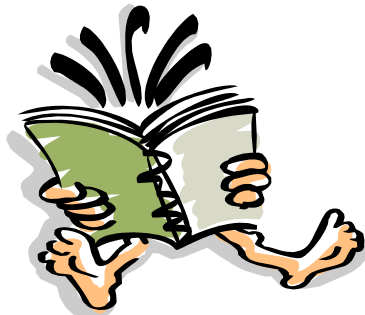


# NP Completeness

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# NP-Completeness

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- So far we've seen a lot of good news!
  - Such-and-such a problem can be solved quickly (i.e., in close to linear time, or at least a time that is some small polynomial function of the input size)
- NP-completeness is a form of bad news!
  - Evidence that many important problems can not be solved quickly.
- NP-complete problems really come up all the time!

# Why should we care?

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- Knowing that they are hard lets you stop beating your head against a wall trying to solve them...
  - **Use a heuristic:** come up with a method for solving a reasonable fraction of the common cases.
  - **Solve approximately:** come up with a solution that you can prove that is close to right.
  - **Use an exponential time solution:** if you really have to solve the problem exactly and stop worrying about finding a better solution.

# Optimization & Decision Problems

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- **Decision problems**

- Given an input and a question regarding a problem, determine if the answer is yes or no

- **Optimization problems**

- Find a solution with the “best” value

- Optimization problems can be cast as decision problems that are easier to study

- *E.g.:* Shortest path:  $G$  = unweighted directed graph
  - Find a path between  $u$  and  $v$  that uses the fewest edges
  - ♦ *Does a path exist from  $u$  to  $v$  consisting of at most  $k$  edges?*

# Algorithmic vs Problem Complexity

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- The **algorithmic complexity** of a computation is some measure of how *difficult* is to perform the computation (i.e., specific to an algorithm)
- The **complexity of a computational *problem* or *task*** is the complexity of the algorithm with the **lowest** order of growth of complexity for solving that problem or performing that task.
  - e.g. the problem of searching an ordered list has *at most  $\lg n$*  time complexity.
- **Computational Complexity**: deals with classifying problems by how hard they are.

# Class of “P” Problems

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- **Class P** consists of (decision) problems that are solvable in polynomial time
- Polynomial-time algorithms
  - Worst-case running time is  $O(n^k)$ , for some constant  $k$
- Examples of polynomial time:
  - $O(n^2)$ ,  $O(n^3)$ ,  $O(1)$ ,  $O(n \lg n)$
- Examples of non-polynomial time:
  - $O(2^n)$ ,  $O(n^n)$ ,  $O(n!)$

# Tractable/Intractable Problems

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- Problems in P are also called **tractable**
- Problems **not** in P are **intractable or unsolvable**
  - Can be solved in reasonable time only for small inputs
  - Or, can not be solved at all
- Are non-polynomial algorithms always worse than polynomial algorithms?
  - $n^{1,000,000}$  is *technically* tractable, but really impossible
  - $n^{\log \log \log n}$  is *technically* intractable, but easy

# Example of Unsolvable Problem

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- Turing discovered in the 1930's that there are problems **unsolvable** by *any* algorithm.
- The most famous of them is the ***halting problem***
  - Given an arbitrary algorithm and its input, will that algorithm eventually halt, or will it continue forever in an “*infinite loop?*”



# Examples of Intractable Problems

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## Hamiltonian Paths

*Optimization Problem:* Given a graph, find a path that passes through every vertex exactly once

*Decision Problem:* Does a given graph have a Hamiltonian Path ?

## Traveling Salesman

*Optimization Problem:* Find a minimum weight Hamiltonian Path

*Decision Problem:* Given a graph and an integer  $k$ , is there a Hamiltonian Path with a total weight at most  $k$  ?

# Intractable Problems

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- Can be classified in various categories based on their degree of difficulty, e.g.,
  - NP
  - NP-complete
  - NP-hard
- Let's define NP algorithms and NP problems ...

