## **Conditional Probability**

### Reading



- Malik Magdon-Ismail. Discrete Mathematics and Computing.
  - Chapter 16

#### **Overview**



- New information changes a probability
- Definition of conditional probability from regular probability
- Conditional probability traps
  - Sampling bias
  - Transposed conditional
- Law of total probability
  - Probabilistic case-by-case analysis

#### Flu Season

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- Suppose I told you that, in general, chances a random person has the flu is about 0.01 (or 1%) (*prior* probability).
- Probability of flu:

 $\mathbb{P}[flu]\approx 0.01$ 

- Suppose now that you also have a slight fever
  - This is new information. Chances of flu "increase".
- Suppose I know that the probability of flu *given* fever:  $\mathbb{P}[flu \mid fever] \approx 0.4$ 
  - New information changes the prior probability to the *posterior* probability.
  - Translate posterior as "After you get the new information."
- $\mathbb{P}[A|B]$  is the (updated) *conditional* probability of *A*, *given* the new information *B*
- Suppose also your roommate has flu (more new information).
  - Flu almost surely!
- Probability of flu *given* fever and roommate flu:

 $\mathbb{P}[flu|fever and rommate flu] \approx 1$ 

### **CS, MATH and Dual CS-MATH Majors**

- Suppose RPI has 5000 students total, with 1000 in CS (vast underestimate!), 100 in MATH, 80 dual CS-MATH
- If you pick a random student from RPI, what's the chance the student is in CS

• Suppose (after you picked the student), I told you the student is a MATH major

 $\mathbb{P}[CS] = \frac{1000}{5000} = 0.2$ 

 $\mathbb{P}[MATH] = \frac{100}{5000} = 0.02$  $\mathbb{P}[CS \text{ and } MATH] = \frac{80}{5000} = 0.016$ 

- New information. What is  $\mathbb{P}[CS|MATH]$ ?
  - Effectively picking a random student from MATH
  - New probability of CS ~ striped area  $|CS \cap MATH|$  $\mathbb{P}[CS|MATH] = \frac{\mathbb{P}[CS \cap MATH]}{\mathbb{P}[MATH]} = \frac{80}{100} = 0.8$
- MATH majors 4 times more likely to be CS majors than a random student.

• Exercise 16.2.





#### Conditional Probability $\mathbb{P}[A|B]$

• Conditional probability is interpreted as follows:

 $\mathbb{P}[A|B]$  = frequency of outcomes known to be in *B* that are also in *A* 

• Suppose event *B* contains  $n_B$  outcomes when you repeat an experiment *n* times:  $\mathbb{P}[B] = \frac{n_B}{n_B}$ 

- Of the 
$$n_B$$
 outcomes in  $B$ , the number also in  $A$  is  $n_{A \cap B}$ 

$$\mathbb{P}[A \cap B] = \frac{n_{A \cap B}}{n}$$

• The frequency of outcomes in A among those outcomes in B is  $n_{A \cap B}/n_B$ 

$$\mathbb{P}[A|B] = \frac{n_{A \cap B}}{n_B} = \frac{n_{A \cap B}}{n} \times \frac{n}{n_B} = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[B]}$$

• The definition of conditional probability is then:

$$\mathbb{P}[A|B] = \frac{n_{A \cap B}}{n_B} = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[B]} = \frac{\mathbb{P}[A \text{ and } B]}{\mathbb{P}[B]}$$



### **Chance of Rain Given Clouds**

- It is cloudy one in five days,  $\mathbb{P}[Clouds] = \frac{1}{5}$
- It rains one in seven days,  $\mathbb{P}[Rain] = \frac{1}{7}$
- What are the chances of rain on a cloudy day?
  - Trick question!
  - We need to know  $\mathbb{P}[Rain \cap Clouds]$
  - It is mostly safe to assume that  $\{Rainy Days\} \subseteq \{Cloudy Days\}$
  - This means that  $\mathbb{P}[Rain \cap Clouds] = \mathbb{P}[Rain]$

$$\mathbb{P}[Rain|Clouds] = \frac{\mathbb{P}[Rain \cap Clouds]}{\mathbb{P}[Clouds]} = \frac{\frac{1}{7}}{\frac{1}{5}} = \frac{5}{7}$$

 $\mathbb{P}[Rain|Clouds] = \frac{\mathbb{P}[Rain \cap Clouds]}{\mathbb{P}[Clouds]}$ 

- 5-times more likely to rain on a cloudy day than on a random day.
- Crucial first step: identify the conditional probability. What is the "new information"?





