Bayesian decision making

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Today two examples:

- 1. Bayesian decision making basics
- 2. Prior probabilities in practice

A:
$$P(X = x_i) = \sum_j \frac{P(X = x_i, Y = y_j)}{P(Y = y_j)}$$

B: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$
C: $P(X = x_i) = \sum_i P(X = x_i, Y = y_j)$
D: $P(Y = y_i) = \sum_j P(X = x_i, Y = y_j)$

What is correct?

A: $P(X = x_i) = \sum_j \frac{P(X = x_i, Y = y_j)}{P(Y = y_j)}$ B: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$ C: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$ D: $P(Y = y_i) = \sum_j P(X = x_i, Y = y_j)$

• Sum rule of probability:
$$P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$$

A:
$$P(X = x_i | Y = y_j) = P(Y = y_j, X = x_i)P(X = x_i)$$

B: $P(X = x_i, Y = y_j) = \frac{P(Y = y_i | X = x_i)}{P(X = x_i)}$
C: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(Y = y_i)$
D: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

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A:
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C: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(Y = y_i)$
D: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

Sum rule of probability:
$$P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$$

Product rule of probability:
$$P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$$

A:
$$P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$$

B: $P(Y = y_i, X = x_j) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_j)}$
C: $P(Y = y_i | X = x_j) = P(X = x_i | Y = y_j)P(Y = y_j)$
D: $P(Y = y_i | X = x_j) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{\sum_i P(X = x_i, Y = y_j)}$

Sum rule of probability:
$$P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$$

▶ Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

What is correct?

A: $P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$ B: $P(Y = y_i, X = x_j) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_j)}$ C: $P(Y = y_i | X = x_j) = P(X = x_i | Y = y_j)P(Y = y_j)$ D: $P(Y = y_i | X = x_j) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{\sum_{i} P(X = x_i, Y = y_j)}$

• Sum rule of probability:
$$P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$$

▶ Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

Bayes' theorem:
$$P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_i)}{P(X = x_i)}$$

What is correct?

A: $\delta^* = \arg \max_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x, s)$ B: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x|s)$ C: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x, s)$ D: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} l(s, x) P(x, s)$

- Sum rule of probability: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$
- ▶ Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

Bayes' theorem:
$$P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$$

- A: $\delta^* = \arg \max_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x, s)$
- **B**: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x|s)$
- C: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} I(s, \delta(x)) P(x, s)$
- **D**: $\delta^* = \arg\min_{\delta} \sum_{x} \sum_{s} l(s, x) P(x, s)$

• Sum rule of probability: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$

• Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

Bayes' theorem:
$$P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$$

▶ Bayes optimal strategy: $\delta^* = \arg \min_{\delta} \sum_x \sum_s I(s, \delta(x)) P(x, s)$

What is correct?

A: $\delta^*(x) = \arg \min_d \sum_s l(s, d) P(s|x)$ B: $\delta^*(x) = \arg \min_d \sum_s l(s, d) P(s, x)$ C: $\delta^*(x) = \arg \min_\delta \sum_x \sum_s l(s, \delta(x)) P(x, s)$ D: $\delta^*(x) = \arg \min_s \sum_d l(s, d) P(s|x)$

• Sum rule of probability: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$

• Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

Bayes' theorem:
$$P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_i)}{P(X = x_i)}$$

▶ Bayes optimal strategy: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x, s)$

What is correct?

A: $\delta^*(x) = \arg\min_d \sum_s l(s, d) P(s|x)$

- **B**: $\delta^*(x) = \arg \min_d \sum_s l(s, d) P(s, x)$
- **C**: $\delta^*(x) = \arg\min_{\delta} \sum_{x} \sum_{s} l(s, \delta(x)) P(x, s)$
- **D**: $\delta^*(x) = \arg\min_s \sum_d I(s, d) P(s|x)$

Sum rule of probability: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$

• Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

► Bayes' theorem: $P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$

▶ Bayes optimal strategy: $\delta^* = \arg \min_{\delta} \sum_x \sum_s I(s, \delta(x)) P(x, s)$

• BOS solution:
$$\delta^*(x) = \arg \min_d \sum_s l(s, d) P(s|x)$$

Assume l(s, d) = 1, if $d \neq s$, l(s, d) = 0 otherwise. What is correct?

