

# The class P: polynomial time

- Theorems 1 and 2 illustrate an important distinction.
- On the one hand, we demonstrated at most a square or polynomial difference between the time complexity of problems measured on deterministic single tape and multi-tape Turing machines.
- On the other hand, we showed at most an exponential difference between the time complexity of the problems on deterministic and non-deterministic Turing machines.
- For our purpose, polynomial difference in running time are considered to be small, whereas exponential differences are considered to be large.
- Polynomial time algorithms are fast enough for many purposes, but exponential time algorithms rarely are useful. (For  $n=1000$ ,  $n^3 = 1$  billion (still manageable number),  $2^n$  is much larger than the number of atoms in the universe. )
- All reasonable deterministic computational models are polynomially equivalent. Any one of them can simulate another with only a polynomial increase in running time.
- From here on we focus on aspects of time complexity theory that are unaffected by polynomial difference in running time. We consider such differences to be insignificant and ignore them.
- *The Question is* whether a given problem is polynomial or non-polynomial.
- So we came to an important definition in the complexity theory, **P** class.

# The class P: definition

- **Definition:** **P** is the class of languages that are decidable in polynomial time on a deterministic single tape Turing machine. That is

$$P = \bigcup_k \text{TIME}(n^k).$$

- The class **P** plays an important role in our theory and is important because
  - **P** is invariant for all models of computation that are polynomially equivalent to the deterministic single tape TM, and
  - **P** roughly corresponds to the class of problems that are realistically solvable on a computer.
- When we analyze an algorithm to show that it runs in polynomial time, we need to do two things
  - First, give a polynomial upper bound (usually in big-O notation) on the number of stages that the algorithm uses when it runs on input of length  $n$ .
  - Then, examine the individual stages in the description of the algorithm to be sure that each can be implemented in polynomial time on a reasonable deterministic model.
- When both tasks have been done, we can conclude that it runs in polynomial time because we have demonstrated that it runs for a polynomial number of stages, each of which can be done in polynomial time, and the composition of polynomials is a polynomial.

# Examples of problems in P

- We had: the problem whether  $w$  is a member of the language  $A = \{0^k 1^k : k \geq 0\}$  is in **P**.
- Fortunately, there are many problems that are in **P**.
- The *PATH* problem is to determine whether a directed path exists from  $s$  to  $t$ .

$PATH(G, s, t) = \{ \langle G, s, t \rangle : G \text{ is a directed graph that has a directed path from } s \text{ to } t \}$ .

**Theorem:**  $PATH \in P$ .

- we use *breadth first search* and successively mark all nodes in  $G$  that are reachable from  $s$  by directed paths of length 1, then 2, then 3, through  $m=|V|$ .

$M =$  “on  $\langle G, s, t \rangle$ : where  $G$  is a directed graph with nodes  $s$  and  $t$ .

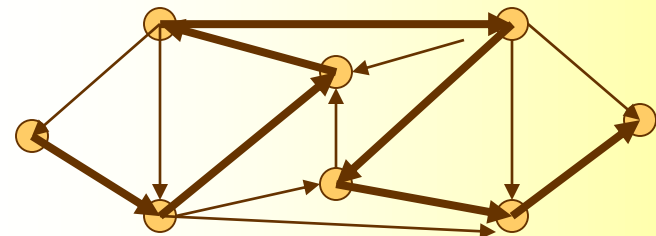
1. Place a mark on node  $s$ .
2. Repeat the following until no additional nodes get marked.
3. Scan all the edges of  $G$ . If an edge  $(a, b)$  is found going from marked node  $a$  to an unmarked node  $b$ , mark  $b$ .
4. If  $t$  is marked, *accept*; otherwise *reject*.”

- **Stages 1, 4** are executed only once. **Stage 3** runs at most  $m=|V|$  times because each time except the last it marks an additional node in  $G$ . Hence, the **total number of stages** is  $1+1+m$ , giving a polynomial in the size of  $G$ .
- **Stages 1, 4** easily implemented in polynomial time on any reasonable deterministic model. **Stage 3** involves a scan of the input and a test whether certain nodes are marked, which also is easily implemented in polynomial time.
- Hence,  $M$  is a polynomial time algorithm for *PATH*.

