15-453

FORMAL LANGUAGES, AUTOMATA AND COMPUTABILITY TIME COMPLEXITY AND POLYNOMIAL TIME; NON DETERMINISTIC TURING MACHINES AND NP

THURSDAY Mar 20

COMPLEXITY THEORY

Studies what can and can't be computed under limited resources such as time, space, etc

Today: Time complexity

MEASURING TIME COMPLEXITY

We measure time complexity by counting the elementary steps required for a machine to halt

Consider the language $A = \{ 0^k 1^k \mid k \ge 0 \}$

On input of length n:

- 1. Scan across the tape and reject if the string is not of the form 0ⁱ1^j
- 2. Repeat the following if both 0s and 1s $\sim n^2$ remain on the tape:

Scan across the tape, crossing off a single 0 and a single 1

3. If 0s remain after all 1s have been crossed off, or vice-versa, reject. Otherwise accept.

Definition: Let M be a TM that halts on all inputs. The running time or time-complexity of M is the function $f: N \to N$, where f(n) is the maximum number of steps that M uses on any input of length n.

ASYMPTOTIC ANALYSIS

 $5n^3 + 2n^2 + 22n + 6 = O(n^3)$

BIG-O

Let f and g be two functions f, g: $N \to R^+$. We say that f(n) = O(g(n)) if there exist positive integers c and n_0 so that for every integer $n \ge n_0$

$$f(n) \le cg(n)$$

When f(n) = O(g(n)), we say that g(n) is an asymptotic upper bound for f(n)

f asymptotically NO MORE THAN g

$$5n^3 + 2n^2 + 22n + 6 = O(n^3)$$

If c = 6 and n_0 = 10, then $5n^3 + 2n^2 + 22n + 6 \le cn^3$

$$2n^{4.1} + 200283n^4 + 2 =$$

 $3n\log_2 n + 5n\log_2\log_2 n =$

$$nlog_{10} n^{78} =$$

$$log_{10} n = log_2 n / log_2 10$$

$$O(nlog_2 n) = O(nlog_{10} n) = O(nlog n)$$

Definition: $TIME(t(n)) = \{ L \mid L \text{ is a language decided by a } O(t(n)) \text{ time Turing Machine } \}$

$$A = \{ 0^k 1^k \mid k \ge 0 \} \in TIME(n^2)$$

$$A = \{ 0^k 1^k \mid k \ge 0 \} \in TIME(nlog n)$$

Cross off every other 0 and every other 1. If the # of 0s and 1s left on the tape is odd, reject

0000000000001111111111111

x0x0x0x0x0x0x0xx1x1x1x1x1x1x1x

xxx0xxx0xxx0xxxx1xxx1x

xxxxxxx0xxxxxxxxxxxx1xxxxx

XXXXXXXXXXXXXXXXXXXXXXXX

We can prove that a TM cannot decide A in less time than O(nlog n)

*7.49 Extra Credit. Let $f(n) = o(n\log n)$. Then Time(f(n)) contains only regular languages.

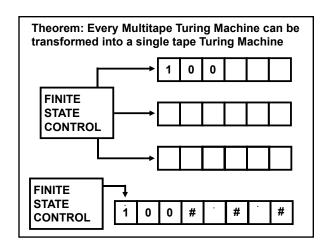
where f(n) = o(g(n)) iff $\lim_{n\to\infty} f(n)/g(n) = 0$ ie, for all c > 0, $\exists n_0$ such that f(n) < cg(n) for all $n \ge n_0$ f asymptotically LESS THAN g Can A = { $0^k1^k \mid k \ge 0$ } be decided in time O(n) with a two-tape TM?

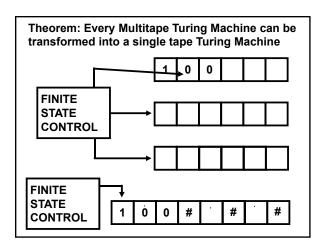
Scan all 0s and copy them to the second tape. Scan all 1s, crossing off a 0 from the second tape for each 1.

Different models of computation yield different running times for the same language!

Theorem: Let t(n) be a function such that $t(n) \ge n$. Then every t(n)-time multi-tape TM has an equivalent $O(t(n)^2)$ single tape TM

Claim: Simulating each step in the multitape machine uses at most O(t(n)) steps on a single-tape machine. Hence total time of simulation is $O(t(n)^2)$.





Theorem: Let t(n) be a function such that $t(n) \ge n$. Then every t(n)-time multi-tape TM has an equivalent $O(t(n)^2)$ single tape TM

Analysis: (Note, k, the # of tapes, is fixed.)

