

Monte Carlo Algorithms^a

- 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 negatives**).

^aMetropolis and Ulam (1949).

Monte Carlo Algorithms (concluded)

- The algorithm makes a false negative with probability ≤ 0.5 .^a
 - Note this probability refers to^b

$\text{prob}[\text{algorithm answers “no”} \mid G \text{ has a perfect matching}]$

not

$\text{prob}[G \text{ has a perfect matching} \mid \text{algorithm answers “no”}]$.

- 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.

^aEquivalently, among the coin flip sequences, at most half of them lead to the wrong answer.

^bIn general, $\text{prob}[\text{algorithm answers “no”} \mid \text{input is a “yes” instance}]$.

The Markov Inequality^a

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 $x = i$.

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

^aAndrei Andreyevich Markov (1856–1922).

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An Application of Markov's Inequality

- Suppose 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 wrong answer with probability $\leq 1/k$.

An Application of Markov's Inequality (concluded)

- By running this algorithm m times, we reduce the error probability to $\leq k^{-m}$.^a
- Suppose, instead, we run the algorithm for the same running time $mkT(n)$ once and rejects the input if it does not stop within the time bound.
- By Markov's inequality, this new algorithm gives the wrong answer with probability $\leq 1/(mk)$.
- This is much worse than the previous algorithm's error probability of $\leq k^{-m}$ for the same amount of time.

^aWith the same input. Thanks to a question on December 7, 2010.

FSAT for k -SAT Formulas (p. 428)

- Let $\phi(x_1, x_2, \dots, x_n)$ be a k -SAT formula.
- If ϕ is satisfiable, then return a satisfying truth assignment.
- Otherwise, return “no.”
- We next propose a randomized algorithm for this problem.

A Random Walk Algorithm for ϕ 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** “ ϕ is satisfiable with T ”;
- 5: **else**
- 6: Let c be an unsatisfied clause in ϕ 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** “ ϕ is unsatisfiable”;

3SAT vs. 2SAT Again

- Note that if ϕ is unsatisfiable, the algorithm will not refute it.
- The random walk algorithm needs expected exponential time for 3SAT.
 - In fact, it runs in expected $O((1.333 \dots + \epsilon)^n)$ time with $r = 3n$,^a much better than $O(2^n)$.^b
- We will show immediately that it works well for 2SAT.
- The state of the art as of 2006 is expected $O(1.322^n)$ time for 3SAT and expected $O(1.474^n)$ time for 4SAT.^c

^aUse this setting per run of the algorithm.

^bSchöning (1999).

^cKwama and Tamaki (2004); Rolf (2006).

Random Walk Works for 2SAT^a

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.*

- Let \hat{T} be a truth assignment such that $\hat{T} \models \phi$.
- Assume our starting T differs from \hat{T} in i values.
 - Their Hamming distance is i .
 - Recall T is arbitrary.
- Let $t(i)$ denote the expected number of repetitions of the flipping step until a satisfying truth assignment is found.

^aPapadimitriou (1991).

The Proof

- 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 any other satisfying truth assignment, then we need to flip the coin at least once.
- We flip a coin 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} .

The Proof (continued)

- 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.
- It must also hold that

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

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

The Proof (continued)

- 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 an absorbing barrier at $i = 0$ and a reflecting barrier at $i = n$.

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 (continued)

- Iteratively, we obtain

$$\begin{aligned}x(2) &= 4n - 4, \\ &\vdots \\ x(i) &= 2in - i^2.\end{aligned}$$

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

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

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 $r = 2n^2$.
 - This amounts to invoking the Markov inequality (p. 460) with $k = 2$, with the consequence of having a probability of 0.5.
- The proof does not yield a polynomial bound for 3SAT.^a

^aContributed by Mr. Cheng-Yu Lee (R95922035) on November 8, 2006.

Boosting the Performance

- We can pick $r = 2mn^2$ to have an error probability of

$$\leq \frac{1}{2m}$$

by Markov's inequality.

