

DCP 1206: Probability

Lecture 10 — Special Continuous Random Variables (I)

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#Fakenews? Fake Videos of Real People



(June, 2019)

Mark: ...*Imagine this for a second: One man, with total control of billions of people's stolen data, all their secrets, their lives, their futures...*
(by Bill Posters @bill_posters_uk)



(July, 2017)

- ▶ Research from University of Washington:
 - ▶ Given only the audio, a video can be synthesized with accurate lip sync
- ▶ How to tell the fake from the true one?
 - ▶ Eye blinking? (arXiv:1806.02877)

Expected Value of a Continuous Random Variable Using CDF

Expected Value via CDF:

Let X be a continuous random variable with CDF $F_X(t)$.

The expected value of X is

$$E[X] = \int_0^{\infty} (1 - F_X(t))dt - \int_0^{\infty} F_X(-t)dt$$

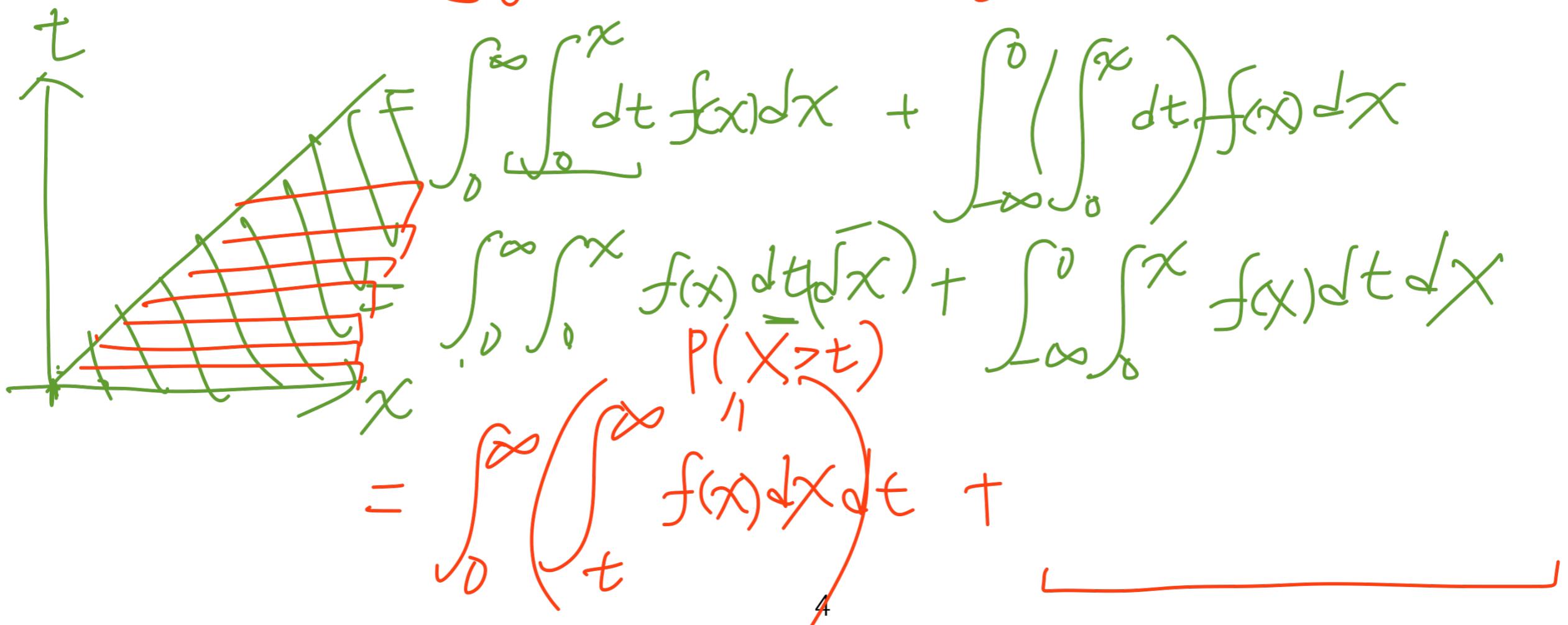
- ▶ What if X is a non-negative random variable?
- ▶ How to prove this?

