

1179: Probability

Lecture 13 — Continuous Random Variables

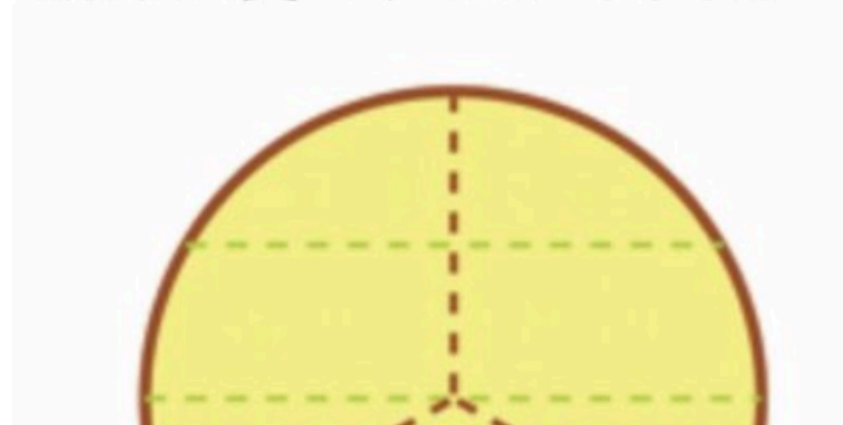
Ping-Chun Hsieh (謝秉均)

October 27, 2021

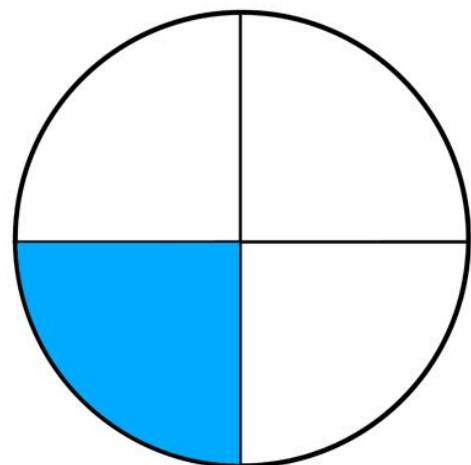
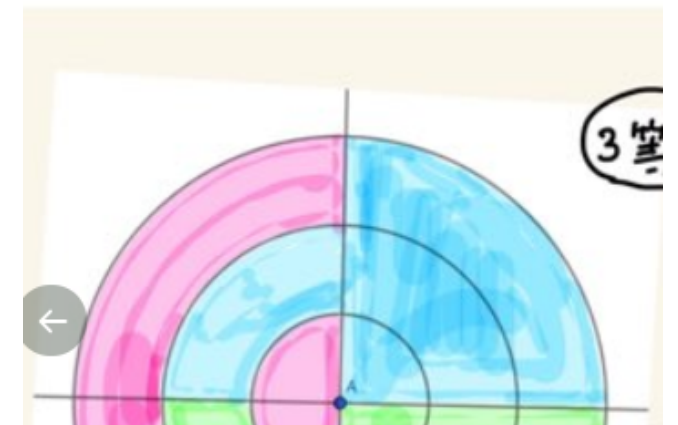
How to Cut a Cake into 3 Equal Pieces?



入賞
4等分線をイメージして切る
講評:入賞の中で唯一実用的

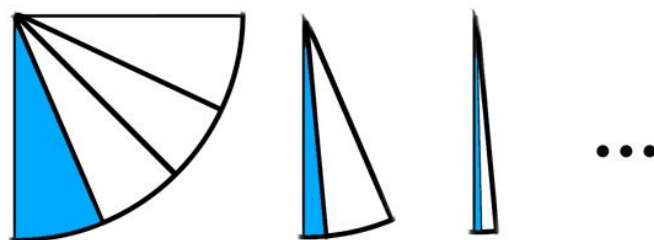


入賞
同心円で切る

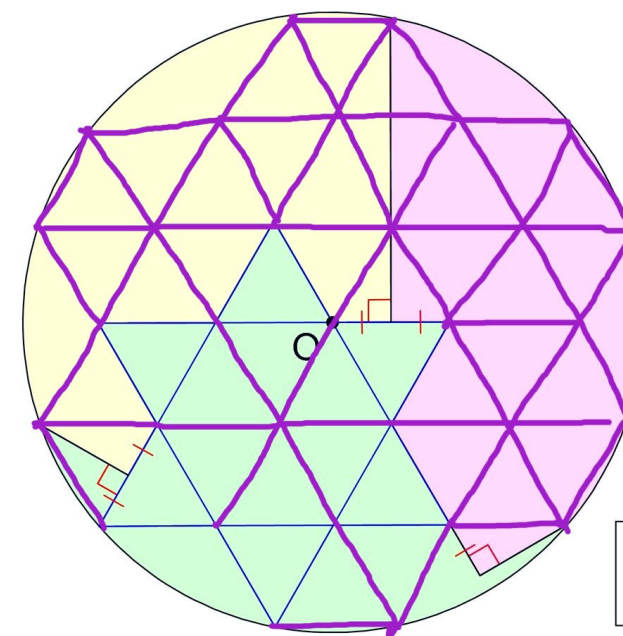


優秀賞
無限に4等分し続ける

講評:何年切り続けても3等分はできない



$$\frac{1}{4} + \frac{1}{4^2} + \frac{1}{4^3} + \frac{1}{4^4} + \dots = \frac{1}{3}$$



Regular Hexagram

Announcements

- ▶ HW2 has been posted on E3 (Due: 11/1, 9pm)
- ▶ About late HW submissions (starting from HW2):
 - ▶ 20% deduction for submissions 0~48 hours past due
 - ▶ HW submitted >48 hours after the deadline will not be graded
- ▶ Midterm on 11/10 (on Wednesday, in class)
 - ▶ 10:10am - 12:10pm
 - ▶ Coverage: Lec 1 - Lec 16
 - ▶ You are allowed to bring a cheat sheet (A4 size, 2-sided, without any attachments)
 - ▶ Locations: EC015 and EC022

EC015

[illegible]

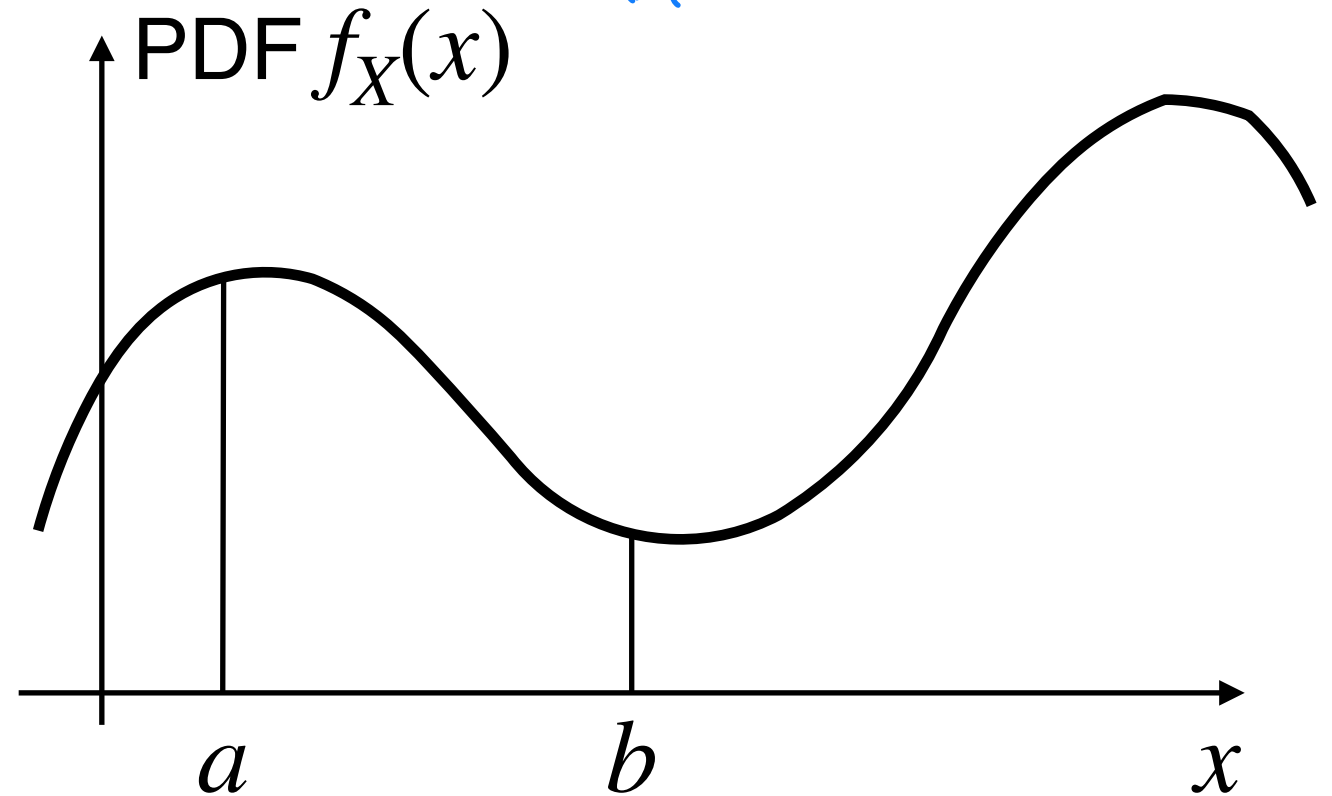
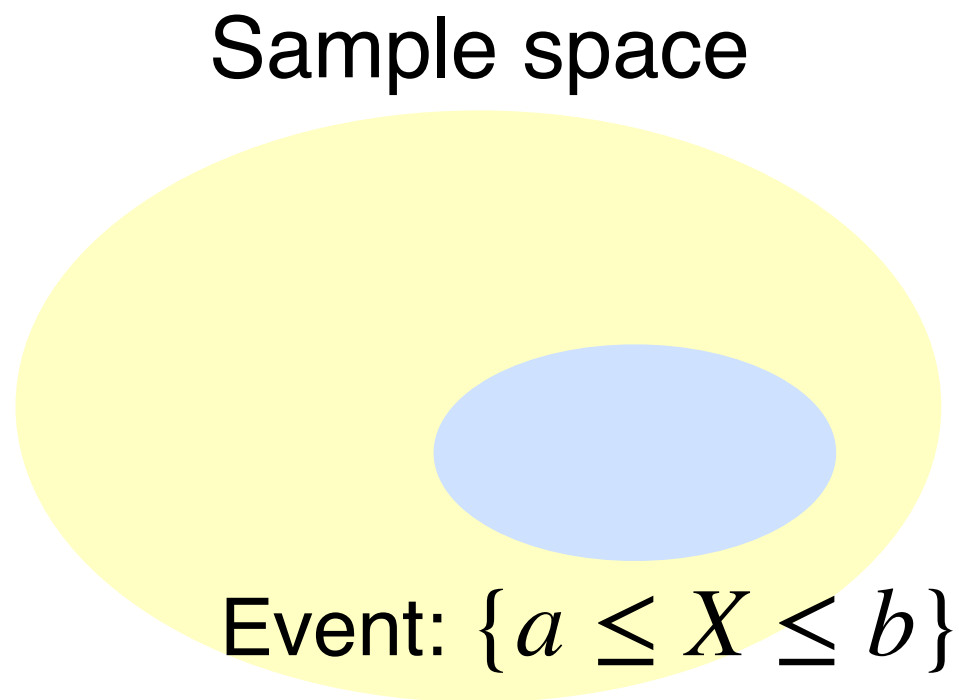
This Lecture

