

# COMS3261: Computer Science Theory

Fall 2013

Mihalis Yannakakis

Lecture 24, 12/2/13

## Time Complexity

- Time complexity of a Deterministic TM  $M$ :  
 $T(n) = \max \# \text{ steps of } M \text{ on inputs of length } \leq n$   
worst-case complexity:  $\max$
- Asymptotics: Big-Oh, small-o notation
- For  $f, g$  positive functions on positive integers,  
 $f = O(g)$  if  $\exists$  constant  $c$ , integer  $n_0$  s.t.  $f(n) \leq cg(n) \forall n \geq n_0$
- Example:  $3n^2 + 7n = O(n^2) = O(n^3)$
- $f = \Theta(g)$  if  $\exists$  constants  $c_1, c_2$ , integer  $n_0$  s.t.  $c_1g(n) \leq f(n) \leq c_2g(n) \forall n \geq n_0$

## Time Complexity Class

- Let  $t: \mathbb{N} \rightarrow \mathbb{N}$
- $\text{TIME}(t(n)) = \{ L \mid \text{language } L \text{ decided by a (multitape) DTM in time } O(t(n)) \}$
- i.e.  $L \in \text{TIME}(t(n)) \Leftrightarrow \exists \text{ DTM } M \exists \text{ constant } c, \text{ integer } n_0 \text{ s.t. } L(M)=L \text{ and } T_M(n) \leq ct(n) \forall n \geq n_0$
- Similarly for other computational problems that are not decision (language) problems
- Are allowed to use multitape TM
- $O(t(n))$  in definition because of **linear speed-up theorem**: can use a TM with more states and tape symbols to speed up computation by a constant factor (above  $O(n)$  time)

## Time Complexity Class

- Are allowed to use multitape TM
- If  $L$  in  $\text{TIME}(t(n))$  then recognized in  $O(t^2(n))$  time by 1-tape TMs
- Simulation of real computers: if  $t(n)$  on RC then  $O(t^3(n))$  by multitape Turing machine (generous – more carefully it is more like  $t^2(n)$  – effects of random access memory versus sequential access as in tape of TMs)
- Different models differ in time complexity by small polynomial changes (e.g.  $t$  vs  $t^2$ )
- Important for low complexities ( $n$ ,  $n \log n$ ,  $n^2$ ), but not important if we want to distinguish polynomial vs. exponential

## Polynomial Time

- **polynomial examples:**  $n$ ,  $n^2$ ,  $3n^2+7n+2$ ,  $6n^4+3n^2$ ,  $n^{100}$
- $p(n)=\Theta(n^d)$  where  $d$  the degree of the polynomial (leading term)
- Defn:  $P = \bigcup_{k=1}^{\infty} TIME(n^k)$
- All reasonable models of computation can simulate each other with polynomial overhead
- **Robust class**
- Can use 1-tape TM for lower bounds
- Can use real computer (RAM) for upper bounds (algorithms)
- Robust also with respect to reasonable input representations

## Examples of Problems in P (1)

- **Sorting** a set of numbers, strings etc:  $n \log n$
- **Membership in a regular language:** linear in length  $n$  of string  
(example apps: pattern matching, lexical analysis, etc)  
- polynomial time also with respect to the size of the representation of the language, for all types of representations (DFA, NFA,  $\epsilon$ -NFA, Regular expression)
- **Membership in a context-free language:**  $n^3$  ( $n$  for DCFL) for string of length  $n$   
(ex. parsing)

## Examples of Problems in P (2)

- Graph problems
  - Graph Reachability, Cycle detection, Shortest paths, Minimum Spanning Tree...
  - Representation of graphs: by adjacency matrix, or adjacency lists, or set of nodes and edges.
    - Important for exact complexity, but all reasonable representations polynomially related, so exact representation does not matter for membership in P

## Examples of Problems in P (3)

- Problems with numbers: numbers represented in binary: size = #bits (magnitude of number exponential in size!)
- Basic operations on numbers: +, -, \*, /
- Divisibility : given  $a, b$  in  $\mathbb{Z}$ , does  $a|b$ ?
- Greatest common divisor (Euclidean algorithm)
- Naive algorithm (try every  $d \leq a, b$ , pick max) not polynomial
- Primality, Compositeness. In P
- Factorization: Given number  $a$ , compute its prime factors.  
Big open problem.
- Basis of cryptographic schemes: assumption: not in P

## Examples of Problems in P (4)

- **Linear algebra:** Solution of linear equations, matrix operations (inversion, multiplication etc)
- Solution of linear inequalities and linear optimization (**Linear Programming**)

## Nondeterministic Time Complexity

- For a Nondeterministic Turing machine  $M$  (that always halts), time complexity  $T_M(n) = \max \# \text{ steps used on any branch of the computation on any input of length } \leq n$ .  
i.e. time  $f(n)$  means that  $M$  halts in time  $\leq f(n)$  for *all* computations on *all inputs* of length  $\leq n$  ; not only accepting computations.
- (If some branch does not halt then time = infinite)
- **NTIME( $t(n)$ )** =  $\{ L \mid \text{language } L \text{ accepted by a (multitape) NTM with time complexity } O(t(n)) \}$

