

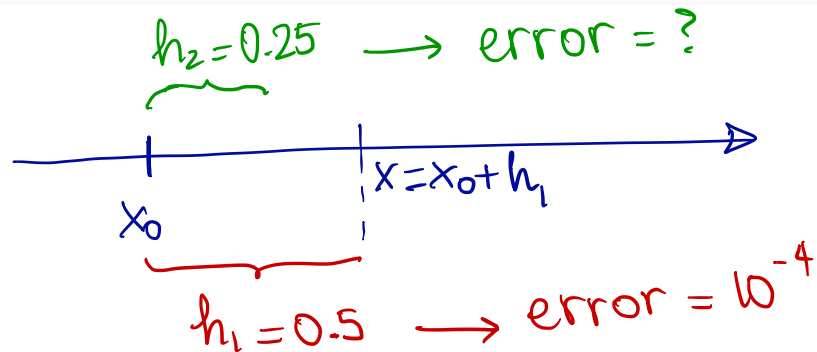
Making error predictions

Suppose you expand $\sqrt{x - 10}$ in a Taylor polynomial of degree 3 about the center $x_0 = 12$. For $h_1 = 0.5$, you find that the Taylor truncation error is about 10^{-4} .

What is the Taylor truncation error for $h_2 = 0.25$?

$$f(x) = \sqrt{x - 10}$$

$$t_3(x) = \sum_{i=0}^3 \frac{f^{(i)}(12)}{i!} \underbrace{(x-12)^i}_h$$



$$\text{error} = O(h^4)$$

$$\frac{e_1}{e_2} = \frac{h_1^4}{h_2^4} \rightarrow e_2 = \left(\frac{h_2}{h_1}\right)^4 e_1 = \left(\frac{0.25}{0.5}\right)^4 10^{-4} =$$

Using Taylor approximations to obtain derivatives

Let's say a function has the following Taylor series expansion about $x = 2$.

$$f(x) = \underbrace{\frac{5}{2} - \frac{5}{2}(x-2)^2 + \frac{15}{8}(x-2)^4 - \frac{5}{4}(x-2)^6 + \frac{25}{32}(x-2)^8 + O((x-2)^9)}_{t_8(x)}$$

$$t_4(x) = \frac{5}{2} - \frac{5}{2}(x-2)^2 + \frac{15}{8}(x-2)^4$$

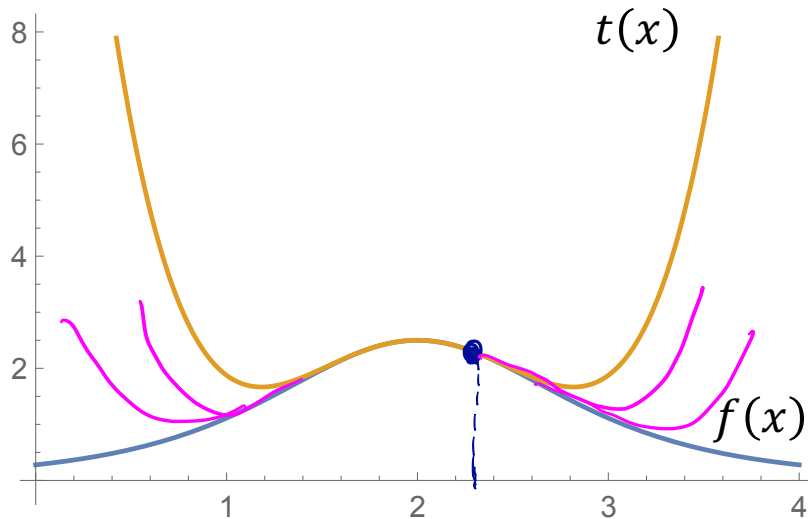
$$t_4'(x) = -\frac{5}{2}(2)(x-2) + \frac{15}{8}(4)(x-2)^3$$

$$t_4'(2.3) = -1.2975$$

(the exact value is $f'(2.3) = -1.3144$)

→ can we use this same approximation
at $x = 3$?

- increase degree
- move expansion
point



Monte Carlo

Randomness

What types of problems can we solve with the help of random numbers?

We can compute (potentially) complicated averages:

1. Where does “the average” web surfer end up? (PageRank)
2. How much is my stock portfolio/option going to be worth?
3. What are my odds to win a certain competition?