# The class NP

- For some interesting and useful problems, polynomial time algorithms that solve them aren't known to exist.
- Why have we been unsuccessful in finding polynomial time algorithms for these problems? We don't know the answer to this important question.
- Perhaps these problems have, as yet undiscovered, polynomial time algorithms that rest on unknown principles.
- Or possibly some of these problems simply cannot be solved in polynomial time. They may be intrinsically difficult.
- One remarkable discovery concerning this question shows that the complexities of many problems are linked. The discovery of a polynomial time algorithm for one such problem can be used to solve an entire class of problems.
- A ***Hamiltonian path*** in a directed graph  $G$  is a directed path that goes through each node exactly once. Consider the problem of testing whether a directed graph contains a Hamiltonian path connecting two specified nodes.
- We can easily obtain an exponential time algorithm for the *HAMPATH* problem by brute-force approach which checks all possible permutations of nodes ( $n!$ ).
- We need only add a check to verify that the potential path is Hamiltonian.
- No one knows whether *HAMPATH* is solvable in polynomial time.

$HAMPATH = \{ \langle G, s, t \rangle : G \text{ is a directed graph with a Hamiltonian path from } s \text{ to } t \}$ .



# The class NP: definition

- Define the **non-deterministic time complexity class**

$NTIME(t(n)) = \{L : L \text{ is a language decided by an } O(t(n)) \text{ time Non-Deterministic Turing machine}\}.$

- **Def:** **NP** is the class of languages that are decidable in polynomial time on a non-deterministic Turing machine. That is

$$NP = \bigcup_k NTIME(n^k).$$

- The class **NP** is insensitive to the choice of reasonable non-deterministic computation model because all such models are polynomially equivalent.

**Theorem:**  $HAMPATH \in NP.$

- The following is a non-deterministic Turing Machine (NTM) that decides the *HAMPATH* problem in non-deterministic polynomial time (we defined the time of a non-deterministic machine to be the time used by the longest computation branch).

$N =$  “on  $\langle G, s, t \rangle$ : where  $G$  is a directed graph with nodes  $s$  and  $t$ .

1. Write a list of  $m$  numbers  $p_1, p_2, \dots, p_m$ , where  $m$  is the number of nodes in  $G$ . Each number in the list is non-deterministically selected to be between 1 and  $m$ .
2. Check for repetitions in the list. If any are found, *reject*.
3. Check whether  $s = p_1$  and  $t = p_m$ . If either fail, *reject*.
4. For each  $i$  between 1 and  $m-1$ , check whether  $(p_i, p_{i+1})$  is an edge of  $G$ . If any are not, *reject*. Otherwise, *accept*.”

- Clearly, this algorithms runs in non-deterministic polynomial time since all stages run in polynomial time.

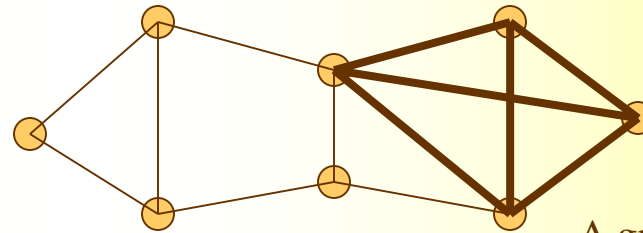
# Polynomial Time Verifiers

- The *HAMPATH* problem does have a feature called *polynomial verifiability* that is important for understanding its complexity.
- Even though we don't know of a fast (i.e., polynomial time) way to determine whether a graph contains a Hamiltonian path, if such a path were discovered somehow (perhaps using the exponential time algorithm), we could easily convince someone else of its existence, simply by presenting it.
- In other words, *verifying* the existence of a Hamiltonian path may be much easier than *determining* its existence.
- We can give an equivalent definition of the **NP** class using the notion *verifier*.
- A *verifier* for a language  $A$  is an algorithm  $V$ , where
$$A = \{w : V \text{ accepts } \langle w, c \rangle \text{ for some string } c\}.$$
- A verifier uses additional information, represented by the symbol  $c$  in definition. This information is called a *certificate*, or *proof*, of membership in  $A$ .
- Example:  $\langle G, s, t \rangle$  belongs to *HAMPATH* if for some path  $p$ ,  $V$  accepts  $\langle \langle G, s, t \rangle, p \rangle$  (that is,  $V$  says “yes,  $p$  is a Hamiltonian path from  $s$  to  $t$  of  $G$ ). For the *HAMPATH* problem, a certificate for a string  $\langle G, s, t \rangle \in \text{HAMPATH}$  simply is the Hamiltonian path  $p$  from  $s$  to  $t$ .
- A *polynomial time verifier* is a verifier that runs in polynomial time in the length of  $w$ .
- A language  $A$  is *polynomially verifiable* if it has a polynomial time verifier.
- **Def:** **NP** is the class of languages that have polynomial time verifiers.
- **The verifier can check in polynomial time that the input is in the language when it is given the certificate.**

# CLIQUE is in NP

- A **clique** in an undirected graph  $G$  is a subgraph, wherein every two nodes are connected by an edge. A  **$k$ -clique** is a clique that contains  $k$  nodes.
- The **clique problem** is to determine whether a graph contains a clique of a specific size.

