## Probability Refresher

CMPUT 397 Fall 2021

### Rolling a 6-sided dice

- The outcomes of a **single** dice roll is {1, 2, 3, 4, 5, 6}
- Some events can be:
  - seeing number 1, seeing number 2, seeing 1 or 2, seeing 1 or 3, etc
- The probability of seeing number 1 is 1/6, the probability of seeing number 1 or 2 is 1/6 + 1/6 = 1/3

## Modelling random phenomena

- Define a set of possible outcomes, a collection of events and probability for each event
- 6-sided dices experiment:
  - the outcomes of a single dice roll are 1, 2, 3, 4, 5, 6
  - A collection of all **events** for 6-sided dice experiment: {{},{1},{2},{3},{4},{5},{6},{1,2}, {1,3}, ..., {1,2,3,4,5,6}}
    - {1} translates to "seeing number 1"
    - {1,2} translates to "seeing number 1 or 2" and the **probability**("seeing number 1 or 2") = 1/3
    - {} translates to "seeing nothing" and the probability{"seeing nothing"} = 0
    - {1, 2, 3, 4, 5, 6} translates to "seeing 1 or 2 or 3 or 4 or 5 or 6" and the probability of this event is 1

## Flipping a 2-sided coin twice

- The possible outcomes are {HH, HT, TH, TT}
- Some examples of events:
  - Seeing head on the first flip = {HH, HT}, seeing only tails = {TT}
- Probability on the events:
  - Prob("seeing head on the first flip") = 1/4 + 1/4 = 1/2
  - Prob("seeing only tails") = 1/4

#### Random variable

 A random variable X assigns a numerical value to each of the outcomes in the outcome space. In other words, it is a function that maps an outcome (not event) to the real

- Examples:
  - 6-sided dice roll:
    - X(outcome 1) = 1, X(outcome 2) = 2, X(outcome 3) = 3, etc

### Random variable continued

- Let's examine the 2-sided coin flip experiment
- Outcomes are {HH, HT, TH, TT}
- A table of events and their corresponding probabilities on the right
- Define X(HH) = 1, X(HT) = 2, X(TH) = 3, X(TT) = 4
- After flipping a coin twice, now we see 1 instead of the outcome HH, HT, TH, or TT, what is the probability of seeing 1?

• 
$$P(X = 1)$$
?

Event	Probability
nothing	0
НН	1/4
HT	1/4
TH	1/4
π	1/4
HH, HT	1/2
HH, TH	1/2
HH, TT	1/2
HT, TH	1/2
HT, TT	1/2
TH, TT	1/2
HH, HT, TH	3/4
HH, HT, TT	3/4
HT, TH, TT	3/4
HH, HT, TH, TT	1

## What does P(X=1) mean?

- Let  $\omega = \{HH, HT, TH, TT\}$ , then P(X = 1) really mean:
  - $P(\{\omega | X(\omega) = 1\}) = P(\{HH\}) = 1/4$
  - Recall X(HH) = 1, X(HT) = 2, X(TH) = 3, X(TT) = 4

Event	Probability
nothing	0
НН	1/4
HT	1/4
TH	1/4
П	1/4
HH, HT	1/2
нн, тн	1/2
HH, TT	1/2
HT, TH	1/2
HT, TT	1/2
TH, TT	1/2
HH, HT, TH	3/4
HH, HT, TT	3/4
HT, TH, TT	3/4
HH, HT, TH, TT	1

### Random variable continued

- Let's examine the 2-sided coin flip experiment again
- Outcomes are {HH, HT, TH, TT}
- A table of events and their corresponding probabilities on the right
- Define X(HT) = 1 = X(TH) and X(HH) = 2 and X(TT) = 3
- After flipping a coin twice, now we see 1 instead of the outcome HH or TH, what is the probability of seeing 1?
  - P(X = 1)?

Event	Probability
nothing	0
нн	1/4
HT	1/4
TH	1/4
П	1/4
нн, нт	1/2
нн, тн	1/2
нн, тт	1/2
HT, TH	1/2
HT, TT	1/2
TH, TT	1/2
HH, HT, TH	3/4
HH, HT, TT	3/4
HT, TH, TT	3/4
HH, HT, TH, TT	1

## Discrete and continuous random variables

- A discrete random variable takes a distinct number of values, which can be finite or countably infinite:
  - Example: 6-sided dice (  $\mathcal{X} = \{1,2,3,4,5,6\}$  )
- A **continuous** random variable takes an uncountably infinite number of values (like an interval)
  - Example: temperature of a thermometer

### Notations

- Discrete random variable:
  - P(X=x)
- Continuous random variable:
  - $P(X \le x)$  or  $P(a \le X \le b)$
  - P(X = x) = 0

## Probability mass function

• Let X be a discrete random variable that takes values in  $\mathcal{X} = \{x_1, x_2, \dots\}$  (finite or countably infinite). The probability

