

Large Deviations

- Suppose you have a *biased* coin.
- One side has probability $0.5 + \epsilon$ to appear and the other $0.5 - \epsilon$, for some $0 < \epsilon < 0.5$.
- But you do not know which is which.
- How to decide which side is the more likely side—with high confidence?
- Answer: Flip the coin many times and pick the side that appeared the most times.
- Question: Can you quantify the confidence?

The Chernoff Bound^a

Theorem 69 (Chernoff (1952)) *Suppose x_1, x_2, \dots, x_n are independent random variables taking the values 1 and 0 with probabilities p and $1 - p$, respectively. Let $X = \sum_{i=1}^n x_i$. Then for all $0 \leq \theta \leq 1$,*

$$\text{prob}[X \geq (1 + \theta)pn] \leq e^{-\theta^2 pn/3}.$$

- The probability that the deviate of a **binomial random variable** from its expected value

$$E[X] = E\left[\sum_{i=1}^n x_i\right] = pn$$

decreases exponentially with the deviation.

^aHerman Chernoff (1923–). The bound is asymptotically optimal.

The Proof

- Let t be any positive real number.
- Then

$$\text{prob}[X \geq (1 + \theta)pn] = \text{prob}[e^{tX} \geq e^{t(1+\theta)pn}].$$

- Markov's inequality (p. 460) generalized to real-valued random variables says that

$$\text{prob}[e^{tX} \geq kE[e^{tX}]] \leq 1/k.$$

- With $k = e^{t(1+\theta)pn} / E[e^{tX}]$, we have

$$\text{prob}[X \geq (1 + \theta)pn] \leq e^{-t(1+\theta)pn} E[e^{tX}].$$

The Proof (continued)

- Because $X = \sum_{i=1}^n x_i$ and x_i 's are independent,

$$E[e^{tX}] = (E[e^{tx_1}])^n = [1 + p(e^t - 1)]^n.$$

- Substituting, we obtain

$$\begin{aligned} \text{prob}[X \geq (1 + \theta)pn] &\leq e^{-t(1+\theta)pn} [1 + p(e^t - 1)]^n \\ &\leq e^{-t(1+\theta)pn} e^{pn(e^t - 1)} \end{aligned}$$

as $(1 + a)^n \leq e^{an}$ for all $a > 0$.

The Proof (concluded)

- With the choice of $t = \ln(1 + \theta)$, the above becomes

$$\text{prob}[X \geq (1 + \theta)pn] \leq e^{pn[\theta - (1+\theta)\ln(1+\theta)]}.$$

- The exponent expands to $-\frac{\theta^2}{2} + \frac{\theta^3}{6} - \frac{\theta^4}{12} + \dots$ for $0 \leq \theta \leq 1$, which is less than

$$-\frac{\theta^2}{2} + \frac{\theta^3}{6} \leq \theta^2 \left(-\frac{1}{2} + \frac{\theta}{6} \right) \leq \theta^2 \left(-\frac{1}{2} + \frac{1}{6} \right) = -\frac{\theta^2}{3}.$$

Power of the Majority Rule

From $\text{prob}[X \leq (1 - \theta)pn] \leq e^{-\frac{\theta^2}{2}pn}$ (prove it):

Corollary 70 *If $p = (1/2) + \epsilon$ for some $0 \leq \epsilon \leq 1/2$, then*

$$\text{prob} \left[\sum_{i=1}^n x_i \leq n/2 \right] \leq e^{-\epsilon^2 n/2}.$$

- The textbook's corollary to Lemma 11.9 seems incorrect.
- Our original problem (p. 519) hence demands, e.g., $n \approx 1.4k/\epsilon^2$ independent coin flips to guarantee making an error with probability $\leq 2^{-k}$ with the majority rule.

BPP^a (Bounded Probabilistic Polynomial)

- The class **BPP** contains all languages L for which there is a precise polynomial-time NTM N such that:
 - If $x \in L$, then at least $3/4$ of the computation paths of N on x lead to “yes.”
 - If $x \notin L$, then at least $3/4$ of the computation paths of N on x lead to “no.”
- So N accepts or rejects by a *clear* majority.

^aGill (1977).

Magic 3/4?

- The number 3/4 bounds the probability (ratio) of a right answer away from 1/2.
- Any constant *strictly* between 1/2 and 1 can be used without affecting the class BPP.
- As with RP,

$$\frac{1}{2} + \frac{1}{q(n)}$$

for any polynomial $q(n)$ can be used in place of 3/4 (p. 514).

