# EECS 336: Introduction to Algorithms P vs. NP Lecture 14 intractability, NP, decision problems

**Reading:** 8.0-8.3

#### Last time:

 $\bullet$  max flow alg / ford-fulkerson

• duality: max flow = min cut

### Today:

- tractibility and intractibility
- P and NP
- decision problems
- INDEP-SET, 3-SAT, TSP, NP, CIRCUIT-SAT

# Intractibility and NP- output: completeness

"when is a problem intractable?"

**Def:**  $\mathcal{P}$  is the class of problems that can be solved in polynomial time.

$$X \in \mathcal{P}$$
 iff

 $\exists$  polynomial  $p(\cdot)$ ,

 $\exists \text{ alg } \mathcal{A},$ 

 $\forall$  instances x of X,

 $\Rightarrow \mathcal{A}$  solves x and in time O(p(|x|))

**Note:** easy to show  $X \in \mathcal{P}$ , just give  $\mathcal{A}$  and prove poly runtime.

**Examples:** network-flow, matching, interval scheduling, etc.

#### Three Infamous Problems

# Problem 1: Independent Set (INDEP-SET)

input: G = (V, E)

output:  $S \subset V$ 

- satisfying  $\forall v \in S, (u, v) \notin E$
- maximizing |S|

#### Problem 2: Satisfiability (SAT)

input: boolean formula  $f(\mathbf{z})$ 

e.g., 
$$f(\mathbf{z}) = (z_1 \vee \bar{z}_2 \vee x_3) \wedge (z_2 \vee \bar{z}_5 \vee z_6) \wedge \cdots$$

• "Yes" if assignment  $\mathbf{z}$  with  $f(\mathbf{z}) = T$  exists

e.g., 
$$\mathbf{z} = (T, T, F, T, F, \ldots)$$

• "No" otherwise.

#### Problem 3: Traveling Salesman (TSP)

input:

- G = (V, E), complete graph.
- $c(\cdot) = \text{costs on edges}$ .

output: cycle C that

- passes through all vertices exactly once.
- minimizes total cost  $\sum_{e \in C} c(e)$ .

No polynomial time algorithm is known for any of these problems!

#### Theory of Intractability

Goal: formal way to argue that no polynomial time algorithm exists (or "unlikely to exist"), i.e.,  $X \notin \mathcal{P}$ .

**Challenge:** must show that all algorithms fail!

**Idea:** to show X is difficult, reduce notoriously hard problem Y to X, i.e., reduce from Y.

**Example:** to show new problem X is hard, e.g., reduce TSP to X, i.e., reduce from TSP.

**Def:**  $\underline{Y}$  reduces to  $\underline{X}$  in polynomial time (notation:  $\underline{Y} \leq_{\mathcal{P}} X$  if any instance of Y can be solved in a polynomial number of computational steps and a polynomial number of calls to black-box that solves instances of X.

### Consequences of $Y \leq_{\mathcal{P}} X$ :

1. if X can be solved in polynomial time then so can Y.

Example: X = network-flow; Y = bipartite matching.

2. if Y cannot be solved in polynomial time then neither can X.

#### **Decision Problems**

Goal: show SAT, INDEP-SET, TSP equivalently hard.

**Challenge:** SAT, INDEP-SET, TSP problem solutions are very different.

Idea: focus on decision version of problem.

**Def:** A <u>decision problem</u> asks "does a feasible solution exist?"

Example: satisfiability.

**Def:** an <u>optimization problem</u> asks "what is the min (or max) value of a feasible solution?"

**Def:** the decision problem  $X_d$  for optimization problem X is has input (x, D) = "does instance x of X have a feasible solution with value at most (or at least) D?"

#### **Examples:**

INDEP-SET<sub>d</sub>: set S with  $|S| \ge D$ 

 $SAT_d$ : **z** such that  $f(\mathbf{z}) = T$ .

