

Conditional Independence; Random variables

CSE 312 Spring 26
Lecture 8

Conditional Independence

Definition. Two events A and B are **independent** conditioned on C if $P(C) \neq 0$ and $P(A \cap B | C) = P(A | C) \cdot P(B | C)$.

- If $P(A \cap C) \neq 0$, equivalent to $P(B|A \cap C) = P(B | C)$
- If $P(B \cap C) \neq 0$, equivalent to $P(A|B \cap C) = P(A | C)$

Contrast to Plain Independence. Two events A and B are **independent** if

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

Example – Tossing Coins (1)

Suppose that Coin 1 has probability of heads 0.3
and Coin 2 has probability of head 0.9.

We choose one of the coins randomly with equal probability and flip that coin 2 times independently. **What is the probability we get all heads?**

C_i = coin i was selected

Tossing Coins (2)

Suppose that Coin 1 has probability of heads 0.3
and Coin 2 has probability of head 0.9.

We choose one of the coins randomly with equal probability and flip that coin 2 times independently. **What is the probability we get all heads?**

C_i = coin i was selected

$$\begin{aligned} P(HH) &= P(HH | C_1) \cdot P(C_1) + P(HH | C_2) \cdot P(C_2) && \text{Law of Total Probability (LTP)} \\ &= P(H|C_1)^2 P(C_1) + P(H | C_2)^2 P(C_2) && \text{Conditional Independence} \\ &= 0.3^2 \cdot 0.5 + 0.9^2 \cdot 0.5 = 0.45 \end{aligned}$$

Tossing coins (3)

Suppose that Coin 1 has probability of heads 0.3
and Coin 2 has probability of head 0.9.

We choose one coin randomly with equal probability and flip that coin 2 times independently. How does the probability we get all heads compare to $P(H)^2$?

Example – Tossing coins (4)

Suppose that Coin 1 has probability of heads 0.3
and Coin 2 has probability of head 0.9.

We choose one coin randomly with equal probability and flip that coin 2 times independently. How does the probability we get all heads compare to $P(H)^2$?

$$P(HH | C_1) = P(H | C_1)^2$$

$$P(HH) \neq P(H)^2$$

$$P(HH) = P(H | C_1)^2 P(C_1) + P(H | C_2)^2 P(C_2) = 0.3^2 \cdot 0.5 + 0.9^2 \cdot 0.5 = 0.45$$

$$P(H) = P(H | C_1) \cdot P(C_1) + P(H | C_2) \cdot P(C_2) = 0.6 \qquad P(HH) = 0.36$$

Pset 3 Coding – Naïve Bayes (Algorithm for Spam Detection)

Will compute probability an email is spam given the words in it

Say email contains words $\{x_1, x_2, \dots, x_n\}$.

Want to compute probability it's spam:

$$\mathbb{P}(S \mid x_1, \dots, x_n)$$

$$\frac{\mathbb{P}(x_1, \dots, x_n \mid S)\mathbb{P}(S)}{\mathbb{P}(x_1, \dots, x_n)} = \frac{\mathbb{P}(x_1, \dots, x_n \mid S)\mathbb{P}(S)}{\mathbb{P}(x_1, \dots, x_n \mid S)\mathbb{P}(S) + \mathbb{P}(x_1, \dots, x_n \mid H)\mathbb{P}(H)}$$

We will assume words in an email are conditionally independent, given email is spam or ham!

$$\begin{aligned}\mathbb{P}(x_1, \dots, x_n, S) &= \mathbb{P}(x_1 \mid x_2, x_3, \dots, x_n, S)\mathbb{P}(x_2 \mid x_3, \dots, x_n, S) \dots \mathbb{P}(x_{n-1} \mid x_n, S)\mathbb{P}(x_n \mid S)\mathbb{P}(S) \\ &\approx \mathbb{P}(x_1 \mid S)\mathbb{P}(x_2 \mid S) \dots \mathbb{P}(x_{n-1} \mid S)\mathbb{P}(x_n \mid S)\mathbb{P}(S) \\ &= \mathbb{P}(S) \prod_{i=1}^n \mathbb{P}(x_i \mid S)\end{aligned}$$

Coding problem will be released on Ed at 11am today.

Agenda (1)

- Random Variables
- Probability Mass Function (PMF)
- Cumulative Distribution Function (CDF)

Random Variables (Idea)

Often: We want to **capture quantitative properties** of the outcome of a random experiment, e.g.:

- *What is the total of two dice rolls?*
- *What is the number of coin tosses needed to see the first head?*
- *What is the number of heads among 5 coin tosses?*

Random Variables

Definition. A **random variable (RV)** for a probability space (Ω, \mathbb{P}) is a function $X: \Omega \rightarrow \mathbb{R}$.

