

1179: Probability

Lecture 7 — Random Variables, CDFs & PMFs, and Discrete Random Variables

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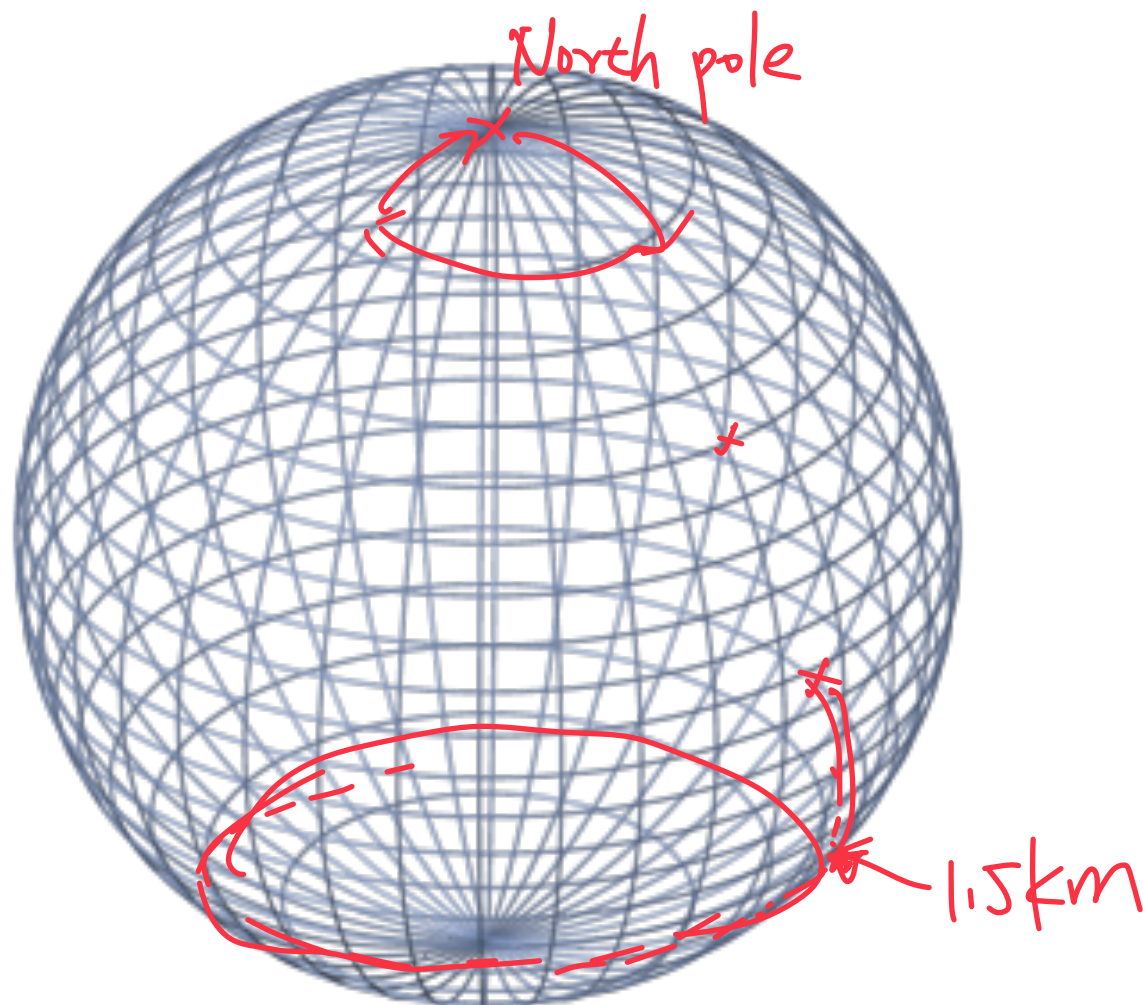
October 6, 2021

Announcement

- ▶ Weekly online office hours: 1pm-1:30pm on Wednesdays
 - ▶ <https://nycu.webex.com/nycu/j.php?MTID=ma2106f2503f60807a6dedb2d5d777756> (same as the Webex link for the lectures)
- ▶ Just send me an email if you would like to meet in some other time slots or in person

An Interview Question

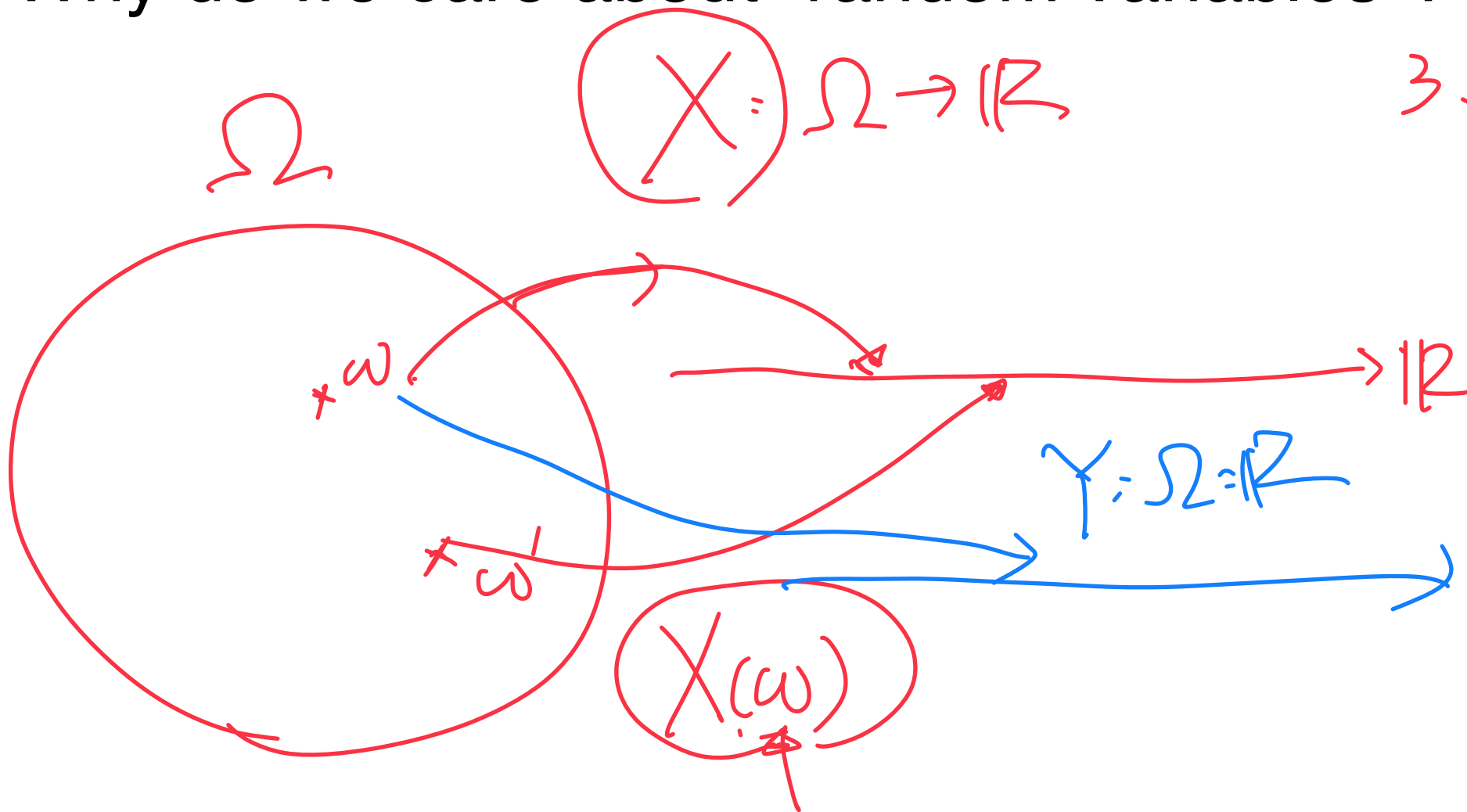
- ▶ Suppose Bill stands somewhere on the surface of the globe.
- ▶ Next, Bill goes 1.5km south, 1.5km west and then 1.5km north. It turns out that Bill returns to the same location.
- ▶ **Question:** Where is Bill?



