

1179: Probability
Lecture 28 — Strong Law of Large
Numbers

Ping-Chun Hsieh (謝秉均)

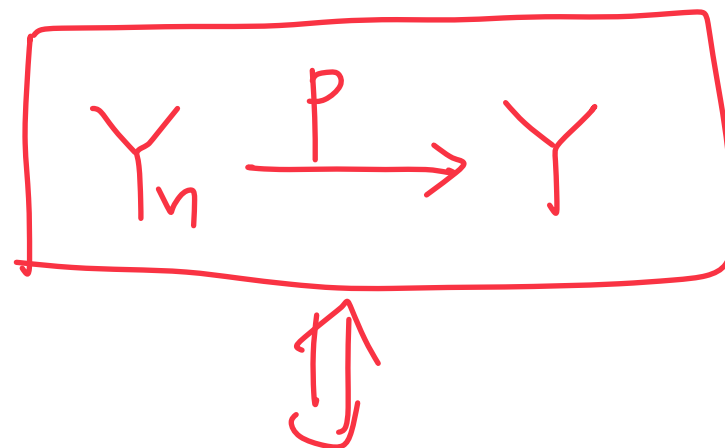
December 24, 2021

Announcements

- ▶ HW4 is now available on E3!
 - ▶ HW4-Part I will be due on 12/30 (Thursday), 9pm
 - ▶ HW4-Part II will be due on 1/3 (Monday), 9pm
- ▶ Final exam on 1/5 (on Wednesday, in class)
 - ▶ 10:10am - 12:10pm
 - ▶ Coverage: Lec 1 - Lec 29
 - ▶ You are allowed to bring a cheat sheet (A4 size, 2-sided, without any attachments)
 - ▶ Locations: EC015 and EC022

"Convergence in probability":

$Y_1, Y_2, \dots, Y_n, \dots$



$$\lim_{n \rightarrow \infty} P\left(\underbrace{\left\{\omega : |Y_n(\omega) - Y(\omega)| > \varepsilon\right\}}_{\text{bad event}}\right) = 0,$$

for any $\varepsilon > 0$.

"WLLN":

$X_1, X_2, X_3, \dots, X_n, \dots$ i.i.d. random variables.

$$Y_n = \frac{X_1 + \dots + X_n}{n}$$

$$\lim_{n \rightarrow \infty} P\left(\omega : |Y_n(\omega) - \underbrace{\mu}_{\text{true mean}}| > \varepsilon\right) = 0, \text{ for any } \varepsilon > 0$$

This Lecture

1. Strong Law of Large Numbers (SLLN) and Almost-Sure Convergence

2. Monte-Carlo Simulation

- Reading material: Chapter 11.4

Convergence in Probability and Almost-Sure Convergence

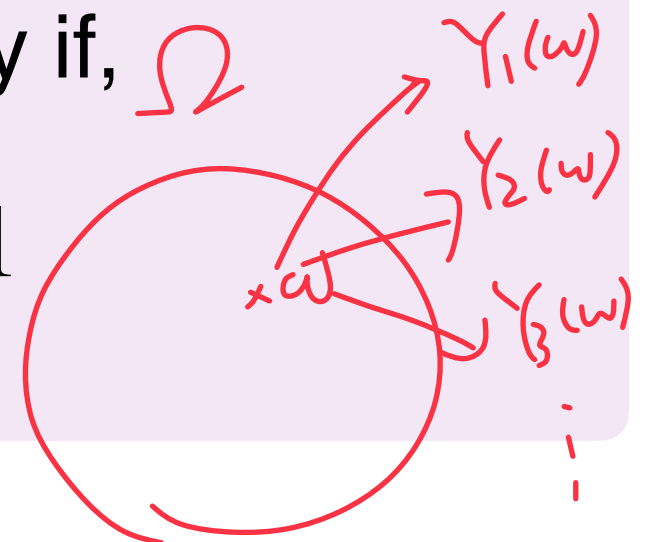
- **Convergence to a Random Variable in Probability**: Let Y_1, Y_2, \dots be a sequence of random variables. We say that Y_n converges to a random variable Y in probability if $\forall \varepsilon > 0$,

$$\lim_{n \rightarrow \infty} P(\{\omega : |Y_n(\omega) - Y(\omega)| > \varepsilon\}) = 0$$