1. Continuous Random Variables

2. Special Continuous Random Variables

- Reading material: Chapter 6.1 and 7.1~7.3

Review: Probability Density Function (PDF)



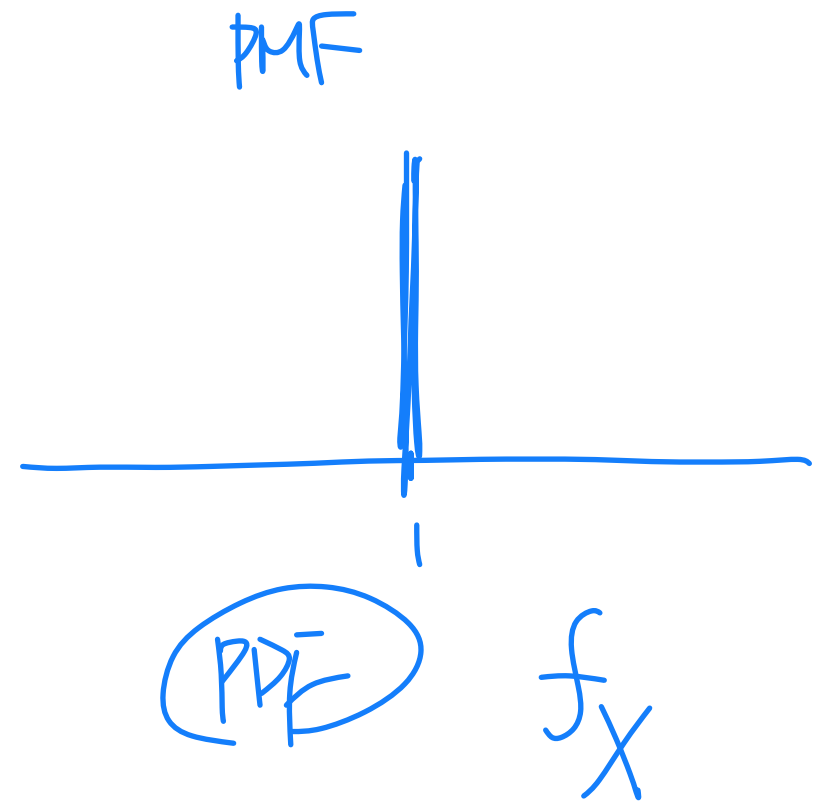
Probability Density Function (PDF):

Let X be a random variable. Then, $f_X(x)$ is the PDF of X if for every subset B of the real line, we have

$$P(X \in B) = \int_B f_X(x) dx$$

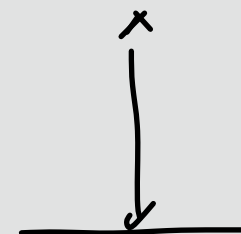
X : discrete random variable

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



$$1 = P(X=1) \stackrel{!}{=} \int_{\{1\}} \underbrace{f_X(x)}_{\text{Delta}} dx$$

Express Other Quantities Using PDF



$$1. \underline{P(X \in \mathbb{R})} = 1 = \int_{-\infty}^{\infty} f_X(x) dx \equiv \int_{\mathbb{R}} f_X(x) dx$$

$$2. \underline{P(X \leq t)} = \int_{-\infty}^t f_X(x) dx$$

$F_X(t)$

$$3. \underline{P(a \leq X \leq b)} = \int_a^b f_X(x) dx$$

$$\int_{x \in \mathbb{R} \setminus \mathbb{Q}} 1 \cdot dx$$

$$4. \underline{P(a \leq X < b)} = \int_a^b f_X(x) dx - \underbrace{P(X=b)}_{=0} = \int_a^b f_X(x) dx$$

$$5. \underline{P(a < X < b)} = \int_a^b f_X(x) dx - \underbrace{P(X=a)}_{=0} - \underbrace{P(X=b)}_{=0}$$

Lebesgue integral
Riemann integral

How to Check if a PDF is Valid?

► **Recall:** 3 Axioms of Probability

1. $P(X \in \mathbb{R}) = 1$ $\Rightarrow \int_{-\infty}^{+\infty} f_X(x) dx = 1$

$$\int_{[0,1] \cup [2,3]} f dx = \int + \int$$

2. $P(X \in A) \geq 0$, for all A

Sufficient condition:

$$f_X(x) \geq 0, \text{ for all } x \in \mathbb{R}$$

$$\Rightarrow \int_A f_X(x) dx \geq 0, \text{ for all } A$$

3. Let A_1, A_2, \dots be mutually exclusive sets of real numbers, then

$$P(X \in \bigcup_{i \geq 1} A_i) = \sum_{i \geq 1} P(X \in A_i)$$

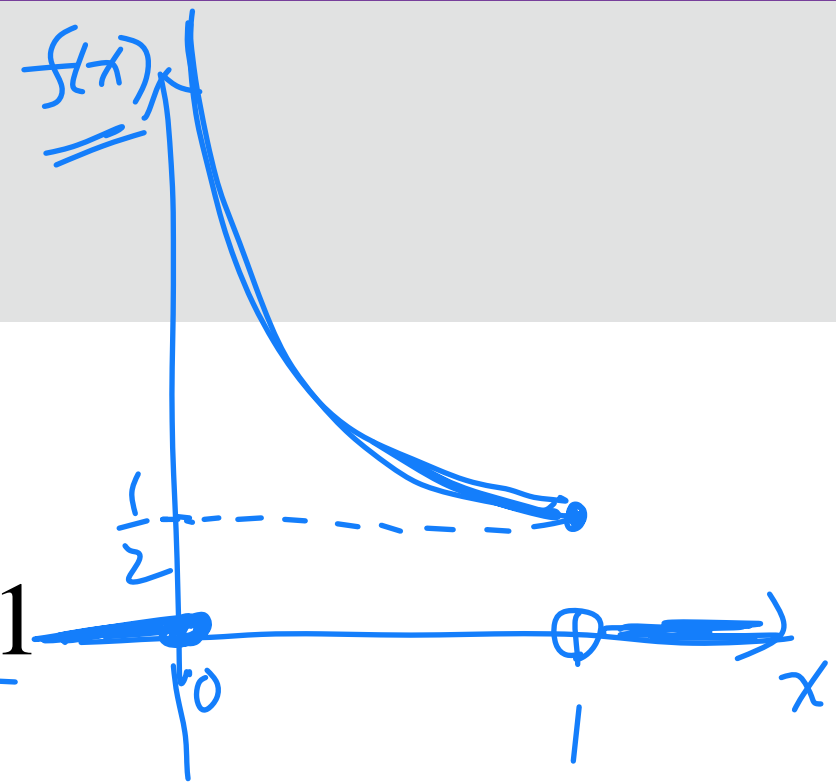
$$\Rightarrow \int_{\bigcup_{i \geq 1} A_i} f_X(x) dx = \sum_{i \geq 1} \int_{A_i} f_X(x) dx$$

This holds by the def. of integration

Example: From PDF to CDF

- ▶ **Example:** Consider the following PDF

$$\underline{f(x)} = \begin{cases} \frac{1}{2\sqrt{x}} & , 0 < x \leq 1 \\ 0 & , \text{otherwise} \end{cases}$$



- ▶ Is $f(x)$ a valid PDF of some random variable?

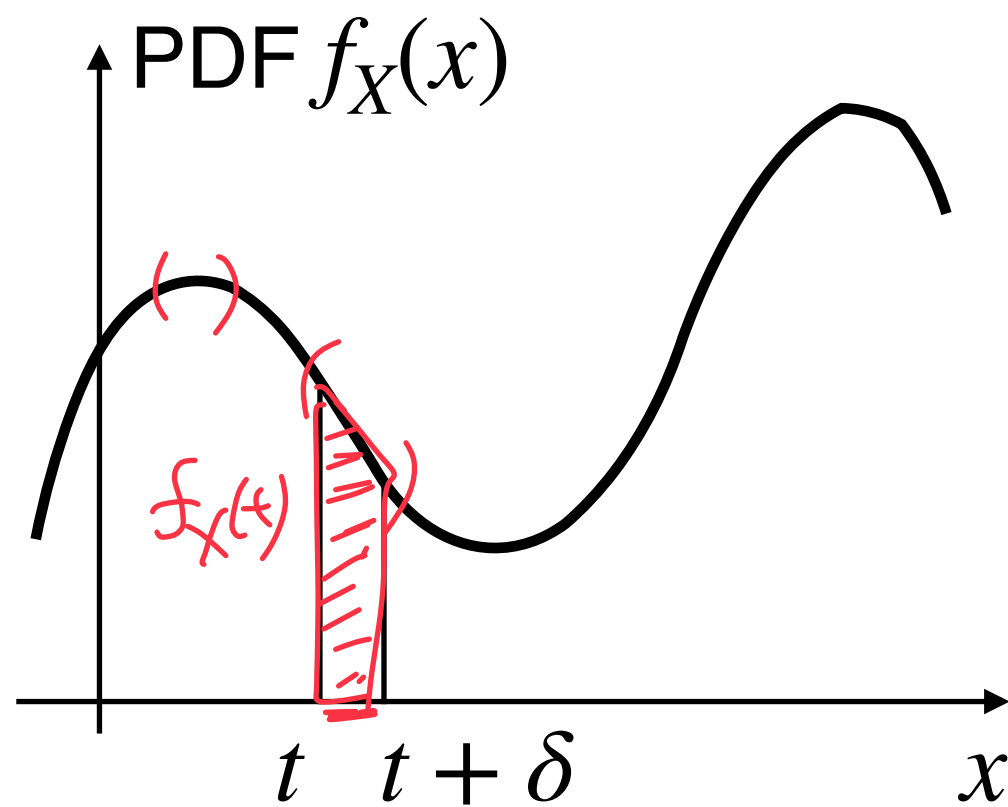
① Nonnegativity: ✓ (since $f(x) \geq 0$ for all $x \in \mathbb{R}$)

