

Lecture 07

Suppose that our sample space contains
2 partitions $\{A_0, A_1\}$



e.g. A_0 : fair coin

A_1 : 2-headed coin

B : event 8 heads in 8 flips

Optimal Decision Rules

Maximum Likelihood Estimation (MLE)

$$\underset{\substack{\uparrow \\ \text{data likelihood}}}{P(B|A_0)} \underset{A_1}{\overset{A_0}{>}} P(B|A_1)$$

\hookrightarrow If $P(B|A_0) > P(B|A_1)$, then
we "choose" / decide A_0 .

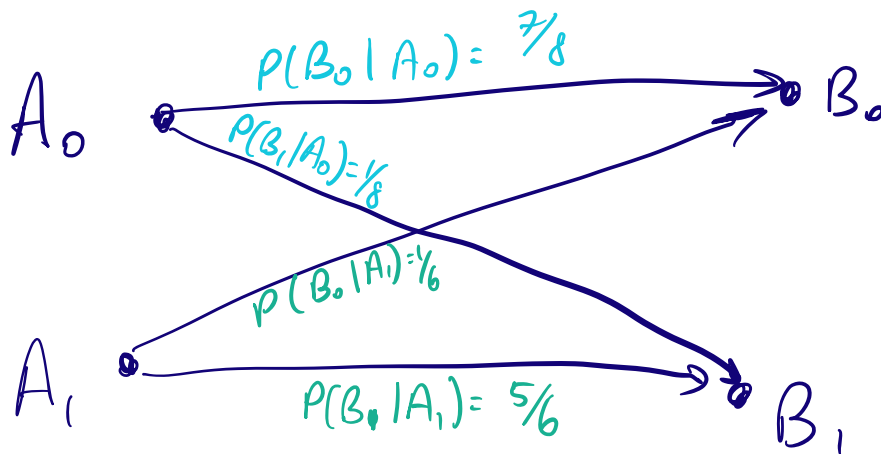
Maximum A Posteriori (MAP)

$$P(A_0 | B) \underset{A_1}{\overset{A_0}{\geq}} P(A_1 | B)$$

\Rightarrow Bayes $\frac{P(B|A_0)P(A_0)}{P(B)} \underset{A_1}{\overset{A_0}{\geq}} \frac{P(B|A_1)P(A_1)}{P(B)}$

Example: Binary Communication System.

- Transmitter sends A_0, A_1
- Receiver receives B_0, B_1



Goal: choose transmitter that maximizes some probability: "Probability of transmitter A_j given that we received B_i ":

MAP! — A_j given that we received B_i :

$$P(A_j | B_i), \forall i, j = \{0, 1\}$$

e.g. if we receive B_0 & observe $P(A_0 | B_0) > P(A_1 | B_0)$, we will decide A_0 .

But! We were not given $P(A_j | B_i)$!

$$\begin{aligned} \text{Bayes: } P(A_j | B_i) &= \frac{P(A_j \cap B_i)}{P(B_i)} \\ &= \frac{P(B_i | A_j) P(A_j)}{P(B_i)} \\ &= \frac{P(B_i | A_j) P(A_j)}{\sum_k P(B_i | A_k) P(A_k)} \end{aligned}$$

Scenario 1:

$$P(A_0) = 2/5, P(A_1) = 3/5 :$$

$$P(A_0 | B_0) \stackrel{A_0}{>} \underset{A_1}{P(A_1 | B_0)}$$

$$\frac{P(B_0 | A_0) P(A_0)}{P(B_0)} \stackrel{A_0}{>} \underset{A_1}{\frac{P(B_0 | A_1) P(A_1)}{P(B_0)}}$$

$$\frac{7/8 \cdot 2/5}{9/20} \stackrel{A_0}{>} \underset{A_1}{\frac{1/6 \cdot 3/5}{9/20}}$$

$$\frac{7}{9} \underset{A_1}{\overset{A_0}{>}} \frac{2}{9} \Rightarrow \text{Decide } A_0!$$

So, when B_0 is received, MAP decision rule is to decide A_0

Similarly, when B_1 is received, MAP rule is to decide A_1 .

If the decision rule is to always decide A_1 , then we will have an error with some probability:

$$P(E) = P(E|B_0)P(B_0) + P(E|B_1)P(B_1)$$

Law of total prob.

$$P(B_0) = P(B_0|A_0)P(A_0) + P(B_0|A_1)P(A_1)$$

$$= \frac{7}{8} \cdot \frac{2}{5} + \frac{1}{6} \cdot \frac{3}{5} = \frac{9}{20}$$

$$P(B_1) = P(B_1|A_0)P(A_0) + P(B_1|A_1)P(A_1)$$

$$= \frac{1}{8} \cdot \frac{2}{5} + \frac{5}{6} \cdot \frac{3}{5} = 1 - P(B_0) = \frac{11}{20}$$

For this decision rule, ^{always H_1} we have:

$$P(E) = \frac{7}{9} \cdot \frac{9}{20} + \left(1 - \frac{10}{11}\right) \cdot \frac{11}{20} = 0.4$$

vs.

MAP Decision rule:

$$\frac{2}{9} \cdot \frac{9}{20} + \left(1 - \frac{10}{11}\right) \cdot \frac{11}{20} = 0.15$$