The Majority Vote Algorithm

Suppose L is decided by N by majority $(1/2) + \epsilon$.

```
1: for  $i = 1, 2, \dots, 2k + 1$  do  
2:   Run  $N$  on input  $x$ ;  
3: end for  
4: if “yes” is the majority answer then  
5:   “yes”;  
6: else  
7:   “no”;  
8: end if
```

Analysis

- The running time remains polynomial, being $2k + 1$ times N 's running time.
- By Corollary 70 (p. 524), the probability of a false answer is at most $e^{-\epsilon^2 k}$.
- By taking $k = \lceil 2/\epsilon^2 \rceil$, the error probability is at most $1/4$.
- Recall that ϵ can be any inverse polynomial, because k remains polynomial in n .

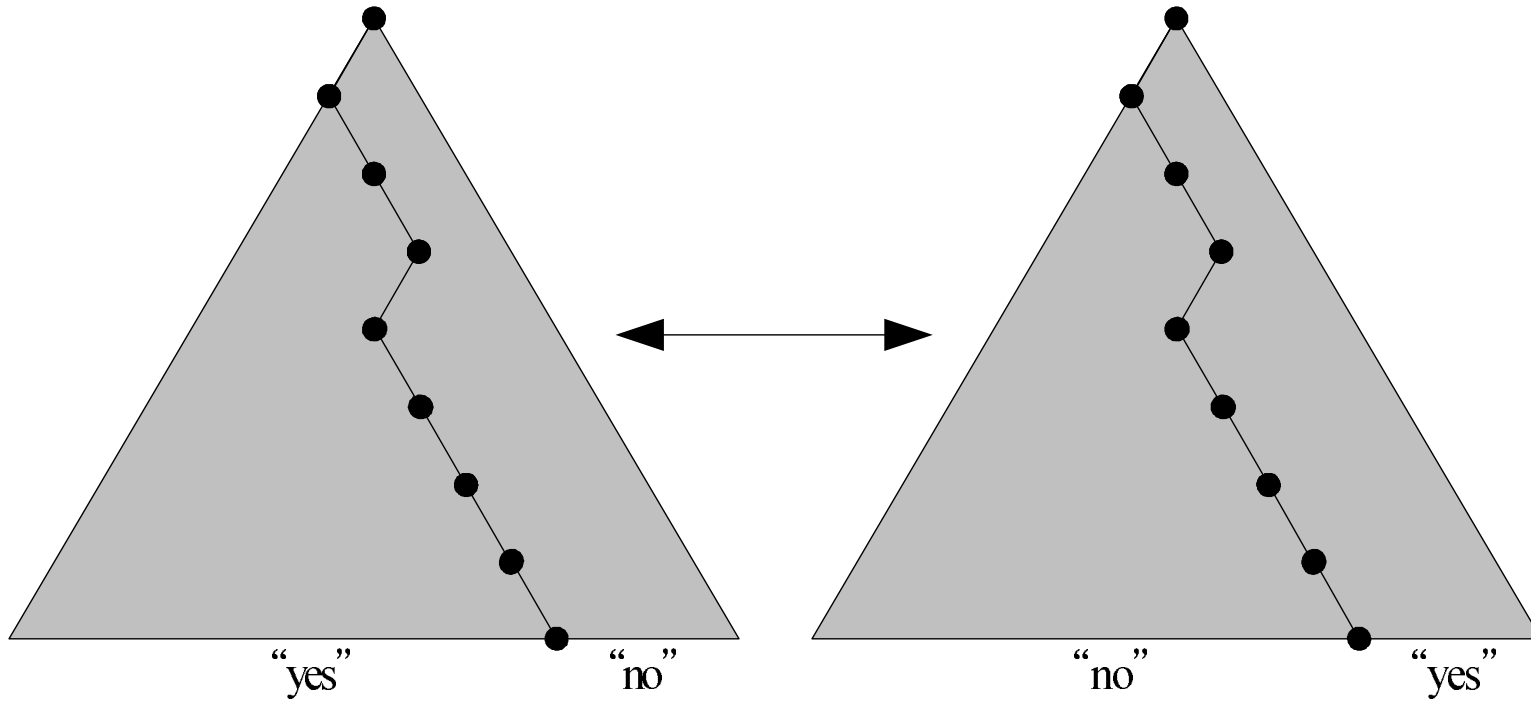
Aspects of BPP

- BPP is the most comprehensive yet plausible notion of efficient computation.
 - If a problem is in BPP, we take it to mean that the problem can be solved efficiently.
 - In this aspect, BPP has effectively replaced P.
- $(\text{RP} \cup \text{coRP}) \subseteq (\text{NP} \cup \text{coNP})$.
- $(\text{RP} \cup \text{coRP}) \subseteq \text{BPP}$.
- Whether $\text{BPP} \subseteq (\text{NP} \cup \text{coNP})$ is unknown.
- But it is unlikely that $\text{NP} \subseteq \text{BPP}$ (see p. 545).

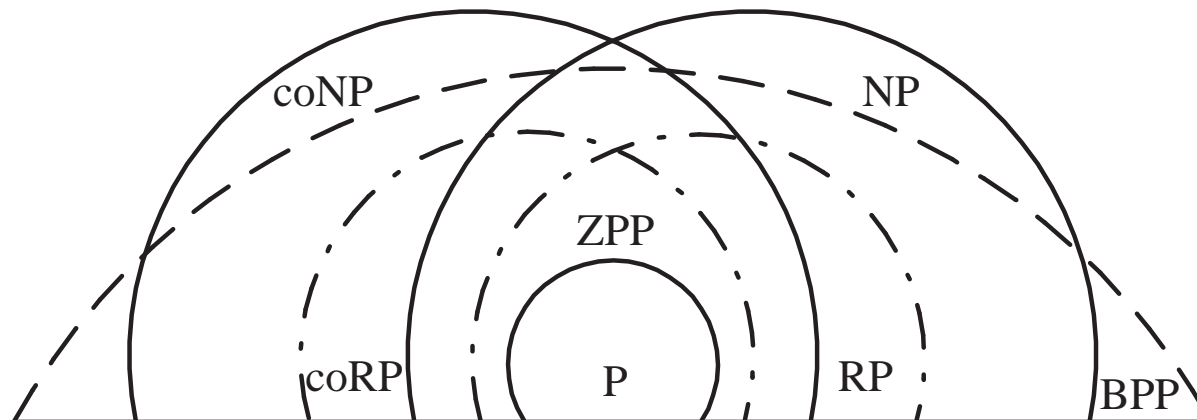
coBPP

- The definition of BPP is symmetric: acceptance by clear majority and rejection by clear majority.
- An algorithm for $L \in \text{BPP}$ becomes one for \bar{L} by reversing the answer.
- So $\bar{L} \in \text{BPP}$ and $\text{BPP} \subseteq \text{coBPP}$.
- Similarly $\text{coBPP} \subseteq \text{BPP}$.
- Hence $\text{BPP} = \text{coBPP}$.
