

CSC 483/583  
Introduction to Conditional Probability  
(to accompany IIR Chapter 11)

Fall 2015

# Outline

## Examples of Conditional Probabilities

### Conditional Probability

### Bayes' Rule

### Independence

# Definition

- ▶ How would you define “probability”?

## Example: Blood Types

- Human blood is classified by the presence or absence of two antigens, called A and B. This gives rise to four types: O, A, B, and AB.

		Antigen B		Marginal
		Absent	Present	
Antigen A	Absent	0.44	0.10	0.54
	Present	0.42	0.04	0.46
Marginal		0.86	0.14	1.00

**Table:** Probability Estimates for U.S. Blood Types

- Let  $A$  be the event “presence of antigen A” and  $B$  be the event “presence of antigen B”

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## Example: Seat Belts

		Child		
		Buck.	Unbuck.	Marginal
Parent	Buck.	0.48	0.12	0.60
	Unbuck.	0.10	0.30	0.40
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- ▶ What is the (estimated) probability of the event “Child is Buckled”?
- ▶ What should our new estimate be if we know that (“given that”) “Parent is Buckled”?

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- ▶ **The set  $B$  becomes our new sample space, with an updated probability of 1!**
- ▶ Instead of asking “In what proportion of  $S$  is  $D$  true?”, we now ask “In what proportion of  $B$  is  $D$  true?”

## Conditioning Changes the Sample Space

- ▶ Recall conditional *proportions*. We calculate conditional proportions from frequencies by restricting attention to (“conditioning on”) a particular category, and dividing the joint frequency by the restricted total.
- ▶ Conditional probability is defined in exactly the same way, replacing proportions with probabilities.

### Conditional Probability

To find the probability of  $A$  *given*  $B$ , consider the ways  $A$  can occur *in the context of*  $B$  (i.e.,  $A \cap B$ ), out of all the ways  $B$  can occur.

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



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### The “Chain Rule” of Probability

For any events,  $A$  and  $B$ , the joint probability  $P(A \cap B)$  can be computed as

$$P(A \cap B) = P(A|B) \times P(B)$$

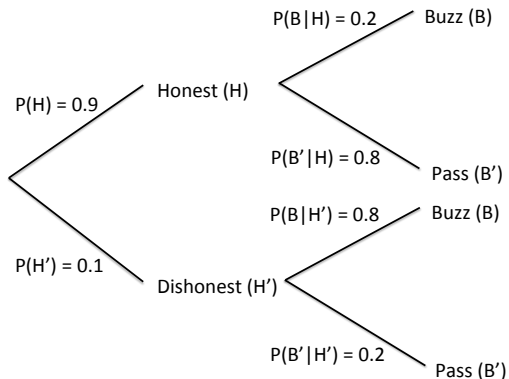
Or, since  $P(A \cap B) = P(B \cap A)$

$$P(A \cap B) = P(B|A) \times P(A)$$

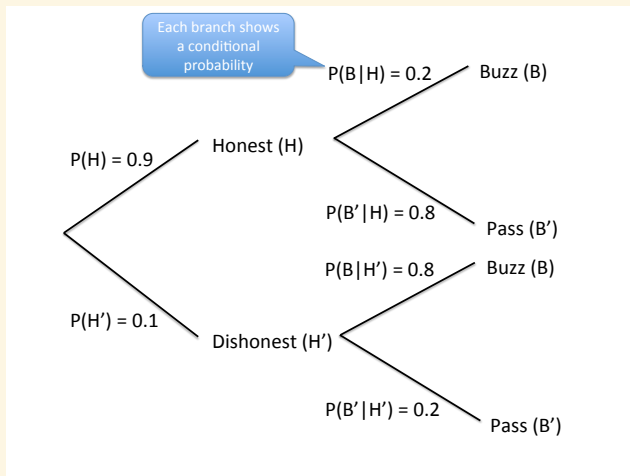
## Example: Lie Detector

- ▶ A store owner discovers that some of her employees have taken cash. She decides to use a lie detector to discover who they are.
- ▶ Suppose that 10% of employees stole, but 100% say they didn't.
- ▶ The lie detector buzzes 80% of the time that someone lies, and 20% of the time that someone is telling the truth.
- ▶ If the detector buzzes, what's the probability that the person was lying?