- **Convergence to a Random Variable Almost Surely**: Let Y_1, Y_2, \dots be a sequence of random variables. We say that Y_n converges to a random variable Y almost surely if,

$$P(\{\omega : \lim_{n \rightarrow \infty} Y_n(\omega) = Y(\omega)\}) = 1$$

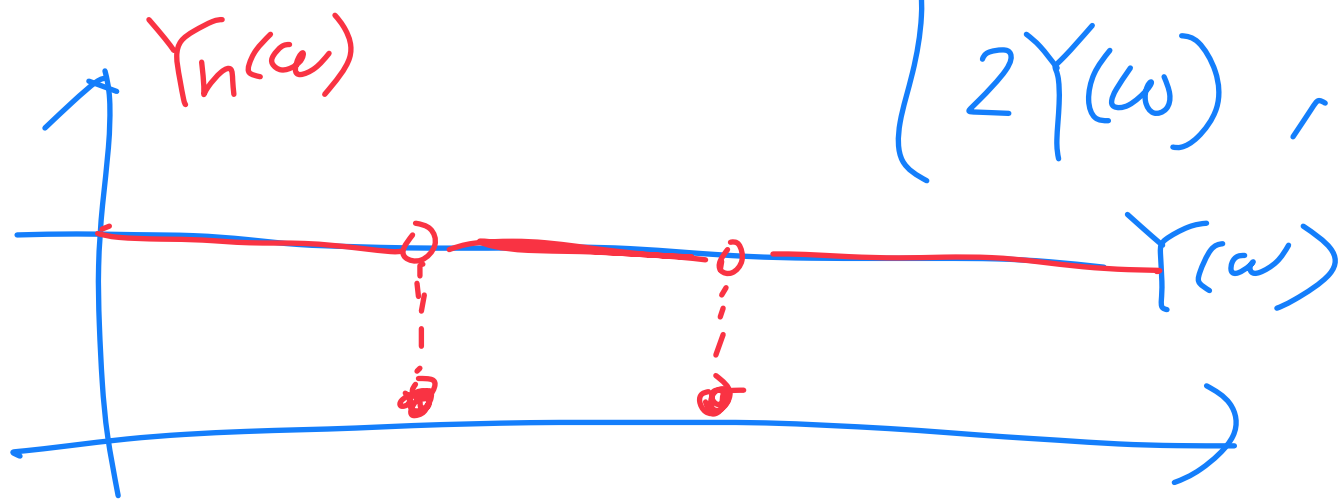
Good event



$$\lim_{n \rightarrow \infty} P(\{\omega: Y_n(\omega) \neq Y(\omega)\}) = 0$$

$$P(\{\omega: \lim_{n \rightarrow \infty} Y_n(\omega) = Y(\omega)\}) = 1$$

$$\omega \quad Y_n(\omega) = \begin{cases} Y(\omega), & \text{if } n \text{ is odd} \\ 2Y(\omega), & \text{if } n \text{ is even} \end{cases}$$



Convergence in Probability, But Not Almost Surely

► **Example:** Let X be a continuous uniform r.v. on $(0,1)$

indicator function

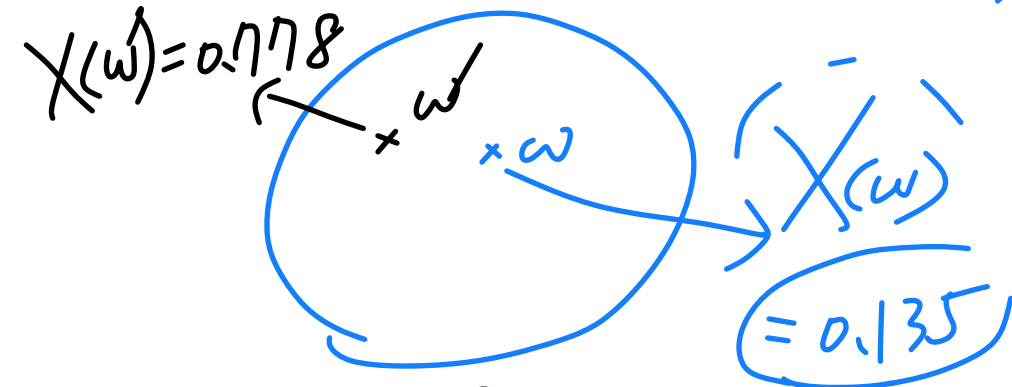
► Consider a sequence of r.v.s X_1, X_2, \dots as follows: $\Omega \rightarrow \text{Real numbers}$

$$X_1 = \mathbb{I}\{X \in [0,1]\}$$

$$X_1(\omega) = 1$$

$$X_2(\omega) = 1$$

$$X_2(\omega') = 0$$



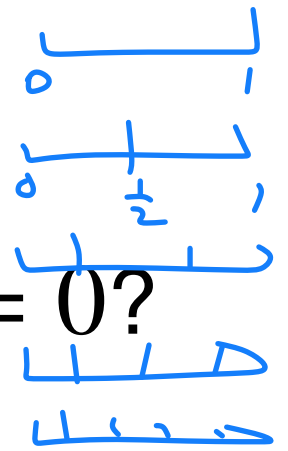
$$X_2 = \mathbb{I}\{X \in [0, \frac{1}{2}]\}$$

$$X_3 = \mathbb{I}\{X \in [\frac{1}{2}, 1]\}$$

$$X_4 = \mathbb{I}\{X \in [0, \frac{1}{3}]\}$$

$$X_5 = \mathbb{I}\{X \in [\frac{1}{3}, \frac{2}{3}]\}$$

$$X_6 = \mathbb{I}\{X \in [\frac{2}{3}, 1]\}$$



► **Question:** Do we have $\lim_{n \rightarrow \infty} P(\{\omega : |X_n(\omega) - 0| > \varepsilon\}) = 0$?