- This approach does not work for RP.
- It did not work for NP either.

BPP and coBPP



“The Good, the Bad, and the Ugly”



Circuit Complexity

- Circuit complexity is based on boolean circuits instead of Turing machines.
- A boolean circuit with n inputs computes a boolean function of n variables.
- By identifying **true**/1 with “yes” and **false**/0 with “no,” a boolean circuit with n inputs accepts certain strings in $\{0, 1\}^n$.
- To relate circuits with an arbitrary language, we need one circuit for each possible input length n .

Formal Definitions

- The **size** of a circuit is the number of *gates* in it.
- A **family of circuits** is an infinite sequence $\mathcal{C} = (C_0, C_1, \dots)$ of boolean circuits, where C_n has n boolean inputs.
- For input $x \in \{0, 1\}^*$, $C_{|x|}$ outputs 1 if and only if $x \in L$.
- In other words,

C_n accepts $L \cap \{0, 1\}^n$.

Formal Definitions (concluded)

- $L \subseteq \{0, 1\}^*$ has **polynomial circuits** if there is a family of circuits \mathcal{C} such that:
 - The size of C_n is at most $p(n)$ for some fixed polynomial p .
 - C_n accepts $L \cap \{0, 1\}^n$.

Exponential Circuits Suffice for All Languages

- Theorem 15 (p. 173) implies that there are languages that cannot be solved by circuits of size $2^n/(2n)$.
- But exponential circuits can solve all problems.

Proposition 71 *All decision problems (decidable or otherwise) can be solved by a circuit of size 2^{n+2} .*

- We will show that for any language $L \subseteq \{0, 1\}^*$, $L \cap \{0, 1\}^n$ can be decided by a circuit of size 2^{n+2} .

