# 305 Lecture 35 - Updating on Multiple Data Points

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• We will end the week by looking at some examples of updating on multiple data points.



Odds and Ends, Chapter 9

## **Conditional Independence**

In a lot of cases, the two data points we get will not be probabilistically independent, but they will be **conditionally independent**.

That is, if  $B_1$  and  $B_2$  are the data points, and X is an arbitrary hypothesis (like A,  $\neg A$ ), we will have

$$\Pr(\mathsf{B}_1 | \mathsf{X}) \Pr(\mathsf{B}_2 | \mathsf{X}) = \Pr(\mathsf{B}_1 \wedge \mathsf{B}_2 | \mathsf{X})$$

#### **Biased Coins**

Here is one kind of case where the happens.

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- If one flip lands heads, that is evidence of a bias towards heads, and hence it increases the probability of heads on the next flip.

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- We have a bunch of biased coins. For each of them, there is a
  probability p of heads on an arbitrary flip, but we don't know what that
  is.
- · The results of two flips of the same coin are not independent.
- If one flip lands heads, that is evidence of a bias towards heads, and hence it increases the probability of heads on the next flip.
- But conditional on a hypothesis about the bias of the coin, the flips are independent.

# **Skilled Activity**

A perhaps more real-life case of this is skilled action, like shooting free throws.

- The success of one attempt is not independent of the success of the previous.
- But conditional on the skill of the actor, the attempts are (probably, more or less) independent.

# Sampling With Replacement

Drawing from a selection with replacement produces conditional independence.

- If I don't know how many black marbles are in an urn, then drawing a
  black marble and replacing it will be evidence that the next marble is
  black.
- But conditional on a hypothesis about the nature of the urn, the draws with replacement will be independent.

## Yesterday, Today, Tomorrow

This is a little off topic, but a lot of real world phenomena satisfy (roughly) the following condition.

- How things were yesterday is a good (probabilistic) guide to how things will be tomorrow.
- So how things will be tomorrow is not independent of how things were yesterday.

## Yesterday, Today, Tomorrow

This is a little off topic, but a lot of real world phenomena satisfy (roughly) the following condition.

- How things were yesterday is a good (probabilistic) guide to how things will be tomorrow.
- So how things will be tomorrow is not independent of how things were yesterday.
- But, conditional on how things are today, how things were yesterday and will be tomorrow are independent.
- Knowing how things were yesterday doesn't tell you any more about how things will be tomorrow once you know how things are today.

#### **Markov Chains**

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- · Lots of real world processes are (more or less) Markov Chains.
- Weather systems, for instance, are probably more or less Markov Chains.
- And lots of ecological models assume that animal populations are Markov Chains.
- · And the core idea is just conditional independence.

# **Conditional Independence**

In cases where the data points  ${\it B}_1$  and  ${\it B}_2$  are independent, we have an easy story about how to work out the probabilities.

$$\Pr(\mathsf{B}_1 \land \mathsf{B}_2 | \mathsf{X}) = \Pr(\mathsf{B}_1 | \mathsf{X}) \Pr(\mathsf{B}_2 | \mathsf{X})$$

#### Same Event

There is an even simpler formula where  $B_1$  and  $B_2$  are the 'same' event, like the coin landing heads both time, or the same color marble being drawn.

$$Pr(B_1 \wedge B_2 | X) = Pr(B_1 | X)^2$$

# An Example

There are two urns in front of us.

- One of them urn A has 4 red marbles, 3 green marbles, and 3 blue marbles.
- The other urn B- has 8 red marbles, 1 green marbles and 1 blue marbles.

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One of the urns will be selected at random, and then a marble drawn from it.

• If the marble is red, what is the probability that Urn A was selected?

## A Table

I'll just do the column for red marble selected.

	Red
Urn A	$0.5 \times 0.4 = 0.2$
Urn B	$0.5 \times 0.8 = 0.4$
Total	0.2 + 0.4 = 0.6

#### A Table

I'll just do the column for red marble selected.

