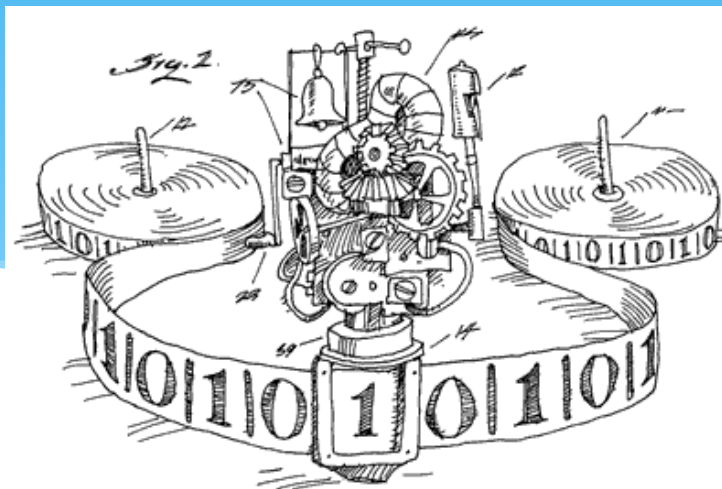


EECS 376: Foundations of Computer Science

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Complexity: Agenda

- 1) What “resources” (time, memory, randomness, ...) are needed to solve a problem on a computer?
- 2) Which problems can be solved **efficiently** on a computer?
- 3) *What does “**efficient**” even mean?*
- 4) The hardest way to make \$1 Million: P vs. NP.

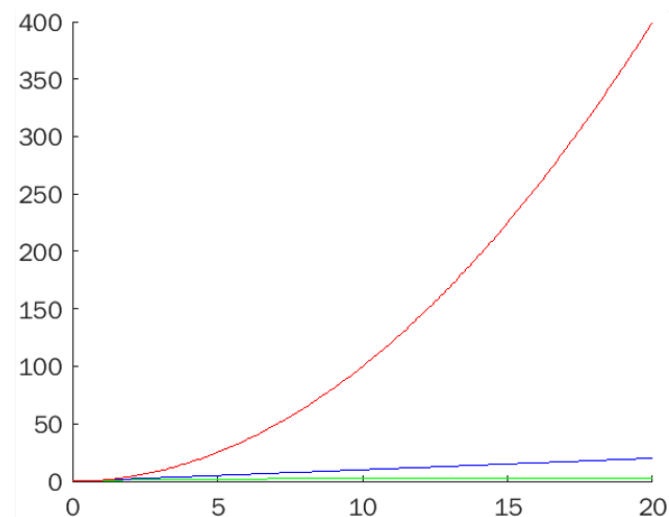
Why Polynomial?

- * “For practical purposes the difference between algebraic [polynomial] and exponential order is often more crucial than the difference between finite and non-finite.”

— Jack Edmonds
(defined complexity class P, 1965)

Review: Running Time

- * We measure “efficiency” of an algorithm by how its (*worst-case*) *runtime* scales with the input “size”
- * We express this growth asymptotically: e.g., $O(\log n)$, $O(n)$, $O(2^n)$, ... where n denotes the input size.
- * Size = # bits to write/represent the input on a computer.



Efficient \equiv “polynomial runtime in input size”;
Very robust definition (as we’ll see)

Polynomial-Time Algorithms

We have seen polynomial-time algorithms for various problems:

GCD(x,y)	Euclid's algorithm ($O(\log(x + y))$ divisions)
Longest Increasing Subseq.	Dynamic Prog. (time $O(n^2)$)
Longest Common Subseq.	Dynamic Prog. (time $O(n^2)$)
All-Pairs Shortest Paths	Dynamic Prog. (time $O(n^3)$)
Minimum Spanning Tree	Greedy (time $O(n \log n)$)

There are several others: Matching, Max-flow/min-cut,

Primality testing [Agarwal, Kayal, Saxena 2002]: Time $O(\log^6 N)$

Based on many clever and ingenious ideas.
Several algorithms still to be discovered!

A Poly-Time Algorithm for Every (Solvable) Problem?