Proof: Expected Value of a Continuous Random Variable Using CDF

$$E[X] = \int_0^\infty (1 - F_X(t))dt - \int_0^\infty F_X(-t)dt$$

1. dt
2. int. change
of integration

Pf: $E[X] = \int_0^\infty xf(x)dx + \int_{-\infty}^0 xf(x)dx$



Existence of Moments (Formally)

Existence of Moments:

Let X be a random variable. Then, the n -th moment of X (i.e. $E[X^n]$) is said to exist if $E[|X^n|] < \infty$

- ▶ Same definition for both discrete and continuous random variables

Example: Pareto Random Variable

- ▶ **Example:** Suppose $n \geq 1$ and the PDF of a r.v. X is

$$f_X(x) = \begin{cases} \frac{C}{x^{n+1}} & , x \geq C \\ 0 & , \text{otherwise} \end{cases}$$

- ▶ What is the value of C ?
- ▶ For what values of m does $E[X^{m+1}]$ exist?

This Lecture

1. (Continuous) Uniform Random Variables

2. Normal Random Variables

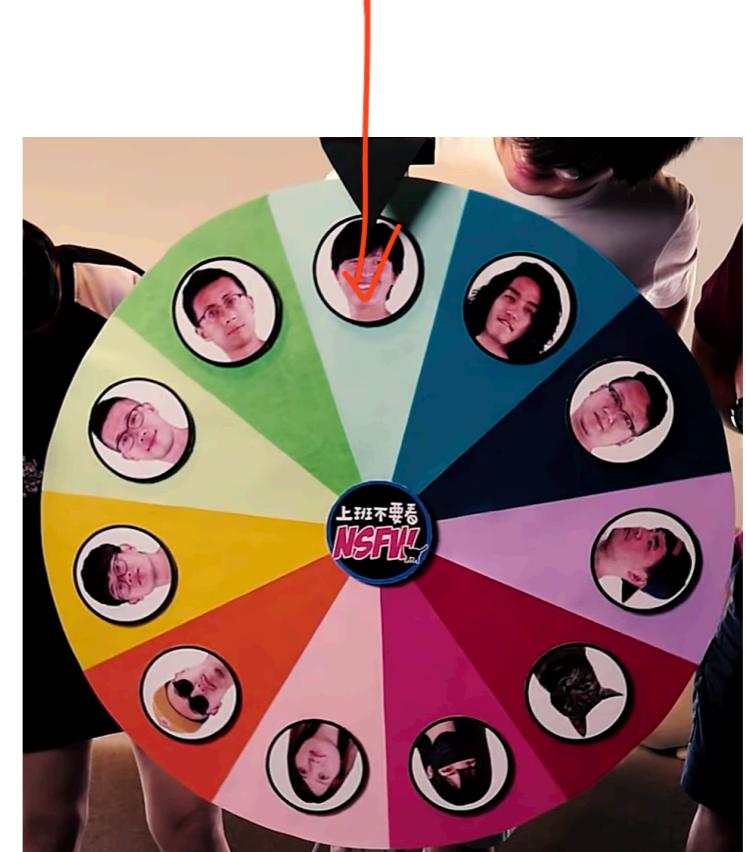
- Reading material: Chapter 7.1-7.2

1. Uniform 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



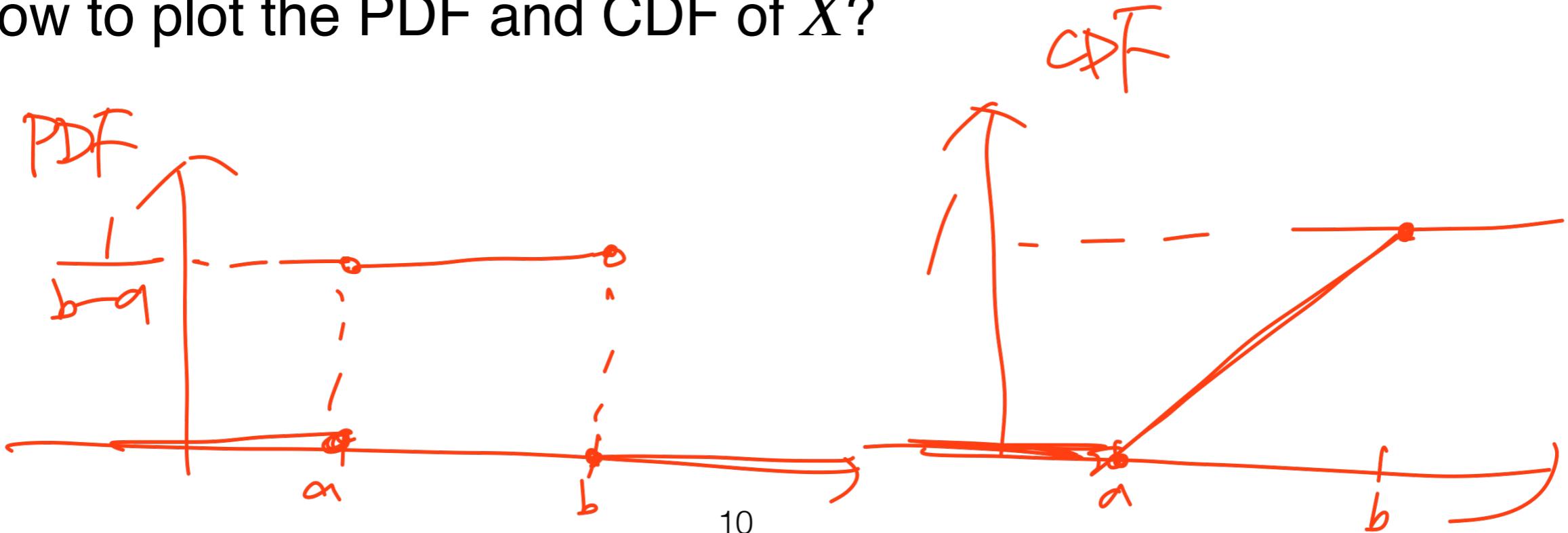
- ▶ What are the common features?
 - ▶ Principle of indifference

1. Uniform Random Variables (Formally)

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

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

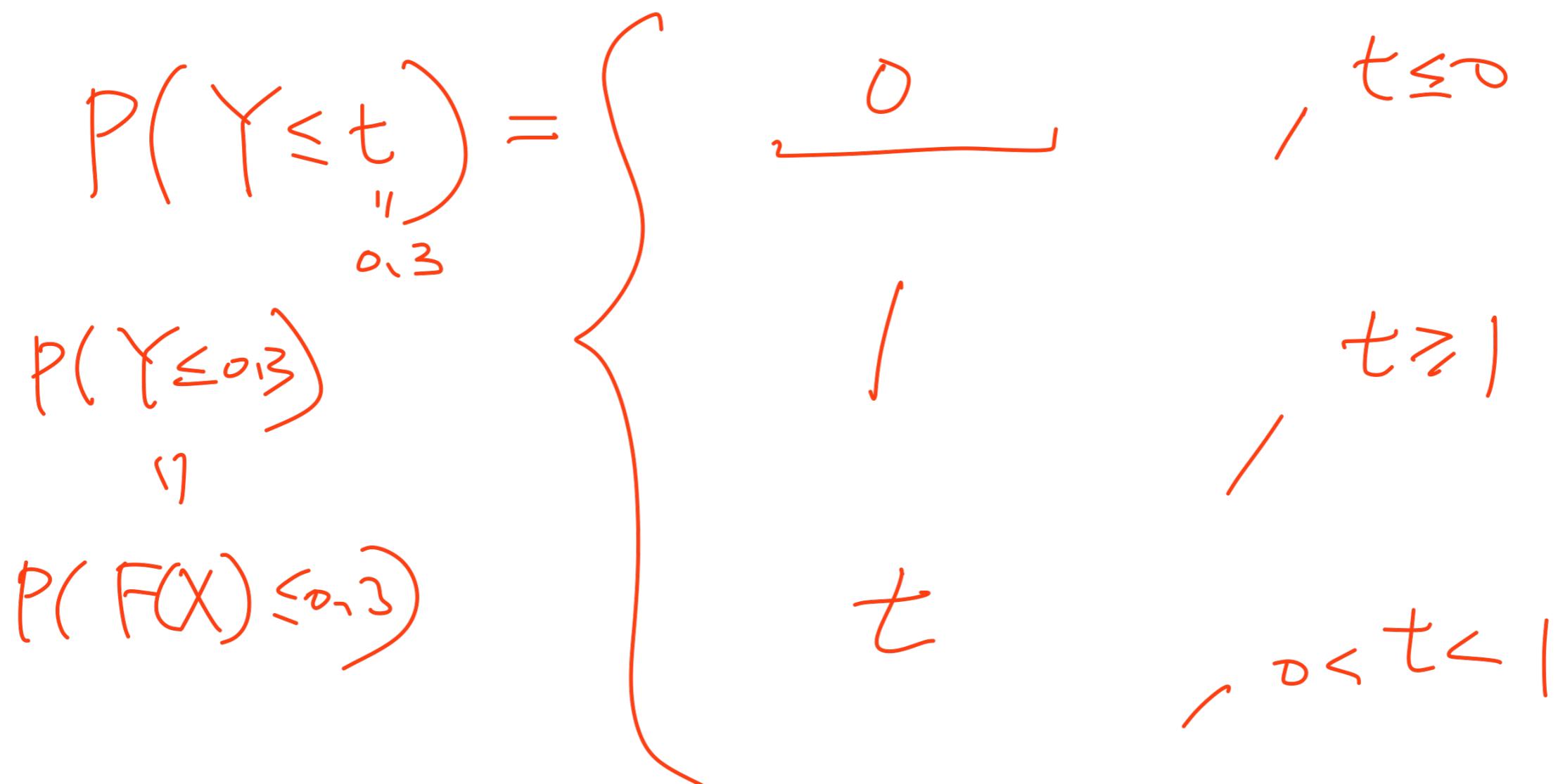
- ▶ How to plot the PDF and CDF of X ?