### **Dicey Conditions**

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### **Dicey Conditions**

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#### **Computing a Conditional Probability**



- 1. Identify that you need a conditional probability  $\mathbb{P}[A|B]$
- 2. Determine the probability space  $(\Omega, \mathbb{P})$  using the outcome-tree method
- 3. Identify the events A and B appearing in  $\mathbb{P}[A|B]$  as subsets of  $\Omega$
- 4. Compute  $\mathbb{P}[A \cap B]$  and  $\mathbb{P}[B]$
- 5. Compute  $\mathbb{P}[A|B] = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[B]}$

#### **Monty Prefers Door 3**



- Recall the Monty Hall outcome tree
  Suppose Monty prefers door 3
- What is the best strategy now?
  - You should still switch
  - Winning outcomes are the same!  $\mathbb{P}[WinBySwitching] = \frac{2}{3}$



- Interestingly, you should feel better if you see Monty opening door 2!
  - Intuitively, why is this true?
  - Monty prefers door 3, so if he opened 2, then 3 very likely contains the prize!

 $\mathbb{P}[Win|Monty \ Opens \ Door \ 2] = \frac{\mathbb{P}[Win \ AND \ Monty \ Opens \ Door \ 2]}{\mathbb{P}[Monty \ Opens \ Door \ 2]}$  $= \frac{\frac{1}{3}}{\frac{1}{9} + \frac{1}{3}} = \frac{3}{4}$ 

• Your chances improved from  $\frac{2}{3}$  to  $\frac{3}{4}$ !

#### **A Pair of Slides**



- Suppose I have to make 2 lecture slides in 2 minutes
- The probability of having a typo on a slide is 0.5
  - What is the probability that both slides have typos?

- Answer:  $\frac{1}{4}$ 

 What is the probability that both slides have typos given the additional information in each of the cases below?



#### **Conditional Probability Traps**



 $\mathbb{P}[A \cap B]$ 

- These four probabilities are all different $\mathbb{P}[A]$  $\mathbb{P}[A|B]$  $\mathbb{P}[B|A]$
- Be careful which one you use!!
- Sampling bias: Using  $\mathbb{P}[A]$  instead of  $\mathbb{P}[A|B]$ 
  - Pollsters have a very tricky job when conducting surveys of public opinion
  - How do they contact people (landline, cell phone, social media, snail mail?)
  - Suppose you ask people "What is the probability that AI goes rogue?" and the only responses you get are through mail
  - What value are you surveying?

 $\mathbb{P}[rogue AI | responder uses mail]$ 

- This number is likely very different from  $\mathbb{P}[rogue AI]$
- Transposed conditional: using  $\mathbb{P}[B|A]$  instead of  $\mathbb{P}[A|B]$ 
  - See book for the famous Lombard study

#### The Covid Test and Transposed Conditionals



- At-home antigen covid tests have different accuracy depending on the case
  - If you have covid, the test will make a mistake ~20% of the time
  - If you don't have covid, the test will make a mistake ~1% of the time
  - Source: https://www.cochrane.org/CD013705/INFECTN\_how-accurate-arerapid-antigen-tests-diagnosing-covid-19
- Suppose you tested positive. What are the chances you actually have covid?
  - If you don't have covid, the test is unlikely to be wrong, so you are likely sick
- It is wrong to look at  $\mathbb{P}[positive|not Covid]$ . We already have this probability!
  - We need  $\mathbb{P}[not \ Covid| positive]$



### The Covid Test and Transposed Conditionals

It is wrong to look at P[positive|not Covid]. We already have this probability!
 We need P[not Covid|positive]

$$\mathbb{P}[not \ Covid | positive] = \frac{\mathbb{P}[not \ Covid \ AND \ YES]}{\mathbb{P}[YES]} \\ = \frac{0.9 \times 0.01}{0.1 \times 0.8 + 0.9 \times 0.01} \\ \approx 10\%$$

- The less accurate test says YES but the chances are 90% that you have covid
  - Two possibilities:
  - You don't have covid (likely) and test made a mistake (very rare)
  - You have covid (rare) and test was correct (likely). Wins!



