

## Conditional Probability

# Math 425

## Introduction to Probability

### Lecture 8

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☞ Very few experiments amount to just one action with random outcomes.

- Sometimes conditions change before the experiment is completed.
- Some experiments have a more complicated structure, in which some partial results are known before the experiment is completed.

☞ We need a rule for recomputing probabilities when the conditions of an experiment change.

This new rule is called the **Conditioning Rule**.

## Example 1

## Example

☞ You are about to roll a **red** and **blue** die.  
What is the probability that the **red** die is larger?

$$P(\text{red larger}) = \frac{15}{36}.$$

☞ The **blue** die is first rolled and shows 5.  
What is the probability that the **red** die is larger?

$$P(\text{red larger}) = \frac{1}{6}.$$

## Example 2: Kidney stones

## Example

Kidney stones are either **small** (< 2 cm diameter) or **large** (> 2 cm diameter).  
Here are the outcomes of one treatment:

Size	Outcomes		
	success (C)	failure	total
small	315	42	357
large (L)	247	96	343
<b>total</b>	562	138	700

## Example 2: Kidney stones

☞ For a patient picked at random from the 700 patients

$$P(L) = \frac{343}{700}.$$

☞ For a patient picked at random, the probability of success is

$$P(C) = \frac{562}{700} \approx 0.8.$$

☞ For a patient picked at random, among those with large stones, the probability of success is

$$\frac{247}{343} \approx 0.72.$$

## Notation

☞ We introduce a notation to make explicit the new conditions:

$$P(C|L) \approx 0.72$$

means the probability of  $C$  (success) given  $L$  (large stones).

☞ Suppose we select a patient at random, among those with successful treatments, to determine if their stones were large.

$$P(L|C) = \frac{247}{562} \approx 0.44.$$

$P(L|C)$  is the conditional probability of  $L$  (large stones) given  $C$  (success).

## Conditional probability

### Definition

Let  $E$  and  $F$  be events.

The conditional probability of  $E$  given  $F$  is defined by:

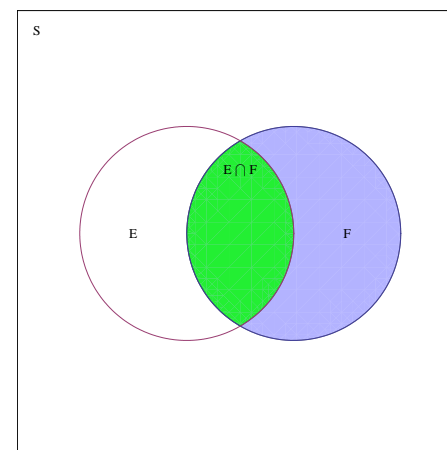
$$P(E|F) = \frac{P(E \cap F)}{P(F)}.$$

provided  $P(F) > 0$ .

## Conditional probability – picture

Conditional probability shrinks the sample space  $S$  to an event  $F$ :

$$P(E|F) = \frac{P(E \cap F)}{P(F)}$$



## Example 1 – revisited

Let  $R$  be the event that the red die is bigger, and  $B$  be the event that the blue die shows 5.

According to the rule,

$$\begin{aligned} P(R|B) &= \frac{P(R \cap B)}{P(B)} = \frac{\frac{1}{36}}{\frac{1}{6}} \\ &= \frac{1}{6} \end{aligned}$$

Note that  $R \cap B$  has one outcome: (5, 6).

## Example 2 – revisited

Size	Outcomes		
	success (C)	failure	total
small	315	42	357
large (L)	247	96	343
<b>total</b>	562	138	700

According to the rule,

$$\begin{aligned} P(C|L) &= \frac{P(C \cap L)}{P(L)} = \frac{\frac{247}{700}}{\frac{343}{700}} \\ &= \frac{247}{343} \end{aligned}$$

## Example 3

## Example

A fair coin is flipped three times. Consider the events

- $A$ : the first flip is heads.
- $B$ : there are exactly two heads overall.

Then,

$$\begin{aligned} S &= \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\} \\ A &= \{HHH, HHT, HTH, HTT\} \\ B &= \{HHT, HTH, THH\} \\ A \cap B &= \{HHT, HTH\}. \end{aligned}$$

Apply the conditioning rule to compute,

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{2}{3} \quad P(B|A) = \frac{P(A \cap B)}{P(A)} = \frac{2}{4}.$$

## Example 4

## Example

A box contains a double-headed coin, a double-tailed coin and a conventional coin. One coin is picked at random, and shows heads.

