



# P, NP, NP-Complete, and NP-Hard

by

Dr. Animesh Chaturvedi

Assistant Professor: LNMIIT Jaipur

Post Doctorate: King's College London & The Alan Turing Institute

PhD: IIT Indore



Indian Institute of Technology Indore  
भारतीय प्रौद्योगिकी संस्थान इंदौर

The  
Alan Turing  
Institute

# Computational complexity theory

- Fields in
  - Theoretical Computer Science
  - Analysis of Algorithms
- An algorithm solves a computation problem by mathematical steps.
- A computational problem (such as an algorithm) is a task solved by a computer.
- Focuses on classifying computational problems according to the resource usage
- Resource usage: amount of resources needed to solve computational problem,
- Resources: such as time and storage.

# Single Source Shortest-Paths Implementation

algorithm	restriction	typical case	worst case	extra space	
topological sort	no directed cycles	$E + V$	$E + V$	$V$	Easy
Dijkstra (binary heap)	no negative weights	$E \log V$	$E \log V$	$V$	
Bellman-Ford	no negative cycles	$E V$	$E V$	$V$	Medium
Bellman-Ford (queue-based)		$E + V$	$E V$	$V$	

Hard

**Remark 1.** Directed cycles make the problem harder.

**Remark 2.** Negative weights make the problem harder.

**Remark 3.** Negative cycles makes the problem intractable.

# Single Source Shortest-Paths Implementation

algorithm	restriction	typical case	worst case	extra space
<b>topological sort</b>	no directed cycles	$E + V$	$E + V$	$V$
<b>Dijkstra (binary heap)</b>	no negative weights	$E \log V$	$E \log V$	$V$
<b>Bellman-Ford</b>	no negative cycles	$E V$	$E V$	$V$
<b>Bellman-Ford (queue-based)</b>		$E + V$	$E V$	$V$

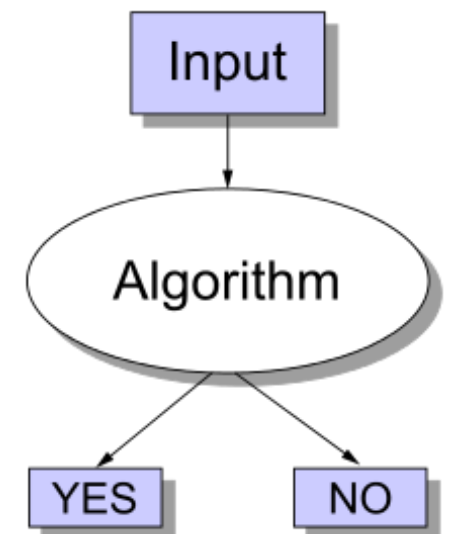
**Remark 1.** Directed cycles make the problem harder.

**Remark 2.** Negative weights make the problem harder.

**Remark 3.** Negative cycles makes the problem intractable.

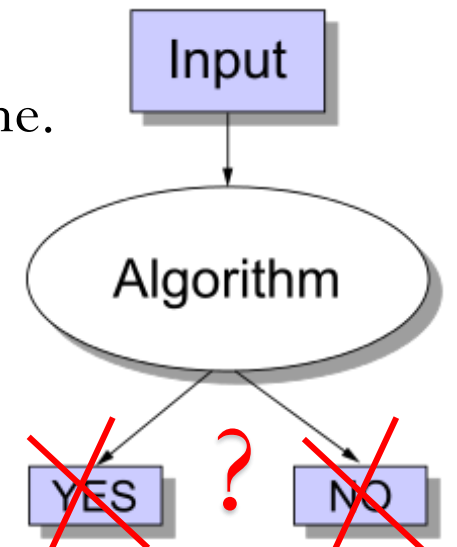
# Deterministic algorithm

- Given a particular input, will always produce the same output, with the underlying machine always passing through the same sequence of states.
- State machine: a state describes what a machine is doing at a particular instant in time.
- State machines pass in a discrete manner from one state to another.
- Enter the input, initial state or start state.
- Current state determines what will be next state, the set of states is predetermined.