- Alternatively, with the same running time, we can run the “ $r = 2n^2$ ” algorithm m times.
- The error probability is now reduced to

$$\leq 2^{-m}.$$

Primality Tests

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

Primality Tests (concluded)

- Suppose $N = PQ$ is a product of 2 distinct primes.
- The probability of success of the density attack (p. 409) is

$$\approx \frac{2}{\sqrt{N}}$$

when $P \approx Q$.

- This probability is exponentially small in terms of the input length $\log_2 N$.

The Fermat Test for Primality

Fermat's "little" theorem on p. 412 suggests the following primality test for any given number N :

- 1: Pick a number a randomly from $\{1, 2, \dots, N - 1\}$;
- 2: **if** $a^{N-1} \not\equiv 1 \pmod{N}$ **then**
- 3: **return** " N is composite";
- 4: **else**
- 5: **return** " N is a prime";
- 6: **end if**

The Fermat Test for Primality (concluded)

- Unfortunately, there are composite numbers called **Carmichael numbers** that will pass the Fermat test for *all* $a \in \{1, 2, \dots, N - 1\}$.^a
 - The Fermat test will return “ N is a prime” for all Carmichael numbers N .
- There are infinitely many Carmichael numbers.^b
- In fact, the number of Carmichael numbers less than N exceeds $N^{2/7}$ for N large enough.

^aCarmichael (1910).

^bAlford, Granville, and Pomerance (1992).

Square Roots Modulo a Prime

- Equation $x^2 = a \pmod{p}$ has at most two (distinct) roots by Lemma 57 (p. 417).
 - The roots are called **square roots**.
 - Numbers a with square roots *and* $\gcd(a, p) = 1$ are called **quadratic residues**.
 - * They are

$$1^2 \pmod{p}, 2^2 \pmod{p}, \dots, (p-1)^2 \pmod{p}.$$

- We shall show that a number either has two roots or has none, and testing which one is true is trivial.^a

^aNo efficient *deterministic* root-finding algorithms are known yet.

Euler's Test

Lemma 63 (Euler) *Let p be an odd prime and $a \not\equiv 0 \pmod{p}$.*

1. *If*

$$a^{(p-1)/2} \equiv 1 \pmod{p},$$

then $x^2 \equiv a \pmod{p}$ has two roots.

2. *If*

$$a^{(p-1)/2} \not\equiv 1 \pmod{p},$$

then

$$a^{(p-1)/2} \equiv -1 \pmod{p}$$

and $x^2 \equiv a \pmod{p}$ has no roots.

The Proof (continued)

- Let r be a primitive root of p .
- By Fermat's "little" theorem,

$$r^{(p-1)/2}$$

is a square root of 1.

- So

$$r^{(p-1)/2} = 1 \text{ or } -1 \pmod{p}.$$

- But as r is a primitive root, $r^{(p-1)/2} \neq 1 \pmod{p}$.
- Hence

$$r^{(p-1)/2} = -1 \pmod{p}.$$

The Proof (continued)

- Let $a = r^k \pmod p$ for some k .
- Then

$$1 = a^{(p-1)/2} = r^{k(p-1)/2} = \left[r^{(p-1)/2} \right]^k = (-1)^k \pmod p.$$

- So k must be even.
- Suppose $a = r^{2j}$ for some $1 \leq j \leq (p-1)/2$.
- Then $a^{(p-1)/2} = r^{j(p-1)} = 1 \pmod p$, and a 's two *distinct* roots are $r^j, -r^j (= r^{j+(p-1)/2} \pmod p)$.
 - If $r^j = -r^j \pmod p$, then $2r^j = 0 \pmod p$, which implies $r^j = 0 \pmod p$, a contradiction.

The Proof (continued)

- As $1 \leq j \leq (p - 1)/2$, there are $(p - 1)/2$ such a 's.
- Each such a has 2 distinct square roots.
- The square roots of all the a 's are distinct.
 - The square roots of different a 's must be different.
- Hence the set of *square roots* is $\{1, 2, \dots, p - 1\}$.
- As a result, $a = r^{2j}$, $1 \leq j \leq (p - 1)/2$, exhaust all the quadratic residues.