(Fix $\varepsilon = 0.01$)

$$P(\{\omega : |X_1(\omega) - 0| > \varepsilon\}) = 1$$

$$P(\{\omega : |X_2(\omega) - 0| > \varepsilon\}) = \frac{1}{2}$$

$$P(\{\omega : |X_3(\omega) - 0| > \varepsilon\}) = \frac{1}{2}$$

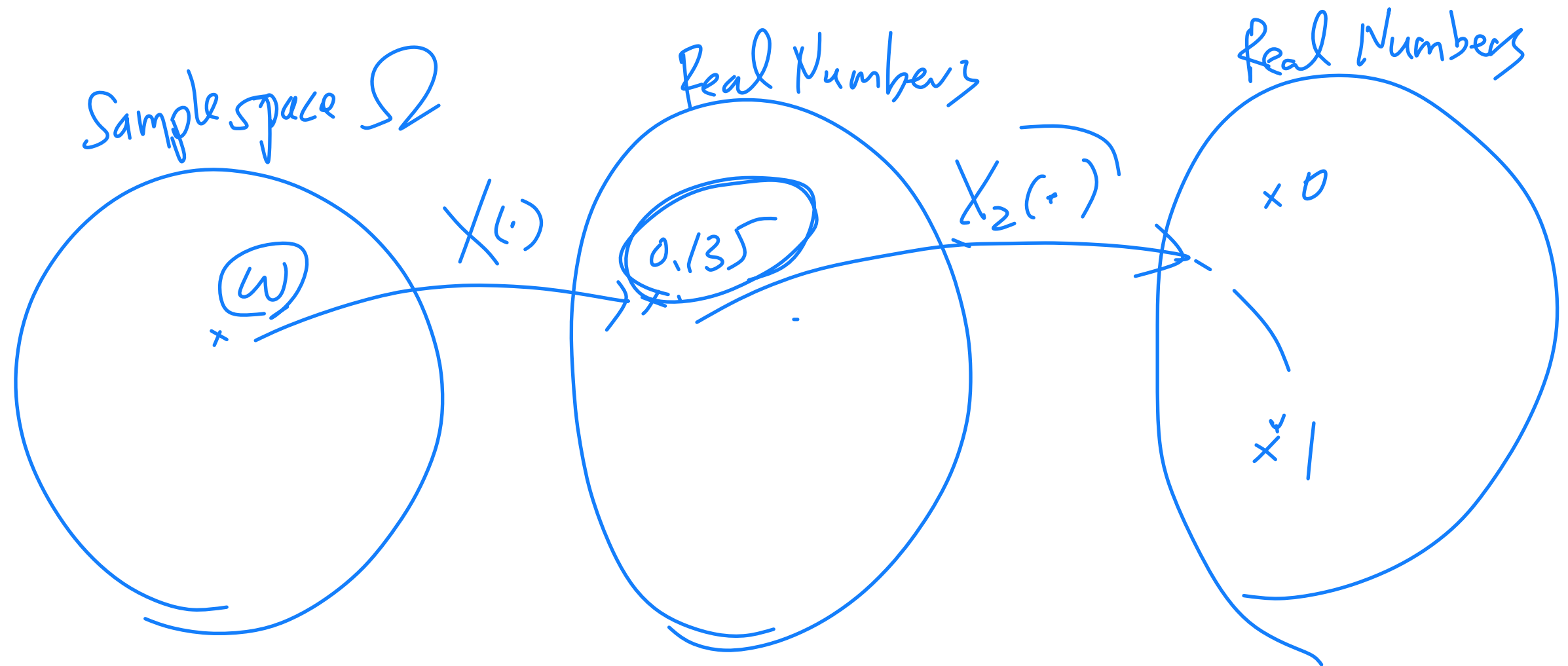
$$P(\{\omega : |X_4(\omega) - 0| > \varepsilon\}) = \frac{1}{3}$$

$$P(\{\omega : |X_5(\omega) - 0| > \varepsilon\}) = \frac{1}{3}$$

bad event

Conclusion:
 $X_n \xrightarrow{P} 0$

$$\Rightarrow \lim_{n \rightarrow \infty} P(\{\omega : |X_n(\omega) - 0| > \varepsilon\}) = 0 \text{ for any } \varepsilon > 0$$



Convergence in Probability, But Not Almost Surely (Cont.)