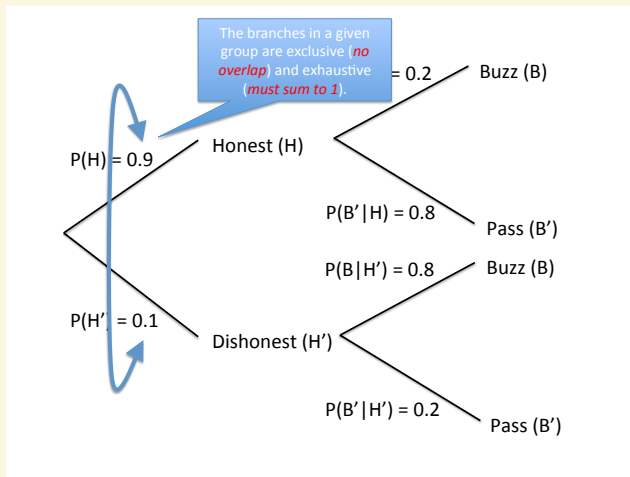
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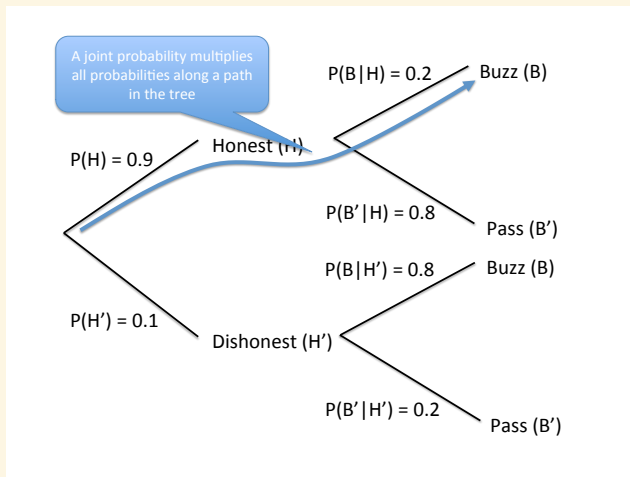


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- ▶ Computing conditional probabilities going against the flow of the tree is a bit tricky. We will see a better way soon!

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- ▶ What is the probability tree for this problem?

## Exercises: Donuts & Coffee

- ▶ Duncan's Donuts are looking into the probabilities of their customers buying donuts *and* coffee. Build the probability tree for them knowing that  $P(\text{Donuts}) = 3/4$ ,  $P(\text{Coffee}|\text{Donuts}') = 1/3$  and  $P(\text{Donuts} \cap \text{Coffee}) = 9/20$ .

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$$P(N|F) = \frac{P(N)P(F|N)}{P(F)} = \frac{0.54 \times 0.16}{0.55}$$

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We can reverse conditional probabilities using **Bayes' Rule**

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For any two events,  $A$  and  $B$ , we have

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- ▶ Remember: use Bayes Rule to reverse conditional probabilities.
- ▶ Useful to infer causes from effects (inferential statistics)!
- ▶ Very easy to derive from the chain rule, so remember that first.

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For any two events,  $A$  and  $B$ , we have

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- ▶ The **prior** (baseline) probability of having Lycanthropy is 1 in 1000.
- ▶ A test has been developed which gives a positive result for 9 in 10 werewolves and 1 in 20 non-werewolves.
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- ▶ We also have  $P(T|L')$ , which we haven't used yet...

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$$P(L|T) = \frac{P(L)P(T|L)}{P(T)}$$

- ▶ Do we have everything we need?
- ▶ We know  $P(L) = 1/1000$ ,  $P(T|L) = 9/10$ . We're missing  $P(T)$ .
- ▶ We also have  $P(T|L')$ , which we haven't used yet...
- ▶ Can we find  $P(T)$  from what we have?

# How to Compute $P(T)$ ?

- ▶ Two equivalent ways of doing it:
  - ▶ Contingency tables
  - ▶ Probability trees
- ▶ Use whichever is easiest for you.

# To the Contingency Table!

		Test		
		Positive	Negative	Marginal
Lycanthropy?	Yes			
	No			
Marginal				1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
- ▶ Sudoku it up!



# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes			0.001
	No			
Marginal				1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
- ▶ Sudoku it up!

# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes			0.001
	No			0.999
Marginal				1

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# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009		0.001
	No			0.999
Marginal				1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
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# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009	0.0001	0.001
	No			0.999
Marginal				1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
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# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009	0.0001	0.001
	No	0.04995		0.999
Marginal				1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
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# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009	0.0001	0.001
	No	0.04995	0.94905	0.999
Marginal				1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
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# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009	0.0001	0.001
	No	0.04995	0.94905	0.999
Marginal		0.05085		1

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# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009	0.0001	0.001
	No	0.04995	0.94905	0.999
Marginal		0.05085	0.94915	1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
- ▶ Sudoku it up!



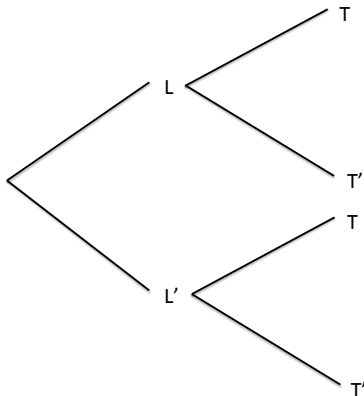
# To the Contingency Table!

		Test		Marginal
		Positive	Negative	
Lycanthropy?	Yes	0.0009	0.0001	0.001
	No	0.04995	0.94905	0.999
Marginal		0.05085	0.94915	1

- ▶  $P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$
- ▶ Sudoku it up!
- ▶  $P(L|T) = \frac{P(L \cap T)}{P(T)} = 0.0009/0.05085 = 0.018$

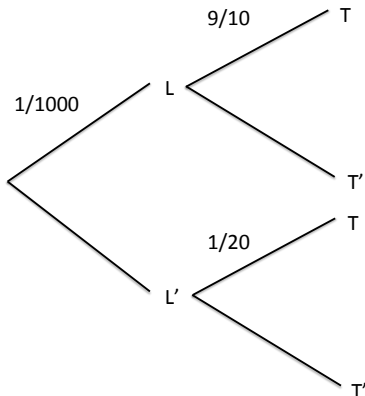
## To the Probability Tree!

$$P(L) = 1/1000, P(T|L) = 9/10, P(T|L') = 1/20$$



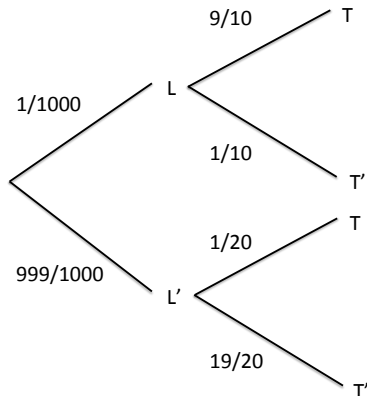
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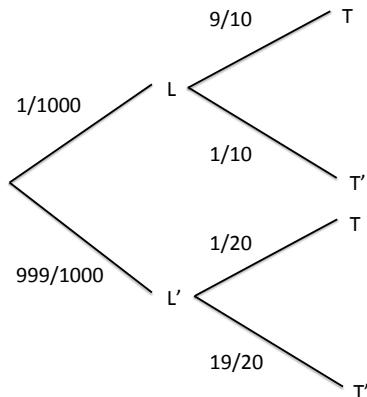
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To obtain  $P(T)$  sum up all joint probabilities (i.e., paths in the tree) that end up having T active:

$$P(T) = \frac{1}{1000} * \frac{9}{10} + \frac{999}{1000} * \frac{1}{20}$$

# The Law of Total Probability

- ▶ What was involved in finding the marginal probability  $P(T)$ ?
  1.  $P(T)$  was the sum of two joint probabilities,  $P(L \cap T)$  and  $P(L' \cap T)$ .

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  2. To find the joint probabilities, we multiplied a marginal by a conditional, according to the “chain rule”.

## The Law of Total Probability

For any events  $A$  and  $B$ , we have  $P(B) = P(A \cap B) + P(A' \cap B)$ . Together with two applications of the chain rule, this gives us

$$P(B) = P(A)P(B|A) + P(A')P(B|A')$$

# Bayes' Rule Revisited

Incorporating the Law of Total Probability into Bayes' Rule, we get

## Bayes Rule (version 2)

For any events  $A$  and  $B$ , we have

$$\begin{aligned} P(A|B) &= \frac{P(A \cap B)}{P(A \cap B) + P(A' \cap B)} \\ &= \frac{P(A)P(B|A)}{P(A)P(B|A) + P(A')P(B|A')} \end{aligned}$$

# Outline

Examples of Conditional Probabilities

Conditional Probability

Bayes' Rule

Independence

# Probabilistic Independence

- ▶ 10% of employees are dishonest.

