

Last lecture

Markov and Chebychev inequalities ([Ch 2.9](#))

- Chebychev inequality
- Confidence interval

Binary Hypothesis Testing ([Ch 2.11](#))

- Definition
- Likelihood table

Correction

Midterm # 1 range until

HWK # 4 \Rightarrow Poisson

NO MLE / confidence interval.

Agenda

Binary Hypothesis Testing (Ch 2.11)

- Likelihood table
- Maximum likelihood decision rule
- Maximum A Posteriori

Maximum Likelihood Table

- Table showing *likelihood* of two hypotheses

$\begin{cases} \rightarrow H_1 \triangleq \text{has tumor} \\ \rightarrow H_0 \triangleq \text{has } X \text{ tumors.} \end{cases}$

	$X = 0$	$X = 1$	$X = 2$	$X = 3$
$P(X H_1) \rightarrow H_1$	0	<u>0.1</u>	<u>0.3</u>	<u>0.6</u>
$P(X H_0) \rightarrow H_0$	<u>0.4</u>	0.3	0.2	0.1

$1 = \sum P(X | H_i)$
 $\Rightarrow \Sigma = 1.$

If Alice has tumor, $P\{\text{Alice ultrasound scan has 2 circles} | H_1\}$

- Decision rule can be shown on the table

by underlining a cell per column

False alarm and missing

Metrics for rules

	$X = 0$	$X = 1$	$X = 2$	$X = 3$
$\Rightarrow H_1$	0	0.1	0.3	0.6
$\Rightarrow H_0$	0.4	0.3	0.2	0.1

- $P_{\text{false alarm}} = P\{\text{Claim } H_1 | H_0\} = 0.3 + 0.2 + 0.1 = 0.6$
(Type I error)
- $P_{\text{miss}} = P\{\text{Claim } H_0 | \underline{H_1}\} = 0$
(Type II error)

Another policy

$$P_{\text{false alarm}} = 0.2 + 0.1 = 0.3$$

$$P_{\text{miss}} = 0 + 0.1 = 0.1$$

	$X = 0$	$X = 1$	$X = 2$	$X = 3$
H_1	0	0.1	0.3	0.6
H_0	0.4	0.3	0.2	0.1

Maximum Likelihood (ML) decision rule

Pick whichever is higher per column!

	$X = 0$	$X = 1$	$X = 2$	$X = 3$
H_1	$p_1(0) = 0$	0.1	<u>0.3</u>	<u>0.6</u>
H_0	<u>0.4</u>	<u>0.3</u>	0.2	0.1

Likelihood Ratio Test (LRT) $\Lambda(k) = \frac{p_1(k)}{p_0(k)}$

A LRT with threshold τ :

$$\begin{cases} H_1 & \text{if } \Lambda(k) > \tau \\ H_0 & \text{else} \end{cases}$$

$\rightarrow \text{MLE} =$
LRT w/ $\tau = 1$

Maximum Likelihood (ML) decision rule

Pick whichever is higher per column!

? $P(H_1) \Rightarrow$

	$X = 0$	$X = 1$	$X = 2$	$X = 3$
H_1	0	0.1	<u>0.3</u>	<u>0.6</u>
H_0	<u>0.4</u>	<u>0.3</u>	0.2	0.1

What's the problem?

\Rightarrow If $P_{\text{false alarm}} \uparrow$ $P(H_0) \uparrow$

$\downarrow P(\text{claim } H_1 | H_0) \Rightarrow \text{large (Type I) error count}$

$P_{\text{miss}} \uparrow$ $P(H_1) \uparrow$

Slidos – Midterm Review and Early Feedback

Midterm Review

- This Friday class time
- Vote for the contents!



#2124287

Early Feedback

- Suggest improvement
- Anonymous by default



#9021295

Maximum A Posteriori (MAP) decision rule

Let's pick based on $P(H|X)$ or joint probability $P(H, X)$

- Pick the higher $P(H, X)$ per column
- Lowest total error rate $p_e = \frac{\text{Total error}}{\text{Total experiments (patients)}} = \sum \bigcirc$
- But how do we get the joint probability?
- Recall conditional probability

$$P(A, B) = P(B|A) P(A)$$

likelihood table

$$P(\underbrace{H_1}_A, \underbrace{X=k}_B) = \underbrace{P(X=k|H_1)}_{\text{likelihood}} \times \underbrace{P(H_1)}_{\text{Prior}}$$

$$\sum = 1$$

$P(H, X)$	$X = 0$	$X = 1$	$X = 2$	$X = 3$
H_1	<u>0</u>	<u>0.02</u>	<u>0.06</u>	<u>0.12</u>
H_0	<u>0.32</u>	<u>0.24</u>	<u>0.16</u>	<u>0.08</u>

$$P(H_1) = 0.2 \quad P(H_0) = 0.8$$

Prior and Posterior

$$P(H_1, X = k) = P(X = k|H_1)P(H_1)$$

- $P(H_1) \triangleq \pi_1$: “prior probability”, probability assumed before observation X .
- $P(H_0) \triangleq \pi_0$
- Bayes’ rule revisited (Param/Hypothesis vs. Observation)

$$P(H|X) = \frac{P(X|H)P(H)}{P(X)}$$

likelihood

prior

evidence

Terminologies

Exam Safe

$$P(\theta|X) = \frac{P(X|\theta)P(\theta)}{P(X)}$$

Likelihood	$P(X \theta)$	Tractable, well defined	Toss $p = 0.3$ coin
Posterior	$P(\theta X)$	Typical goal	$x_{1:3} = \{H, T, T\}, p = ?$
Prior	$P(\theta)$	Domain knowledge	# coins $p = 0.3$ in world
Evidence	$P(X)$	Approx. with large data	# H in world

Likelihood table to joint probability

Assume $\pi_1 = P(H_1) = 0.2$, $\pi_0 = P(H_0) = 1 - \pi_1 = 0.8$

- Decide on joint probability is same as posterior probability

- MAP rule = LRT rule with $\tau = \frac{\pi_0}{\pi_1} = \frac{0.8}{0.2} = 4$

$$P(X|H_1)P(H_1) > \tau P(X|H_0)P(H_0)$$

Handwritten notes: $P(X, H)$ above the first term, $P(H|X)$ above the second term, and $P(H_0)$ below the threshold τ .

\wedge 0 0.33 1.5 6

$P(X H)$	$X = 0$	$X = 1$	$X = 2$	$X = 3$
H_1	0	<u>0.1</u>	<u>0.3</u>	<u>0.6</u>
H_0	<u>0.4</u>	<u>0.3</u>	0.2	0.1

$P(H, X)$	$X = 0$	$X = 1$	$X = 2$	$X = 3$
H_1	0	0.02	0.06	<u>0.12</u>
H_0	<u>0.32</u>	<u>0.24</u>	<u>0.16</u>	0.08