► **Example:** Let X be a continuous uniform r.v. on $(0,1)$

► Consider a sequence of r.v.s X_1, X_2, \dots as follows:

$$X_1 = \mathbb{I}\{X \in [0,1]\}$$

$$X_2 = \mathbb{I}\{X \in [0, \frac{1}{2}]\} \quad X_3 = \mathbb{I}\{X \in [\frac{1}{2}, 1]\}$$

$$X_4 = \mathbb{I}\{X \in [0, \frac{1}{3}]\} \quad X_5 = \mathbb{I}\{X \in [\frac{1}{3}, \frac{2}{3}]\} \quad X_6 = \mathbb{I}\{X \in [\frac{2}{3}, 1]\}$$

$$X_7 = \dots [0, \frac{1}{4}]$$

► **Question:** Do we have $P(\{\omega : \lim_{n \rightarrow \infty} X_n(\omega) = 0\}) = 1$?

$$X_1(\omega^*) = 1$$

$$X_2(\omega^*) = 1 \quad X_3(\omega^*) = 0$$

$$X_4(\omega^*) = 1 \quad X_5(\omega^*) = 0$$

$$X_7(\omega^*) = 1 \quad X_8(\omega^*) = 0$$

$$X_6(\omega^*) = 0$$

$$X_9(\omega^*) = 0$$

$$X_{10}(\omega^*) = 0$$

Conclusion:

$$X_n \xrightarrow{\text{a.s.}} 0$$

Equivalent Definition of Almost-Sure Convergence

Def ①

$Y_n \xrightarrow{a.s.} Y$

► **Almost-Sure Convergence:** $P(\{\omega : \lim_{n \rightarrow \infty} Y_n(\omega) = Y(\omega)\}) = 1$

$n \rightarrow \infty$

Good event

► **Equivalent Definition of Almost-Sure Convergence:** Let Y_1, Y_2, \dots be a sequence of random variables. We say that Y_n converges to a random variable Y almost surely if for all $\varepsilon > 0$,

Def ②

$$P\left(\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} \underbrace{\{\omega : |Y_n(\omega) - Y(\omega)| > \varepsilon\}}_{A_n}\right) = 0$$

$$\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} A_n$$

$$P\left(\left\{\omega : \lim_{n \rightarrow \infty} Y_n(\omega) \neq Y(\omega)\right\}\right) = 0$$

(Fix $\varepsilon > 0$)

bad event

$|Y_n(\omega) - Y(\omega)| > \varepsilon$ for infinitely many times

WLLN vs SLLN

- **The Weak Law of Large Numbers (Khinchin's Law)**: Let X_1, \dots, X_n be a sequence of independent and identically distributed (i.i.d.) random variables with mean μ . Define $S_n = (X_1 + \dots + X_n)$. Then, for every $\varepsilon > 0$, we have