Random number generators

- Computers are deterministic - operations are reproducible
- How do we get random numbers out of a deterministic machine?

Demo “Playing around with random number generators”

- Pseudo-random numbers
 - Numbers and sequences appear random, but they are in fact reproducible
 - Good for algorithm development and debugging
- How truly random are the pseudo-random numbers?

Example: Linear congruential generator

$$x_0 = \textit{seed}$$

a: multiplier

c: increment

$$x_{n+1} = (a x_n + c) \pmod{M}$$

M: modulus

- If we keep generating numbers using this algorithm, will we eventually get the same number again? Can we define a period?

Good random number generator

- Random pattern
- Long period
- Efficiency
- Repeatability
- Portability

Random variables

We can think of a random variable X as a function that maps the outcome of unpredictable (random) processes to numerical quantities.

Examples:

- How much rain are we getting tomorrow?
- Will my buttered bread land face-down?

random variable
 $X = 80\%$

We don't have an exact number to represent these random processes, but we can get something that represents the **average** case.

To do that, we need to know how likely each individual value of X is.

Discrete random variables

Each random value X takes values x_i with probability p_i

for $i = 1, \dots, m$ and $\sum_{i=1}^m p_i = 1$

Example:



Random variable $\implies X = \#$ top of die
after each roll

Possible values x_i :

$$x_1 = 1 \longrightarrow p_1 = \frac{1}{6}$$

$$x_2 = 2 \longrightarrow p_2 = \frac{1}{6}$$

⋮

$$x_6 = 6 \longrightarrow p_6 = \frac{1}{6}$$

Coin toss example

Random variable X : result of a toss can be heads or tails

$$x_1 = X = 1: \text{toss is heads} \rightarrow p_1 = 0.5$$

$$x_2 = X = 0: \text{toss is tail} \rightarrow p_2 = 0.5$$

Expected value : $E(x) = \sum_{i=1}^m p_i x_i$

Roll : $E(x) = \frac{1}{6}(1) + \frac{1}{6}(2) + \frac{1}{6}(3) + \dots + \frac{1}{6}(6) = \frac{7}{2}$

Toss : $E(x) = \frac{1}{2}(1) + \frac{1}{2}(0) = 0.5$

Coin toss example

(*) expected value

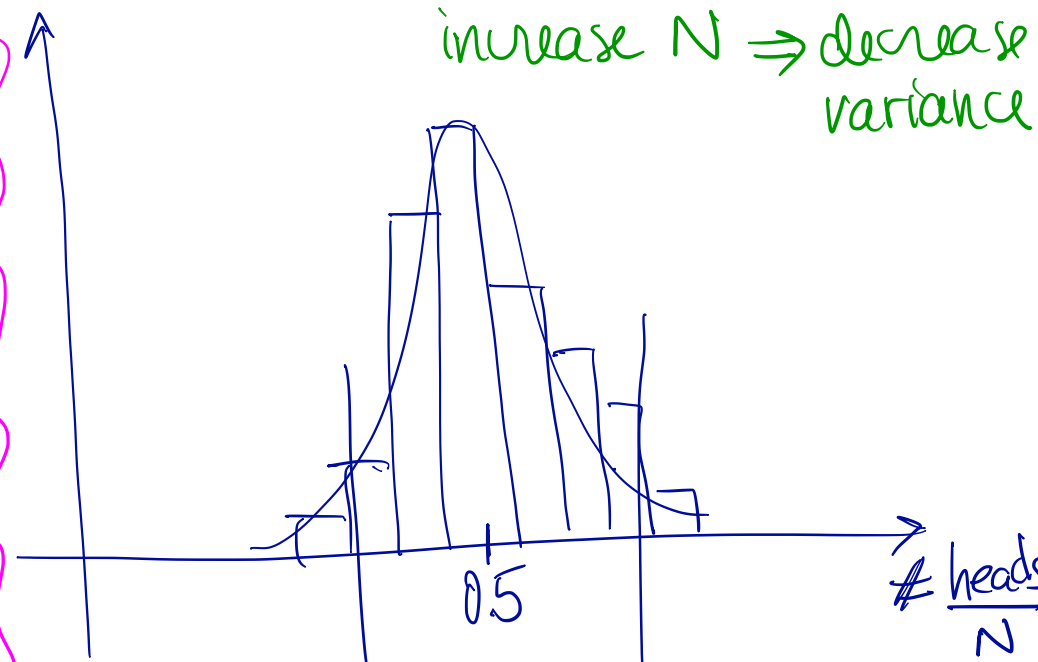
Numerical experiment:

