



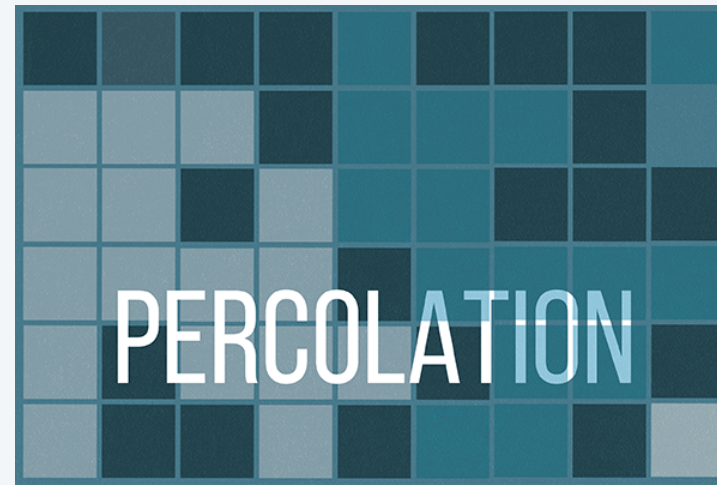
<https://algs4.cs.princeton.edu>

RANDOMNESS

- ▶ *what it is and what it isn't*
- ▶ *Las Vegas and Monte Carlo*
- ▶ *approximate counting*
- ▶ *context*

A brief recap: where we've already encountered randomness

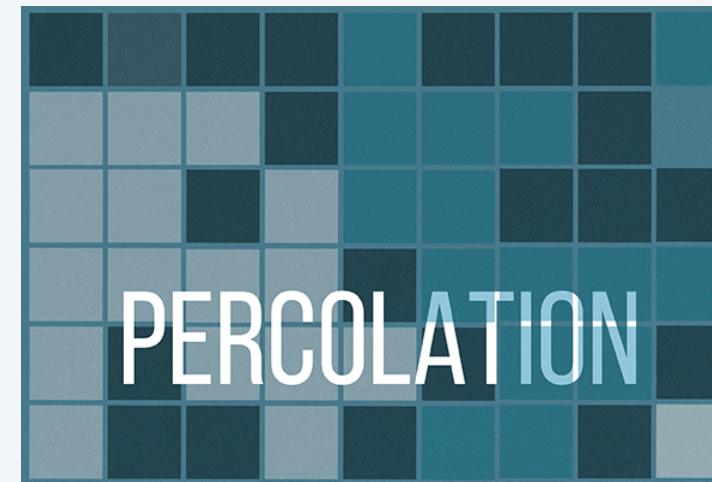
Percolation. Monte Carlo simulation: open **random** blocked sites.



Randomized queues. Remove item chosen uniformly at **random**.



A brief recap: where we've already encountered randomness



Test 2: open **random** sites until the system percolates

Test 7: open **random** sites with large n

Test 12: call `open()`, `isOpen()`, and `numberOfOpenSites()`

in **random** order until just before system percolates

Test 13: call `open()` and `percolates()` in **random** order until just before system percolates

Test 14: call `open()` and `isFull()` in **random** order until just before system percolates

Test 15: call all methods in **random** order until just before system percolates

Test 16: call all methods in **random** order until almost all sites are open

(with inputs not prone to backwash)

Test 20: call all methods in **random** order until all sites are open

(these inputs are prone to backwash)

A brief recap: where we've already encountered randomness



Tests 1-8 make **random** intermixed calls to `addFirst()`, `addLast()`, `removeFirst()`, `removeLast()`, `isEmpty()`, and `size()`, and `iterator()`.

Test 12: check `iterator()` after **random** calls to `addFirst()`, `addLast()`, `removeFirst()`, and `removeLast()` with probabilities (p_1 , p_2 , p_3 , p_4)

Tests 1-6 make **random** intermixed calls to `enqueue()`, `dequeue()`, `sample()`, `isEmpty()`, `size()`, and `iterator()`.

Test 16: **check randomness** of `sample()` by enqueueing n items, repeatedly calling `sample()`, and counting the frequency of each item

Test 17: **check randomness** of `dequeue()` by enqueueing n items, dequeueing n items, and seeing whether each of the $n!$ permutations is equally likely

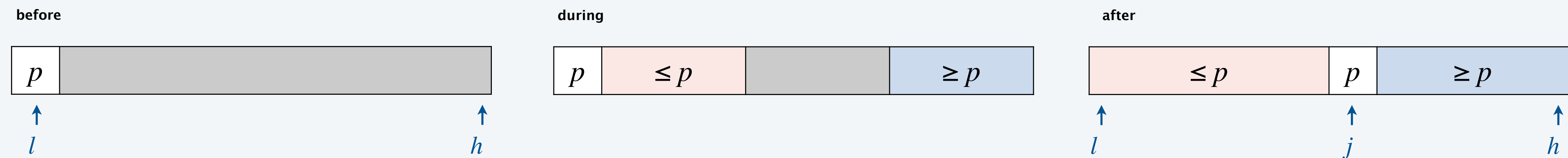
Test 18: **check randomness** of `iterator()` by enqueueing n items, iterating over those n items, and seeing whether each of the $n!$ permutations is equally likely

A brief recap: where we've already encountered randomness

Quicksort is a (Las Vegas) **randomized algorithm**.

Shuffling is needed for performance guarantee.

Equivalent alternative: pick a random pivot in each subarray.



Hash tables.