The Proof (concluded)

- Define boolean function $f : \{0, 1\}^n \rightarrow \{0, 1\}$, where

$$f(x_1x_2 \cdots x_n) = \begin{cases} 1 & x_1x_2 \cdots x_n \in L, \\ 0 & x_1x_2 \cdots x_n \notin L. \end{cases}$$

- $f(x_1x_2 \cdots x_n) = (x_1 \wedge f(1x_2 \cdots x_n)) \vee (\neg x_1 \wedge f(0x_2 \cdots x_n))$.
- The circuit size $s(n)$ for $f(x_1x_2 \cdots x_n)$ hence satisfies

$$s(n) = 4 + 2s(n - 1)$$

with $s(1) = 1$.

- Solve it to obtain $s(n) = 5 \times 2^{n-1} - 4 \leq 2^{n+2}$.

The Circuit Complexity of P

Proposition 72 *All languages in P have polynomial circuits.*

- Let $L \in P$ be decided by a TM in time $p(n)$.
- By Corollary 32 (p. 264), there is a circuit with $O(p(n)^2)$ gates that accepts $L \cap \{0, 1\}^n$.
- The size of the circuit depends only on L and the length of the input.
- The size of the circuit is polynomial in n .

Polynomial Circuits vs. P

- Is the converse of Proposition 72 true?
 - Do polynomial circuits accept only languages in P?
- No.
- Polynomial circuits can accept *undecidable* languages!

Languages That Polynomial Circuits Accept (concluded)

- Let $L \subseteq \{0, 1\}^*$ be an undecidable language.
- Let $U = \{1^n : \text{the binary expansion of } n \text{ is in } L\}$.^a
 - For example, $11111_1 \in U$ if $101_2 \in L$.
- U is also undecidable.
- $U \cap \{1\}^n$ can be accepted by the trivial circuit C_n that outputs 1 if $1^n \in U$ and outputs 0 if $1^n \notin U$.^b
- The family of circuits (C_0, C_1, \dots) is polynomial in size.

^aAssume n 's leading bit is always 1 without loss of generality.

^bWe may not know which is the case for *general* n .

A Patch

- Despite the simplicity of a circuit, the previous discussions imply the following:
 - Circuits are *not* a realistic model of computation.
 - Polynomial circuits are *not* a plausible notion of efficient computation.
- What is missing?
- The *effective and efficient constructibility* of

$$C_0, C_1, \dots$$

Uniformity

- A family (C_0, C_1, \dots) of circuits is **uniform** if there is a $\log n$ -space bounded TM which on input 1^n outputs C_n .
 - Note that n is the length of the input to C_n .
 - Circuits now cannot accept undecidable languages (why?).
 - The circuit family on p. 540 is not constructible by a *single* Turing machine (algorithm).
- A language has **uniformly polynomial circuits** if there is a *uniform* family of polynomial circuits that decide it.

Uniformly Polynomial Circuits and P

Theorem 73 *$L \in P$ if and only if L has uniformly polynomial circuits.*

- One direction was proved in Proposition 72 (p. 538).
- Now suppose L has uniformly polynomial circuits.
- A TM decides $x \in L$ in polynomial time as follows:
 - Calculate $n = |x|$.
 - Generate C_n in $\log n$ space, hence polynomial time.
 - Evaluate the circuit with input x in polynomial time.
- Therefore $L \in P$.

Relation to P vs. NP

- Theorem 73 implies that $P \neq NP$ if and only if NP-complete problems have no *uniformly* polynomial circuits.
- A stronger conjecture: NP-complete problems have no polynomial circuits, *uniformly or not*.
- The above is currently the preferred approach to proving the $P \neq NP$ conjecture—without success so far.

BPP's Circuit Complexity

Theorem 74 (Adleman (1978)) *All languages in BPP have polynomial circuits.*

- Our proof will be *nonconstructive* in that only the existence of the desired circuits is shown.
 - Recall our proof of Theorem 15 (p. 173).
 - Something exists if its probability of existence is nonzero.
- It is not known how to efficiently generate circuit C_n .
- If the construction of C_n can be made efficient, then $P = BPP$, an unlikely result.