② Total probability = 1:
$$\int_{-\infty}^{\infty} f(x) dx = \int_0^1 f(x) dx = \int_0^1 \frac{1}{2\sqrt{x}} dx$$
$$= \sqrt{x} \Big|_0^1 = \sqrt{1} - \sqrt{0} = 1$$

From CDF to PDF

► CDF: $F_X(t) = P(X \leq t) = \int_{-\infty}^t f_X(x) dx$

$$\lim_{\delta \rightarrow 0} \frac{P(t < X \leq t + \delta)}{\delta} = \lim_{\delta \rightarrow 0} \frac{F_X(t + \delta) - F_X(t)}{\delta} = f'_X(t) = f_X(t)$$



► Suppose PDF is continuous

► $F_X(t + \delta) - F_X(t) = ?$

$$P(t < X \leq t + \delta) = \int_t^{t+\delta} f_X(x) dx = f_X(c) \cdot \delta$$

Mean value theorem of integration
for some $c \in [t, t + \delta]$

From CDF to PDF (Formally)

✓ Derivative of CDF is PDF:

Let X be a random variable with a CDF $F_X(\cdot)$ and a PDF $f_X(\cdot)$. If $f_X(\cdot)$ is continuous at x_0 , then

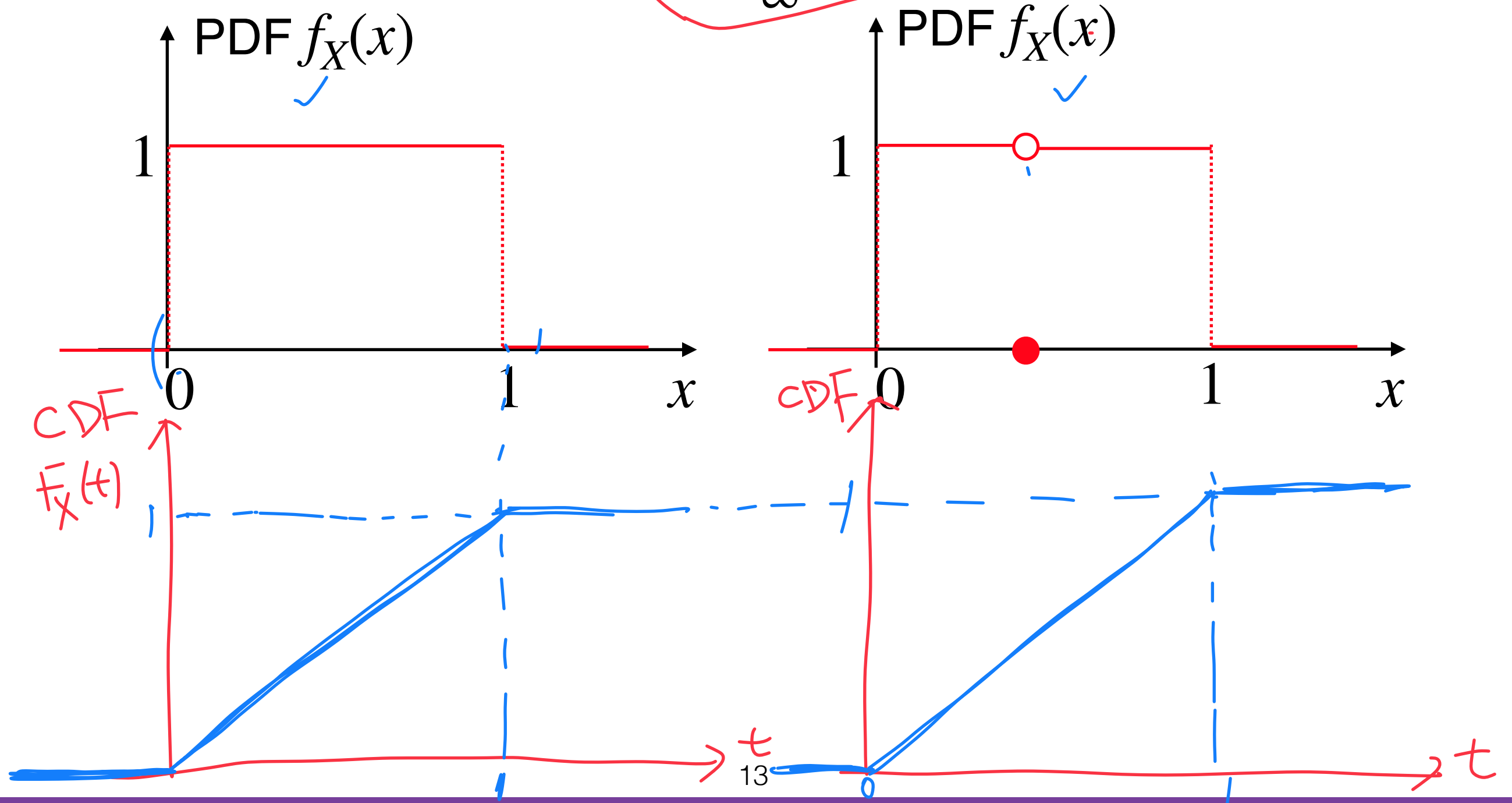
$$\underline{F'_X(x_0)} = \underline{f_X(x_0)}$$

- Any similar results in calculus?

Fundamental Theorem of Calculus

From CDF to PDF: Why Continuity is Required?

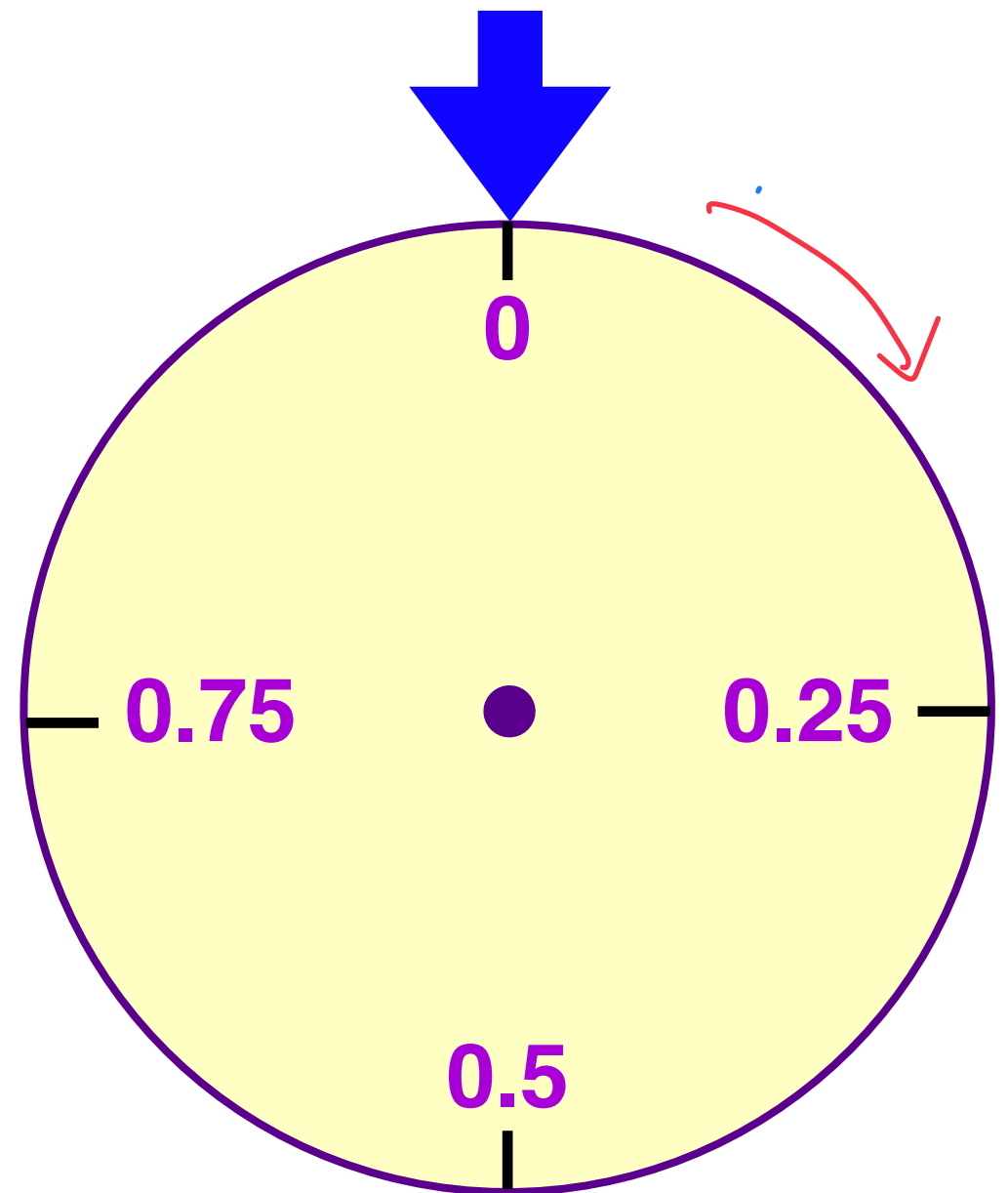
► CDF: $F_X(t) = P(X \leq t) = \int_{-\infty}^t f_X(x) dx$ "identifiability"



Special Continuous Random Variables

1. (Continuous) Uniform Random Variables

- ▶ **Example:** A bus arrive at a random time between 9:15am and 9:30am
- ▶ **Example:** Play wheel of fortune
- ▶ What are the common features?
 - ▶ Principle of indifference



1. Uniform Random Variables (Formally)

$$X \sim \text{Unif}(a, b)$$

Uniform Random Variables: A random variable X is uniform with parameters a, b ($a < b$) if its PDF is