The set of values that X can take on is called its range/support Ω_X

Example. Number of heads in 2 independent coin flips $\Omega = \{HH, HT, TH, TT\}$

$$\Omega_X = \{0, 1, 2\}$$

Drawing Balls

20 balls labeled 1, 2, ..., 20 in an urn

- Draw a subset of 3 uniformly at random
- What is $|\Omega|$?

Drawing Balls – new random variable

20 balls labeled 1, 2, ..., 20 in an urn

– Draw a subset of 3 uniformly at random

– $|\Omega| = \binom{20}{3}$

– Let $X =$ maximum of the 3 numbers on the balls

- Example: $X(2, 7, 5) = 7$

- Example: $X(15, 3, 8) = 15$

Drawing Balls – what is $|\Omega_X|$?

20 balls labeled 1, 2, ..., 20 in an urn

- Draw a subset of 3 uniformly at random
- Let $X =$ maximum of the 3 numbers on the balls
 - Example: $X(2, 7, 5) = 7$
 - Example: $X(15, 3, 8) = 15$
- What is $|\Omega_X|$?

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- A. 20^3
- B. 20
- C. 18
- D. $\binom{20}{3}$

Example: Returning Homeworks

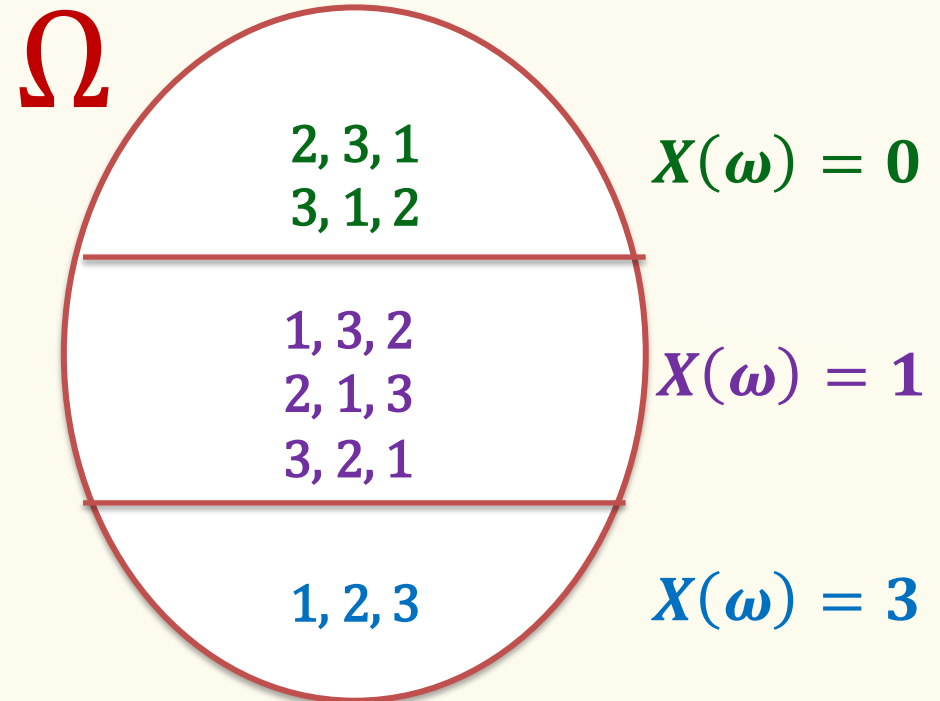
- Class with 3 students, randomly hand back homeworks. All permutations equally likely.
- Let X be the number of students who get their own HW

$\Pr(\omega)$	ω	$X(\omega)$
1/6	1, 2, 3	
1/6	1, 3, 2	
1/6	2, 1, 3	
1/6	2, 3, 1	
1/6	3, 1, 2	
1/6	3, 2, 1	

Example: Returning Homeworks (2)

- Class with 3 students, randomly hand back homeworks. All permutations equally likely.
- Let X be the number of students who get their own HW

$\Pr(\omega)$	ω	$X(\omega)$
1/6	1, 2, 3	3
1/6	1, 3, 2	1
1/6	2, 1, 3	1
1/6	2, 3, 1	0
1/6	3, 1, 2	0
1/6	3, 2, 1	1