$CLIQUE = \{ \langle G, k \rangle : G \text{ is an undirected graph with a } k\text{-clique} \}$ .



A graph with 4-clique.

**Theorem:**  $CLIQUE \in NP$ .

- **Proof:** The following is a verifier  $V$  for  $CLIQUE$ .

$V =$  “on input  $\langle \langle G, k \rangle, c \rangle$ :

1. Test whether  $c$  is a set of  $k$  nodes in  $G$ .
2. Test whether  $G$  contains all edges connecting nodes in  $c$ .
3. If both pass, *accept*; otherwise, *reject*.”

- **Alternative proof:** If you prefer to think of **NP** in terms of non-deterministic polynomial Turing machine ...

$N =$  “on  $\langle G, k \rangle$ : where  $G$  is an undirected graph,  $k$  is an integer.

1. Non-deterministically select a subset  $c$  of  $k$  nodes in  $G$ .
2. Test whether  $G$  contains all edges connecting nodes in  $c$ .
3. If yes, *accept*; otherwise, *reject*.”

# SUBSET-SUM is in NP

- We have a collection of numbers,  $x_1, x_2, \dots, x_k$ , and a target number  $t$ . We want to determine whether the collection contains a subcollection that adds up to  $t$ .

$SUBSET - SUM = \{ \langle S, t \rangle : S = \{x_1, x_2, \dots, x_k\} \text{ and} \\ \text{for some } \{y_1, y_2, \dots, y_l\} \subseteq \{x_1, x_2, \dots, x_k\}, \text{ we have } \sum y_i = t \}.$

- For example  $\langle \{4, 11, 16, 21, 27\}, 25 \rangle$  is in  $SUBSET-SUM$  since  $4+21=25$ .
- Note that  $\{x_1, x_2, \dots, x_k\}$  and  $\{y_1, y_2, \dots, y_l\}$  are multisets (we allow repetitions).

**Theorem:**  $SUBSET - SUM \in NP$ .

- **Proof:** The following is a verifier  $V$  for  $SUBSET-SUM$ .

$V =$  “on input  $\langle \langle S, t \rangle, c \rangle$ :

1. Test whether  $c$  is a collection of numbers that sum to  $t$ .
2. Test whether  $S$  contains all the numbers in  $c$ .
3. If both pass, *accept*; otherwise, *reject*.”

- **Alternative proof:** If you prefer to think of **NP** in terms of non-deterministic polynomial Turing machine ...

$N =$  “on  $\langle S, t \rangle$ :

1. Non-deterministically select a subset  $c$  of the numbers in  $S$ .
2. Test whether  $c$  is a collection of numbers that sum to  $t$ .
3. If yes, *accept*; otherwise, *reject*.”



# The P versus NP question

**P** = the class of languages that are decidable by polynomial time *deterministic* TMs.

**NP** = the class of languages that are decidable by polynomial time *non-deterministic* TMs.

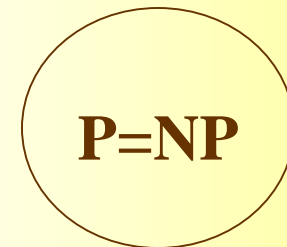
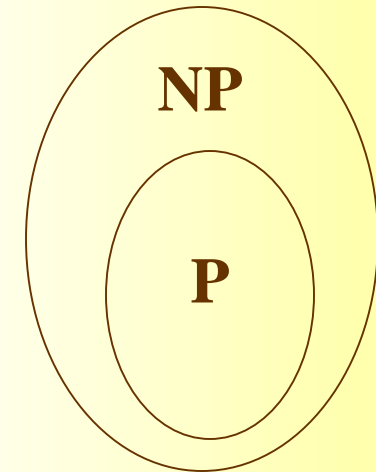
OR EQUIVALENTLY

**P** = the class of languages where membership can be *decided* quickly (in pol. time).

**NP** = the class of languages where membership can be *verified* quickly (in pol. time).

- We presented examples of languages, such as *HAMPATH* and *CLIQUE*, that are members of **NP** but that are not known to be in **P**.
- No polynomial time algorithms are known for those problems.
- We are unable to *prove* the existence of a single language in **NP** that is not in **P**.
- The *question* of whether **P** = **NP** is one of the greatest unsolved problems in theoretical computer science.
- Most researchers believe that the two classes are not equal because people have invested enormous effort to find polynomial time algorithms for certain problems in **NP**, without success.
- The best method known for solving problems in **NP** deterministically uses exponential time. In other words, one can show that

$$\bigcup_k \text{NTIME}(n^k) = \boxed{\text{NP} \subseteq \text{EXPTIME} = \bigcup_k \text{TIME}(2^{n^k})}.$$



One of these two possibilities is correct.