

- ① Suppose a coin is either: $H_1 \rightarrow$ biased with heads prob. $\frac{2}{3}$
 $H_0 \rightarrow$ fair.

Measurement is five flips of coin. Let X be # times heads.

Describe ML and MAP decision rules, find P_F , P_M , and P_E for each using prior $(\pi_0, \pi_1) = (0.2, 0.8)$.

Likelihood matrix for each row is binomial:

	$X=0$	$X=1$	$X=2$	$X=3$	$X=4$	$X=5$
H_1	$(\frac{1}{3})^5$	$5(\frac{2}{3})(\frac{1}{3})^4$	$10(\frac{2}{3})^2(\frac{1}{3})^3$	$10(\frac{2}{3})^3(\frac{1}{3})^2$	$5(\frac{2}{3})^4(\frac{1}{3})$	$(\frac{2}{3})^5$
H_0	$(\frac{1}{2})^5$	$5(\frac{1}{2})^5$	$10(\frac{1}{2})^5$	$10(\frac{1}{2})^5$	$5(\frac{1}{2})^5$	$(\frac{1}{2})^5$

in computing Likelihood ratio, binomial coefficients cancel:

$$\Lambda(k) = \frac{\binom{5}{k} (\frac{2}{3})^k (\frac{1}{3})^{5-k}}{\binom{5}{k} (\frac{1}{2})^5} = 2^k \left(\frac{2}{3}\right)^5 \approx \frac{2^k}{7.6}$$

ML rule is to declare H_1 whenever $\Lambda(x) \geq 1$, or equiv. $X \geq 3$.

For ML rule:

$$P_F = 10\left(\frac{1}{2}\right)^5 + 5\left(\frac{1}{2}\right)^5 + \left(\frac{1}{2}\right)^5 = \frac{1}{2}$$

$$P_M = \left(\frac{1}{3}\right)^5 + 5\left(\frac{2}{3}\right)\left(\frac{1}{3}\right)^4 + 10\left(\frac{2}{3}\right)^2\left(\frac{1}{3}\right)^3 = \frac{51}{243} \approx 0.201$$

$$P_E = \pi_0 P_F + \pi_1 P_M = \frac{1}{5} \cdot \frac{1}{2} + \frac{4}{5} (0.201) \approx 0.26$$

MAP rule is to decide H_1 when $\Lambda(x) \geq \frac{\pi_0}{\pi_1} = \frac{1}{4}$, or equiv. $X \geq 1$.

so MAP rule declares H_0 only if $X=0$.

$$P_F = 1 - \left(\frac{1}{2}\right)^5 \approx 0.97$$

$$P_M = \left(\frac{1}{3}\right)^5 = \frac{1}{243} \approx 0.041$$

$$P_E = \frac{1}{5} P_F + \frac{4}{5} P_M \approx 0.227$$

As expected P_E for MAP is smaller than for ML.

- ② Detection with Poisson observations: A deep space transmitter uses laser-based binary signaling. If zero, # photons X is Poisson ($\lambda_0=2$). If one, # photons X is Poisson ($\lambda_1=6$). Based on X , decide if 0 or 1 transmitter. Find ML, MAP, with $\pi_0/\pi_1=5$. Express in terms of X .

ML, MAP rules require $\Lambda(X)$:

$$\Lambda(k) = \frac{P(X=k | \text{one sent})}{P(X=k | \text{zero sent})} = \frac{e^{-\lambda_1} \lambda_1^k / k!}{e^{-\lambda_0} \lambda_0^k / k!} = \left(\frac{\lambda_1}{\lambda_0}\right)^k e^{-(\lambda_1 - \lambda_0)} = 3^k e^{-4} \frac{3^k}{54.6}$$

So ML rule is to decide 'one' if $X \geq 4$, since want $\Lambda(X) \geq 1$.

MAP rule is $\Lambda(k) \geq \frac{\pi_0}{\pi_1} = 5$, i.e. $\frac{3^k}{54.6} \geq 5$ or $X \geq 6$.

- ③ Sensor Fusion Two motion detectors are used to detect the presence of a person in a room, as part of an energy-saving device. First sensor outputs X and second Y . Both have possible outputs $\{0,1,2\}$, with larger numbers indicating person is present.

Let H_0 : absent
 H_1 : present.

Likelihoods as:

	$X=0$	$X=1$	$X=2$
H_1	0.1	0.3	0.6
H_0	0.8	0.1	0.1

	$Y=0$	$Y=1$	$Y=2$
H_1	0.1	0.1	0.8
H_0	0.7	0.2	0.1

Assume sensors provide conditionally independent readings given hypothesis.

- Find likelihood matrix for joint observation (X, Y) , and find ML rule. break ties in favor of H_1 .
- find P_E, P_M , for ML rule
- Let $\pi_1=0.2, \pi_0=0.8$. Compute joint prob. matrix and find MAP rule.

(a) for MAP rule, compute P_f , P_m , $P_c = \pi_0 P_f + \pi_1 P_m$.

(c) compute P_c for ML rule.

(a) likelihood matrix (X, Y) is

(X, Y)	(0,0)	(0,1)	(0,2)	(1,0)	(1,1)	(1,2)	(2,0)	(2,1)	(2,2)
H_1	0.01	0.01	<u>0.08</u>	0.03	<u>0.03</u>	0.24	0.06	<u>0.06</u>	<u>0.48</u>
H_0	<u>0.56</u>	<u>0.16</u>	0.08	<u>0.07</u>	0.02	0.01	<u>0.07</u>	0.02	0.01

ML rule by underlining.

(b)

P_m is sum of entries in H_1 row not underlined:

$$P_m = 0.01 + 0.01 + 0.03 + 0.06 = 0.11$$

P_f is sum of entries in H_0 row not underlined:

$$P_f = 0.08 + 0.02 + 0.01 + 0.02 + 0.01 = 0.14$$

(c) joint prob matrix:

(X, Y)	(0,0)	(0,1)	(0,2)	(1,0)	(1,1)	(1,2)	(2,0)	(2,1)	(2,2)
H_1	0.002	0.002	0.016	0.006	0.006	<u>0.048</u>	0.012	0.012	<u>0.096</u>
H_0	<u>0.448</u>	<u>0.128</u>	<u>0.064</u>	<u>0.056</u>	<u>0.016</u>	0.008	<u>0.056</u>	<u>0.016</u>	0.008

MAP rule by underlining.

$$(d) P_f = \Pr[(X, Y) \in \{(1,2), (2,2)\} / H_0] = 0.01 + 0.01 = 0.02$$

$$P_m = \Pr[(X, Y) \notin \{(1,2), (2,2)\} / H_1] = 1 - \Pr[(X, Y) \in \{(1,2), (2,2)\} / H_1]$$

$$= 1 - 0.24 - 0.48 = 0.28$$

$$\text{so } P_c = (0.8)(0.02) + (0.2)(0.28) = 0.072$$

(also sum of not underlined entries in matrix).

(c) for ML,

$$P_c = (0.8)(0.14) + (0.2)(0.11) = 0.134.$$

of course max than error for ML.