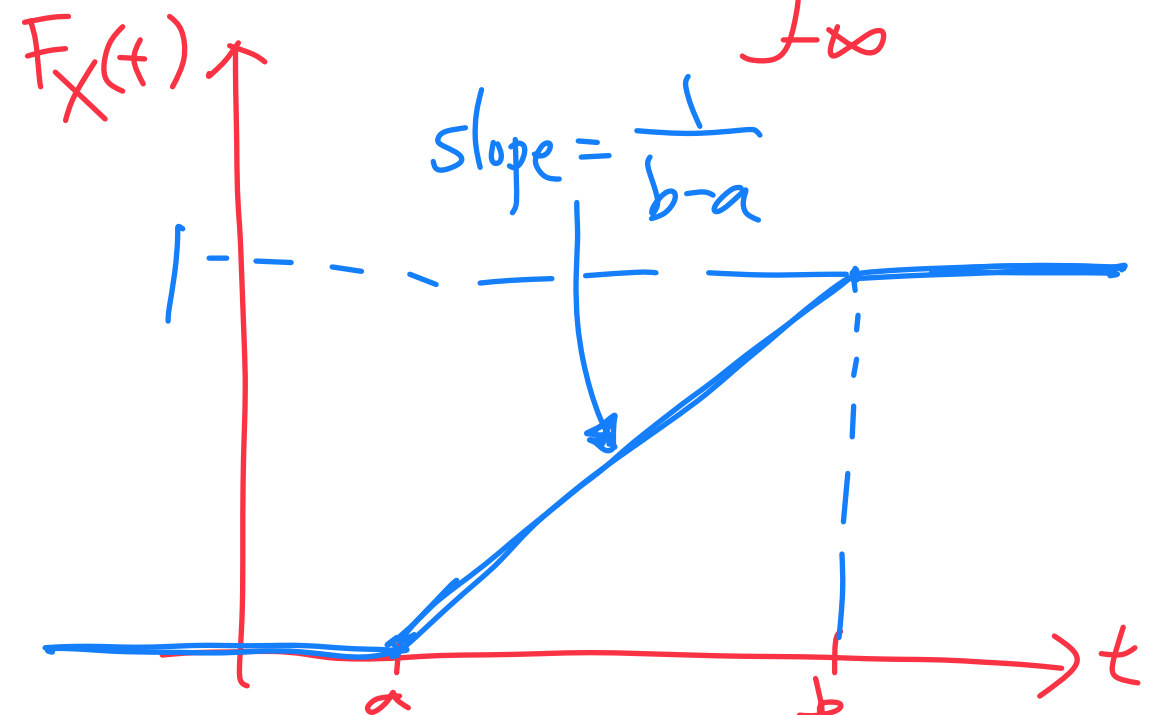
$$f_X(x) = \begin{cases} \frac{1}{b-a} & , \text{ if } a < x < b \\ 0 & , \text{ otherwise} \end{cases}$$

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

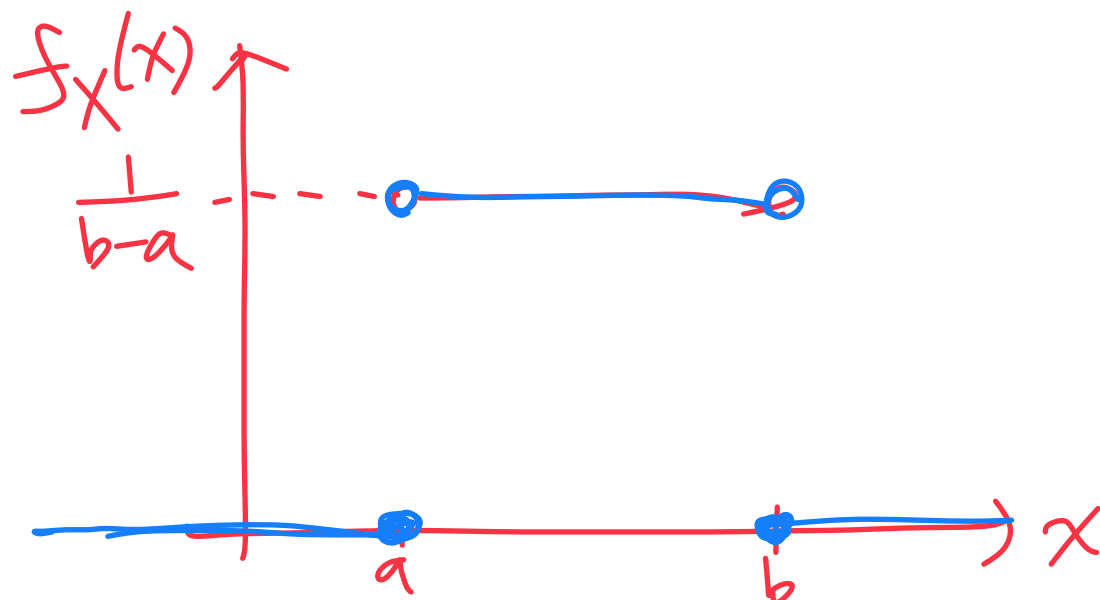
$$= \int_{-\infty}^t f_X(x) dx$$

- How to plot the PDF and CDF of X ?

CDF:



PDF:



Example: Uniform Distribution

- ▶ **Example:** Let X be a random variable with CDF $F(t)$.
- ▶ Define another random variable $Y = F(X)$.
- ▶ What type of random variable is Y ?

$$P(Y \leq 0.2) = P(F(X) \leq 0.2)$$

$$= P(\{\omega : F(X(\omega)) \leq 0.2\}) = 0.2$$

For any $t^* \in [0, 1]$:

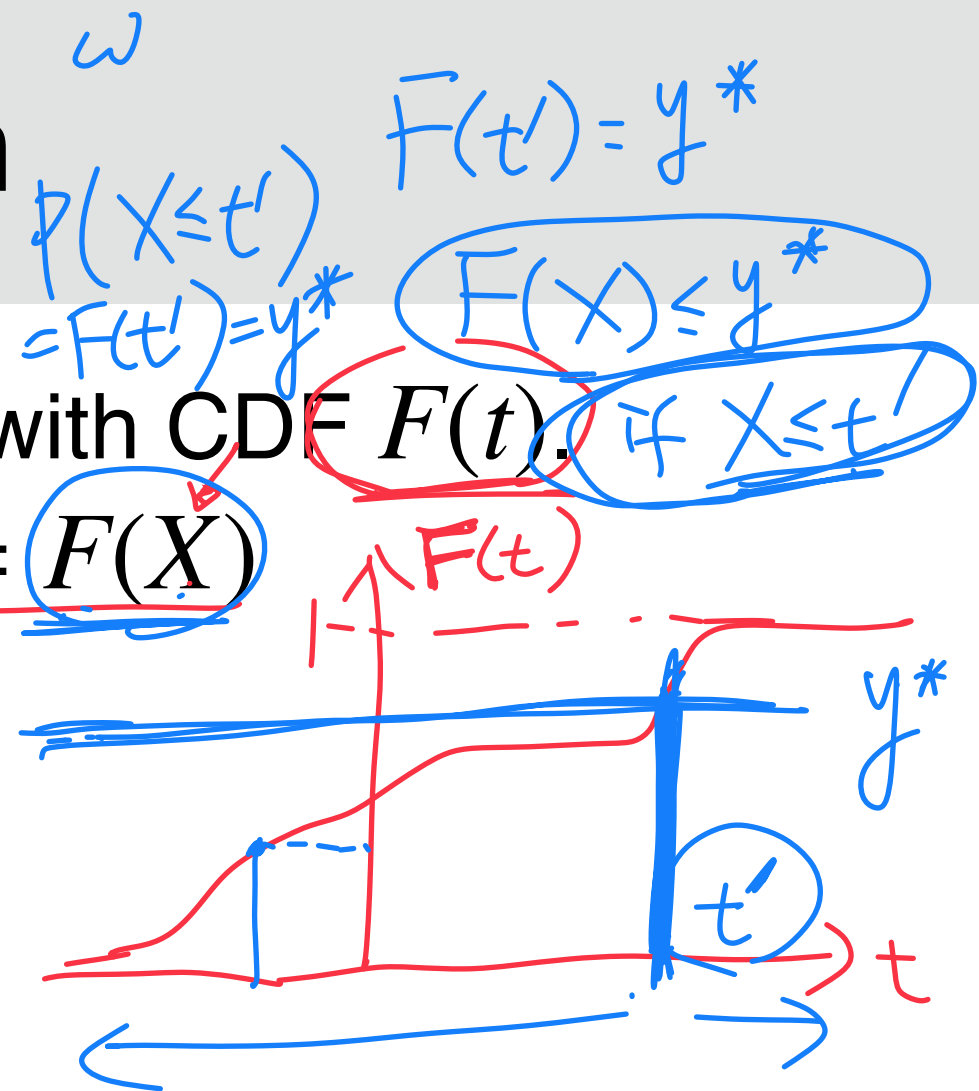
$$P(Y \leq t^*) = t^*$$

For $t^* < 0$:

$$P(Y \leq t^*) = 0$$

For $t^* > 1$:

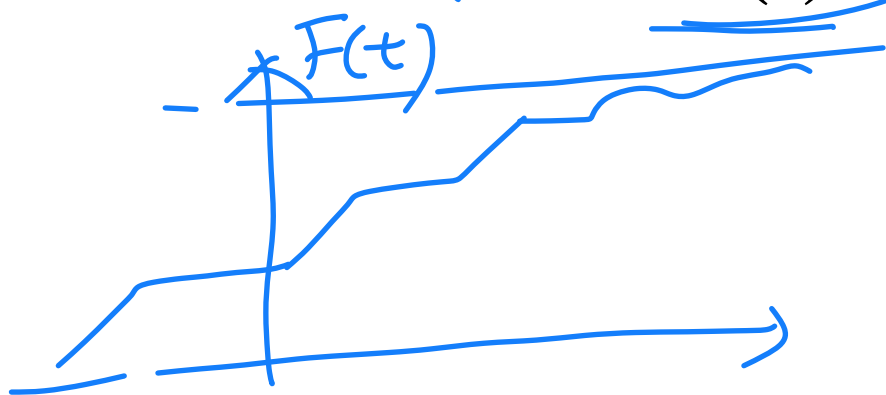
$$P(Y \leq t^*) = 1$$



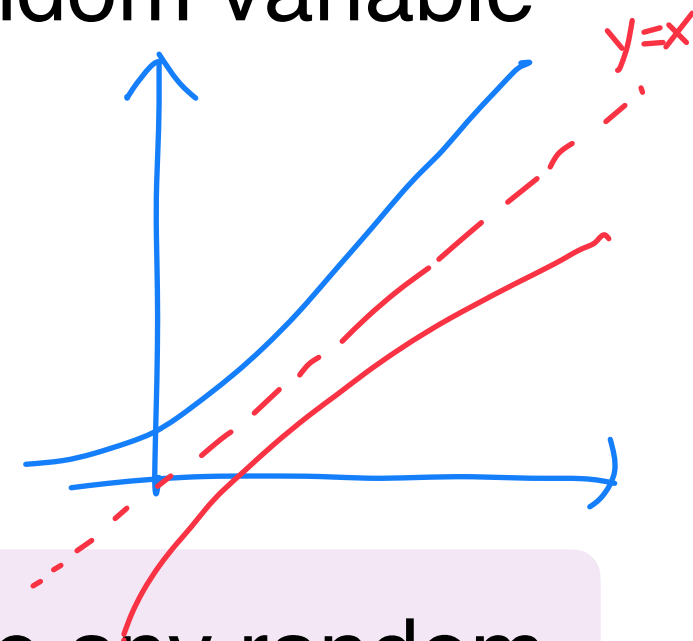
$$Y \sim \text{Unif}(0, 1)$$

Why Are Uniform Random Variables Useful?

