2.4: Conditional Probability

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Motivation

Sometimes the probabilities of events change with the set of available information that we have.

Example: Let $A = \{ \text{Google's share price increases tomorrow} \}$. Then $P(A) \approx .5$

But if we let $C = \{\text{the government bans access to google.com}\}$, what's P(A given C)?

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Conditional Probability

The notation: we'll write P(A given C) like P(A|C). And here's the formula:

$$P(A|C) = \frac{P(A \cap C)}{P(C)}$$

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Note

- 1 It helps to think of this in the context of venn diagrams
- ② $C \neq \emptyset$, otherwise we're dividing by 0 using this definition
- **3** sometimes we'll use $P(A|C)P(C) = P(A \cap C)$



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Example 2.26 (pg. 76)

$$P(A|B \cup C) = \frac{P(A \cap (B \cup C))}{P(B \cup C)}$$

$$= \frac{P[(A \cap B) \cup (A \cap C)]}{P(B \cup C)}$$

$$= \frac{P(A \cap B) + P(A \cap C) - P((A \cap B) \cap (A \cap C))}{P(B \cup C)}$$

$$= \frac{P(A \cap B) + P(A \cap C) - P(A \cap B \cap A \cap C)}{P(B \cup C)}$$

$$= \frac{P(A \cap B) + P(A \cap C) - P(A \cap B \cap C)}{P(B) + P(C) - P(B \cap C)}$$

$$= .255$$

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Remember we said we would use $P(A|C)P(C) = P(A \cap C)$ a lot? This is useful in situations where there is some sort of sequential thing going on, or there are several stages of some random process.

Example: you're only interested in if the stock market goes up or down. The events that it goes up on Monday, Tuesday, Wednesday, Thursday and Friday are A_1 , A_2 , A_3 , A_4 and A_5 , respectively. The changes that it goes up or goes down only depend on what happened the day before (unless it's a Monday in which case let's assume it goes up or down with probability .5). Say there's a 75% chance that it does the same thing as it did the day before. This means that there is a 25% chance it does the opposite. What's the probability the market goes up every day of the week?

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$$P(\text{goes up every day}) = P(A_5 \cap A_4 \cap \dots \cap A_1)$$

$$= P(A_5 | A_4 \cap A_3 \cap A_2 \cap A_1) P(A_4 | A_3 \cap A_2 \cap A_1) \times$$

$$P(A_3 | A_2 \cap A_1) P(A_2 | A_1) P(A_1)$$

$$= P(A_5 | A_4) P(A_4 | A_3) P(A_3 | A_2) P(A_2 | A_1) P(A_1)$$

$$= (.75)(.75)(.75)(.5)$$

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Motivation

Conditional probability is also a precursor for a thing called **Bayes' Theorem**. We need a few more ideas before we get there though...

If A_1, \ldots, A_k are all pairwise disjoint and if $\bigcup_{i=1}^k A_i = \mathcal{S}$, then

$$P(B) = \sum_{i=1}^{k} P(B|A_i)P(A_i)$$

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Proof

Notice that

$$P(B) = P(B \cap S)$$

$$= P(B \cap (\bigcup_{i=1}^{k} A_i))$$

$$= P\left(\bigcup_{i=1}^{k} (A_i \cap B)\right)$$

$$= \sum_{i=1}^{k} P(A_i \cap B)$$

$$= \sum_{i=1}^{k} P(B|A_i) P(A_i)$$

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Bayes' Rule

Here's Bayes' theorem:

$$P(A_j|B) = \frac{P(B|A_j)P(A_j)}{\sum_{i=1}^k P(B|A_i)P(A_i)}$$

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Note

A key take-away here is that there's a switch thing going on...the left hand side has probabilities conditioning on B, whereas the right hand side has probabilities conditioning on As.

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Example 2.30 pops up in STAT 2020/2120 a lot...

Let D be the event that a person has a rare disease. Let T be the event that you test positive for this disease. The bottom of page 80 gives us P(D) = .001, P(T|D) = .99 and P(T|D') = .02. What's the probability that a person has the disease if he/she is told that he/she has it?

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$$P(D|T) = \frac{P(T|D)P(D)}{P(T|D)P(D) + P(T|D')P(D')} = .047$$

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