# Nondeterministic and NP Algorithms

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**Nondeterministic algorithm** = two stage procedure:

1) Nondeterministic (“guessing”) stage:

generate randomly an arbitrary string that can be thought of as a candidate solution (“certificate”)

2) Deterministic (“verification”) stage:

take the certificate and the instance to the problem and returns YES if the certificate represents a solution

**NP algorithms (Nondeterministic polynomial)**

verification stage is polynomial

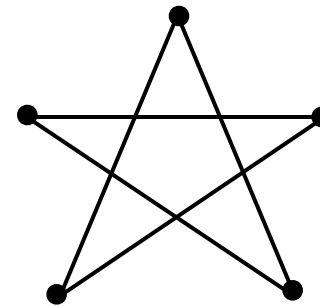
# Class of “NP” Problems

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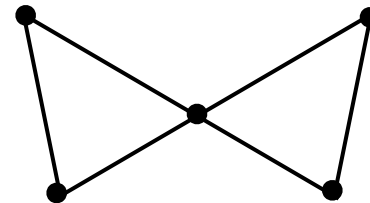
- **Class NP** consists of problems that could be solved by NP algorithms
  - i.e., verifiable in polynomial time
- If we were given a “certificate” of a solution, we could verify that the certificate is correct in time polynomial to the size of the input
- Warning: NP does **not** mean “non-polynomial”

# *E.g.:* Hamiltonian Cycle

- **Given:** a directed graph  $G = (V, E)$ , determine a simple cycle that contains each vertex in  $V$ 
  - Each vertex can only be visited once
- **Certificate:**
  - Sequence:  $\langle v_1, v_2, v_3, \dots, v_{|V|} \rangle$



hamiltonian



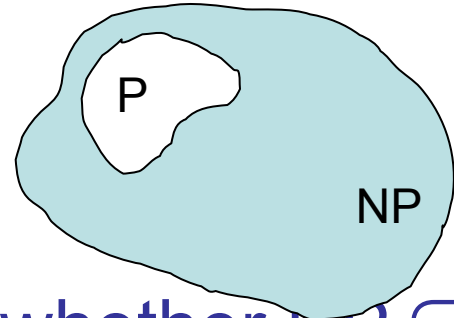
not  
hamiltonian

# Is $P = NP$ ?

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- Any problem in  $P$  is also in  $NP$ :

$$P \subseteq NP$$

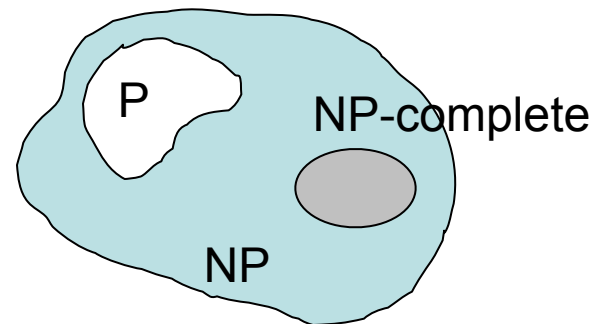


- The big (and **open question**) is whether  $NP \subseteq P$  or  $P = NP$ 
  - i.e., if it is always easy to check a solution, should it also be easy to find a solution?
- Most computer scientists believe that this is false but we do not have a proof ...

# NP-Completeness (informally)

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- **NP-complete** problems are defined as the hardest problems in NP



- Most practical problems turn out to be either P or NP-complete.

