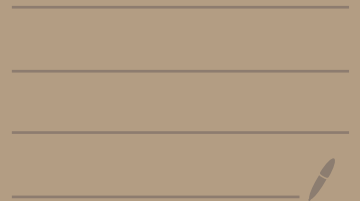


Principles of Digital Communication

Feb 26th, 2021



Binary Hypothesis Testing

Setup:

Object of Interest

Observation

Guess

H



γ

$$\hat{H} = \text{func}(\gamma)$$

i

z

$$\hat{c} = \phi(z)$$

Reading for next week:

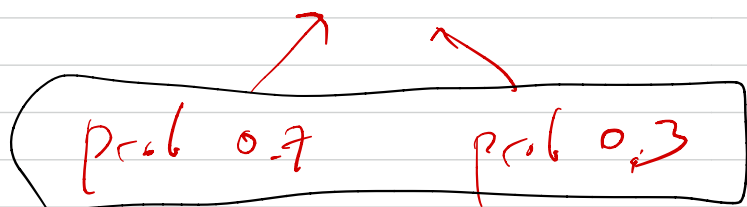
Chapter 1: 1.2 - 1.6

2: 2.1 - 2.2.1

Ex: γ is nothing

$$H \in \{0, 1\}$$

$$\hat{H} = ?$$



Proposal 1: set $\hat{H} = 1$

Proposal 2: " $\hat{H} = 0$

Proposal 3: " $\hat{H} = \begin{cases} 0 & \text{wp } 0.7 \\ 1 & \text{wp } 0.3 \end{cases}$

↙ with probability

Q: $\Pr(\text{correct}) = ?$ with proposal 3
 $= 0.7$

Copy from prev:

$$H = \begin{cases} 0 & \text{w.p. } 0.7 \\ 1 & \text{w.p. } 0.3 \end{cases}$$

$Y = \text{nothing}$

process 1 $\hat{H} = 1 \longrightarrow \Pr(\text{correct}) = 0.3$

2 $\hat{H} = 0 \longrightarrow \Pr(\text{correct}) = 0.7$

3 $\hat{H} = \begin{cases} 0 & \text{w.p. } 0.7 \\ 1 & \text{w.p. } 0.3 \end{cases}$

$\longrightarrow \Pr(\text{correct}) = (0.7)^2 + (0.3)^2$

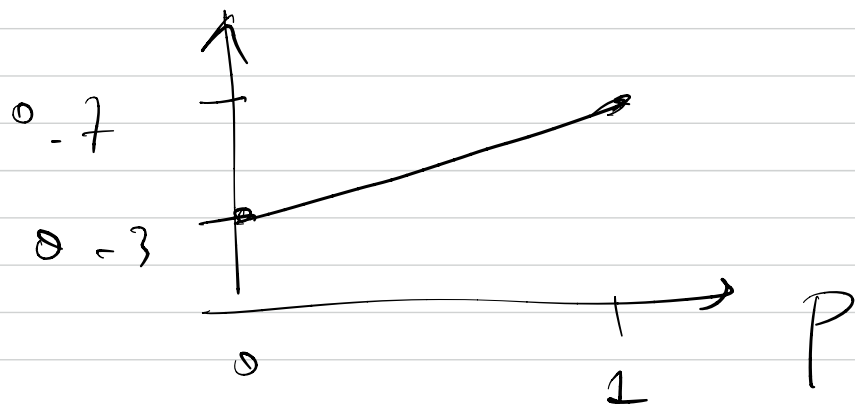
H & $\hat{H}$ are indep.

$= 0.58$

Can we do better? Suppose

$$\hat{H} = \begin{cases} 0 & \text{w.p. } p \\ 1 & \text{w.p. } 1-p \end{cases}$$

$$\begin{aligned} \Pr(H = \hat{H}) &= \Pr(H, \hat{H} = (0,0)) + \Pr(H, \hat{H} = (1,1)) \\ &= (0.7)p + (0.3)(1-p) \end{aligned}$$



General (Binary) Case:

$$H \in \{0, 1\} \quad Y \in \mathcal{Y}$$

$$\text{A-priori probability } p \quad H = \begin{cases} 0 & \text{w.p. } 1-p = P_H(0) \\ 1 & \text{w.p. } p = P_H(1) \end{cases}$$