<https://algs4.cs.princeton.edu>

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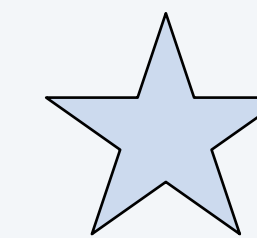
Which of these outcomes is most likely to occur in a sequence of 6 coin flips?



D. All of the above.

E. Both B and C.

The uniform distribution

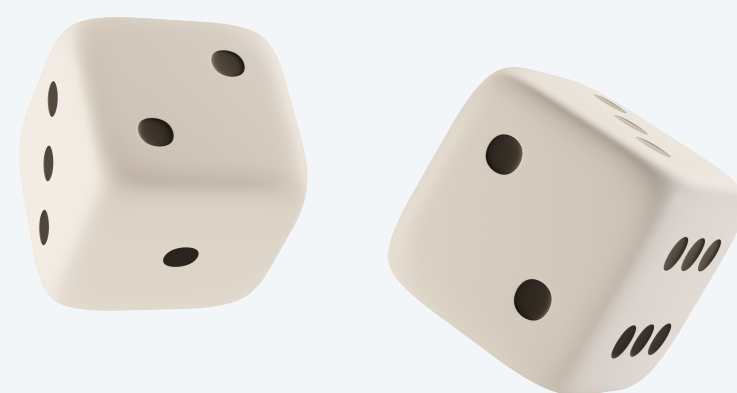


Coin flip.



$$\mathbb{P}[C \text{ lands heads}] = \mathbb{P}[C \text{ lands tails}] = \frac{1}{2}.$$

Roll of a die.



$$\mathbb{P}[D = 1] = \mathbb{P}[D = 2] = \dots = \mathbb{P}[D = 6] = \frac{1}{6}.$$



$$\mathbb{P}[D = 1] = \dots = \mathbb{P}[D = 20] = \frac{1}{20}.$$

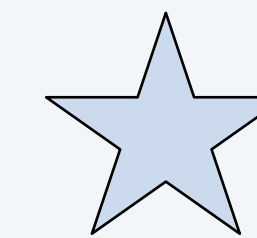
Notation.

C and D are **random variables**.

“ C lands heads,” “ $D = 4$ ” or “ D is even” are **events** with probabilities $\mathbb{P}[C \text{ lands heads}]$, etc.

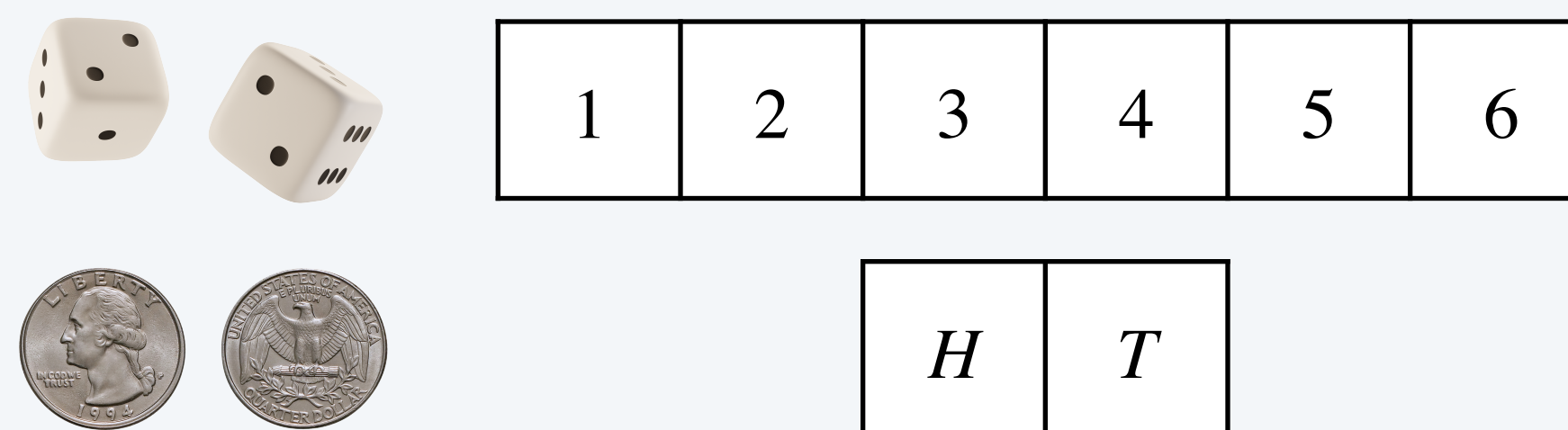
A **distribution** consists of all outcome-probability pairs.

[uniform distribution: all probabilities equal]



Generating uniform distributions.

- Over (small) domain of size n :
place outcomes in array, return random element.



- Over large domains:
 - Bit strings of length n : [size 2^n]
flip n coins, output sequence of outcomes ($H = 0, T = 1$).
 - Permutations of n items: [size $n!$]
sample n elements from $\{ 1, 2, \dots, n \}$ without replacement.
 - Spanning trees of n -vertex graph? [size $\leq n^{n-2}$]



Flip a coin 6 times and count how often it lands heads. Which count is most likely?

- A. 2
- B. 3
- C. 4
- D. All of the above.
- E. None of the above.

Pseudorandomness

Computers can't generate randomness (without specialized hardware).



Pseudorandom functions.



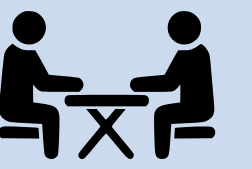
`random` — Generate pseudo-random numbers

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

This module implements pseudo-random number generators for various distributions.