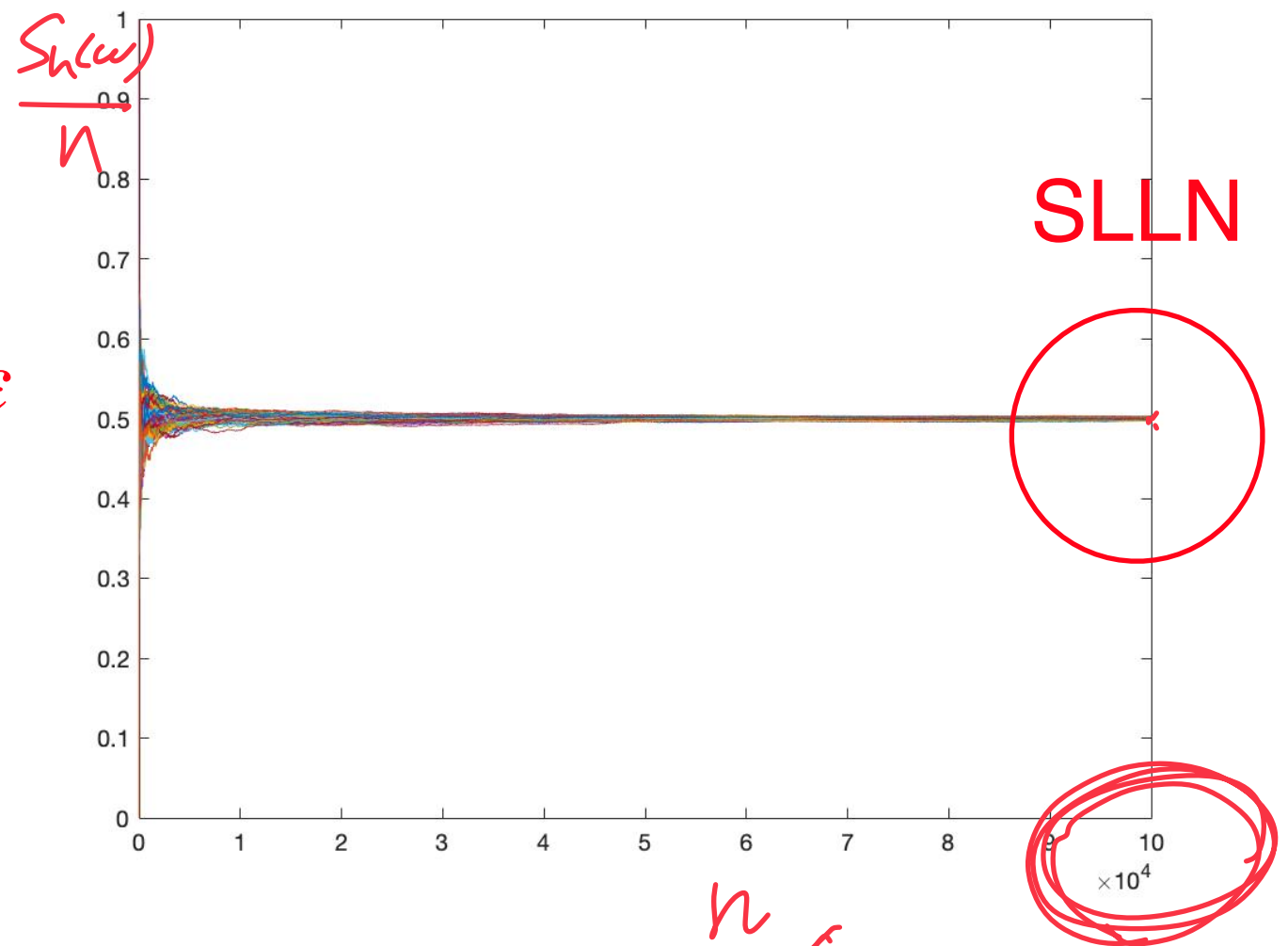
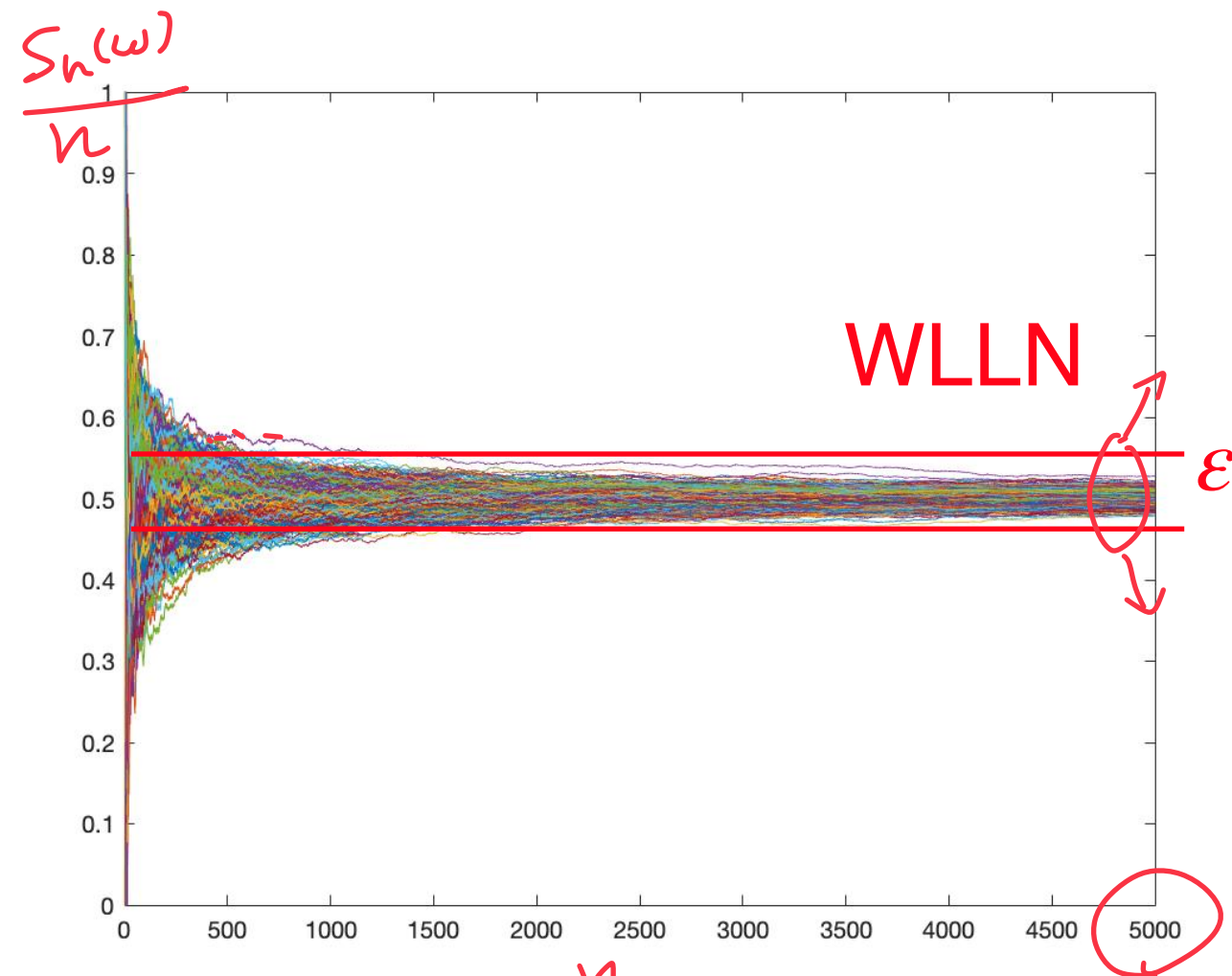
$$\lim_{n \rightarrow \infty} P\left(\left\{\omega : \left|\frac{S_n(\omega)}{n} - \mu\right| \geq \varepsilon\right\}\right) = 0$$