The Proof (concluded)

- If $a = r^{2j+1}$, then it has no roots because all the square roots have been taken.
- Now,

$$a^{(p-1)/2} = \left[r^{(p-1)/2} \right]^{2j+1} = (-1)^{2j+1} = -1 \pmod{p}.$$

The Legendre Symbol^a and Quadratic Residuacity Test

- By Lemma 63 (p. 480) $a^{(p-1)/2} \pmod p = \pm 1$ for $a \not\equiv 0 \pmod p$.
- For odd prime p , define the **Legendre symbol** $(a | p)$ as

$$(a | p) = \begin{cases} 0 & \text{if } p | a, \\ 1 & \text{if } a \text{ is a quadratic residue modulo } p, \\ -1 & \text{if } a \text{ is a **quadratic nonresidue** modulo } p. \end{cases}$$

- Euler's test implies $a^{(p-1)/2} \equiv (a | p) \pmod p$ for any odd prime p and any integer a .
- Note that $(ab | p) = (a | p)(b | p)$.

^aAndrien-Marie Legendre (1752–1833).

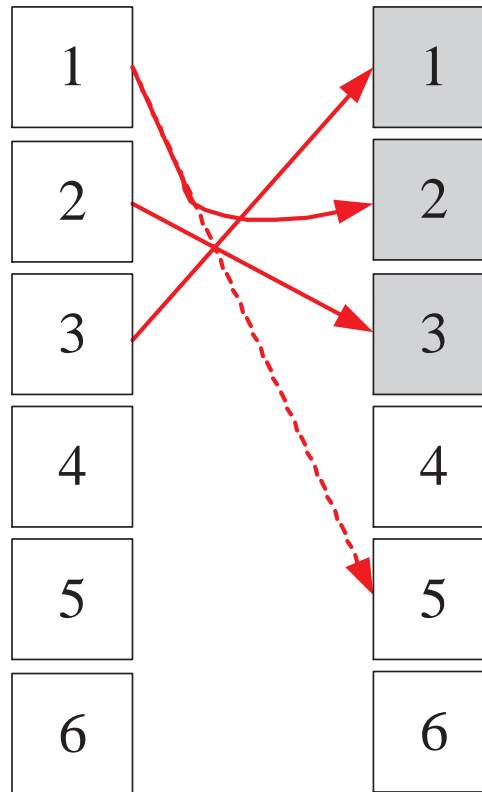
Gauss's Lemma

Lemma 64 (Gauss) *Let p and q be two odd primes. Then $(q|p) = (-1)^m$, where m is the number of residues in $R = \{iq \bmod p : 1 \leq i \leq (p-1)/2\}$ that are greater than $(p-1)/2$.*

- All residues in R are distinct.
 - If $iq = jq \bmod p$, then $p|(j-i)q$ or $p|q$.
- No two elements of R add up to p .
 - If $iq + jq = 0 \bmod p$, then $p|(i+j)$ or $p|q$.
 - But neither is possible.

The Proof (continued)

- Consider the set R' of residues that result from R if we replace each of the m elements $a \in R$ such that $a > (p - 1)/2$ by $p - a$.
 - This is equivalent to performing $-a \pmod{p}$.
- All residues in R' are now at most $(p - 1)/2$.
- In fact, $R' = \{1, 2, \dots, (p - 1)/2\}$ (see illustration next page).
 - Otherwise, two elements of R would add up to p , which has been shown to be impossible.



$p = 7$ and $q = 5$.

The Proof (concluded)

- Alternatively, $R' = \{\pm iq \bmod p : 1 \leq i \leq (p-1)/2\}$, where exactly m of the elements have the minus sign.
- Take the product of all elements in the two representations of R' .
- So

$$[(p-1)/2]! = (-1)^m q^{(p-1)/2} [(p-1)/2]! \bmod p.$$

- Because $\gcd([(p-1)/2]!, p) = 1$, the above implies

$$1 = (-1)^m q^{(p-1)/2} \bmod p.$$

Legendre's Law of Quadratic Reciprocity^a

- Let p and q be two odd primes.
- The next result says their Legendre symbols are distinct if and only if both numbers are 3 mod 4.

