

Announcements

- Problem Set 7 graded; will be returned at end of lecture.
- Unclaimed problem sets and midterms moved!
 - Now in cabinets in the Gates open area near the drop-off box.

Previously on CS103...

The Complexity Class ${\bf P}$

- The complexity class **P** (**p**olynomial time) contains all problems that can be solved in polynomial time.
- Formally:

$$\mathbf{P} = \bigcup_{k=0}^{\infty} \mathsf{TIME}(n^k)$$

• The **Cobham-Edmonds Thesis**: A decision problem can be solved efficiently iff it is in **P**.

Examples of Problems in ${\bf P}$

- All regular languages are in **P**.
 - Belong to TIME(*n*).
- All DCFLs are in **P**.
 - Belong to $TIME(n^2)$.
- All CFLs are in **P**.
 - Belong to $TIME(n^{18})$.
- Many other problems are in **P**:
 - POWER2
 - SEARCH

Regular Languages DCFLs CFLs P

Undecidable Languages

R

Proving Languages are in P

- Directly prove the language is in P.
 - Build a decider for the language *L*.
 - Prove that the decider runs in time $O(n^k)$.
- Use closure properties.
 - Prove that the language can be formed by appropriate transformations of languages in ${\bf P}.$

• Reduce the language to a language in P.

• Show how a polynomial-time decider for some language L' can be used to decide L.

- Let $A \subseteq \Sigma_1^*$ and $B \subseteq \Sigma_2^*$ be languages.
- A **polynomial-time reduction** is a function $f: \Sigma_1^* \to \Sigma_2^*$ with the following properties:
 - f(w) can be computed in polynomial time.
 - $w \in A$ iff $f(w) \in B$.
- Notation: $A \leq_{P} B$.
- Informally:
 - A way of turning inputs to A into inputs to B
 - that can be computed in polynomial time
 - that preserves the correct answer.

- Suppose that we know that $B \in \mathbf{P}$.
- Suppose that $A \leq_{P} B$.





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- Suppose that we know that $B \in \mathbf{P}$.
- Suppose that $A \leq_{P} B$.
- Then $A \in \mathbf{P}$ as well.



Theorem: If $B \in \mathbf{P}$ and $A \leq_{P} B$, then $A \in \mathbf{P}$.

Proof: Let H be a polynomial-time decider for *B*. Consider the following TM:

 $\begin{aligned} M &= \text{``On input } w: \\ & \text{Compute } f(w). \\ & \text{Run } H \text{ on } f(w). \\ & \text{If } H \text{ accepts, accept; if } H \text{ rejects, reject.''} \end{aligned}$

We claim that *M* is a polynomial-time decider for *A*. To see this, we prove that *M* is a polynomial-time decider, then that $\mathscr{L}(M) = A$. To see that *M* is a polynomial-time decider, note that because *f* is a polynomial-time reduction, computing f(w) takes time $O(n^k)$ for some *k*. Moreover, because computing f(w) takes time $O(n^k)$, we know that $|f(w)| = O(n^k)$. *M* then runs *H* on f(w). Since *H* is a polynomial-time decider, *H* halts in $O(m^r)$ on an input of size *m* for some *r*. Since $|f(w)| = O(n^k)$, *H* halts after $O(|f(w)|^r) = O(n^{kr})$ steps. Thus *M* halts after $O(n^k + n^{kr})$ steps, so *M* is a polynomial-time decider.

To see that $\mathscr{L}(M) = A$, note that M accepts w iff H accepts f(w) iff $f(w) \in A$. Since f is a polynomial-time reduction, $f(w) \in B$ iff $w \in A$. Thus M accepts w iff $w \in A$, so $\mathscr{L}(M) = A$.

A Sample Reduction

- Given an undirected graph *G*, a **matching** in *G* is a set of edges such that no two edges share an endpoint.
- A **maximum matching** is a matching with the largest number of edges.

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Maximum matchings are not necessarily unique.

- Jack Edmonds' paper "Paths, Trees, and Flowers" that describes a polynomialtime algorithm for finding maximum matchings.
 - (This is the same Edmonds as in "Cobham-Edmonds Thesis.)
- Using this fact, what other problems can we solve?











A Domino Tiling Reduction

• Let *MATCHING* be the language defined as follows:

 $MATCHING = \{ \langle G, k \rangle \mid G \text{ is an undirected graph}$ with a matching of size at least $k \}$

- **Theorem** (Edmonds): $MATCHING \in \mathbf{P}$.
- Let *DOMINO* be this language:

DOMINO = { $\langle D, k \rangle$ | D is a grid and k nonoverlapping dominoes can be placed on D. }

• We'll prove $DOMINO \leq_{P} MATCHING$ to show that $DOMINO \in \mathbf{P}$.