A:
$$\delta^*(x) = \arg\min_d P(d|x)$$

- B: $\delta^*(x) = \arg \max_d P(d|x)$
- C: $\delta^*(x) = \arg \max_d P(d|x)P(x)$
- D: $\delta^*(x) = \arg \max_d P(d|x)P(s)$

Sum rule of probability: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$

• Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

► Bayes' theorem: $P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$

▶ Bayes optimal strategy: $\delta^* = \arg \min_{\delta} \sum_x \sum_s I(s, \delta(x)) P(x, s)$

• BOS solution:
$$\delta^*(x) = \arg \min_d \sum_s l(s, d) P(s|x)$$

Assume l(s, d) = 1, if $d \neq s$, l(s, d) = 0 otherwise. What is correct?

A: $\delta^*(x) = \arg\min_d P(d|x)$

B: $\delta^*(x) = \arg \max_d P(d|x)$

C: $\delta^*(x) = \arg \max_d P(d|x)P(x)$

D: $\delta^*(x) = \arg \max_d P(d|x)P(s)$

• Sum rule of probability: $P(X = x_i) = \sum_j P(X = x_i, Y = y_j)$

• Product rule of probability: $P(X = x_i, Y = y_j) = P(Y = y_j | X = x_i)P(X = x_i)$

Bayes' theorem:
$$P(Y = y_j | X = x_i) = \frac{P(X = x_i | Y = y_j)P(Y = y_j)}{P(X = x_i)}$$

- ▶ Bayes optimal strategy: $\delta^* = \arg \min_{\delta} \sum_{x} \sum_{s} I(s, \delta(x)) P(x, s)$
- ► BOS solution: $\delta^*(x) = \arg \min_d \sum_s l(s, d) P(s|x)$
- $L_{0,1}$ classification: $\delta^*(x) = \arg \max_d P(d|x)$

The probability distribution of height of men and women is known (see table).

_	x cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
-	P(x male)	0.05	0.15	0.2	0.25	0.3	0.05	1
	P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female.

A: Male

B: Female

The probability distribution of height of men and women is known (see table).

x 	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
P(x male)	0.05	0.15	0.2	0.25	0.3	0.05	1
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Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female.

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P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female.

A: Male

B: Female (if we assume that there are same the number of men and women.)

The probability distribution of height of men and women is known (see table).

x cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
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Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female. **Female** Task 2: What would be the minimum ratio of men to change the previous answer into the opposite one.

Right step?

A:
$$P(X = male, Y = L) = P(X = female, Y = L)$$

B:
$$P(X = male|Y = L) = P(X = female|Y = L)$$

C:
$$P(X = male | Y > L) = P(X = female | Y < L)$$

D: P(X = male|Y > L) > P(X = female|Y < L)

The probability distribution of height of men and women is known (see table).

x cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
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C: P(X = male | Y > L) = P(X = female | Y < L)

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The probability distribution of height of men and women is known (see table).

cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
P(x male)	0.05	0.15	0.2	0.25	0.3	0.05	1
P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female. **Female** Task 2: What would be the minimum ratio of men to change the previous answer into the opposite one. P(X = male|Y = L) = P(X = female|Y = L)

From the equation get value of?

- A: P(X = male)
- B: P(X = male | Y = L)
- C: P(X = female | Y < L)
- D: P(X = male | Y > L)

The probability distribution of height of men and women is known (see table).

cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
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Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female. Female Task 2: What would be the minimum ratio of men to change the previous answer into the opposite one. P(X = male|Y = L) = P(X = female|Y = L)

From the equation get value of?

- A: P(X = male)
- **B**: P(X = male | Y = L)
- C: P(X = female | Y < L)
- **D:** P(X = male|Y > L)

The probability distribution of height of men and women is known (see table).

x cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
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Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female. Female

Task 2: What would be the minimum ratio of men to change the previous answer into the opposite one. P(X = male|Y = L) = P(X = female|Y = L)

From the equation get value of? P(X = male)Calculate P(X = male):

A: $\frac{5}{11}$ B: $\frac{6}{11}$ C: $\frac{6}{10}$ D: $\frac{7}{12}$

The probability distribution of height of men and women is known (see table).