- **Question:** How to generate a customized random variable X with CDF $F(t)$?



np.random - uniform
geometric
binomial



Inverse Transform Sampling (ITS): Generate any random variable with CDF $F(t)$ from a uniform random variable

1. Generate a random variable $U \sim \text{Unif}(0,1)$
2. Let $X = F^{-1}(U)$, where $F^{-1}(u) := \inf\{z : F(z) \geq u\}$

Proof: Inverse Transform Sampling $P(X \leq t) = F(t)$

Inverse Transform Sampling: Generate any random variable with CDF $F(t)$ from a uniform random variable

1. Generate a random variable $U \sim \text{Unif}(0,1)$

2. Let $X = F^{-1}(U)$, where $F^{-1}(u) := \inf\{z : F(z) \geq u\}$

F is non-decreasing

$$\begin{aligned} \blacktriangleright \underbrace{P(F^{-1}(U) \leq x)}_{\uparrow} &= P(F(F^{-1}(U)) \leq F(x)) \\ &= P(U \leq \underbrace{F(x)})_{\uparrow} = \underbrace{F(x)}_{\downarrow} \\ &= P(\{2X \leq 2\}) \end{aligned}$$

The CDF of X is indeed $F(x)$

Example: Inverse Transform Sampling

- ▶ **Example:** Generate a random variable X with CDF

$$F_X(t) = \begin{cases} 1 - \exp(-t^2) & , t \geq 0 \\ 0 & , \text{otherwise} \end{cases}$$

Normal Random Variables

2. Standard Normal Random Variables

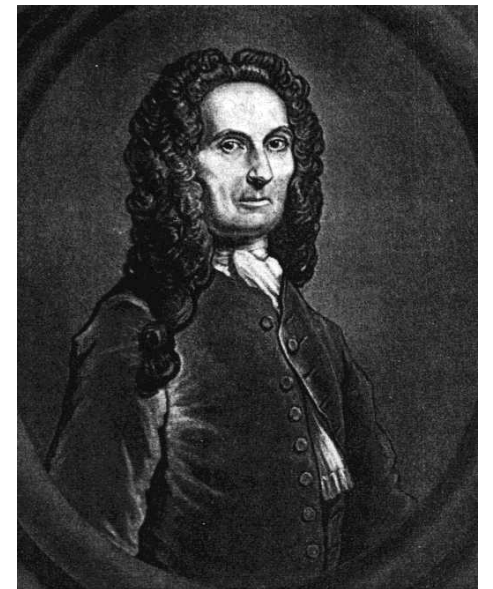
- **Motivation:** Consider $X \sim \text{Binomial}(n, \frac{1}{2})$ $\xrightarrow{\text{PMF}} P(X=k) = \begin{cases} C_k^n \cdot (\frac{1}{2})^k \cdot (\frac{1}{2})^{n-k}, & k=0,1,\dots,n \\ 0, & \text{otherwise} \end{cases}$

Define $Y = \frac{X - \frac{1}{2}n}{\frac{1}{2}\sqrt{n}}$. What is the CDF of Y vs n ? $n \rightarrow \infty$

$$E[X] = \frac{1}{2} \cdot n$$

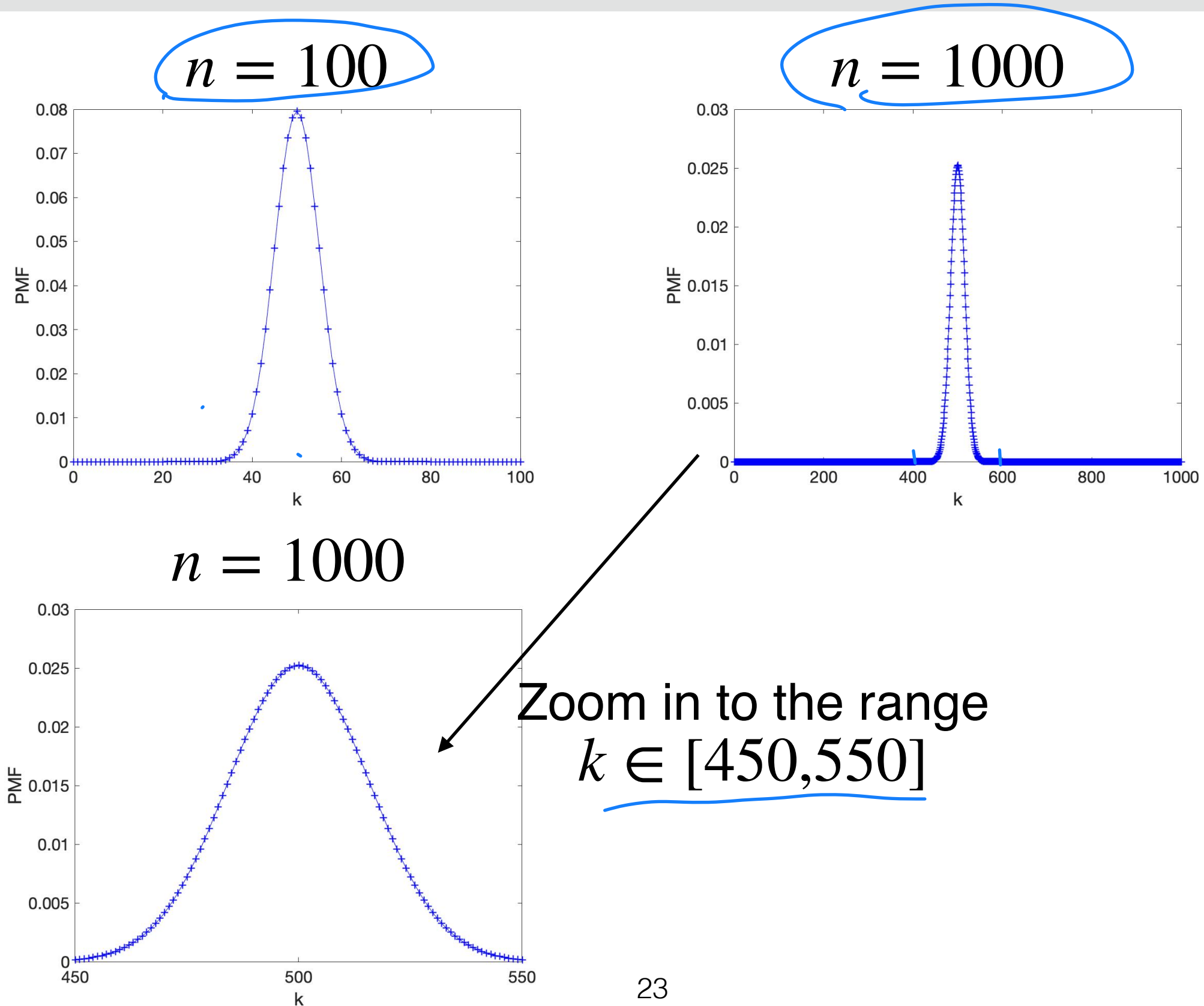
$$\text{Var}[X] = n \cdot \frac{1}{2} \cdot (1 - \frac{1}{2}) = \frac{1}{4} \cdot n$$

$$Y = \frac{X - (E[X])}{\sqrt{\text{Var}[X]}}$$



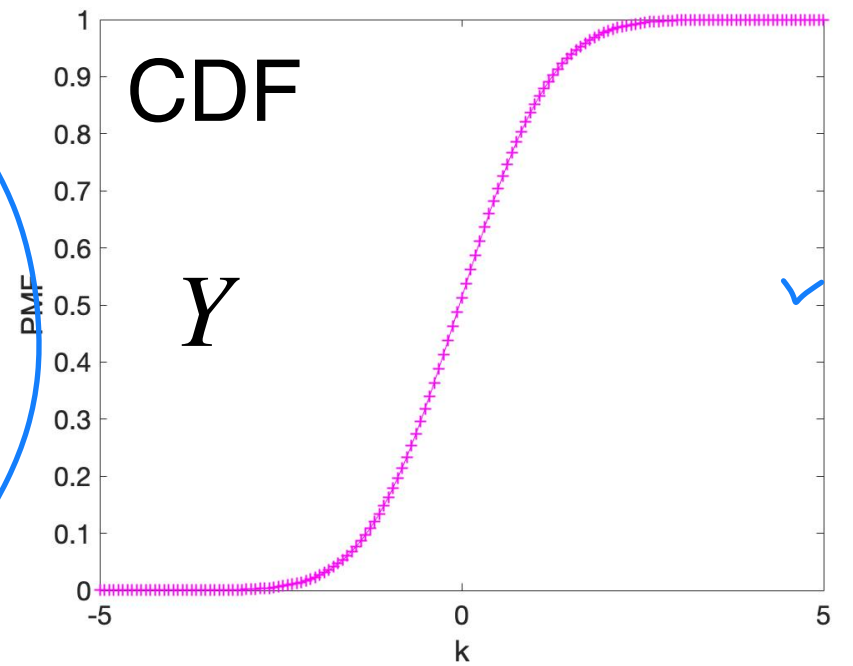
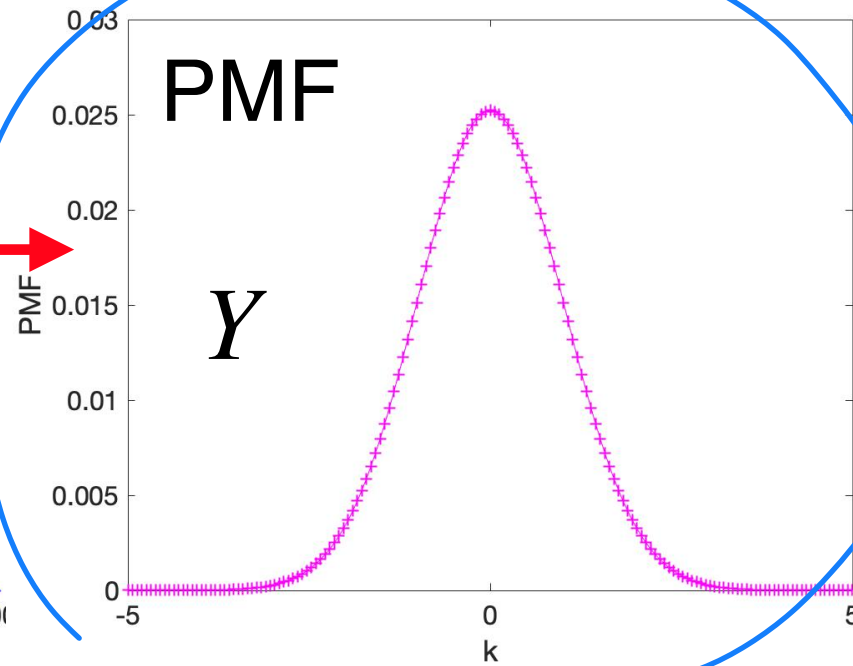
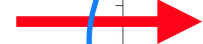
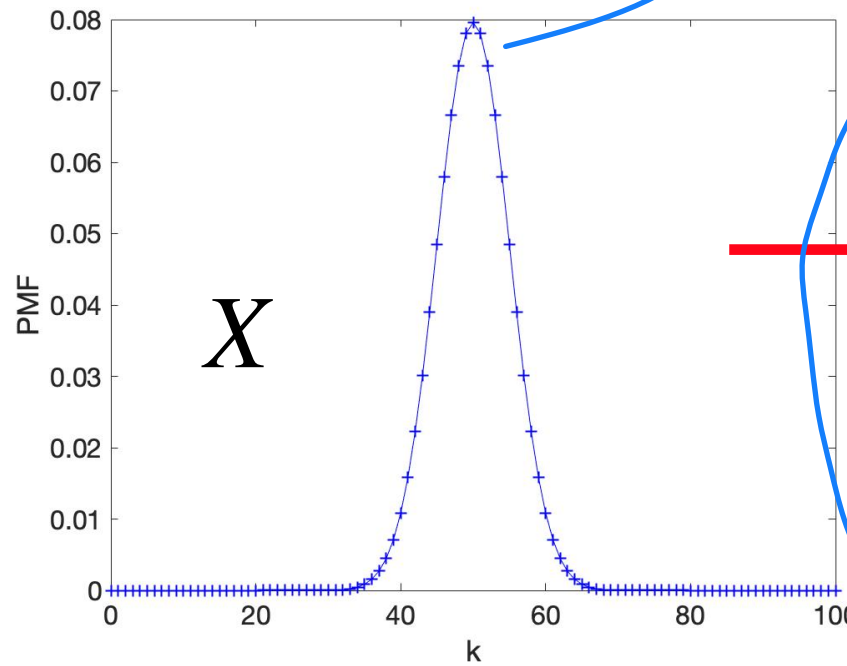
Abraham de Moivre