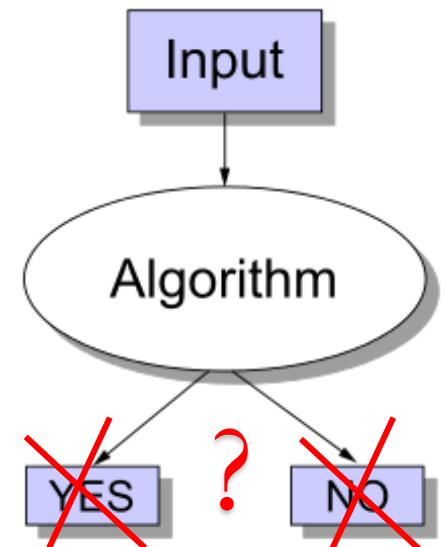
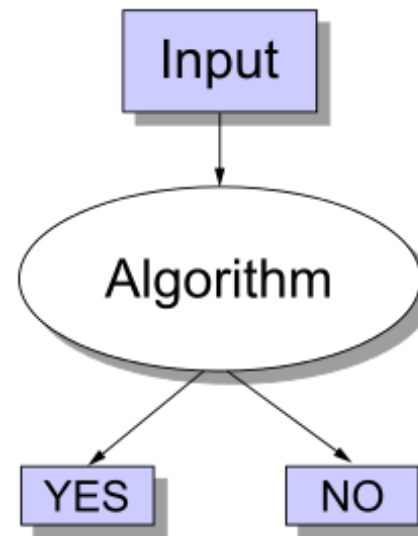
# Non-deterministic Algorithms

- If it uses external state other than the input, such as
  - user input,
  - a global variable,
  - a hardware timer value,
  - a random value, or
  - stored disk data.
- If it is timing-sensitive,
  - e.g. if it has multiple processors writing to the same data at the same time.
- If a hardware error causes its state to change in an unexpected way.
- The order each processor writes data will affect the result.



# Deterministic and Non-deterministic Algorithms

- Disadvantages of Determinism
  - predictable future by players or predictable security by hacker
  - e.g. predictable card shuffling program or security key
- Pseudorandom number generator is often not sufficient,
  - thus cryptographically secure pseudo-random number generator,
  - hardware random number generator.



# P (Polynomial) Time Problems

- Contains all decision problems that can be solved deterministically using a polynomial time (polynomial amount of computation time).
- A problem is in P-complete, if it is in P
- P is the class of computational problems that are "efficiently solvable" or "tractable".
- Class P, typically take all the "tractable" problems for a sequential algorithm,
- But, in reality, some problems not known to be solved in polynomial P time.