Solving Domino Tiling














Our Reduction

- Given as input (D, k), construct the graph G as follows:
 - For each empty cell x_i , construct a node v_i .
 - For each pair of adjacent empty cells x_i and x_j, construct an edge (v_i, v_i)



• Let $f(\langle D, k \rangle) = \langle G, k \rangle$.

A Polynomial-Time Reduction

- To prove that f is a polynomial-time reduction, we will show that the size of f(w) is a polynomial in the size of w.
 - Technically, this is **not** sufficient to prove that *f* runs in polynomial time.
 - However, most reductions that construct a polynomially-large object take polynomial time.
 - We will gloss over the fact that the polynomialsize object can be constructed in polynomial time; barring very unusual reductions, this is almost always true.

A Polynomial-Time Reduction

- Given a grid D and a number k, how large is the constructed graph G?
 - One node per empty cell in *D*.
 - One edge per pair of adjacent empty cells in *D*.
- There are O(|D|) empty cells in D.
- Each empty cell may have up to four neighbors.
- So there are at most O(|D|) constructed edges.
- Each node and edge can be built in polynomial time, so the overall reduction takes polynomial time.

Lemma: *f* is computable in polynomial time.

Proof: We show that $f(\langle D, k \rangle) = \langle G, k \rangle$ has size that is a polynomial in the size of $\langle D, k \rangle$.

For each empty cell x_i in D, we construct a single node v_i in G. Since there are O(|D|) cells, there are O(|D|) nodes in the graph. For each pair of adjacent, empty cells x_i and x_j in D, we add the edge (x_i, x_j) . Since each cell in D has four neighbors, the maximum number of edges we could add this way is O(|D|) as well. Thus the total size of the graph G is O(|D|). Consequently, the total size of $\langle G, k \rangle$ is O(|D| + |k|), which is a polynomial in the size of the input.

Since each part of the graph could be constructed in polynomial time, the overall graph can be constructed in polynomial time.

Summary of ${\bf P}$

- **P** is the complexity class of yes/no questions that can be solved in polynomial time.
- **P** is closed under polynomial-time reductions.

What *can*'t you do in polynomial time?



How many simple paths are there from the start node to the end node? How many subsets of this set are there?

7

7

1 2 3 4 5 6 7 8

How many binary search trees can you form from these numbers?

An Interesting Observation

- There are (at least) exponentially many objects of each of the preceding types.
- However, each of those objects is not very large.
 - Each simple path has length no longer than the number of nodes in the graph.
 - Each subset of a set has no more elements than the original set.
 - Each binary search tree made from some elements has exactly one node per element.
- This brings us to our next topic...



NTMs

- A **nondeterministic Turing machine** (NTM) is a generalization of the Turing machine.
- An NTM may have multiple transitions defined for a given state/symbol combination.
- The NTM accepts if **any** choice of transitions enters an accepting state.
- The NTM rejects if **all** choices of transitions enter a rejecting state.
- Otherwise, the NTM loops.

Nondeterminism Revisited

- If we add nondeterminism to the DFA, we get the NFA.
 - NFAs are no more powerful than DFAs.
- If we add nondeterminism to the DPDA, we get the PDA.
 - PDAs are more powerful than DPDAs.
- Adding nondeterminism to a TM produces the equivalently powerful NTM.
 - NTMs are no more powerful than TMs.

Nondeterminism Revisited

- Converting an NFA to a DFA might introduce exponentially more space.
- It is sometimes impossible to convert an NPDA to a DPDA.
- Converting an NTM to a TM might dramatically slow down the TM.

Designing NTMs

- Nondeterminism is a **very** powerful tool for solving problems.
- Many problems can be solved simply with nondeterminism using the following template:
 - **Nondeterministically** guess some important piece of information.
 - **Deterministically** check that the guess was correct.

