

# COMPx270: Randomised and Advanced Algorithms

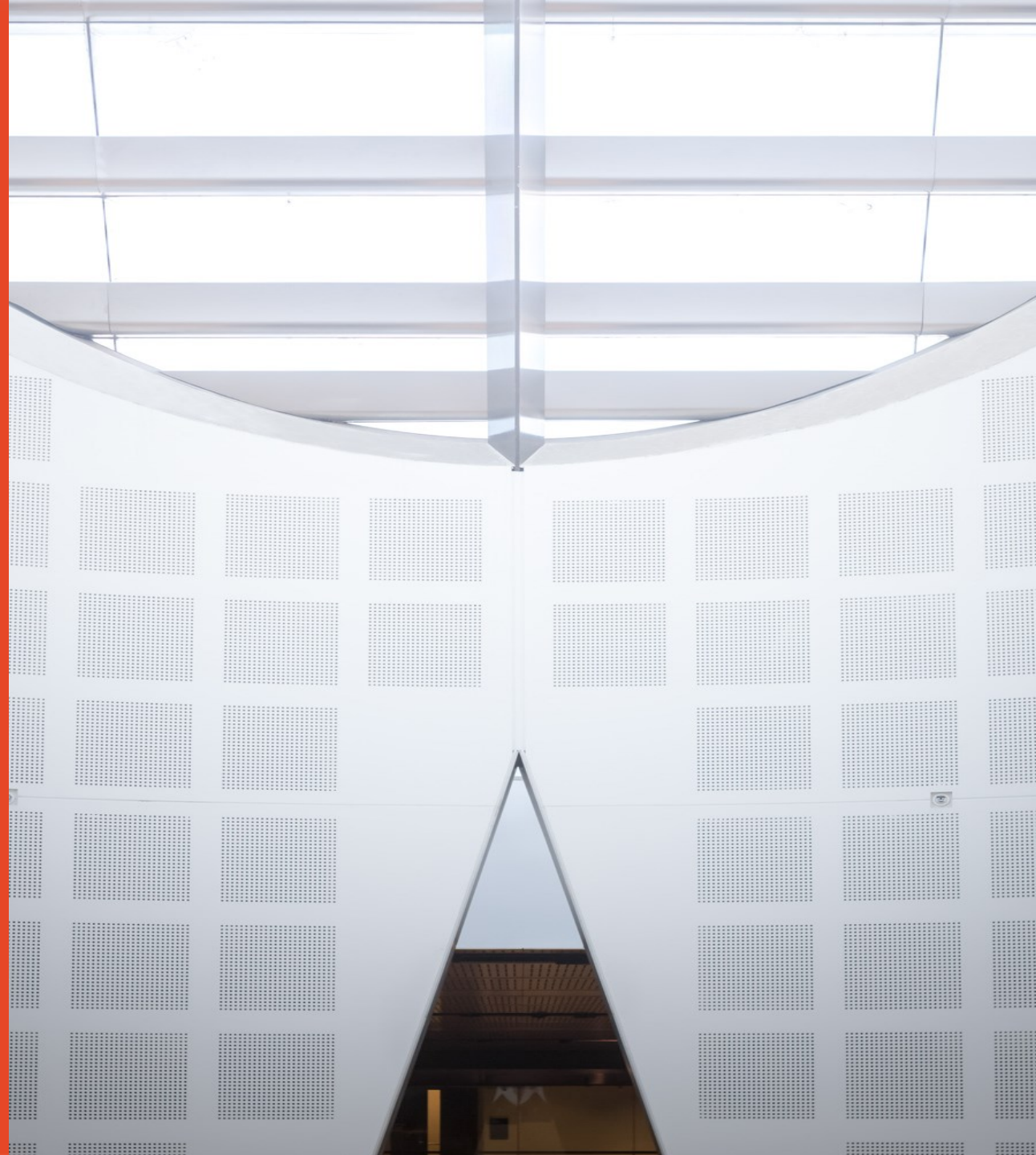
## Lecture 1: Randomness, Probability, and Algorithms 🎲

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## An experiment

Shuffle a deck of cards: then go through them in order. How many times do two consecutive cards have the same suit?

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4♥, 3♥, 8♣, 2♣, 3♠, 10♥, 8♦, 7♠, K♥, 5♦, 8♥, J♥, 9♣, 5♣, J♠, 2♥,  
Q♠, 2♠, 10♠, 6♠, 6♣, 5♥, 4♣, 9♠, Q♦, 8♠, 6♦, 10♦, 7♣, J♣, K♣, 4♦,  
K♦, K♠, A♦, A♠, A♣, 4♠, A♥, 3♣, 9♦, 3♦, J♦, 9♥, Q♥, Q♣, 2♦, 10♣,  
5♠, 7♦, 6♥, 7♥

## An experiment

Shuffle a deck of cards: then go through them in order. How many times do two consecutive cards have the same suit in expectation?