$N = 100$ toss

	#heads	#heads/ N (*)
1	45	0.45
2	52	0.52
3	54	0.54
...		
M	51	0.51

M = number of numerical experiments

occurrences



increase M \Rightarrow better estimate for expected value

increase N \Rightarrow decrease variance

LIVE DEMO

decrease variance
 \Downarrow
increase N

Texas Holdem Game

Question: for each starting pair of cards, what is the probability of winning?

1 Numerical experimental \Rightarrow play N games

Game: set of 7 cards

\Downarrow
tie, win
or loss

Starting hand

dealer hand

Opponent hand

BLIND BET	
ONLY HIGHEST WIN AWARDED WHEN DEALER IS BEATEN	
Royal Flush	500:1
Straight Flush	50:1
Four of a Kind	10:1
Full House	3:1
Flush	3:2
Straight	1:1
All Other	Push

TRIPS BET	
ONLY HIGHEST WIN AWARDED BET PAYS EVEN IF YOU FOLD	
Royal Flush	50:1
Straight Flush	40:1
Four of a Kind	30:1
Full House	8:1
Flush	7:1
Straight	4:1
Three of a Kind	3:1

Texas Holdem Game

Question: for each starting pair of cards, what is the probability of winning?

Starting hand (deterministic variable **S**):



→ Fixed

Dealer hand (random variable **D**):



Opponent hand (random variable **O**):



↳ each "game" generates these 7 cards at random

for $i = 1, N$ (games)

generate D, O

who win(S, D, O)

→ use poker rules to decide who wins.

Texas Holdem Game

$$X = \text{Win}(S, O, D)$$

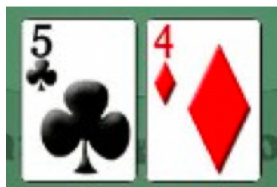
$X = [1, 0, 0]$: starting hand wins

$X = [0, 1, 0]$: starting hand loses (opponent wins)

$X = [0, 0, 1]$: tie

odd of start hand

$$\text{winning} = \frac{\# \text{time } S \text{ win}}{N}$$



Numerical experiment of $N=50$ games

game 1 $\rightarrow X = \text{Who Win}(S, O, D) = [0, 1, 0]$

2 $\rightarrow X = [1, 0, 0]$

⋮

$N \rightarrow$

$X = [0, 1, 0]$

(+) $[\# \text{time } S \text{ win}, \# \text{time } O \text{ win}, \# \text{time tie}]$