- **Longest Path:** Given G and vertices (s,t) , find a *longest $s \rightarrow t$ path* (w/o cycle).
- **Hamiltonian Cycle:** Given G , find a *cycle that visits each vertex exactly once*.
- **Independent Set:** Given G , find a *largest subset of vertices that have no edges among them*.
- **Subset Sum:** Given $a_1, a_2, \dots, a_n \in \mathbb{Z}$ and target t , find a *subset that sums to t*
 - (what about the DP solution we saw in the class?)
- ...

Most believe that none of these problems has a poly-time algorithm.
We know that if any one of them does, then all of them do!

Theory of “P vs. NP”: a crown jewel of Computer Science—
with a million dollar reward!

The Class P

- * **Definition:** \mathbf{P} = the set of all languages that can be decided by a TM in (some) polynomial time in the input size.
- * In other words: $L \in \mathbf{P}$ if L is *efficiently decidable*, i.e., it has a polynomial-time decision algorithm.
- * **Formally:** $\mathbf{P} = \bigcup_{k \geq 1} DTIME(n^k)$
- * **Properties:**
 - * Model agnostic: can replace TM with any “realistic” (deterministic) model.
 - * Composition: if an efficient M has an oracle, which is instantiated by an efficient M' , then the resulting program is also efficient.
 - * *Proof idea*: $(n^k)^{k'} = n^{k \cdot k'}$ is also polynomial (for constants k, k').

Search vs. Decision Problems

Definition: P = the set of all languages that can be decided by a TM in (some) polynomial time in the input size.

We are (still) talking about **decision** problems, with “yes/no” answers.

What about $\text{gcd}(x,y)$? Shortest path $s \rightarrow t$? $\text{LIS}(x)$? Etc...
These don't have yes/no answers...

Not a problem! Can *recast* them as decision problems, and solve the original search problems with only polynomial “slowdown.”

Recasting Search as Decision: GCD

- * **Problem:** Given integers x, y, z , is $\gcd(x, y) \geq z$?
 - * $L_{GCD} = \left\{ (x, y, z) : x, y, z \in \mathbb{N} \text{ and } z \geq \gcd(x, y) \right\}$
- * **Claim:** L_{GCD} is efficiently decidable, i.e., $L_{GCD} \in \mathbf{P}$.
- * **Proof:** Run Euclid's algorithm and answer accordingly.
- * **Claim:** Using any L_{GCD} -decider, we can efficiently compute $\gcd(x, y)$ itself.
- * **Proof:** binary search on $z \in \{1 \dots \min(x, y)\}$, asking decider if $\gcd(x, y) \geq z$.
Runtime: $\log \min(x, y) = O(\text{input size})$ calls to decider.

Recasting: Shortest Paths

- * **Problem:** Given an (integer-)weighted graph G , vertices s and t , and $z \in \mathbb{Z}$, is there an $s \rightarrow t$ path of length at most z ?
 - * $L_{path} = \{ \langle G, s, t, z \rangle : G \text{ has an } s \rightarrow t \text{ path of length } \leq z \}$
- * **Claim:** L_{path} is efficiently decidable, i.e., $L_{path} \in \mathbf{P}$.
- * **Proof:** run Floyd-Warshall APSP and answer accordingly.
- * **Claim:** using any L_{path} -decider, we can compute the length of a shortest $s \rightarrow t$ path (with no negative-weight cycles).
- * **Proof:** binary search on $z \in \{-Z, \dots, Z\}$ where $Z = \sum_{(u,v) \in E} |w(u, v)|$.

Recasting Search to Decision In General

For a search problem with a **numerical output** A , recast it as a language with an **extra numerical input** z that asks: is $z \geq A$?

Using any decider, can solve the search problem via binary search.
Only $O(\log(\text{RANGE}))$ calls to the decider.

For a search problem with an **array output** A , recast it as a language with **extra inputs** z, i that asks: is $z \geq A[i]$?

Using any decider, can solve the search problem via binary search for each index i .

Similarly for **string outputs**, **set outputs**, ...

Introducing: The Class NP

Common misconception: NP does not stand for “Non/Not Polynomial” !

NP means “**Nondeterministic** Polynomial time.”

We won't cover nondeterminism in this course.
Instead, we'll use a conceptually simpler definition of NP.

“A better name would have been **VP: verifiable** in polynomial time.”