TSP<sub>d</sub>: tour C with  $\sum_{c \in C} c(e) \leq D$ 

## Deciding is as hard as optimizing

Theorem:  $X \leq_{\mathcal{P}} X_d$ 

**Proof:** (reduction via binary search)

- given
  - $\bullet$  instance x of X
  - black-box  $\mathcal{A}$  to solve  $X_d$
- search(A, B) = find optimal value in [A, B].
  - D = (A + B)/2
  - run  $\mathcal{A}(x,D)$
  - if "yes", search(A, D)
  - if "no",  $\operatorname{search}(D, B)$

# Finding solution is as hard as deciding

Example: satisfiability

- 1. if f is satisfiable  $\exists \mathbf{z}$  s.t.  $f(\mathbf{z}) = T$
- 2. guess  $z_n = T$
- 3. let  $f'(z_1, ..., z_{n-1}) = f(z_1, ..., z_{n-1}, T)$
- 4. if f' is satisfiable, repeat (2) on f'
- 5. if f' is unsatisfiable, repeat (2) on  $f''(z_1, \ldots, z_{n-1}) = f(z_1, \ldots, z_{n-1}, F)$ .

**Note:** since  $X_d =_{\mathcal{P}} X$ , we write "X" but we mean " $X_d$ "

## A notoriously hard problem

**Note:** all example problem have short certificates that could easily verify "yes" instance.

**Def:**  $\mathcal{NP}$  is the class of problems that have short (polynomial sized) certificates that can easily (in polynomial time) verify "yes" instances.

# Historical Note: $\mathcal{NP} = \underline{\text{non-deterministic}}$ polynomial time

"a nondeterministic algorithm could guess the certificate and then verify it in polynomial time"

**Note:** Not all problems are in  $\mathcal{NP}$ .

E.g., unsatisfiability.

#### Def:

- Problem  $\underline{X}$  is in  $\mathcal{NP}$  if exists short easily-verifiable certificate.
- Problem X is  $\mathcal{NP}$ -hard if  $\forall Y \in \mathcal{NP}$ ,  $Y \leq_{\mathcal{P}} X$ .
- Problem X is  $\mathcal{NP}$ -complete if  $X \in \mathcal{NP}$  and X is  $\mathcal{NP}$ -hard.

Lemma: INDEP-SET  $\in \mathcal{NP}$ .

Lemma: SAT  $\in \mathcal{NP}$ .

Lemma:  $TSP \in \mathcal{NP}$ .

Goal: show INDEP-SET, SAT, TSP are

 $\mathcal{NP}$ -complete.

Notorious Problem: NP

input:

- decision problem verifier program VP.
- polynomial  $p(\cdot)$ .
- $\bullet$  decision problem instance: x

output:

- "Yes" if exists certificate c such that VP(x,c) has "verified = true" at computational step p(|x|).
- "No" otherwise.

Fact: NP is  $\mathcal{NP}$ -complete.

**Note:** Unknown whether  $\mathcal{P} = \mathcal{NP}$ .

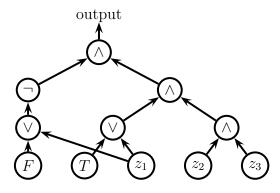
**Note:**  $\leq_{\mathcal{P}}$  is transitive: if  $Y \leq_{\mathcal{P}} X$  and

 $X \leq_{\mathcal{P}} Z$  then  $Y \leq_{\mathcal{P}} Z$ .

**Plan:** NP  $\leq_{\mathcal{P}}$  CIRCUIT-SAT  $\leq_{\mathcal{P}}$  SAT.

### Circuit Satisfiability

#### Example:



#### Problem 4: CIRCUIT-SAT

input: boolean circuit  $Q(\mathbf{z})$ 

- directed acyclic graph G = (V, E)
- internal nodes labeled by logical gates:

• leaves labeled by variables or constants

$$T, F, z_1, \ldots, z_n$$
.

 $\bullet$  root r is output of circuit

output:

- "Yes" if exists **z** with  $Q(\mathbf{z}) = T$
- "No" otherwise.

**Lemma:** CIRCUIT-SAT is  $\mathcal{NP}$ -hard.

**Proof:** (reduce from NP)

- goal: convert NP instance (VP, p, x) to CIRCUIT-SAT instance Q
- $VP(\cdot, \cdot)$  polynomial time

- $\Rightarrow$  computer can run it in poly steps.
- each step of computer is circuit.
- output of one step is input to next step
- unroll p(|x|) steps of computation
  - $\Rightarrow \exists \text{ poly-size circuit } Q'(\mathbf{x}, \mathbf{c}) = VP(x, c)$
- hardcode **x**:  $Q(\mathbf{c}) = Q'(\mathbf{x}, \mathbf{c})$
- Conclusion: Q is sat iff exists c with VP(x,c) = "verified".

QED