Lecture 32:
Monte Carlo Method
&
Hypothesis Testing - Part I

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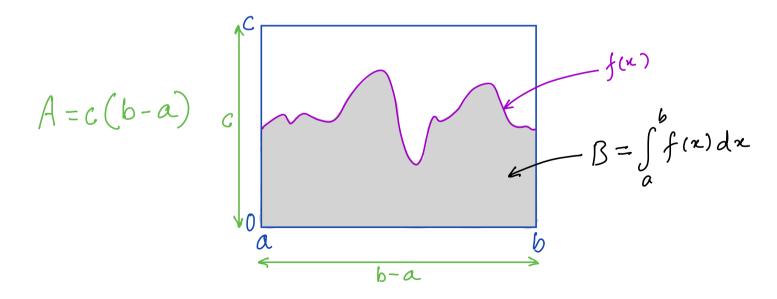
- ► <u>Monte Carlo method</u> is a set of algorithms which <u>use</u> randomness to solve problems either exactly or approximately.
- We will study Monte Carlo integration a method used to approximate a definite integral using random point generation.
- Consider the problem of approximating the definite integral

$$\int_{a}^{b} f(x)dx$$

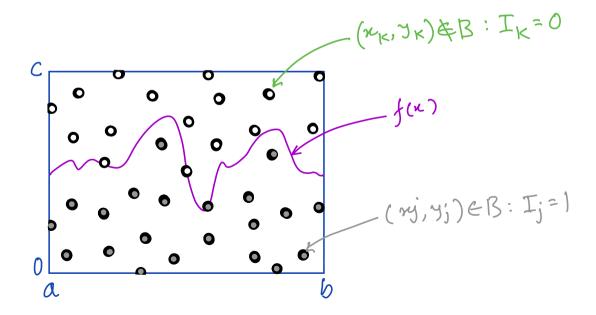
where, f is some complicated function so that exact integration is not possible with any existing method.

- ▶ Assume that $0 \le f(x) \le c$ and hence the integral is finite.
- We can generate and <u>use random numbers to approximate this</u> integral!

- ➤ We can generate and use random numbers to approximate this integral!
- Let A be the area of the rectangle in the (x, y)-plane defined by $a \le x \le b$ and $0 \le y \le c$.
- ▶ Let B be the region under the curve y = f(x) for $a \le x \le b$,
- \triangleright Thus, the area of B is the desired integral.



- Method: take random samples from A, then calculate the proportion of the samples that also fall into the area B.
- ▶ Generate iid points $(X_1, Y_1), \ldots, (X_n, Y_n)$ uniformly over A. Now let I_j be the Bernoulli r.v. such that $I_j = 1$ if $(X_j, Y_j) \in B$ and 0 otherwise.



► Then

$$p = E(I_j) = P(I_j = 1) = \frac{B}{A} = \frac{\int_a^b f(x)dx}{c(b-a)}.$$

ightharpoonup By the WLLN, for large n we can approximate p as

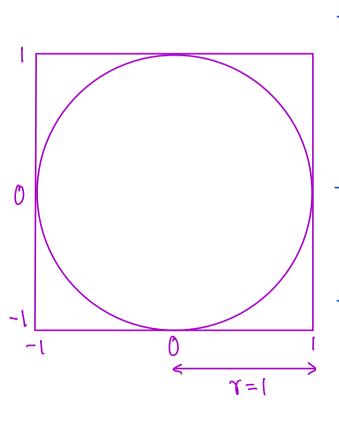
$$\frac{1}{n}\sum_{i=1}^{n}I_{j}.$$

► Hence,

$$\frac{1}{n}\sum_{i=1}^{n}T_{i}\approx\frac{\int_{a}^{b}f(x)dx}{c(b-a)}$$

$$\Rightarrow \int_{a}^{b} f(x) dx \approx \frac{c(b-a)}{n} \sum_{j=1}^{n} I_{j}$$

 \triangleright Example: Approximate the value of the number π .



- crenerate samples (24,4,1), ..., (xn, yn) ild uniformly in the square.

- Let
$$T_j = \begin{cases} 1 & \text{if } r_j^2 + \gamma_j^2 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

-Then, $\pi r^2 = \pi$ $\approx 4 \cdot \frac{1}{n} \sum_{j=1}^{n} I_j$ area of the square

Hypothesis testing

- ▶ We, the scientists/engineers do not only come across problems of point/interval estimation in practice.
- ▶ Often, given a sample, we need to <u>decide</u> whether certain "statement" is true.
- ➤ For example, a medical researcher may decide on the basis of experimental evidence whether coffee drinking increases the risk of cancer in humans.
- ▶ Here "coffee drinking increases the risk of cancer in humans" is a <u>statement</u> or conjecture.
- ▶ A conjecture is a statement which a scientist proposes but correctness of which is yet to be established.
- A data scientist attempts to "test" correctness of the statement based on the available sample/evidence.

Hypothesis testing

- ► A statistical hypothesis is a conjecture concerning a population.
- Whether a hypothesis is true or false is never known with absolute certainty unless we examine the entire population or underlying distribution.
- ▶ We take a random sample from the population of interest and use the data contained in this sample to provide evidence that either supports or does not support the hypothesis.
- Evidence from the sample that is inconsistent with the stated hypothesis leads to a rejection or nullification of the hypothesis.

Hypothesis testing

- Another simple story to understand hypothesis testing: A person is arrested based on a suspicion of committing a crime.
- ▶ The court has the following hypothesis to test: The person is innocent.
- ▶ The goal of the court is to nullify this hypothesis based on evidence.
- ▶ In the court, evidence is presented and examined.
- ▶ Based on the evidence, the jury either fails to reject the hypothesis (the person is innocent) or rejects it (the person is guilty of the crime).
- ► Important thing to note:
- ▶ If the jury fails to reject the hypothesis based on the evidence presented, it does not imply that the person is innocent.
- ▶ It only implies that the evidence was insufficient to convict.