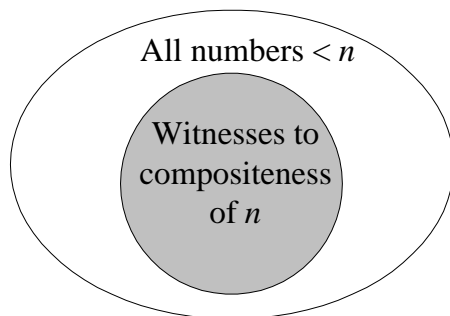


## The Density Attack for PRIMES



- It works, but does it work well?

## The Chinese Remainder Theorem

- Let  $n = n_1 n_2 \cdots n_k$ , where  $n_i$  are pairwise relatively prime.
- For any integers  $a_1, a_2, \dots, a_k$ , the set of simultaneous equations

$$x = a_1 \pmod{n_1}$$

$$x = a_2 \pmod{n_2}$$

$$\vdots$$

$$x = a_k \pmod{n_k}$$

has a unique solution modulo  $n$  for the unknown  $x$ .

## Fermat's "Little" Theorem<sup>a</sup>

**Lemma 56** For all  $0 < a < p$ ,  $a^{p-1} = 1 \pmod{p}$ .

- Consider  $a\Phi(p) = \{am \pmod{p} : m \in \Phi(p)\}$ .
- $a\Phi(p) = \Phi(p)$ .
  - Suppose  $am = am' \pmod{p}$  for  $m > m'$ , where  $m, m' \in \Phi(p)$ .
  - That means  $a(m - m') = 0 \pmod{p}$ , and  $p$  divides  $a$  or  $m - m'$ , which is impossible.
- Hence  $(p-1)! = a^{p-1}(p-1)! \pmod{p}$ .
- Finally,  $(a^{p-1} - 1) = 0 \pmod{p}$  because  $p \nmid (p-1)!$ .

---

<sup>a</sup>Pierre de Fermat (1601–1665).

## The Fermat-Euler Theorem

**Corollary 57** For all  $a \in \Phi(n)$ ,  $a^{\phi(n)} = 1 \pmod n$ .

- As  $12 = 2^2 \times 3$ ,

$$\phi(12) = 12 \times \left(1 - \frac{1}{2}\right) \left(1 - \frac{1}{3}\right) = 4$$

- In fact,  $\Phi(12) = \{1, 5, 7, 11\}$ .
- For example,

$$5^4 = 625 = 1 \pmod{12}.$$

## Exponents and Primitive Roots

- From Fermat's "little" theorem, all exponents divide  $p - 1$ .
- A primitive root of  $p$  is thus a number with exponent  $p - 1$ .
- Let  $R(k)$  denote the total number of residues in  $\Phi(p)$  that have exponent  $k$ .
- We already knew that  $R(k) = 0$  for  $k \nmid (p - 1)$ .
- Any  $a \in \Phi(p)$  of exponent  $k$  satisfies  $x^k = 1 \pmod p$ .
- Hence there are at most  $k$  residues of exponent  $k$ , i.e.,  $R(k) \leq k$ .

## Exponents

- The **exponent** of  $m \in \Phi(p)$  is the least  $k \in \mathbb{Z}^+$  such that

$$m^k = 1 \pmod p.$$

- Every residue  $s \in \Phi(p)$  has an exponent.
  - $1, s, s^2, s^3, \dots$  eventually repeats itself, say  $s^i = s^j \pmod p$ , which means  $s^{j-i} = 1 \pmod p$ .
- If the exponent of  $m$  is  $k$  and  $m^\ell = 1 \pmod p$ , then  $k \mid \ell$ .
  - Otherwise,  $\ell = qk + a$  for  $0 < a < k$ , and  $m^\ell = m^{qk+a} = m^a = 1 \pmod p$ , a contradiction.

**Lemma 58** Any nonzero polynomial of degree  $k$  has at most  $k$  distinct roots modulo  $p$ .

## Size of $R(k)$

- Let  $s$  be a residue of exponent  $k$ .
- $1, s, s^2, \dots, s^{k-1}$  are all distinct modulo  $p$ .
  - Otherwise,  $s^i = s^j \pmod p$  with  $i < j$  and  $s$  is of exponent  $j - i < k$ , a contradiction.
- As all these  $k$  distinct numbers satisfy  $x^k = 1 \pmod p$ , they are all the solutions of  $x^k = 1 \pmod p$ .
- But do all of them have exponent  $k$  (i.e.,  $R(k) = k$ )?
- And if not (i.e.,  $R(k) < k$ ), how many of them do?