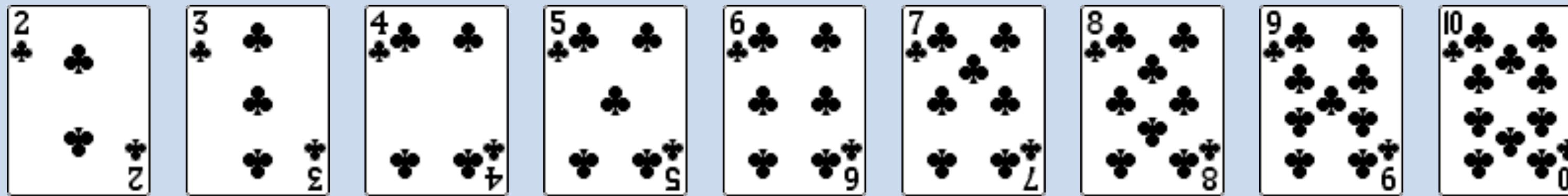
For integers, there is uniform selection from a range. For sequences, there is uniform selection of a random element, a function to generate a random permutation of a list in-place, and a function for random sampling without replacement.

Interview question: shuffle an array

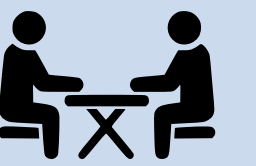


Goal. Rearrange array so that result is a uniformly random permutation.

*all $n!$ permutations
equally likely*

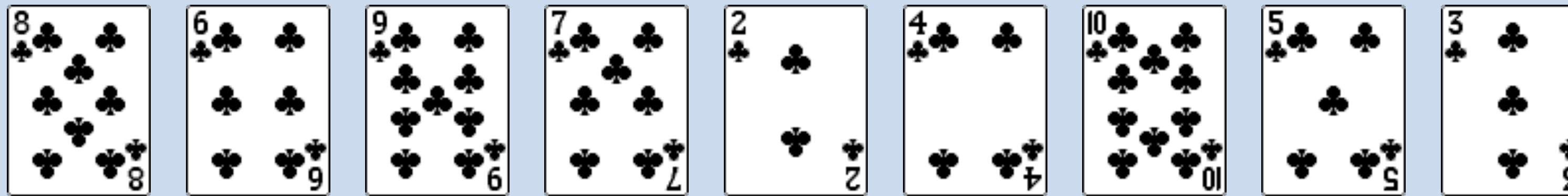


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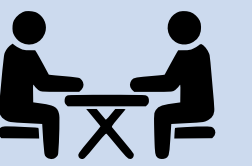


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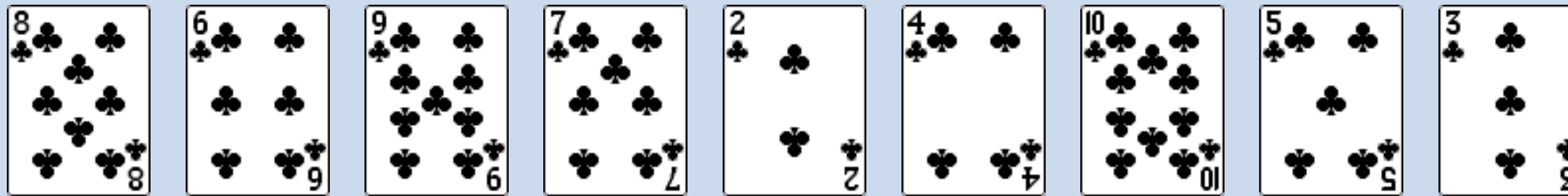


Interview question: shuffle an array



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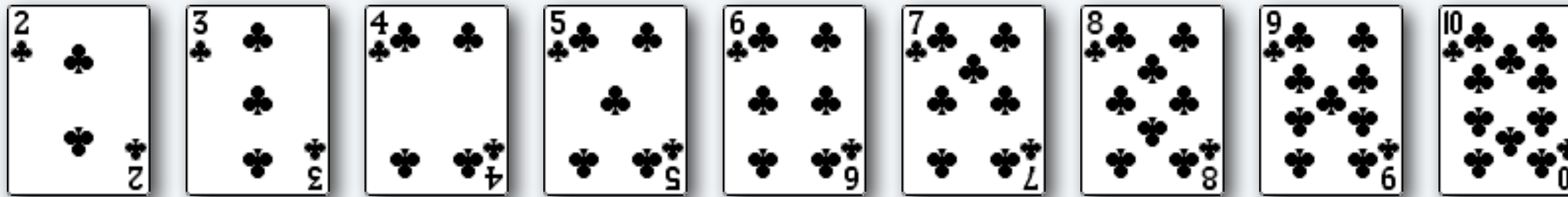
*all $n!$ permutations
equally likely*



Challenge. Design a linear-time algorithm.

Knuth shuffle

- In iteration i , pick integer r between 0 and i uniformly at random.
- Swap $a[i]$ and $a[r]$.



Proposition. [Fisher–Yates 1938] Knuth shuffling algorithm produces a uniformly random permutation of the input array in linear time.

*← assuming integers
uniformly at random*

Knuth shuffle

- In iteration i , pick integer r between 0 and i uniformly at random.
- Swap $a[i]$ and $a[r]$.

common bug: between 0 and $n - 1$
correct variant: between i and $n - 1$

```
public class Knuth {
    public static void shuffle(Object[] a) {
        int n = a.length;
        for (int i = 0; i < n; i++) {
            int r = StdRandom.uniform(i + 1);
            exch(a, i, r);
        }
    }
}
```

between 0 and i

<http://algs4.cs.princeton.edu/11model/Knuth.java.html>

Broken Knuth shuffle

Q. What happens if integer is chosen between 0 and $n - 1$?

A. Not uniformly random!

*instead of between
0 and i*

permutation	Knuth shuffle	broken shuffle
A B C	1 / 6	4 / 27
A C B	1 / 6	5 / 27
B A C	1 / 6	5 / 27
B C A	1 / 6	5 / 27
C A B	1 / 6	4 / 27
C B A	1 / 6	4 / 27

*$3^3 = 27$ possible outcomes
(but 27 is not a multiple of 6)*

probability of each permutation when shuffling { A, B, C }

Industry story (online poker)

Texas hold'em poker. Software must shuffle electronic cards.



