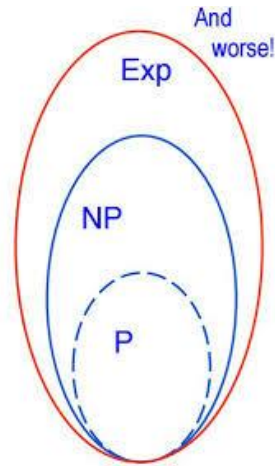


# Lecture 22

## Computational Complexity II



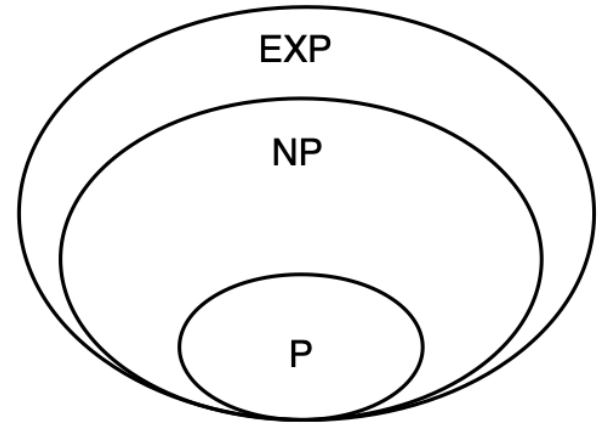
# Announcements

- ❖ Quiz 2 releases Friday; due by end of semester (5/8)
- ❖ Group Project problems now up; choose one soon!

# Complexity Classes

Nested classification of problems

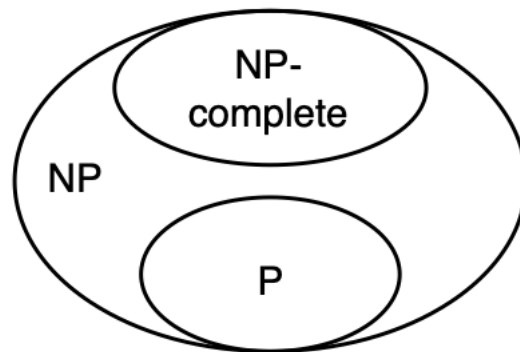
- ❖ **P**: solvable in polynomial time
- ❖ **NP**: includes most natural and interesting problems
- ❖ **EXP**: solvable in exponential time



# NP-Completeness

## NP-Complete

- ❖ Problems that are "as difficult" as every other problem in **NP**
- ❖ A polynomial time algorithm for any NP-Complete problem implies one for every problem in **NP**



# Outline

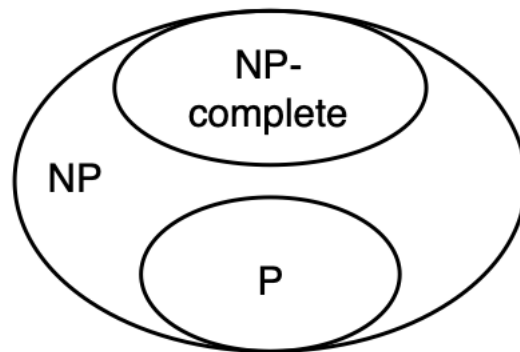
Goal: make sense of this picture

Polynomial-time reductions

❖ What it means for one problem to be "as difficult as" another problem

Define NP and NP-Complete

Work through examples



# Polynomial-Time Reduction

Problem Y is **polynomial-time reducible** to Problem X if any instance of Problem Y can be solved with

- ❖ A polynomial number of standard computational steps
- ❖ A polynomial number of calls to a black box that solves problem X

```
solveY(yInput)
```

```
    Construct xInput          // poly-time  
    foo = solveX(xInput)     // poly # of calls to solveX  
    return yes/no based on foo // poly-time
```

Notation:  $Y \leq_p X$

Useful for discussing problems in **P** or complexity higher classes but erases distinctions between subclasses of **P**

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When thinking about the complexity of problems, we usually frame the problems in terms of answering a YES/NO question

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All the problems we've looked at this semester  
can be framed as decision problems

# P and NP

## Complexity Class **P**

- ❖ Decision problems for which there is a polynomial time algorithm

## Complexity Class **NP**

- ❖ Decision problems for which there is a polynomial time certifier
- ❖ A solution can be "verified" in polynomial time
- ❖ NP means non-deterministic polynomial time

# Solver vs Certifier

Let  $X$  be a decision problem and  $s$  be an input to the problem

- ❖ Example: Subset sum problem with input set  $S$

Solver

- ❖ Takes as input a problem input and correctly returns YES or NO
- ❖ Types of algorithms we've been developing all semester

Certifier

- ❖ Takes as input a problem input and a "certificate" or "witness" and correctly returns YES

# Certifier Example: Subset Sum

**Input:** set of integers  $S = \{7, 3, -1, -24, 6, -2\}$

**Problem:** does  $S$  have a subset that sums to zero? Type equation here.

**Witness:** a witness  $w$  is a specific subset of  $S$

**Idea:** we can check if  $w$  verifies that the answer is YES

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- ❖ Likely, not much

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There are problems with polynomial-time certifiers but may not have polynomial-time solvers

Subset Sum problem

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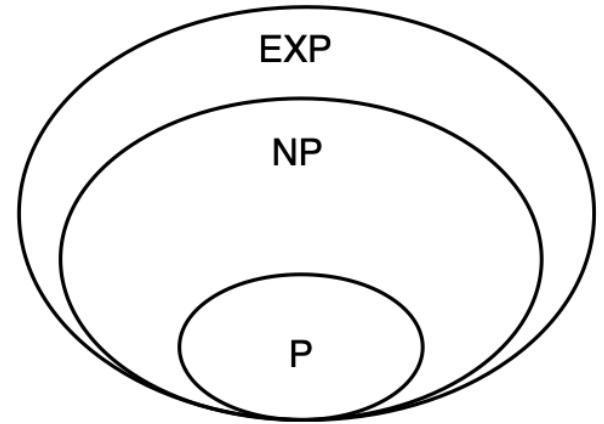
Subset Sum problem

- ❖ Baseline algorithm has running time:  $O(2^n)$
- ❖ Best known dynamic programming algorithm:  $O(nT)$  meaning pseudo-polynomial
- ❖ If a given subset sums to zero, we can verify it in running time  $O(n)$

Therefore, the Subset Sum problem is in **NP**

# P, NP, EXP

- ❖ **P**: easy to solve
- ❖ **NP**: easy to verify (may be hard to solve)
- ❖ **EXP**: may be hard to solve or verify



# NP-Completeness

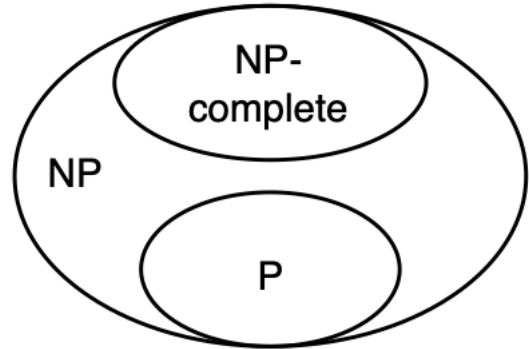
Problems that are "as difficult" as every other problem in **NP**

A problem  $X$  is NP-Complete if

- ❖  $X$  is in **NP**
- ❖ If  $Y$  is in **NP**, then  $Y \leq_p X$  (known as NP-hard)

The Subset Sum Problem is NP-Complete

[More examples](#)



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