### Size of $R(k)$ (continued)

- Suppose  $\ell < k$  and  $\ell \notin \Phi(k)$  with  $\gcd(\ell, k) = d > 1$ .
- Then

$$(s^\ell)^{k/d} = 1 \pmod{p}.$$

- Therefore,  $s^\ell$  has exponent at most  $k/d$ , which is less than  $k$ .
- We conclude that

$$R(k) \leq \phi(k).$$

### A Few Calculations

- Let  $p = 13$ .
- From p. 338, we know  $\phi(p - 1) = 4$ .
- Hence  $R(12) = 4$ .
- And there are 4 primitives roots of  $p$ .
- As  $\Phi(p - 1) = \{1, 5, 7, 11\}$ , the primitive roots are  $g^1, g^5, g^7, g^{11}$  for any primitive root  $g$ .

### Size of $R(k)$ (concluded)

- Because all  $p - 1$  residues have an exponent,

$$p - 1 = \sum_{k|(p-1)} R(k) \leq \sum_{k|(p-1)} \phi(k) = p - 1$$

by Lemma 54 on p. 331.

- Hence

$$R(k) = \begin{cases} \phi(k) & \text{when } k|(p-1) \\ 0 & \text{otherwise} \end{cases}$$

- In particular,  $R(p - 1) = \phi(p - 1) > 0$ , and  $p$  has at least one primitive root.
- This proves one direction of Theorem 50 (p. 324).

### The Other Direction of Theorem 50 (p. 324)

- Suppose  $p$  is not a prime.
- We proceed to show that no primitive roots exist.
- Suppose  $r^{p-1} = 1 \pmod{p}$ , the 1st condition of the primitive root on p. 324.
- We will show that the 2nd condition must be violated.
- $r^{\phi(p)} = 1 \pmod{p}$  by the Fermat-Euler theorem (p. 338).
- Because  $p$  is not a prime,  $\phi(p) < p - 1$ .

### The Other Direction of Theorem 50 (concluded)

- Let  $k$  be the smallest integer such that  $r^k = 1 \pmod p$ .
- As  $k|\phi(p)$ ,  $k < p - 1$ .
- Let  $q$  be a prime divisor of  $(p - 1)/k > 1$ .
- Then  $k|(p - 1)/q$ .
- Therefore, by virtue of the definition of  $k$ ,

$$r^{(p-1)/q} = 1 \pmod p.$$

- But this violates the 2nd condition of the primitive root on p. 324.

### Bipartite Perfect Matching

- We are given a **bipartite graph**  $G = (U, V, E)$ .
  - $U = \{u_1, u_2, \dots, u_n\}$ .
  - $V = \{v_1, v_2, \dots, v_n\}$ .
  - $E \subseteq U \times V$ .
- We are asked if there is a **perfect matching**.
  - A permutation  $\pi$  of  $\{1, 2, \dots, n\}$  such that

$$(u_i, v_{\pi(i)}) \in E$$

for all  $u_i \in U$ .

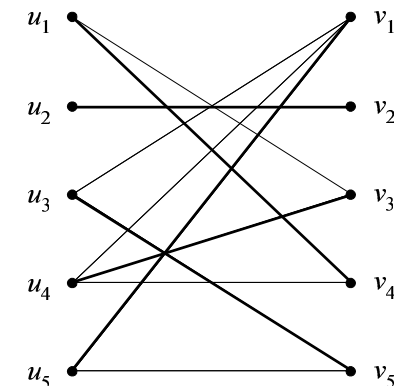
### Randomized Algorithms<sup>a</sup>

- Randomized algorithms flip unbiased coins.
- There are important problems for which there are no known efficient *deterministic* algorithms but for which very efficient randomized algorithms exist.
  - Primality tests, extraction of square roots, etc.
- There are problems where randomization is *necessary*.
  - Secure protocols.
- Are randomized algorithms algorithms<sup>b</sup>?

<sup>a</sup>Rabin, 1976, Solovay and Strassen, 1977.

<sup>b</sup>“Truth is so delicate that one has only to depart the least bit from it to fall into error.” — *The Provincial Letters*, Pascal (1623–1662).

