

Overview

- Sampling / LLN
 - ↳ Computing
 - expected values
 - integrals
- Error in Sampling
- RNG
- Errors in general

Announcements

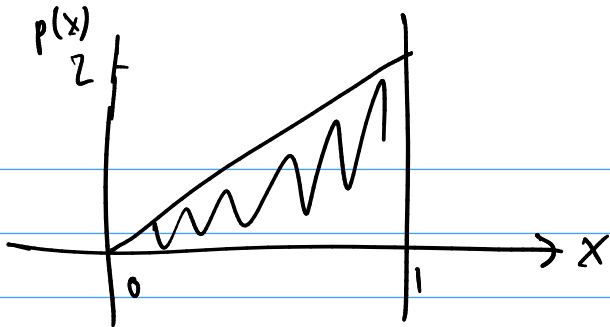
- Quiz 8 : oops
- Examled grades today

LLN:

$$E[X] \approx \frac{1}{N} \sum s_i$$

s_i : samples

↑ from the same distribution
as X



Monte - Carlo algorithm,

- get a random result
- always get a result

Cas Vegas algorithm:

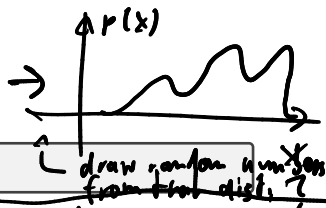
- always computes the correct result (if it computes one)
- but it may only sometimes (randomly) produce a result

Switching Distributions for Sampling

Found:

$$\frac{1}{N} \sum s_i$$

$$E[X] = E \left[\tilde{X} \cdot \frac{p(\tilde{X})}{\tilde{p}(\tilde{X})} \right]$$



Why is this useful for sampling?

$$E[X] = \int x \cdot p(x) dx = \int x \cdot \frac{p(x)}{\tilde{p}(x)} \tilde{p}(x) dx$$

$$= E \left[\tilde{X} \cdot \frac{p(\tilde{X})}{\tilde{p}(\tilde{X})} \right]$$

$$\approx \frac{1}{N} \sum_{i=1}^N s_i \cdot \frac{p(s_i)}{\tilde{p}(s_i)}$$

\tilde{X} distributed according to \tilde{p}

Need $\tilde{p} \neq 0$
wherever $p \neq 0$

In-class activity: Monte-Carlo Methods

$$\int g(x) dx = \int g(x) \cdot \frac{1}{\tilde{p}(x)} \cdot \tilde{p}(x) dx$$

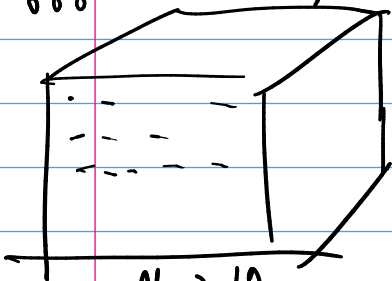
$$= E \left[\frac{g(\tilde{x})}{\tilde{p}(\tilde{x})} \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \frac{g(\tilde{s}_i)}{\tilde{p}(\tilde{s}_i)}$$

$$\int_0^1 \int_0^1 \int_0^1$$

$$f(x, y, z, t)$$

$$dx dy dz$$



$$N \rightarrow 10$$

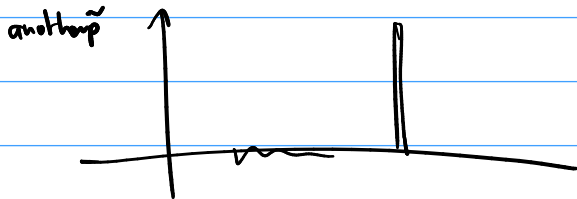
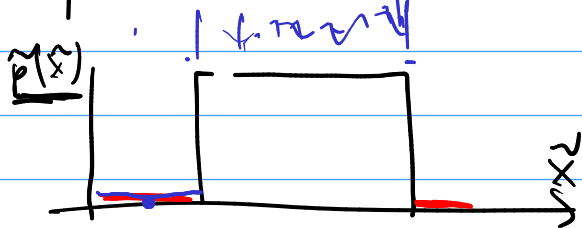
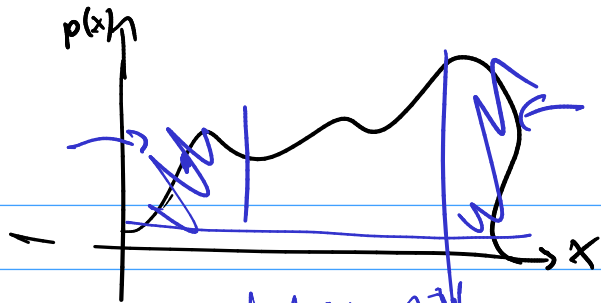
For regular grid: N^3 cost

For MC:



$$\frac{1}{N} \sum_{i=1}^N f(\beta_i)$$

cost: $\sim N$ samples



Expected Value: Example II

What is the expected snowfall in Illinois?

$$E[\text{snow}] = \int_0^1 \int_0^1 \text{Snow}(x,y) p(x,y) dy dx$$

Snow(x,y)

: Expected snowfall
at (x,y)

$\rightarrow p(x,y) = C \cdot g(x,y)$

$p(x,y)$ = prob. with
which
(x,y)
is in IL

$$g(x,y) = \begin{cases} 1 & \text{in IL} \\ 0 & \text{outside of IL} \end{cases}$$



$C = \frac{1}{\iint g(x,y)} \leftarrow = \text{Area of IL?}$ (not quite; need with units not (0,0))

Choose \tilde{p} uniform on
[0,1] x [0,1]

$$\frac{1}{C} = \int_0^1 \int_0^1 g(x,y) dx dy$$

$$= \int_0^1 \int_0^1 \frac{g(x,y)}{\tilde{p}(x,y)} \tilde{p}(x,y) dx dy$$

$$= E \left[\frac{g(\tilde{x}, \tilde{y})}{\tilde{p}(\tilde{x}, \tilde{y})} \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \frac{g(\tilde{x}_i, \tilde{y}_i)}{\tilde{p}(\tilde{x}_i, \tilde{y}_i)}$$

Dealing with Unknown Scaling

What if a distribution function is only known up to a constant factor, e.g.

$$p(x) = C \cdot \underbrace{\begin{cases} 1 & \text{point } x \text{ is in } \mathbb{L}, \\ 0 & \text{it isn't.} \end{cases}}_{q(x)}$$

Typically $\int_{\mathbb{R}} q \neq 1$. We need to find C so that $\int p = 1$, i.e.

$$C = \frac{1}{\int_{\mathbb{R}} q(x) dx}.$$

Idea: Use sampling.

Demo: Computing π using Sampling

Demo: Errors in Sampling

Sampling: Error

The **Central Limit Theorem** states that with

$$S_N := X_1 + X_2 + \cdots + X_n$$

for the (X_i) independent and identically distributed according to random variable X with variance σ^2 , we have that

$$\frac{S_N - NE[X]}{\sqrt{\sigma^2 N}} \rightarrow \mathcal{N}(0, 1),$$

i.e. that term approaches the normal distribution. As we increase N , σ^2 stays fixed, so the asymptotic behavior of the error is

$$\left| \frac{1}{N} S_N - E[X] \right| = O\left(\frac{1}{\sqrt{N}}\right).$$