## **Randomized Algorithms**

- Algorithms that make random choices during the computation
- Often faster, simpler than traditional algorithms

### Miller-Rabin primality test

**Input:** *n*-bit number *x*.

**Goal**: decide whether *x* is a prime number or not.

- Extremely important problem: many applications in cryptography.
- There is a deterministic polynomial time algorithm (AKS-2000), running time is  $O(n^{12})$

### The test (running time $O(n^2)$ ):

- **1.** Express  $x 1 = 2^s \cdot d$ , where *d* is odd.
- **2.** Pick  $a \in \{1, 2, \dots, x 1\}$  uniformly at random.

**3.** If for some t = 1, 2, ..., s,  $a^{2^t \cdot d} = 1 \mod x$ , yet  $a^{2^{t-1} \cdot d} \neq -1 \mod x$ , conclude that *x* is not prime. Otherwise conclude that *x* is prime.

**Theorem:** If x is prime, the test concludes that x is prime with probability 1. If x is not prime, the test concludes not prime with probability at least 3/4.

### Min-Cut

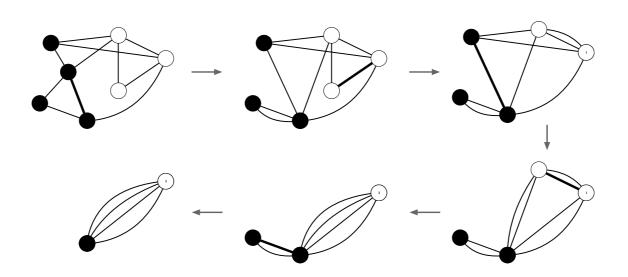
Input: An undirected graph.

**Goal**: Partition the vertices of the graph in two sets A, B, to minimize the number of edges going from A to B.

You can use flows and cuts, but there is a simpler randomized algorithm

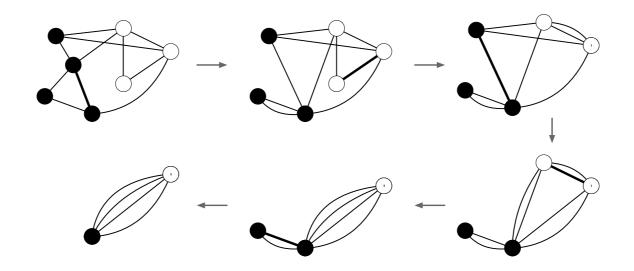
### Karger's Algorithm:

- 1. In each step, pick a uniformly random edge and contract it.
- 2. Stop when you have just two vertices.
- **3.** Output the corresponding cut.



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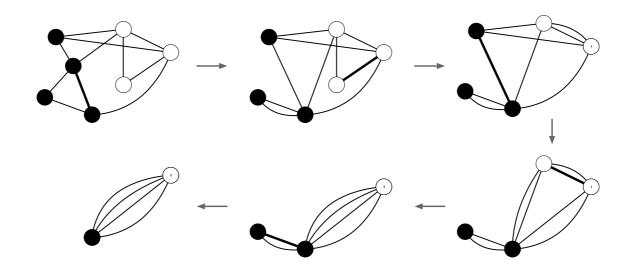


**Thm:** The algorithm finds the min-cut with probability at least 2/(n(n-1)). **Pf:** 

- Suppose the min-cut cuts k edges.
- Then every vertex must degree  $\geq k$ , or else that vertex would already give a smaller min-cut.
- So, the number of edges in the graph is at least nk/2.
- The probability we pick one of the edges of the min-cut is at most k/(nk/2) = 2/n.
- The probability that an edge of the min-cut is never picked is at least (1 2/n)(1 2/(n 1))...(1 2/3)=  $((n - 2)/n) \cdot ((n - 3)/(n - 1)) \cdot ((n - 4)/(n - 2))... = 2/(n(n - 1)).$

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**Final algorithm:** Repeat the above algorithm 100n(n - 1) times. Output the best cut that you find.

# Graph coloring

Input: An undirected graph.

**Goal**: Find a 3-coloring of vertices that maximizes the number of edges that get 2 colors.

#### Algorithm:

Randomly color the vertices of the graph red, blue, green.

**Thm:** The expected number of vertices that are properly colored is at least 2m/3. Pf: For each edge e, define  $X_e = 1$  if the edge e gets two colors, and  $X_e = 0$  otherwise.

 $\mathbb{E}[X_e] = \Pr[X_e = 1] \cdot 1 = 2/3.$ So, by linearity of expectation,  $\mathbb{E}[\sum_e X_e] = \sum_e \mathbb{E}[X_e] = 2m/3.$ 

No known poly time algorithm achieves > 2m/3.

# **Dominating set**

**Input:** An undirected graph, every vertex has degree  $\geq \Delta$ . **Goal**: Find a small set of vertices *S* such that every vertex is either in *S* or is a neighbor of *S*.

### **Algorithm:**

- 1. Randomly include each vertex in the set X, with probability p.
- 2. Let Y be the set vertices not in X and not a neighbor of X.
- 3. Output  $X \cup Y$ .

**Claim:** The expected size of  $X \cup Y$  is at most  $pn + n(1-p)^{1+\Delta} \le pn + e^{-p(1+\Delta)}n$ . Set  $p = \ln(1 + \Delta)/(1 + \Delta)$ , to get expected size at most  $n(1 + \ln(1 + \Delta))/(1 + \Delta)$ . **Pf of Claim:** 

- 1. The expected size of X is pn.
- 2. For each vertex, the probability that it is included in Y is at most  $(1 p)^{1+\Delta}$ .
- 3. So the expected size of *Y* is  $n(1-p)^{1+\Delta}$ .

# Matrix product checking in $O(n^2)$ time.

**Input:**  $n \times n$  matrices A, B, C**Goal:** Check that AB = C

### **Algorithm:**

1. Pick  $x \in \{0,1\}^n$  uniformly at random. 2. Check ABx = Cx

**Claim:** If  $AB \neq C$ , then  $Pr[ABx = Cx] \leq 1/2$ .

#### Pf of Claim:

Let D = (AB - C)Suppose  $D_{i,j} \neq 0$ , then  $(Dx)_i = \sum_k D_{i,k} x_k = D_{i,j} x_j + \sum_{k \neq j} D_{i,k} x_k$ , so for every fixing of  $\sum_{k \neq j} D_{i,k} x_k$ , the probability that  $(Dx)_i = 0$  is at most 1/2.