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 ?

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



1. Uniform Random Variables: Mean and Variance

- ▶ Example: $X \sim \text{Unif}(a, b)$
 - ▶ What is $E[X]$?
 - ▶ How about $\text{Var}[X]$?

Why Are Uniform Random Variables Useful?

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



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 $F^{-1}(u) := \inf\{z : F(z) \geq u\}$

inverse of $F(\cdot)$

Proof: Inverse Transform Sampling

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 $F^{-1}(u) := \inf\{z : F(z) \geq u\}$

$$\Pr(F^{-1}(U) \leq x) = \Pr(U \leq F(x)) = F(x)$$

~~+~~ apply $F(\cdot)$ on both sides and
use the fact: $F(\cdot)$ is non-decreasing

Example: Inverse Transform Sampling

- Example: Generate a random variable X with CDF

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

Step 1.

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

Step 2:

inverse function $F_X^{-1}(u)$

$$u = 1 - e^{-x^2} \Leftrightarrow e^{-x^2} = 1 - u$$

$$\Leftrightarrow -x^2 = \ln(1-u) \quad \text{inverse function}$$

$$\Leftrightarrow x = \sqrt{\ln \frac{1}{1-u}}$$

2. Standard Normal Random Variables

Gaussian

► Motivation: Consider $X \sim \text{Binomial}(n, \frac{1}{2})$

$$\frac{1}{2}$$

(Cosine) $\stackrel{n}{\rightarrow}$
= Cosine + isin

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

1730s

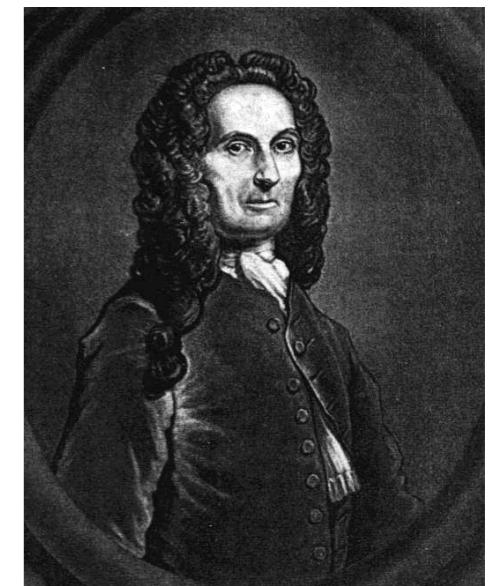
$$X - \frac{1}{2}n$$

$$\frac{1}{2}\sqrt{n}$$

mean

$$n \rightarrow \infty$$

Euler



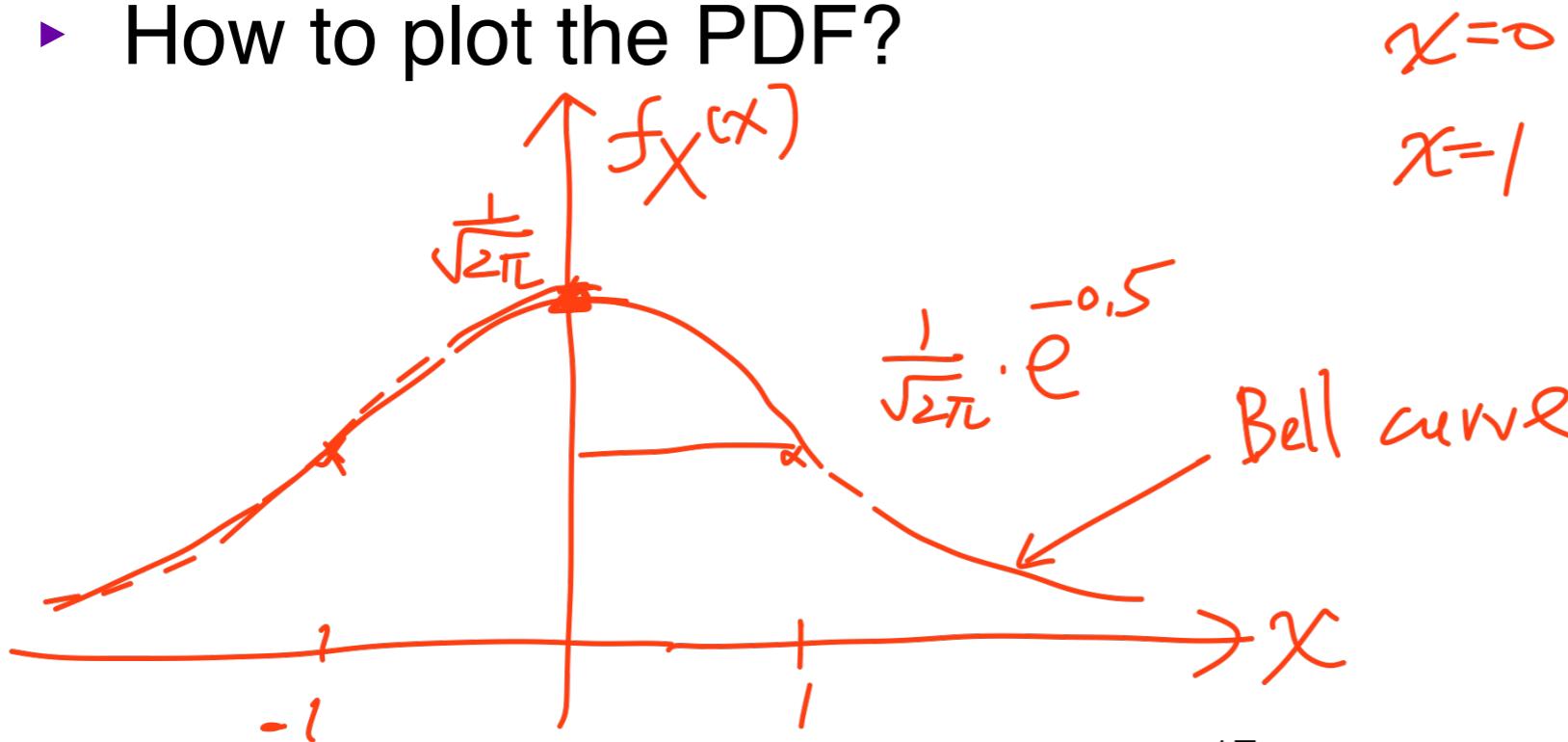
Abraham de Moivre

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?