-Clyde Kruskal

Quote of The Day

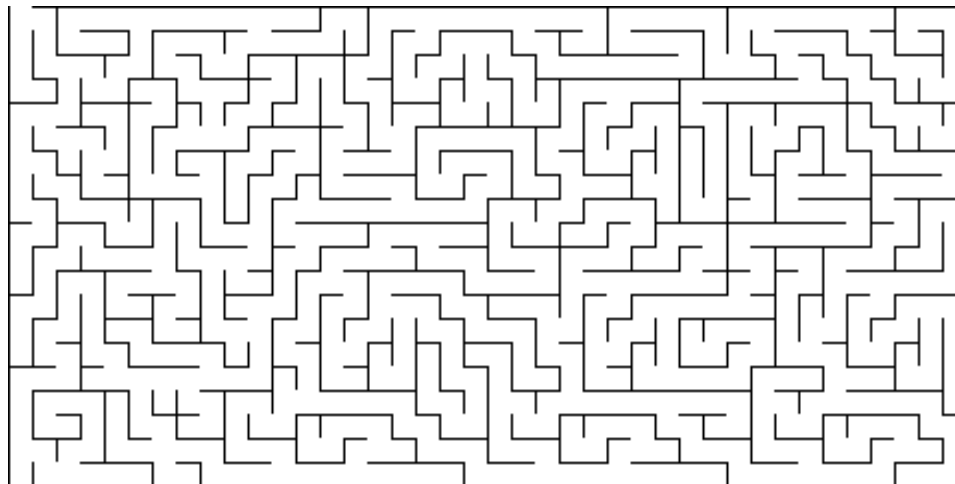
“Doveryai, no proveryai”

(Trust, but verify)

— Old Russian Proverb,
Used by Ronald Reagan

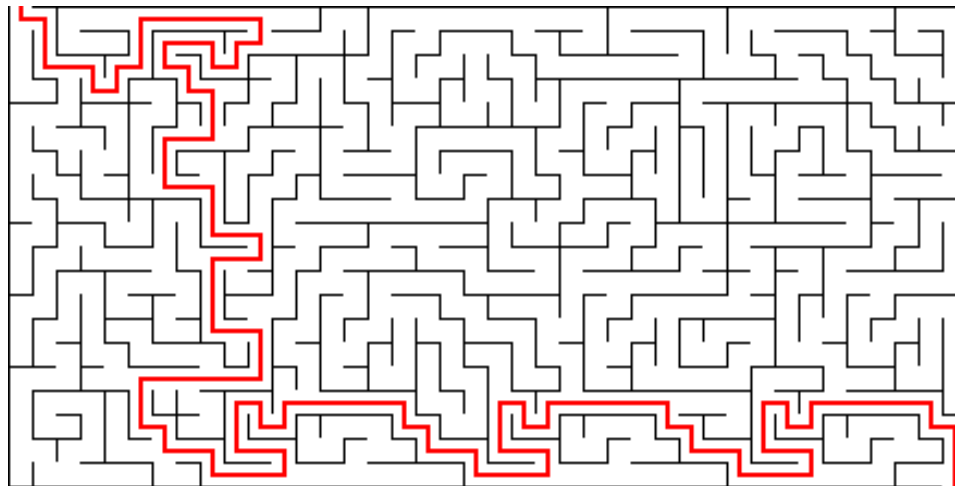
Verifiable Problems

- * **Example 1:** Given a maze, is there a route through the maze from the start to finish?
- * **Answer:** Yes.
- * **Response:** We're not convinced—you could be lying!



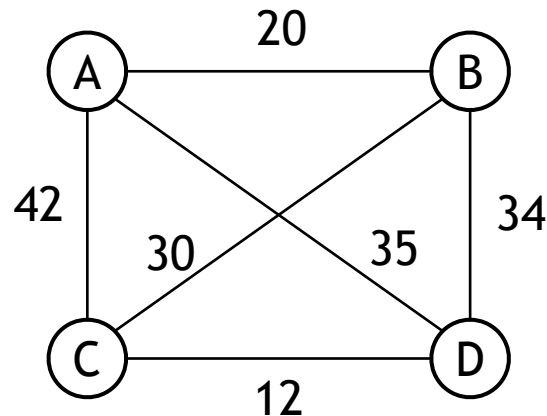
Verifiable Problems

- * **Example 1:** Given a maze, is there a route through the maze from the start to finish?
- * **Answer:** Yes; see the route below.
- * **Response:** We follow the route and are convinced.