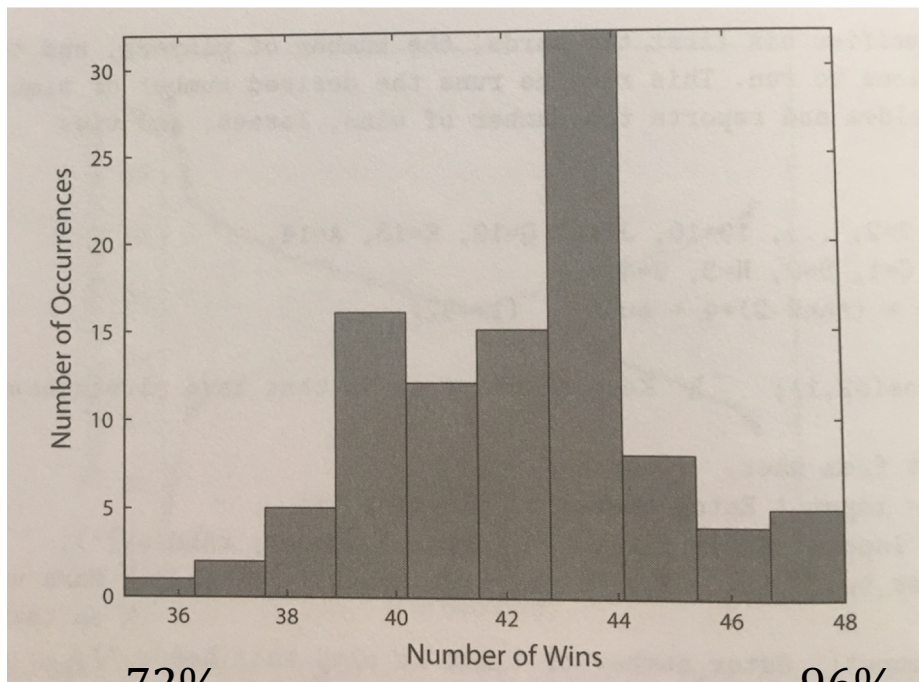
Texas Holdem Game

increase $N \rightarrow$ reduce variance

Starting hand: pair of aces

$M =$

Plotting the number of wins for 100 numerical experiments

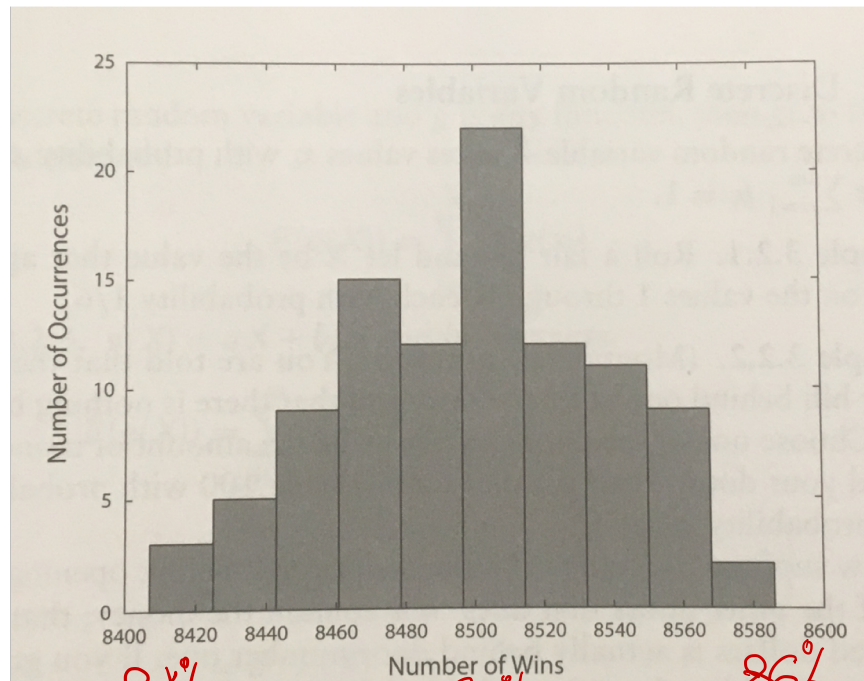


72%

84%

96%

$N = 50$ games



84%

85%

86%

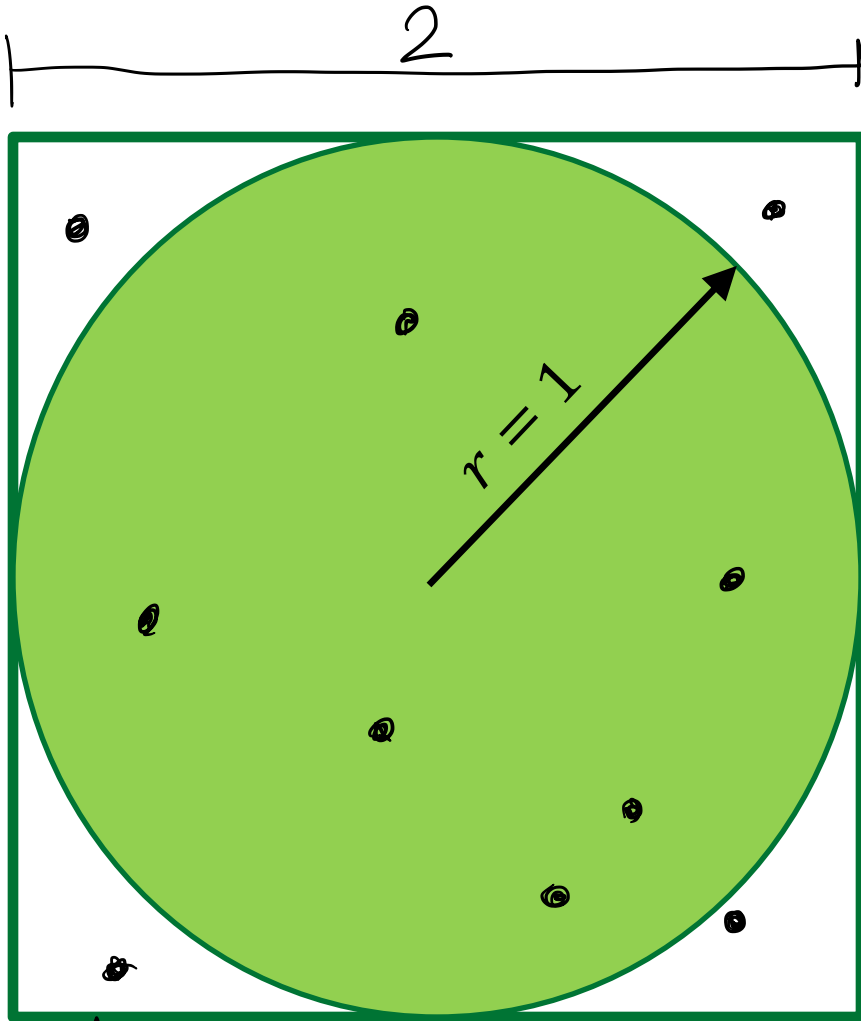
$N = 10,000$ games

Monte Carlo methods

- You just implemented an example of a Monte Carlo method!
- Algorithm that compute APPROXIMATIONS of desired quantities based on randomized sampling

→ Often used to approximate areas/volumes of complicated surfaces.

Example: Approximate the number π



1 numerical experiment:

- sample N points inside domain

- count # points that are inside circle $\rightarrow N_o$

- $A_{\square} \propto N_{\square} = N$

- $A_o \propto N_o$