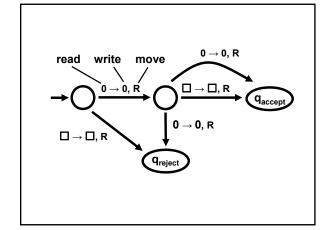
Let S be simulator

- Put S's tape in proper format: O(n) steps
- Two scans to simulate one step,
 1. to optain info for next move O(t(n)) steps, why?
 2. to simulate it (may need to shift everything over to right possibly k times): O(t(n)) steps, why?

Therefore, $O(n) + t(n) O(t(n)) = O(t(n)^2)$ steps in simulation,

$$P = \bigcup_{k \in N} TIME(n^k)$$

NON-DETERMINISTIC TURING MACHINES AND NP



Definition: A Non-Deterministic TM is a 7-tuple T = (Q, Σ , Γ , δ , q_0 , q_{accept} , q_{reject}), where:

Q is a finite set of states

 Σ is the input alphabet, where $\square \notin \Sigma$

 Γ is the tape alphabet, where $\square \in \Gamma$ and $\Sigma \subseteq \Gamma$

 $\delta: Q \times \Gamma \to 2^{(Q \times \Gamma \times \{L,R\})}$

 $q_0 \in Q$ is the start state

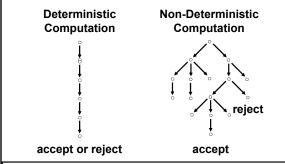
 $q_{accept} \in Q$ is the accept state

 $q_{reject} \in Q$ is the reject state, and $q_{reject} \neq q_{accept}$

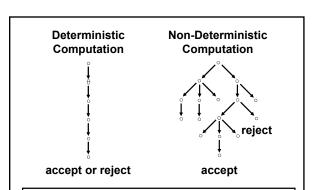
NON-DETERMINISTIC TMs

...are just like standard TMs, except:

- 1. The machine may proceed according to several possibilities
- 2. The machine accepts a string if there exists a path from start configuration to an accepting configuration



Definition: Let M be a NTM that is a decider (all branches halt on all inputs). The running time or time-complexity of M is the function $f: N \to N$, where f(n) is the maximum number of steps that M uses on any branch of its computation on any input of length n.

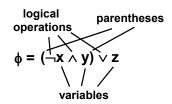


Theorem: Let t(n) be a function such that $t(n) \ge n$. Then every t(n)-time nondeterministic single-tape TM has an equivalent $2^{O(t(n))}$ deterministic single tape TM

Definition: NTIME(t(n)) = { L | L is decided by a O(t(n))-time non-deterministic Turing machine }

 $\mathsf{TIME}(\mathsf{t}(\mathsf{n})) \subseteq \mathsf{NTIME}(\mathsf{t}(\mathsf{n}))$

BOOLEAN FORMULAS



A satisfying assignment is a setting of the variables that makes the formula true

x = 1, y = 1, z = 1 is a satisfying assignment for ϕ

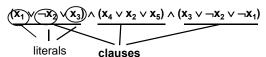
A Boolean formula is satisfiable if there exists a satisfying assignment for it

YES
$$a \wedge b \wedge c \wedge \neg d$$

NO $\neg (x \vee y) \wedge x$

SAT = { $\phi \mid \phi$ is a satisfiable Boolean formula }

A 3cnf-formula is of the form:



YES $(x_1 \lor \neg x_2 \lor x_1)$

NO
$$(x_3 \lor x_1) \land (x_3 \lor \neg x_2 \lor \neg x_1)$$

NO
$$(x_1 \lor x_2 \lor x_3) \land (\neg x_4 \lor x_2 \lor x_1) \lor (x_3 \lor x_1 \lor \neg x_1)$$

NO
$$(x_1 \lor \neg x_2 \lor x_3) \land (x_3 \land \neg x_2 \land \neg x_1)$$

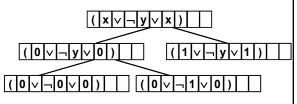
3SAT = $\{ \phi \mid \phi \text{ is a satisfiable 3cnf-formula } \}$

3SAT = { $\phi \mid \phi$ is a satisfiable 3cnf-formula }

Theorem: $3SAT \in NTIME(n^2)$

On input ϕ :

- 1. Check if the formula is in 3cnf
- 2. For each variable, non-deterministically substitute it with 0 or 1



3. Test if the assignment satisfies ϕ

 $NP = \bigcup_{k \in N} NTIME(n^k)$

Theorem: $L \in NP \Leftrightarrow if$ there exists a poly-time Turing machine V(erifier) with

L = { $x \mid \exists y \text{(witness)} \mid y \mid = poly(\mid x \mid) \text{ and } V(x,y) \text{ accepts }}$

- (1) If L = { x | ∃y |y| = poly(|x|) and V(x,y) accepts } then L ∈ NP
- (2) If L ∈ NP then L = { x | ∃y |y| = poly(|x|) and V(x,y) accepts }