What is the probability that it is the double-headed coin?

 **Warning.** The following reasoning is incorrect:

- Since the coin shows heads, it is either double-headed or conventional. Since the coin was picked at random, these are equally likely, so the probability is  $\frac{1}{2}$ .

**This is wrong!!**

## Example 4 – continued

☞ Let the sample space be the 6 possible faces on the three coins. Consider the events

- $H$ : coin shows heads,
- $D$ : coin is double-headed.

Since 3 faces yield  $H$ ,

$$\mathbf{P}(H) = \frac{1}{2} \quad \mathbf{P}(H \cap D) = \frac{1}{3}.$$

Apply the conditioning rule,

$$\mathbf{P}(D|H) = \frac{\mathbf{P}(H \cap D)}{\mathbf{P}(H)} = \frac{2}{3}.$$

## Example 4 – Sample space

☞ Consider the sample space, where the outcomes are  $H$ ,  $T$  and  $I$  (double-tailed),  $II$  (double-headed) and  $III$  (Normal):

$$S = \{(I, T_1), (I, T_2), (II, H_1), (II, H_2), (III, H_1), (III, T_2)\}$$

All outcomes are equiprobable.

☞ After we see a heads, the sample space is **reduced**:

$$\{(I, T_1), (I, T_2), (II, H_1), (II, H_2), (III, H_1), (III, T_2)\}$$

The outcomes remaining are still equiprobable.

$$\mathbf{P}(D|H) = \frac{2}{3}.$$

## Example 4 – Simulation

☞ I ran a simulation of Example 4 in Mathematica.

- 1 Choose a coin at random and simulate a toss.
- 2 If Tails shows, go on to next trial.
- 3 If Heads shows, record which coin was chosen.
- 4 After 10,000 trials, output

$$\frac{|D \cap H|}{|H|}$$

☞ I ran the simulation 1000 times (our calculation was 0.666667):

- Mean: 0.667024 (of ratios)
- Maximum ratio: 0.688831
- Minimum ratio: 0.645387
- Standard Deviation: 0.006673

You can be 95% confident that the correct probability is within two decimal places of the mean.

## Conjunctions

☞ We can often compute  $\mathbf{P}(F)$  and  $\mathbf{P}(E|F)$  easily. If so, we can compute the conjunction  $\mathbf{P}(E \cap F)$  as well.

### Lemma (Multiplication Rule)

For any events  $E$  and  $F$  (where  $\mathbf{P}(F) > 0$ ),

$$\mathbf{P}(E \cap F) = \mathbf{P}(E) \cdot \mathbf{P}(F|E)$$

## Example 5

## Example

A drawer contains 5 red socks and 3 blue socks.

If you remove the socks at random, what is the probability of holding a blue pair?

You could do this by enumerating all the possible combinations in the sample space for this experiment, but conditioning provides a simpler method.

## Example 5 – continued

5 red and 3 blue socks.

Consider the two events

- $B_i$ : the  $i$ th sock picked is blue ( $i = 1, 2$ ).

We want to determine  $\mathbf{P}(B_1 \cap B_2)$ .

Use the multiplication rule,

$$\begin{aligned} \mathbf{P}(B_1 \cap B_2) &= \mathbf{P}(B_1) \cdot \mathbf{P}(B_2 | B_1) \\ &= \frac{3}{8} \cdot \frac{2}{7} = \frac{3}{28}. \end{aligned}$$

**Reason:** If you choose 1 blue sock then there are 2 blue socks and 7 socks remaining. So,

$$\mathbf{P}(B_2 | B_1) = \frac{2}{7}.$$

## Multiplication Rule for 3 events

We can extend the Multiplication Rule to three events.

## Lemma

For any events  $E, F, G$  (provided  $\mathbf{P}(E \cap F \cap G) > 0$ )

$$\mathbf{P}(E \cap F \cap G) = \mathbf{P}(E) \cdot \mathbf{P}(F | E) \cdot \mathbf{P}(G | E \cap F)$$

**Proof.** Use the Multiplication Rule twice,

$$\begin{aligned} \mathbf{P}(E \cap F \cap G) &= \mathbf{P}(E \cap F) \cdot \mathbf{P}(G | E \cap F) \\ &= \mathbf{P}(E) \cdot \mathbf{P}(F | E) \cdot \mathbf{P}(G | E \cap F) \end{aligned}$$

We need  $\mathbf{P}(E \cap F \cap G) \neq 0$  to ensure the conditional probabilities exist.