	Red
Urn A	$0.5 \times 0.4 = 0.2$
Urn B	$0.5 \times 0.8 = 0.4$
Total	0.2 + 0.4 = 0.6

$$\Pr(\textit{A|Red}) = \frac{\Pr(\textit{A} \land \textit{Red})}{\Pr(\textit{Red})} = \frac{0.2}{0.6} = \frac{1}{3}$$

#### **Another Example**

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One of the urns will be selected at random, and then two marbles drawn from it with replacement.

• If both draws are red, what is the probability that Urn A was selected?

## A Table

	Red-Red
Urn A	$0.5 \times 0.4^2 = 0.08$
Urn B	$0.5 \times 0.8^2 = 0.32$
Total	0.08 + 0.32 = 0.4

#### A Table

$$\begin{tabular}{c|cccc} \hline & Red-Red \\ \hline Urn A & 0.5 \times 0.4^2 = 0.08 \\ Urn B & 0.5 \times 0.8^2 = 0.32 \\ \hline Total & 0.08 + 0.32 = 0.4 \\ \hline \end{tabular}$$

$$\Pr(\textit{A|Red-Red}) = \frac{\Pr(\textit{A} \land \textit{Red-Red})}{\Pr(\textit{Red-Red})} = \frac{0.08}{0.4} = \frac{1}{5}$$

The probability of Urn A fell by a lot.

## Yet Another Example

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One of the urns will be selected at random, and then two marbles drawn from it with replacement.

 If the first draw is red and the second green, what is the probability that Urn A was selected?

# A Table

	Red-Green
Urn A	$0.5 \times 0.4 \times 0.3 = 0.06$
Urn B	$0.5 \times 0.8 \times 0.1 = 0.04$
Total	0.06 + 0.04 = 0.1

#### A Table

	Red-Green
Urn A	$0.5 \times 0.4 \times 0.3 = 0.06$
Urn B	$0.5 \times 0.8 \times 0.1 = 0.04$
Total	0.06 + 0.04 = 0.1

$$\Pr(\text{A|Red-Green}) = \frac{\Pr(\text{A} \land \text{Red-Green})}{\Pr(\text{Red-Green})} = \frac{0.06}{0.1} = \frac{3}{5}$$

The probability of Urn A rose by a lot.

**Dependent Events** 

## Dependence

What happens if the events  ${\it B}_1$  and  ${\it B}_2$  are dependent on one or other of the hypotheses?

- The typical case is that they will be dependent on none or all of the hypotheses.
- But it's possible in principle to have independence on some and dependence on others.
- And in that case we have to use the more complicated procedure I'm about to describe.

# Sampling Without Replacement

The paradigm example of conditional dependence is sampling **without** replacement.

- · Assume you know which urn I'm using.
- Then the draws without replacement won't be independent because every time you draw a marble, there are fewer marbles of that color to draw the next time.

## Example

Assume that I am using urn A. (Or assume that we are working out conditional probabilities conditional on urn A.)

- For the first draw, the probability of red is 4 in 10, or 0.4.
- Conditional on the first draw being red, the probability of the second draw being red is 3 in 9, or  $\frac{1}{3}$ .
- That's because there are now 9 marbles left, and 3 of them are red.

# Continuing the Example

So to work out the probability of some sequence of draws  ${\it D}_1, {\it D}_2$  given a hypothesis X about the urn, we need to use the more complicated rule.

$$\Pr(\mathsf{D}_1 \land \mathsf{D}_2 | \mathsf{X}) = \Pr(\mathsf{D}_1 | \mathsf{X}) \Pr(\mathsf{D}_2 | \mathsf{X} \land \mathsf{D}_1)$$

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For example

$$\Pr(\mathit{Red}_1 \land \mathit{Red}_2 | \mathit{A}) = \Pr(\mathit{Red}_1 | \mathit{A}) \Pr(\mathit{Red}_2 | \mathit{A} \land \mathit{Red}_1) = \frac{4}{10} \times \frac{3}{9} = \frac{2}{15}$$

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For example

$$\Pr(\mathit{Red}_1 \land \mathit{Red}_2 | \mathit{B}) = \Pr(\mathit{Red}_1 | \mathit{B}) \Pr(\mathit{Red}_2 | \mathit{B} \land \mathit{Red}_1) = \frac{8}{10} \times \frac{7}{9} = \frac{28}{45}$$

## **Another Example**

There are two urns in front of us.