How We Learned to Cheat at Online Poker: A Study in Software Security

<https://www.developer.com/tech/article.php/616221/How-We-Learned-to-Cheat-at-Online-Poker-A-Study-in-Software-Security.htm>

Industry story (online poker)

```
for i := 1 to 52 do begin
  r := random(51) + 1; ← between 1 and 51
  swap := card[r];
  card[r] := card[i];
  card[i] := swap;
end;
```

Shuffling algorithm in FAQ at www.planetpoker.com

Bug 1. Random number r is never 52 \Rightarrow 52nd card can't end up in 52nd place.

Bug 2. Shuffle not uniform (should be between 1 and i).

Bug 3. random() uses 32-bit seed \Rightarrow 2^{32} possible shuffles.

Bug 4. Seed = milliseconds since midnight \Rightarrow 86.4 million shuffles.

“ The generation of random numbers is too important to be left to chance. ”

— Robert R. Coveyou

Industry story (online poker)

Best practices for shuffling (if your business depends on it).

- Use a hardware random-number generator that has passed both the FIPS 140-2 and the NIST statistical test suites.
- Continuously monitor statistical properties: hardware random-number generators are fragile and fail silently.
- Use an unbiased shuffling algorithm.



RANDOM.ORG

Bottom line. Shuffling a deck of cards is hard!

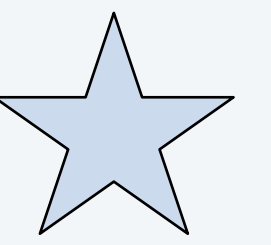


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RANDOMNESS

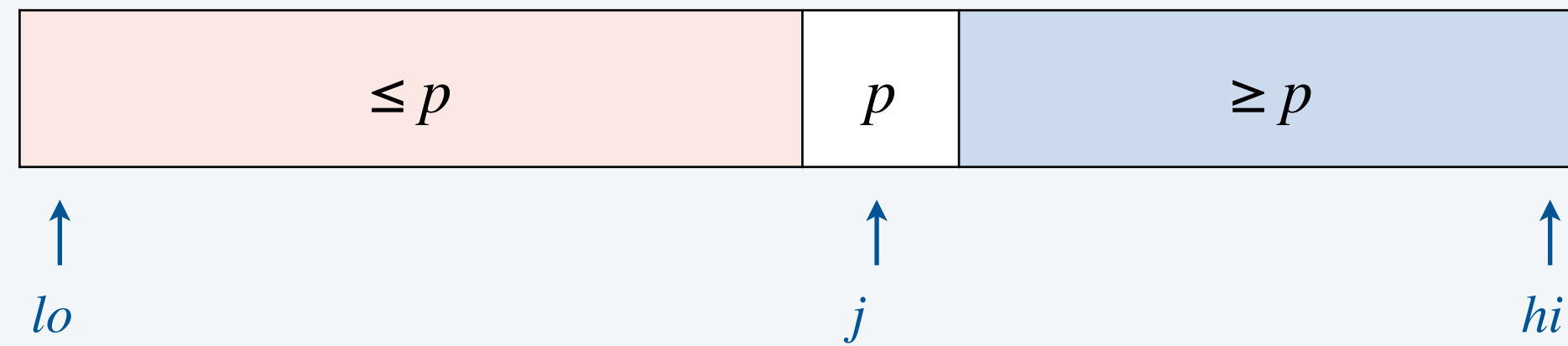
- ▶ *what it is and what it isn't*
- ▶ ***Las Vegas and Monte Carlo***
- ▶ *approximate counting*
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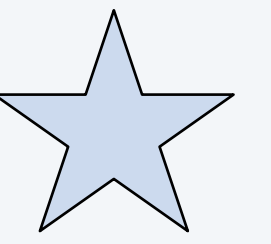
Las Vegas algorithms



- Guaranteed to be correct.
- Running time depends on outcomes of random coin flips.

Ex. Quicksort, quickselect.





Monte Carlo algorithm.

- Not guaranteed to be correct.
- Running time is deterministic.
[doesn't depend on coin flips]

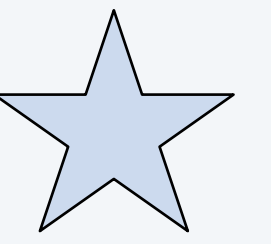


Amplification. If $\mathbb{P}[A \text{ is correct}] = 1\%$, repeat 500 times.