• 
$$P(X = x_k)$$
 for  $k = 1, 2, 3, ...,$ 

is called the probability mass function (PMF) of X

• Equivalently we write: p(x) = P(X = x)

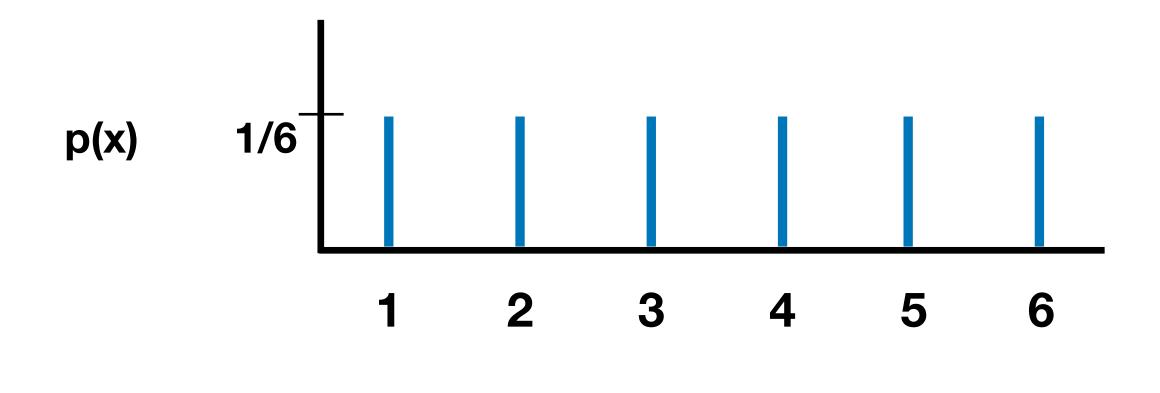
## Properties of the PMF

1. 
$$p: \mathcal{X} \to [0, 1]$$
  
i.e.,  $0 \le p(x) \le 1$ 

$$\sum_{x \in \mathcal{X}} p(x) = 1$$

## An example of PMF

- Recall the 6-sided dice roll
- p(x) = 1/6 for all x in  $\{1, 2, 3, 4, 5, 6\}$

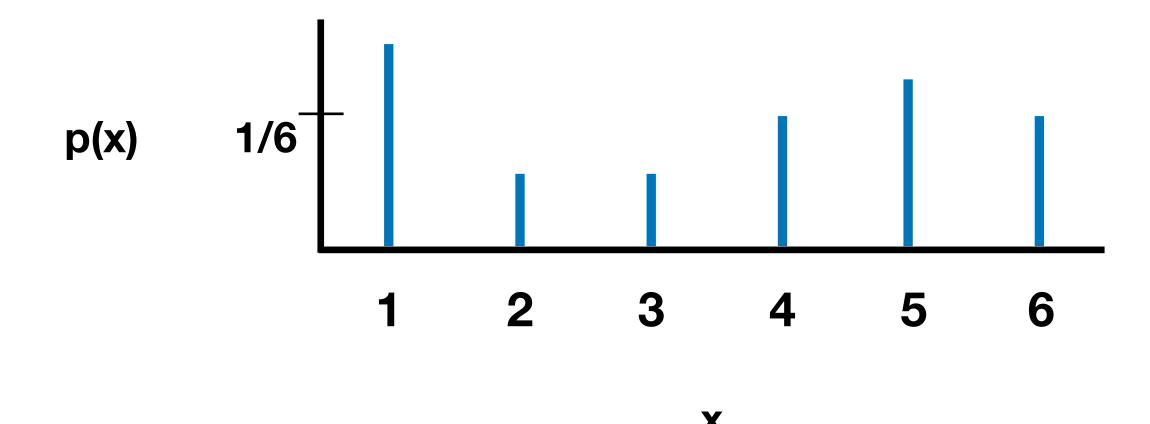


X

What is  $\sum_{x \in \mathcal{X}} p(x)$ ?

#### Another PMF

- Recall the 6-sided dice roll
- What does this PMF say?



### Another example

- Imagine that you are a medical doctor, prescribing treatments to patients
- You give the treatment for 100 patients and note the treatment outcome as {Bad, Neutral, Good}.
- 20 patients have a Bad outcome
- 50 a Neutral outcome
- 30 a Good outcome
- $p(x) = \begin{cases} 0.2 & \text{if } x \text{ is Bad} \\ 0.5 & \text{if } x \text{ is Neutral} \\ 0.3 & \text{if } x \text{ is Good.} \end{cases}$

Obtain p(x) using 20/100, 50/100 and 30/100

### Another example

- Imagine that you are a medical doctor, prescribing treatments to patients
- You give the treatment for many patients and note the outcome X = treatment outcome in {Bad, Neutral, Good}. You find that you have the following probabilities:

$$p(x) = \begin{cases} 0.2 & \text{if } x \text{ is Bad} \\ 0.5 & \text{if } x \text{ is Neutral} \\ 0.3 & \text{if } x \text{ is Good.} \end{cases}$$

• But that's pretty stochastic...