Theorem: $L \in NP \Leftrightarrow if$ there exists a poly-time Turing machine V(erifier) with

L = { $x \mid \exists y \text{ (witness) } |y| = \text{poly(}|x|\text{)} \text{ and } V(x,y) \text{ accepts }}$ Proof:

(1) If L = { x | ∃y |y| = poly(|x|) and V(x,y) accepts } then L ∈ NP

Because we can guess y and then run V

(2) If $L \in NP$ then $L = \{ x \mid \exists y \mid y \mid = poly(|x|) \text{ and } V(x,y) \text{ accepts } \}$

Theorem: $L \in NP \Leftrightarrow if$ there exists a poly-time Turing machine V(erifier) with

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(1) If L = { $x \mid \exists y \mid y \mid = poly(\mid x \mid)$ and V(x,y) accepts } then L $\in NP$

Because we can guess y and then run V

(2) If L ∈ NP then L = { x | ∃y |y| = poly(|x|) and V(x,y) accepts }

Let N be a non-deterministic poly-time TM that decides L and define V(x,y) to accept if y is an accepting computation history of N on x

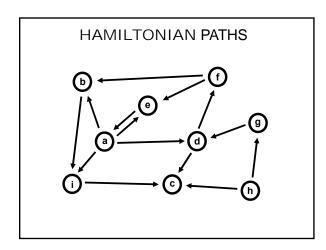
3SAT = $\{ \phi \mid \exists y \text{ such that } y \text{ is a satisfying assignment to } \phi \text{ and } \phi \text{ is in 3cnf} \}$

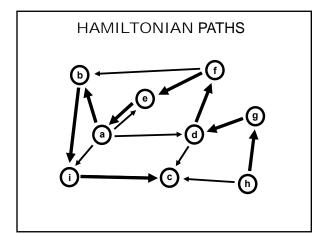
SAT = $\{ \phi \mid \exists y \text{ such that } y \text{ is a satisfying assignment to } \phi \}$

A language is in NP if and only if there exist polynomial-length certificates* for membership to the language

SAT is in NP because a satisfying assignment is a polynomial-length certificate that a formula is satisfiable

* that can be verified in poly-time

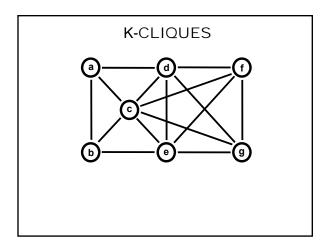


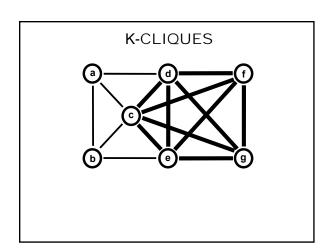


HAMPATH = { (G,s,t) | G is a directed graph with a Hamiltonian path from s to t }

Theorem: HAMPATH ∈ NP

The Hamilton path itself is a certificate





CLIQUE = { (G,k) | G is an undirected graph with a k-clique }

Theorem: CLIQUE ∈ NP

The k-clique itself is a certificate

NP = all the problems for which once you have the answer it is easy (i.e. efficient) to verify

P = NP? \$\$\$

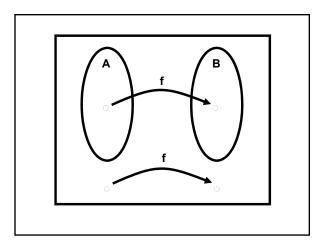
POLY-TIME REDUCIBILITY

 $f: \Sigma^* \to \Sigma^*$ is a polynomial time computable function if some poly-time Turing machine M, on every input w, halts with just f(w) on its tape

Language A is polynomial time reducible to language B, written $A \leq_P B$, if there is a polytime computable function $f: \Sigma^* \to \Sigma^*$ such that:

$$w \in A \Leftrightarrow f(w) \in B$$

f is called a polynomial time reduction of A to B



Theorem: If $A \leq_P B$ and $B \in P$, then $A \in P$

Proof: Let $M_{\rm B}$ be a poly-time (deterministic) TM that decides B and let f be a poly-time reduction from A to B

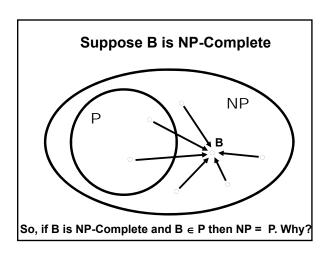
We build a machine M_A that decides A as follows:

On input w:

- 1. Compute f(w)
- 2. Run M_B on f(w)

Definition: A language B is NP-complete if:

- 1. B ∈ NP
- 2. Every A in NP is poly-time reducible to B (i.e. B is NP-hard)



Theorem (Cook-Levin): SAT is NP-complete

Corollary: SAT ∈ P if and only if P = NP

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Read Chapter 7.3 of the book for next time