Quick Review

- ▶ What is a “random variable”?
- ▶ Why do we care about “random variables”?

1. Simpler notations
2. Multiple ^{random} variables
3. R.V. capture common features of experiments



This Lecture

1. Random Variables and Cumulative Distribution Function (CDF)

2. Probability Mass Function (PMF)

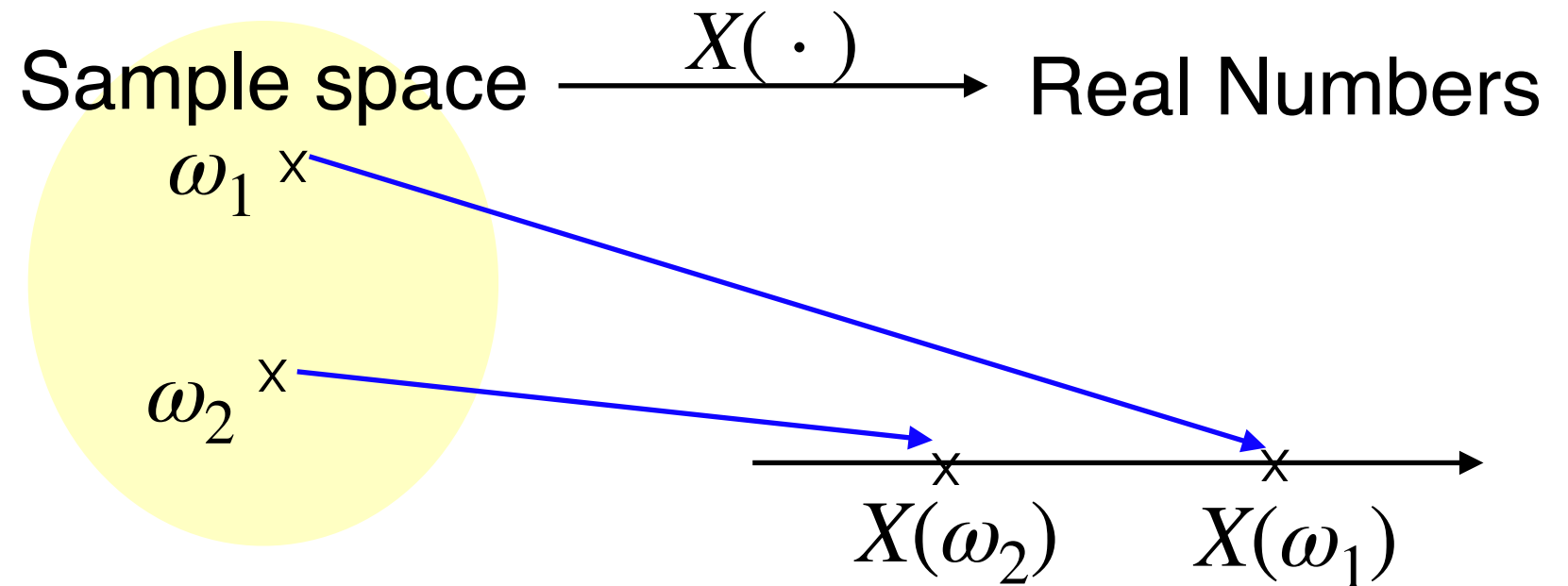
3. Special Discrete Random Variables

- Reading material: Chapter 4.1~4.3 and 5.1~5.2

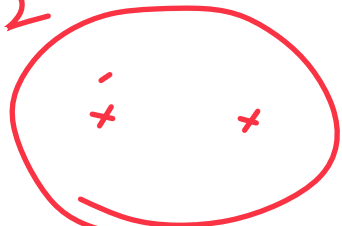
1. Random Variables

What is a Random Variable (Formally)?

- **Random variable**: a function that maps each outcome to a real number



- **Example**: Whether NCTU will merge with NYMU

Ω  $X(\omega = \underline{\text{"Yes"}}) = 1, \quad X(\omega = \underline{\text{"No"}}) = 0$

- **Example**: # of people waiting in line at Shinemood

$Y(n \text{ people}) = N$

Function of a Random Variable

$$\Omega \xrightarrow{X} \mathbb{R} \xrightarrow{f} \mathbb{R}$$

► **Example:** Buy a waffle at Shinemood

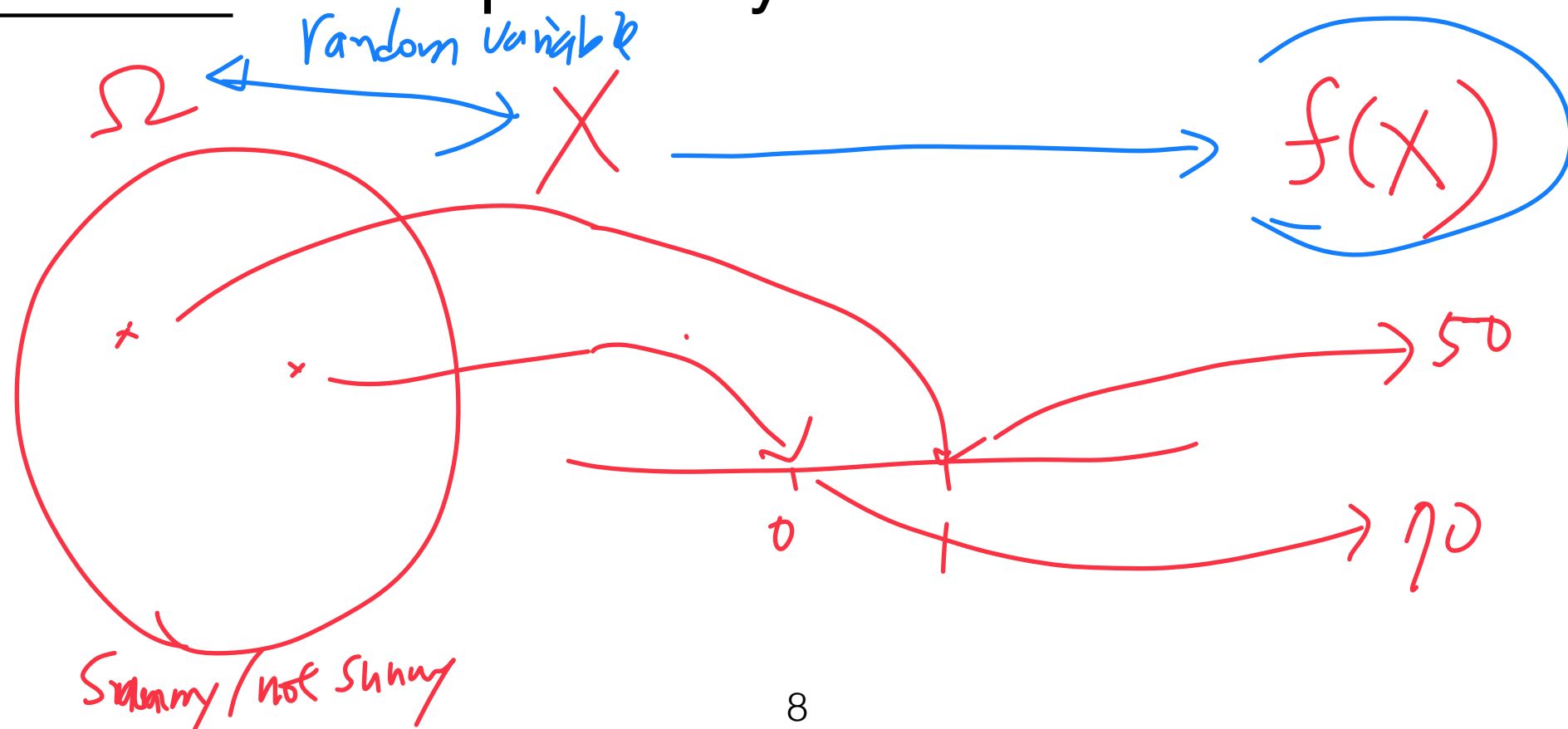
*X is a random variable
f(X) "also" " " " "*