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Task 1: Estimate whether a 168cm tall (i.e. L) person is male or female. Female Task 2: What would be the minimum ratio of men to change the previous answer into the opposite one. P(X = male|Y = L) = P(X = female|Y = L)

From the equation get value of? P(X = male)Calculate P(X = male):

B: $\frac{6}{11}$

$$\begin{array}{l} P(X = male|Y = L) = P(X = female|Y = L) \\ \frac{P(L|male) \cdot P(male)}{P(L)} = \frac{P(L|female) \cdot P(female)}{P(L)}, P(female) = 1 - P(male) \\ P(L|male) \cdot P(male) = P(L|female) \cdot (1 - P(male)) \\ 0.25 \cdot P(male) = 0.3 - 0.3 \cdot P(male) \Rightarrow P(male) = \frac{6}{11} \end{array}$$

The probability distribution of height of men and women is known (see table).

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Task 3: Assuming there are 70% men and 30% women, consider the loss function I (s - state, d - decision): I(s = female, d = male) = 2, I(s = male, d = female) = 1, I(s = male, d = male) = I(s = female, d = female) = 0. How do you classify a person under consideration of L? How?

A:
$$\delta^*(X = L) = \operatorname{argmin}_s \sum_s l(s, d) \cdot P(s|X = L)$$

B: $\delta^*(X = L) = \operatorname{argmin}_d l(s, d) \cdot P(s|X = L)$
C: $\delta^*(X = L) = \operatorname{argmin}_s \sum_d l(s, d) \cdot P(s|X = L)$
D: $\delta^*(X = L) = \operatorname{argmin}_d \sum_s l(s, d) \cdot P(s|X = L)$

The probability distribution of height of men and women is known (see table).

× cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
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P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 3: Assuming there are 70% men and 30% women, consider the loss function I (s = state, d = decision): I(s = female, d = male) = 2, I(s = male, d = female) = 1, I(s = male, d = male) = I(s = female, d = female) = 0. How do you classify a person under consideration of L? How?

D:
$$\delta^*(X = L) = \operatorname{argmin}_d \sum_s l(s, d) \cdot P(s|X = L)$$

The probability distribution of height of men and women is known (see table).

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P(x male)	0.05	0.15	0.2	0.25	0.3	0.05	1
P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 3: Assuming there are 70% men and 30% women, consider the loss function I (s = state, d = decision): I(s = female, d = male) = 2, I(s = male, d = female) = 1, I(s = male, d = male) = I(s = female, d = female) = 0. How do you classify a person under consideration of L? How? $\delta^*(X = L) = \operatorname{argmin}_d \sum_s I(s, d) \cdot P(s|X = L)$ Result?

- A: female
- B: male

The probability distribution of height of men and women is known (see table).

x cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
P(x male)	0.05	0.15	0.2	0.25	0.3	0.05	1
P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 3: Assuming there are 70% men and 30% women, consider the loss function I (s = state, d = decision): I(s = female, d = male) = 2, I(s = male, d = female) = 1, I(s = male, d = male) = I(s = female, d = female) = 0. How do you classify a person under consideration of L? How? $\delta^*(X = L) = \operatorname{argmin}_d \sum_s I(s, d) \cdot P(s|X = L)$ Result?

A: female

B: male

The probability distribution of height of men and women is known (see table).

x cm	XS (0–100)	S (100–125)	M (125–150)	L (150–175)	XL (175–200)	XXL (200-∞)	\sum
P(x male)	0.05	0.15	0.2	0.25	0.3	0.05	1
P(x female)	0.05	0.1	0.3	0.3	0.25	0.0	1

Task 3: Assuming there are 70% men and 30% women, consider the loss function l (s = state, d = decision): l(s = female, d = male) = 2, l(s = male, d = female) = 1, l(s = male, d = male) = l, l(s = male, d = male) = 0. How do you classify a person under consideration of L? $P(male|L) = \frac{P(L|male) \cdot P(male)}{P(L)} = \frac{P(L|male) \cdot P(male)}{P(L|male) \cdot P(male) + P(L|female) \cdot P(female)} = \frac{0.25 \cdot 0.7}{0.25 \cdot 0.7 + 0.3 \cdot 0.3} = 0.66$ P(female|L) = 1 - 0.66 = 0.34 $\delta^*(X) = \operatorname{argmin}_d(l(female, d) \cdot P(female|L) + l(male, d) \cdot P(male|L))$

$$\delta^*(X) = \operatorname{argmin}_d \left\{ \begin{array}{l} d = \textit{female} : 0 \cdot 0.34 + 1 \cdot 0.66 = 0.66 \\ d = \textit{male} : 2 \cdot 0.34 + 0 \cdot 0.66 = 0.68 \end{array} \right\} \Rightarrow d = \textit{female}$$