$z_\alpha$  is the upper  $\alpha * 100\%$  point of the  $N(0, 1)$ .

Then  $\bar{C} = \theta_0 + \frac{z_\alpha}{\sqrt{n}}$  and  $\varphi^*$  becomes:

$$\varphi^*(X) = \begin{cases} 1 & \text{if } \bar{X} > \theta_0 + \frac{z_\alpha}{\sqrt{n}} \\ 0 & \text{if } \bar{X} \leq \theta_0 + \frac{z_\alpha}{\sqrt{n}} \end{cases}$$

NOW WE NOTICE that the resulting  $\varphi^*(X)$  above, although having been constructed for a particular

$H_1: \theta = \theta_1$ , DOES NOT involve this  $\theta_1 > \theta_0$  in its shape.

Hence the SAME test  $\varphi^*$  will be the most powerful  $\alpha$ -test for any chosen  $\theta_1 > \theta_0$ ! Therefore,  $\varphi^*(X)$  will be the UNIFORM most powerful for testing also  $H_0: \theta = \theta_0$  versus  $H_1: \theta > \theta_0$ .



Notice that we used the monotonicity of the Likelihood ratio in our argument.

This example was generalized in the Blackwell-Girshick (BG) Theorem (p.52 of the notes).

I also gave an example of applying the BG theorem to derive UMP  $\alpha$  tests.

Example: Assume that  $X = (X_1, X_2, \dots, X_n)$  are i.i.d. from  $f(x, \theta) = \begin{cases} 2x/\theta^2, & 0 < x < \theta \\ 0 & \text{else.} \end{cases}$  Construct a UMP  $\alpha$  test.

of  $H_0: \theta \leq \theta_0$  versus  $H_1: \theta > \theta_0$ .

Solution: First we want to show that the family  $L(X, \theta)$  in this case is a MLR family in the statistic  $T = X_{(n)}$ .

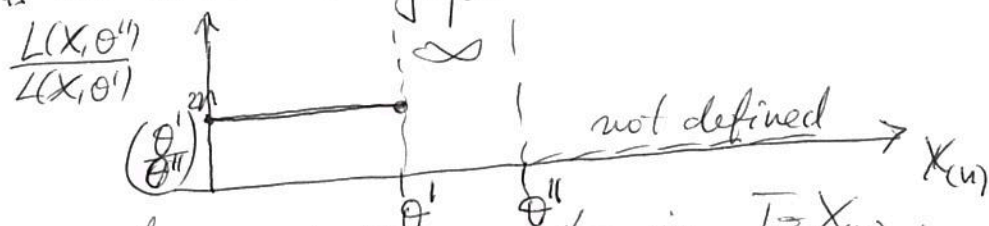
Indeed:  $L(X, \theta) = \frac{2^n \prod_{i=1}^n x_i I_{(X_{(n)}, \infty)}(\theta)}{\theta^{2n}}$  ( $\theta$ ). Now take

two values  $0 < \theta' < \theta''$  and consider

$$\frac{L(X, \theta'')}{L(X, \theta')} = \left(\frac{\theta'}{\theta''}\right)^{2n} \frac{I_{(X_{(n)}, \infty)}(\theta'')}{I_{(X_{(n)}, \infty)}(\theta')}$$

Putting  $X_{(n)}$  on the

OX axis we have the graph:



Hence we have MLR property in  $T = X_{(n)}$ . Then BG theorem tells us that UMP  $\alpha$  test of  $H_0$  vs  $H_1$  exists and is given by

$$\varphi^* = \begin{cases} 1 & \text{if } X_{(n)} > K \\ 0 & \text{if } X_{(n)} \leq K \end{cases}$$

To find  $K$  we need to exhaust the level, i.e. must satisfy  $E_{\theta_0} \varphi^* = \alpha$ . However  $E_{\theta_0} \varphi^* = P_{\theta_0}(X_{(n)} > K) = 1 - P_{\theta_0}(X_{(n)} \leq K)$   
 $= 1 - [P_{\theta_0}(X_1 \leq K)]^n = 1 - \left(\frac{K}{\theta_0}\right)^{2n} = \alpha \Rightarrow \boxed{K = \theta_0 (1 - \alpha)^{\frac{1}{2n}}}$  and  $\varphi^*$  is completely determined.