## CS 341: Algorithms

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## **Decision Problems**

**Decision Problem:** Given a problem instance I, answer a certain question "yes" or "no".

**Problem Instance:** Input for the specified problem.

Problem Solution: Correct answer ("yes" or "no") for the specified problem instance. I is a ves-instance if the correct answer for the instance I is "yes". I is a **no-instance** if the correct answer for the instance I is "no".

**Size of a problem instance:** Size(I) is the number of bits required to specify (or encode) the instance I.

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# The Complexity Class P

**Algorithm Solving a Decision Problem:** An algorithm A is said to solve a decision problem  $\Pi$  provided that A finds the correct answer ("yes" or "no") for every instance I of  $\Pi$  in finite time.

**Polynomial-time Algorithm:** An algorithm A for a decision problem  $\Pi$  is said to be a polynomial-time algorithm provided that the complexity of A is  $O(n^k)$ , where k is a positive integer and n = Size(I).

The Complexity Class P denotes the set of all decision problems that have polynomial-time algorithms solving them. We write  $\Pi \in P$  if the decision problem  $\Pi$  is in the complexity class P.

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# **Cycles in Graphs**

## **Problem**

## Cycle

**Instance:** An undirected graph G = (V, E).

**Question:** Does G contain a cycle?

### **Problem**

## Hamiltonian Cycle

**Instance:** An undirected graph G = (V, E).

**Question:** Does G contain a hamiltonian cycle?

A hamiltonian cycle is a cycle that passes through every vertex in  ${\cal V}$  exactly once.

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# **Knapsack Problems**

#### **Problem**

#### 0-1 Knapsack-Dec

```
Instance: a list of profits, P = [p_1, \ldots, p_n]; a list of weights, W = [w_1, \ldots, w_n]; a capacity, M; and a target profit, T. Question: Is there an n-tuple [x_1, x_2, \ldots, x_n] \in \{0, 1\}^n such that \sum w_i x_i < M and \sum p_i x_i > T?
```

#### **Problem**

## Rational Knapsack-Dec

**Instance:** a list of profits,  $P = [p_1, \ldots, p_n]$ ; a list of weights,  $W = [w_1, \ldots, w_n]$ ; a capacity, M; and a target profit, T. Question: Is there an n-tuple  $[x_1, x_2, \ldots, x_n] \in [0, 1]^n$  such that

 $\sum w_i x_i \leq M$  and  $\sum p_i x_i \geq T$ ?

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## **Polynomial-time Turing Reductions**

Suppose  $\Pi_1$  and  $\Pi_2$  are problems (not necessarily decision problems). A (hypothetical) algorithm  $\mathbf{A_2}$  to solve  $\Pi_2$  is called an **oracle** for  $\Pi_2$ .

Suppose that A is an algorithm that solves  $\Pi_1$ , assuming the existence of an oracle  $A_2$  for  $\Pi_2$ . ( $A_2$  is used as a subroutine within the algorithm A.)

Then we say that **A** is a **Turing reduction** from  $\Pi_1$  to  $\Pi_2$ , denoted  $\Pi_1 \leq^T \Pi_2$ .

A Turing reduction  $\bf A$  is a polynomial-time Turing reduction if the running time of  $\bf A$  is polynomial, under the assumption that the oracle  $\bf A_2$  has unit cost running time.

If there is a polynomial-time Turing reduction from  $\Pi_1$  to  $\Pi_2$ , we write  $\Pi_1 \leq_P^T \Pi_2$ .

Informally: Existence of a polynomial-time Turing reduction means that if we can solve  $\Pi_2$  in polynomial time, then we can solve  $\Pi_1$  in polynomial time.

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# **Travelling Salesperson Problems**

#### **Problem**

## **TSP-Optimization**

**Instance:** A graph G and edge weights  $w: E \to \mathbb{Z}^+$ .

**Find:** A hamiltonian cycle H in G such that  $w(H) = \sum_{e \in H} w(e)$  is minimized.

#### **Problem**

## **TSP-Optimal Value**

**Instance:** A graph G and edge weights  $w: E \to \mathbb{Z}^+$ .

Find: The minimum T such that there exists a hamiltonian cycle H in G

with w(H) = T.