## Nondeterministic vs. Deterministic Time

- Let  $t(n) \geq n$ .
- Every  $t(n)$  time NTM has an equivalent (1-tape) DTM with time complexity  $c^{t(n)}$  for some constant  $c$  (equivalently time  $2^{O(t(n))}$  )
- Showed this in the simulation of NTM with DTM.
- **Big Open Question:** Is the exponential blowup unavoidable?
- Especially important in the special case of polynomial time complexity: **P=NP ? question**

## Nondeterministic Polynomial Time: NP

- **Definition of NP:** Class of languages  $L$  (decision problems: Yes/No problems) that can be accepted by some Nondeterministic Turing Machine that runs in polynomial time, i.e. time  $O(n^k)$  for some constant  $k$ .

$$NP = \bigcup_{k=1}^{\infty} NTIME(n^k)$$

- **Alternative equivalent view of NP:** Using certificates (witnesses, proofs,...): All guessing done up front at the beginning.

## Alternative view of NP: Certificates

- **Theorem:** A language  $L \subseteq A^*$  is in NP iff there is a 2-ary relation  $R \subseteq A^* \times B^*$  ( $B$  could be  $A$  or  $\{0,1\}$  or any alphabet) that is
  1. **polynomially balanced:**  $R(x,y) \Rightarrow |y| \leq |x|^c$  for some  $c$   
(could also require  $=$  instead of  $\leq$ , equivalent)
  2. **polynomially decidable:** there is a polynomial-time (deterministic) TM that accepts  $(x,y)$  iff  $R(x,y)$such that  $L = \{ x \mid \exists y \text{ s.t. } R(x,y) \}$

## Alternative view of NP: Certificates

$$L = \{ x \mid \exists y \text{ s.t. } R(x,y) \}$$

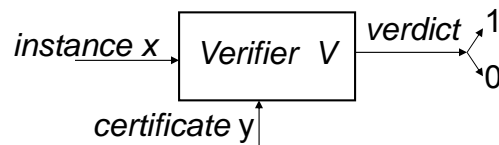
Means:

- An input  $x$  is in  $L$  iff there is a **certificate** (witness, solution, proof)  $y$  such that  $R(x,y)$  holds.
- Two important properties:
  1. certificates are **short** (polynomially bounded)
  2. certificates are **easy to check** (polynomial-time checkable)

## NP = short, verifiable certificates

- **Verifier V**: Polynomial time algorithm for the relation  $R(x,y)$ ; V checks that y is a certificate for x

$$L = \{x \in A^* \mid \exists y \in B^*, |y| \leq |x|^c, V(x,y) = 1\}$$



- $x \in L \Rightarrow \exists y \in B^*, |y| \leq |x|^c \text{ s.t. } V(x,y) = 1$
- $x \notin L \Rightarrow \forall y \in B^*, |y| \leq |x|^c. V(x,y) = 0$

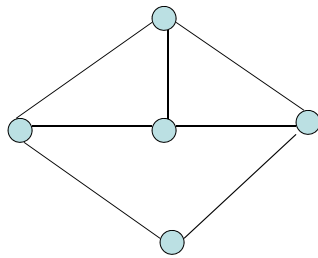
## Proof

1. Suppose L is accepted by a polynomial time NTM M.  
certificate y = sequence of choices of M that makes M accept  
(Could also use certificate = complete accepting computation )  
 $R(x,y)$ : M accepts input x with sequence of choices y
2. Suppose L has a relation R that is polynomially balanced and polynomially decidable. Then L accepted by NTM M that guesses first a certificate y of length  $\leq |x|^c$  and then verifies that  $R(x,y)$  holds.



## Examples

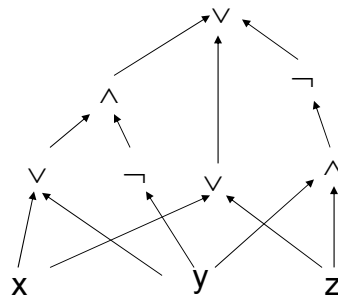
- **Graph 3 Colorability:** Instance: Graph G  
certificate  $y$  = assignment of a “color”  $\in \{1,2,3\}$  to each node  
s.t. adjacent nodes are assigned different colors
- **Constraint Satisfaction problems**, e.g. schedule a set of events (eg. exams) in a given number of slots so that no conflicts



## Example: Boolean Circuit Satisfiability

- Input: Boolean (combinational) circuit using AND, OR, NOT gates with 1 output, many inputs
- Output: Yes, if there is an assignment of True/False (1/0) to the inputs that makes the circuit output true (1)

Example:



$x=y=z=0$  satisfies the circuit

## Fundamental Question

$$P = NP?$$

Is it always as easy to generate a proof/certificate as it is to check a proof/certificate that is given to us?