### A Perfect Matching



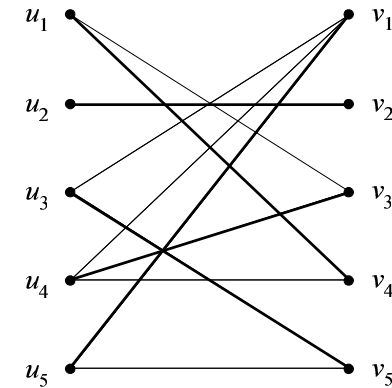
## Symbolic Determinants

- Given a bipartite graph  $G$ , construct the  $n \times n$  matrix  $A^G$  whose  $(i, j)$ th entry  $A_{ij}^G$  is a variable  $x_{ij}$  if  $(u_i, v_j) \in E$  and zero otherwise.
- The **determinant** of  $A^G$  is

$$\det(A^G) = \sum_{\pi} \sigma(\pi) \prod_{i=1}^n A_{i, \pi(i)}^G, \quad (5)$$

where  $\pi$  ranges over all permutations of  $n$  elements and  $\sigma(\pi)$  is 1 if  $\pi$  is the product of an even number of transpositions and  $-1$  otherwise.

## A Perfect Matching in a Bipartite Graph



## Determinant and Bipartite Perfect Matching

In  $\sum_{\pi} \sigma(\pi) \prod_{i=1}^n A_{i, \pi(i)}^G$ , note the following:

- Each summand corresponds to a possible perfect matching  $\pi$ .
- As all variables appear only *once*, all of these summands are different monomials and will not cancel.

**Proposition 59 (Edmonds, 1967)**  $G$  has a perfect matching if and only if  $\det(A^G)$  is not identically zero.

## The Perfect Matching in the Determinant

- The matrix is

$$A^G = \begin{bmatrix} 0 & 0 & x_{13} & \boxed{x_{14}} & 0 \\ 0 & \boxed{x_{22}} & 0 & 0 & 0 \\ x_{31} & 0 & 0 & 0 & \boxed{x_{35}} \\ x_{41} & 0 & \boxed{x_{43}} & x_{44} & 0 \\ \boxed{x_{51}} & 0 & 0 & 0 & x_{55} \end{bmatrix}.$$

- $\det(A^G)$  contains term  $x_{14}x_{22}x_{35}x_{43}x_{51}$ , which denotes a perfect matching.

### How To Test If a Polynomial Is Identically Zero?

- $\det(A^G)$  is a polynomial in  $n^2$  variables.
- There are exponentially many terms in  $\det(A^G)$ .
- Expanding the determinant polynomial is not feasible.
  - Too many terms.
- Observation: If  $\det(A^G)$  is *identically zero*, then it remains zero if we substitute *arbitrary* integers for the variables  $x_{11}, \dots, x_{nn}$ .
- What is the likelihood of obtaining a zero when  $\det(A^G)$  is *not* identically zero?

### Density Attack

- The density of roots in the domain is at most

$$\frac{mdM^{m-1}}{M^m} = \frac{md}{M}.$$

- This suggests a sampling algorithm.

### Number of Roots of a Polynomials

**Lemma 60 (Schwartz, 1980)** Let  $p(x_1, x_2, \dots, x_m) \not\equiv 0$  be a polynomial in  $m$  variables each of degree at most  $d$ . Let  $M \in \mathbb{Z}^+$ . Then the number of  $m$ -tuples

$$(x_1, x_2, \dots, x_m) \in \{0, 1, \dots, M-1\}^m$$

such that  $p(x_1, x_2, \dots, x_m) = 0$  is

$$\leq mdM^{m-1}.$$

- By induction on  $m$ .

### A Randomized Bipartite Perfect Matching Algorithm<sup>a</sup>

- 1: Choose  $n^2$  integers  $i_{11}, \dots, i_{nn}$  from  $\{0, 1, \dots, b-1\}$  randomly;
- 1: Calculate  $\det(A^G(i_{11}, \dots, i_{nn}))$  by Gaussian elimination;
- 2: **if**  $\det(A^G(i_{11}, \dots, i_{nn})) \neq 0$  **then**
- 3:     **return** “ $G$  has a perfect matching”;
- 4: **else**
- 5:     **return** “ $G$  has no perfect matchings”;
- 6: **end if**

---

<sup>a</sup>Lovász, 1979.

## Analysis

- Pick  $b$  such that  $b^{n^2} = 2n^2$ .
- If  $G$  has no perfect matchings, the algorithm will always be correct.
- Suppose  $G$  has a perfect matching.
  - The algorithm will answer incorrectly with probability at most  $n^2d/b = 0.5$  because  $d = 1$ .
  - Repeat the algorithm *independently*  $k$  times and output “ $G$  has no perfect matchings” if all of the  $k$  runs say so.
  - The error probability is now reduced to at most  $2^{-k}$ .

