Chapter 7, Part 2

# **Classes P and NP**

# The Complexity Class P

Juris Hartmanis and Dick Stearns [1965] : proposed *computational complexity* — measuring complexity of problems by the number of steps (or the number of cells) expended in the worst case under the TM model

Fundamental results in the Hartmanis-Stearns paper:

1. Time Hierarchy Theorem (see Section 9.1)  $\cdots$ TIME $(t(n)) \neq$  TIME $(t(n)^2)$  for all reasonable t(n)

# 2. Linear Speed-up Theorem ···

TIME(t(n)) = TIME(ct(n)) for all c > 0 and all reasonable t(n)

A better hierarchy theorem is proven by Harry Lewis and Stearns

# The Complexity Class P (continued)

Alan Cobham [1964], Jack Edmonds [1965], and Michael Rabin [1966] suggested the "**polynomial time**" as a broad classification of problems that are solvable in a *reasonable amount of time*  $\mathbf{P} = \bigcup_{k>0} \text{TIME}(n^k)$ 

Why polynomial, why not, say  $n^3$ ?

Because the "polynomial time" is invariant under the model of computation

**NP** is the nondeterministic counterpart of **P**  $\mathbf{NP} = \bigcup_{k>0} \operatorname{NTIME}(n^k)$ 

### Problems in P

#### **The Path Problem**

Input A directed graph G = (V, E) and  $s, t, 1 \le s, t \le |V|$ 

**Question** Does the graph has a directed path from s to t?

We define PATH to be the set of all positive instances  $\langle G, s, t \rangle$  to the Path Problem.

#### The Path Problem

An encoding of a graph can be its **adjacency matrix**  $(a_{ij})$ : for every  $i, j, 1 \le i, j \le n$ ,  $a_{ij} = 1$  if  $(i, j) \in E$  and 0 otherwise

The entire encoding can be

 $0^n \# a_1 a_2 \cdots a_n \# 0^s \# 1^t$ ,

where  $a_1, a_2, \ldots, a_n$  are the rows of the adjacency matrix

# **A Polynomial Time Algorithm for** *PATH*

Let G = (V, E) be an instance of PATH, n = |V|, and A the adjacency matrix of G.

For each  $k \ge 1$ , let  $A^{(k)}$  be the k-th power of the matrix A, where  $\lor$  and  $\land$  replace + and  $\times$ .

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For each  $k \ge 1$ , let  $A^{(k)}$  be the k-th power of the matrix A, where  $\lor$  and  $\land$  replace + and  $\times$ .

Then for every  $k \ge 1$  and every  $i, j, 1 \le i, j \le n$ , the (i, j)th entry of  $A^{(k)}$  is a 1 if and only if there is a directed path from i to j of length at most k in G.

# Algorithm for *PATH*

- Compute  $B = A^{(n)}$  by iterative multiplication; that is, compute  $A^{(1)} = A, A^{(2)}, A^{(3)}, A^{(4)}, \ldots$  by multiplying A by the previous matrix.
- if the (s,t)th entry of B = 1 accept ; else reject

# Running Time on a Multi-tape Turing Machine

- We have only to compute the transpose of  $A^{(i)}$ .
  - The initial one can be obtained by transposing the input matrix A, which requires  ${\cal O}(n^3)$  steps.
  - For the other matrices, that is done by controlling the order in which the entries are computed.
- There are  $n^2$  entries per matrix.
- There are n-1 matrix multiplications.
- 2n bits are examined to compute an entry of one product.

Thus, the running time is  ${\cal O}(n^4)$  step algorithm

#### **Testing Relative Primality of Two Numbers**

# The Relative Primality Problem

Input Integers  $x, y \ge 1$ .

Question Are x and y relatively prime to each other, i.e., gcd(x,y) = 1?

Define RELPRIME to be the set of all positive instances  $\langle x, y \rangle$  of the Relative Primality Problem.

**Note:** x and y should not be encoded in unary

# **A Polynomial Time Algorithm for** *RELPRIME*

Use the Euclidean Algorithm: On input  $\langle x, y \rangle$ :

- 1. repeat  $x \leftarrow x \mod y$ ; swap x and y; until y = 0
- 2. output x

### How Quickly Does x Decrease?

Suppose x has value u, y has value v,  $v \le u$ , after one iteration of the above algorithm, the value of x becomes u' and the value of y becomes v'. and after another iteration of the above algorithm, the value of x becomes u'' and the value of y becomes v''.

We have:

- u' = v,
- if v > u/2, then  $v' = u \mod v = u v < u/2$ ;
- if  $v \le u/2$ , then  $v' \le v 1 < u/2$ .

So, we have

- u'' = v' < u/2,
- v'' < u'/2 = v/2.

This implies that in two iterations, both x and y will be less than half of what they are now.

# **Running Time Analysis**

If  $\max\{|x|, |y|\} = n$ , then the running time is  $O(n^3)$ .

(\*) if the Euclid algorithm on  $\langle x,y\rangle$  outputs 1 then accept ; else reject

The Running Time Analysis:  $O(n^3)$ .

**Polynomial Time Decidability of Context-Free Languages** 

**Theorem.** Every context-free language is in P.

### **Polynomial Time Decidability of Context-Free Languages**

#### **Theorem.** Every context-free language is in P.

**Proof** Let *L* be context-free. Let *G* be a CNF grammar for *L*. Suppose  $w = w_1 \cdots w_n$  be a string whose membership in *L* we are testing.

The case when  $w = \epsilon$  is easy: we accept if and only if  $S \to \epsilon$  is a in G.

#### The Nonempty Case

So, assume  $w \neq \epsilon$  and let  $w_1, \ldots, w_n$  be the symbols of w.