Lemma 65 (Legendre (1785), Gauss)

$$(p|q)(q|p) = (-1)^{\frac{p-1}{2} \frac{q-1}{2}}.$$

^aFirst stated by Euler in 1751. Legendre (1785) did not give a correct proof. Gauss proved the theorem when he was 19. He gave at least 6 different proofs during his life. The 152nd proof appeared in 1963.

The Proof (continued)

- Sum the elements of R' in the previous proof in mod 2.
- On one hand, this is just $\sum_{i=1}^{(p-1)/2} i \pmod{2}$.
- On the other hand, the sum equals

$$\begin{aligned} & \sum_{i=1}^{(p-1)/2} \left(qi - p \left\lfloor \frac{qi}{p} \right\rfloor \right) + mp \pmod{2} \\ &= \left(q \sum_{i=1}^{(p-1)/2} i - p \sum_{i=1}^{(p-1)/2} \left\lfloor \frac{qi}{p} \right\rfloor \right) + mp \pmod{2}. \end{aligned}$$

- Signs are irrelevant under mod 2.
- m is as in Lemma 64 (p. 486).

The Proof (continued)

- Ignore odd multipliers to make the sum equal

$$\left(\sum_{i=1}^{(p-1)/2} i - \sum_{i=1}^{(p-1)/2} \left[\frac{qi}{p} \right] \right) + m \pmod{2}.$$

- Equate the above with $\sum_{i=1}^{(p-1)/2} i \pmod{2}$ to obtain

$$m = \sum_{i=1}^{(p-1)/2} \left[\frac{qi}{p} \right] \pmod{2}.$$

The Proof (concluded)

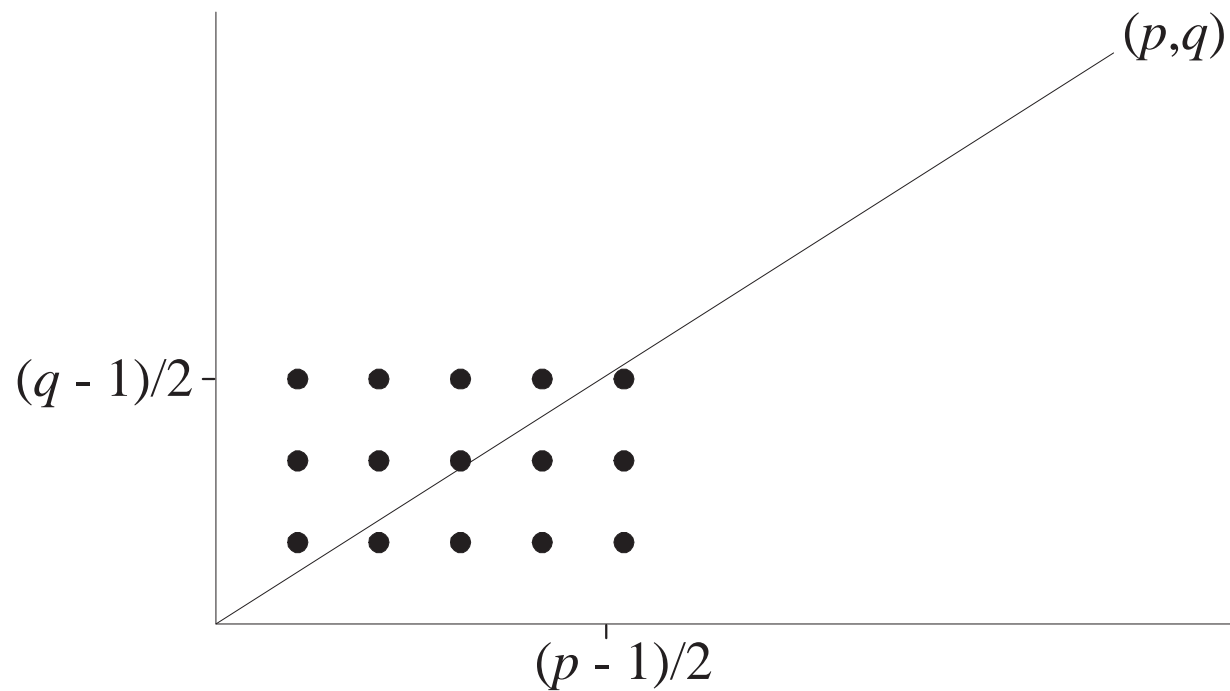
- $\sum_{i=1}^{(p-1)/2} \lfloor \frac{qi}{p} \rfloor$ is the number of integral points under the line