## The Markov Inequality<sup>a</sup>

**Lemma 61** *Let  $x$  be a random variable taking nonnegative integer values. Then for any  $k > 0$ ,*

$$\text{prob}[x \geq kE[x]] \leq 1/k.$$

- Let  $p_i$  denote the probability that  $x = i$ .

$$\begin{aligned} E[x] &= \sum_i ip_i \\ &= \sum_{i < kE[x]} ip_i + \sum_{i \geq kE[x]} ip_i \\ &\geq kE[x] \times \text{prob}[x \geq kE[x]]. \end{aligned}$$

---

<sup>a</sup>Andrei Andreyevich Markov (1856–1922).

## Monte Carlo Algorithms

- The randomized bipartite perfect matching algorithm is called a **Monte Carlo algorithm** in the sense that
  - If the algorithm finds that a matching exists, it is always correct (no **false positives**).
  - If the algorithm answers in the negative, then it may make an error (**false negative**).
- The probability that the algorithm makes a false negative is at most 0.5.
- This probability is *not* over the space of all graphs or determinants, but *over* the algorithm’s own coin flips.
  - It holds for *any* bipartite graph.

## An Application of Markov’s Inequality

- Algorithm  $C$  runs in expected time  $T(n)$  and always gives the right answer.
- Consider an algorithm that runs  $C$  for time  $kT(n)$  and rejects the input if  $C$  does not stop within the time bound.
- By Markov’s inequality, this new algorithm runs in time  $kT(n)$  and gives the correct answer with probability at least  $1 - (1/k)$ .
- By running this algorithm  $m$  times, we reduce the error probability to  $\leq k^{-m}$ .

### A Random Walk Algorithm for $\phi$ in CNF Form

- 1: Start with an *arbitrary* truth assignment  $T$ ;
- 2: **for**  $i = 1, 2, \dots, r$  **do**
- 3:   **if**  $T \models \phi$  **then**
- 4:     **return** “ $\phi$  is satisfiable”;
- 5:   **else**
- 6:     Let  $c$  be an unsatisfiable clause in  $\phi$  under  $T$ ; {All of its literals are false under  $T$ .}
- 7:     Pick any  $x$  of these literals *at random*;
- 8:     Modify  $T$  to make  $x$  true;
- 9:   **end if**
- 10: **end for**
- 11: **return** “ $\phi$  is unsatisfiable”;

### The Proof

- Let  $\hat{T}$  be a truth assignment such that  $\hat{T} \models \phi$ .
- Let  $t(i)$  denote the expected number of repetitions of the flipping step until a satisfying truth assignment is found if our starting  $T$  differs from  $\hat{T}$  in  $i$  values.
  - Their Hamming distance is  $i$ .
- It can be shown that  $t(i)$  is finite.
- $t(0) = 0$  because it means that  $T = \hat{T}$  and hence  $T \models \phi$ .
- If  $T \neq \hat{T}$  or  $T$  is not equal to any other satisfying truth assignment, then we need to flip at least once.

### 3SAT and 2SAT Again

- Note that if  $\phi$  is unsatisfiable, the algorithm will not refute it.
- The random walk algorithm runs in exponential time for 3SAT.
- But we will show that it works well for 2SAT.

**Theorem 62** *Suppose the random walk algorithm with  $r = 2n^2$  is applied to any satisfiable 2SAT problem with  $n$  variables. Then a satisfying truth assignment will be discovered with probability at least 0.5.*

### The Proof (continued)

- We flip to pick among the 2 literals of a clause not satisfied by the present  $T$ .
- At least one of the 2 literals is true under  $\hat{T}$ , because  $\hat{T}$  satisfies all clauses.
- So we have at least 0.5 chance of moving closer to  $\hat{T}$ .
- Thus

$$t(i) \leq \frac{t(i-1) + t(i+1)}{2} + 1$$

for  $0 < i < n$ .

- Inequality is used because, for example,  $T$  may differ from  $\hat{T}$  in both literals.



### The Proof (continued)

- It must also hold that

$$t(n) \leq t(n-1) + 1$$

because at  $i = n$ , we can only decrease  $i$ .

- As we are only interested in upper bounds, we solve

$$x(0) = 0$$

$$x(n) = x(n-1) + 1$$

$$x(i) = \frac{x(i-1) + x(i+1)}{2} + 1, \quad 0 < i < n$$

- This is one-dimensional random walk with a reflecting and an absorbing barrier.