For each  $i, j, 1 \le i \le j \le n$ , let t(i, j) be the set of all variables from which  $w_i \cdots w_j$  can be produced.

# The Nonempty Case

So, assume  $w \neq \epsilon$ .

For each  $i, j, 1 \le i \le j \le n$ , let t(i, j) be the set of all variables from which  $w_i \cdots w_j$  can be produced

We can compute t(i,j) for all i,j,  $1 \le i \le j \le n$ , using dynamic programming.

Then test the membership by examining whether  $S \in t(1, n)$ 

# **Dynamic Programming for Computing the Table**

Set  $t(i,i) \leftarrow$  the set of all A such that  $A \rightarrow w_i$  is in G. Then execute the following:

for 
$$\ell = 2$$
 to  $n$   
for  $i = 1$  to  $n - \ell + 1$   
 $j = i + \ell - 1$ ;  $t(i, j) = \emptyset$ ;  
for  $k = i$  to  $j - 1$   
if  $\exists A, B \in t(i, k), C \in t(k + 1, j)$   
such that  $A \rightarrow BC$  is in  $G$   
then add  $A$  to  $t(i, j)$ 

The running time is  $O(n^3)$  since  $\ell, i$ , and k have at most n possible values.

The size of t(i, j) is at most the number of variables of G, but that is a constant since G is fixed.

# **Examples of NP Languages**

#### **The Hamilton Path Problem**

Input A directed graph G = (V, E) and  $s, t \in V$ ,  $s \neq t$ 

Question Is there a Hamilton Path from s to t in G, i.e., a directed path from s to t that visits all the nodes exactly once?

Define HAMPATH to be the set of all positive instances  $\langle G, s, t \rangle$  to the Hamilton Path Problem.

#### The Class NP

# **The Compositeness Problem**

Input Integer  $x \ge 1$ 

**Question** Does x a composite number, i.e., have an integer divisor other than 1 and x?

Define COMPOSITES to be the set of all composite numbers x.

# A Characterization of NP by Verifiers

A verifier of a language A is an algorithm V such that  $A = \{w \mid V \text{ accepts } \langle w, c \rangle \text{ for some } c\}.$ 

That is, a verifier is an algorithm that takes two inputs w and c and decides whether to accept or reject in such a way that:

- If  $w \in A,$  there is an auxiliary input c that makes the verifier accept and
- If  $w \not\in A$ , there is no auxiliary input c that makes the verifier accept .

For a fixed V, the string c witnessing to  $w \in A$  (that is, one such that V accepts  $\langle w, c \rangle$ ) is called a **certificate** or a **proof**.

# An Alternative Definition of NP

We will measure the time of V in terms of the length of w.

# **Definition.** (alternate) NP is the class of languages that have polynomial time verifiers.

This means that polynomial time verifies reject (input, proofcandidate) pairs in which the proof candidate is exceedingly long.

### Equivalence Between the Two Definitions of $\ensuremath{\mathbf{NP}}$

# **Theorem.** The alternative definition is equivalent to the first definition of NP.

**Proof** (Sketch) If L has a polynomial time verifier, then we can construct a nondeterministic Turing machine that nondeterministic guesses a proof of length bounded by some fixed polynomial and then verifies the proof.

If L is accepted by a polynomial time nondeterministic Turing machine, we can use the accepting computation paths of the machine as the proofs of membership.

# Membership of HAMPATH in NP

Define a certificate for each  $\langle G, s, t \rangle \in HAMPATH$  to be any sequence  $\langle v_1, ..., v_n \rangle$  of nodes such that (i) for every  $i, 1 \leq i \leq n, i = v_j$  for some j, (ii)  $s = v_1$ , (iii)  $t = v_n$ , and (iv) for every  $i, 1 \leq i \leq n - 1$ ,  $(v_i, v_{i+1}) \in E$ .

A correct certificate can be of length  $O(n \log n)$  and verification can be done in  $O(n^3)$  steps.

#### Membership in NP

Define a certificate for each  $x \in COMPOSITES$  to be any number y such that y divides x and 1 < y < x. Then a correct certificate can be of **length** O(n)

#### **The Clique Problem**

Input A graph G = (V, E) and  $k \ge 1$ .

**Question** Does G contain a complete graph of size  $\geq k$ ?

Define CLIQUE to be the set of all positive instances  $\langle G, k \rangle$  to the Clique Problem.

# **Theorem.** *CLIQUE* is in NP.

**Proof** (Sketch) Define a certificate for an instance  $\langle G, k \rangle$ , where G is an n node graph, to be an n bit sequence  $c = c_1 \cdots c_n$  such that:

Exactly k of  $c_1, \ldots, c_n$  are 1s and for every  $i, j, 1 \le i < j \le n$ , if  $c_i = c_j = 1$ , then  $(i, j) \in E$ 

Then verification can be done in  $O(n^3)$  steps.

#### More Problems in NP: Subset Sum

# The Subset Sum Problem

**Input** integers  $x_1, \ldots, x_k$  and t

**Question** Is there a subset of  $\{x_1, \ldots, x_k\}$  that adds up to t?

Define SUBSET-SUM to be the set of all positive instances  $\langle S, t \rangle$  to the Subset Sum Problem.

#### Membership in NP

# **Theorem.** SUBSET-SUM is in NP.

**Proof** (Sketch) Define a certificate for an instance  $\langle S, t \rangle$  with |S| = n in SUBSET-SUM to be an n bit sequence such that  $\sum_{i=1}^{n} c_i x_i = t$ 

Then verification can be done in  $O(n^2)$  steps.