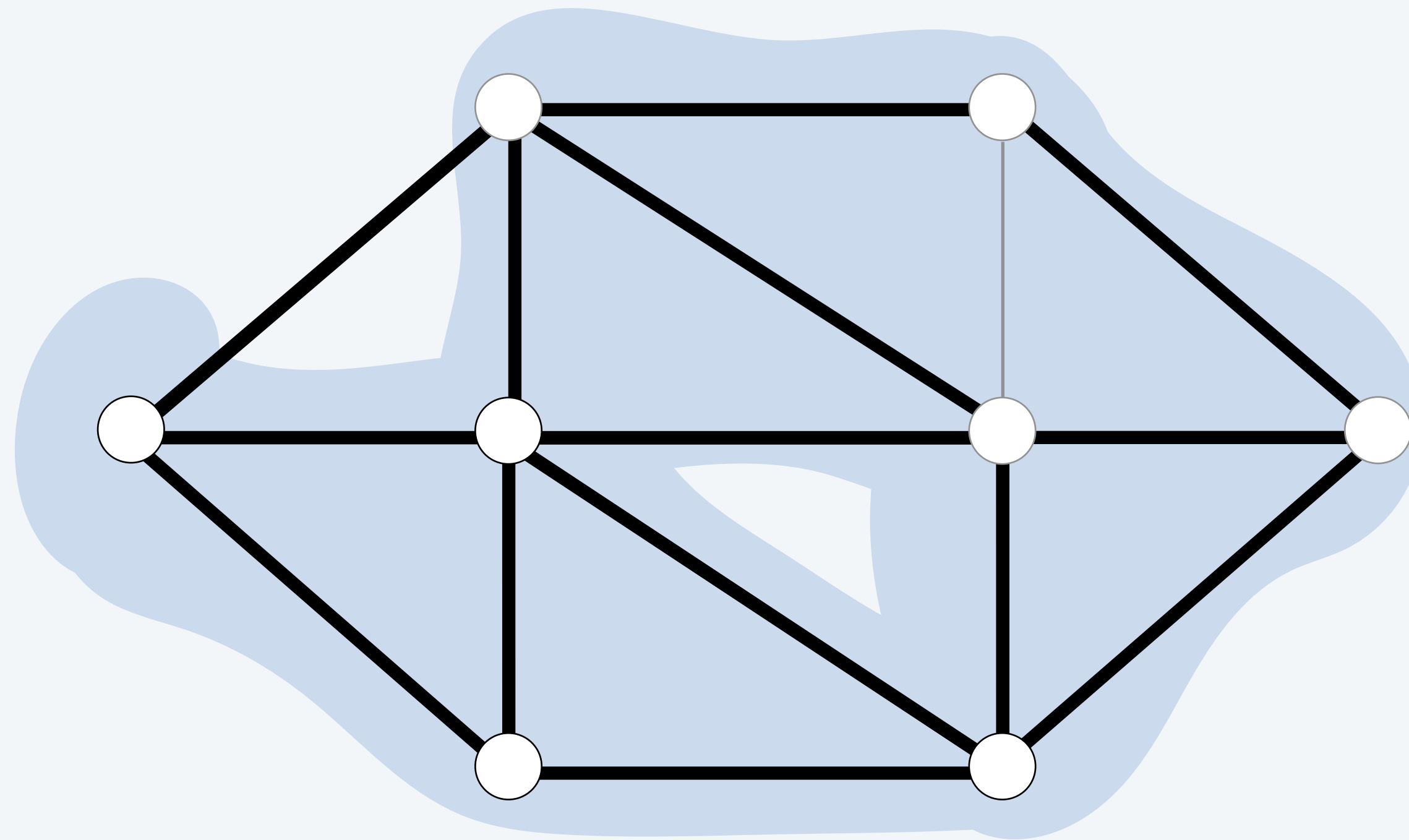
Then, $\mathbb{P}[A_1, A_2, \dots, A_{500} \text{ are all incorrect}] \leq \left(\frac{99}{100}\right)^{500} < 1\%$

↑
independence

Global mincut problem



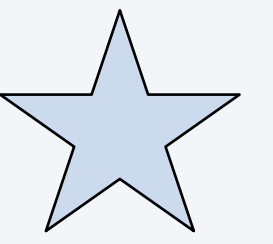
Goal. Find cut in undirected graph with fewest edges (for any source and sink).



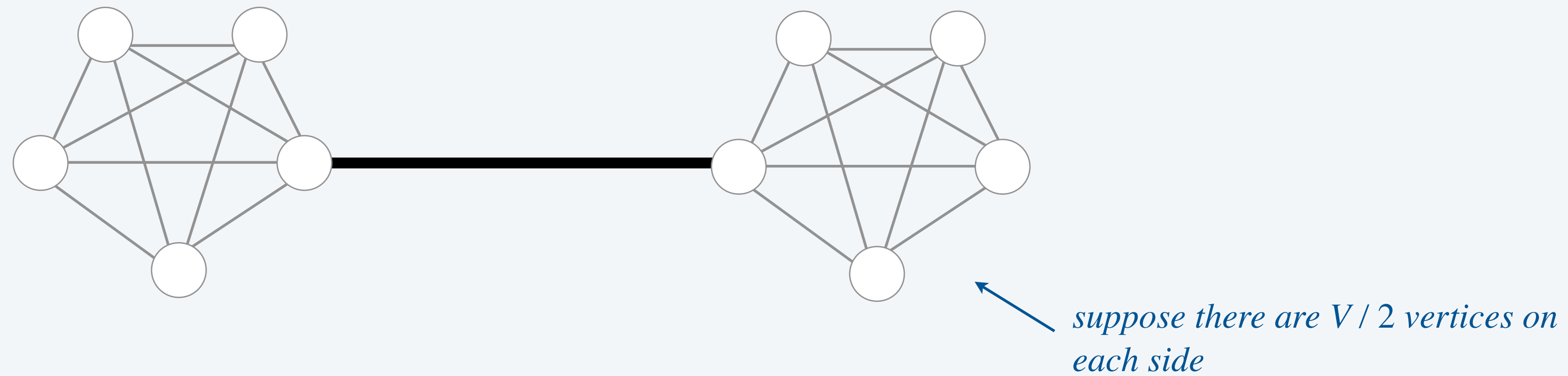
Idea. Pick a random cut.