► If it is sunny today, then you spend \$50 to order a Matcha-red-bean waffle

► Otherwise, you spend \$70 to order a Fried-chicken waffle

► Question: Is the price of your waffle a r.v.?

Yes!

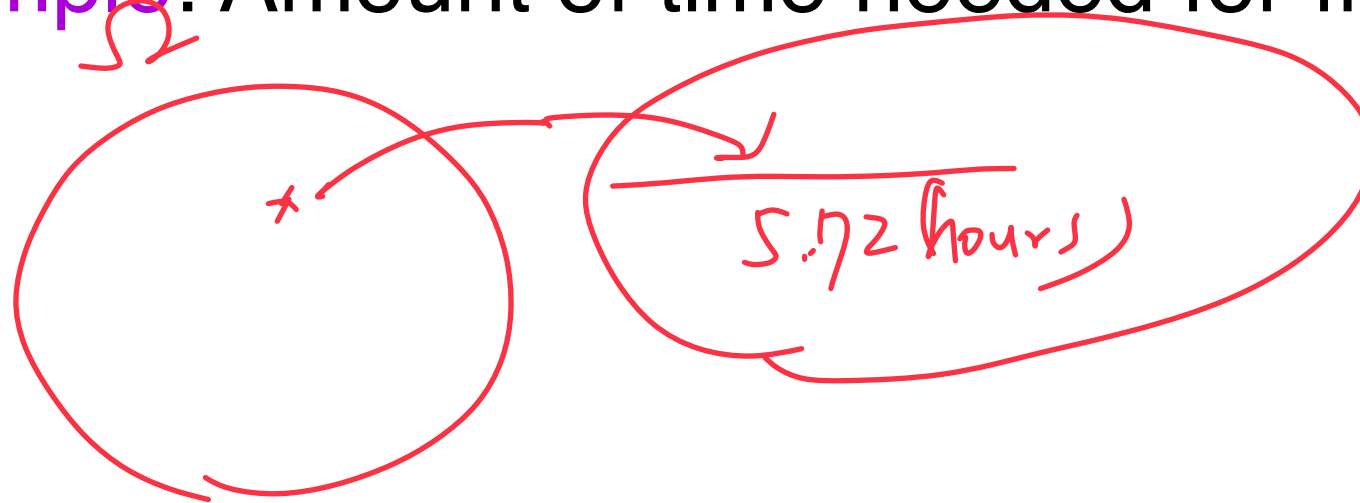


Discrete and Continuous Random Variables

- ▶ **Example:** # of people waiting in line at Shinemood

Discrete

- ▶ **Example:** Amount of time needed for finishing HW1



Continuous

Cumulative Distribution Function (CDF)

- Random variables are used to calculate the probabilities of events.

Cumulative Distribution Function (CDF): For any random variable X , the CDF of X is defined as:

$$F_X(t) = P(X \leq t), \text{ for all } t \in \mathbb{R}$$

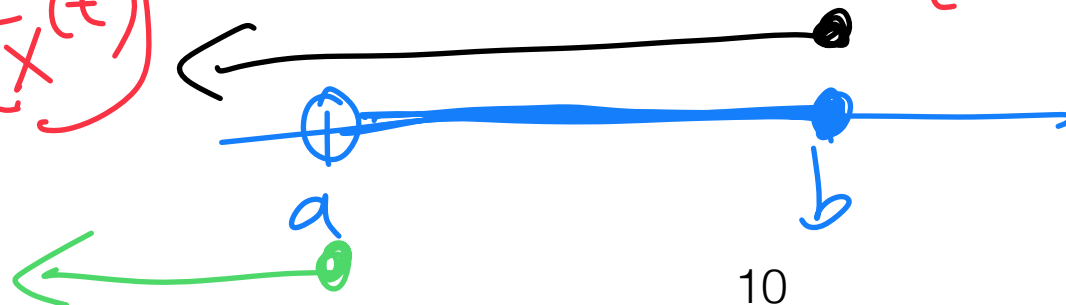
- What's the range of $F_X(t)$?

$\in [0, 1]$

- How to use the CDF?

- Example:** $P(a < X \leq b) = ?$

CDF of X : $F_X(t)$



CDF of a Discrete Random Variable

- ▶ **Example:** Roll a fair 4-sided die

- ▶ $P(X = 1) = P(X = 2) = P(X = 3) = P(X = 4) = 1/4$
- ▶ What is the CDF of X ?

$$F_X(t) = P(X \leq t)$$

$$F_X(0) = 0$$

$$F_X(1) = \frac{1}{4}$$

$$F_X(1.5) = \frac{1}{4}$$

$$F_X(1.9999) = \frac{1}{4}$$

$$F_X(2) = \frac{2}{4}$$

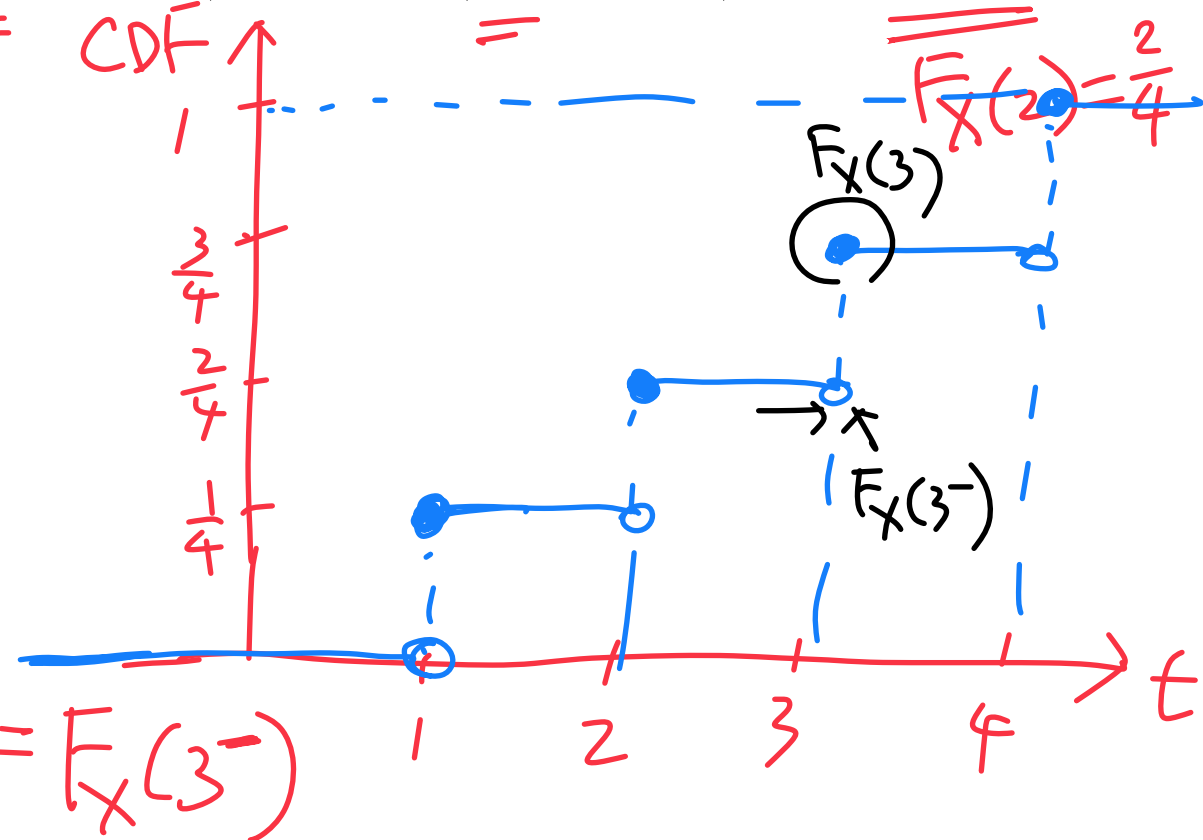
$$1. \ P(X \leq 3) = \frac{3}{4} = F_X(3)$$

