

## OI 3.2: Conditional Probability (Part 2)

## Review

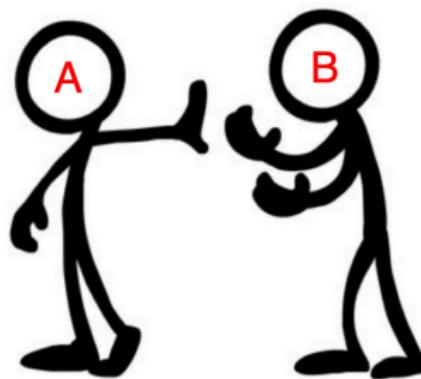
- ▶ What are some things we have already learned about for random processes?

## Worksheet

- ▶ In your worksheet we calculated the probability of an event for a random process, *given* some information about a different random process.
- ▶ Did probabilities change when you learned information about the pet that was adopted?

## Conditional Probability

- ▶  $P(A|B)$ : The probability of event  $A$  happening *given* event  $B$
- ▶ In other words: suppose we know  $B$  happens, what is the probability of  $A$ ?



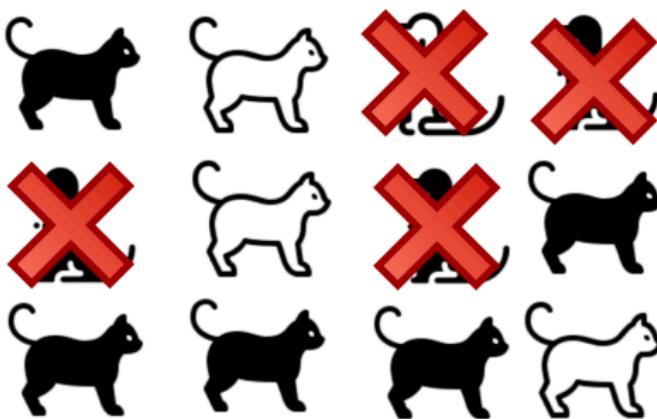
## Conditional Probability

When we have outcomes that are all equally likely (count data):

$$P(A|B) = \frac{\text{\# of occurrences of variable } A \text{ and variable } B}{\text{\# of occurrences of variable } B}$$

## Conditional Probability

Suppose you learn that the animal adopted is a cat. Given this information, what is the probability that it has a black coat?



# Conditional Probability

Suppose you learn that the animal adopted is a cat. Given this information, what is the probability that it has a black coat?

Species\Coat	Black	White	Total
Dog	45	12	57
Cat	101	55	156
Total	146	67	213

## Conditional Probability

When we have outcomes that *may not* be equally likely (probabilities/proportions):

$$P(A|B) = \frac{P(A \text{ and } B)}{P(B)}$$

The above formula is the **conditional probability rule**. This rule *always* works. The previous version of the rule only works with count data.

# Conditional Probability

Suppose you learn that the animal adopted is a cat. Given this information, what is the probability that it has a black coat?

Species\Coat	Black	White	Total
Dog	.45	.06	.51
Cat	.38	.11	.49
Total	.83	.17	1

## Conditional Probability

Lets try Questions 1d and 2d on the handout again with the **conditional probability rule**.

Do the answers match?

## Conditional Probability

- ▶ Conditional probabilities are essentially *rescaling* a probability.
- ▶ If the two variables  $A$  and  $B$  are dependent, then knowing information about  $B$  changes the probability of  $A$ .
- ▶ The notation  $A|B$  reflects this newfound knowledge.
- ▶ For example:
  - ▶ What is the probability the pet adopted has a black coat?
  - ▶ What is the probability the pet adopted has black coat, given it is a cat?
- ▶ Knowing the pet was a cat changed the information!
- ▶ All conditional probability problems can be thought of in this way.

## Multiplication Rule

- ▶ We can multiply and divide probabilities across the equal sign.
- ▶ For example, a simple rearrangement can give us the **multiplication rule**. If  $A$  and  $B$  represent two outcomes or events, then

$$P(A \text{ and } B) = P(A|B)P(B)$$

- ▶ The multiplication rule is the same thing as the conditional probability rule, this is just a rearrangement!

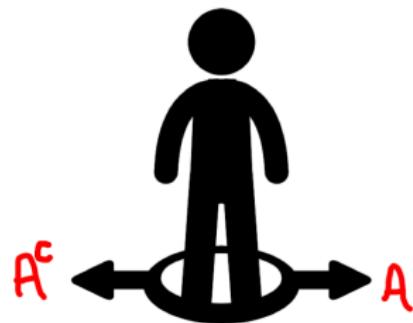
## Multiplication Rule and Independence

- ▶ We have seen a little bit of the multiplication rule before when we talked about independent events.
- ▶ Recall, we said two events  $A$  and  $B$  are independent if  $P(A \text{ and } B) = P(A)P(B)$ . This is equivalent to saying that two events are independent if  $P(A|B) = P(A)$ .