#### How can we make this less stochastic?

- What if we knew something about the patient?
- What information could help?

$$p(x) = \begin{cases} 0.2 & \text{if } x \text{ is Bad} \\ 0.5 & \text{if } x \text{ is Neutral} \\ 0.3 & \text{if } x \text{ is Good.} \end{cases}$$

### Conditional Probabilities

- X = outcome of treatment
- Y = age of the patient
- P(X = Good | Y = 22) can a different value than P(X = Good | Y = 78)
- P(X = x | Y = y) can be read as the probability that the random variable 'X' takes the value 'x', given that the random variable 'Y" has taken the value 'y'.

# Another example of conditional probabilities

- Often the value of a random variable is dependent on or correlated with another random variable
- Example: a store restocks on Wednesday, so the probability that they have pencils in stock depends on the day of the week

1	Wed	Pencil Delivery!
2	Thu	
3	Fri	
4	Sat	
5	Sun	
6	Mon	
7	Tues	

 $P(\text{Pencils in stock} \mid \text{Tuesday}) = 0.2$   $P(\text{Pencils in stock} \mid \text{Wednesday}) = 1.0$ 

## Same properties for conditional probabilities as for unconditioned probabilities

1. 
$$p(\cdot|Y=y): \mathcal{X} \to [0,1]$$

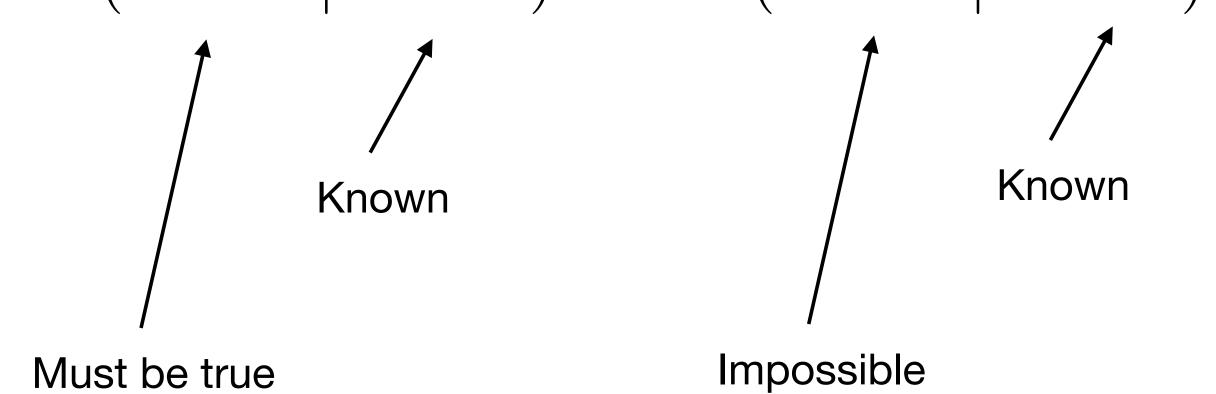
Could write it  $p_{Y=y}(x)$ 

$$\sum_{x \in \mathcal{X}} p(x|Y=y) = 1$$

e.g. If  $P(\text{Pencils in stock} \mid \text{Tuesday}) = 0.2$ , then  $P(\text{Pencils NOT in stock} \mid \text{Tuesday}) = 0.8$ 

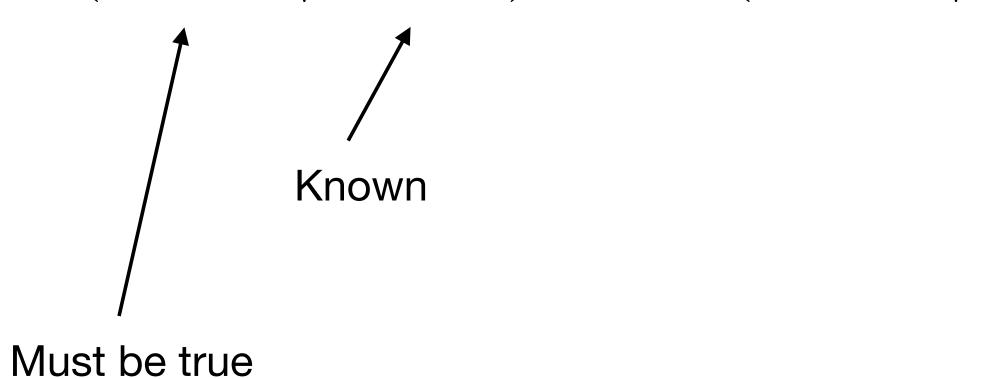
### Exercise 1

• Find  $P(X = 3 \mid X = 3)$  and  $P(X = 9 \mid X = 4)$ .