$$y = (q/p)x$$

for $1 \leq x \leq (p-1)/2$.

- Gauss's lemma (p. 486) says $(q|p) = (-1)^m$.
- Repeat the proof with p and q reversed.
- So $(p|q) = (-1)^{m'}$, where m' is the number of integral points *above* the line $y = (q/p)x$ for $1 \leq y \leq (q-1)/2$.
- As a result, $(p|q)(q|p) = (-1)^{m+m'}$.
- But $m + m'$ is the total number of integral points in the $\frac{p-1}{2} \times \frac{q-1}{2}$ rectangle, which is $\frac{p-1}{2} \frac{q-1}{2}$.

Eisenstein's Rectangle



$p = 11$ and $q = 7$.

The Jacobi Symbol^a

- The Legendre symbol only works for odd *prime* moduli.
- The **Jacobi symbol** $(a | m)$ extends it to cases where m is not prime.
- Let $m = p_1 p_2 \cdots p_k$ be the prime factorization of m .
- When $m > 1$ is odd and $\gcd(a, m) = 1$, then

$$(a | m) = \prod_{i=1}^k (a | p_i).$$

- Note that the Jacobi symbol equals ± 1 .
- It reduces to the Legendre symbol when m is a prime.
- Define $(a | 1) = 1$.

^aCarl Jacobi (1804–1851).

Properties of the Jacobi Symbol

The Jacobi symbol has the following properties, for arguments for which it is defined.

1. $(ab | m) = (a | m)(b | m)$.
2. $(a | m_1 m_2) = (a | m_1)(a | m_2)$.
3. If $a = b \pmod{m}$, then $(a | m) = (b | m)$.
4. $(-1 | m) = (-1)^{(m-1)/2}$ (by Lemma 64 on p. 486).
5. $(2 | m) = (-1)^{(m^2-1)/8}$.^a
6. If a and m are both odd, then
$$(a | m)(m | a) = (-1)^{(a-1)(m-1)/4}.$$

^aBy Lemma 64 (p. 486) and some parity arguments.

Properties of the Jacobi Symbol (concluded)

- These properties allow us to calculate the Jacobi symbol without factorization.
- This situation is similar to the Euclidean algorithm.
- Note also that $(a | m) = 1/(a | m)$ because $(a | m) = \pm 1$.^a

^aContributed by Mr. Huang, Kuan-Lin (B96902079, R00922018) on December 6, 2011.

Calculation of $(2200|999)$

$$\begin{aligned}(202|999) &= (2|999)(101|999) \\ &= (-1)^{(999^2-1)/8}(101|999) \\ &= (-1)^{124750}(101|999) = (101|999) \\ &= (-1)^{(100)(998)/4}(999|101) = (-1)^{24950}(999|101) \\ &= (999|101) = (90|101) = (-1)^{(101^2-1)/8}(45|101) \\ &= (-1)^{1275}(45|101) = -(45|101) \\ &= -(-1)^{(44)(100)/4}(101|45) = -(101|45) = -(11|45) \\ &= -(-1)^{(10)(44)/4}(45|11) = -(45|11) \\ &= -(1|11) = -1.\end{aligned}$$

A Result Generalizing Proposition 10.3 in the Textbook

Theorem 66 *The group of set $\Phi(n)$ under multiplication mod n has a primitive root if and only if n is either 1, 2, 4, p^k , or $2p^k$ for some nonnegative integer k and an odd prime p .*

This result is essential in the proof of the next lemma.