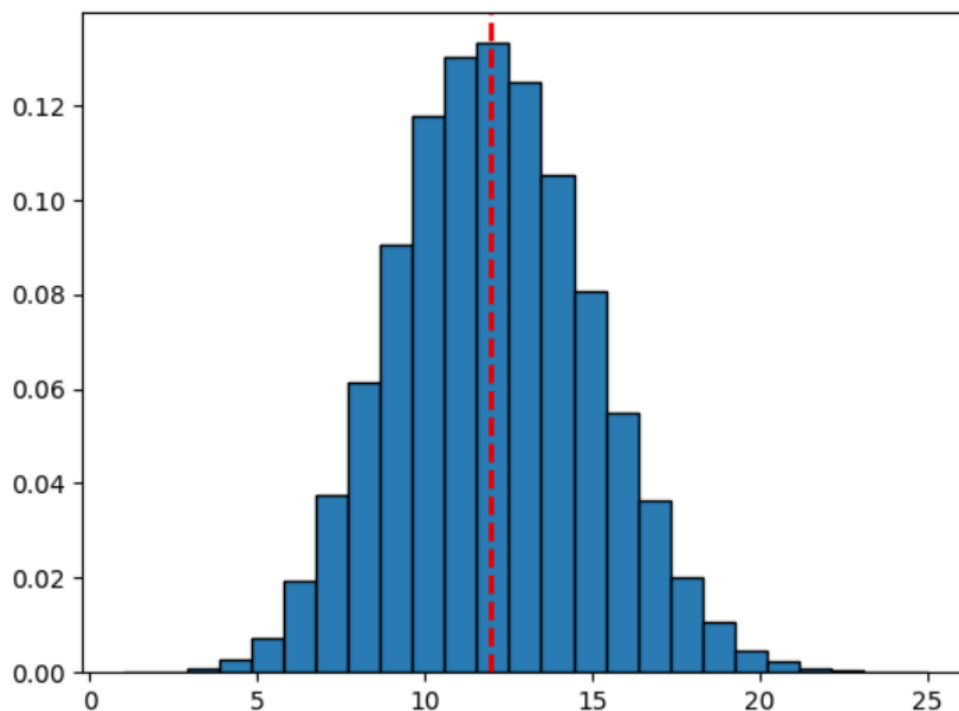
## An experiment 🎲

Shuffle a deck of cards: then go through them in order. How many times do two consecutive cards have the same suit in expectation?

```
1 import numpy as np
2 import random
3 deck = 13*['S', 'H', 'D', 'C']
4 consecutives = []
5 for _ in range(50000):
6     shuffled_deck = random.sample(deck, len(deck));
7     consecutives += [np.sum([shuffled_deck[i] == shuffled_deck[i+1] for i
8                             in range(len(deck)-1)])]
9 print("Empirical mean: %f" % np.mean(consecutives))
```

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*Can we prove it?*



## An experiment 🃏

Shuffle a deck of cards: then go through them in order. How many times do two consecutive cards have the same suit in expectation?

**Theorem (Linearity of expectation).**


$$\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$$

No assumption of independence, or anything. Surprisingly useful!

## An experiment

Shuffle a deck of cards: then go through them in order. How many times do two consecutive cards have the same suit in expectation?

# Randomised algorithms



**Standard algorithms:** “recipes.” Input = ingredients, output =  .

Given input, follow steps, get  .

Given same ingredients, get same  .

# Randomised algorithms

**Standard algorithms:** “recipes.” Input = ingredients, output = cake .

- Given input, follow steps, get .
- Given same ingredients, get same .

**Randomised algorithms:** “recipes with randomness” Input = ingredients, output = cake , randomness = unpredictable oven

- Given input, follow steps, get .
- Given same ingredients, get .

# Randomised algorithms

Randomized algorithms are algorithms where the behaviour doesn't depend solely on the input. It also depends (in part) on random choices or the values of a number of random bits. 🎲

Important distinctions: what is (and isn't) a randomized algo

- the input is assumed to be “random” ❌
- we average the time complexity over many calls to the algo ❌
- the input is worst-case, but the algo makes random choices ✅

# Randomised algorithms

(cartoon definition)

# Randomised algorithms, Monte Carlo, Las Vegas

(details)

## Why randomisation? 🎲

- Avoid pathological corner cases
- Get approximate result very fast
- Avoid predictable outcomes
- **Get faster, simpler algorithms**
- Break ties or bypass “impossibility results”
- Cryptography! Privacy!



## Why not randomisation? 🎲

- Randomness is not always good or desirable
- Random bits don't grow on trees!
- Bad random bits? Bad outputs.

# secrets — Generate secure random numbers for managing secrets ¶

*Added in version 3.6.*

**Source code:** [Lib/secrets.py](#)

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The [secrets](#) module is used for generating cryptographically strong random numbers suitable for managing data such as passwords, account authentication, security tokens, and related secrets.

In particular, [secrets](#) should be used in preference to the default pseudo-random number generator in the [random](#) module, which is designed for modelling and simulation, not security or cryptography.

## **Get faster, simpler algorithms? (An example)**

Given an  $n$ -bit integer, decide whether it is a prime number.

