

Conditional Probability

①

• Many times, we want to consider the chance that an event **A** occurs under the assumption that **B** occurs.

• There are three possibilities:

- (1) The occurrence of **B** might increase the chances of **A**, so that **A** and **B** are positively correlated.
- (2) The occurrence of **B** might decrease the chances of **A**, so that **A** and **B** are negatively correlated.
- (3) The occurrence of **B** might not change the chances of **A**, so that **A** and **B** are independent.

Ex: Suppose we flip a fair coin 3 times. (2)

Our sample space is

$$\Omega = \{HHH, HHT, HTH, HTT, \\ TTH, THT, TTH, TTT\}$$

and each outcome $\omega \in \Omega$ is equally likely.

Let B be the event that the first coin flip is heads. (Recall: this means $B = \{HHH, HHT, HTH, HTT\}$.)

Also, ~~we~~ define events A_1, A_2, A_3 as follows:

$A_1 =$ majority of the flips ~~is~~ ^{are} heads

$A_2 =$ majority of the flips ~~is~~ ^{are} tails

$A_3 =$ odd number of the flips are heads.

Then A_1 and B are positively correlated

A_2 and B are negatively correlated

A_3 and B are independent.

def The probability of an event A given that an event B occurs, written $Pr(A|B)$ is defined as

$$Pr(A|B) = \frac{Pr(A \cap B)}{Pr(B)}$$

(assuming $Pr(B) > 0$.)

Think: $Pr(A|B)$ means, "If I restrict myself to considering outcomes $\omega \in \Omega$ where B occurs (i.e. $\omega \in B$), what are the odds that A also occurs?"

We divide by $Pr(B)$ to renormalize so that B and $Pr(\cdot | B)$ give a new probability space.

Ex: • $Pr(B|B) = \frac{Pr(B \cap B)}{Pr(B)} = \frac{Pr(B)}{Pr(B)} = 1$

• $Pr(A|B) + Pr(\bar{A}|B) = \frac{Pr(A \cap B)}{Pr(B)} + \frac{Pr(\bar{A} \cap B)}{Pr(B)}$
= $\frac{Pr(B)}{Pr(B)} = 1$

Let us return to our 3-coin example. (4)

$$\bullet \Pr(B) = \frac{|B|}{|\Omega|} = \frac{4}{8} = \frac{1}{2}$$

$$\bullet \Pr(A_1) = \frac{|A_1|}{|\Omega|} = \frac{4}{8} = \frac{1}{2}$$

$$\bullet \Pr(A_2) = \frac{|A_2|}{|\Omega|} = \frac{4}{8} = \frac{1}{2}$$

$$\bullet \Pr(A_3) = \frac{|A_3|}{|\Omega|} = \frac{4}{8} = \frac{1}{2}$$

$$\begin{aligned} (1) \Pr(A_1 | B) &= \frac{\Pr(A_1 \cap B)}{\Pr(B)} = \frac{|A_1 \cap B|/8}{1/2} \\ &= \frac{3/8}{1/2} = \frac{6}{8} = \frac{3}{4} \end{aligned}$$

Because $\Pr(A_1 | B) = 3/4 > 1/2 = \Pr(A_1)$, we see that A_1 and B are positively correlated.

$$\begin{aligned} (2) \Pr(A_2 | B) &= 1 - \Pr(\bar{A}_2 | B) \\ &= 1 - \Pr(A_1 | B) = 1 - 3/4 = 1/4 \end{aligned}$$

so $\Pr(A_2 | B) = 1/4 < \Pr(A_2)$ and A_2 and B are negatively correlated.

$$\begin{aligned} (3) \Pr(A_3 | B) &= \frac{\Pr(A_3 \cap B)}{\Pr(B)} = \frac{|A_3 \cap B|/8}{1/2} \\ &= \frac{2/8}{1/2} = \frac{4}{8} = \frac{1}{2} \end{aligned}$$

so $\Pr(A_3 | B) = \frac{1}{2} = \Pr(A_3)$ and A_3 and B are independent. (5)

Note: $\Pr(A|B) = \Pr(A) \iff \frac{\Pr(A \cap B)}{\Pr(B)} = \Pr(A)$

$\iff \Pr(A \cap B) = \Pr(A) \cdot \Pr(B)$

$\iff \frac{\Pr(A \cap B)}{\Pr(A)} = \Pr(B)$

$\iff \Pr(B|A) = \Pr(B)$

def Two events A and B are independent if $\Pr(A \cap B) = \Pr(A) \cdot \Pr(B)$.

