

EE6 310

May 10, 2018

x_1, x_2, \dots, x_n IID unknown μ, σ^2

$$\hat{\mu} = \bar{x}_n = \frac{\sum_{i=1}^n x_i}{n} = \text{sample average}$$

$E\hat{\mu} = \mu \Rightarrow \text{unbiased}$

$\text{Var } \hat{\mu} = \frac{\sigma^2}{n} \rightarrow 0 \text{ as } n \rightarrow \infty \text{ consistency}$

$$\hat{\sigma}^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x}_n)^2 \quad E\hat{\sigma}^2 = \sigma^2$$

Order statistics $x_1, \dots, x_n \rightarrow \boxed{\text{sort}}$

$x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$

$\uparrow \quad \quad \quad \uparrow$

min $x_{(n/2)}$ max

median

Hypothesis Testing

null

$$H_0: X \sim N(0, \sigma^2)$$

hypothesis

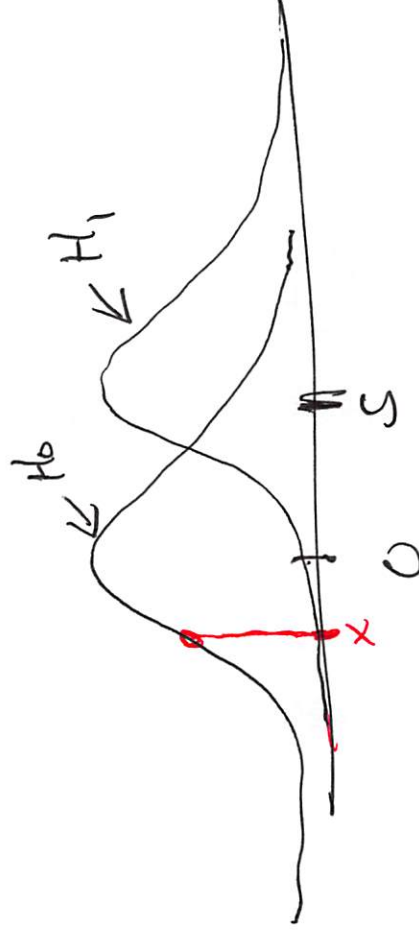
alternative

$$H_1: X \sim N(s, \sigma^2)$$

hypothesis

s = strength of reflection

$$s > 0$$



$$\hat{H} = \begin{cases} 1 \\ 0 \end{cases}$$

guess H_0 is True

guess H_1 is True

Defini:

True Positive
rate

$$P(\hat{H}=1 \mid H_1 = \text{True})$$

True Neg

$$\hat{H}=0 \text{ and } H_0 \text{ is True}$$

False Alarm - False Pos

$$\hat{H}=1 \text{ and } H_0 \text{ is False}$$

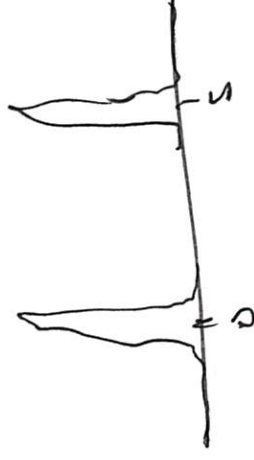
Miss - False Neg

$$\hat{H}=0 \text{ and } H_1 \text{ is True}$$

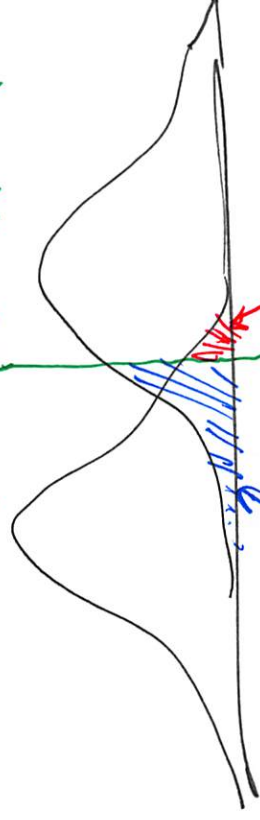
$$P(TP) + P(FN) = 1$$

$$P(FP) + P(TN) = 1$$

GLS



$\hat{H}=0 \rightarrow \hat{H}=1$



$$P(X < X_T \mid H_1) = P(\hat{H}=0 \mid H_1)$$

= miss prob

$$P(X > X_T \mid H_0) = P(\hat{H}=1 \mid H_0)$$

= false alarm prob

Easy: Communications

$$P(H_0) = P(H_1) = \frac{1}{2}$$

$$\begin{matrix} \nearrow & b\hat{H}=1 \\ \nwarrow & b\hat{H}=0 \end{matrix}$$

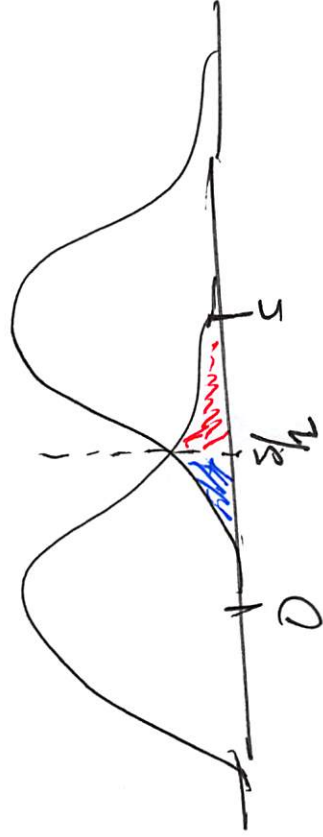
$$\text{minimize } P(FP) + P(FN)$$

$$P(\text{error}) = P(\hat{H} \neq 1 | H_0) P(H_0) + P(\hat{H} \neq 0 | H_1) P(H_1)$$