$$2. \ P(X < 3) = \frac{2}{4} = F_X(3) - P(X=3) = F_X(3^-)$$

$$3. \ P(1 < X \leq 3) = \frac{2}{4} = F_X(3) - F_X(1)$$

$$4. \ P(1 < X < 3) = \frac{1}{4} = F_X(3^-) - F_X(1)$$

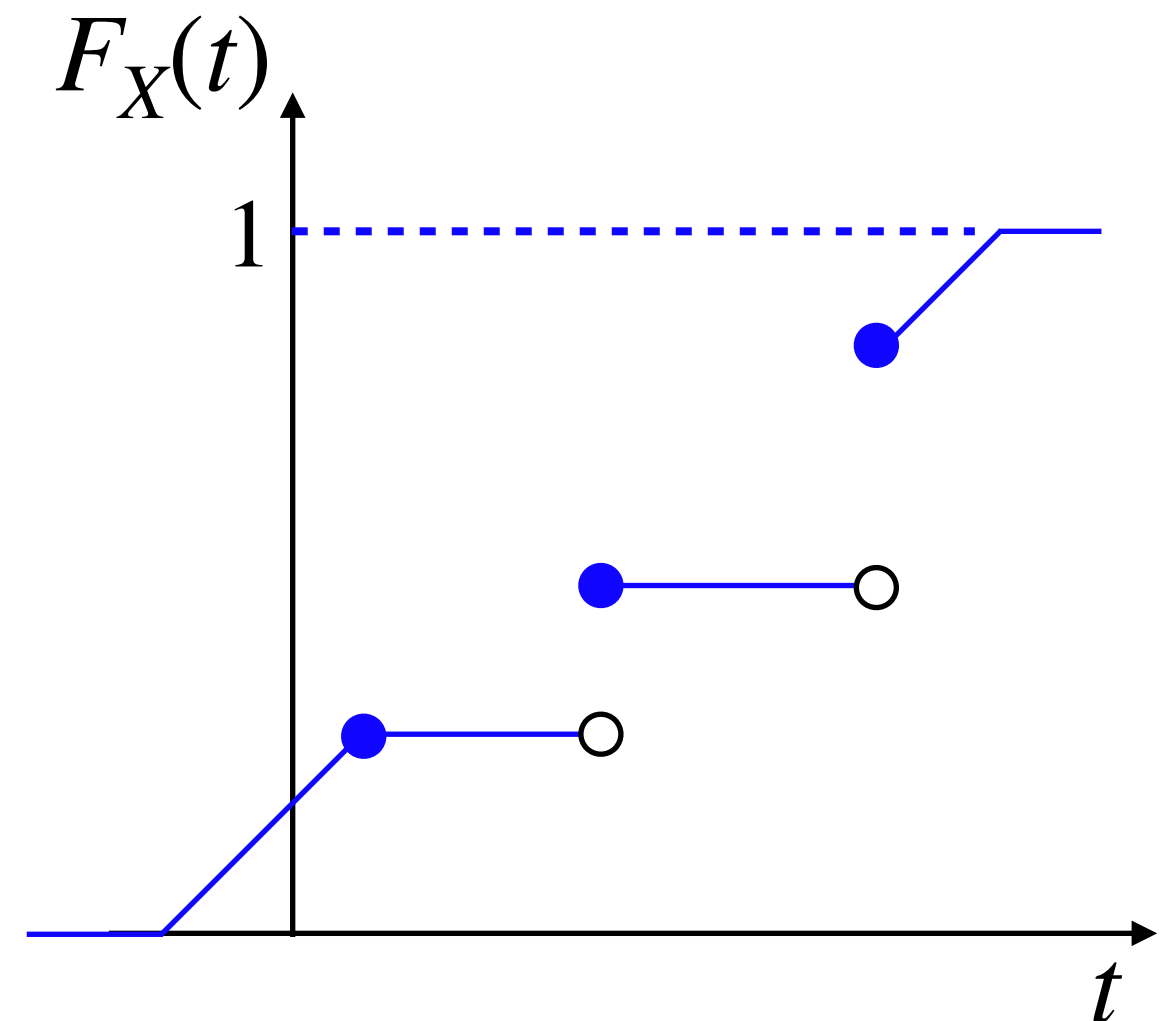
$$5. \ P(X = 3) = \frac{1}{4} = F_X(3) - F_X(3^-)$$



Use CDF to Find Probability of an Event (I)