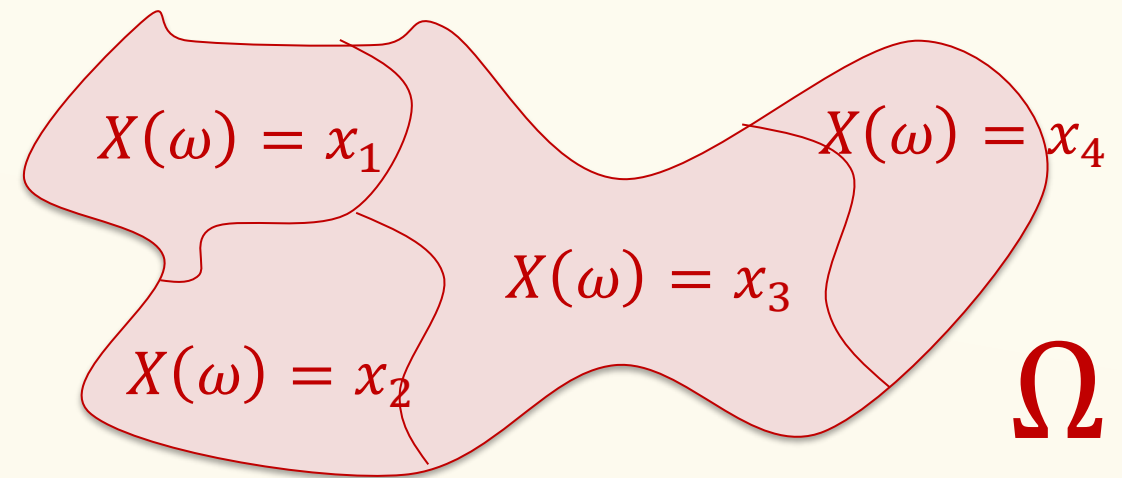


Random variables partition sample space

Definition. A **random variable (RV)** for a probability space (Ω, \mathbb{P}) is a function $X: \Omega \rightarrow \mathbb{R}$.

The set of values that X can take on is called its range/support Ω_X

Random variables partition the sample space.



Definition. For a RV $X: \Omega \rightarrow \mathbb{R}$, we define the event

$$\{X = x\} \stackrel{\text{def}}{=} \{\omega \in \Omega \mid X(\omega) = x\}$$

Agenda (2)

- Random Variables
- Probability Mass Function (pmf)
- Cumulative Distribution Function (CDF)

Probability Mass Function (PMF)

Definition. For a RV $X: \Omega \rightarrow \mathbb{R}$, we define the event

$$\{X = x\} \stackrel{\text{def}}{=} \{\omega \in \Omega \mid X(\omega) = x\}$$

The **probability mass function** (PMF) of X tells us the probabilities of these events, i.e., the probability that X takes each value in Ω_X

We use the notation

$$p_X(x) = \mathbb{P}(X = x) = \mathbb{P}(\{\omega \in \Omega \mid X(\omega) = x\})$$

For the probability mass function

$$\sum_{x \in \Omega_X} \mathbb{P}(X = x) = 1$$

Probability Mass Function – example 1

Flipping two independent coins $\Omega = \{HH, HT, TH, TT\}$

X = number of heads in the two flips

$$X(HH) = 2 \quad X(HT) = 1 \quad X(TH) = 1 \quad X(TT) = 0$$

$$\Omega_X = \{0, 1, 2\}$$

What is the pmf of X ?