## Example 6

## Example

An urn is filled with 6 red balls, 5 blue balls, and 4 green balls. Three balls chosen at random are removed from the urn.

What is the probability that the balls are of the same color?

We are interested in the events (where  $i = 1, 2, 3$ )

- $R_i$ :  $i$ th ball drawn is red,
- $B_i$ :  $i$ th ball drawn is blue,
- $G_i$ :  $i$ th ball drawn is green.
- $C$ : three balls are the same color.

## Example 6 – continued

☞ Urn: 6 red, 5 blue, and 4 green. Use the Multiplication Rule,

$$\begin{aligned} \mathbf{P}(R_1 \cap R_2 \cap R_3) &= \mathbf{P}(R_1) \cdot \mathbf{P}(R_2 | R_1) \cdot \mathbf{P}(R_3 | R_1 \cap R_2) \\ &= \frac{6}{15} \cdot \frac{5}{14} \cdot \frac{4}{13} = \frac{4}{91} \end{aligned}$$

$$\begin{aligned} \mathbf{P}(B_1 \cap B_2 \cap B_3) &= \mathbf{P}(B_1) \cdot \mathbf{P}(B_2 | B_1) \cdot \mathbf{P}(B_3 | B_1 \cap B_2) \\ &= \frac{5}{15} \cdot \frac{4}{14} \cdot \frac{3}{13} = \frac{2}{91} \end{aligned}$$

$$\begin{aligned} \mathbf{P}(G_1 \cap G_2 \cap G_3) &= \mathbf{P}(G_1) \cdot \mathbf{P}(G_2 | G_1) \cdot \mathbf{P}(G_3 | G_1 \cap G_2) \\ &= \frac{4}{15} \cdot \frac{3}{14} \cdot \frac{2}{13} = \frac{4}{455} \end{aligned}$$

Since these events are mutually exclusive,

$$\mathbf{P}(C) = \frac{4}{91} + \frac{2}{91} + \frac{4}{455} = \frac{34}{455} \approx 0.0747.$$

## General multiplication Rule

☞ The Multiplication rule is the probabilistic version of the product rule for counting.

**Theorem (Generalized Multiplication Rule)**

Let  $E_1, E_2, \dots, E_n$  be any events such that

$$\mathbf{P}(E_1 \cap E_2 \cap \dots \cap E_n) > 0.$$

Then

$$\begin{aligned} \mathbf{P}(E_1 \cap E_2 \cap \dots \cap E_n) &= \mathbf{P}(E_1) \cdot \mathbf{P}(E_2 | E_1) \cdot \mathbf{P}(E_3 | E_1 \cap E_2) \cdots \\ &\quad \cdots \mathbf{P}(E_n | E_1 \cap E_2 \cap \dots \cap E_{n-1}). \end{aligned}$$

(See Ross p. 71.)

## Example 7

### Example

In Pick-Six Lottery: A person purchases a ticket, and can choose 6 distinct numbers in the set  $\{1, 2, 3, \dots, 49\}$ .

Later a Lottery Machine picks 6 distinct numbers at random in the set  $\{1, 2, 3, \dots, 49\}$ .

A winning ticket is one which matches the six numbers chosen by the Machine (in any order of selection).

☞ What are the odds of winning with one ticket?

## Example 7 – solution

**Solution.** We solved this before (Lecture 5) by counting using the Product Rule for counting. We conditionalize here.

☞ Let  $E_i$  be the event that there are  $i$  matches. We want to compute

$$\begin{aligned} \mathbf{P}(E_6) &= \mathbf{P}(E_1 \cap E_2 \cap \dots \cap E_6) \\ &= \mathbf{P}(E_1) \cdot \mathbf{P}(E_2 | E_1) \cdots \mathbf{P}(E_6 | E_1 \cap \dots \cap E_5) \\ &= \frac{6}{49} \cdot \frac{5}{48} \cdot \frac{4}{47} \cdot \frac{3}{46} \cdot \frac{2}{45} \cdot \frac{1}{44} \\ &= \frac{1}{13,983,816} \end{aligned}$$