Uniformly? Since there are $2^V - 1$ cuts, may succeed with only $\sim \frac{1}{2^V}$ probability.

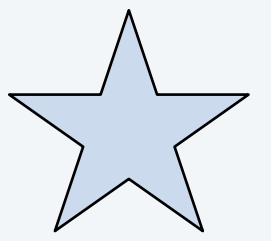
Global mincut problem



Example. Bad graph for the “pick a uniformly random cut” algorithm.



Problem. There is only 1 mincut, but $2^V - 1$ total cuts, we need to be lucky to find it.



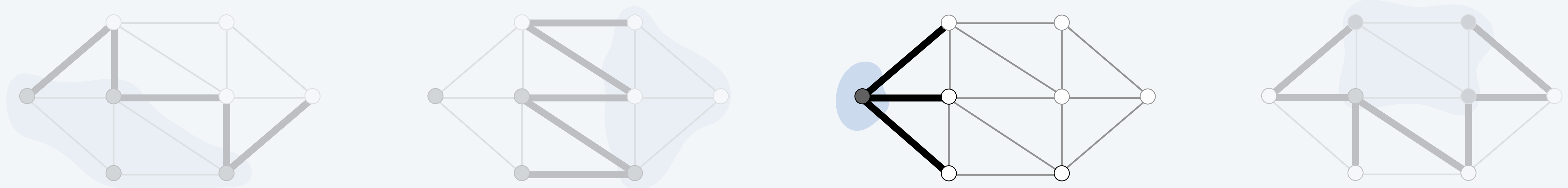
Karger's global mincut algorithm

Algorithm.

- Assign a random weight (uniform between 0 and 1) to each edge e .
- Run Kruskal's MST algorithm until 2 connected components left.
- 2 connected components defines the cut.

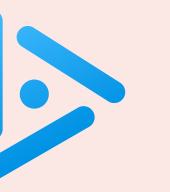
Probability of finding a mincut: $\geq \frac{1}{V^2}$. [no mincut edges in each connected component]

Run algorithm many times and return best cut.



Remark 1. Finds global mincut in $\Theta(EV^2 \log E)$ time — better than $\Theta(V)$ runs of Ford-Fulkerson!

Remark 2. With clever idea, improved to $\Theta(V^2 \log^3 V)$ time (still randomized).



Smallest # of repetitions of Karger's algorithm to get correct answer with 99% probability?

- A. $\Theta(1)$
- B. $\Theta(V)$
- C. $\Theta(V^2)$
- D. $\Theta(V^3)$
- E. None of the above.

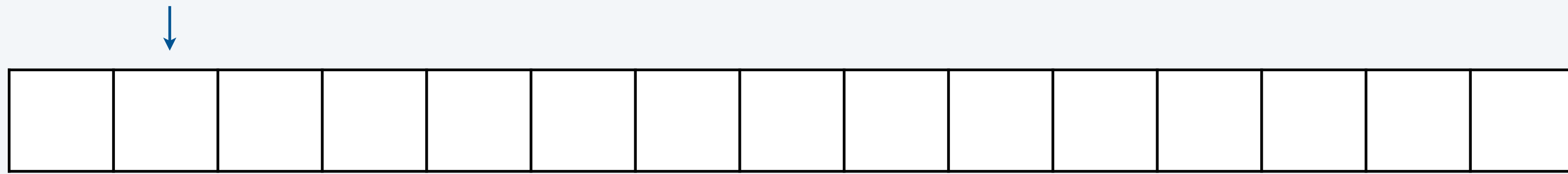


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RANDOMNESS

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Packet counting



15



Fix $n \in \mathbb{N}$. How many bits must a counter have to count from 0 to $n - 1$?

- A. $\log_2 n$
- B. $\lfloor \log_2 n \rfloor$ ← *round down*
- C. $\lceil \log_2 n \rceil$ ← *round up*
- D. $\lfloor \log_2 n \rfloor + 1$
- E. n

