# COMPSCI 630 Randomized Algorithm

Spring 2016

Rong Ge

#### Basic Info

 Web Page: <a href="http://www.cs.duke.edu/courses/spring16/compsci">http://www.cs.duke.edu/courses/spring16/compsci</a>
630/

Contact Me:

LSRC D226

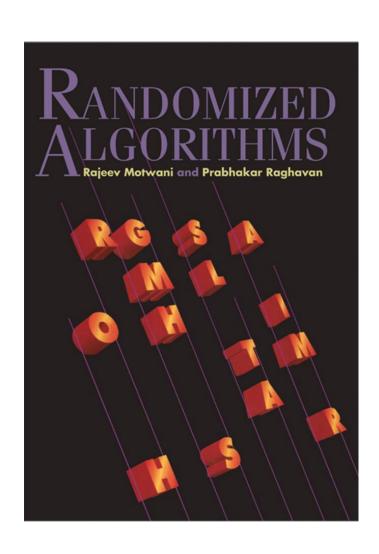
Email: rongge@cs.duke.edu

Office Hour:
 Wednesdays after class at 3:00 – 5:00 pm.

#### Basic Info

- Homework:
   3 or 4 problem sets, due 2 weeks after posted.
   Discussions are allowed, but must acknowledge.
- Piazza: Use Piazza for clarifications http://piazza.com/duke/spring2016/cps630
- Final Exam:
   There will be a take-home final for this course.
- Homework/Final will mostly involve proofs for correctness/running time of algorithms.

# Textbook



# Lecture 1: Overview

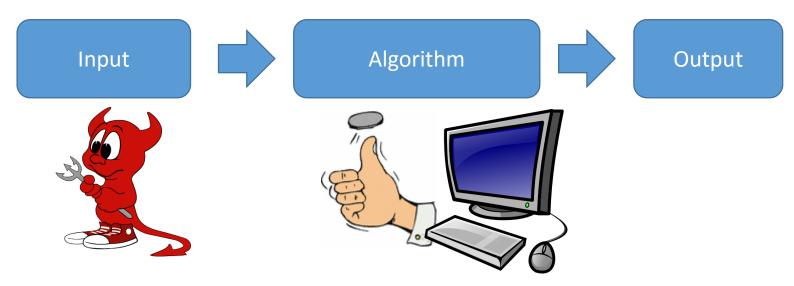
#### Outline

What are randomized algorithms?

Why do we need randomness?

How do we analyze randomized algorithms?

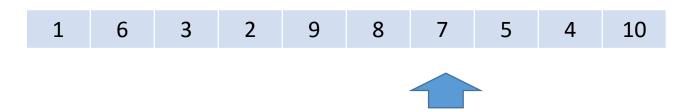
# Randomized Algorithm



- Input is still deterministic (worst-case analysis).
- Output is "correct" with probability at least 2/3 for every input.

# Example 1

Randomized quicksort



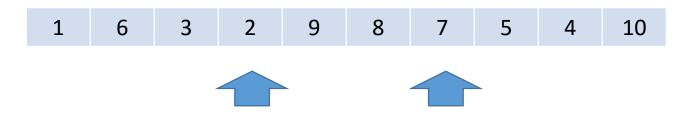
• Randomly pick a pivot, partition into two parts.



Read intro in first Chapter for analysis

# Example 2

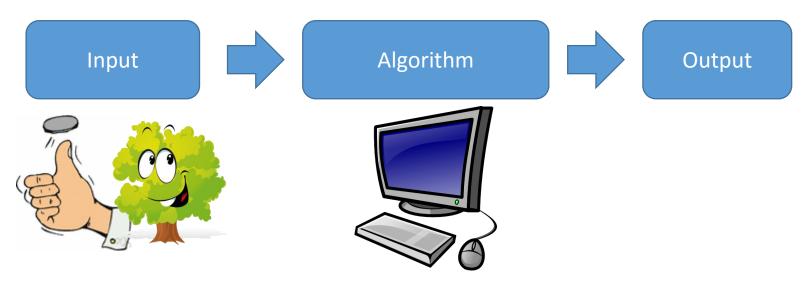
• Find an number that's within top 50% in an array.



Randomly pick two numbers, output the larger one

• Probability of Failure = 
$$\frac{1}{2} \times \frac{1}{2} = \frac{1}{4}$$
!

# Average Case Analysis



- Input is randomly generated (by nature).
- Output is "correct" for most inputs.

# Average vs. Randomized Algorithm

Average case is usually easier
 (If there is a randomized algorithm, there is also a good algorithm for any input distribution.)

They are sometimes related (Min-Max Principal).

Many tools apply to both settings.

# Las Vegas Algorithm



- Always terminates with correct output.
- Running time is random, usually want to analyze the expected running time.