The Jacobi Symbol and Primality Test^a

Lemma 67 *If $(M|N) = M^{(N-1)/2} \pmod N$ for all $M \in \Phi(N)$, then N is a prime. (Assume N is odd.)*

- Assume $N = mp$, where p is an odd prime, $\gcd(m, p) = 1$, and $m > 1$ (not necessarily prime).
- Let $r \in \Phi(p)$ such that $(r|p) = -1$.
- The Chinese remainder theorem says that there is an $M \in \Phi(N)$ such that

$$M = r \pmod p,$$

$$M = 1 \pmod m.$$

^aMr. Clement Hsiao (R88526067) pointed out that the textbook's proof for Lemma 11.8 is incorrect while he was a senior in January 1999.

The Proof (continued)

- By the hypothesis,

$$M^{(N-1)/2} = (M | N) = (M | p)(M | m) = -1 \pmod{N}.$$

- Hence

$$M^{(N-1)/2} = -1 \pmod{m}.$$

- But because $M = 1 \pmod{m}$,

$$M^{(N-1)/2} = 1 \pmod{m},$$

a contradiction.

The Proof (continued)

- Second, assume that $N = p^a$, where p is an odd prime and $a \geq 2$.
- By Theorem 66 (p. 499), there exists a primitive root r modulo p^a .
- From the assumption,

$$M^{N-1} = \left[M^{(N-1)/2} \right]^2 = (M|N)^2 = 1 \pmod{N}$$

for all $M \in \Phi(N)$.

The Proof (continued)

- As $r \in \Phi(N)$ (prove it), we have

$$r^{N-1} = 1 \pmod{N}.$$

- As r 's exponent modulo $N = p^a$ is $\phi(N) = p^{a-1}(p-1)$,

$$p^{a-1}(p-1) \mid N-1,$$

which implies that $p \mid N-1$.

- But this is impossible given that $p \mid N$.

The Proof (continued)

- Third, assume that $N = mp^a$, where p is an odd prime, $\gcd(m, p) = 1$, $m > 1$ (not necessarily prime), and a is even.
- The proof mimics that of the second case.
- By Theorem 66 (p. 499), there exists a primitive root r modulo p^a .
- From the assumption,

$$M^{N-1} = \left[M^{(N-1)/2} \right]^2 = (M|N)^2 = 1 \pmod{N}$$

for all $M \in \Phi(N)$.

The Proof (continued)

- In particular,

$$M^{N-1} = 1 \pmod{p^a} \quad (7)$$

for all $M \in \Phi(N)$.

- The Chinese remainder theorem says that there is an $M \in \Phi(N)$ such that

$$M = r \pmod{p^a},$$

$$M = 1 \pmod{m}.$$

- Because $M = r \pmod{p^a}$ and Eq. (7),

$$r^{N-1} = 1 \pmod{p^a}.$$

The Proof (concluded)

- As r 's exponent modulo $N = p^a$ is $\phi(N) = p^{a-1}(p - 1)$,

$$p^{a-1}(p - 1) \mid N - 1,$$

which implies that $p \mid N - 1$.

- But this is impossible given that $p \mid N$.

The Number of Witnesses to Compositeness

Theorem 68 (Solovay and Strassen (1977)) *If N is an odd composite, then $(M|N) = M^{(N-1)/2} \pmod N$ for at most half of $M \in \Phi(N)$.*

- By Lemma 67 (p. 500) there is at least one $a \in \Phi(N)$ such that $(a|N) \neq a^{(N-1)/2} \pmod N$.
- Let $B = \{b_1, b_2, \dots, b_k\} \subseteq \Phi(N)$ be the set of *all* distinct residues such that $(b_i|N) = b_i^{(N-1)/2} \pmod N$.
- Let $aB = \{ab_i \pmod N : i = 1, 2, \dots, k\}$.

