

3 Detection

4.5.22

①

We apply theory from Ch 2 to detection problems.

Given : $\left. \begin{array}{l} \cdot \text{data } \underline{x} \\ \cdot \text{likelihood} \end{array} \right\} \text{mandatory}$
 $\left. \begin{array}{l} \cdot \text{prior} \\ \cdot \text{loss} \end{array} \right\} \text{optional}$

Desired : $\left. \begin{array}{l} \cdot \text{Detector} \\ \cdot \text{Detector Performance} \\ \cdot \text{How to optimize system parameters for an optimized system} \end{array} \right\}$

Comments : $\left. \begin{array}{l} \cdot \text{Only binary detection} \\ \cdot \text{Use detection notation } H_0, H_1 \text{ instead of } w_1, w_2 \end{array} \right\}$

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A general Log Likelihood Ratio Test (LRT) detector:

$$\ln \left(\frac{P(\underline{x} | H_1)}{P(\underline{x} | H_0)} \right) = -\frac{1}{2\sigma^2} \left[\sum_{n=0}^{d-1} (x(n) - A)^2 - \sum_{n=0}^{d-1} (x(n))^2 \right]$$

$$= -\frac{1}{2\sigma^2} \left[-2A \sum_n x(n) + dA^2 \right] \cdot \frac{-2dA^2}{-2dA^2}$$

• SNR is best for
comparison in detection
of noise in AWGN

$$= \underbrace{\frac{dA^2}{\sigma^2}}_{P > 0} \left[\underbrace{\frac{1}{A} \cdot \frac{1}{d} \sum_n x(n)}_{\bar{x}} - \frac{1}{2} \right]$$

$$= \underbrace{t(\underline{x})}_{t(\underline{x})}$$

$$= \underbrace{P}_{H_0} \left(t(\underline{x}) - \frac{1}{2} \right) \underset{H_0}{\gtrless} \underbrace{\ln \tilde{\gamma}}_{\substack{\text{original} \\ \text{threshold} \\ \text{for LR}}}$$

$$= t(\underline{x}) \underset{H_0}{\gtrless} \frac{1}{\rho} \ln \tilde{\gamma} + \frac{1}{2} := \gamma : \text{threshold for } t(\underline{x})$$

with
• test $t(\underline{x}) = \frac{\bar{x}}{A}$

• $\frac{A^2}{\sigma^2}$: Signal to noise ratio (SNR) of $x(n) | H_1 = A + z(n) \sim N(A, \sigma^2)$

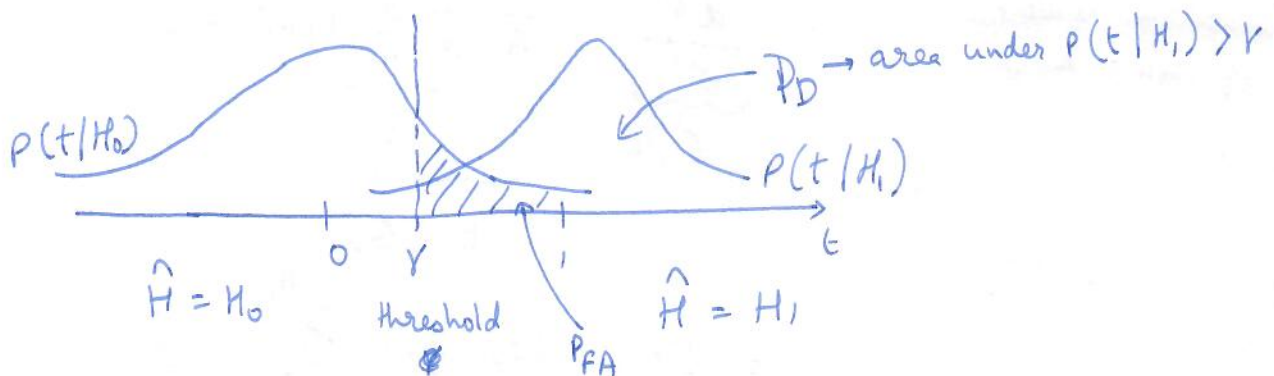
• $\rho = d \cdot \frac{A^2}{\sigma^2}$: SNR of $\bar{x} | H_1 = A + \bar{z} \sim N(A, \frac{\sigma^2}{d})$

Distribution of $t(x)$

$$H_0: t(x) = \frac{1}{A} \bar{z} \sim N\left(0, \frac{1}{A} \frac{\sigma^2}{d}\right) = N\left(0, \frac{1}{\rho}\right)$$

$$H_1: t(x) = \frac{1}{A} (A + \bar{z}) \sim N\left(1, \frac{1}{A} \frac{\sigma^2}{d}\right) \cong N\left(1, \frac{1}{\rho}\right)$$

Detailed performance measures P_{FA} , P_D :



Thus we can't only reduce P_{FA} and increase P_D as there is an overlap of the area of operations. Hence we need to make some compromise b/w them.