$$F_X(t) = P(X \leq t), \text{ for all } t \in \mathbb{R}$$

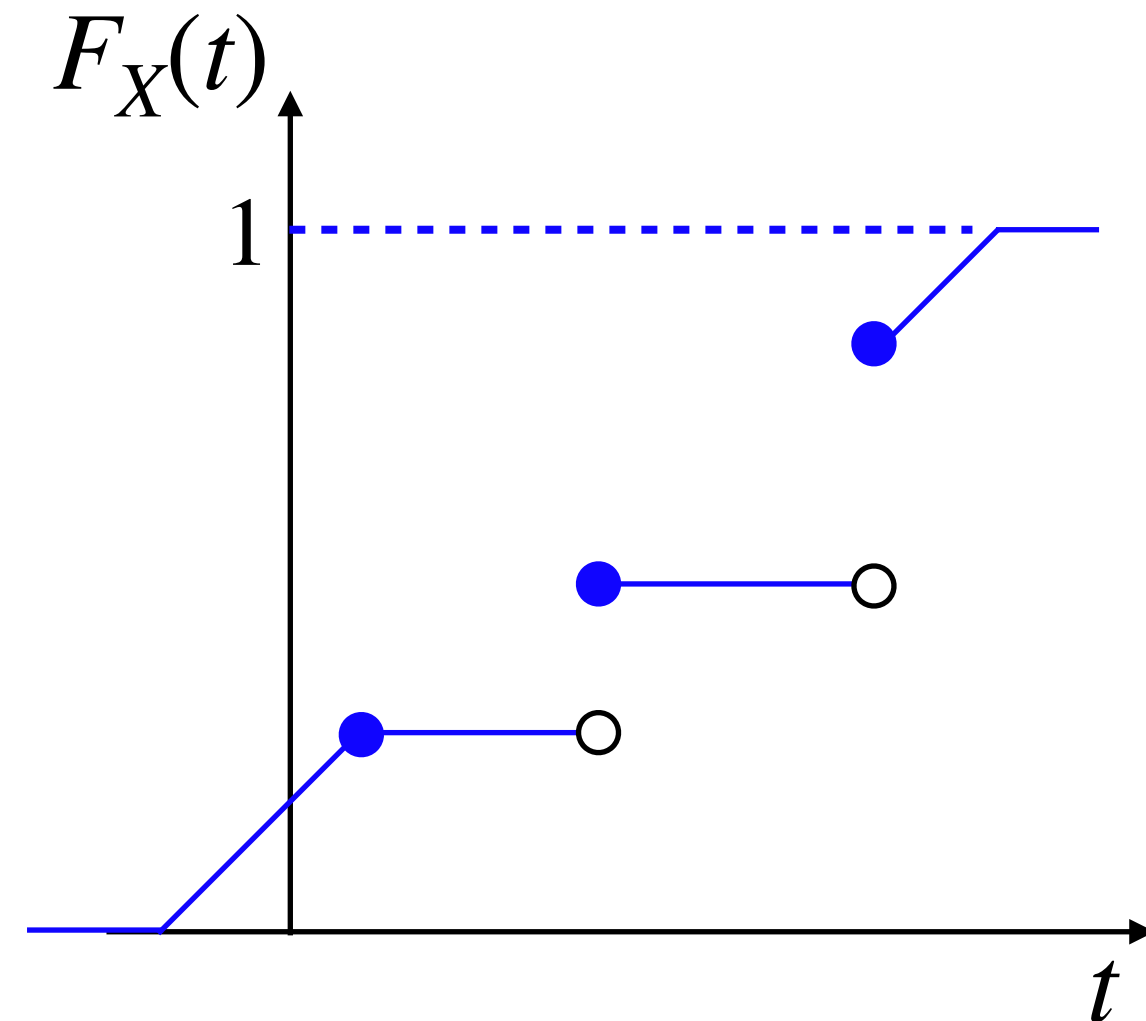
Event	Probability of the event
$X \leq a$	$F_X(a)$
$X > a$	$1 - F_X(a)$
$X < a$	$F_X(a^-) \equiv F_X(a) - P(X=a)$
$X \geq a$	$1 - F_X(a^-)$
$X = a$	$F_X(a) - F_X(a^-)$



Use CDF to Find Probability of an Event (II)

$$F_X(t) = P(X \leq t), \text{ for all } t \in \mathbb{R}$$

Event	Probability of the event
$a < X \leq b$	$F_X(b) - F_X(a)$
$a < X < b$	$F_X(b^-) - F_X(a)$
<u>$a \leq X \leq b$</u>	$F_X(b) - F_X(a^-)$
$a \leq X < b$	$F_X(b^-) - F_X(a^-)$



Example: Use CDF to Find Probability

► **Example:** The CDF of a random variable X is given by

$$F_X(x) = \begin{cases} 0, & x < 0 \\ x/4, & 0 \leq x < 1 \\ 1/2, & 1 \leq x < 2 \\ \frac{1}{8}x + \frac{1}{2}, & 2 \leq x < 3 \\ 1, & x \geq 3 \end{cases}$$

1. $P(X < 2) =$

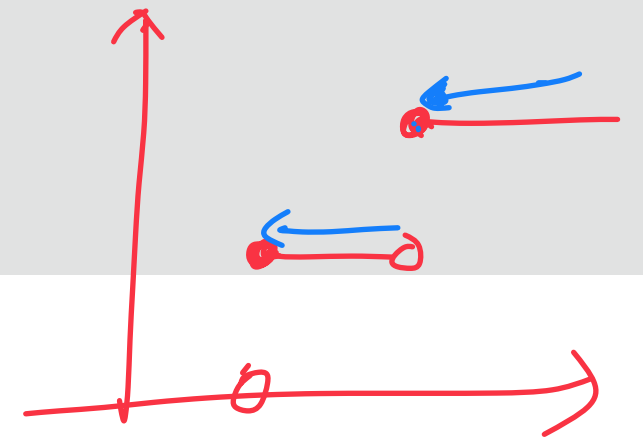
2. $P(1 \leq X < 3) =$

3. $P(X > 3/2) =$

4. $P(X = 2) =$

Properties of a Valid CDF

$$F_X(t) = P(X \leq t), \text{ for all } t \in \mathbb{R}$$



Properties of a CDF:

1. $F_X(t)$ is non-decreasing

$$t' \geq t \Rightarrow F_X(t') \geq F_X(t)$$

2. $\lim_{t \rightarrow \infty} F_X(t) = 1$

$$F_X(\infty) \equiv P(X \leq \infty) = P(\{\omega : X(\omega) \leq \infty\}) = P(\Omega) = 1$$

3. $\lim_{t \rightarrow -\infty} F_X(t) = 0$

$$F_X(-\infty) = P(\{\omega : X(\omega) \leq -\infty\}) = P(\emptyset) = 0$$

4. $F_X(t)$ is **right-continuous** ($\underline{F_X(t^+) = F_X(t)}$)

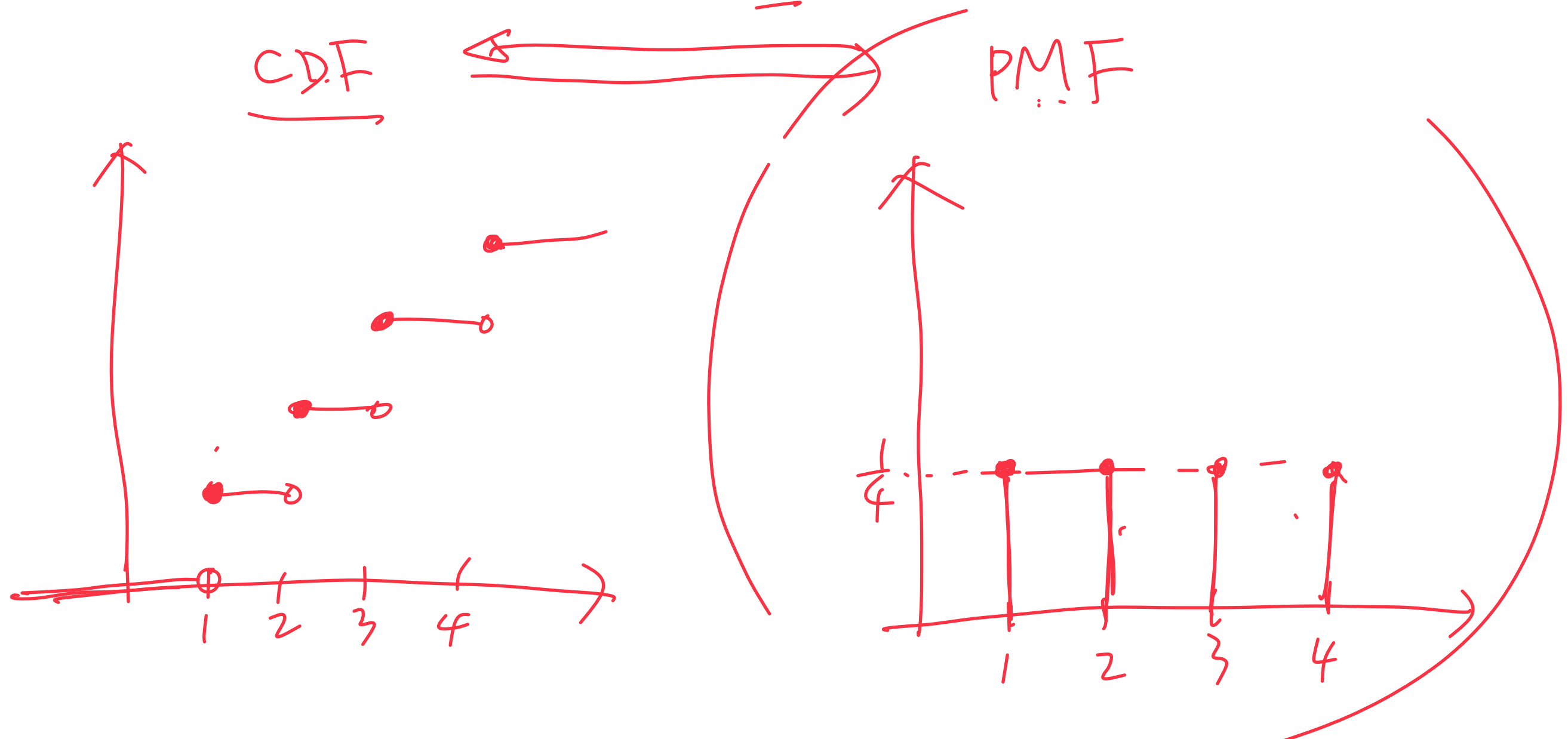
2. Probability Mass Function (PMF)