- **The Strong Law of Large Numbers**: Let X_1, \dots, X_n be a sequence of independent and identically distributed (i.i.d.) random variables with mean μ . Define $S_n = (X_1 + \dots + X_n)$. Then, we have

$$P\left(\left\{\omega : \lim_{n \rightarrow \infty} \frac{S_n(\omega)}{n} = \mu\right\}\right) = 1$$

Visualization of WLLN and SLLN

- Example: $X_i \sim \text{Bernoulli}(0.5)$ and $S_n = X_1 + \dots + X_n$

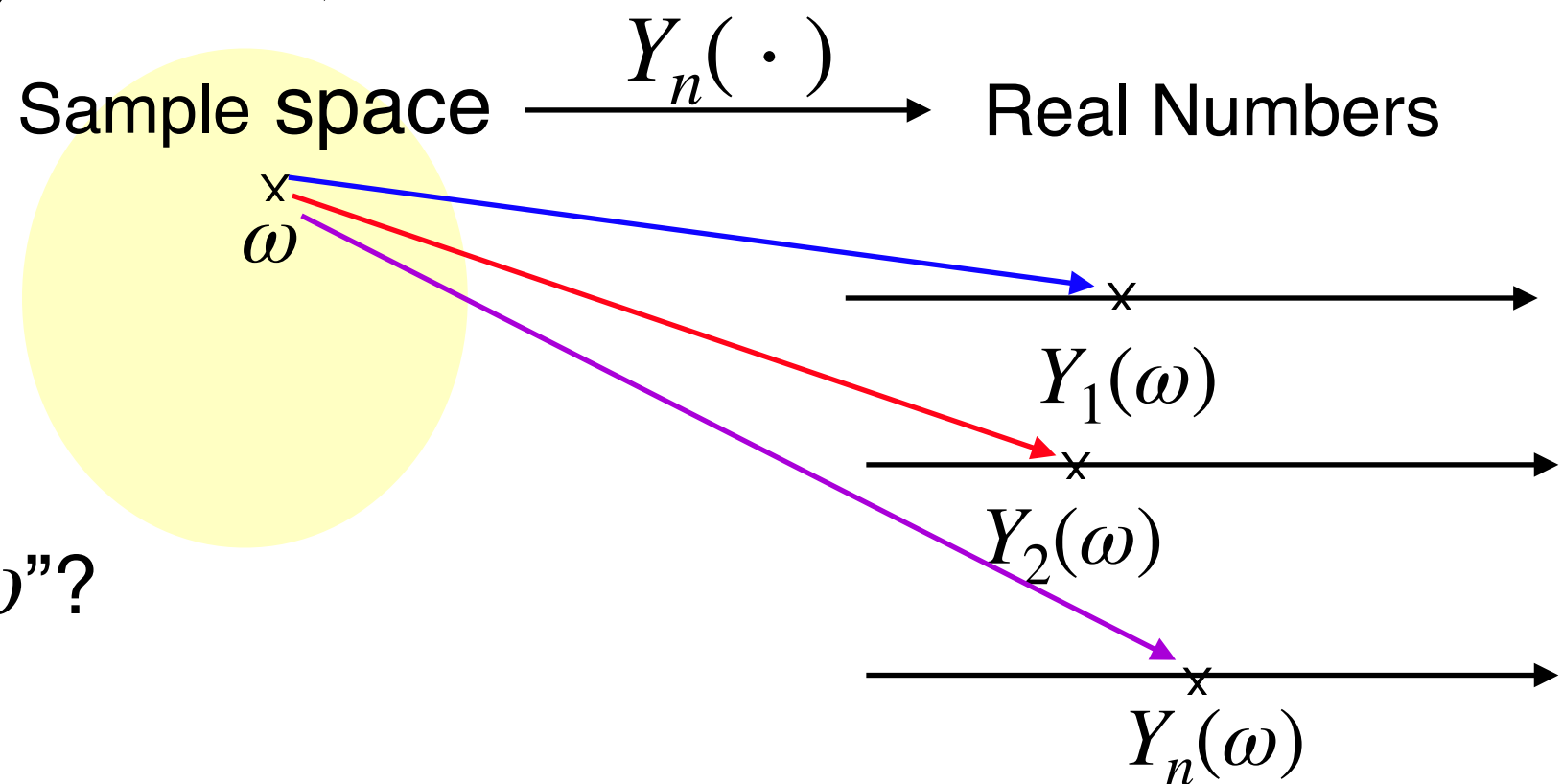


$$\lim_{n \rightarrow \infty} P\left(\underbrace{\left\{\omega : \left|\frac{S_n(\omega)}{n} - \mu\right| \geq \epsilon\right\}}_{\text{bad event}}\right) = 0$$

$$P\left(\left\{\omega : \lim_{n \rightarrow \infty} \frac{S_n(\omega)}{n} = \mu\right\}\right) = 1$$

How to Interpret SLLN?

- ▶ Let X_1, X_2, \dots be a sequence of i.i.d. random variables with mean μ
- ▶ Define $Y_n = (X_1 + X_2 \dots + X_n)/n$
- ▶ **SLLN**: $P\left(\left\{\omega : \lim_{n \rightarrow \infty} \frac{S_n(\omega)}{n} = \mu\right\}\right) = 1$



- ▶ **Question**: What is an " ω "?

How to Prove SLLN (Under a Mild Condition)?

1. Borel-Cantelli Lemma
2. A Bound for the 4-th Moment Condition
3. Markov's Inequality

1. Borel-Cantelli Lemma

► Recall: HW1, Problem 3

Problem 3 (Continuity of Probability Functions)

(12+12=24 points)

(a) Let A_1, A_2, A_3, \dots be a countably infinite sequence of events. Prove that if $\sum_{n=1}^{\infty} P(A_n) < \infty$, then $P(\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} A_n) = 0$. This property is known as the **Borel-Cantelli Lemma**. (Hint: Consider the continuity of probability function for $\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} A_n$ and then apply the union bound)

(b) Consider a countably infinite sequence of coin tosses. The probability of having a head at the k -th toss is p_k , with $p_k = 100 \cdot k^{-N}$ (Note: different tosses are NOT necessarily independent). We use I to denote the event

► **Borel-Cantelli Lemma:** Let $\{A_n\}$ be any sequence of events.

If $\sum_{n=1}^{\infty} P(A_n) < \infty$, then we have

$$P\left(\left\{\omega : \omega \in A_n \text{ for infinitely many } n\right\}\right) = P\left(\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} A_n\right) = 0$$

Review: Proof of Borel-Cantelli Lemma

- ▶ **Borel-Cantelli Lemma:** Let $\{A_n\}$ be any sequence of events. If $\sum_{n=1}^{\infty} P(A_n) < \infty$, then we have
$$P\left(\left\{\omega : \omega \in A_n \text{ for infinitely many } n\right\}\right) = P\left(\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} A_n\right) = 0$$

- ▶ **Proof:**

2. A Bound For 4-th Moment

- ▶ **A Bound on 4-th Moment:** Let X_1, \dots, X_n be a sequence of independent and identically distributed (i.i.d.) random variables with mean μ and $E[X_1^4] < \infty$. Define $S_n = (X_1 + \dots + X_n)$. Then, there exists a constant $K < \infty$ such that

$$E[(S_n - n\mu)^4] \leq Kn^2$$

- ▶ **Proof:** Please see the supplemental on E3
- ▶ **Question:** How about $E[(\frac{S_n}{n} - \mu)^4] \leq ?$