Probability Mass Function

Flipping two independent coins

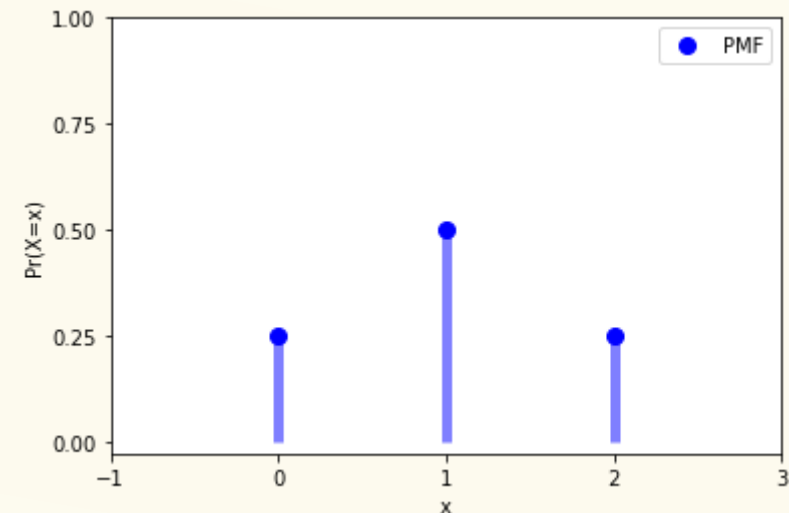
$$\Omega = \{HH, HT, TH, TT\}$$

X = number of heads in the two flips

$$X(HH) = 2 \quad X(HT) = 1 \quad X(TH) = 1 \quad X(TT) = 0$$

$$\Omega_X = \{0, 1, 2\}$$

$$p_X(x) = \Pr(X = x) = \begin{cases} \frac{1}{4}, & x = 0 \\ \frac{1}{2}, & x = 1 \\ \frac{1}{4}, & x = 2 \\ 0, & \text{o.w.} \end{cases}$$



RV Example

20 balls labeled 1, 2, ..., 20 in a bin

- Draw a subset of 3 uniformly at random
- Let $X =$ maximum of the 3 numbers on the balls

What is $p_X(20) = P(X = 20)$?

Poll:

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- A. $\frac{\binom{20}{2}}{\binom{20}{3}}$
- B. $\frac{\binom{19}{2}}{\binom{20}{3}}$
- C. $\frac{19^2}{\binom{20}{3}}$
- D. $\frac{19 \cdot 18}{\binom{20}{3}}$

Agenda (3)

- Random Variables
- Probability Mass Function (PMF)
- Cumulative Distribution Function (CDF)

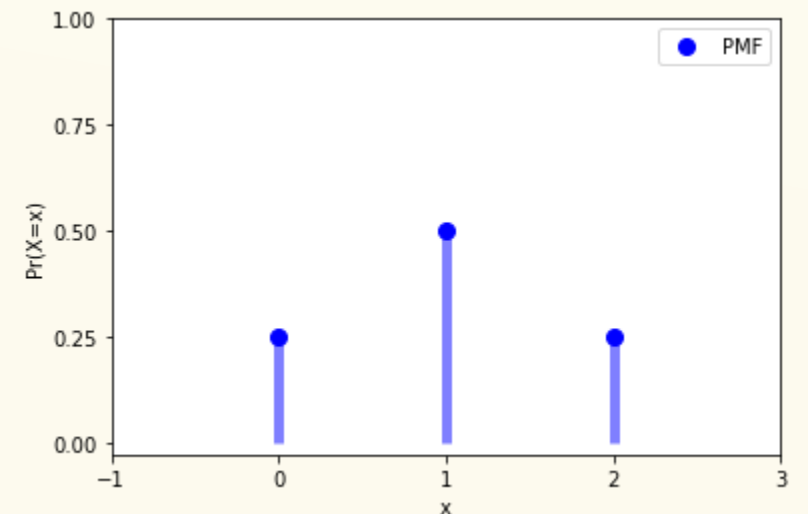
Cumulative Distribution Function (CDF)

Definition. For a RV $X: \Omega \rightarrow \mathbb{R}$, the **cumulative distribution function** of X specifies for any real number x , the probability that $X \leq x$.

$$F_X(x) = \Pr(X \leq x)$$

Recall the **probability mass function** of X , where X is the number of heads in 2 independent coin tosses.

$$p_X(x) = \Pr(X = x) = \begin{cases} 1 & x = 0 \\ \frac{1}{4}, & x = 0 \\ \frac{1}{2}, & x = 1 \\ \frac{1}{4}, & x = 2 \\ 0, & o.w. \end{cases}$$