PMF: Another Way to Specify CDF of a Discrete Random Variable

- ▶ **Example:** Roll a fair 4-sided die once

$$P(X=1) = P(X=2) = P(X=3) = P(X=4) = \frac{1}{4}$$

- ▶ Define a random variable X = the number that we observe



Probability Mass Function (PMF)

Probability Mass Function (PMF): For any discrete random variable X with possible values $\{x_1, x_2, x_3, \dots\}$, the PMF $p(\cdot)$ of X is a function that satisfies: *finite or countably infinite*

$$(1) \underline{p(x_i)} = \underline{P(X = x_i)}$$

$$(2) \underline{p(x) = 0, \text{ if } x \notin \{x_1, x_2, x_3, \dots\}}$$

$$(3) \sum_{i=1}^{\infty} p(x_i) = 1$$

- For discrete random variables: CDF \Leftrightarrow PMF

Example: From CDF to PMF

- ▶ **Example:** Given the CDF of a discrete random variable X as:

$$F_X(t) = \begin{cases} 0, & x < 1 \\ 1/36, & 1 \leq x < 2 \\ 4/36, & 2 \leq x < 3 \\ 9/36, & 3 \leq x < 4 \\ 16/36, & 4 \leq x < 5 \\ 25/36, & 5 \leq x < 6 \\ 1, & x \geq 6 \end{cases}$$

- ▶ What is the PMF of X ?

PMF:

$$p(1) = \frac{1}{36}$$

$$p(2) = \frac{3}{36}$$

$$p(3) = \frac{5}{36}$$

$$p(4) = \frac{7}{36}$$

$$p(5) = \frac{9}{36}$$

$$p(6) = \frac{11}{36}$$

$p(x) = 0$ for all
 $x \notin \{1, 2, 3, 4, 5, 6\}$

$$p(x) = \begin{cases} \frac{1}{36}, & \text{if } x=1 \\ \frac{3}{36}, & \text{if } x=2 \\ \frac{5}{36}, & \text{if } x=3 \\ \frac{7}{36}, & \text{if } x=4 \\ \frac{9}{36}, & \text{if } x=5 \\ \frac{11}{36}, & \text{if } x=6 \\ 0, & \text{else} \end{cases}$$

Example: From PMF to CDF

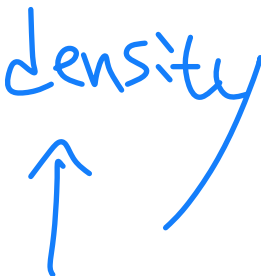
- **Example:** Given the PMF of a discrete random variable X as:

$$p(x) = \begin{cases} \underline{1/6}, & x = -2 \checkmark \\ \underline{1/3}, & x = 1 \checkmark \\ 1/4, & x = 5 \checkmark \\ 1/4, & x = 10 \checkmark \end{cases}$$

- What is the CDF of X ?

$$F_X(t) = \begin{cases} 0 & \text{if } t < -2 \\ \frac{1}{6} & \text{if } -2 \leq t < 1 \\ \frac{1}{6} + \frac{1}{3} & \text{if } 1 \leq t < 5 \\ \frac{1}{6} + \frac{1}{3} + \frac{1}{4} & \text{if } 5 \leq t < 10 \\ \frac{1}{6} + \frac{1}{3} + \frac{1}{4} + \frac{1}{4} & \text{if } t \geq 10 \end{cases}$$

Probability Distribution

- ▶ Discrete random variables: CDF or PMF
 - ▶ Continuous random variables: CDF or PDF (will be discussed in the next few lectures)
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2. Special Discrete Random Variables

Experiments With 2 Possible Outcomes

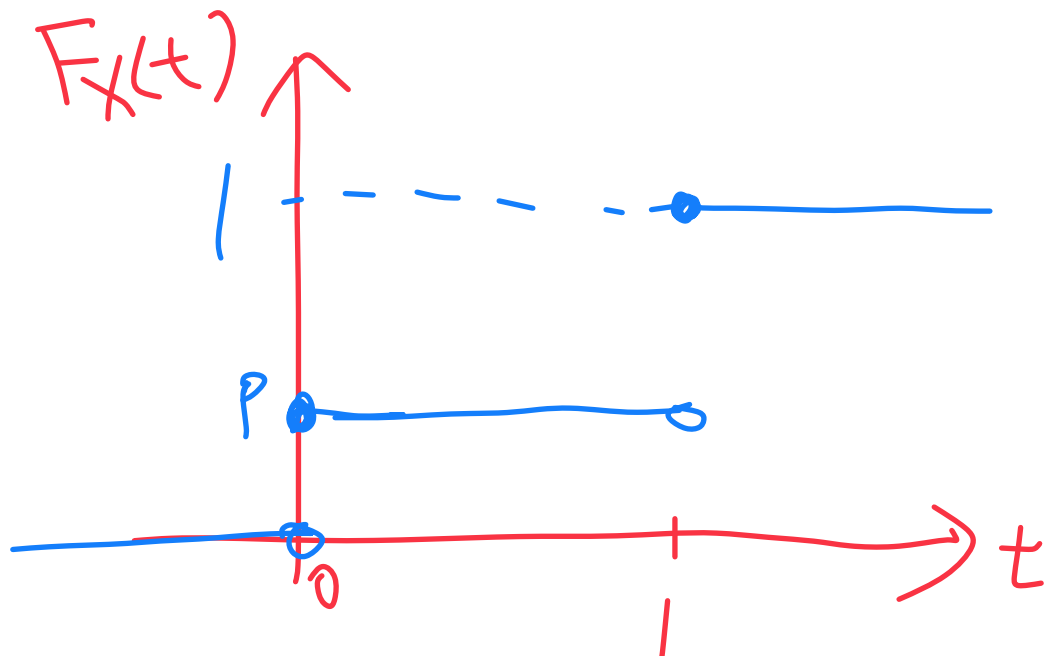
- ▶ **Example:** Whether an image of dog is classified correctly
- ▶ **Example:** Whether NCTU will merge with NYMU
- ▶ **Example:** Whether the 3rd student leaving the classroom wears glasses
- ▶ **Example:** Toss a coin once. Head or tail? $\overset{p}{\text{Head}}$ $\overset{1-p}{\text{Tail}}$ $X = 0, 1$
- ▶ What are the common features?
 - ▶ 1 experiment trial (no repetition) with 2 possible outcomes
 - ▶ 1 probability parameter
 - ▶ Want: Whether a specific outcome occurs

1. Bernoulli Random Variables (Formally)

Bernoulli Random Variables: A random variable X is Bernoulli with parameter p if its PMF is given by

$$P(X = k) = \begin{cases} p, & \text{if } k = 1 \\ 1 - p, & \text{if } k = 0 \\ 0, & \text{otherwise} \end{cases}$$

- How about its CDF?



