Computational Complexity Theory

Lecture 7: Class EXP; Time Hierarchy Theorem

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Recap: Nondeterministic Turing Machines

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

Recap: Nondeterministic Turing Machines

- Definition. An NTM M accepts a string $x \in \{0, I\}^*$ iff on input x there **exists** 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 decides L in T(|x|) time if
 M accepts x → x∈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.

Recap: Alternate characterization of NP

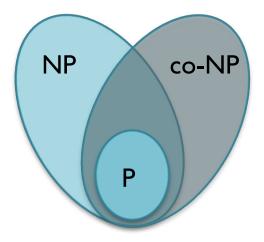
- Definition. A language L is in NTIME(T(n)) if there's an NTM M that decides L in c. T(n) time on inputs of length n, where c is a constant.
- Theorem. NP = $\bigcup_{c>0}$ NTIME (n^c).

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

 $x \in L \implies \exists u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = I$

Recap: Class co-NP

- 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 = $\{\phi : \phi \text{ is } \underline{not} \text{ satisfiable}\}$.



Recap: Class co-NP : Alternate definition

Recall, a language L ⊆ {0,1}* is in NP if there's a poly function p and a poly-time verifier M such that

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

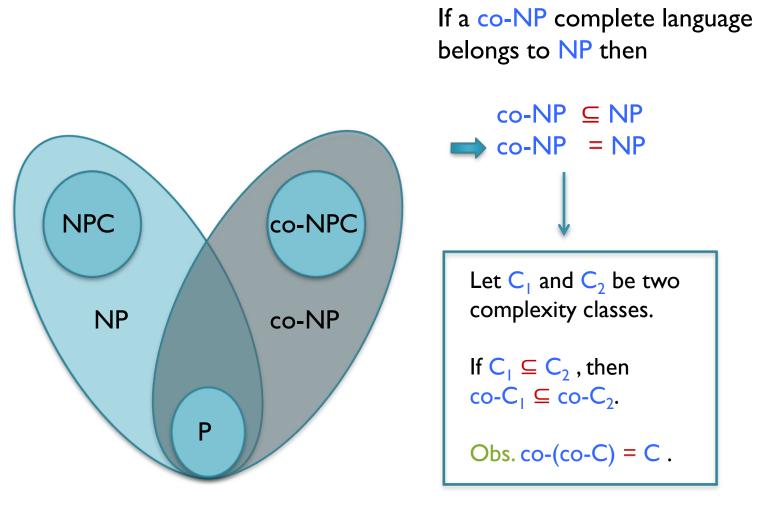
Definition. A language L ⊆ {0,1}* is in co-NP if there's a poly function p and a poly-time TM M such that
 x∈L ⇔∀u ∈{0,1}^{p(|x|)} s.t. M(x, u) = 1

for NP this was \exists

Recap: co-NP-completeness

- 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. SAT and TAUTOLOGY are co-NP-complete.
- Theorem. If L in NP-complete then L is co-NP-complete

The diagram again



The diagram again co-NP = NPNPC co-NPC co-NP NP ?? Ρ

If a co-NP-complete language belongs to NP then

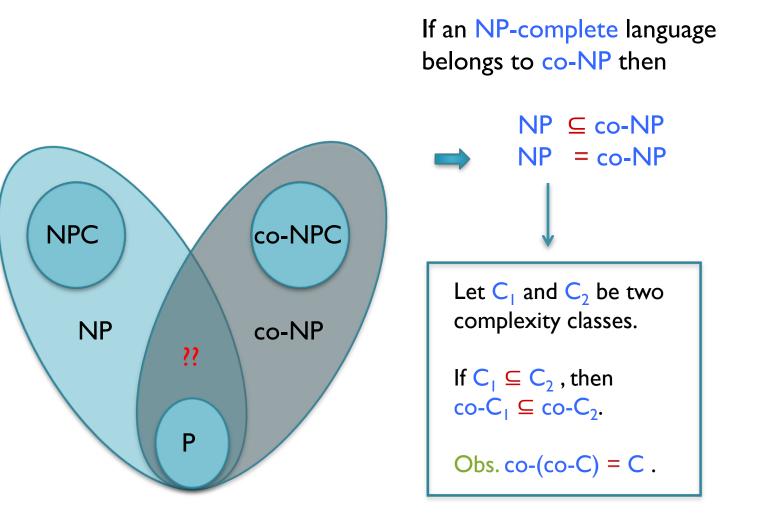
 $co-NP \subseteq NP$

Let C_1 and C_2 be two complexity classes.

If $C_1 \subseteq C_2$, then $co-C_1 \subseteq co-C_2$.

Obs. co-(co-C) = C.

The diagram again



• 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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- 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 the 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.

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

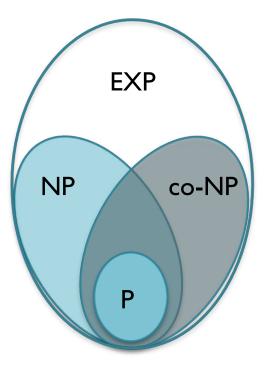
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• <u>Exponential Time Hypothesis</u>. (Impagliazzo & Paturi 1999) Any algorithm for 3-SAT takes $\geq 2^{\delta.n}$ time, where $\delta \geq 0$ is <u>some fixed constant</u> and n is the no. of variables.

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

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ETH \implies P \neq NP

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Homework: Read about Strong Exponential Time Hypothesis (SETH).

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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., $f(n) \cdot \log f(n) = o(g(n)) \cdot e.g. f(n) = n, g(n) = n^2$

Let f(n) and g(n) be time-constructible functions s.t.,
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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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 - 1. On input x, compute $|x|^2$.

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D outputs the **<u>opposite</u>** of what M_x outputs.

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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).

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 |x| steps. And D outputs the opposite of what M_x outputs!

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Contradiction! M does not decide L.

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- Theorem. P ⊊ EXP
 Proof. Similar (homework)

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 3 numbers in the list that sum to zero.
- Conjecture. No algorithm solves 3SUM in $O(n^{2-\epsilon})$ time for some constant $\epsilon > 0$.
- However, there's a ~O(n² / (log n)²) time algorithm for 3SUM. ("~" suppressing a poly(log log n) factor.)

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 k numbers in the list that sum to zero.
- Theorem (Patrascu & Williams 2010). ETH implies kSUM requires $n^{\Omega(k)}$ time.