Undergraduate Complexity Theory 15-455 @ CMU TOC

Marcythm

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1 Course Overview

resources; efficient; famous open problems; notations $(\langle X \rangle)$. Three types of problems: decision / function / search.

2 Turing Machines

language; Turing machine; decider; computation trace; Church-Turing Thesis (extended ver.)

3 Simulation and Turing Machine Variants

Standard model: 1-tape TM.

Variations: 1-way / alphabet / multitape / RAM; Boolean Circuits; TM operation tricks; simulation.

4 Time Complexity and Universal Turing Machines

(time) complexity class TIME(f(n)); Speedup Theorem; P; Universal TM; diagonalization method.

5 The Time Hierarchy Theorem

Time Hierarchy Theorem (more time, more power). diagonalization method (important technique: simulation then do the opposite). time constructible.

6 Problems in P

by THT \exists more than P: EXP. PATH, 2-COL, LCS, c-CLIQUE.

7 SAT

different forms: CKT-SAT, FORMULA-SAT a.k.a. SAT, CNF-SAT, k-SAT (descending on generality). All NP-complete; CKT-EVAL not parallelizable; formula; time complexity about c-SAT.

8 NP

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conjectures: P \neq NP;

ETH: \exists \delta > 0 : 3\text{-SAT} \notin \mathsf{TIME}((1+\delta)^n);

SETH: \forall \delta > 0 : \exists k : k\text{-SAT} \notin \mathsf{TIME}((2-\delta)^n).

verifier-based def for NP. (In fact a trivial interactive proof system w/o randomness.)
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9 Nondeterminism

nondeterminism (a non-realistic feature for computation models); NTIME; NP. the equivalence between different views/defs about NP (nondeterminism / verifier).

10 Reductions

These reductions are transitive; P is closed under both reductions, but NP only under mapping reduction.

11 NP-Completeness and the Cook-Levin Theorem

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\mathsf{NP}\text{-}hard; \ \mathsf{NP}\text{-}complete. \mathsf{Cook}\text{-}\mathbf{Levin} \mathsf{Theorem}: \ \forall L \in \mathsf{NP}: L \leq_m^\mathsf{P} \mathsf{SAT}. (proof idea: computation is local.) Circuit-based proof of Cook-Levin. Reading: boolean circuit, circuit family, size/depth complexity.
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12 NP-Completeness Reductions

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3SAT \leq_m^P E3SAT \leq_m^P NAE-3SAT \leq_m^P 3COL \leq_m^P INDSET; CSP; gadget.
Reading: NP-complete problems, VERTEX-COVER, HAMPATH, UHAMPATH, SUBSET-SUM.
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13 Search-to-Decision, Padding, Dichotomy Theorems

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SAT is downward-self-reducible; search-to-decision reduction. padding (common technique: add meaningless contents to blow input size up); Dichotomy Theorem: Every boolean CSP is either in P (2SAT, XOR-SAT, HornSAT) or NP-complete. Dichtomy Conjecture: every CSP is either in P or NP-complete.
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14 Ladner's Theorem and Mahaney's Theorem

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Ladner's Theorem: P \neq NP \implies \exists L \in NP \backslash P \text{ s.t. } L \text{ is not NP-complete.}
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On class gives a weaker proof assuming ETH, with the idea basically the same: "water down" SAT to make it not so hard, thus get a valid algo for SAT which contradicts the assumption.

Mahaney's Theorem: $P \neq NP \implies sparse NP$ -complete language not exists.

Proof idea: reduce SAT to L, the width of DSR's search tree is poly (restricted by L), thus SAT $\in P$.