Jacob Bernoulli

2. Binomial Random Variables

(all trials are independent from each other)

- ▶ **Example:** Play the same claw machine for 5 times, and each trial is successful with probability 0.7. What is $P(\text{win 3 toys})$?
- ▶ **Example:** Stephen Curry makes 20 free throws, and each throw is good with probability 0.95. What is $P(\text{he missed 2 throws})$?
- ▶ **Example:** Consider n independent coin tosses with head probability p . What is $P(\text{we observe 2 heads})$?
- ▶ What are the common features?
 - ▶ n repetitions of the same Bernoulli experiment
 - ▶ Want: how many successes in n repetitions

2. PMF of Binomial Random Variables

- ▶ **Example:** Play the same claw machine for 5 times, and each trial is successful with probability 0.7. All trials are independent.

- ▶ Define a r.v. X = the number of toys we get X can be 0, 1, 2, 3, 4, 5

- ▶ What is the PMF of X ? $P(X=k)$

$$P(X=k) = \begin{cases} C_0^5 \cdot (0.7)^0 \cdot (0.3)^5 & , \text{ if } X=0 \\ C_1^5 \cdot (0.7)^1 \cdot (0.3)^4 & , \text{ if } X=1 \\ C_2^5 \cdot (0.7)^2 \cdot (0.3)^3 & , \text{ if } X=2 \\ C_3^5 \cdot (0.7)^3 \cdot (0.3)^2 & , \text{ if } X=3 \\ C_4^5 \cdot (0.7)^4 \cdot (0.3)^1 & , \text{ if } X=4 \\ C_5^5 \cdot (0.7)^5 \cdot (0.3)^0 & , \text{ if } X=5 \\ 0 & , \text{ otherwise} \end{cases}$$

2. Binomial Random Variables (Formally)

Claw machine: $(X \sim \text{Binomial}(5, 0.7))$

Binomial Random Variables: A random variable X is Binomial with parameters (n, p) if its PMF is given by

$$\underline{P(X = k)} = \begin{cases} C_k^n p^k (1-p)^{n-k}, & \text{if } k = 0, 1, 2, \dots, n \\ 0, & \text{otherwise} \end{cases}$$

Do we have $\sum_{k=0}^n P(X = k) = 1$?

$$\sum_{k=0}^n C_k^n p^k (1-p)^{n-k} = 1$$

Binomial expansion $(p + (1-p))^n$

What is a Binomial r.v. with parameter $n = 1$?

$$\text{Binomial}(n=1, p) \equiv \text{Bernoulli}(p)$$

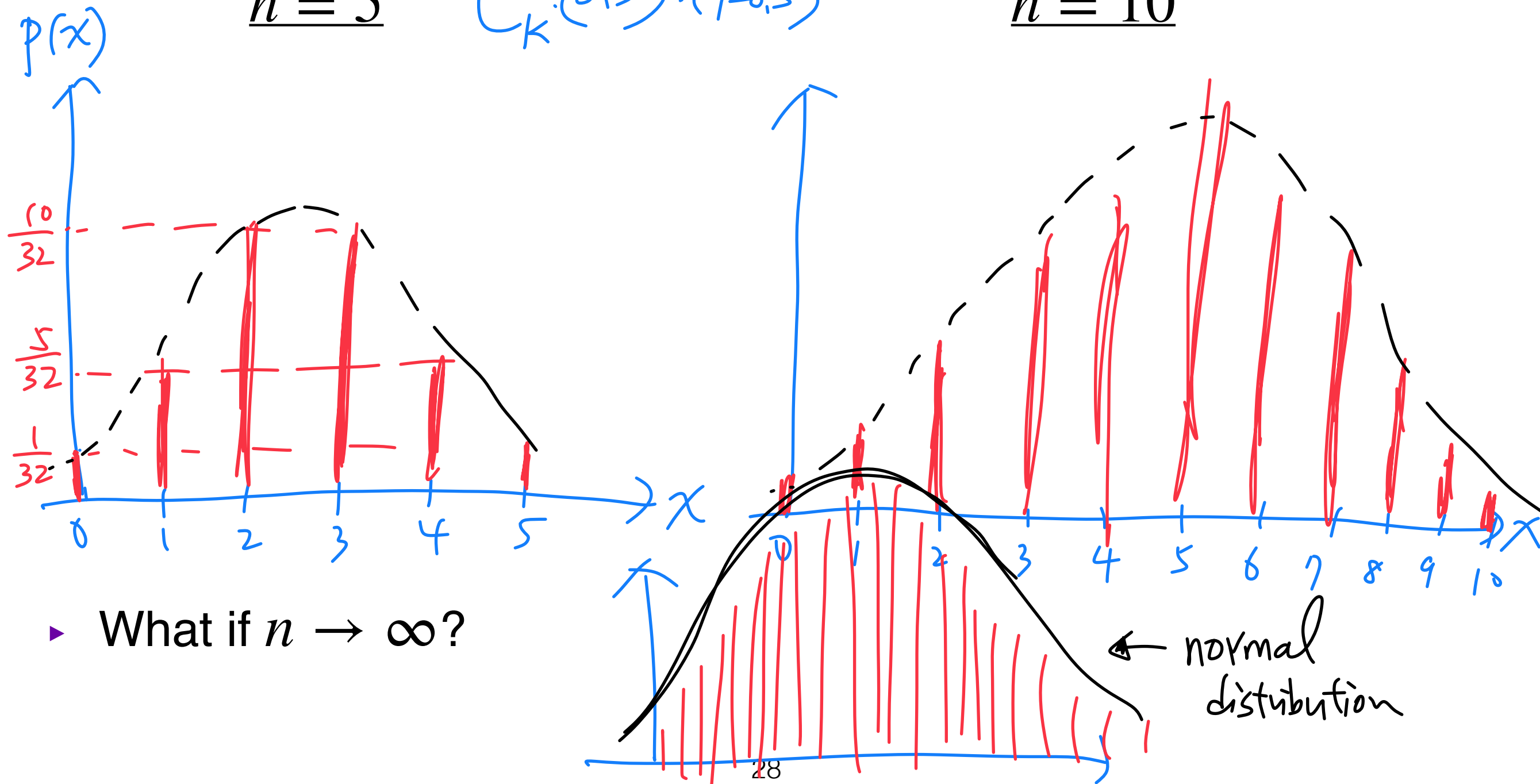
PMFs of Binomial Random Variables

- ▶ **Example:** Let's plot the PMF of $X \sim \text{Binomial}(n, p = 0.5)$

$n = 5$

$$C_K^5 \cdot (0.5)^K \cdot (1-0.5)^{5-K}$$

$n = 10$



- ▶ What if $n \rightarrow \infty$?