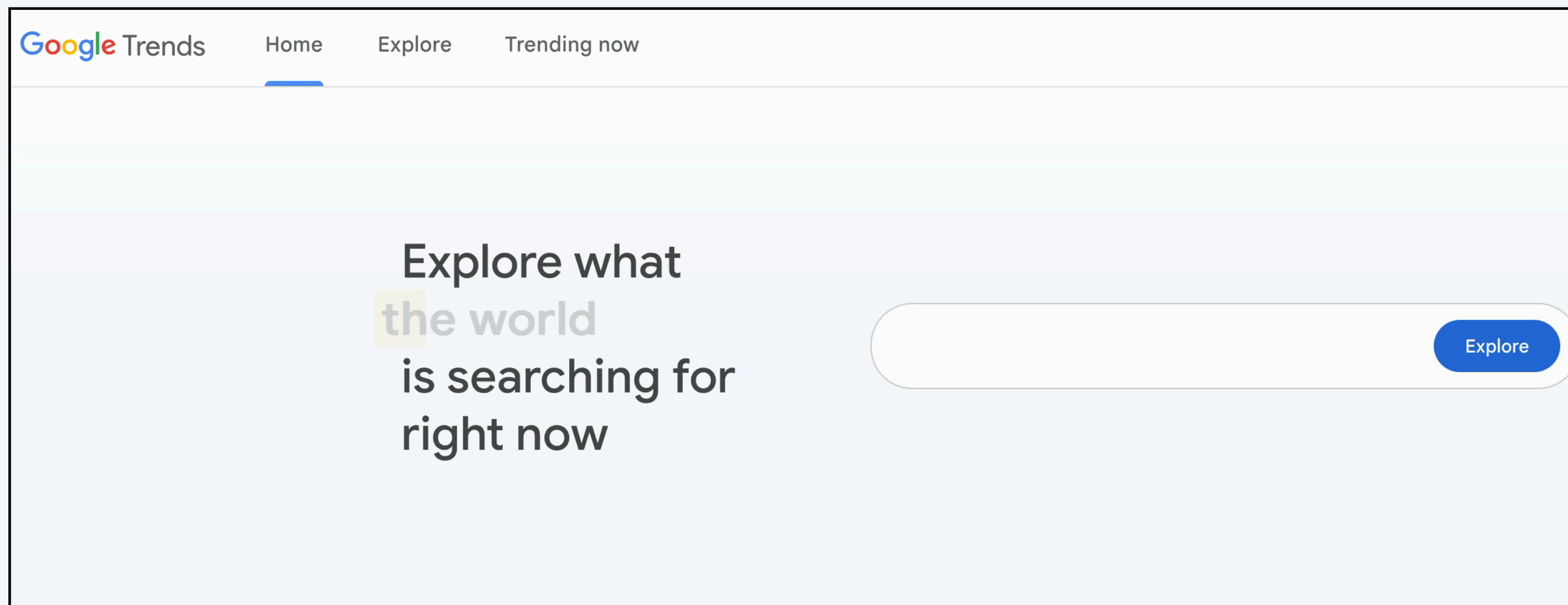
Approximate counting

Goal. Count with less memory: from $\sim \log_2 n$ to $\Theta(\log \log n)$.

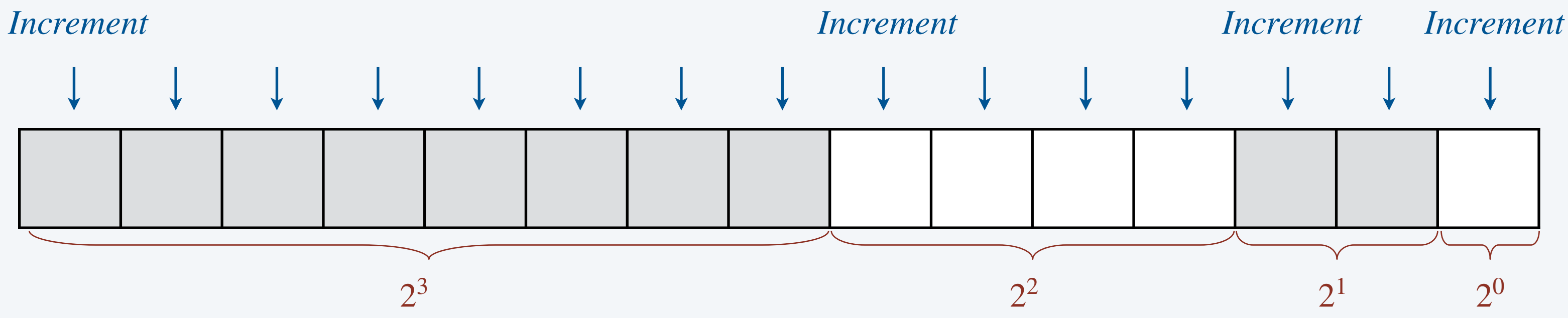
Why bother?

Database with 1 billion entries: $\log_2 10^9 \approx 30$ bits, but $\log_2 \log_2 10^9 \approx 5$ bits.

Factor-6 improvement matters a lot.



Approximate counting

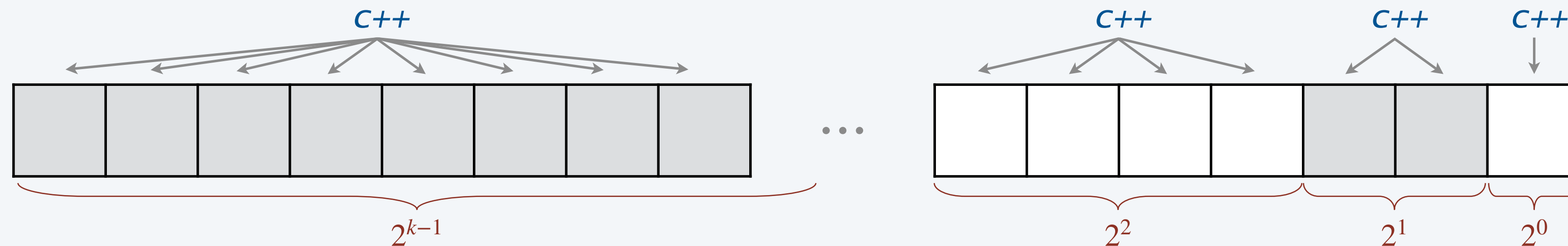


4

Approximate counting

```
public class ApproximateCounter() {  
    private byte c;  
  
    public void increment() {  
        if (StdRandom.uniformInt(1 << c) == 0)  
            c++;  
    }  
  
    public int count() {  
        return (1 << c) - 1;  
    }  
}
```

Returns 2^c



Value of counter around k after $n = 2^0 + 2^1 + \dots + 2^{k-1} = 2^k - 1$ packets.

Memory requirement: $\sim \log_2 k \sim \log_2 \log_2 n$.

Approximate counting: probabilistic analysis

Proposition. The value c_n of the counter after n packets satisfies $\mathbb{E}[2^{c_n}] = n + 1$.

Pf. [by induction on n]

Base case: initially, $\mathbb{E}[2^{c_0}] = 2^0 = 0 + 1$.