# Reductions

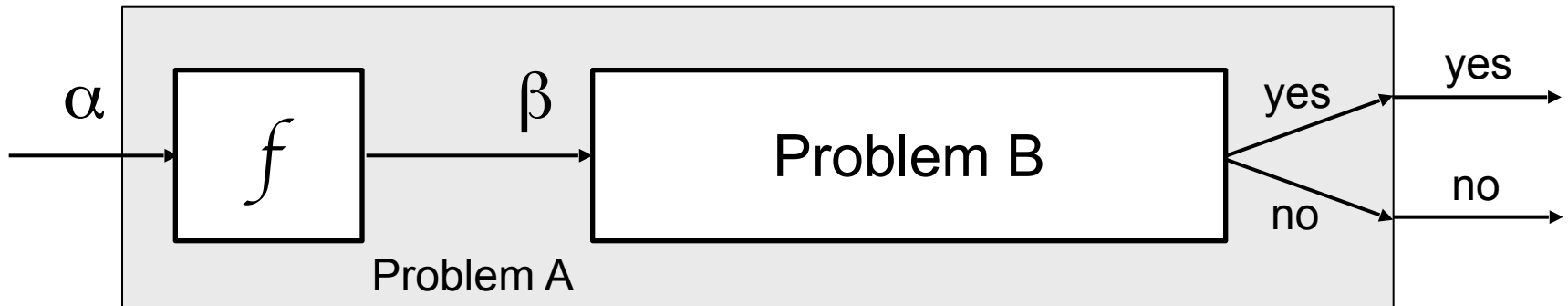
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- Reduction is a way of saying that one problem is “**easier**” than another.
- We say that problem A is easier than problem B, (i.e., we write “ **$A \leq B$** ”)  
if we can solve A using the algorithm that solves B.
- **Idea:** transform the inputs of A to inputs of B



# Polynomial Reductions

- Given two problems A, B, we say that A is polynomially **reducible** to B ( $A \leq_p B$ ) if:
  - There exists a function  $f$  that converts the input of A to inputs of B in polynomial time
  - $A(i) = \text{YES} \Leftrightarrow B(f(i)) = \text{YES}$



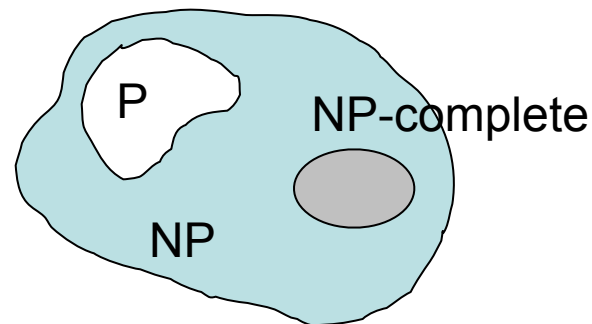
# NP-Completeness (formally)

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- A problem B is **NP-complete** if:

