# **Computational Complexity Theory**

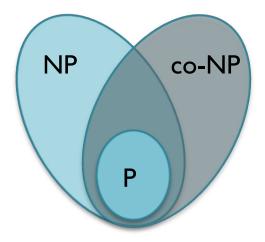
#### Lecture 8: Class EXP; Time Hierarchy; Ladner's theorem

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#### Class co-NP and EXP

# Recap: Class co-NP

- Definition. For every L ⊆ {0,1}\* let L = {0,1}\* \ L.
   A language L is in co-NP if L is in NP.
- Example. SAT =  $\{\phi : \phi \text{ is } \underline{not} \text{ satisfiable}\}$ .



# Recap: Alternate definition of co-NP

- Recall, a language L ⊆ {0, I}\* is in NP if there's a poly-time verifier M such that
  - $\begin{array}{ll} x \in L & \Longleftrightarrow \exists u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = I \\ x \in \overline{L} & \Longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } M(x, u) = 0 \\ x \in \overline{L} & \longleftrightarrow \forall u \in \{0, I\}^{p(|x|)} \text{ s.t. } \overline{M}(x, u) = I \end{array}$
- Definition. A language  $L \subseteq \{0, I\}^*$  is in co-NP if there's a polynomial function p and a poly-time TM M such that

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

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 is co-NP-complete.

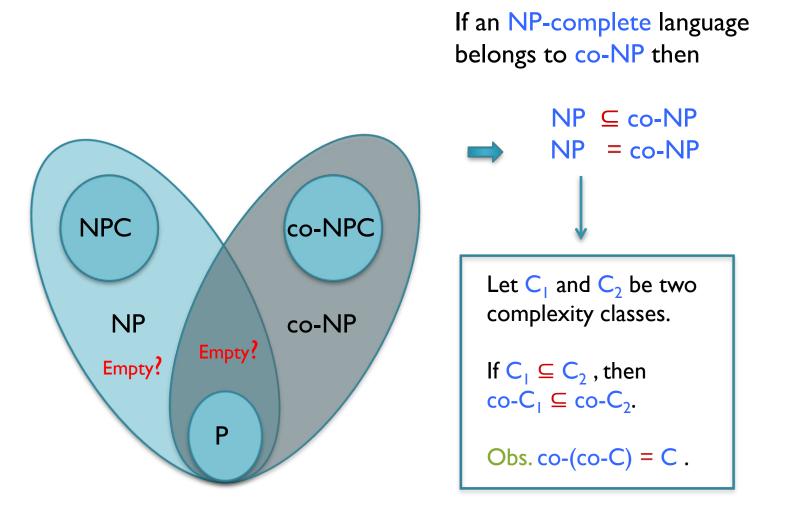
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  - Every language L in co-NP is polynomial-time (Karp) reducible to L'.
- Theorem. Let
  - TAUTOLOGY = { $\phi$  : every assignment satisfies  $\phi$  }. TAUTOLOGY is co-NP-complete.
  - Proof. Similar (homework)

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- Definition. A language L'  $\subseteq$  {0,1}\* 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)

# Recap: The diagram again



# Recap: FACT 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.

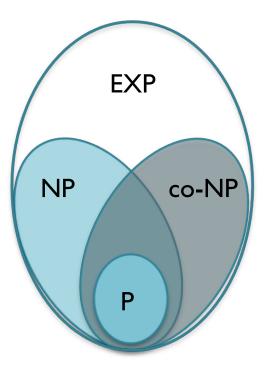
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• <u>Exponential Time Hypothesis</u>. (Impagliazzo & Paturi 1999) Any algorithm for 3-SAT takes  $\geq 2^{\delta.n}$  time, where  $\delta \geq 0$ is <u>some fixed constant</u> and n is the no. of variables.

In other words,  $\delta$  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).

- Definition. Class EXP is the exponential time analogue of class P.
   EXP = ∪ DTIME (2<sup>n<sup>C</sup></sup>) We'll address this using diagonalization
   Observation. P ⊆ NP ⊆ EXP Is P ⊊ EXP?
- <u>Exponential Time Hypothesis</u>. (Impagliazzo & Paturi 1999) Any algorithm for 3-SAT takes  $\geq 2^{\delta.n}$  time, where  $\delta \geq 0$ is some fixed constant and n is the no. of variables.

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If  $M_{\alpha}$  takes T time on x then U takes O(T log T) time to simulate  $M_{\alpha}$  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  $\alpha$  and x, simulates  $M_{\alpha}$  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)).
- Theorem. (Hartmanis & Stearns 1965)
   DTIME(f(n)) ⊊ DTIME(g(n))
- Theorem.  $P \subsetneq EXP$
- This type of results are called **lower bounds**.

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Theorem. DTIME(f(n)) ⊊ DTIME(g(n)) Proof. We'll prove with f(n) = n and g(n) = n<sup>2</sup>.