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#### **Total Probability: Case by Case Probability**

- Two types of outcomes in any event *A*:
  - The outcomes in *B* (green)
  - The outcomes not in B (red)

 $\mathbb{P}[A] = \mathbb{P}[A \cap B] + \mathbb{P}[A \cap \overline{B}] \quad (*)$ 

- (Similar to sum rule from counting)
- (Also known as marginalization)
- From the definition of conditional probability:

$$\mathbb{P}[A \cap B] = \mathbb{P}[A \text{ AND } B] = \mathbb{P}[A|B] \times \mathbb{P}[B]$$
$$\mathbb{P}[A \cap \overline{B}] = \mathbb{P}[A \text{ AND } \overline{B}] = \mathbb{P}[A|\overline{B}] \times \mathbb{P}[\overline{B}]$$

- Plugging these in (\*), we get a **FUNDAMENTAL** result for case-by-case analysis
- LAW OF TOTAL PROBABILITY:

 $\mathbb{P}[A] = \mathbb{P}[A|B] \times \mathbb{P}[B] + \mathbb{P}[A|\overline{B}] \times \mathbb{P}[\overline{B}]$ 

 (Weight conditional probabilities for each case by probabilities of each case and add.)





#### Three Coins: Two Are Fair, One is 2-Headed



- Pick a random coin and flip. What is the probability of H?
- First, let's use the outcome-tree method as before



• What is the probability of H?

$$\mathbb{P}[H] = \frac{1}{3} + \frac{1}{6} + \frac{1}{6} = \frac{2}{3}$$

# Three Coins: Two Are Fair, One is 2-Headed, cont'd



- Pick a random coin and flip. What is the probability of H?
- Now, let's use the law of total probability:
  - Case 1. *B*: You picked one of the fair coins
  - Case 2.  $\overline{B}$ : You picked the two-headed coin

$$\mathbb{P}[H] = \mathbb{P}[H|B] \times \mathbb{P}[B] + \mathbb{P}[H|\overline{B}] \times \mathbb{P}[\overline{B}]$$
$$= \frac{1}{2} \times \frac{2}{3} + 1 \times \frac{1}{3} = \frac{2}{3}$$

- Note that we don't have to draw the (potentially exponentially growing) outcome tree anymore
- Exercise. A box has 10 coins: 6 fair and 4 biased (probability of heads  $\frac{2}{3}$ ). What is  $\mathbb{P}[2 \text{ heads}]$  in each case?
  - Pick a single random coin and flip it 3 times.
  - Flip 3 times. For each flip, pick a random coin, flip it and then put the coin back.

# Fair Toss from Biased Coin (unknown probability p of heads)?



- Suppose I would like to use a biased coin in order to obtain a fair outcome
  - The challenge is that I don't even know the probability of H or T for this coin
  - How do I do it?
  - One option would be to toss  $10^{1000}$  times and estimate p
  - A more pragmatic method is to notice that  $\mathbb{P}['ht'] = \mathbb{P}['th'] = p(1-p)$ 
    - (Lower case 'h' and 't' denote the outcomes of a toss.)
  - This suggests that an H is as likely as a T.
  - I can now design the following algorithm:
- 1. Make two tosses of the biased coin
- 2. If you get 'ht' output H; 'th' output T; otherwise RESTART.



# Fair Toss from Biased Coin (*unknown* probability *p* of heads)?, cont'd



- Let's use the law of total probability to estimate the value of H output by our algorithm
- $$\begin{split} \mathbb{P}[H] &= \mathbb{P}[H|RESTART] \cdot \mathbb{P}[RESTART] + \mathbb{P}[H|'ht'] \cdot \mathbb{P}['ht'] + \mathbb{P}[H|'th'] \cdot \mathbb{P}['th'] \\ &= \mathbb{P}[H] \cdot \left(p^2 + (1-p)^2\right) + 1 \cdot p(1-p) + 0 \cdot p(1-p) \\ &= \mathbb{P}[H] \cdot \left(p^2 + (1-p)^2\right) + p(1-p) \end{split}$$
- Solve for  $\mathbb{P}[H]$ :

$$\mathbb{P}[H] = \frac{p(1-p)}{1 - (p^2 + (1-p)^2)} = \frac{p(1-p)}{2p - 2p^2} = \frac{p(1-p)}{2p(1-p)} = \frac{1}{2}$$

(You can also solve this problem using an infinite outcome tree and computing an infinite sum.)