15 coNP

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coNP; P is closed under complement; UNSAT is coNP-complete. L \in \mathsf{NP} \cap \mathsf{coNP} has "good characterization".
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16 Space Complexity

model for space complexity: individual work tape; space complexity class $\mathsf{SPACE}(f(n))$, L, PSPACE. Savitch's Theorem: $\mathsf{ST-PATH} \in \mathsf{SPACE}(\log^2 n)$. (for generalized version, see lec17) **Space Hierarchy Theorem**. (proof is similar to THT: diagonalization method)

17 Savitch's Theorem

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Savitch's Theorem: \forall f(n) \geq n : \mathsf{NSPACE}(f(n)) \subseteq \mathsf{SPACE}(f^2(n)). (proof idea: "middle-first search") Nondeterminism-based def of NL: \mathsf{NL} = \mathsf{NSPACE}(\log n); \mathsf{ST-PATH} \in \mathsf{NL}. Reading: \mathsf{NL} \subseteq \mathsf{SPACE}(\log^2 n), \mathsf{NL} \subseteq \mathsf{P}; Can extend Savitch's Theorem to f(n) \geq \log n.
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18 NL-completeness and Logspace Reductions

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\forall f(n) \geq \log n : \mathsf{NSPACE}(f(n)) \subseteq \mathsf{TIME}(2^{O(f(n))}), \mathsf{NSPACE}(f(n)) \subseteq \mathsf{SPACE}(f(n)^2). logspace reduction (to reason about logspace classes, so must be as weak as logspace): closure, transitivity. (logspace algos are usually insane since they can use crazy time for saving space.) ST-PATH is NL-complete.
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19 From P-completeness to PSPACE-completeness

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P-complete languages: HornSAT, LP, CKT-EVAL;
Empirically, polytime reduction implies logspace reduction. (also Cook-Levin's)
TQBF is PSPACE-complete. (proof idea: use quantifiers to reduce size from exp to poly)
(PSPACE's essence seems to be "games", like TQBF.)
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20 The Immerman-Szelepcsényi Theorem

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NPSPACE = coNPSPACE = PSPACE by Savitch's. What about scaling down to logspace? Immerman-Szelepcsényi Theorem: \overline{\mathsf{ST-PATH}} \in \mathsf{NL}(\mathsf{NL} = \mathsf{coNL}). Proof idea: one step each time from s to exploit the locality of "certificates".
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21 Randomized Complexity: RP, coRP, and ZPP

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new computation resource / power: randomness; can greatly reduce running time with error allowed. randomized\ time\ complexity\ class\ (one-sided\ error): RTIME(f(n)), RP, coRP. zero\text{-}sided\ error: ZPTIME(f(n)), ZPP. COMPOSITES, PRIMES \in RP; PRIMES \in coRP; P \subseteq RP \subseteq NP, P \subseteq coRP \subseteq coNP; ZPP = RP \cap coRP. Reading: amplification lemma.
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22 BPP

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two-sided error: \mathsf{BPTIME}(f(n)), \mathsf{BPP}. (named by "bounded error probabilistic time") \mathsf{BPP} \subseteq \mathsf{PSPACE} \subseteq \mathsf{EXP}; \mathsf{P} = \mathsf{NP} \implies \mathsf{P} = \mathsf{BPP}; \mathsf{BPP} \subseteq \mathsf{P/poly}. (by amplification, then union bound) derandomization result: If \mathsf{3SAT} requires \mathsf{SIZE}(2^{\delta n}) for some \delta > 0, then \mathsf{P} = \mathsf{BPP}. (worst-case hardness \implies strong average-case hardness \implies PRNG) Reading: Read-Once Branching Programs, EQ_{\mathsf{ROBP}} \in \mathsf{BPP} (proof idea: arithmetization on \mathbb{F}_p.)
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23 The Polynomial Hierarchy

