

In this class:

- A quick review of probability: random variables and expectation
- How to compute π with a Monte Carlo method
- How to compute an integral with a Monte Carlo method
- The central limit theorem and what it says about Monte Carlo

September 14, 2016

Monte Carlo Methods

Next class

HW Due, Catchup and Review!
G&C – Chapter 3

Next next class

Review for Midterm, Variance

Random Variables

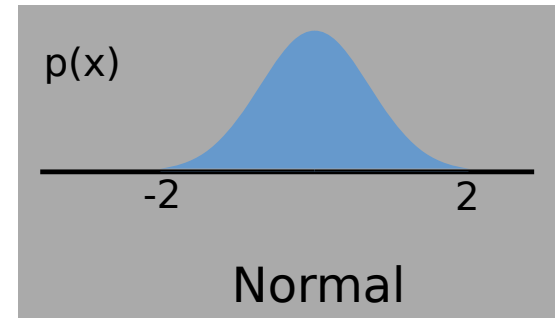
X is a random variable (R.V.)

- Discrete RV

coin toss, dice roll, Monty Hall problem

- Continuous RV

uniform, normal



Expectation

the “expectation” is akin to the “average”

$$E[X] = \sum_{v=\text{values of } X} v \cdot \text{Prob}[X = v]$$

$$E[\text{dice}] = 1 \cdot 1/6 + 2 \cdot 1/6 + 3 \cdot 1/6 + \dots$$

$$E[X] = \int_{\mathcal{D}} xp(x) dx$$

Expectation

Linear $E[aX + bY] = E[aX] + E[bY]$

if a and b are non-random constants

Non-multiplicative $E[XY] \neq E[X]E[Y]$

(unless X and Y are independent)

Expectation Example

$X \sim \text{Uniform}[0, 1], Y \sim \text{Uniform}[0, 1]$

$$Z = XY$$

$$E[Z] = E[XY] = \int_0^1 \int_0^1 xy \, dx \, dy$$

$$E[Z] = 1/4$$

Preview of Central Limit Theorem

The CLT is why Monte Carlo works!

X_1, X_2, X_3, \dots are iid copies

independent, identically distributed

think independent dice rolls

$X_1 \sim X$ dice roll 1

$X_2 \sim X$ dice roll 2

Preview of Central Limit Theorem

The CLT is why Monte Carlo works!

X_1, X_2, X_3, \dots are iid copies

independent, identically distributed

The CLT gives a relationship between

$$\frac{1}{N} \sum_{i=1}^N X_i \quad \leftrightarrow \quad E[X]$$

Average of N samples of X.
What we compute with MC!

The formal expectation!