

# Randomized Algorithms

## Introduction

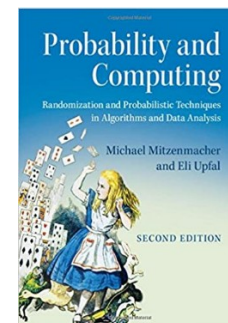
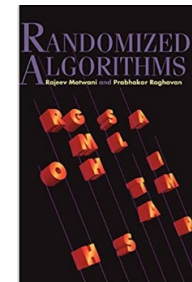
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# Self Introduction

- **Ph.D.:** CSIE, National Chung Cheng University, 2011.
  - DAAD-NSC Sandwich Program (2007–2008).
  - Dissertation supervisors: Maw-Shang Chang & Peter Rossmanith (RWTH Aachen)
- **Postdoc** in Genomics Research Center, Academia Sinica (2011–2014).
- **Postdoc** in Institute of Information Science, Academia Sinica (2014–2018).
- **Quantitative Analyst (intern)** of Point72/Cubist Systematic Strategies (2018–2019).
- **Quantitative Analyst** of Seth Technologies Inc. (2020–2021/01).

# Textbooks and Materials

- Textbooks:
  - *Randomized Algorithms*. Motwani, R. and Raghavan, P., 1995. Cambridge University Press.
  - *Probability and Computing: Randomized Algorithms and Probabilistic Analysis*. M. Mitzenmacher and E. Upfal, 2005.
- Other materials:
  - Prepared slides.



# Prerequisites

- Basic undergraduate courses in
  - Algorithms
  - Data structures
  - Probability theory
  - Discrete mathematics
- Motivation.
- Curiosity.

# Topics

- Examples of Probability Paradoxes
- Las Vegas and Monte Carlo
- Randomized Quicksort
- Chernoff Bounds
- The Stable Marriage Problem
- The Coupon Collector's Problem
- The Secretary Problem
- Random Graphs
- Random Treaps
- Markov Chains (Optional)
- Monte Carlo Simulation (Optional)

# Grading Policy

- Assignments (x 10, 50%)
- Programming Project bonus (x  $n$ , 5%)
- Final Team Report (50%)
  - Peer-Grading Mechanism.
  - Election Game Equilibrium.
  - Presentation for other selected topics.

# Traditional deterministic algorithms



# Randomized algorithms

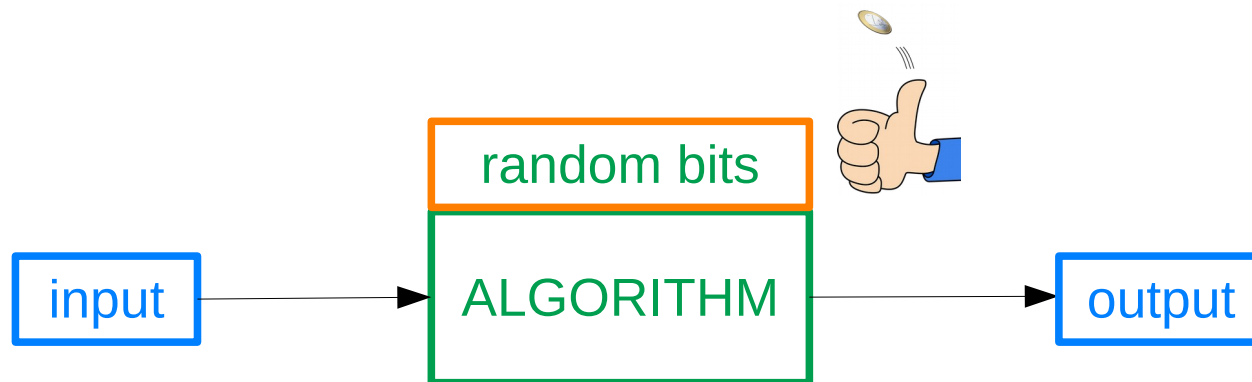


figure freely downloaded from <https://pixabay.com/illustrations/coin-flipping-coin-hand-flip-flick-5822271/>  
*Randomized Algorithms, CSIE, TKU, Taiwan*



# Why?

- Randomized algorithms are
  - often much ***simpler*** than the best known deterministic ones.
  - often much ***more efficient*** (faster or using less space) than the best known deterministic ones.
- Sometimes ideas from the randomized algorithms lead to good deterministic algorithms.

# Comparisons

- It's different from the *average-case* analysis of deterministic algorithms.
  - e.g., expected running time of a deterministic algorithm on input sampled from a distribution.
- In most scenarios, it's NOT a heuristic algorithm.
  - The **accuracy** is guaranteed, or
  - The **running time** is guaranteed.

# An illustrating example:

- **Problem:** find a grade-‘A’ student in a class of  $n$  students where half of them get ‘A’.
- What is the time complexity for the best deterministic algorithm?
  - I mean, in the “worst case”.

# A randomized algorithm (from Wikipedia)

```
findingA_LV(array  $L$ ,  $n$ )
```

```
begin
```

```
  repeat
```

```
    Randomly select one element out of  $n$  elements.
```

```
  until 'A' is found
```

```
end
```

**Assignment:** Prove that the expected number of iterations is  $\lim_{n \rightarrow \infty} \sum_{i=1}^n \frac{i}{2^i} \leq 2$ .

# A randomized algorithm (from Wikipedia)

```
findingA_MC(array  $L$ ,  $n$ ,  $k$ )  
begin  
   $i \leftarrow 0$   
  repeat  
    Randomly select one element out of  $n$  elements.  
     $i \leftarrow i + 1$   
  until  $i = k$  or 'A' is found  
end
```

After  $k$  iterations,  $\Pr[\text{find } A] = 1 - (1/2)^k$ .

# Birthday problem (paradox)

- There are  $n$  randomly chosen people in a room.
- How **large** should  $n$  be such that there is at least one pair of them having the same birthday (mm/dd)?
- By the pigeonhole principle,  $n = 367$ ? or 366?
- Let us consider this problem in the other way around.  
How **large** should  $n$  be such that there is at least one pair of them having the same birthday (mm/dd) with probability  $\geq 0.5$ ?

# Birthday problem (paradox)

- $n$  people:  $x_1, x_2, \dots, x_n$
- Event  $i$ : some pair of  $x_1, x_2, \dots, x_i$  have the same birthday.
- $\Pr[\text{Event2}] = 1 - \frac{364}{365}$
- $\Pr[\text{Event3}] = 1 - \frac{364}{365} \cdot \frac{363}{365}$
- ...
- $\Pr[\text{Event23}] = 1 - \frac{364}{365} \cdot \frac{363}{365} \cdot \dots \cdot \frac{343}{365} \approx 0.507297.$
- 23 is much less than 366 or 367.