Define $P_{n,k} = \mathbb{P}[c_n = k]$. They satisfy the recurrence $P_{0,0} = P_{0,1} = 1$ and

$$P_{n+1,k} = \frac{1}{2^{k-1}} \times P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times P_{n,k}$$

counter was $k-1$ *counter was k*

increased *didn't increase*

analysis beyond
scope of this course

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counter was $k-1$ *counter was k*
↓ ↓
↑ ↑
increased *didn't increase*

Decompose $\mathbb{E}[2^{c_{n+1}}]$ and rearrange:

$$\mathbb{E}[2^{c_{n+1}}] = \sum_{k=0}^{n+1} 2^k \times P_{n+1,k}$$

analysis beyond
scope of this course

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$$P_{n+1,k} = \underbrace{\frac{1}{2^{k-1}}}_{\text{increased}} \times \underbrace{P_{n,k-1}}_{\text{counter was } k-1} + \left(1 - \frac{1}{2^k}\right) \times \underbrace{P_{n,k}}_{\text{counter was } k} \quad \underbrace{\left(1 - \frac{1}{2^k}\right)}_{\text{didn't increase}}$$

Decompose $\mathbb{E}[2^{c_{n+1}}]$ and rearrange:

$$\mathbb{E}[2^{c_{n+1}}] = \sum_{k=0}^{n+1} 2^k \times \left(\frac{1}{2^{k-1}} \times P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times P_{n,k} \right)$$

analysis beyond
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↑ ↓
increased *didn't increase*

Decompose $\mathbb{E}[2^{c_{n+1}}]$ and rearrange:

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$$= 2 \sum_{k=0}^{n+1} P_{n,k-1}$$

analysis beyond
scope of this course

Approximate counting: probabilistic analysis

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increased *didn't increase*

counter was k-1 *counter was k*

Decompose $\mathbb{E}[2^{c_{n+1}}]$ and rearrange:

$$\begin{aligned} \mathbb{E}[2^{c_{n+1}}] &= \sum_{k=0}^{n+1} 2^k \times \left(\frac{1}{2^{k-1}} \times P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times P_{n,k} \right) \\ &= 2 \sum_{k=0}^{n+1} P_{n,k-1} + \sum_{k=0}^{n+1} 2^k \times P_{n,k} \end{aligned}$$

analysis beyond
scope of this course

Approximate counting: probabilistic analysis

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counter was $k-1$ *counter was k*
↓ ↓
↑ ↑
increased *didn't increase*

Decompose $\mathbb{E}[2^{c_{n+1}}]$ and rearrange:

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analysis beyond
scope of this course

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$$P_{n+1,k} = \underbrace{\frac{1}{2^{k-1}}}_{\text{increased}} \times \overset{\text{counter was } k-1}{\downarrow} P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times \overset{\text{counter was } k}{\downarrow} P_{n,k}$$

\uparrow didn't increase

Decompose $\mathbb{E} [2^{c_{n+1}}]$ and rearrange:

$$\begin{aligned}
 \mathbb{E} [2^{c_{n+1}}] &= \sum_{k=0}^{n+1} 2^k \times \left(\frac{1}{2^{k-1}} \times P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times P_{n,k} \right) \\
 &= 2 \sum_{k=0}^{n+1} P_{n,k-1} + \sum_{k=0}^{n+1} 2^k \times P_{n,k} - \sum_{k=0}^{n+1} P_{n,k} \\
 &= \mathbf{2} + \mathbf{\mathbb{E} [2^{c_n}] - 1}
 \end{aligned}$$

analysis beyond
scope of this course

Approximate counting: probabilistic analysis

Proposition. The value c_n of the counter after n packets satisfies $\mathbb{E}[2^{c_n}] = n + 1$.

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$$P_{n+1,k} = \underbrace{\frac{1}{2^{k-1}}}_{\text{increased}} \times \overset{\text{counter was } k-1}{\downarrow} P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times \overset{\text{counter was } k}{\downarrow} P_{n,k}$$

↑ *didn't increase*

Decompose $\mathbb{E}[2^{c_{n+1}}]$ and rearrange:

$$\mathbb{E}[2^{c_{n+1}}] = \sum_{k=0}^{n+1} 2^k \times \left(\frac{1}{2^{k-1}} \times P_{n,k-1} + \left(1 - \frac{1}{2^k}\right) \times P_{n,k} \right)$$

$$= 2 \sum_{k=0}^{n+1} P_{n,k-1} + \sum_{k=0}^{n+1} 2^k \times P_{n,k} - \sum_{k=0}^{n+1} P_{n,k}$$

$$= 2 + \mathbb{E}[2^{c_n}] - 1$$

$$= (n + 1) + 1$$

inductive hypothesis →

analysis beyond scope of this course



<https://algs4.cs.princeton.edu>

RANDOMNESS

- *what it is and what it isn't*
- *Las Vegas and Monte Carlo*
- *approximate counting*
- *context*

Beyond this course

- Approximation algorithms [intractability: stay tuned!]
- Cryptography [average-case hardness]
- Complexity theory: $P \stackrel{?}{=} BPP$ [derandomization]
- Mathematics: the Probabilistic Method
E.g., graph with E edges has a cut with $E/2$ edges. [approximate maxcut]
To prove that there exists an object with property T :
 - sample a random object;
 - show that $\mathbb{P}[T \text{ is satisfied}] > 0$.
- Quantum computation



IBM Quantum System One

Credits

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<i>Random number generator</i>	<u>XKCD</u>	<u>CC BY-NC 2.5</u>

```
int getRandomNumber()
{
    return 4;    // chosen by fair dice roll.
                // guaranteed to be random.
}
```

<https://xkcd.com/221/>