#### **Problem**

#### **TSP-Decision**

**Instance:** A graph G, edge weights  $w: E \to \mathbb{Z}^+$ , and a target T.

**Question:** Does there exist a hamiltonian cycle H in G with  $w(H) \leq T$ ?

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# **TSP-Optimal Value** $\leq_{\mathcal{D}}^T$ **TSP-Dec**

```
Algorithm: TSP-OptimalValue-Solver(G, w)
 external TSP-Dec-Solver
 hi \leftarrow \sum_{e \in E} w(e)
  lo \leftarrow 0
 if not TSP-Dec-Solver(G, w, hi) then return (\infty)
 while hi > lo
    \mathbf{do} \ \begin{cases} mid \leftarrow \left \lfloor \frac{hi + lo}{2} \right \rfloor \\ \mathbf{if} \ TSP\text{-}Dec\text{-}Solver(G, w, mid) \\ \mathbf{then} \ hi \leftarrow mid \\ \mathbf{else} \ lo \leftarrow mid + 1 \end{cases}
  return (hi)
```

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# **TSP-Optimization** $\leq_{\mathcal{D}}^T$ **TSP-Dec**

```
Algorithm: TSP-Optimization-Solver(G = (V, E), w)
external TSP-OptimalValue-Solver, TSP-Dec-Solver
T^* \leftarrow TSP-OptimalValue-Solver(G, w)
if T^* = \infty then return ("no hamiltonian cycle exists")
w_0 \leftarrow w
H \leftarrow \emptyset
 for all e \in E
 return (H)
```

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## **Certificates**

**Certificate:** Informally, a certificate for a yes-instance I is some "extra information" C which makes it easy to **verify** that I is a yes-instance.

**Certificate Verification Algorithm:** Suppose that Ver is an algorithm that verifies certificates for yes-instances. Then Ver(I,C) outputs "yes" if I is a yes-instance and C is a valid certificate for I. If Ver(I,C) outputs "no", then either I is a no-instance, or I is a yes-instance and C is an invalid certificate.

Polynomial-time Certificate Verification Algorithm: A certificate verification algorithm Ver is a polynomial-time certificate verification algorithm if the complexity of Ver is  $O(n^k)$ , where k is a positive integer and n = Size(I).

# The Complexity Class NP

Certificate Verification Algorithm for a Decision Problem: A certificate verification algorithm Ver is said to solve a decision problem  $\Pi$  provided that

- for every yes-instance I, there exists a certificate C such that Ver(I,C) outputs "yes".
- for every no-instance I and for every certificate C, Ver(I,C) outputs "no".

The Complexity Class NP denotes the set of all decision problems that have polynomial-time certificate verification algorithms solving them. We write  $\Pi \in NP$  if the decision problem  $\Pi$  is in the complexity class NP.

Finding Certificates vs Verifying Certificates: It is not required to be able to find a certificate C for a yes-instance in polynomial time in order to say that a decision problem  $\Pi \in NP$ 

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# Certificate Verification Algorithm for Hamiltonian Cycle

A certificate consists of an n-tuple,  $X = [x_1, \ldots, x_n]$ , that might be a hamiltonian cycle for a given graph G = (V, E) (where n = |V|).

```
Algorithm: Hamiltonian Cycle Certificate Verification(G,X) flag \leftarrow true Used \leftarrow \{x_1\} j \leftarrow 2 while (j \leq n) and flag \begin{cases} flag \leftarrow (x_j \not\in Used) \text{ and } (\{x_{j-1}, x_j\} \in E) \\ \text{if } (j=n) \text{ then } flag \leftarrow flag \text{ and } (\{x_n, x_1\} \in E) \\ Used \leftarrow Used \cup \{x_j\} \end{cases} return (flag)
```

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## **Polynomial-time Reductions**

For a decision problem  $\Pi$ , let  $\mathcal{I}(\Pi)$  denote the set of all instances of  $\Pi$ . Let  $\mathcal{I}_{ves}(\Pi)$  and  $\mathcal{I}_{no}(\Pi)$  denote the set of all yes-instances and no-instances (respectively) of  $\Pi$ .