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- ▶ What's the probability an employee is dishonest if it rains tomorrow?

# Probabilistic Independence

- ▶ 10% of employees are dishonest.
- ▶ There's a 5% chance of rain tomorrow.
- ▶ What's the probability an employee is dishonest if it rains tomorrow?
- ▶ Probably your intuition is that one conveys no information about the other. What does this mean about the relationship between the conditional probability of  $D$  given  $R$ , and the marginal probability of  $D$ ?

# Probabilistic Independence

## Independent Events

We say that event  $A$  is **independent** of event  $B$  if conditioning on  $B$  does not change the probability of  $A$ , that is if

$$P(A|B) = P(A)$$



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$$P(A|B) = P(A)$$

- ▶ If the employee is dishonest, what's the probability that it will rain tomorrow?
- ▶ It seems like independence should be symmetric. Is it?

# Probabilistic Independence

- ▶ If  $A$  is independent of  $B$ , then  $P(A|B) = P(A)$ . Is  $P(B|A)$  also equal to  $P(B)$ ?

# Probabilistic Independence

- ▶ If  $A$  is independent of  $B$ , then  $P(A|B) = P(A)$ . Is  $P(B|A)$  also equal to  $P(B)$ ?
- ▶ Using Bayes' rule, we have

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

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- ▶ Using Bayes' rule, we have

$$\begin{aligned}P(B|A) &= \frac{P(B)P(A|B)}{P(A)} \\&= \frac{P(B)P(A)}{P(A)} \\&= P(B)\end{aligned}$$

- ▶ So independence is in fact symmetric.

# Probabilistic Independence

- It turns out that if  $A$  and  $B$  are independent, then their joint probability has a particularly simple form

$$\begin{aligned}P(A \cap B) &= P(A)P(B|A) \\ &= P(A)P(B)\end{aligned}$$



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## Independence (version 2)

If  $A$  and  $B$  are independent events, then

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## Independence (version 2)

If  $A$  and  $B$  are independent events, then

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

- What is  $P(A \cup B)$  for independent events?

# Are these Independent Events?

1.  $E$ : First coin comes up heads,  $F$ : Second coin comes up heads

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4.  $E$ : Draw a yellow ball on first pick,  $F$ : Draw a yellow ball on second pick (with replacement)

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5.  $E$ : Draw a yellow ball on first pick,  $F$ : Draw a yellow ball on second pick (without replacement)

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1. We remove socks from a drawer with black and white, one after another, until we get a match. Is the color of the first sock independent of the color of the second?



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1. We remove socks from a drawer with black and white, one after another, until we get a match. Is the color of the first sock independent of the color of the second?
2.  $E_1$ : it rains today.  $E_2$ : today is Thursday.

## Exercise: the Case of the Two Classes

- ▶ The Health Club is wondering how best to market their new yoga class, and the Head of Marketing wonders if someone who goes swimming is more likely to go to a yoga class. "Maybe we could offer some sort of discount to the swimmers to get them to try our yoga."
- ▶ The CEO disagrees: "I think you're wrong. I think people who go swimming and people who go to yoga are independent. I don't think people who go swimming are more likely to do yoga than anybody else."
- ▶ They ask a group of 96 people whether they go to swimming or yoga classes. Out of this group, 32 go to yoga and 72 go swimming. 24 people go to both.
- ▶ Are the two classes dependent or independent?

## Exercise: the Absent-minded Diners

- ▶ Three friends decide to go out for a meal, but they forget where they're going to meet.
- ▶ Fred decides to throw a coin. If it lands heads, he'll go to the diner; tails, and he'll go to the Italian restaurant.
- ▶ George throws a coin, too: heads, it's the Italian restaurant; tails, it's the diner.
- ▶ Ron decides he'll just go to the Italian restaurant because he likes the food.
- ▶ What's the probability all three friends meet?
- ▶ What's the probability one of them eats alone?

# Summary

## Conditional Probability Summary

- ▶ Representing conditional probabilities using contingency tables and probability trees.
- ▶ The chain rule
- ▶ Bayes rule
- ▶ The law of total probability (aka the partition rule in IIR)
- ▶ Independent events

For more see this course + textbook:

Math 363:

<http://math.arizona.edu/~jwatkins/math363f15.htm>