### Exercise 1: Answer

• Find  $P(X = 3 \mid X = 3)$  and  $P(X = 9 \mid X = 4)$ .



$$P(X = 3 \mid X = 3) = 1$$

### Multiple random variables

- We now have seen two random variables, X and Y, we can reason about more than just p(x | y)
- We can ask about their joint probability: p(x, y)
- We might want to ask about the other conditional: p(y | x)

## Back to the Pencils example

- Let X in {0, 1} (Pencils not in stock, Pencils in stock)
- Let Y in {Monday, Tuesday, ..., Sunday} (Notice p(y) = 1/7 for all y)
- We can ask P(X = 1, Y = Tuesday) or P(Y = Tuesday | X = 1)
- Do you think P(Y = Wednesday | X = 1) > P(Y = Tuesday | X = 1)

1	Wed	Pencil Delivery!
2	Thu	
3	Fri	
4	Sat	
5	Sun	
6	Mon	
7	Tues	

 $P(\text{Pencils in stock} \mid \text{Tuesday}) = 0.2$   $P(\text{Pencils in stock} \mid \text{Wednesday}) = 1.0$ 

### Chain Rule

- Can express joint probability using conditional probabilities
- p(x, y) = p(x | y) p(y)
- p(x, y) = p(y | x) p(x)

### Back to the Pencils example

- Do you think P(Y = Wednesday, X = 1) > P(Y = Tuesday, X = 1)?
- Why or why not? Use P(Y = Wednesday, X = 1) = P(Y = Wednesday | X = 1) P(X = 1)

1	Wed	Pencil Delivery!
2	Thu	
3	Fri	
4	Sat	
5	Sun	
6	Mon	
7	Tues	

 $P(\text{Pencils in stock} \mid \text{Tuesday}) = 0.2$   $P(\text{Pencils in stock} \mid \text{Wednesday}) = 1.0$ 

# Independence and Conditional Independence

- X and Y are independent if and only if p(x, y) = p(x) p(y)
  - equivalently, if p(x | y) = p(x)
  - recall: p(x, y) = p(x | y) p(y)
- X and Y are conditionally independent, given Z, if and only if p(x, y | z) = p(x | z) p(y | z)

## Example of Independence

- 2-sided fair coin
- Let X denote the first flip, which results in any of the outcome {H, T}
- Let Y denote the second flip, which results in any of the outcome {H, T}
- Both flips are independent of each other, and if the probability of flipping H is 1/2, then
  - p(x, y) = p(x) p(y) = 1/2 x 1/2 = 1/4
  - i.e., P(X = H, Y = H) = P(X = H) P(Y = H), and P(X = T, Y = H) = P(X = T) P(Y = H), ...

### Example of Conditional Independence

- For the pencils example, imagine you had an random variable Z where
  - Z = 0 if the day of the week is in the range Saturday to Tuesday
  - Z = 1 if the day of the week is in the range Wednesday to Friday
- Z is not independent of X (recall, X is whether Pencils are in stock or not)
  - $P(Z = 0 \mid X = 1)$  not equal to P(Z = 0) (P(Z = 0) = 4/7 whereas  $P(Z = 0 \mid X = 1) < 4/7$ )
- Z is conditionally independent of X, given Y (Y is the day of the week)

• 
$$P(Z=0|X=1,Y=D)=P(Z=0|Y=D)=\left\{ egin{array}{ll} 0 & \mbox{if D in Sat to Tues} \\ 1 & \mbox{if D in Wed to Fri} \end{array} \right.$$

### Marginalization

• If we have the joint distribution p(x, y), we can find the marginals p(x) and p(y)

$$p(x) = \sum_{y \in \mathcal{Y}} p(x, y) \qquad p(y) = \sum_{x \in \mathcal{X}} p(x, y)$$

## Exercise 2 (Marginalization)

- Imagine someone gives you P(X = 1 | Y = y) for y in {Monday,..., Sunday}
- You already know p(y) = 1/7
- How would you get p(X = 1)? (i.e., the probability that Pencils are in stock)

## Exercise 3 (Conditional Probs)

 You flip a coin and get two heads in a row. What is the probability that your third coin flip also results in heads?



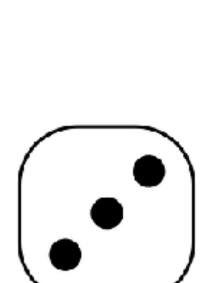






## Exercise 4 (Formalizing probability questions)

You roll two standard dice. One of the die shows a 3.
What is the probability that the sum of the dice is greater than 7?



Roll 1