A collection of events \mathcal{A} is mutually independent if, for each finite subset

$$\{A_1, A_2, \dots, A_n\} \subseteq \mathcal{A}$$

we have

$$\Pr(A_1 \cap A_2 \cap \dots \cap A_n) = \Pr(A_1) \cdot \Pr(A_2) \cdot \dots \cdot \Pr(A_n)$$

A collection of events \mathcal{A} is pairwise independent

if each pair $\{A_1, A_2\} \in \mathcal{A}$ of events in \mathcal{A} are independent. (6)

Warning:

• \mathcal{A} mutually independent \Rightarrow \mathcal{A} pairwise independent is true, but

• \mathcal{A} pairwise independent \Rightarrow \mathcal{A} mutually independent is false!

Ex: Suppose we flip a fair coin twice:

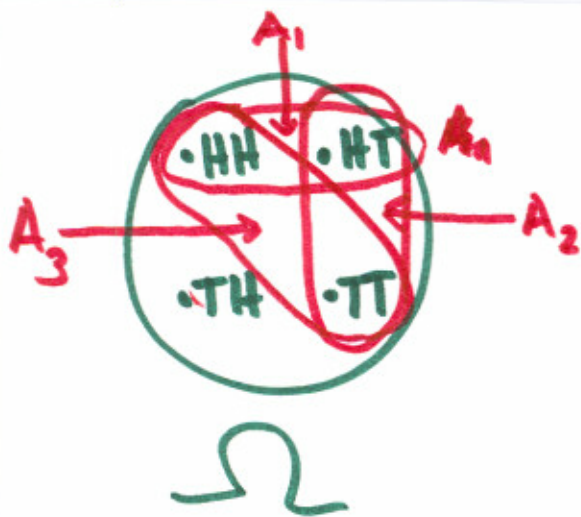
$$\Omega = \{HH, HT, TH, TT\}$$

and let

$A_1 =$ first flip is H

$A_2 =$ second flip is T

$A_3 =$ both flips the same.



Although $\mathcal{A} = \{A_1, A_2, A_3\}$ is pairwise independent, \mathcal{A} is not mutually independent.

Ex: By our claim, A_1 and A_2 are independent.

$$Pr(A_1 \cap A_2) = \frac{|A_1 \cap A_2|}{|\Omega|} = \frac{1}{4}$$

$$Pr(A_1) \cdot Pr(A_2) = \frac{|A_1|}{|\Omega|} \cdot \frac{|A_2|}{|\Omega|} = \frac{2}{4} \cdot \frac{2}{4} = \frac{1}{4} \checkmark$$

Note: \mathcal{A} is not mutually independent because the product rule does not hold for A_1, A_2, A_3 :

$$Pr(A_1 \cap A_2 \cap A_3) = \frac{|A_1 \cap A_2 \cap A_3|}{|\Omega|} = \frac{0}{4} = 0$$

$$Pr(A_1) \cdot Pr(A_2) \cdot Pr(A_3) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}$$

An Important Inequality

- What is $\Pr(A_1 \cap A_2 \cap \dots \cap A_n)$?
- If $\{A_1, \dots, A_n\}$ are mutually independent, then $\Pr(A_1 \cap A_2 \cap \dots \cap A_n) = \Pr(A_1) \cdot \Pr(A_2) \cdot \dots \cdot \Pr(A_n)$.
- If A_1, \dots, A_n are not mutually independent, we can still sometimes compute $\Pr(A_1 \cap \dots \cap A_n)$ with this formula:

Prop ~~not~~ ^{If} A_1, A_2, \dots, A_n be events, then

$$\Pr(A_1 \cap A_2 \cap \dots \cap A_n) = \Pr(A_1) \cdot \Pr(A_2 | A_1) \cdot \Pr(A_3 | A_1 \cap A_2) \cdot \Pr(A_4 | A_1 \cap A_2 \cap A_3) \cdot \dots \cdot \Pr(A_n | A_1 \cap \dots \cap A_{n-1})$$