- Recall: a number n > 1 is composite if it is not prime.
- Let $\Sigma = \{ 1 \}$ and consider the language

 $COMPOSITE = \{ \mathbf{1}^n \mid n \text{ is composite } \}$

- We can build a multitape, nondeterministic TM for COMPOSITE as follows:
- $M = "On input 1^n:$
 - **Nondeterministically** write out *q* **1**s on a second tape $(2 \le q < n)$
 - Nondeterministically write out $r \mathbf{1}s$ on a third tape $(2 \le r < n)$
 - **Deterministically** check if qr = n.
 - If so, accept.
 - Otherwise, reject"











Guess q and r (Nondeterministic) Compute qr (Deterministic) Check if n = qr (Deterministic)















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1

1

1

1

1

Guess q and r (Nondeterministic) Compute qr (Deterministic) Check if n = qr (Deterministic)



. . .






























































































Nondeterministic Algorithms Guess q and r (Nondeterministic) . . . **Compute qr** (Deterministic) Check if n = qr (Deterministic)


Nondeterministic Algorithms



Nondeterministic Algorithms



Nondeterministic Algorithms











- When discussing deterministic TMs, the notion of time complexity is (reasonably) straightforward.
- **Recall:** One way of thinking about nondeterminism is as a tree.
- In a **deterministic** computation, the tree is a straight line.
- The time complexity is the height of that straight line.



- When discussing deterministic TMs, the notion of time complexity is (reasonably) straightforward.
- **Recall:** One way of thinking about nondeterminism is as a tree.
- The time complexity is the height of the tree (the length of the **longest** possible choice we could make).

- $M = "On input 1^n:$
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 - Nondeterministically write out r 1son a third tape $(2 \le r < n)$
 - **Deterministically** check if qr = n.
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O(*n*) steps

O(*n*) steps

- O(n²) steps
- + O(1) steps

```
O(n<sup>2</sup>) steps
```

- Our multitape NTM can decide COMPOSITEin time $O(n^2)$.
- Using a similar construction to the deterministic case, a single-tape NTM can decide COMPOSITE in $O(n^4)$.
- The best known deterministic algorithm for deciding *COMPOSITE* runs *much* more slowly.
 - Runs in time around $O(n^8)$.
- Just how much more powerful are NTMs?

From NTMs to TMs

- NTMs are at least as powerful as TMs.
 - Just don't use any nondeterminism!
- TMs are at least as powerful as NTMs.
 - Idea: Simulate the NTM with a multitape TM.
 - Run a breadth-first search on possible options.



From NTMs to TMs



From NTMs to TMs

- **Theorem**: For any NTM with time complexity f(n), there is a TM with time complexity $2^{O(f(n))}$.
- It is unknown whether it is possible to do any better than this in the general case.
- NTMs are capable of exploring multiple options in parallel; this "seems" inherently faster than deterministic computation.

TIME and NTIME

- **Recall:** TIME(*f*(*n*)) is the class of languages that can be decided in O(*f*(*n*)) time by a single-tape TM.
- **NTIME**(f(n)) is the class of languages that can be decided in O(f(n)) time by a single-tape NTM.
 - All possible options terminate in O(f(n)) steps.
- For any f(n), TIME $(f(n)) \subseteq$ NTIME(f(n)).
 - Can always convert a TM to an NTM.

The Complexity Class $\ensuremath{\mathbf{NP}}$

- The complexity class NP (nondeterministic polynomial time) contains all problems that can be solved in polynomial time by a single-tape NTM.
- Formally:

$$NP = \bigcup_{k=0}^{\infty} NTIME(n^k)$$

- What types of problems are in $\mathbf{NP}?$

- Does a Sudoku grid have a solution?
 - $M = "On input \langle S \rangle$, an encoding of a Sudoku puzzle:
 - **Nondeterministically** guess how to fill in all the squares.
 - **Deterministically** check whether the guess is correct.
 - If so, accept; if not, reject."

		7		6		1		
					3		5	2
3			1		5	9		7
6		5		3		8		9
	1						2	
8		2		1		5		4
1		3	2		7			8
5	7		4					
		4		8		7		

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2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
3	8	6	1	2	5	9	4	7
6	4	5	7	3	2	8	1	9
7	1	9	5	4	8	3	2	6
8	3	2	6	1	9	5	7	4
1	6	3	2	5	7	4	9	8
5	7	8	4	9	6	2	3	1
9	2	4	3	8	1	7	6	5

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If we allow for a generalized Sudoku board of arbitrary size:

There are polynomially many grid cells to fill in.

Checking the grid takes polynomial time.

Overall algorithm takes polynomial time.

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
3	8	6	1	2	5	9	4	7
6	4	5	7	3	2	8	1	9
7	1	9	5	4	8	3	2	6
8	3	2	6	1	9	5	7	4
1	6	3	2	5	7	4	9	8
5	7	8	4	9	6	2	3	1
9	2	4	3	8	1	7	6	5

- A **graph coloring** is a way of assigning colors to nodes in an undirected graph such that no two nodes joined by an edge have the same color.
 - Applications in compilers, cell phone towers, etc.
- Question: Can graph *G* be colored with at most *k* colors?