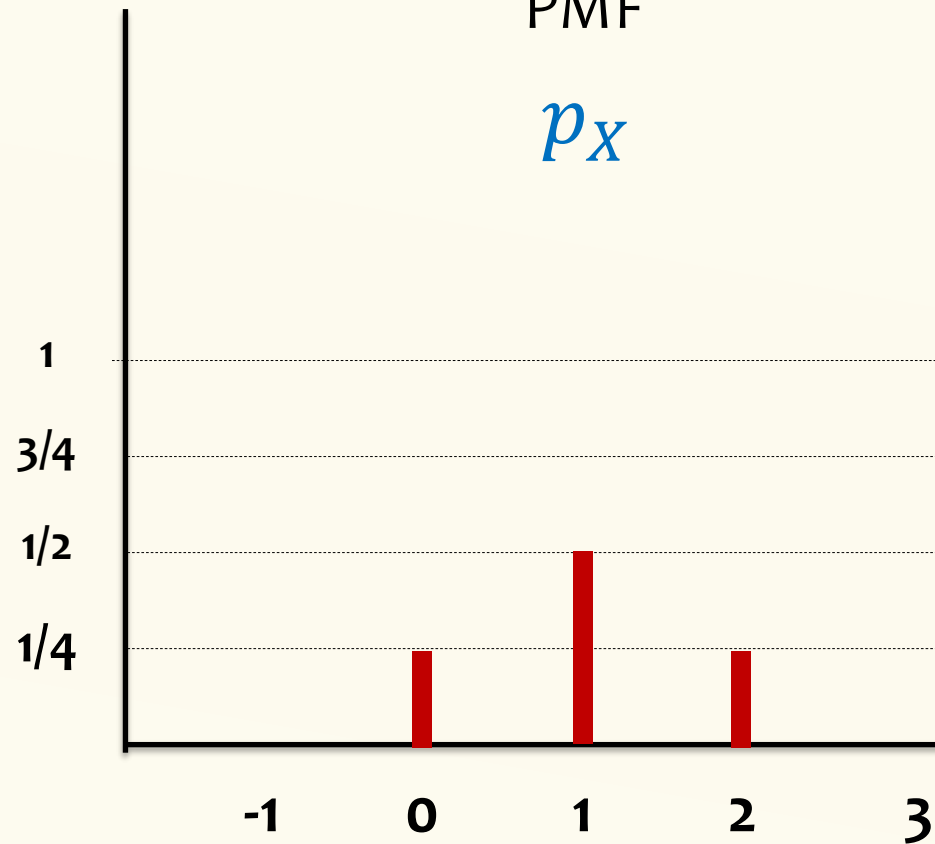
Example – Two fair independent coin flips

$X =$ number of heads

Probability Mass Function

PMF

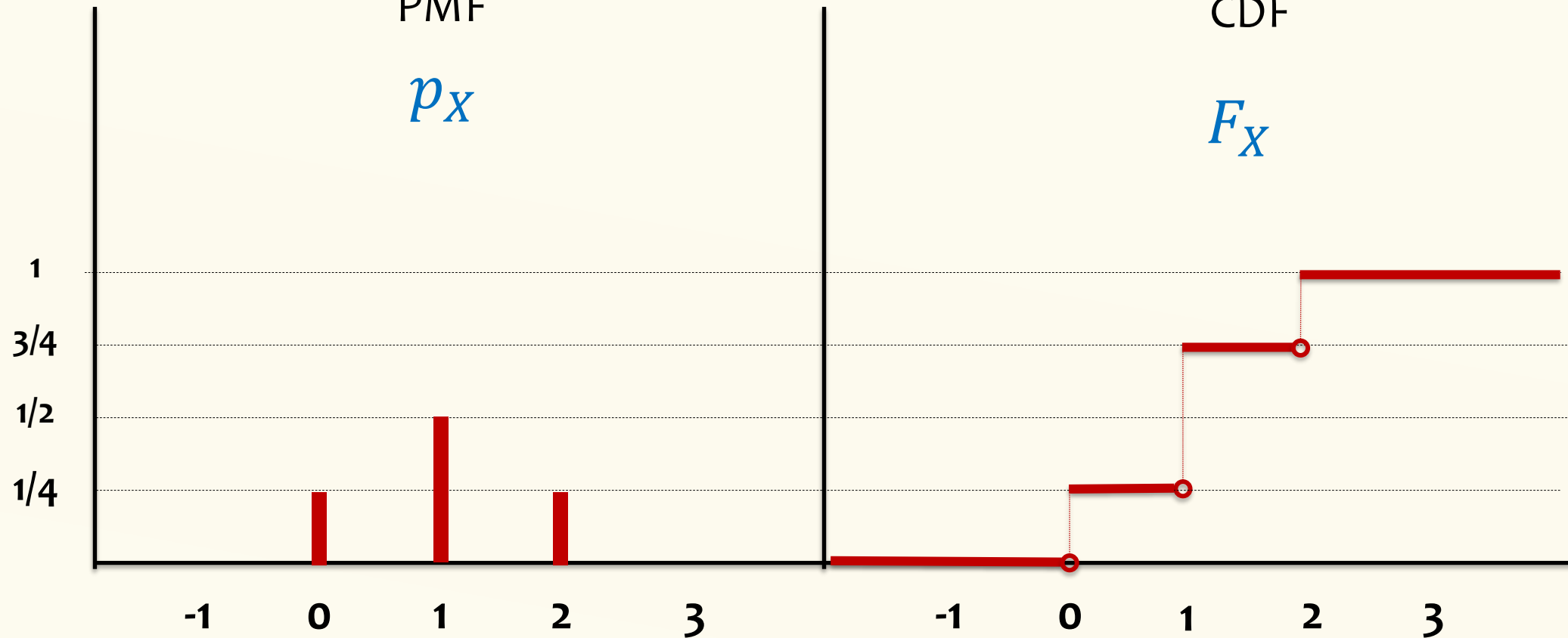
p_X



Cumulative Distribution Function

CDF

F_X



Cumulative Distribution Function (pmf and CDF for two coin tosses)