Verifiable Problems

- * **Example 2: TSP (decision version):**
Given 4 cities and pair-wise distances between them, is there a cycle of length at most 100 through all the cities?
- * **Answer:** Yes; $A \rightarrow B \rightarrow C \rightarrow D \rightarrow A$.
- * **Response:** We check that this visits all the cities in a cycle, compute the cost $20+30+12+35 = 97 < 100$, and are convinced.



Verifiable Problems

- * **Example 3: Subset Sum**
- * Given integers a_1, a_2, \dots, a_n and target t , is there a subset of the a_i that sums to t ?
- * **Answer:** Yes; here are indices i of some a_i that do.
- * **Response:** We check that all given i are distinct, that the specified a_i sum to t , and are convinced.

Efficiently Verifiable Problems

- * **Intuition (informal):** A decision problem is *efficiently verifiable* if:
 1. When the answer is “yes”: we can be efficiently convinced given some additional information: “certificate”, “proof”, “witness”.
 2. When the answer is “no”: no additional information—even maliciously generated—could fool us into thinking that the answer is yes.
- * **Asymmetric:** we never wish to be convinced that the answer is “no”.
- * **TSP:** If there is a cheap enough tour, we can be convinced (give one). If not, even a “malicious adversary” cannot fool us.
- * **Subset Sum:** if some subset sums to t , we can be convinced (give one). If not, even a “malicious adversary” cannot fool us.

The Class NP

- * **Definition:** A decision problem L is *efficiently verifiable* if there exists an algorithm $V(x, c)$, called a *verifier*, satisfying:
1. $V(x, c)$ is efficient with respect to x , i.e., polynomial time in $|x|$.
 2. For every $x \in L$, there exists some c such that $V(x, c)$ accepts.
 - (When the answer is “yes”, we can be efficiently convinced given some certificate/witness/proof.)
 3. For every $x \notin L$, $V(x, c)$ rejects for all c .
 - (When the answer is “no”, we won’t be fooled by any [bogus] proof.)

Definition: the class **NP** = set of all efficiently verifiable languages.
I.e., $L \in \mathbf{NP}$ if L is efficiently verifiable.

Example: TSP

- * **TSP:** Given n cities and pair-wise distances, is there a cycle that visits every city exactly once, and has length at most k ?
- * $\text{TSP} = \{(G, k) : G \text{ is a weighted graph w/ a } \textit{tour} \text{ of length } \leq k\}$
- * **Claim:** TSP is efficiently verifiable, i.e., $\text{TSP} \in \text{NP}$.

Consider certificate c in the form of a sequence of vertices (which is supposed to be a tour of length $\leq k$, if one exists)

```
bool verifyTSP(graph  $G = (V, E)$ , int  $k$ ,  
               certificate  $c = (v_1, \dots, v_m)$ ):  
1. if  $c$  is not a permutation of  $V$ : reject  
2. return  $\text{length}(v_1, \dots, v_m, v_1) \leq k$ 
```

Correctness Analysis of `verifyTSP`

```
bool verifyTSP(graph  $G = (V, E)$ , int  $k$ ,  
               certificate  $c = (v_1, \dots, v_m)$ ):  
  1. if  $c$  is not a permutation of  $V$ : reject  
  2. return  $\text{length}(v_1, \dots, v_m, v_1) \leq k$ 
```

- * **Case 1:** If $(G, k) \in \text{TSP}$, then some certificate c makes `verifyTSP` $((G, k), c)$ return *true*.
- * Let $c = (v_1, \dots, v_m)$ be the vertices of G , ordered according to some tour of length $\leq k$ (which exists by hypothesis).
 - * Since c is a permutation of V , line 1 does not reject.
- * Then, $\text{length}(v_1, \dots, v_m, v_1) \leq k$ by assumption and `verifyTSP` (x, c) returns *true* on line 2, as desired.