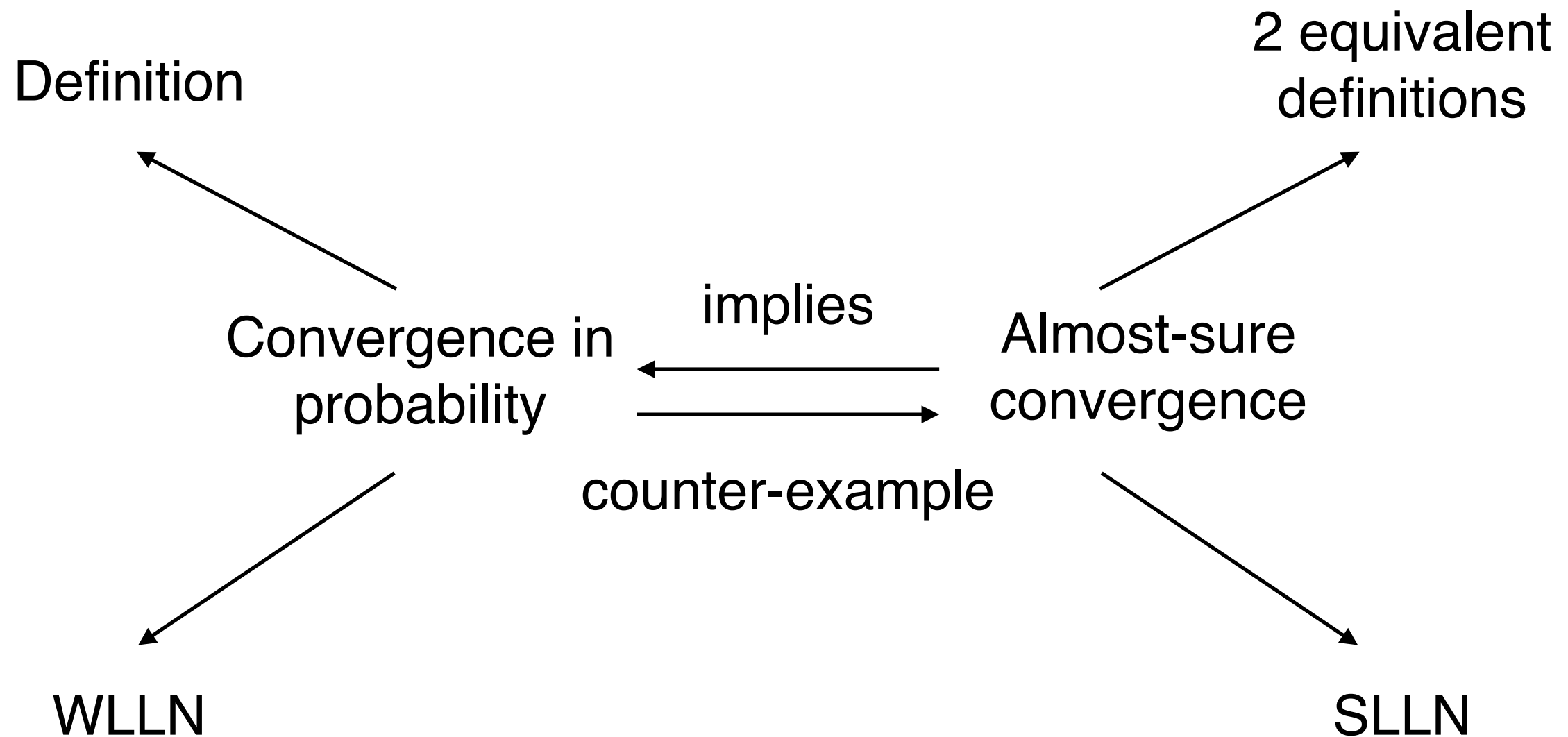
Put Everything Together: Proof of SLLN

► **SLLN:** $P\left(\bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} \left\{\omega : \left|\frac{S_n(\omega)}{n} - \mu\right| > \varepsilon\right\}\right) = 0, \forall \varepsilon > 0$

► **Proof:**

$$P\left(\underbrace{\left\{\left|\frac{S_n}{n} - \mu\right| \geq n^{-\gamma}\right\}}_{A_n}\right) = P\left(\left|\frac{S_n}{n} - \mu\right|^4 \geq n^{-4\gamma}\right) \leq$$

A Quick Summary



Application of SLLN: Monte-Carlo Simulation

- ▶ **A Motivating Example:** Find the following integration

$$I = \int_0^1 e^{-x^3} dx$$

- ▶ **Question:** Is there a closed-form expression for I ?
- ▶ **Question:** How about Riemann integration?

Application of SLLN: Monte-Carlo Simulation (Cont.)

- ▶ **A Motivating Example:** Find the following integration

$$I = \int_0^1 e^{-x^3} dx$$

- ▶ **Monte-Carlo method:** Let $U \sim \text{Unif}(0,1)$

1. Let $U \sim \text{Unif}(0,1)$. Rewrite $\int_0^1 e^{-x^3} dx = E[e^{-U^3}]$

2. Draw K i.i.d. random variables $U_1, \dots, U_K \sim \text{Unif}(0,1)$

$$E[e^{-U^3}] \approx \frac{1}{K} \sum_{i=1}^K e^{-U_i^3} \quad (\text{Why?})$$

Monte-Carlo Simulation (Formally)

► **Objective:** Find the integration $I = \int g(x)dx$

► **Monte-Carlo Simulation:**

1. Let X be a random variable with PDF $p(x)$. Rewrite I as

$$I = \int \frac{g(x)}{p(x)} p(x) dx = E_{p(x)} \left[\frac{g(X)}{p(X)} \right]$$

2. Draw K i.i.d. random variables X_1, \dots, X_K with PDF $p(x)$

Construct
$$\hat{I}_K = \frac{1}{K} \sum_{i=1}^K \frac{g(X_i)}{p(X_i)} \approx I$$

► **Question:** $E[\hat{I}_K] = ?$ $\text{Var}[\hat{I}_K] = ?$

► **Question:** How to choose K and $p(x)$?