Example: Ensemble Learning With Voting

- ▶ **Example:** Suppose that we train an image classifier with either 1 or 3 ensembles and then apply majority voting

- ▶ Suppose each classifier makes the correct prediction with probability p

- ▶ Shall we choose 1 or 3 ensembles?

$$(p > \frac{1}{2})$$

$$p > p^3 + 3p^2(1-p)$$

$$(p < \frac{1}{2})$$

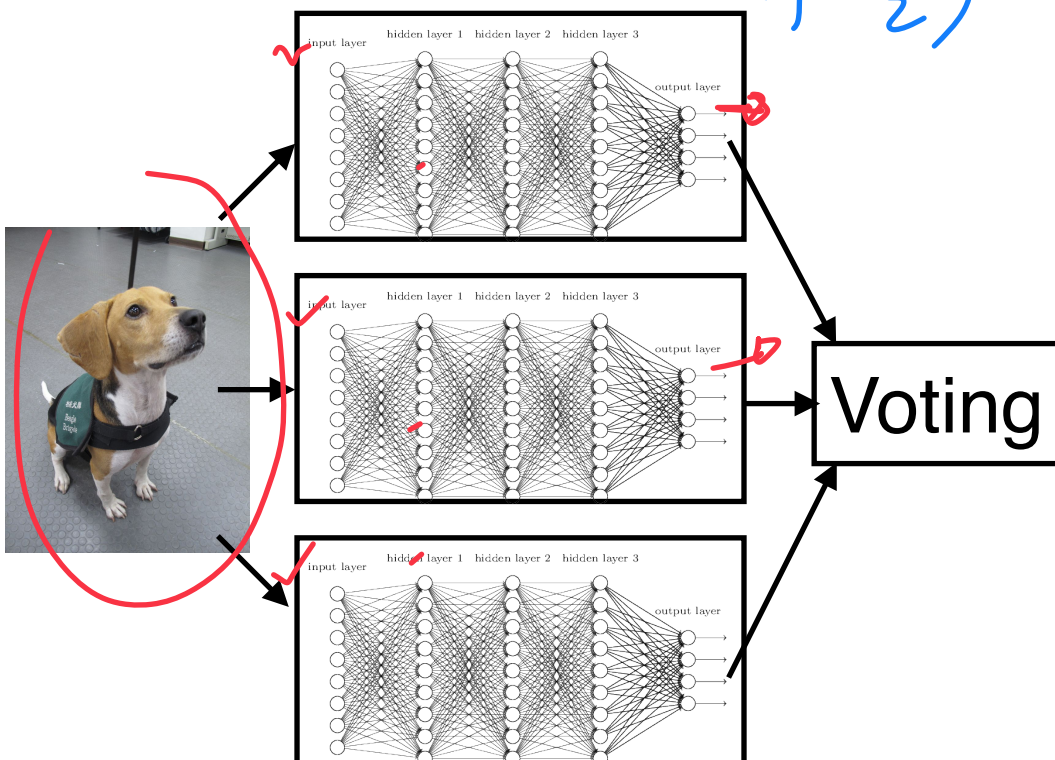
① 1 ensemble =

Prediction accuracy = p

② 3 ensembles:

• 3 classifiers are all correct: $p \cdot p \cdot p = p^3$
 $= C_3^3 \cdot p^3$

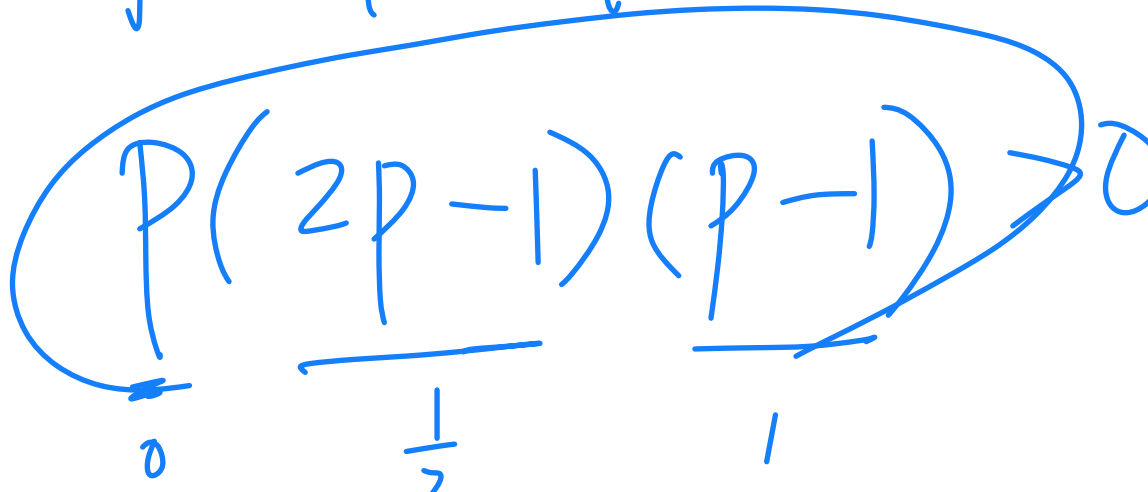
• 2 out of them are correct: $C_2^3 \cdot p^2 \cdot (1-p)$
 $= 3p^2(1-p)$



$$(p > p^3 + 3p^2(1-p)) = -2p^3 + 3p^2$$

$$\Leftrightarrow \underbrace{2p^3 - 3p^2 + p}_{> 0}$$

$$\Leftrightarrow p(2p^2 - 3p + 1) > 0$$


$$\Leftrightarrow p(2p-1)(p-1) > 0$$


$$\Leftrightarrow \cancel{p > 1} \text{ or } \underline{0 < p < \frac{1}{2}}$$

Example: A Poll of Coriander Lovers

- ▶ **Example:** Let p = probability that a random person likes coriander.
 - ▶ Suppose we randomly sample N people and define a random variable $X = \{\text{number of coriander lovers in } N \text{ people}\}$
 - ▶ For a fixed integer k , under what value of p is $P(X = k)$ maximized?



3. Poisson Random Variables

- ▶ **Example**: On average, 20 people stop by Shinemood every hour. What is $P(\text{exactly 100 people visit Shinemood in 3 hours})$?
- ▶ **Example**: On average, 1000 MayDay's concert tickets are sold every second. What is $P(\text{all 50k tickets are sold out in 1 min})$?
- ▶ What are the common features?
 - ▶ **Average rate** is known and static
 - ▶ Want: how many occurrences in an observation window?

Poisson: Limiting Case of Binomial

- ▶ **Example:** Consider $X \sim \text{Binomial}(n, p = \lambda/n)$, λ is a constant
 - ▶ What is $P(X = k)$?
 - ▶ What if $n \rightarrow \infty$?

3. Poisson Random Variables (Formally)

Poisson Random Variables: Given parameters

- λ : average rate
- T : duration of the observation window

A random variable X is Poisson with parameter λT if its PMF is given by

$$P(X = n) = \frac{e^{-\lambda T} (\lambda T)^n}{n!}, \quad n = 0, 1, 2, 3, \dots$$

► Do we have $\sum_{n=0}^{\infty} P(X = n) = 1$?

1-Minute Summary

1. Random Variables and CDF

- Function from outcomes to real numbers
- Use CDF to find the probability of an event

2. Probability Mass Function (PMF)

- An alternative way to specify the distribution of a discrete random variable

3. Special Discrete Random Variables

- Bernoulli / Binomial / Poisson