The Proof (concluded)

- $|aB| = k$.
 - $ab_i = ab_j \pmod N$ implies $N|a(b_i - b_j)$, which is impossible because $\gcd(a, N) = 1$ and $N > |b_i - b_j|$.
- $aB \cap B = \emptyset$ because
$$(ab_i)^{(N-1)/2} = a^{(N-1)/2} b_i^{(N-1)/2} \neq (a|N)(b_i|N) = (ab_i|N).$$
- Combining the above two results, we know

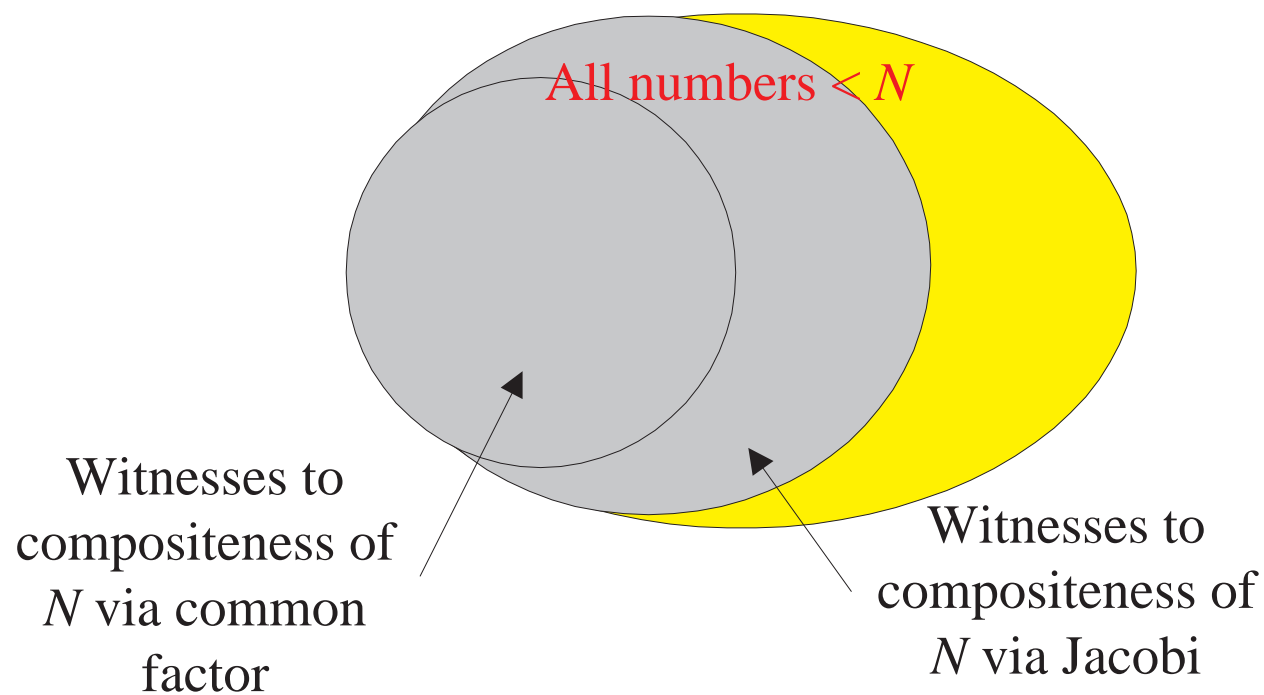
$$\frac{|B|}{\phi(N)} \leq \frac{|B|}{|B \cup aB|} = 0.5.$$

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1: if  $N$  is even but  $N \neq 2$  then
2:   return “ $N$  is composite”;  
3: else if  $N = 2$  then
4:   return “ $N$  is a prime”;  
5: end if
6: Pick  $M \in \{2, 3, \dots, N - 1\}$  randomly;  
7: if  $\gcd(M, N) > 1$  then
8:   return “ $N$  is composite”;  
9: else
10:  if  $(M|N) \neq M^{(N-1)/2} \pmod N$  then
11:    return “ $N$  is composite”;  
12:  else
13:    return “ $N$  is a prime”;  
14:  end if
15: end if
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Analysis