Definition. For a RV $X: \Omega \rightarrow \mathbb{R}$, the **cumulative distribution function** of X specifies for any real number x , the probability that $X \leq x$.

$$F_X(x) = \Pr(X \leq x)$$

Go back to 2 coin clips, where X is the number of heads

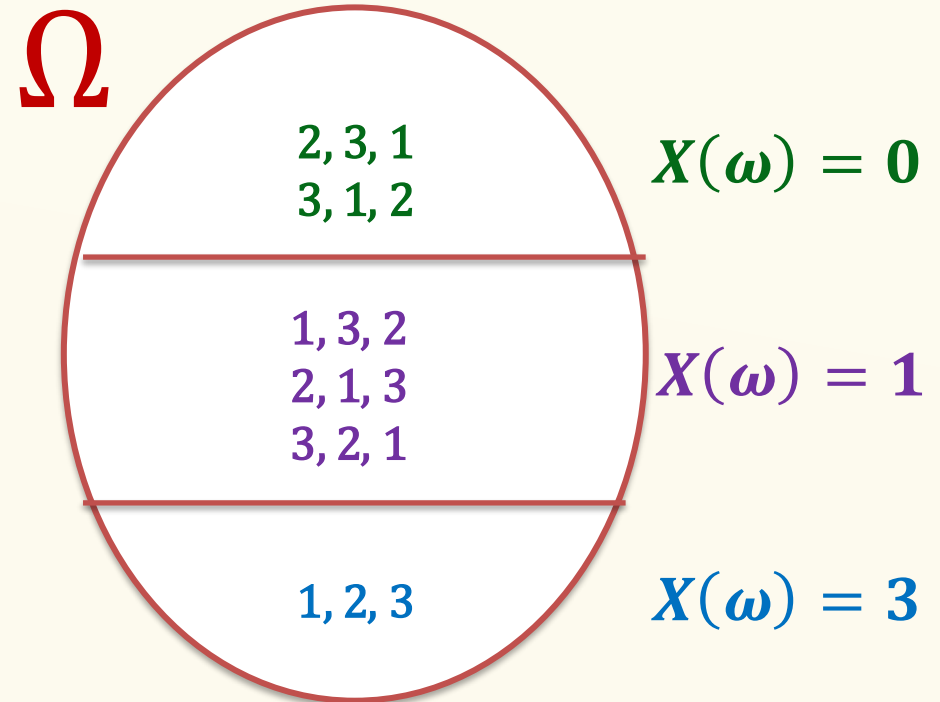
$$p_X(x) = \Pr(X = x) = \begin{cases} \frac{1}{4}, & x = 0 \\ \frac{1}{2}, & x = 1 \\ \frac{1}{4}, & x = 2 \\ 0, & \text{o.w.} \end{cases}$$

$$F_X(x) = \begin{cases} 0, & x < 0 \\ \frac{1}{4}, & 0 \leq x < 1 \\ \frac{3}{4}, & 1 \leq x < 2 \\ 1, & 2 \leq x \end{cases}$$

Example: Returning Homeworks – Partition of sample space

- Class with 3 students, randomly hand back homeworks. All permutations equally likely.
- Let X be the number of students who get their own HW

$\Pr(\omega)$	ω	$X(\omega)$
1/6	1, 2, 3	3
1/6	1, 3, 2	1
1/6	2, 1, 3	1
1/6	2, 3, 1	0
1/6	3, 1, 2	0
1/6	3, 2, 1	1



