Computational Complexity Theory

Lecture 4: NTM, Class co-NP and EXP; Diagonalization

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NTM: An alternate characterization of NP

- A nondeterministic Turing machine is like a deterministic Turing machines but with two transition functions.
- It is formally defined by a tuple $(\Gamma, Q, \delta_0, \delta_1)$. It has a special state q_{accept} in addition to q_{start} and q_{halt} .

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this is different from *randomly*

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- At every step of computation, the machine applies one of two functions δ_0 and δ_1 *arbitrarily*.
- Unlike DTMs, NTMs are **not intended to be physically realizable** (because of the arbitrary nature of application of the transition functions).

- Definition. An NTM M <u>accepts</u> a string $x \in \{0, I\}^*$ iff on input x there <u>exists</u> a sequence of applications of the transition functions δ_0 and δ_1 (beginning from the start configuration) that makes M reach q_{accept} .
- Definition. An NTM M <u>decides</u> a language L ⊆ {0,1}* if
 M accepts x → x∈L

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remember in this course we'll always be dealing with TMs that halt on every input.

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- Definition. An NTM M decides L in T(|x|) time if
 - > M accepts x \iff x \in L

> On <u>every sequence</u> of applications of the transition functions on input x, M either reaches q_{accept} or q_{halt} within T(|x|) steps of computation.

Class NTIME

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- Theorem. NP = $\bigcup_{c>0}$ NTIME (n^c).

Proof sketch: Let L be a language in NP. Then, there's a poly-time verifier M s.t,

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Think of an NTM M' that on input x, at first <u>guesses</u> a u $\in \{0, I\}^{p(|x|)}$ by applying δ_0 and δ_1 nondeterministically

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.... and then simulates M on (x, u) to verify M(x, u) = 1.

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- Definition. For every $L \subseteq \{0, I\}^*$ let $\overline{L} = \{0, I\}^* \setminus L$. A language L is in co-NP if \overline{L} is in NP.
- Example. SAT = $\{\phi : \phi \text{ is } \underline{not} \text{ satisfiable}\}$.

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- Note: co-NP is <u>not</u> complement of NP. Every language in P is in both NP and co-NP.

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 A language L is in co-NP if L is in NP.
- Example. SAT = $\{\phi : \phi \text{ is } \underline{not} \text{ satisfiable}\}$.
- Note: SAT is Cook reducible to SAT. But, there's a fundamental difference between the two problems that is captured by the fact that SAT is <u>not</u> known to be Karp reducible to SAT. In other words, there's no known poly-time verification process for SAT.

Recall, a language L ⊆ {0, I}* is in NP if there's a poly-time verifier M such that

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 $x \in L \quad \Longleftrightarrow \exists u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = I$ $x \in \overline{L} \quad \longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = 0$ $x \in \overline{L} \quad \longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } \overline{M}(x, u) = I$ $\bigcap \\ \overline{M} \text{ outputs the opposite of } M$

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 - $\begin{array}{ll} x \in L & \Longleftrightarrow \exists u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = I \\ x \in \overline{L} & \longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = 0 \\ x \in \overline{L} & \longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } \overline{M}(x, u) = I \end{array}$
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$$x \in L \quad \Longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = I$$

- Definition. A language $L' \subseteq \{0, I\}^*$ is co-NP-complete if
 - L' is in co-NP
 - Every language L in co-NP is polynomial-time (Karp) reducible to L'.
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 - ¯L ∈ NP

$$\implies \overline{L} \leq_{p} SAT \\ \implies L \leq_{D} \overline{SAT}$$

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 - Every language L in co-NP is polynomial-time (Karp) reducible to L'.
- Theorem. Let
 - TAUTOLOGY = { ϕ : every assignment satisfies ϕ }. TAUTOLOGY is co-NP complete.
 - Proof. Similar (homework)

- Definition. A language $L' \subseteq \{0, I\}^*$ is co-NP-complete if
 - L' is in co-NP
 - Every language L in co-NP is polynomial-time (Karp) reducible to L'.
- Theorem. If L in NP-complete then L is co-NP-complete
 Proof. Similar (homework)

The diagram again



The diagram again



Integer factoring in NP \cap co-NP

• Integer factoring.

 $FACT = \{(N, U): there's a prime in [U] dividing N\}$

- Claim. FACT \in NP \cap co-NP
- So, FACT is NP-complete implies NP = co-NP.

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FACT = {(N, U): there's a prime in [U] dividing N}

- Claim. FACT \in NP \cap co-NP
- Proof. FACT ∈ NP : Give p as a certificate. The verifier checks if p is prime (AKS test), I ≤ p ≤ U and p divides N.
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Proof. FACT ∈ NP : Give complete prime factorization of N as a certificate. The verifier checks the correctness of the factorization, and then checks if none of the prime factors is in [U].

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 Homework: If FACT ∈ P, then there's a algorithm to find the prime factorization a given n-bit integers in poly(n) time.

