

Independence

Two events are independent if the occurrence of one provides no info about the occurrence of the other.

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

$$P(A \cap B) = P(A) \cdot P(B)$$

Remarks:

1/ Disjoint \neq Independence - A & B are disjoint iff $P(A \cap B) = 0$

If 2 events are disjoint & indep, then $P(A) = 0$ or $P(B) = 0$.

2/ The base outcomes of any exp are disjoint & they have nonzero prob \Rightarrow always dependent

Ex: 1/ A: 1st die 6

B: sum of 2 dice = 7

$$P(A|B) = \frac{1}{6}, P(A) = \frac{1}{6} \Rightarrow A, B \text{ indep}$$

2/ A: 1st die 6

C: sum of 2 dice = 11

$$P(A|C) = \frac{1}{2}, P(A) = \frac{1}{6} \Rightarrow A, C \text{ dep}$$

Conditional Independence

$$P(A \cap B|C) = P(A|C) \cdot P(B|C)$$

Notes: 1/ Dependent events can be cond. indep

2/ Indep events can be cond. dep

Ex: 2 indistinguishable coins - one is two-tailed, other is two-headed

You pick one of the 2 coins at random & flip it twice.

H_i : i th flip is heads

1/ Are H_1 & H_2 indep? No

2/ If not, what's $P(H_2|H_1)$? = 1

3/ Conditioned on picking a coin, are the 2 flips indep?

A = Pick 2-headed coin

$$P(H_1 \cap H_2 | A) = P(H_1 | A) \cdot P(H_2 | H_1 \cap A)$$

$$= P(H_1 | A) \cdot P(H_2 | A) \leftarrow \text{indep}$$

Indep events can be conditionally dependent.

Ex: 2 tosses of a fair coin

H_1 : 1st toss heads

H_2 : 2nd toss heads

D: 1st \neq 2nd toss

1/ Are H_1 & H_2 indep? Yes

2/ Are H_1 & H_2 indep, given D?

$$P(H_1 \cap H_2 | D) = 0 \neq P(H_1 | D) \cdot P(H_2 | D) = \frac{1}{2} \cdot \frac{1}{2}$$

Independence of collection of events:

$$P(\bigcap_{i \in S} A_i) = \prod_{i \in S} P(A_i)$$

Pairwise indep \neq joint indep

Ex (prev): $\left. \begin{array}{l} - H_1 \& H_2 \text{ indep} \\ - H_1 \& D \text{ indep} \\ - H_2 \& D \text{ indep} \end{array} \right\} \text{ BUT } H_1, H_2, D \text{ not indep}$

Ex: Rolling a 6 before a 7 (sum)

Roll 2 dice. What's the prob of rolling a 6 before a 7?

Use indep:

- Condition on 1st roll

① Let S be the event that 1st roll is 6

② Let T be the event that 1st roll is 7

$$P(E) = \underbrace{P(E|S)}_1 \underbrace{P(S)}_{5/36} + \underbrace{P(E|T)}_0 P(T) + \underbrace{P(E|(SUT)^c)}_{P(E)} \underbrace{P((SUT)^c)}_{25/36}$$

Solve: $P(E) = \frac{5}{11}$

$\left\{ \begin{array}{l} \{(1,5), (2,4), \dots, (5,1)\} \text{ all outcomes that sum to 6} \\ \{(1,6), \dots, (6,1)\} \text{ all outcomes that sum to 7} \\ + \{(1,5), (2,4), \dots, (5,1)\} \text{ all outcomes that sum to 6} \end{array} \right\} 11 \text{ total}$

Bayes' Theorem

- A_1, \dots, A_n partition Ω

- we know $P(A_i)$

- we also know $P(B|A_i)$

- we want $P(A_i|B)$

Ex: Medical Test

A_1 = malignant tumor

A_2 = benign tumor

A_3 = other

B = test is positive

We know $P(B|A_i)$ ← accuracy of test

We want $P(A_i|B)$ ← what are the odds I have a bad tumor given positive test?

$$P(A_i|B) = \frac{P(A_i \cap B)}{P(B)} = \frac{P(B|A_i) \cdot P(A_i)}{\sum_{i=1}^n P(A_i) \cdot P(B|A_i)}$$

False Positive Text

- If person has disease, test is pos w.p. 0.95

- If person does not have disease, test is neg w.p. 0.95

- Random person has disease w.p. 0.001

← prior

A person tests pos. $P(\text{person has disease}) = ?$

A : has disease

B : tests pos

$$P(A|B) = \frac{P(B|A)P(A)}{P(A)P(B|A) + P(A^c)P(B|A^c)} = \frac{(0.001)(0.95)}{(0.001)(0.95) + (0.999)(0.05)}$$

$$= 0.0187 \text{ } (< 2\%)$$

Note: $P(A|B)$ heavily influenced by the prior $P(A)$.

Discrete RVs

From events to RVs

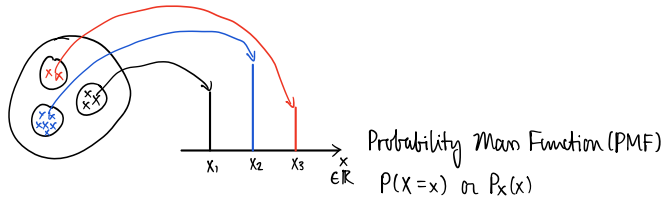
- RVs associate a real # w/ each event

Ex: ① The RV X has value i if the throw of a die is i .

② X^2 : perfectly legit RV

③ Assigning real #s allows us to do statistics

→ More formally, a RV is a func from $\Omega \rightarrow \mathbb{R}$



Ex: 2 4-sided dice

| | 1 | 2 | 3 | 4 |
|---|---|---|---|---|
| 1 | 1 | 1 | 1 | 1 |
| 2 | 1 | 2 | 2 | 2 |
| 3 | 1 | 2 | 3 | 3 |
| 4 | 1 | 2 | 3 | 4 |

M_k : Event that min of the 2 is k

M : RV that is equal to min of the 2 dice