The Proof

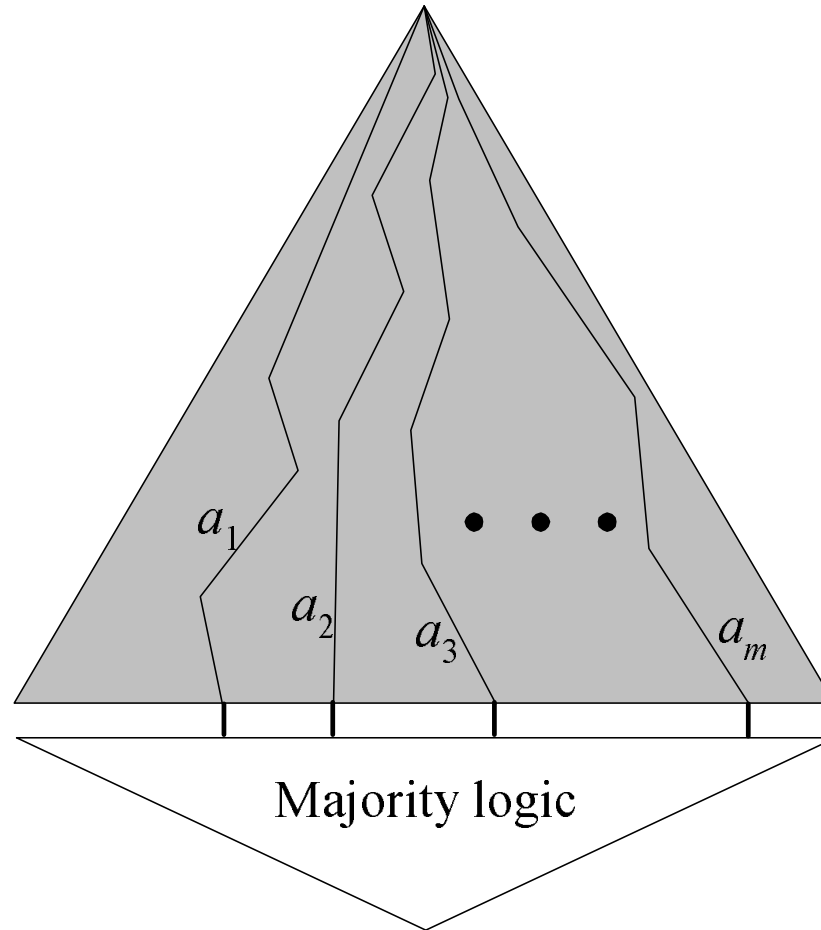
- Let $L \in \text{BPP}$ be decided by a precise NTM N by clear majority.
- We shall prove that L has polynomial circuits C_0, C_1, \dots
 - These circuits cannot make mistakes.
- Suppose N runs in time $p(n)$, where $p(n)$ is a polynomial.
- Let $A_n = \{a_1, a_2, \dots, a_m\}$, where $a_i \in \{0, 1\}^{p(n)}$.
- Each $a_i \in A_n$ represents a sequence of nondeterministic choices (i.e., a computation path) for N .
- Pick $m = 12(n + 1)$.

The Proof (continued)

- Let x be an input with $|x| = n$.
- Circuit C_n simulates N on x with each sequence of choices in A_n and then takes the majority of the m outcomes.^a
- As N with a_i is a polynomial-time deterministic TM, it can be simulated by polynomial circuits of size $O(p(n)^2)$.
 - See the proof of Proposition 72 (p. 538).
- The size of C_n is therefore $O(mp(n)^2) = O(np(n)^2)$.
 - This is a polynomial.

^aAs m is even, there may be no clear majority. Still, the probability of that happening is very small and does not materially affect our general conclusion. Thanks to a lively class discussion on December 14, 2010.

The Circuit



The Proof (continued)

- We now prove the existence of an A_n making C_n correct on *all* n -bit inputs.
- Call a_i **bad** if it leads N to a false positive or a false negative.
- Select A_n *uniformly randomly*.
- For each $x \in \{0, 1\}^n$, $1/4$ of the computations of N are erroneous.
- Because the sequences in A_n are chosen randomly and independently, the expected number of bad a_i 's is $m/4$.

The Proof (continued)

- By the Chernoff bound (p. 520), the probability that the number of bad a_i 's is $m/2$ or more is at most

$$e^{-m/12} < 2^{-(n+1)}.$$

- The error probability of using majority rule is thus $< 2^{-(n+1)}$ for each $x \in \{0, 1\}^n$.
- The probability that there is an x such that A_n results in an incorrect answer is $< 2^n 2^{-(n+1)} = 2^{-1}$.
 - $\text{prob}[A \cup B \cup \dots] \leq \text{prob}[A] + \text{prob}[B] + \dots$.
- Note that each A_n yields a circuit.