2. CDF of Standard Normal

- 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

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

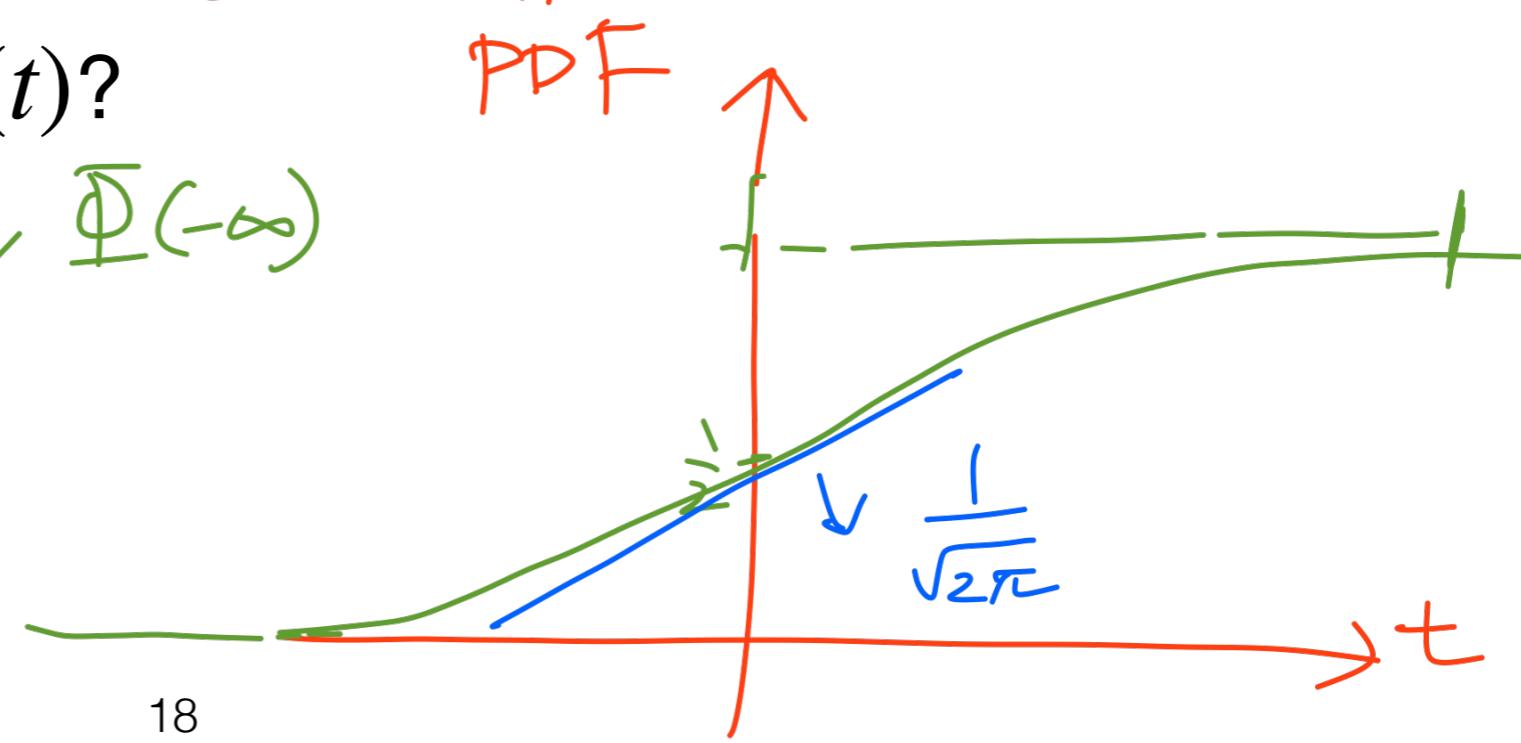
- Question: How to plot $\Phi(t)$?