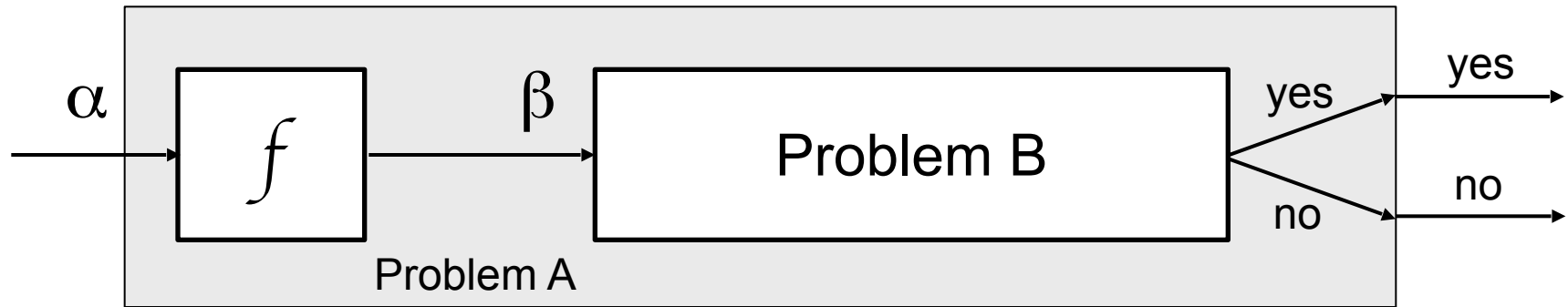
(1)  $B \in \mathbf{NP}$

(2)  $A \leq_p B$  for all  $A \in \mathbf{NP}$



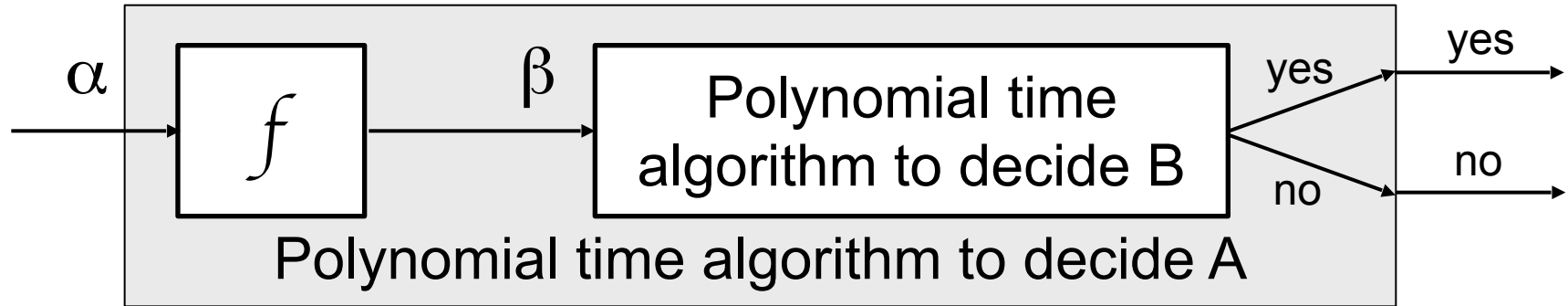
- If B satisfies only property (2) we say that B is **NP-hard**
- No polynomial time algorithm has been discovered for an **NP-Complete** problem
- No one has ever proven that no polynomial time algorithm can exist for any **NP-Complete** problem

# Implications of Reduction



- If  $A \leq_p B$  and  $B \in P$ , then  $A \in P$
- if  $A \leq_p B$  and  $A \notin P$ , then  $B \notin P$

# Proving Polynomial Time



1. Use a **polynomial time** reduction algorithm to transform A into B
2. Run a known **polynomial time** algorithm for B
3. Use the answer for B as the answer for A

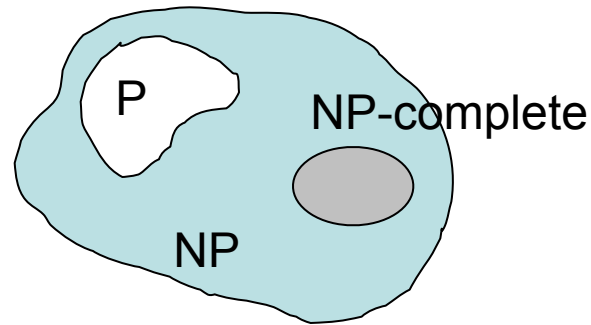
# Proving NP-Completeness In Practice

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- Prove that the problem B is in NP
  - A randomly generated string can be checked in polynomial time to determine if it represents a solution
- Show that **one known** NP-Complete problem can be transformed to B in polynomial time
  - No need to check that **all** NP-Complete problems are reducible to B

# Revisit “Is $P = NP$ ?”

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*Theorem:* If any NP-Complete problem can be solved in polynomial time  $\Rightarrow$  then  $P = NP$ .

# P & NP-Complete Problems

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- **Shortest simple path**

- Given a graph  $G = (V, E)$  find a **shortest** path from a source to all other vertices
- Polynomial solution:  $O(VE)$

- **Longest simple path**

- Given a graph  $G = (V, E)$  find a **longest** path from a source to all other vertices
- NP-complete

# P & NP-Complete Problems

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- **Euler tour**

- $G = (V, E)$  a connected, directed graph find a cycle that traverses each edge of  $G$  exactly once (may visit a vertex multiple times)
- Polynomial solution  $O(E)$

- **Hamiltonian cycle**

- $G = (V, E)$  a connected, directed graph find a cycle that visits each vertex of  $G$  exactly once
- NP-complete