Correctness Analysis of `verifyTSP`

```
bool verifyTSP(graph  $G = (V, E)$ , int  $k$ ,  
               certificate  $c = (v_1, \dots, v_m)$ ):  
  1. if  $c$  is not a permutation of  $V$ : reject  
  2. return  $\text{length}(v_1, \dots, v_m, v_1) \leq k$ 
```

- * **Case 2:** If $(G, k) \notin \text{TSP}$, then `verifyTSP`(x, c) rejects for any certificate c .
 - * If c is not a permutation of V , then line 1 rejects, as desired.
 - * Now suppose $c = (v_1, v_2, \dots, v_m)$ is a permutation of V .
 - * By assumption, G has no tour of length $\leq k$.
 - * Therefore, $\text{length}(v_1, \dots, v_m, v_1) > k$.
 - * Thus, line 2 rejects, as desired.

Runtime Analysis of `verifyTSP`

```
bool verifyTSP(graph  $G = (V, E)$ , int  $k$ ,  
               certificate  $c = (v_1, \dots, v_m)$ ):  
1. if  $c$  is not a permutation of  $V$ : reject  
2. return  $\text{length}(v_1, \dots, v_m, v_1) \leq k$ 
```

- * Suppose the input (G, k) has (bit) length n .
- * There are at most n vertices in G .
- * Line 1 takes $O(n)$ time: use a boolean array to check if each vertex appears exactly once in c .
- * Line 2 takes $O(n)$ time: sum the weights of the edges on the tour.
- * Therefore, the runtime is polynomial in n , as desired.

Practice with Verifiers

$L_{Comp} = \{n : n \text{ is composite (not prime)}\}$

$L_{HAM} = \{G : G \text{ has a Hamiltonian cycle}\}$

$L_{Primes} = \{n : n \text{ is prime}\}$ (complement of L_{Comp})

Not obvious. But there is a clever verifier! (next slide)

$L_{non-HAM} = \{G : G \text{ has no Hamiltonian cycle}\}$

We **do not expect** the problem to have an efficient verifier!
(Would have very surprising consequences.)

Aside: Certificate for Primes

$L_{Primes} = \{n : n \text{ is prime}\}$

Theorem (Pratt 75): L_{Primes} is efficiently verifiable.

Proof: Key property of primes: n is prime if there is an x such that $x^{n-1} = 1 \pmod{n}$ but $x^e \neq 1 \pmod{n}$ for all $e = (n-1)/p_i$,

where p_i are the prime divisors of $n-1$.

Certificate: $(x, p_1, \dots, p_k, \text{certificate that each } p_i \text{ is prime})$

By induction: Certificate has size $O(\log n)$.

P vs NP

- * $L \in \mathbf{P}$ if there is a polynomial-time (in $|x|$) algorithm M where:
 - * $x \in L \implies M(x)$ accepts
 - * $x \notin L \implies M(x)$ rejects
- * $L \in \mathbf{NP}$ if there is a polynomial-time (in $|x|$) algorithm V where:
 - * $x \in L \implies V(x, c)$ accepts for some c
 - * $x \notin L \implies V(x, c)$ rejects for every c
- * **Observe:** $\mathbf{P} \subseteq \mathbf{NP}$ ($V(x, c)$ can ignore c and just run $M(x)$.)
- * **\$1,000,000 question:** Is $\mathbf{P} = \mathbf{NP}$?
Is every efficiently verifiable problem
also efficiently solvable?
Seems not... but we don't know for sure!

P vs NP

- * ... Let $p(n)$ be the number of steps to find a proof of length n . The question is, how rapidly does $p(n)$ grow for an optimal machine? It is possible to show that $p(n) > Kn$. If there really were a machine with $p(n) \sim Kn$ (or even just $\sim Kn^2$) then that would have consequences of the greatest significance. Namely, this would clearly mean that the thinking of a mathematician in the case of yes-or-no questions could be completely replaced by machine ...
 - Kurt Godel's letter to von Neumann in 1956 (15 years before P vs NP was formalized!)
- * "If $P = NP$, then the world would be a profoundly different place than we usually assume it to be. There would be no special value in 'creative leaps', no fundamental gap between solving a problem and recognizing the solution once it's found."
 - Scott Aaronson

The Major Open Problem of Computer Science