$$\text{min } P(\text{error})$$

in this example $H_0: X \sim N(0, \sigma^2)$
 $H_1: X \sim N(s, \sigma^2)$

$$X_T = \sum$$



Neyman-Pearson Test

$$\text{Likelihood Ratio } L(x) = \frac{f_1(x)}{f_0(x)} \quad \begin{matrix} H_1 \\ \geq \\ H_0 \end{matrix}$$

if $L(x) > L_0$ decide $H=1$

$L(x) < L_0$ decide $H=0$

$$\text{Ex. } L(x) = \frac{\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}}{\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x}{\sigma}\right)^2}} = \frac{e^{-\frac{1}{2\sigma^2}(x^2 - 2\mu x + \sigma^2)}}{e^{-\frac{1}{2\sigma^2}(x^2)}}$$

$$= e^{\frac{\mu x}{\sigma^2} - \frac{\sigma^2}{2\sigma^2}} = e^{\frac{\mu x}{\sigma^2} - \frac{1}{2}}$$

X_T gives equality $\geq L_0$

$$l(x) = \log(L(x)) \stackrel{?}{\geq} l_0 = \log(L_0)$$

$$+ \frac{Sx}{\sigma^2} - \frac{s^2}{2\sigma^2} \stackrel{?}{\geq} l_0$$

$$x = l_0 + \frac{s^2}{2\sigma^2}$$

$$+ \frac{Sx}{\sigma^2} \geq l_0 + \frac{s^2}{2\sigma^2}$$

$$x \geq \frac{\sigma^2}{S} \left(l_0 + \frac{s^2}{2\sigma^2} \right) = x_T$$

$$\text{if } x > x_T, \hat{H} = 1$$

$$\text{else, } \hat{H} = 0$$

FP: H_0 true

$$P(FP) = P(X > \bar{x}_T | H_0)$$

$$= 1 - P(X < \bar{x}_T | H_0)$$

$$= 1 - P\left(\frac{X - \mu}{\sigma} < \frac{\bar{x}_T - \mu}{\sigma} \mid H_0\right)$$

$$= 1 - \Phi\left(\frac{\bar{x}_T - \mu}{\sigma}\right)$$

$$P(FN) = P(X < \bar{x}_T | H_1)$$

$$= P\left(\frac{X - \mu}{\sigma} < \frac{\bar{x}_T - \mu}{\sigma} \mid H_1\right)$$

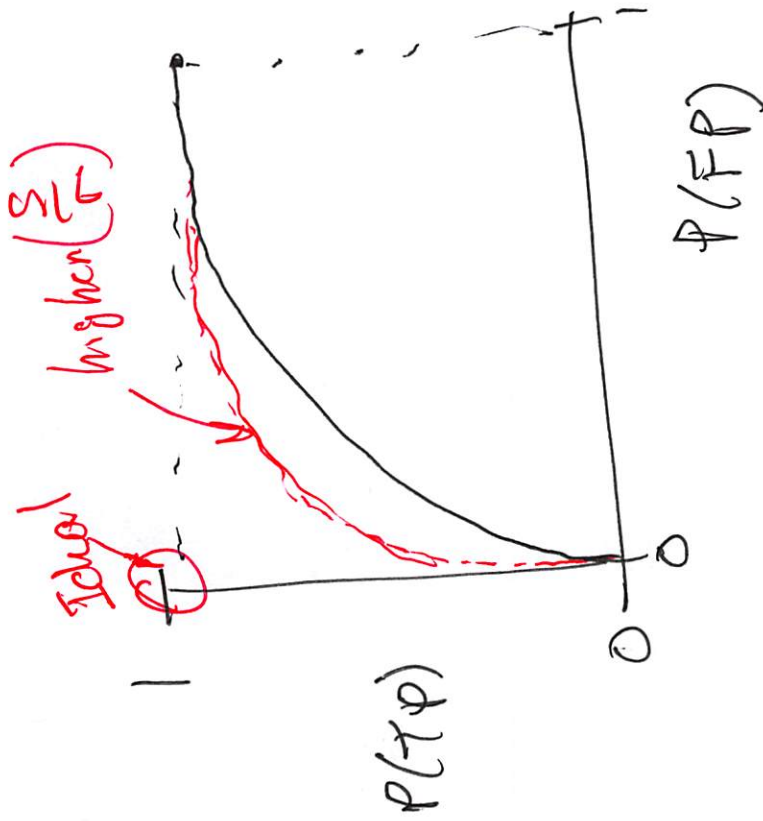
$$= \Phi\left(\frac{\bar{x}_T - \mu}{\sigma}\right)$$

↑

solve for \bar{x}_T

subs into

$$P(TP) = 1 - \Phi\left(\Phi^{-1}(1 - P(FP)) - \frac{z}{\sigma}\right)$$



ROC curve

(Receiver Operating
Characteristic)