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\begin{array}{l} \mathsf{P} = \mathsf{NP} \implies \mathsf{NP} = \mathsf{coNP}; \ \mathrm{quantifier\text{-}based} \ \mathrm{def} \ \mathrm{of} \ \mathsf{PH}; \ \Sigma_i\text{-}\mathsf{SAT} \ \mathrm{is} \ \Sigma_i\mathsf{P}\text{-}\mathrm{complete}. \\ \mathsf{PH} \ ``collapses" \ \mathrm{to} \ \mathrm{the} \ i\text{-}\mathrm{th} \ \mathrm{level} \ \mathrm{if} \ \Sigma_i\mathsf{P} = \Pi_i\mathsf{P}. \ (\mathrm{e.g.}, \ \mathrm{if} \ \mathsf{P} = \mathsf{NP} \ \mathrm{or} \ \mathsf{NP} = \mathsf{coNP}) \\ \mathsf{Reading} : \ Alternating \ TM, \ \mathsf{ATIME}, \ \mathsf{ASPACE}, \ \mathsf{AP}, \ \mathsf{APSPACE}, \ \mathsf{AL}. \\ \forall f(n) \geq n : \ \mathsf{ATIME}(f(n)) \subseteq \mathsf{SPACE}(f(n)) \subseteq \mathsf{ATIME}(f^2(n)); \ \forall f(n) \geq \log n : \ \mathsf{ASPACE}(f(n)) = \mathsf{TIME}(2^{O(f(n))}). \\ \mathsf{corollary} : \ \mathsf{AL} = \mathsf{P}, \ \mathsf{AP} = \mathsf{PSPACE}, \ \mathsf{APSPACE} = \mathsf{EXP}. \end{array}
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24 Oracle TMs & P^{NP}

Oracle TM; $\mathsf{P}^{\mathsf{NP}} \subseteq \Sigma_2 \mathsf{P}$ (and thus $\mathsf{P}^{\mathsf{NP}} = \mathsf{co} \mathsf{-} \mathsf{P}^{\mathsf{NP}} \subseteq \mathsf{co} \mathsf{-} \Sigma_2 \mathsf{P} = \Pi_2 \mathsf{P}$). In fact P^{NP} is just $\Delta_2 \mathsf{P} = \Sigma_2 \mathsf{P} \cap \Pi_2 \mathsf{P}$.

25 Interactive Proofs: IP = PSPACE

A glimpse past the 80s from this lec; from mid 60s to the end of 80s (space/time/randomness) in prev. Interactive Proof System (interaction + randomness): IP[k], IP = IP[poly(n)]. $NP \subseteq IP[1]$; $BPP \subseteq IP[0]$; $IP \subseteq PSPACE$.

The two-sided error def of IP can be automatically upgraded to one-sided error. [Shamir '89] TQBF \in IP is motivated by [LFKN '89] #3SAT \in ' IP. (proof idea: arithmetization on \mathbb{F}_p .) Reading: IP is kind of a probabilistic analog of NP, like the probabilistic analog RP of P.

26 Beyond Worst-Case Analysis

Assume $P \neq NP$ in next lecs. (Thus $3SAT \notin P$.)

To solve 3SAT: 1. allow more time; 2. allow errors (correct for $\geq 99\%$ inputs); 3. approx algos (satisfies 90% clauses for every input). Dream is to show hardness for these 3 possibilities, just with $P \neq NP$.

27 Hardness within P

What is really efficient maybe not P, but $O(n \operatorname{polylog}(n))$; For these the $\operatorname{\underline{model}}$ matters! (RAM here) SETH $\Longrightarrow \forall \epsilon > 0 : \mathsf{LCS} \notin \mathsf{TIME}(n^{2-\epsilon})$. Similar results for 3SUM, APSP, k-CLIQUE, etc. fine-grained complexity; care about the exact complexity of reductions, not just P. some reductions between problems, with assumptions SETH / CNF-SETH.

28 Why is P vs. NP difficult?

History of P vs. NP, & some negative results: HALTS not computable; THT; both use diagonalization. **Baker-Gill-Solovay's Theorem**: $\exists A, B : \mathsf{P}^A = \mathsf{NP}^A, \mathsf{P}^B \neq \mathsf{NP}^B$. (negative result about *proof tech!*) Proof idea: $A = \mathsf{TQBF}$, construct B using diagonalization.

Approaches after BGS: In '80s theorists try to prove harder statements; In '88 Hästad shows ckt lower-bound for PARITY; In '94 Razborov shows limitations of "natural" proof strategy.

Assuming good PRNG exists, ∄ "natural" proof that NP has no poly-size ckts.