## 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  $O(n^{12})$

#### Miller-Rabin primality test (running time $O(n^2)$ ):

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

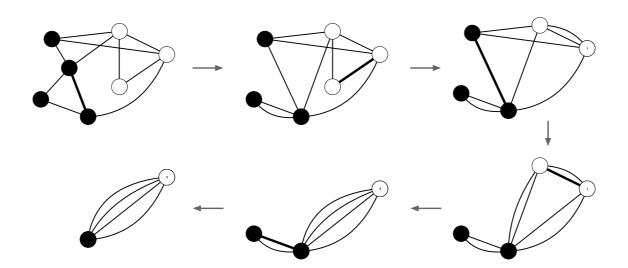
Input: An undirected graph.

**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

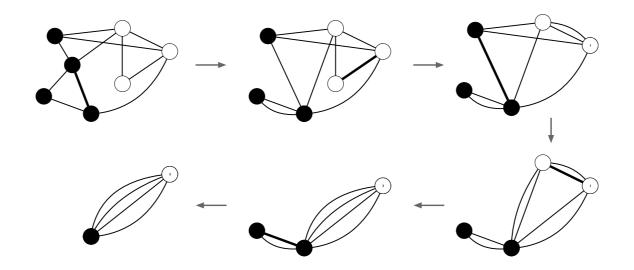
#### **Karger's Algorithm:**

- 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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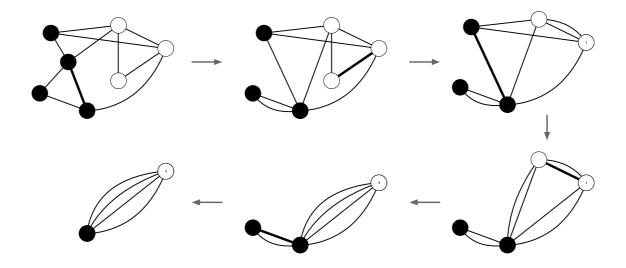


**Thm:** The algorithm finds the min-cut with probability at least 2/(n(n-1)). **Pf:** 

- 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  $(Dx)_i = \sum_k D_{i,k} x_k = D_{i,j} x_j + \sum_{k \neq i} D_{i,k} x_k$ , so for every fixing

of 
$$\sum_{k \neq j} D_{i,k} x_k$$
, the probability that  $(Dx)_i = 0$  is at most  $1/2$ .