### The Proof (continued)

- Iteratively, we obtain

$$x(2) = 4n - 4$$

$$\vdots$$

$$x(i) = 2in - i^2$$

- The worst case happens when  $i = n$ , in which case

$$x(n) = n^2.$$

### The Proof (continued)

- Add the equations up to obtain

$$\begin{aligned} & x(1) + x(2) + \cdots + x(n) \\ = & \frac{x(0) + x(1) + 2x(2) + \cdots + 2x(n-2) + x(n-1) + x(n)}{2} \\ & + n + x(n-1). \end{aligned}$$

- Simplify to yield

$$\frac{x(1) + x(n) - x(n-1)}{2} = n.$$

- As  $x(n) - x(n-1) = 1$ , we have

$$x(1) = 2n - 1.$$

### The Proof (concluded)

- We therefore reach the conclusion that

$$t(i) \leq x(i) \leq x(n) = n^2.$$

- So the expected number of steps is at most  $n^2$ .
- The algorithm picks a running time  $2n^2$ .
- This amounts to invoking the Markov inequality (p. 360) with  $k = 2$ , with the consequence of having a probability of 0.5.

## Boosting the Performance

- We can pick  $r = 2mn^2$  to have an error probability of  $\leq (2m)^{-1}$  by Markov's inequality.
- Alternatively, with the same running time, we can run the " $r = 2n^2$ " algorithm  $m$  times.
- But the error probability is reduced to  $\leq 2^{-m}$ !
- The gain comes from the fact that Markov's inequality does not take advantage of any specific feature of the random variable.
- The gain also comes from the fact that the two algorithms are different.

## The Density Attack for PRIMES

- 1: Pick  $k \in \{2, \dots, p-1\}$  randomly; {Assume  $p > 2$ .}
- 2: **if**  $k | p$  **then**
- 3:     **return** " $N$  is a composite";
- 4: **else**
- 5:     **return** " $N$  is a prime";
- 6: **end if**

The probability of success when  $p$  is composite is  $1 - \phi(p)/p$ .

## Primality Tests

- PRIMES asks if a number  $p$  is a prime.
- The classic algorithm tests if  $k | p$  for  $k = 2, 3, \dots, \sqrt{p}$ .
- But it runs in  $\Omega(2^{n/2})$  steps, where  $n = |p| = \log_2 p$ .

## The Fermat Test for Primality

- Fermat's "little" theorem on p. 337 suggests the following primality test for any given number  $p$ :
  - Pick a number  $a$  randomly from  $\{1, 2, \dots, p-1\}$ .
  - If  $a^{p-1} \neq 1 \pmod p$ , then declare " $p$  is composite."
  - Otherwise, declare " $p$  is probably prime."
- Unfortunately, there are composite numbers called **Carmichael numbers** that will pass the Fermat test for all  $a \in \{1, 2, \dots, p-1\}$ .
- It is only recently that Carmichael numbers are known to be infinite in number.

## Euler's Test

**Lemma 63 (Euler)** Let  $p$  be an odd prime and  $a \neq 0 \pmod p$ .

1. If  $a^{(p-1)/2} = 1 \pmod p$ , then  $x^2 = a \pmod p$  has two roots.
  2. If  $a^{(p-1)/2} \neq 1 \pmod p$ , then  $a^{(p-1)/2} = -1 \pmod p$  and  $x^2 = a \pmod p$  has no roots.
- Let  $r$  be a primitive root of  $p$ .
  - If  $a = r^{2j}$ , then  $a^{(p-1)/2} = r^{j(p-1)} = 1 \pmod p$  and its two distinct roots are  $r^j, -r^j (= r^{j+(p-1)/2})$ .

## Square Roots Modulo a Prime

- Equation  $x^2 = a \pmod p$  has at most two (distinct) roots by Lemma 58 on p. 339.
  - The roots are called **square roots**.
  - Numbers  $a$  with square roots and  $\gcd(a, p) = 1$  are called **quadratic residues**:  
 $1^2 \pmod p, 2^2 \pmod p, \dots, (p-1)^2 \pmod p$ .
- We shall show that a number either has two roots or has none, and testing which is true is trivial.
- We remark that there are no known efficient *deterministic* algorithms to find the roots.

## The Proof (concluded)

- Since there are  $(p-1)/2$  such  $a$ 's, and each such  $a$  has two distinct roots, we have run out of *square roots*.
  - $\{c : c^2 = a \pmod p\} = \{1, 2, \dots, p-1\}$ .
- If  $a = r^{2j+1}$ , then it has no roots because all the square roots are taken.
- By Fermat's "little" theorem,  $r^{(p-1)/2}$  is a square root of 1, so  $r^{(p-1)/2} = \pm 1 \pmod p$ .
- But as  $r$  is a primitive root,  $r^{(p-1)/2} = -1 \pmod p$ .
- $a^{(p-1)/2} = (r^{(p-1)/2})^{2j+1} = (-1)^{2j+1} = -1 \pmod p$ .