Example: Returning Homeworks –pmf and CDF

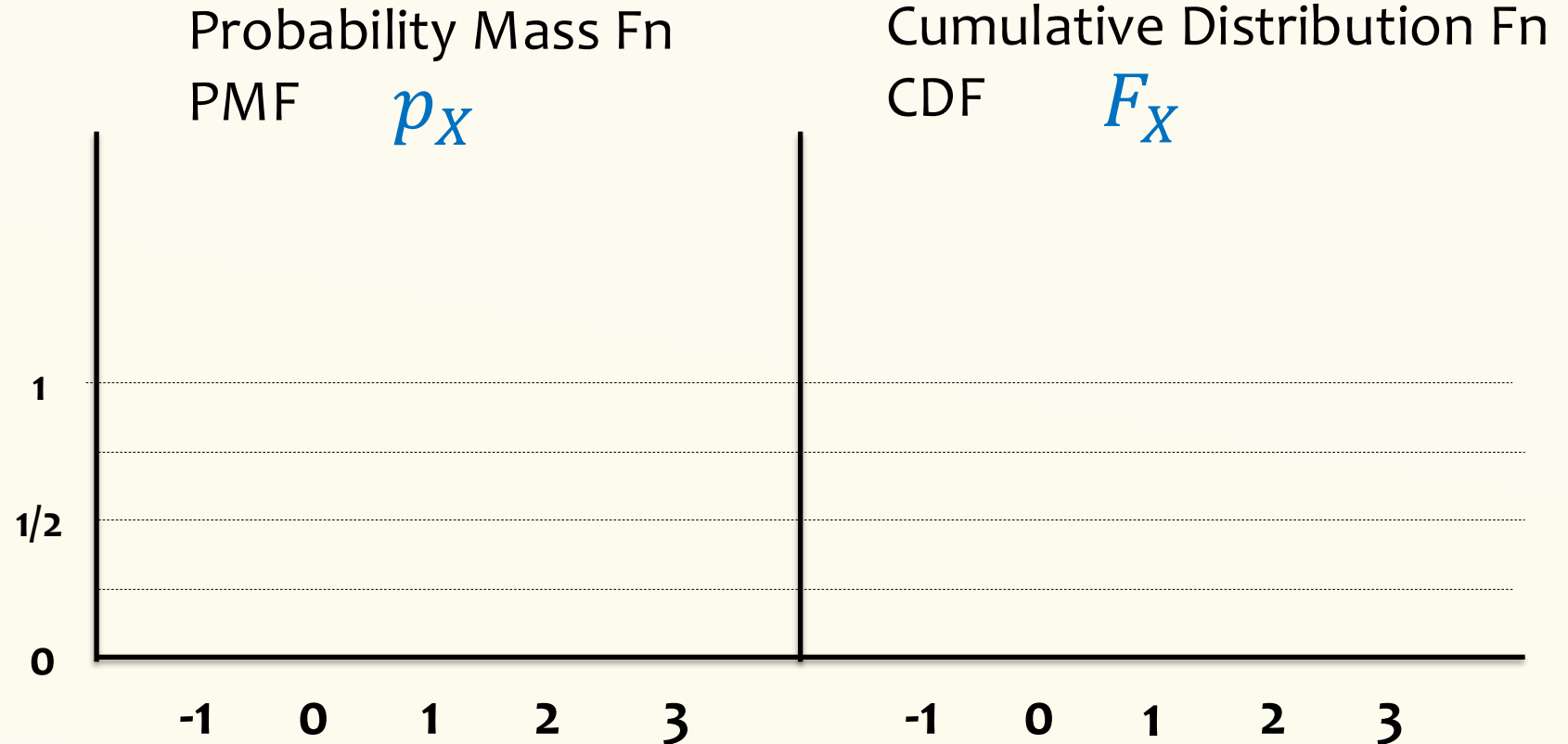
- Class with 3 students, randomly hand back homeworks. All permutations equally likely.
- Let X be the number of students who get their own HW

$\Pr(\omega)$	ω	$X(\omega)$
1/6	1, 2, 3	3
1/6	1, 3, 2	1
1/6	2, 1, 3	1
1/6	2, 3, 1	0
1/6	3, 1, 2	0
1/6	3, 2, 1	1

$$p_X(0) = P(X = 0) = 1/3$$

$$p_X(1) = P(X = 1) = 1/2$$

$$p_X(2) = P(X = 3) = 1/6$$



Example – Number of Heads

We flip n coins, independently, each heads with probability p

$$\Omega = \{\text{HH} \cdots \text{HH}, \text{HH} \cdots \text{HT}, \text{HH} \cdots \text{TH}, \dots, \text{TT} \cdots \text{TT}\}$$