- $\Phi(\infty) = ?$ $\Phi(0) = ?$, $\Phi(-\infty)$

"
 $P(X < \infty)$

"
 $P(X \leq 0)$

$\Phi(0)$



2. Standard Normal: Mean and Variance

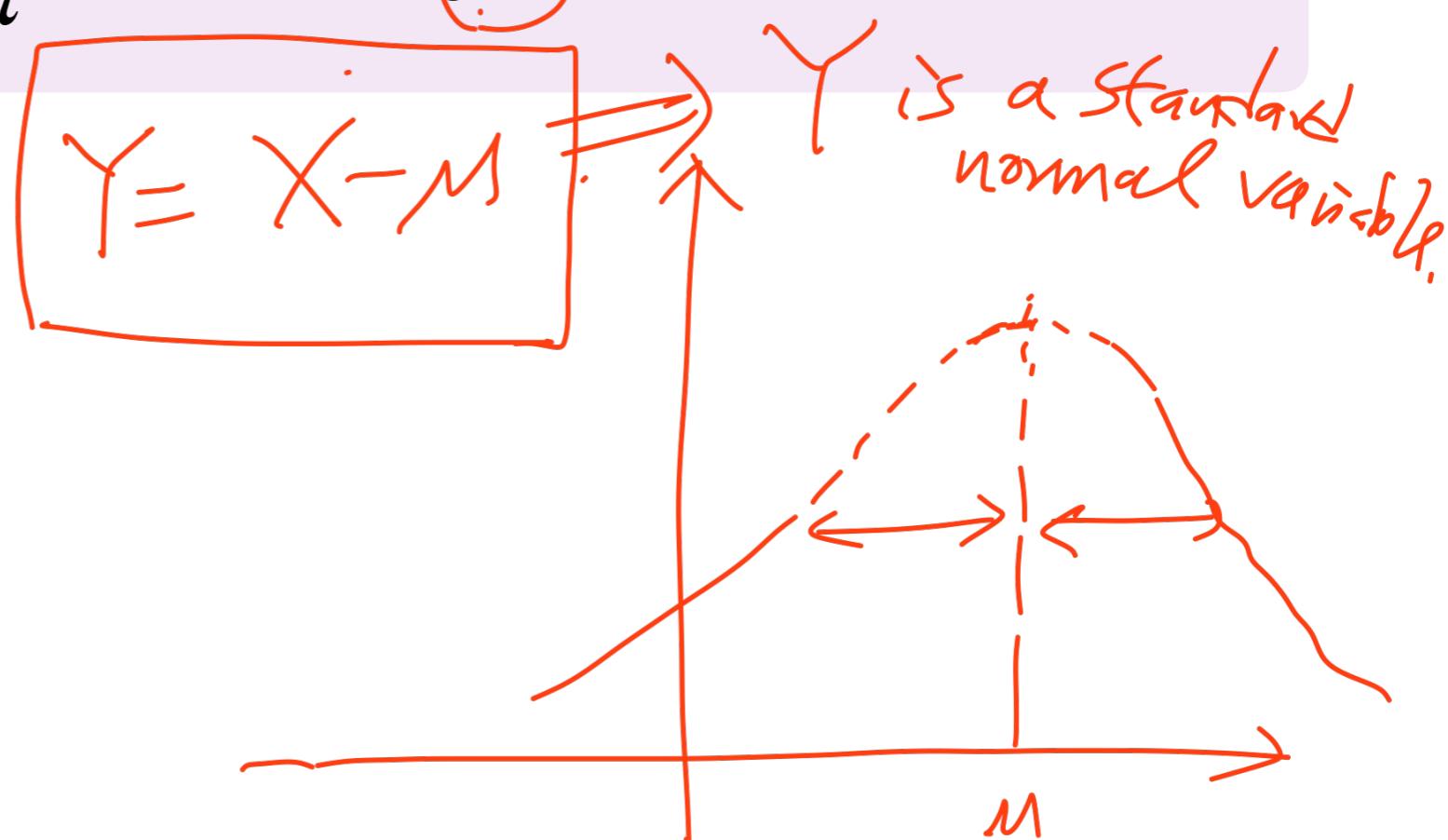
- ▶ **Example:** X is a standard normal random variable
 - ▶ What is $E[X]$?
 - ▶ How about $\text{Var}[X]$?

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 $P(|X| < 4)$?

Why is Normal Distribution Useful?

1. **Central Limit Theorem:**
2. **Gaussian Process and Black-Box Optimization:**

Next Lecture

1. Special continuous random variables

- Exponential
- Gamma
- Beta

1-Minute Summary

1. Special Continuous Random Variables

- Uniform and inverse transform sampling
- Standard normal and normal