The Proof (concluded)

- We just showed that at least half of them are correct.
- So with probability ≥ 0.5 , a random A_n produces a correct C_n for *all* inputs of length n .
- Because this probability exceeds 0, an A_n that makes majority vote work for all inputs of length n exists.
- Hence a correct C_n exists.^a
- We have used the **probabilistic method**.

^aQuine (1948), “To be is to be the value of a bound variable.”

Leonard Adleman^a (1945–)



^aTuring Award (2002).

Cryptography

Whoever wishes to keep a secret
must hide the fact that he possesses one.
— Johann Wolfgang von Goethe (1749–1832)

Cryptography

- **Alice** (A) wants to send a message to **Bob** (B) over a channel monitored by **Eve** (eavesdropper).
- The protocol should be such that the message is known only to Alice and Bob.
- The art and science of keeping messages secure is **cryptography**.



Encryption and Decryption

- Alice and Bob agree on two algorithms E and D —the **encryption** and the **decryption algorithms**.
- Both E and D are known to the public in the analysis.
- Alice runs E and wants to send a message x to Bob.
- Bob operates D .
- Privacy is assured in terms of two numbers e, d , the **encryption** and **decryption keys**.
- Alice sends $y = E(e, x)$ to Bob, who then performs $D(d, y) = x$ to recover x .
- x is called **plaintext**, and y is called **ciphertext**.^a

^aBoth “zero” and “cipher” come from the same Arab word.

Some Requirements

- D should be an inverse of E given e and d .
- D and E must both run in (probabilistic) polynomial time.
- Eve should not be able to recover x from y without knowing d .
 - As D is public, d must be kept secret.
 - e may or may not be a secret.

Degrees of Security

- **Perfect secrecy:** After a ciphertext is intercepted by the enemy, the a posteriori probabilities of the plaintext that this ciphertext represents are identical to the a priori probabilities of the same plaintext before the interception.
 - The probability that plaintext \mathcal{P} occurs is independent of the ciphertext \mathcal{C} being observed.
 - So knowing \mathcal{C} yields no advantage in recovering \mathcal{P} .
- Such systems are said to be **informationally secure**.
- A system is **computationally secure** if breaking it is theoretically possible but computationally infeasible.

Conditions for Perfect Secrecy^a

- Consider a cryptosystem where:
 - The space of ciphertext is as large as that of keys.
 - Every plaintext has a nonzero probability of being used.
- It is perfectly secure if and only if the following hold.
 - A key is chosen with uniform distribution.
 - For each plaintext x and ciphertext y , there exists a unique key e such that $E(e, x) = y$.

^aShannon (1949).

The One-Time Pad^a

- 1: Alice generates a random string r as long as x ;
- 2: Alice sends r to Bob over a secret channel;
- 3: Alice sends $r \oplus x$ to Bob over a public channel;
- 4: Bob receives y ;
- 5: Bob recovers $x := y \oplus r$;

^aMauborgne and Vernam (1917); Shannon (1949). It was allegedly used for the hotline between Russia and U.S.

Analysis

- The one-time pad uses $e = d = r$.
- This is said to be a **private-key cryptosystem**.
- Knowing x and knowing r are equivalent.
- Because r is random and private, the one-time pad achieves perfect secrecy (see also p. 559).
- The random bit string must be new for each round of communication.
 - **Cryptographically strong pseudorandom generators** require exchanging only the seed once.
- The assumption of a private channel is problematic.

Public-Key Cryptography^a

- Suppose only d is private to Bob, whereas e is public knowledge.
- Bob generates the (e, d) pair and publishes e .
- Anybody like Alice can send $E(e, x)$ to Bob.
- Knowing d , Bob can recover x by $D(d, E(e, x)) = x$.
- The assumptions are complexity-theoretic.
 - It is computationally difficult to compute d from e .
 - It is computationally difficult to compute x from y without knowing d .

^aDiffie and Hellman (1976).