Plotting $X \sim \text{Binomial}(n, 1/2)$

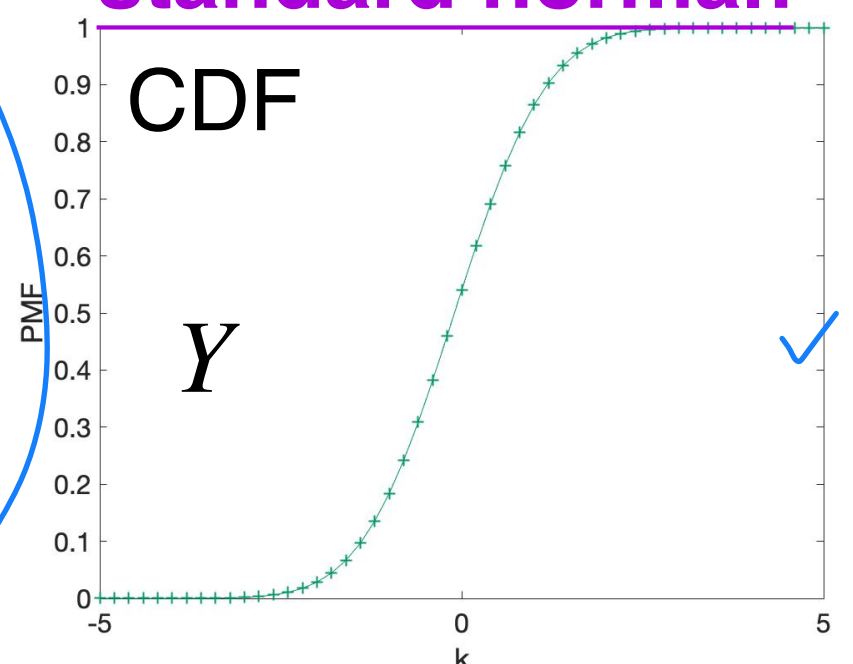
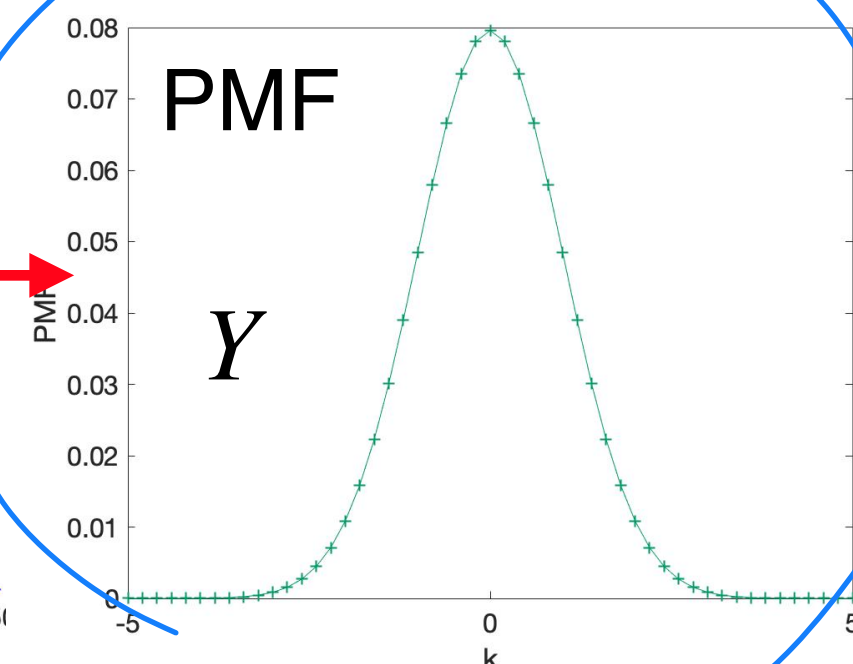
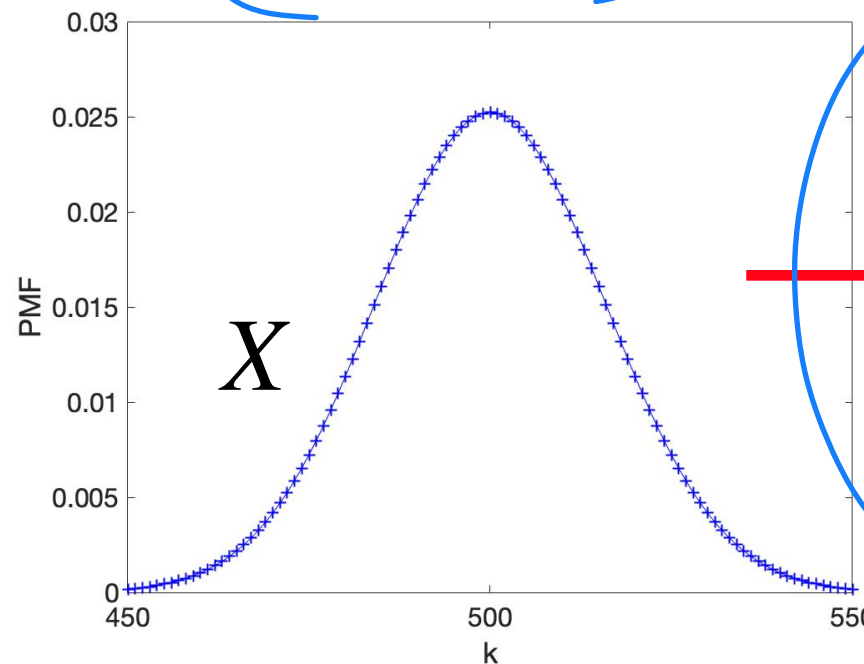


Plotting $Y = (X - 0.5n)/(0.5\sqrt{n})$

$n = 100$



$n = 1000$



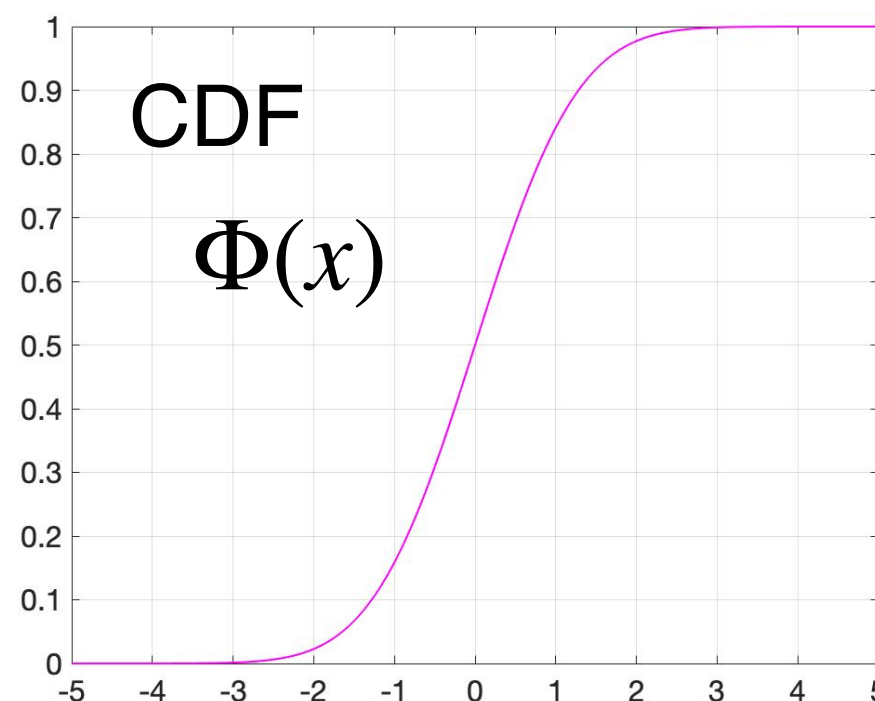
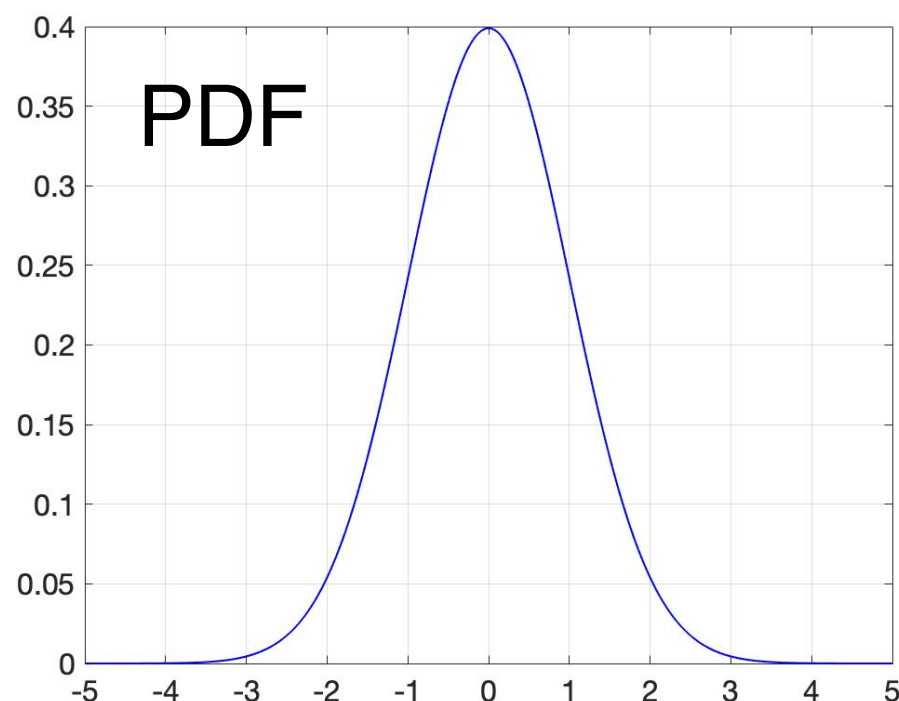
\approx CDF of a
standard normal!