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- The other urn B- has 8 red marbles, 1 green marbles and 1 blue marbles.

## **Another Example**

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- The other urn B- has 8 red marbles, 1 green marbles and 1 blue marbles.

One of the urns will be selected at random, and then two marbles drawn from it without replacement.

• If both draws are red, what is the probability that Urn A was selected?

	Red-Red
Jrn A	$0.5 \times \frac{4}{10} \times \frac{3}{9} = \frac{1}{15}$
Jrn B	$0.5 \times \frac{8}{10} \times \frac{7}{9} = \frac{14}{45}$
Total	$\frac{1}{15} + \frac{14}{45} = \frac{17}{45}$

$$\Pr(\mathsf{A}|\mathit{Red}-\mathit{Red}) = \frac{\Pr(\mathsf{A} \land \mathit{Red}-\mathit{Red})}{\Pr(\mathit{Red}-\mathit{Red})} = \frac{\frac{1}{15}}{\frac{17}{45}} = \frac{3}{17}$$

The probability of Urn A fell by a bit more.

## Yet Another Example

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One of the urns will be selected at random, and then two marbles drawn from it with replacement.

 If the first draw is red and the second green, what is the probability that Urn A was selected?

## The General Conjunction Rule

To work out the probability of some sequence of draws  $D_1, D_2$  given a hypothesis X about the urn, we need to use the more complicated rule.

$$\Pr(\mathsf{D}_1 \land \mathsf{D}_2 | \mathsf{X}) = \Pr(\mathsf{D}_1 | \mathsf{X}) \Pr(\mathsf{D}_2 | \mathsf{X} \land \mathsf{D}_1)$$

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$$\Pr(\mathsf{D}_1 \land \mathsf{D}_2 | \mathsf{X}) = \Pr(\mathsf{D}_1 | \mathsf{X}) \Pr(\mathsf{D}_2 | \mathsf{X} \land \mathsf{D}_1)$$

So in this case we get

$$\Pr(\mathit{Red}_1 \land \mathit{Green}_2 | \mathit{A}) = \Pr(\mathit{Red}_1 | \mathit{A}) \Pr(\mathit{Green}_2 | \mathit{A} \land \mathit{Red}_1) = \frac{4}{10} \times \frac{3}{9} = \frac{2}{15}$$

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To work out the probability of some sequence of draws  ${\it D}_1, {\it D}_2$  given a hypothesis X about the urn, we need to use the more complicated rule.

$$\Pr(\mathsf{D}_1 \land \mathsf{D}_2 | \mathsf{X}) = \Pr(\mathsf{D}_1 | \mathsf{X}) \Pr(\mathsf{D}_2 | \mathsf{X} \land \mathsf{D}_1)$$

And for Urn B we get

$$\Pr(\mathit{Red}_1 \land \mathit{Green}_2 | \mathit{B}) = \Pr(\mathit{Red}_1 | \mathit{B}) \Pr(\mathit{Green}_2 | \mathit{B} \land \mathit{Red}_1) = \frac{8}{10} \times \frac{1}{9} = \frac{4}{45}$$

	Red-Green
Jrn A Jrn B <b>Total</b>	$0.5 \times \frac{4}{10} \times \frac{3}{9} = \frac{1}{15}$ $0.5 \times \frac{8}{10} \times \frac{1}{9} = \frac{2}{45}$ $\frac{1}{15} + \frac{2}{45} = \frac{5}{45}$

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Urn B	$0.5 \times \frac{8}{10} \times \frac{1}{9} = \frac{2}{45}$
Total	$\frac{1}{15} + \frac{2}{45} = \frac{5}{45}$

$$\Pr(A|Red - Green) = \frac{\Pr(A \land Red - Red)}{\Pr(Red - Red)} = \frac{\frac{1}{15}}{\frac{5}{45}} = \frac{3}{5}$$

Which, interestingly, is exactly the same as in the with replacement case.