$X = \#$ of heads

$$\Omega_X =$$

$$p_X(k) = P(X = k) =$$

Example – Number of Heads - pmf

We flip n coins, independently, each heads with probability p

$$\Omega = \{\text{HH} \cdots \text{HH}, \text{HH} \cdots \text{HT}, \text{HH} \cdots \text{TH}, \dots, \text{TT} \cdots \text{TT}\}$$

$X = \#$ of heads

$$p_X(k) = P(X = k) = \binom{n}{k} \cdot p^k \cdot (1 - p)^{n-k}$$

of sequences with k heads

Prob of sequence w/ k heads

Agenda (4)

- Random Variables
- Probability Mass Function (PMF)
- Cumulative Distribution Function (CDF)
- Expectation

Expectation (Idea)

Example. Toss a coin 20 times independently with probability $\frac{1}{4}$ of coming up heads on each toss.

X = number of heads

How many heads do you *expect* to see?

What if you toss it independently n times and it comes up heads with probability p each time?

Expected Value of a Random Variable

Definition. Given a discrete RV $X: \Omega \rightarrow \mathbb{R}$, the **expectation** or **expected value** or **mean** of X is

$$\mathbb{E}[X] = \sum_{\omega \in \Omega} X(\omega) \cdot P(\omega)$$

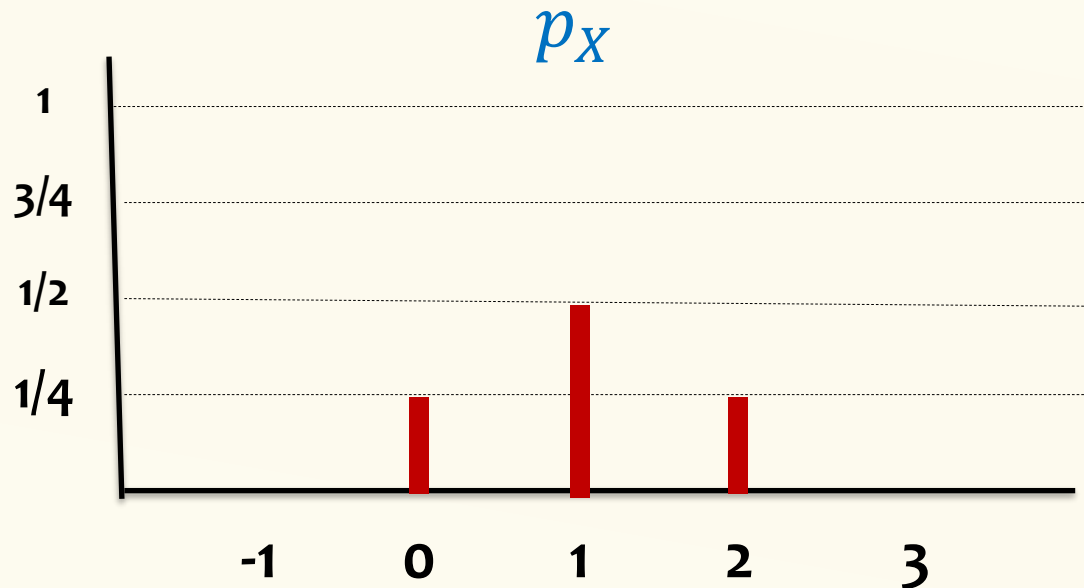
or equivalently

$$\mathbb{E}[X] = \sum_{x \in \Omega_X} x \cdot P(X = x) = \sum_{x \in \Omega_X} x \cdot p_X(x)$$

Intuition: “Weighted average” of the possible outcomes (weighted by probability)

Expectation

Example. Two fair coin flips
 $\Omega = \{TT, HT, TH, HH\}$
 $X =$ number of heads



$$\mathbb{E}[X] = \sum_{\omega \in \Omega} X(\omega) \cdot P(\omega)$$

$$\mathbb{E}[X] = \sum_{x \in \Omega_X} x \cdot P(X = x)$$

What is $\mathbb{E}[X]$?

$$\begin{aligned} \mathbb{E}[X] &= X(TT) P(TT) + X(HT) P(HT) \\ &\quad + X(TH) P(TH) + X(HH) P(HH) \end{aligned}$$

$$= 0 \cdot \frac{1}{4} + 1 \cdot \frac{1}{4} + 1 \cdot \frac{1}{4} + 2 \cdot \frac{1}{4} = 1$$

$$\mathbb{E}[X] = 0 \cdot p_X(0) + 1 \cdot p_X(1) + 2 \cdot p_X(2)$$

$$= 0 \cdot \frac{1}{4} + 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{4} = \frac{1}{2} + \frac{1}{2} = 1$$