2. Standard Normal Random Variables (Formally)

Standard Normal Random Variables: A random variable X is called standard normal if its PDF is

$$f_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), \text{ for all } x \in \mathbb{R}$$

- How to plot the PDF and CDF?



$$E[X] = 0$$
$$\text{Var}[X] = 1$$

2. CDF of Standard Normal (Formally)

- ▶ As standard normal is widely applicable, we use a special notation $\Phi(\cdot)$ for its CDF

CDF of Standard Normal: The CDF of a standard normal random variable X is

$$\Phi(t) := P(X \leq t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} \exp\left(\frac{-x^2}{2}\right) dx$$

- ▶ **Question:** How to plot $\Phi(t)$?
 - ▶ $\Phi(\infty) = ?$ $\Phi(0) = ?$

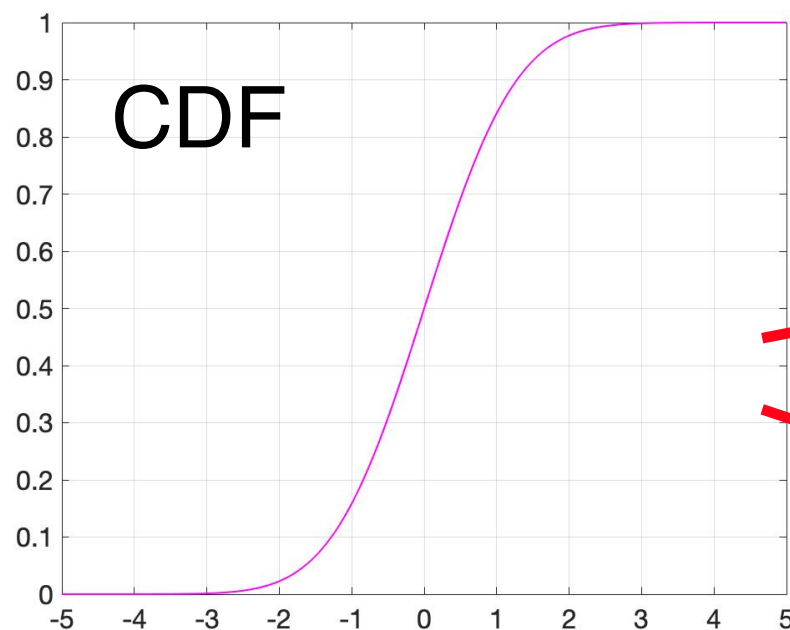
Why is Normal Distribution Useful?

1. Central Limit Theorem:

2. Gaussian Process and Black-Box Optimization:

From Standard Normal to Normal: CDF

- ▶ X is standard normal



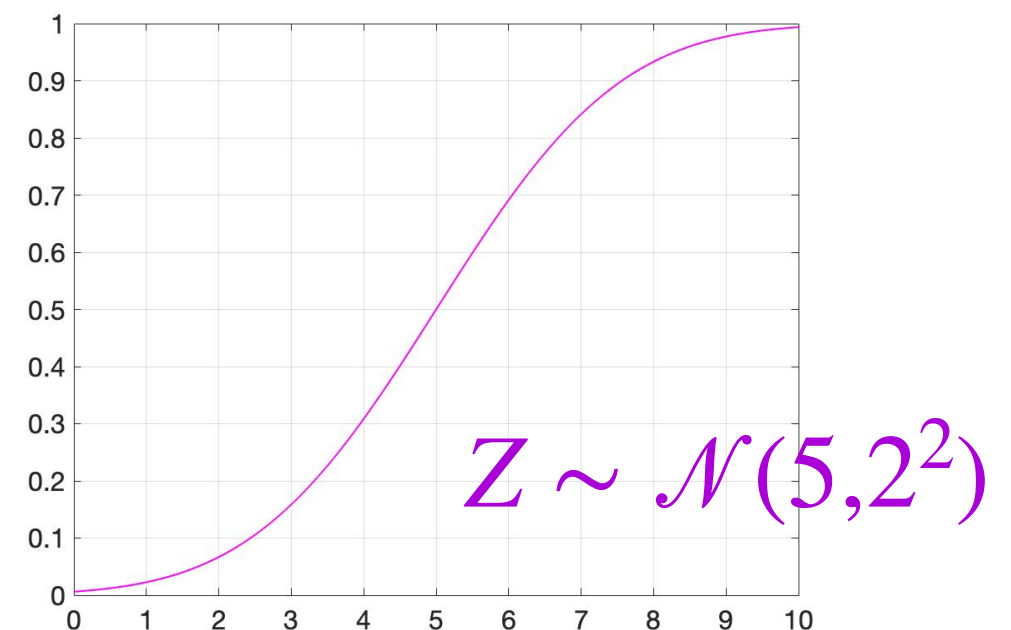
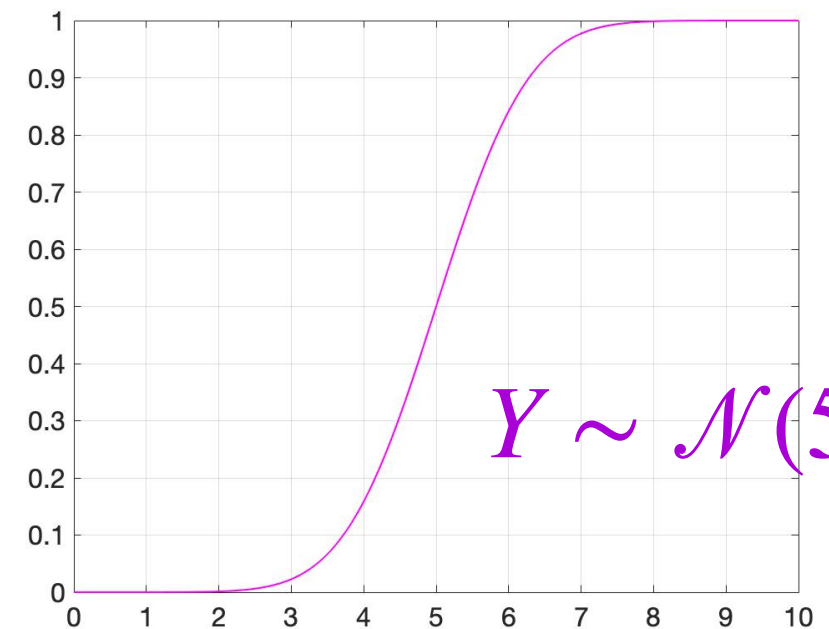
$$F_X(t) = \Phi(t)$$