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False alarm rate

$$\begin{aligned} P_{FA}(\gamma) &= P(\hat{H} = H_1 | H_0 = H_0) = P(t > \gamma | H_0); t \sim N\left(0, \frac{1}{\rho}\right) \\ &= P\left(\underbrace{\sqrt{\rho} t}_{\sim N(0, 1)} > \sqrt{\rho} \gamma\right) \\ &= Q(\sqrt{\rho} \gamma) \end{aligned}$$

Detection rate

$$\begin{aligned} P_D(\gamma) &= P(\hat{H} = H_1 | H = H_1) = P(t > \gamma | H_1); t \sim N\left(1, \frac{1}{\rho}\right) \\ &= P\left(\underbrace{\sqrt{\rho} (t-1)}_{\sim N(0, 1)} > \sqrt{\rho} (\gamma-1) | H_1\right) = Q(\sqrt{\rho} (\gamma-1)) \end{aligned}$$

$\gamma \uparrow: \begin{matrix} P_{FA} \downarrow \\ P_D \downarrow \end{matrix} \left. \vphantom{\begin{matrix} P_{FA} \downarrow \\ P_D \downarrow \end{matrix}} \right\} \begin{matrix} \text{max } P_D \text{ and min } P_{FA} \\ \text{impossible, only compromise} \end{matrix}$

Special optimum detectors

(a) MBR: $Y_{MBR} = \frac{1}{P} \ln \tilde{Y}_{MBR} + \frac{1}{2}$

$$= \frac{1}{P} \ln \left(\frac{\lambda_0 - \lambda_{00}}{\lambda_{01} - \lambda_{11}} \cdot \frac{P_0}{P_1} \right) + \frac{1}{2}$$

(b) ML: 0/1 loss, $P_0 = P_1 = \frac{1}{2}$ $\tilde{Y}_{ML} = 1$, $Y_{ML} = \frac{1}{2}$

i.e. $\frac{\bar{x}}{A} \underset{H_0}{\underset{H_1}{>}} \frac{1}{2}$

$$P_{FA}(Y_{ML}) = Q\left(\frac{1}{2}\sqrt{P}\right)$$

$$P_D(Y_{ML}) = Q\left(-\frac{1}{2}\sqrt{P}\right)$$

$$= 1 - Q\left(\frac{1}{2}\sqrt{P}\right)$$

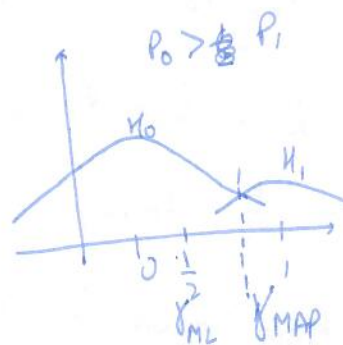
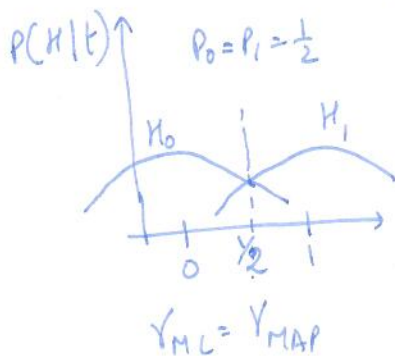
$$P_{FA} + P_D = 1$$

(c) MAP: 0/1 loss $P_0 \neq \frac{1}{2}$

$$Y_{MAP} = \frac{1}{P} \log \left(\frac{P_0}{P_1} \right) + \frac{1}{2}$$

$$\begin{aligned} P_0 &> P_1 - \epsilon \\ &> Y_{ML} = \frac{1}{2} \\ &< \\ P_0 &< P_1 - \epsilon' \end{aligned}$$

MAP decision: $\max_H P(H|t) \sim P(t|H) \cdot P(H)$



(3)

(d) Neyman-Pearson : no priors, no loss.

$$P_{FA}(\gamma) = Q(\sqrt{\rho} \gamma_{NP}) \stackrel{!}{=} \alpha \leftarrow \text{system threshold set by us.}$$

$$\Rightarrow \gamma_{NP} = \frac{1}{\sqrt{\rho}} Q^{-1}(\alpha)$$

How to improve the detector performance?

$$\text{increase } \rho = d \cdot \frac{A^2}{\sigma^2}$$

- increase signal power A^2 : not always possible.
- increase the number of measurements d : not always possible
- reduce noise power σ^2

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E3.2 : Bit detection

$$H_0 : x(n) = -1 + z(n) \quad z(n) \text{ iid } N(0, \sigma^2)$$

$$H_1 : x(n) = 1 + z(n)$$

$$\text{Let } y(n) = x(n) + 1 \begin{cases} H_0 = z(n) \\ H_1 = A + z(n) \end{cases} \quad A = 2$$

Same soln as E3.1

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