$$\pi r^2 = 4N_o/N$$

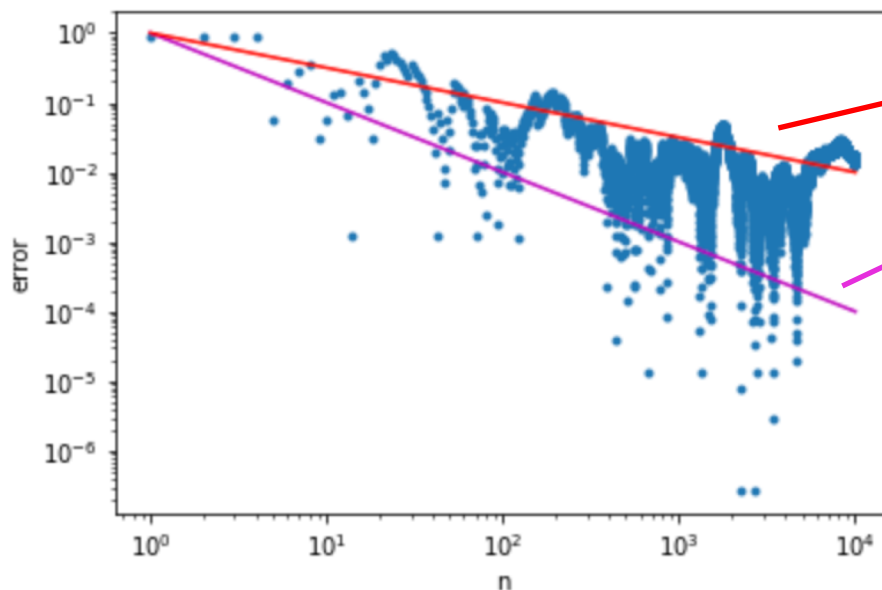
$$\frac{A_{\square}}{A_o} = \frac{N}{N_o} \Rightarrow A_o = \frac{N_o}{N} A_{\square} \Rightarrow$$

$$A_o = \frac{4N_o}{N}$$

$$(\pi)_{\text{approx}} = \frac{4N_o}{N}$$

What can we learn about this simple numerical experiment?

- What is the cost of this numerical experiment? What happens to the cost when we increase the number of sampling points (n)?
- Does the method converge? What is the error?



$$\text{error} = O\left(\frac{1}{\sqrt{n}}\right) = O(n^{-1/2})$$

$$\text{error} = O\left(\frac{1}{n}\right) = O(n^{-1})$$

- CONS: Slow convergence rate when using Monte Carlo Methods
- PROS: Efficiency does not degrade with increase in the dimension of the problem (try to modify the demo to approximate the area of an sphere)