- The algorithm certainly runs in polynomial time.
- There are no false positives (for COMPOSITENESS).
 - When the algorithm says the number is composite, it is always correct.
- The probability of a false negative is at most one half.
 - If the input is composite, then the probability that the algorithm says the number is a prime is ≤ 0.5 .
- So it is a Monte Carlo algorithm for COMPOSITENESS.

The Improved Density Attack for COMPOSITENESS



Randomized Complexity Classes; RP

- Let N be a polynomial-time precise NTM that runs in time $p(n)$ and has 2 nondeterministic choices at each step.
- N is a **polynomial Monte Carlo Turing machine** for a language L if the following conditions hold:
 - If $x \in L$, then at least half of the $2^{p(n)}$ computation paths of N on x halt with “yes” where $n = |x|$.
 - If $x \notin L$, then all computation paths halt with “no.”
- The class of all languages with polynomial Monte Carlo TMs is denoted **RP** (**randomized polynomial time**).^a

^aAdleman and Manders (1977).

Comments on RP

- Nondeterministic steps can be seen as fair coin flips.
- There are no false positive answers.
- The probability of false negatives, $1 - \epsilon$, is at most 0.5.
- But any constant between 0 and 1 can replace 0.5.
 - By repeating the algorithm $k = \lceil -\frac{1}{\log_2 1 - \epsilon} \rceil$ times, the probability of false negatives becomes $(1 - \epsilon)^k \leq 0.5$.
- In fact, ϵ can be arbitrarily close to 0 as long as it is of the order $1/q(n)$ for some polynomial $q(n)$.
 - $-\frac{1}{\log_2 1 - \epsilon} = O(\frac{1}{\epsilon}) = O(q(n))$.

Where RP Fits

- $P \subseteq RP \subseteq NP$.
 - A deterministic TM is like a Monte Carlo TM except that all the coin flips are ignored.
 - A Monte Carlo TM is an NTM with extra demands on the number of accepting paths.
- $COMPOSITENESS \in RP$; $PRIMES \in coRP$; $PRIMES \in RP$.^a
 - In fact, $PRIMES \in P$.^b
- $RP \cup coRP$ is an alternative “plausible” notion of efficient computation.

^aAdleman and Huang (1987).

^bAgrawal, Kayal, and Saxena (2002).

ZPP^a (Zero Probabilistic Polynomial)

- The class **ZPP** is defined as $\text{RP} \cap \text{coRP}$.
- A language in ZPP has *two* Monte Carlo algorithms, one with no false positives and the other with no false negatives.
- If we repeatedly run both Monte Carlo algorithms, *eventually* one definite answer will come (unlike RP).
 - A *positive* answer from the one without false positives.
 - A *negative* answer from the one without false negatives.

^aGill (1977).

The ZPP Algorithm (Las Vegas)

- 1: {Suppose $L \in \text{ZPP}$.}
- 2: { N_1 has no false positives, and N_2 has no false negatives.}
- 3: **while true do**
- 4: **if** $N_1(x) = \text{"yes"}$ **then**
- 5: **return** "yes";
- 6: **end if**
- 7: **if** $N_2(x) = \text{"no"}$ **then**
- 8: **return** "no";
- 9: **end if**
- 10: **end while**

ZPP (concluded)

- The *expected* running time for the correct answer to emerge is polynomial.
 - The probability that a run of the 2 algorithms does not generate a definite answer is 0.5 (why?).
 - Let $p(n)$ be the running time of each run of the while-loop.
 - The expected running time for a definite answer is

$$\sum_{i=1}^{\infty} 0.5^i i p(n) = 2p(n).$$

- Essentially, ZPP is the class of problems that can be solved without errors in expected polynomial time.