Suppose that  $\Pi_1$  and  $\Pi_2$  are decision problems. We say that there is a polynomial-time reduction (AKA polynomial transformation) from  $\Pi_1$ to  $\Pi_2$  (denoted  $\Pi_1 \leq_P \Pi_2$ ) if there exists a function  $f: \mathcal{I}(\Pi_1) \to \mathcal{I}(\Pi_2)$ such that the following properties are satisfied:

- f(I) is computable in polynomial time (as a function of size (I), where  $I \in \mathcal{I}(\Pi_1)$ )
- if  $I \in \mathcal{I}_{\mathbf{ves}}(\Pi_1)$ , then  $f(I) \in \mathcal{I}_{\mathbf{ves}}(\Pi_2)$
- if  $I \in \mathcal{I}_{\mathbf{no}}(\Pi_1)$ , then  $f(I) \in \mathcal{I}_{\mathbf{no}}(\Pi_2)$

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# Two Graph Theory Decision Problems

## **Problem**

## Clique

**Instance:** An undirected graph G = (V, E) and an integer k, where  $1 \le k \le |V|$ .

**Question:** Does G contain a clique of size  $\geq k$ ? (A clique is a subset of vertices  $W \subseteq V$  such that  $uv \in E$  for all  $u, v \in W$ ,  $u \neq v$ .)

#### **Problem**

#### Vertex Cover

**Instance:** An undirected graph G=(V,E) and an integer k, where  $1 \le k \le |V|$ .

Question: Does G contain a vertex cover of size  $\leq k$ ? (A vertex cover is a subset of vertices  $W \subseteq V$  such that  $\{u,v\} \cap W \neq \emptyset$  for all edges  $uv \in E$ .)

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## Clique $\leq_P$ Vertex-Cover

Suppose that I=(G,k) is an instance of **Clique**, where G=(V,E),  $V=\{v_1,\ldots,v_n\}$  and  $1\leq k\leq n$ .

Construct an instance  $f(I)=(H,\ell)$  of Vertex Cover, where H=(V,F),  $\ell=n-k$  and

$$v_i v_j \in F \Leftrightarrow v_i v_j \notin E$$
.

H is called the **complement** of G, because every edge of G is a non-edge of H and every non-edge of G is an edge of H.

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# **Properties of Polynomial-time Reductions**

Suppose that  $\Pi_1,\Pi_2,\ldots$  are decision problems.

#### **Theorem**

If  $\Pi_1 \leq_P \Pi_2$  and  $\Pi_2 \in P$ , then  $\Pi_1 \in P$ .

#### **Theorem**

 $\Pi_1 <_P \Pi_2$  and  $\Pi_2 <_P \Pi_3$ , then  $\Pi_1 <_P \Pi_3$ .

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## The Complexity Class NPC

The complexity class NPC denotes the set of all decision problems  $\Pi$  that satisfy the following two properties:

- $\Pi \in NP$
- For all  $\Pi' \in NP$ ,  $\Pi' \leq_P \Pi$ .

NPC is an abbreviation for NP-complete.

#### **Theorem**

If  $P \cap NPC \neq \emptyset$ , then P = NP.

## Satisfiability and the Cook-Levin Theorem

#### **Problem**

#### **CNF-Satisfiability**

**Instance:** A boolean formula F in n boolean variables  $x_1, \ldots, x_n$ , such that F is the **conjunction** (logical "and") of m clauses, where each clause is the **disjunction** (logical "or") of literals. (A literal is a boolean variable or its negation.)

**Question:** Is there a truth assignment such that F evaluates to true?

#### **Theorem**

**CNF-Satisfiability**  $\in$  *NPC*.

## **Proving Problems NP-complete**

Now, given any NP-complete problem, say  $\Pi_1$ , other problems in NP can be proven to be NP-complete via polynomial reductions from  $\Pi_1$ , as stated in the following theorem:

#### **Theorem**

Suppose that the following conditions are satisfied:

- $\Pi_1 \in NPC$ ,
- $\Pi_1 \leq_P \Pi_2$ , and
- $\Pi_2 \in NP$ .

Then  $\Pi_2 \in NPC$ .

# **More Satisfiability Problems**

#### **Problem**

## 3-CNF-Satisfiability

**Instance:** A boolean formula F in n boolean variables, such that F is the conjunction of m clauses, where each clause is the disjunction of exactly three literals.

**Question:** Is there a truth assignment such that F evaluates to true?

#### **Problem**

## 2-CNF-Satisfiability

**Instance:** A boolean formula F in n boolean variables, such that F is the conjunction of m clauses, where each clause is the disjunction of exactly two literals.