Pf:

$$\begin{aligned} & \Pr(A_1) \cdot \Pr(A_2 | A_1) \cdot \dots \cdot \Pr(A_n | A_1 \cap \dots \cap A_{n-1}) \\ &= \Pr(A_1) \cdot \frac{\Pr(A_2 \cap A_1)}{\Pr(A_1)} \cdot \frac{\Pr(A_3 \cap A_2 \cap A_1)}{\Pr(A_1 \cap A_2)} \cdot \dots \cdot \frac{\Pr(A_n \cap A_1 \cap \dots \cap A_{n-1})}{\Pr(A_1 \cap \dots \cap A_{n-1})} \\ &= \Pr(A_1 \cap A_2 \cap \dots \cap A_n) \end{aligned}$$

Application: Suppose r balls are assigned to n bins. The balls are assigned

(-) uniformly: a ball is equally likely to be in any of the bins, and

(-) independently: the location of a ball is not influenced by the locations of other balls.

- Let A be the event that ~~no~~ ^{each} bin contains at most one ball. What is $Pr(A)$?
- Note: if $r > n$, the pigeonhole principle tells us $Pr(A) = 0$ because $A = \emptyset$.
- What if $r \leq n$?
- Let

- For each $1 \leq k \leq r$, let A_k be the event that the first k balls are placed into bins so that each bin contains at most one ball.

- Note: $A = A_r \subseteq A_{r-1} \subseteq A_{r-2} \subseteq \dots \subseteq A_2 \subseteq A_1 = \Omega$
 so $A_1 \cap A_2 \cap \dots \cap A_k = A_k$.

- Therefore

$$\begin{aligned}
 Pr(A) &= Pr(A_1 \cap \dots \cap A_r) \\
 &= Pr(A_1) \cdot Pr(A_2 | A_1) \cdot Pr(A_3 | A_1 \cap A_2) \cdot \dots \cdot Pr(A_r | A_1 \cap \dots \cap A_{r-1}) \\
 &= Pr(A_1) \cdot Pr(A_2 | A_1) \cdot Pr(A_3 | A_2) \cdot \dots \cdot Pr(A_r | A_{r-1})
 \end{aligned}$$

- What is $Pr(A_k | A_{k-1})$? We could try to evaluate this directly from the definition, but that will put us right back where we started.

• Instead, think: ~~WHAT IF~~ If A_{k-1} occurs,
(i.e. the first $k-1$ balls land in different bins),
what are the chances that A_k occurs
(i.e. the k th ball lands in an unoccupied
bin)?

• Well, the location of the k th ball is
independent of the first $k-1$ balls, and
it is equally likely to go in any bin.

There are $n - (k-1) = n - k + 1$ unoccupied bins,
so the chance that the k th ball lands
in one of the unoccupied bins is $\frac{n-k+1}{n}$.

• Therefore $P_r(A_k | A_{k-1}) = \frac{n-k+1}{n}$ and

$$\begin{aligned} P_r(A) &= P_r(A_1) \cdot P_r(A_2 | A_1) \cdot P_r(A_3 | A_2) \cdot \dots \cdot P_r(A_r | A_{r-1}) \\ &= 1 \cdot \frac{n-1}{n} \cdot \frac{n-2}{n} \cdot \dots \cdot \frac{n-r+1}{n} \end{aligned}$$

(12)

• This is a special case of what is called the birthday "paradox":

How many people do we need to gather before it is more likely than not that two have the same birthday?

- We view the calendar days as $n=365$ bins and people as balls, which are equally likely to have any date as his/her birthday.
- Plugging in $n=365$ and r into our formula, we get:

r	$\Pr(A) = \Pr(\text{no common birthdays})$
5	0.972..
10	0.880
15	0.742
20	0.581
25	0.422
30	0.284
35	0.178
40	0.103

45	0.055
50	0.027

(13)

· It turns out that when $r=23$, the probability that no two people have a common birthday is 0.484 , so it is more likely than not that there will be a common birthday if at least 23 people are gathered.



$$Pr(A|B) = \frac{Pr(A \cap B)}{Pr(B)}$$

$A \cap B$
 $\overline{A \cap B}$

$$Pr(A \cap B) + Pr(\overline{A \cap B}) = Pr((A \cap B) \cup (\overline{A \cap B})) = Pr(B)$$