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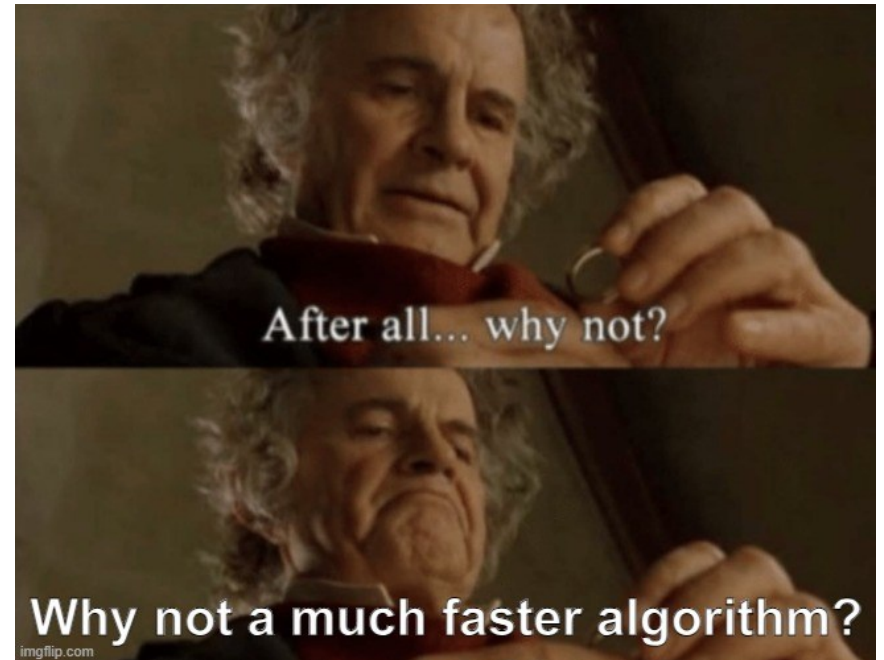
*The algorithm was the first one which is able to determine in polynomial time, whether a given number is prime or composite and this without relying on mathematical conjectures such as the generalized Riemann hypothesis. [...] In 2006 the authors received both the Gödel Prize and Fulkerson Prize for their work. (Wikipedia)*

## Get faster, simpler algorithms? (An example)

Given an  $n$ -bit integer, decide whether it is a prime number.

There exists a randomised algorithm! Since 1980 (Miller-Rabin). 🎲

Runs in time  $\tilde{O}(n^2)$ .



## Now, QuickSort

Given an array  $A$  of  $n$  distinct numbers, sort  $A$ .

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**Theorem.** There are deterministic sorting algorithms with running time  $O(n \log n)$ .

**Theorem.** Every (comparison-based) sorting algorithm must have worst-case running time  $\Omega(n \log n)$ .



## Now, QuickSort

---

**Require:** Input array  $A$  of size  $n$

- 1: **if**  $n \leq 1$  **then return**  $A$
  - 2: Select an index  $1 \leq i \leq n$ , and let  $p \leftarrow A[i]$  be the *pivot*
  - 3: Partition  $A$  into 3 subarrays:  $A_1$  (elements smaller than  $p$ ),  $A_2$  (equal to  $p$ ), and  $A_3$  (greater than  $p$ )  $\triangleright O(n)$  time
  - 4: Recursively call QuickSort on  $A_1$  and  $A_3$  to sort them
  - 5: Merge the (sorted)  $A_1, A_2, A_3$  into  $A$   $\triangleright O(n)$  time
  - 6: **return**  $A$
-

## Now, QuickSort

Given an array  $A$  of  $n$  distinct numbers, sort  $A$ .

**Theorem.** QuickSort is a deterministic sorting algorithm with running time  $O(n^2)$ .

(But it is simple, and nice, and does well in practice.)

# Randomised QuickSort

---

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## Now, QuickSort

Given an array  $A$  of  $n$  distinct numbers, sort  $A$ .

**Theorem.** Randomised QuickSort is a sorting algorithm with *expected* running time  $O(n \log n)$ .

(And it is simple, and still nice, and still does well in practice.)

# Now, QuickSort

(proof)

(proof)

# Recap, and looking forward

- Randomised, linearity of expectation, applications
- Concentration bounds, probability amplification, median trick
- Coupon Collector, Load Balancing, Power of Two Choices
- Derandomisation: Max-Cut, Method of Conditional Expectations
- Randomized Min-Cut (Karger's algorithm)
- Probabilistic data structures I: Hashing and Bloom filters
- Probabilistic data structures II: Johnson-Lindenstrauss, LSH
- Streaming and Sketching I: definitions, examples, frequency estimation
- Streaming and Sketching II: CountSketch, Count-min Sketch
- Linear Programming and Randomised Rounding
- Embeddings: FRT algorithm, and applications
- Sampling and Counting

**To conclude: something completely different!**

If  $X$  is a non-negative integer-valued random variable, then

$$\mathbb{E}[X] = \sum_{n=0}^{\infty} n \Pr[X = n] = \sum_{n=1}^{\infty} \Pr[X \geq n]$$

(This is useful!) See tutorial.