$$X \sim \mathcal{N}(0,1)$$

$$Y = X + 5$$

$$Z = 2X + 5$$

$$F_Y(t) = \Phi(t - 5)$$

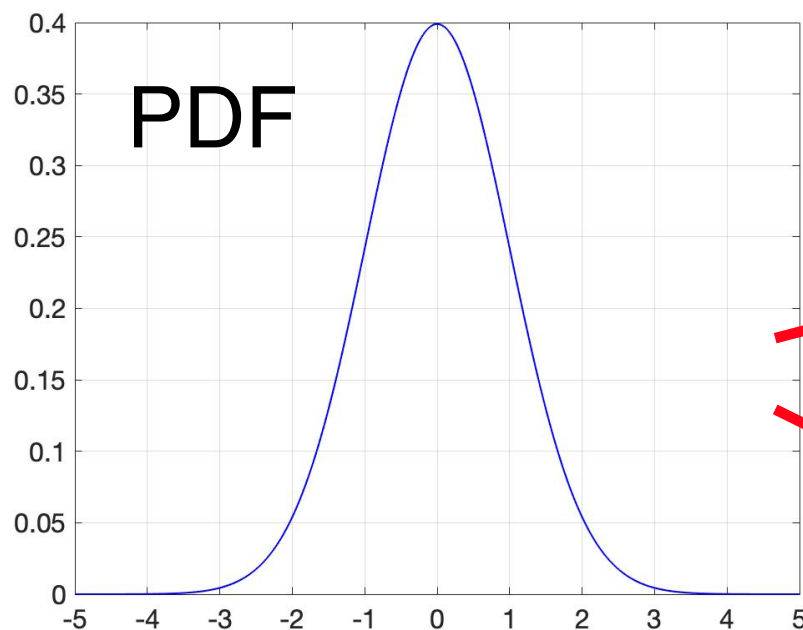


$$F_Z(t) = \Phi\left(\frac{t - 5}{2}\right)$$

From Standard Normal to Normal: PDF

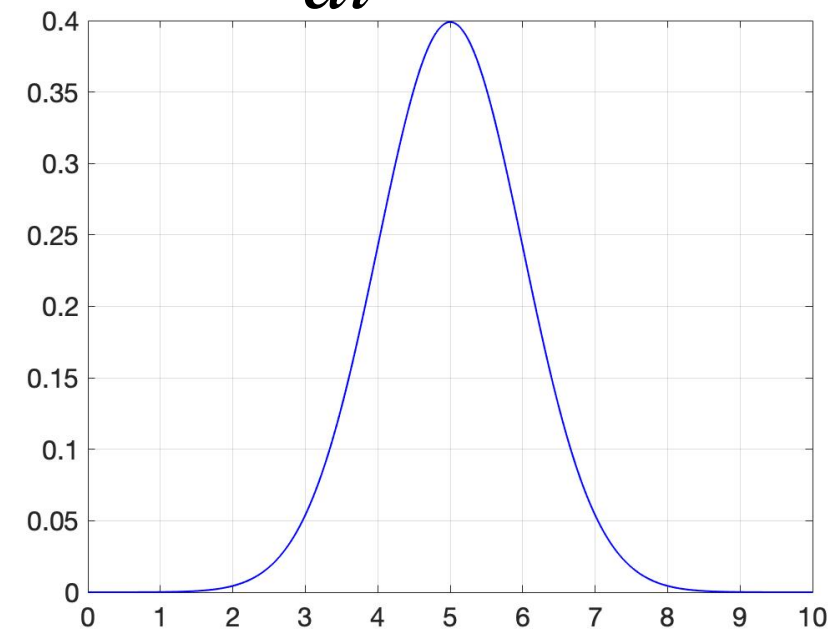
- ▶ X is standard normal

$$f_X(t) = \frac{d\Phi(t - 5)}{dt} = \Phi'(t - 5)$$

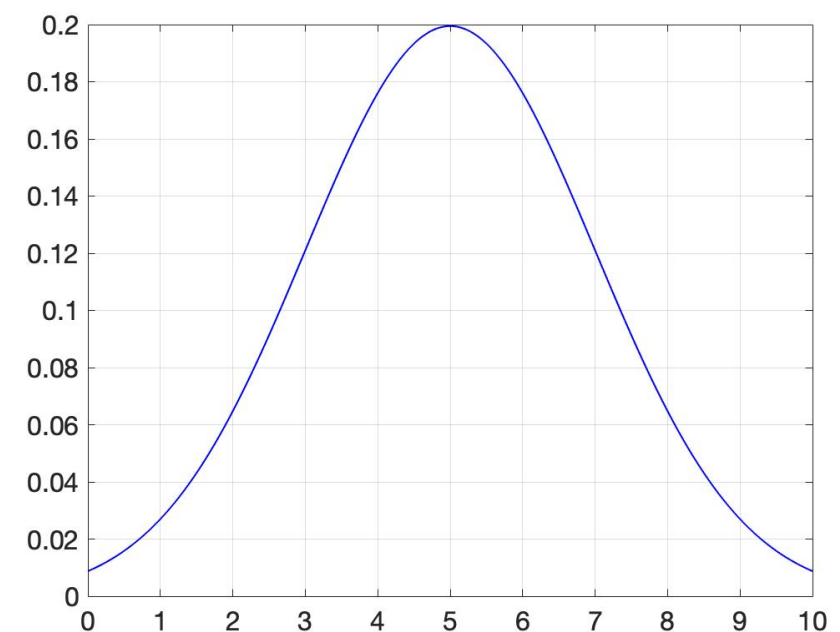


$$f_X(t) = \Phi'(t)$$

$$Y = X + 5$$



$$Z = 2X + 5$$

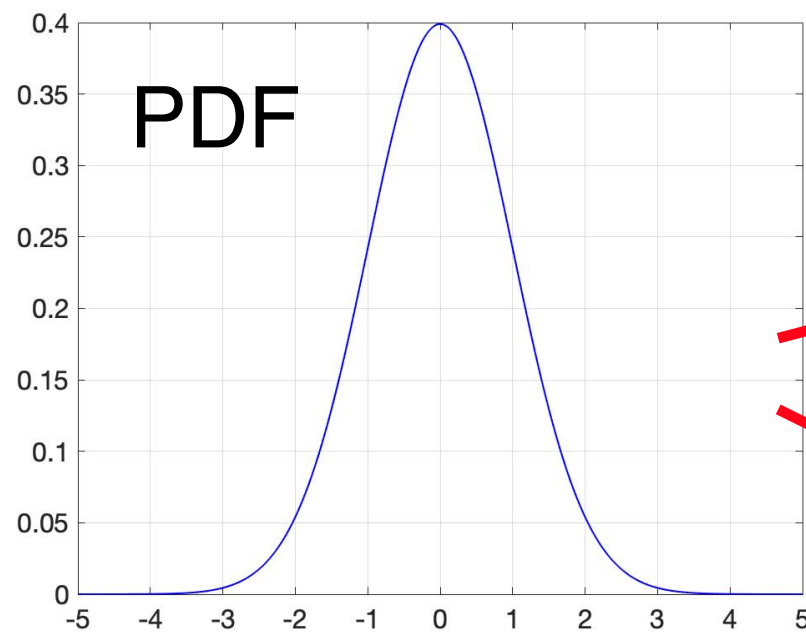


$$f_X(t) = \frac{d\Phi(\frac{t-5}{2})}{dt} = \frac{1}{2}\Phi'(\frac{t-5}{2})$$

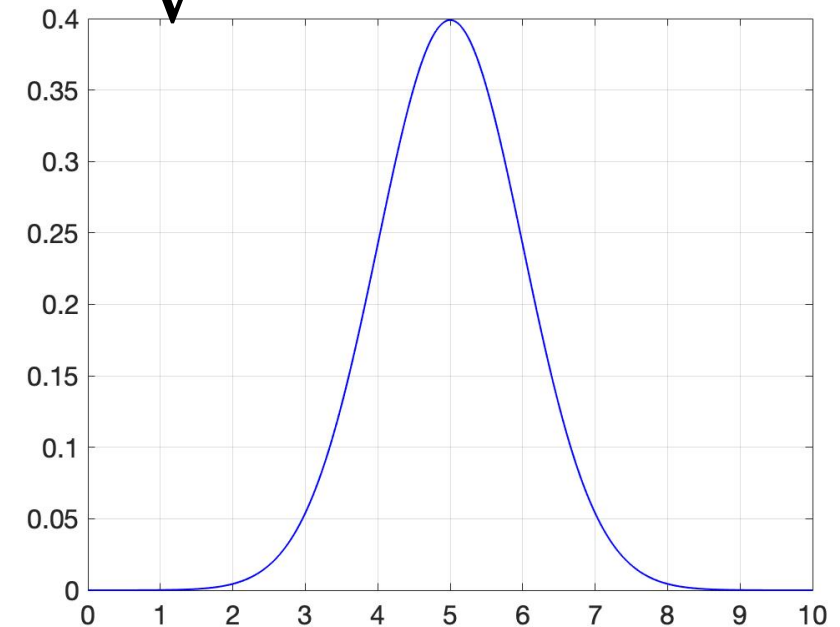
From Standard Normal to Normal: PDF

- ▶ X is standard normal

$$f_Y(t) = \frac{1}{\sqrt{2\pi}} \exp\left(\frac{-(t-5)^2}{2}\right)$$

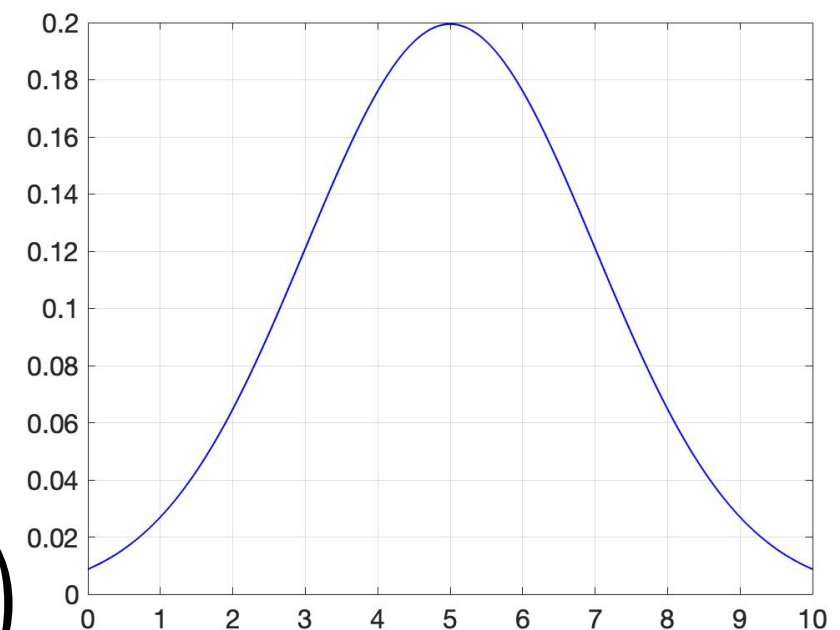


$$Y = X + 5$$



$$f_X(t) = \frac{1}{\sqrt{2\pi}} \exp\left(\frac{-t^2}{2}\right)$$

$$Z = 2X + 5$$



$$f_Z(t) = \frac{1}{2\sqrt{2\pi}} \exp\left(\frac{-(t-5)^2}{2 \cdot 2^2}\right)$$

A General Recipe for Linear Transformation

- ▶ X is a continuous random variable
 - ▶ CDF: $F_X(t)$
 - ▶ PDF: $f_X(t) = \frac{dF_X(t)}{dt}$
- ▶ Consider $Y = aX + b$, $a, b \in \mathbb{R}, a \neq 0$
 - ▶ CDF $F_Y(t)$?
 - ▶ PDF $f_Y(t)$?
 - ▶ If $X \sim \mathcal{N}(0,1)$, then $F_Y(t) = ?$

Normal Random Variables

Normal Random Variables: A random variable X is called normal with parameters μ, σ if its PDF is

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(\frac{-(x - \mu)^2}{2\sigma^2}\right)$$

- ▶ **Notation:** $X \sim \mathcal{N}(\mu, \sigma^2)$
- ▶ How to plot the PDF?

Example: Normal Distribution

- ▶ **Example:** Let $X \sim \mathcal{N}(-2, 5)$
 - ▶ What is $P(|X| < 4)$?

Exponential Random Variables

Recall: Geometric Random Variables

- ▶ Suppose $X \sim \text{Geometric}(p)$
 - ▶ What is the PMF of X ?
 - ▶ Memoryless property?
- ▶ Question: Is there a continuous counterpart of a geometric random variable?

3. Exponential Random Variables

Exponential Random Variables: A random variable X is exponential with parameters $\lambda > 0$ if its PDF is

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & , \text{if } x \geq 0 \\ 0 & , \text{otherwise} \end{cases}$$

- ▶ How to plot the PDF of $\text{Exp}(\lambda = 1)$?

3. Exponential Random Variables

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & , \text{if } x \geq 0 \\ 0 & , \text{otherwise} \end{cases}$$

- ▶ What is the CDF of X ?

Memoryless Property

- ▶ Suppose $X \sim \text{Exp}(\lambda)$
 - ▶ What is $P(X > s + t \mid X > t)$?

Example: Nokia 3310

- ▶ **Example:** Suppose the lifetime of a Nokia 3310 is an exponential random variable with mean = 10 years.
 - ▶ Suppose a Nokia 3310 was bought 15 years ago.
 - ▶ $P(\text{it will last another 5-10 years})?$



Exponential Distribution: A Good Model for Occurrence of Events

- ▶ **Communication networks**: Inter-arrival time between two data packets
- ▶ **Survival analysis**: User's lifetime (App, social network...)
- ▶ **Reliability modeling**: Amount of time until the hardware on AWS EC2 fails

1-Minute Summary

1. Continuous Random Variables

- Probability density function
- PDF vs CDF

2. Special Continuous Random Variables

- Uniform and Inverse Transform Sampling (ITS)
- Standard Normal
- Exponential and Memoryless Property