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Theorem. DTIME(f(n)) ⊊ DTIME(g(n)) Proof. We'll prove with f(n) = n and g(n) = n<sup>2</sup>. Task: Show that there's a language L decided by a TM D with time complexity O(n<sup>2</sup>) s.t., any TM M with runtime O(n) cannot decide L.

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D's time steps not  $M_x$ 's time steps.

2. Simulate  $M_x$  on x for  $|x|^2$  steps.

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D outputs the **<u>opposite</u>** 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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- Theorem.  $DTIME(f(n)) \subsetneq DTIME(g(n))$
- Theorem. P  $\subsetneq$  EXP
- **No** EXP-complete problem (under poly-time Karp reduction) is in P.

E.g., Decide if a TM halts in k steps; generalized versions of games such as chess, checkers, Go, etc.

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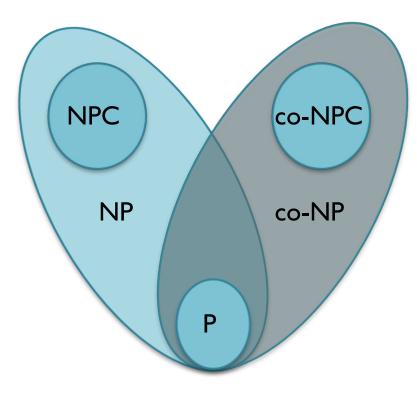
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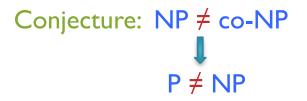
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- However, there's a ~O(n<sup>2</sup> / (log n)<sup>2</sup>) time algorithm for 3SUM. ("~" suppressing a poly(log log n) factor.)

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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.

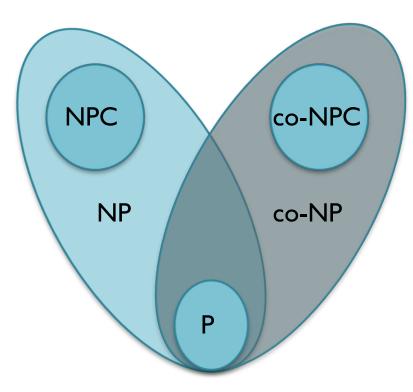
# Revisiting NP\co-NP

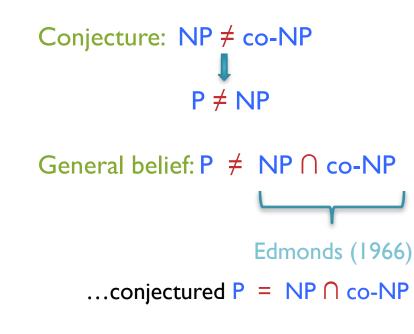




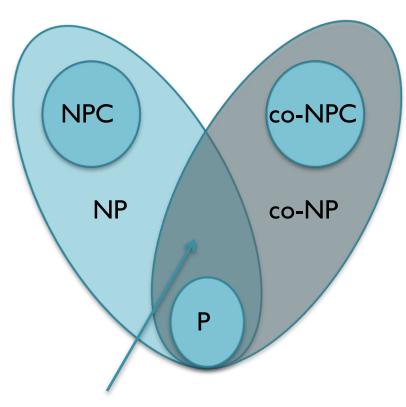
General belief:  $P \neq NP \cap co-NP$ 

# Revisiting NP∩co-NP





# Revisiting NP∩co-NP



Conjecture: NP  $\neq$  co-NP  $\downarrow$ P  $\neq$  NP

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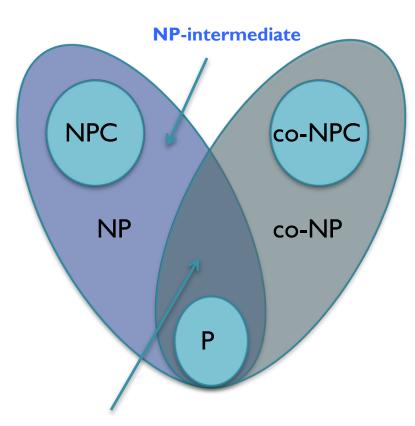
Check:

https://cstheory.stackexchange.com/questions/20 021/reasons-to-believe-p-ne-np-cap-conp-or-not

Check if the shortest non-zero vector in an n-dimensional lattice has length at most I or at least  $\sqrt{n}$ .

- Integer factoring (FACT)
- Approximate shortest vector in a lattice

Ref: "Lattice problems in NP∩co-NP" by Aharonov & Regev (2005)



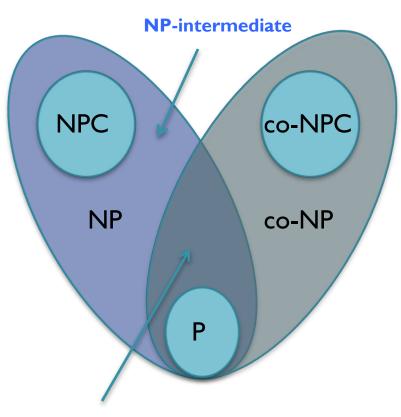
Conjecture:  $NP \neq co-NP$   $\downarrow$  $P \neq NP$ 

General belief:  $P \neq NP \cap co-NP$ 

Obs: If NP  $\neq$  co-NP and FACT  $\notin$  P then FACT is NP-intermediate.