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- Question: Can graph *G* be colored with at most *k* colors?
- M ="On input $\langle G, k \rangle$:
 - Nondeterministically guess a *k*-coloring of the nodes of *G*.
 - **Deterministically** check whether it is legal.
 - If so, accept; if not, reject."



- Suppose you want to start a delivery service. You want to place depots such that each customer is within some distance of the depot.
- Given a set of candidate locations for depots, can you place k depots and guarantee that each customer is covered?



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- Given a set of candidate locations for depots, can you place k depots and guarantee that each customer is covered?
- M = "On input $\langle D, C, \delta, k \rangle$ (depot locations, customer locations, minimum distance required, and number of depots desired):
 - **Nondeterministically** guess *k* depots from *D*.
 - **Deterministically** verify each $c \in C$ is within δ distance of some depot.
 - If so, accept; otherwise reject."

A General Pattern

- The NTMs we have seen so far always follow this pattern:
 - M = "On input w:
 - **Nondeterministically** guess some string *x*.
 - **Deterministically** check whether *x* solves *w*.
 - If so, accept; otherwise, reject."
- Is there a different way of characterizing NP?

Polynomial-Time Verifiers

- A **polynomial-time verifier** is a deterministic TM of the form
 - M ="On input $\langle w, x \rangle$:
 - **Deterministically** check whether *x* solves *w*.

- If so, accept; otherwise, reject."

such that M runs in time polynomial in the length of w (not the length of x).

• The string x is called a **certificate** or a **witness** for *w*.

An Efficiently Verifiable Puzzle

An Efficiently Verifiable Puzzle



An Efficiently Verifiable Puzzle



Question: Can this lock be opened?

Verifiers, Formally

• Formally, a **verifier** is a TM V such that

w $\in L$ iff $\exists x \in \Sigma^*$. V accepts $\langle w, x \rangle$

• In other words

 $L = \{ w \in \Sigma^* \mid \exists x \in \Sigma^*. \text{ V accepts } \langle w, x \rangle \}$

- If $w \in L$, the verifier can check this easily if we know the proper x.
- If $w \notin L$, the verifier does not help much.
 - Just because V rejects $\langle w, x \rangle$ does not mean that $w \notin L$.
- Note that $\mathscr{L}(V) \neq L$.

Verification is Powerful

- Many undecidable languages can still be verified.
- Here is a verifier for *HALT*:
 - $V = "On input \langle M, w, n \rangle$, where M is a TM, w is a string, and n is a natural number:
 - Run M on w for n steps.
 - If *M* halts *w* within that time, accept; otherwise reject."
- V always halts on all inputs (even if M loops on w).
- If *M* halts on *w*, there is some choice of *n* for which *V* accepts (namely, the number of steps *M* takes before it halts on *w*).
- Thus *HALT* can be **verified** but not **decided**.

- Does a Sudoku grid have a solution?
 - M = "On input (S, A), an encoding of a Sudoku puzzle and an alleged solution to it:
 - **Deterministically** check whether *A* is a solution to *S*.
 - If so, accept; if not, reject."

		7		6		1		
					3		5	2
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5	7		4					
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- A **graph coloring** is a way of assigning colors to nodes in an undirected graph such that no two nodes joined by an edge have the same color.
 - Applications in compilers, cell phone towers, etc.
- Question: Can graph *G* be colored with at most *k* colors?
- M = "On input $\langle G, k, C \rangle$, where C is an alleged coloring:
 - **Deterministically** check whether *C* is a legal *k*-coloring of *G*.
 - If so, accept; if not, reject."



Two Equivalent Formulations of $\ensuremath{\mathbf{NP}}$

- **Theorem**: A language *L* has a polynomial-time verifier iff $L \in \mathbf{NP}$.
- Proof sketch:
 - Any polynomial-time verifier can be turned into a polynomial-time NTM by having the NTM nondeterministically guess the certificate for *w*, then check it (deterministically) by running the verifier.
 - If an NTM can decide *L* in polynomial time, a verifier could work by having a certificate saying which nondeterministic choices the original NTM made, then simulating those choices of the NTM to check it.






Next Time

• NP-Completeness

- What are the hardest problems in $\ensuremath{\mathbf{NP}}\xspace$?
- How do you prove a problem is NPcomplete?