Whitfield Diffie (1944–)



Martin Hellman (1945–)



Complexity Issues

- Given y and x , it is easy to verify whether $E(e, x) = y$.
- Hence one can always guess an x and verify.
- Cracking a public-key cryptosystem is thus in NP.
- A necessary condition for the existence of secure public-key cryptosystems is $P \neq NP$.
- But more is needed than $P \neq NP$.
- For instance, it is not sufficient that D is hard to compute in the worst case.
- It should be hard in “most” or “average” cases.

One-Way Functions

A function f is a **one-way function** if the following hold.^a

1. f is one-to-one.
2. For all $x \in \Sigma^*$, $|x|^{1/k} \leq |f(x)| \leq |x|^k$ for some $k > 0$.
 - f is said to be **honest**.
3. f can be computed in polynomial time.
4. f^{-1} cannot be computed in polynomial time.
 - Exhaustive search works, but it is too slow.

^aDiffie and Hellman (1976); Boppana and Lagarias (1986); Grollmann and Selman (1988); Ko (1985); Ko, Long, and Du (1986); Watanabe (1985); Young (1983).

Existence of One-Way Functions

- Even if $P \neq NP$, there is no guarantee that one-way functions exist.
- No functions have been proved to be one-way.
- Is breaking glass a one-way function?

Candidates of One-Way Functions

- Modular exponentiation $f(x) = g^x \bmod p$, where g is a primitive root of p .
 - **Discrete logarithm** is hard.^a
- The RSA^b function $f(x) = x^e \bmod pq$ for an odd e relatively prime to $\phi(pq)$.
 - Breaking the RSA function is hard.

^aConjectured to be 2^{n^ϵ} for some $\epsilon > 0$ in both the worst-case sense and average sense. It is in NP in some sense (Grollmann and Selman (1988)).

^bRivest, Shamir, and Adleman (1978).

Candidates of One-Way Functions (concluded)

- Modular squaring $f(x) = x^2 \bmod pq$.
 - Determining if a number with a Jacobi symbol 1 is a quadratic residue is hard—the **quadratic residuacity assumption (QRA)**.^a

^aDue to Gauss.

The RSA Function

- Let p, q be two distinct primes.
- The RSA function is $x^e \bmod pq$ for an odd e relatively prime to $\phi(pq)$.
 - By Lemma 52 (p. 404),

$$\phi(pq) = pq \left(1 - \frac{1}{p}\right) \left(1 - \frac{1}{q}\right) = pq - p - q + 1. \quad (8)$$

- As $\gcd(e, \phi(pq)) = 1$, there is a d such that

$$ed \equiv 1 \pmod{\phi(pq)},$$

which can be found by the Euclidean algorithm.

A Public-Key Cryptosystem Based on RSA

- Bob generates p and q .
- Bob publishes pq and the encryption key e , a number relatively prime to $\phi(pq)$.
 - The encryption function is $y = x^e \bmod pq$.
 - Bob calculates $\phi(pq)$ by Eq. (8) (p. 570).
 - Bob then calculates d such that $ed = 1 + k\phi(pq)$ for some $k \in \mathbb{Z}$.
- The decryption function is $y^d \bmod pq$.
- It works because $y^d = x^{ed} = x^{1+k\phi(pq)} = x \bmod pq$ by the Fermat-Euler theorem when $\gcd(x, pq) = 1$ (p. 414).

The “Security” of the RSA Function

- Factoring pq or calculating d from (e, pq) seems hard.
 - See also p. 410.
- Breaking the last bit of RSA is as hard as breaking the RSA.^a
- Recommended RSA key sizes:^b
 - 1024 bits up to 2010.
 - 2048 bits up to 2030.
 - 3072 bits up to 2031 and beyond.

^aAlexi, Chor, Goldreich, and Schnorr (1988).

^bRSA (2003).

The “Security” of the RSA Function (concluded)

- Recall that problem A is “harder than” problem B if solving A results in solving B.
 - Factorization is “harder than” breaking the RSA.
 - Calculating Euler’s phi function is “harder than” breaking the RSA.
 - Factorization is “harder than” calculating Euler’s phi function (see Lemma 52 on p. 404).
 - So factorization is harder than calculating Euler’s phi function, which is harder than breaking the RSA.
- Factorization cannot be NP-hard unless $NP = coNP$.^a
- So breaking the RSA is unlikely to imply $P = NP$.

^aBrassard (1979).

Adi Shamir, Ron Rivest, and Leonard Adleman



Ron Rivest^a (1947–)



^aTuring Award (2002).

Adi Shamir^a (1952–)



^aTuring Award (2002).