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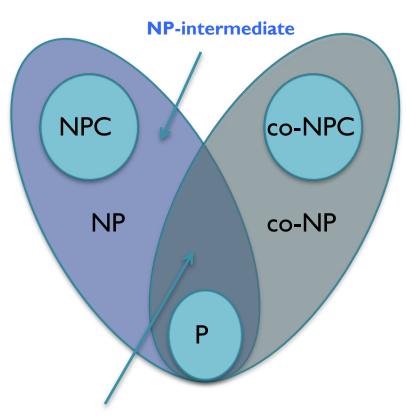
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Ladner's theorem:  $P \neq NP$  implies existence of a NP-intermediate language.



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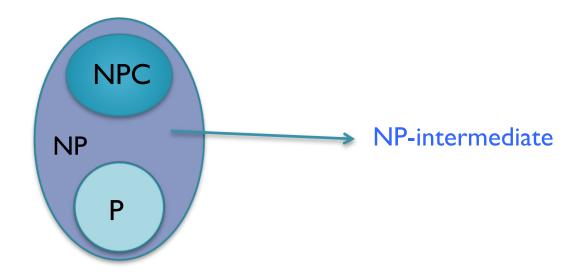
Ladner's theorem:  $P \neq NP$  implies existence of a NP-intermediate language.

(proved using **diagonalization**)

#### Ladner's Theorem

- Another application of Diagonalization

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- Theorem. (Ladner 1975) If P ≠ NP then there is a NPintermediate language.
  - **Proof.** A delicate argument using diagonalization.

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Let 
$$SAT_H = \{\Psi 0 \mid m^{H(m)} : \Psi \in SAT \text{ and } |\Psi| = m\}$$

H would be defined in such a way that  $SAT_{H}$  is NP-intermediate (assuming P  $\neq$  NP )

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**Proof:** Later (uses <u>diagonalization</u>).

Let's see the proof of Ladner's theorem assuming the existence of such a "special" H.

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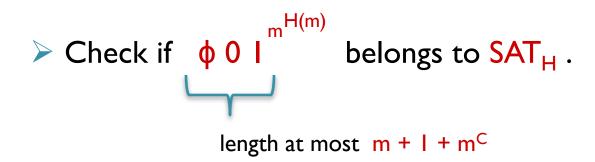
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• As  $P \neq NP$ , it must be that  $SAT_H \notin P$ .

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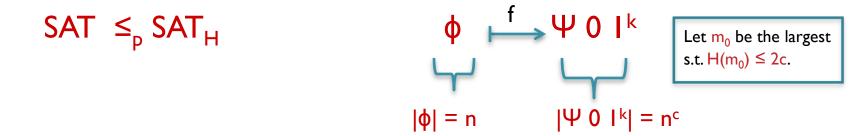
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 $|\varphi| = n$   $|\Psi 0 I^{k}| = n^{c}$ 

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Either  $m \le m_0$  (in which case the task reduces to checking if a constant-size  $\Psi$  is satisfiable),

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> Compute H(m) and check if  $k = m^{H(m)}$ .

or H(m) > 2c (as H(m) tends to infinity with m).

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 $\Rightarrow$  Compute H(m) and check if  $k = m^{H(m)}$ .  
 $\Rightarrow$  Hence,  $\sqrt{n} \geq m$ .

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- ≻ Hence,  $\sqrt{n} \ge m$ . Also  $φ \in SAT$  iff  $Ψ \in SAT$

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Thus, checking if an n-size formula  $\phi$ is satisfiable reduces to checking if a  $\sqrt{n}$ -size formula  $\Psi$  is satisfiable.

 $P \neq NP$ 

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- > Compute H(m) and check if  $k = m^{H(m)}$ .
- ➢ Hence, √n ≥ m. Also φ ∈ SAT iff Ψ ∈ SAT

Do this recursively! Only O(log log n) recursive steps required.

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- ≻ Hence,  $\sqrt{n} \ge m$ . Also  $\phi \in SAT$  iff  $\Psi \in SAT$
- Hence  $SAT_H$  is not NP-complete, as P  $\neq$  NP.

# Natural NP-intermediate problems ??

- Integer factoring
- Approximate shortest vector in a lattice
- Minimum Circuit Size Problem

("Multi-output MCSP is NP-hard", Ilango, Loff & Oliveira 2020)

• Graph isomorphism

("GI in QuasiP time", Babai 2015)