## Complement Rule with Conditional Probabilities

Recall our rule for complements:

$$1 = P(A^C) + P(A)$$

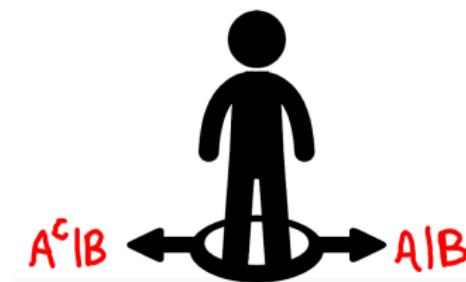


Since  $A^C$  is anything other than  $A$ , if we have  $P(A)$  we can find  $P(A^C)$ .

## Complement Rule with Conditional Probabilities

The same concept holds with conditional probabilities!

$$1 = P(A^C|B) + P(A|B)$$



There is only two options once we are given  $B$ :

$$A^C|B \quad \text{and} \quad A|B$$

Therefore, if we know  $P(A|B)$  we can find  $P(A^C|B)$  just as before!

## Example

### [OI 3.13] Joint and conditional probabilities.

Let  $P(A) = 0.3$  and  $P(B) = 0.7$ .

- a) Can you compute  $P(A \text{ and } B)$  if you only know  $P(A)$  and  $P(B)$ ?
- b) Assuming that events  $A$  and  $B$  arise from independent random processes,
  - i. what is  $P(A \text{ and } B)$ ?
  - ii. what is  $P(A \text{ or } B)$ ?
  - iii. what is  $P(A|B)$ ?
- c) If we are given that  $P(A \text{ and } B) = 0.1$ , are the random variables giving rise to events  $A$  and  $B$  independent?
- d) If we are given that  $P(A \text{ and } B) = 0.1$ , what is  $P(A|B)$ ?
- e) If  $P(A|B) = .14$ , what is  $P(A^C|B)$ ?

## Total Probability Rule

Let  $A_1, \dots, A_k$  represent all disjoint outcomes for a variable or process. Then if  $B$  is an event, possibly for another variable or process, we have

$$\begin{aligned} P(B) &= P(B \text{ and } A_1) + P(B \text{ and } A_2) + \dots + P(B \text{ and } A_k) \\ &= P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_k)P(A_k) \end{aligned}$$

We get the second line from substituting the multiplication rule.

## Total Probability Rule - Contingency Tables

We saw the total probability rule in with contingency tables!

MPAA Rating	Box Office Gross		Total
	Low	High	
Not Rated	.001	.006	.007
G	.008	.014	.022
PG	.055	.101	.156
PG-13	.101	.267	.368
R	.140	.302	.442
NC-17	.003	.002	.005
Total	.308	.692	1

This specific example:

$$P(\text{PG-13}) = P(\text{PG-13 and Low}) + P(\text{PG-13 and High})$$

Generally:

$$P(B) = P(B \text{ and } A_1) + \dots + P(B \text{ and } A_k)$$

## Example (Part 1 - Together)

Let  $D$  indicate being infected with Covid19, and  $A$  represent a positive test result. Suppose scientists have concluded that rate of Covid19 infections for adults in the US is  $P(D) = 0.01$ . The probability of correctly detecting the Covid19 when it is present in the system is  $P(A|D) = 0.95$ , and the probability Covid19 is correctly not detected is  $P(A^C|D^C) = 0.931$ . Suppose we pick someone at random from the US. We want to find the probability that someone has Covid19 given that they test positive  $P(D|A)$ .



Testing statistics reported by John Hopkins in 2020.

## Example (Part 1 - Together)

Did the answer surprise you?

Do you think Covid19 tests you take are accurate? Why or why not?

## Example (Part 1 - Together)

- ▶ We found the probability that the test correctly detects if a person has Covid19 when testing someone in the **general population**.
- ▶ When actually performing medical tests, we do not always test the general population, instead we test **those who are suspected to have the disease**.

## Example (Part 2 - In groups)

Suppose that the Covid19 infection rate among those who suspect they have Covid19 (or have been exposed to it recently) is  $P(D) = .40$ . If  $P(A|D) = 0.95$ , and  $P(A^C|D^C) = 0.931$ . What is  $P(D|A)$ ?



## A Few Notes

- ▶ The medical field often reports *sensitivity* and *specificity* for a test.
- ▶ **Sensitivity**: a test's ability to designate an individual with disease as positive

$$P(A|D)$$

- ▶ **Specificity**: a test's ability to designate an individual who does not have a disease as negative

$$P(A^C|D^C)$$

- ▶ Most people want to know if the test is accurate in general, or if the test is accurate for those at risk. Why report  $P(A|D)$  or  $P(A^C|D^C)$  instead?

## Definitions and Rules

- ▶ Mutually Exclusive:

$$P(A \text{ and } B) = 0$$

- ▶ Independence:

$$P(A \text{ and } B) = P(A)P(B)$$

- ▶ Complement:

$$P(A) = 1 - P(A^C)$$

## Definitions and Rules

- ▶ General Additivity:

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

- ▶ Conditional:

$$P(A|B) = \frac{P(A \text{ and } B)}{P(B)}$$

- ▶ General Multiplication:

$$P(A \text{ and } B) = P(A|B)P(B)$$

- ▶ Total Probability:

$$P(B) = P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_k)P(A_k)$$