# Last (Difficult) Example

- There are four urns in the room, three of type X, one of type Y.
- The type X urns have 4 blue marbles and 2 yellow marbles.
- The type Y urn has 5 blue marbles and 3 yellow marbles.
- · One of the four urns was selected at random.
- Then two marbles were selected without replacement from the randomly selected urn.
- · The first was blue, the second was yellow.
- · A third marble is about to be selected.
- · What is the probability that it is blue?

### The Table

Urn	Blue-then-Yellow
Type X	$\frac{3}{4} \times \frac{4}{6} \times \frac{2}{5} = \frac{1}{5}$
Type Y	
Total	

 $\Pr(\mathbf{X} \land \mathit{Blue}_1 \land \mathit{Yellow}_2) = \Pr(\mathbf{X}) \times \Pr(\mathit{Blue}_1 | \mathbf{X}) \times \Pr(\mathit{Yellow}_2 | \mathbf{X} \land \mathit{Blue}_1)$ 

### The Table

Urn	Blue-then-Yellow
Type X Type Y Total	$\frac{\frac{3}{4} \times \frac{4}{6} \times \frac{2}{5} = \frac{1}{5}}{\frac{1}{4} \times \frac{5}{8} \times \frac{3}{7} = \frac{15}{224}}$

 $\Pr(\mathbf{Y} \land \mathit{Blue}_1 \land \mathit{Yellow}_2) = \Pr(\mathbf{Y}) \times \Pr(\mathit{Blue}_1 | \mathbf{Y}) \times \Pr(\mathit{Yellow}_2 | \mathbf{Y} \land \mathit{Blue}_1)$ 

#### The Table

Urn	Blue-then-Yellow
Type X Type Y Total	$ \begin{array}{r} \frac{1}{5} \\ 15 \\ 224 \\ 299 \\ 1120 \end{array} $

You should double check this, but I think

$$\frac{1}{5} + \frac{15}{224} = \frac{299}{1120}$$

So that's  $Pr(Blue_1 \land Yellow_2)$ 

### **Conditional Probabilities**

$$\Pr(\textit{X}|\textit{Blue}_1 \land \textit{Yellow}_2) = \frac{\Pr(\textit{X} \land \textit{Blue}_1 \land \textit{Yellow}_2)}{\Pr(\textit{Blue}_1 \land \textit{Yellow}_2)} = \frac{\frac{1}{5}}{\frac{299}{1120}} = \frac{224}{299}$$

$$\Pr(\mathsf{Y}|\mathit{Blue}_1 \land \mathit{Yellow}_2) = \frac{\Pr(\mathsf{Y} \land \mathit{Blue}_1 \land \mathit{Yellow}_2)}{\Pr(\mathit{Blue}_1 \land \mathit{Yellow}_2)} = \frac{\frac{15}{224}}{\frac{299}{1120}} = \frac{75}{299}$$

The probability of Y is ever so fractionally higher than when we started.

#### **Next Marble**

- If X (and Blue-followed-by-Yellow), the probability of next marble being blue is  $\frac{3}{4}$ .
- If Y (and Blue-followed-by-Yellow), the probability of next marble being blue is  $\frac{2}{3}$ .

#### **Next Marble**

- If X (and Blue-followed-by-Yellow), the probability of next marble being blue is  $\frac{3}{4}$ .
- If Y (and Blue-followed-by-Yellow), the probability of next marble being blue is  $\frac{2}{3}$ .
- · So overall probability of next marble being blue is

$$\frac{224}{299} \times \frac{3}{4} + \frac{75}{299} \times \frac{2}{3} = \frac{218}{299} \approx 0.729$$

## General Strategy of Last Slide

 If there are two hypotheses X and Y, and you want to know the probability of some event E, it will be given by

$$Pr(E) = Pr(X) Pr(E|X) + Pr(Y) Pr(E|Y)$$

And that generalises to the case where there are multiple hypotheses  ${\cal H}_1, \dots {\cal H}_n$ 

$$\Pr(\mathit{E}) = \Pr(\mathit{H}_1) \Pr(\mathit{E}|\mathit{H}_1) + \dots + \Pr(\mathit{H}_n) \Pr(\mathit{E}|\mathit{H}_n)$$