Integer factoring in NP \cap co-NP

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FACT = {(N, U): there's a prime in [U] dividing N}

• Factoring algorithm. Dixon's randomized algorithm factors an n-bit number in $\exp(O(\sqrt{n \log n}))$ time.

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- Exponential Time Hypothesis. (Impagliazzo & Paturi 1999) Any algorithm for 3-SAT takes $\geq 2^{\delta \cdot n}$ time, where $\delta \geq 0$ is <u>some fixed constant</u> and n is the no. of variables.

In other words, δ cannot be made arbitrarily close to 0.

- Definition. Class EXP is the exponential time analogue of class P. EXP = \bigcup DTIME (2ⁿ ^c)
- Observation. $P \subseteq NP \subseteq EXP$
- Exponential Time Hypothesis. (Impagliazzo & Paturi 1999) Any algorithm for 3-SAT takes $\geq 2^{\delta.n}$ time, where $\delta > 0$ is some fixed constant and n is the no. of variables.

ETH \implies P \neq NP

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If M_{α} takes T time on x then U takes O(T log T) time to simulate M_{α} on x.

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- These techniques are characterized by <u>two</u> main features:
 - I. There's a universal TM U that when given strings α and x, simulates M_{α} on x with only a <u>small</u> overhead.
 - 2. Every string represents some TM, and every TM can be represented by *infinitely many* strings.

- An application of Diagonalization

• Let f(n) and g(n) be <u>time-constructible</u> functions s.t., $f(n) . \log f(n) = o(g(n)).$ e.g. $f(n) = n, g(n) = n^2$

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Theorem. DTIME(f(n)) ⊊ DTIME(g(n)) Proof. We'll prove with f(n) = n and g(n) = n². Task: Show that there's a language L decided by a TM D with time complexity O(n²) s.t., any TM M with runtime O(n) cannot decide L.

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I. On input x, compute $|x|^2$.

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 - b. Otherwise, output 1.

Let f(n) and g(n) be time-constructible functions s.t., f(n) . log f(n) = o(g(n)).
Theorem. DTIME(f(n)) ⊊ DTIME(g(n)) Proof. We'll prove with f(n) = n and g(n) = n². D runs in O(n²) time as n² is <u>time-constructible</u>.

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Theorem. DTIME(f(n)) ⊊ DTIME(g(n)) Proof. We'll prove with f(n) = n and g(n) = n². Claim. There's no TM M with running time O(n) that decides L (the language accepted by D).

Let f(n) and g(n) be time-constructible functions s.t.,
 f(n) . log f(n) = o(g(n)).

- Theorem. $DTIME(f(n)) \subsetneq DTIME(g(n))$
 - Proof. We'll prove with f(n) = n and $g(n) = n^2$.
 - For contradiction, suppose M decides L and runs for at most c.n steps on inputs of length n.

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- > Think of a <u>sufficiently large</u> x such that $M = M_x$.
- > Suppose $M(x) = M_x(x) = b$.
- D on input x, simulates M_x on x for |x|² steps. Since M_x stops within c.|x| steps, D's simulation also stops within c'.c. |x|. log |x| steps.

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> Suppose $M(x) = M_x(x) = b$.

D on input x, simulates M_x on x for |x|² steps. Since M_x stops within c.|x| steps, D's simulation also stops within c'.c. |x|. log |x| steps. (as c'.c. |x|. log |x| < |x|² for sufficiently large x)

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Proof. We'll prove with f(n) = n and $g(n) = n^2$.

- For contradiction, suppose M decides L and runs for at most c.n steps on inputs of length n.
- > Think of a <u>sufficiently large</u> x such that $M = M_x$.
- > Suppose $M(x) = M_x(x) = b$.
- D on input x, simulates M_x on x for |x|² steps. Since M_x stops within c.|x| steps, D's simulation also stops within c'.c. |x|. log
 |x| steps. And D outputs the opposite of what M_x outputs!

- Let f(n) and g(n) be time-constructible functions s.t.,
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- For contradiction, suppose M decides L and runs for at most c.n steps on inputs of length n.
- > Think of a <u>sufficiently large</u> x such that $M = M_x$.
- > Suppose $M(x) = M_x(x) = b$.
- \rightarrow Hence, D(x) = I-b.
- Let f(n) and g(n) be time-constructible functions s.t.,
 f(n) . log f(n) = o(g(n)).
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Contradiction! M does not decide L.

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 f(n) . log f(n) = o(g(n)).

- Theorem. $DTIME(f(n)) \subsetneq DTIME(g(n))$
- Theorem. P ⊊ EXP
 Proof. Similar (homework)

• Are there natural problems that take $\Omega(n^2)$ time?

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- 3SUM: Given a list of n numbers, check if there exists
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- Theorem (Patrascu & Williams 2010). ETH implies kSUM requires $n^{\Omega(k)}$ time.