**Question:** Is there a truth assignment such that F evaluates to true?

**3-CNF-Satisfiability**  $\in$  *NPC*, while **2-CNF-Satisfiability**  $\in$  *P*.

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# **CNF-Satisfiability** $\leq_P$ **3-CNF-Satisfiability**

Suppose that  $(X, \mathcal{C})$  is an instance of **CNF-SAT**, where  $X = \{x_1, \dots, x_n\}$  and  $\mathcal{C} = \{C_1, \dots, C_m\}$ . For each  $C_j$ , do the following:

case 1 If  $|C_j| = 1$ , say  $C_j = \{z\}$ , construct four clauses

$$\{z,a,b\},\{z,a,\overline{b}\},\{z,\overline{a},b\},\{z,\overline{a},\overline{b}\}.$$

case 2 If  $|C_j| = 2$ , say  $C_j = \{z_1, z_2\}$ , construct two clauses

$$\{z_1, z_2, c\}, \{z_1, z_2, \overline{c}\}.$$

case 3 If  $|C_j| = 3$ , then leave  $C_j$  unchanged.

case 4 If  $|C_j| \ge 4$ , say  $C_j = \{z_1, z_2, \dots, z_k\}$ , then construct k-2 new clauses

$$\{z_1, z_2, d_1\}, \{\overline{d_1}, z_3, d_2\}, \{\overline{d_2}, z_4, d_3\}, \dots, \{\overline{d_{k-4}}, z_{k-2}, d_{k-3}\}, \{\overline{d_{k-3}}, z_{k-1}, z_k\}.$$

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# **3-CNF-Satisfiability** $\leq_P$ Clique

Let I be the instance of **3-CNF-SAT** consisting of n variables,  $x_1, \ldots, x_n$ , and m clauses,  $C_1, \ldots, C_m$ . Let  $C_i = \{z_1^i, z_2^i, z_3^i\}$ ,  $1 \le i \le m$ .

Define f(I) = (G, k), where G = (V, E) according to the following rules:

- $V = \{v_j^i : 1 \le i \le m, 1 \le j \le n\}$ ,
- $v^i_j v^{i'}_{j'} \in E$  if and only if  $i \neq i'$  and  $z^i_j \neq \overline{z^{i'}_{j'}}$ , and
- $\bullet$  k=m.

## Subset Sum

#### **Problem**

#### Subset Sum

**Instance:** A list of sizes  $S = [s_1, ..., s_n]$ ; and a target sum, T. These are all positive integers.

**Question:** Does there exist a subset  $J \subseteq \{1, ..., n\}$  such that

$$\sum_{i \in J} s_i = T?$$

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# **Vertex Cover** <<sub>P</sub> **Subset Sum**

Suppose I = (G, k), where G = (V, E), |V| = n, |E| = m and  $1 \le k \le n$ . Suppose  $V = \{v_1, \dots, v_n\}$  and  $E = \{e_0, \dots, e_{m-1}\}$ . For  $1 \le i \le n$ , 0 < i < m - 1. let

$$c_{ij} = \begin{cases} 1 & \text{if } e_j \text{ is incident with } v_i \\ 0 & \text{otherwise.} \end{cases}$$

Define n+m sizes and a target sum W as follows:

$$a_{i} = 10^{m} + \sum_{j=0}^{m-1} c_{ij} 10^{j} \quad (1 \le i \le n)$$

$$b_{j} = 10^{j} \quad (0 \le j \le m - 1)$$

$$W = k \cdot 10^{m} + \sum_{j=0}^{m-1} 2 \cdot 10^{j}$$

Then define  $f(I) = (a_1, \dots, a_n, b_0, \dots, b_{m-1}, W)$ .

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# Reductions among NP-complete Problems (summary)

In the above diagram, arrows denote polynomial reductions.

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## **Undecidable Problems**

A decision problem  $\Pi$  is **undecidable** if there does not exist an algorithm that solves  $\Pi$ .

If  $\Pi$  is undecidable, then for every algorithm A, there exists at least one instance  $I \in \mathcal{I}(\Pi)$  such that A(I) does not find the correct answer ("yes" or "no") in finite time.

#### **Problem**

Halting

**Instance:** A computer program A and input x for the program A.

Question: When program A is executed with input x, will it halt in

finite time?

# **Undecidability of the Halting Problem**

Suppose that *Halt* is a program that solves the **Halting Problem**. Consider the following algorithm *Strange*.

What happens when we run Strange(Strange)?

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