The observation has two possible statistics

$$P_{Y|H}(y|0), \quad P_{Y|H}(y|1) = P_Y(Y=y | H=1)$$

We want: rule: $\hat{H} = \phi(Y)$ so as to maximize the probability that $H = \hat{H}$.

$$P_r(H = \hat{H}) = \sum_y P_r(Y=y) P_r(H = \hat{H} | Y=y)$$

$$= P_r(H = \hat{H}, Y \in \mathcal{Y})$$

$$= \sum_{y \in \mathcal{Y}} P_r(\hat{H} = H, Y=y)$$

$$= \sum_{y \in \mathcal{Y}} P_r(Y=y) P_r(H = \hat{H} | Y=y)$$

$$= \sum_{y \in \mathcal{Y}} P_r(Y=y) \underbrace{P_r(H = \phi(y) | Y=y)}$$

To maximize $P_r(H = \hat{H})$ we should maximize for each y

$$P_r(H = \phi(y) | Y=y).$$

Note that conditioned on $Y=y$,

$$\rightarrow P_r(H=0 | Y=y) = \frac{P_r(Y=y | H=0) \cdot P(H=0)}{P(Y=y)}$$

$$Pr(H=1 | Y=5) = Pr(Y=5 | H=1) \frac{Pr(H=1)}{Pr(Y=2)}$$

Consequently we should choose

$$\phi(\gamma) = \begin{cases} 0 & \text{if } \underline{\Pr(H=0/Y=0)} > \underline{\Pr(H=1/Y=2)} \\ 1 & "<" \\ \text{don't care} & "=" \\ \text{between 0 \& 1} & \end{cases}$$

This rule

① it maximizes the $P(H = \hat{H})$

② it is known as the "MAP" rule



a posteriori probability
maximum.

Example :

$$H = \begin{pmatrix} 0 & 1-p \\ 1 & p \end{pmatrix} \quad \gamma = \begin{pmatrix} \alpha \\ \beta \end{pmatrix}$$

$$Pr(\gamma = \alpha | H = 0) = \underline{\underline{0.9}}, \quad Pr(\gamma = \beta | H = 0) = \underline{\underline{0.1}}$$

$$Pr(Y = \alpha | I = 1) = 0.2, \quad Pr(Y = \beta | I = 1) = 0.8$$

Conditional on $Y = \alpha$, what is

$$\Pr(H=0 | Y=\alpha) = (1-p) \Pr(Y=\alpha | H=0) / \Pr(Y=\alpha)$$

$\swarrow 0.9$

(also note $\Pr(Y=\alpha) = \Pr(H=0) \Pr(Y=\alpha | H=0) + \Pr(H=1) \Pr(Y=\alpha | H=1)$)

$$= (1-p)(0.9) + p(0.2)$$

$$\Pr(H=1 | Y=\alpha) = p \cdot \underbrace{\Pr(Y=\alpha | H=0)}_{0.2} / \Pr(Y=\alpha)$$

Decision when $Y = \alpha$:

$$\Pr(H=0 | Y=\alpha) \quad \text{vs} \quad \Pr(H=1 | Y=\alpha)$$

$$\equiv (1-p)(0.9) \quad \text{vs} \quad p(0.2)$$

$$\equiv \frac{0.9}{0.2} \quad \begin{array}{c} \swarrow \text{decide } \hat{H}=0 \\ \downarrow p \\ \downarrow 1-p \\ \swarrow \text{decide } \hat{H}=1 \end{array}$$

$$\equiv \frac{\Pr(Y=\alpha | H=0)}{\Pr(Y=\alpha | H=1)} \quad \begin{array}{c} \hat{H}=0 \\ \swarrow \\ \searrow \\ \hat{H}=1 \end{array} \quad \frac{\Pr(H=1)}{\Pr(H=0)}$$

{ "ML" rule
↑ maximum likelihood

$$\frac{\Pr(Y=\alpha | H=0)}{\Pr(Y=\alpha | H=1)}$$

$$\begin{array}{c} \hat{H}=0 \\ > 1 \\ < \\ \hat{H}=1 \end{array}$$

in general sub optimal,
but is ML
if $\Pr(H=0) = \Pr(H=1)$.