Example: randomized quicksort

	1	6	3	2	9	8	7	5		4	10	
1	6	3	2	5	4			9	8	7	10	0

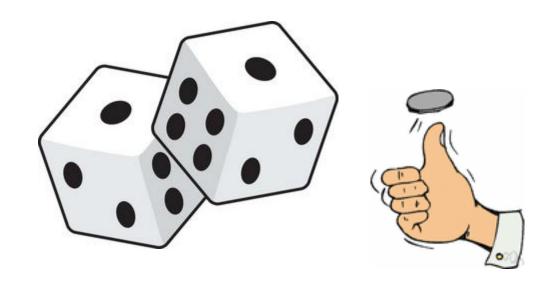


# Monte Carlo Algorithm

- May produce incorrect outputs.
- Sometimes finish in fixed amount of time,
   sometimes can also have a random running time.

Example: Finding a top 50% number in an array.

 Think: Can we transform between Las Vegas and Monte Carlo algorithms?

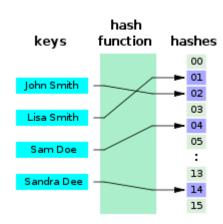


Why is randomness useful?

- Avoid Worst-Case Scenarios.
- Randomized Quicksort

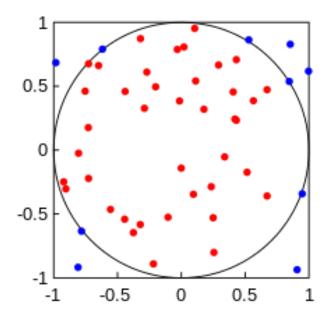


Hashing



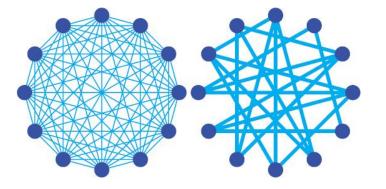
Give a rough estimate very quickly.

Monte-Carlo algorithm for computing volume.

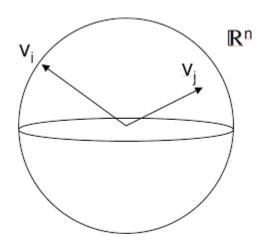


Compress data while preserving crucial information.

- Dimension reduction
- Graph sparsification



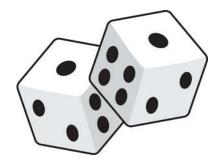
Round Linear Program/Semidefinite Programs.



# cithms

# Tools to Analyze Randomized Algorithms

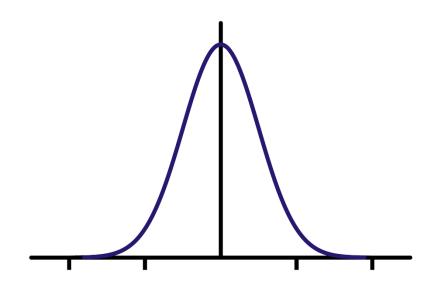
- Basic properties of random variables.
  - Linearity of expectation
  - Independence
  - Union bound
  - Markov/Chebyshev inequalities.
- Chapters 1, 3 in book.



Concentration Bounds, Martingales

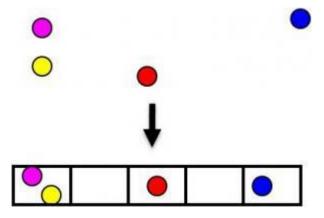
• 
$$f(X) \approx \mathbb{E}[f(X)]$$

• Chapter 4 in book.



Dimension Reduction

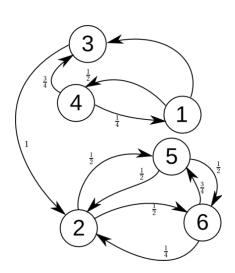
- Useful for fast algorithms with approximate solutions.
- We will see examples in many different settings.



Markov Chains

Powerful tool for sampling/counting.

• Chapters 6, 11 in book



Complexity, derandomization.

 Is there a problem that has a polynomial time randomized algorithm but no deterministic polynomial time algorithm?

#### What's not covered in detail

- Game theory
- Randomization is inherent in the solution space!
- CPS 590.4 Computational Microeconomics: Game Theory, Social Choice, and Mechanism Design

Read Chapter 2 for a short intro.

#### What's not covered in detail

- Online Algorithms
- Randomization is crucial in algorithms/analysis.
- CPS590 Optimization and Decision-making under Uncertainty

Read Chapter 13 for a short intro.

#### What's not covered in detail

- Geometry Algorithms
- Many elegant applications of randomness.
- CPS634 Geometric Algorithms

Read Chapter 10 for a few examples.