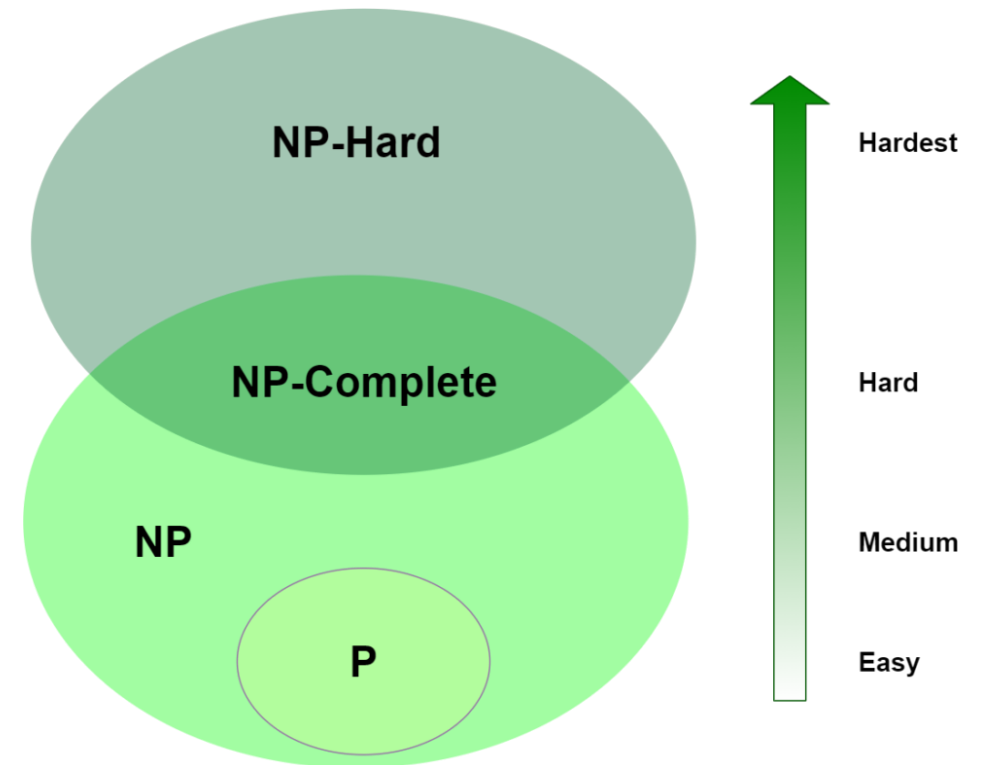


# P (Polynomial) Time Problems

- Programmable Function (or method) is polynomial-time
  - if completes in constant-time or polynomial time,
  - then the entire algorithm takes polynomial time.
- Polynomial-time algorithms:
  - Minimum Spanning Tree: Kruskal's  $O(E \lg V)$  and Prim's  $O(E + V \lg V)$  algorithm
  - Shortest Path Algorithms: Dijkstra's  $O(E \lg V)$  and Bellman-Ford's  $O(EV)$  algorithm

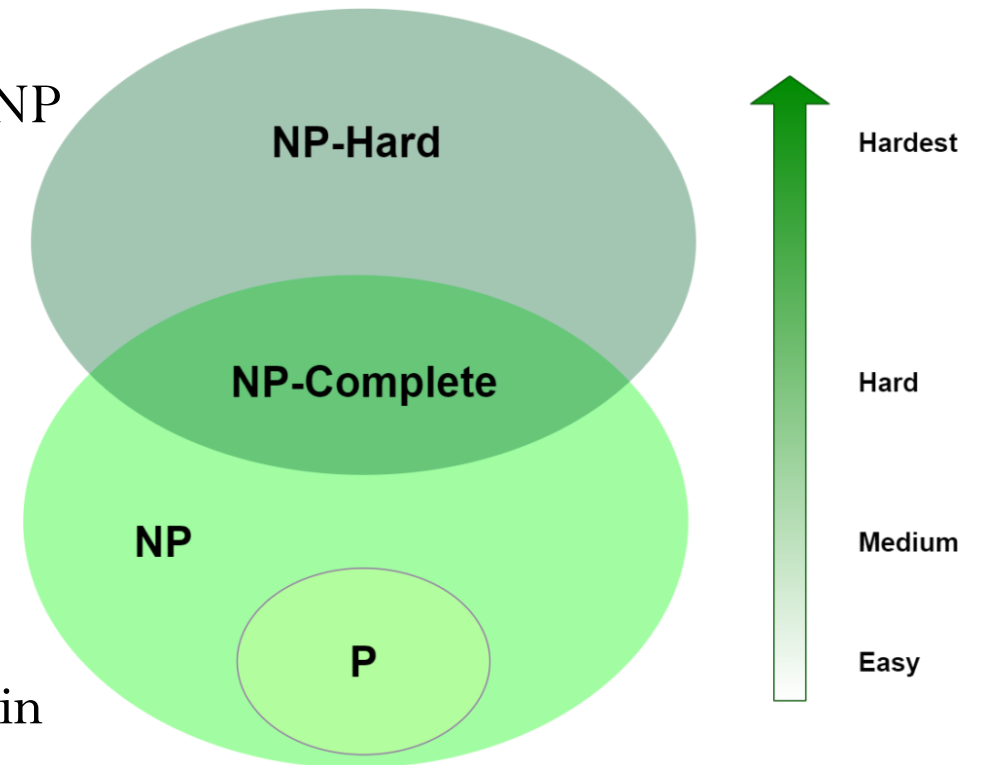
# NP - Naming convention

- Classification
  - Hardest  $\rightarrow$  NP-Hard
  - Hard  $\rightarrow$  NP-Complete
  - Medium  $\rightarrow$  NP
  - Easy  $\rightarrow$  P
- Order of N inputs
  - $O(1)$  – constant-time
  - $O(\log_2(n))$  – logarithmic-time
  - $O(n)$  – linear-time
  - $O(n^2)$  – quadratic-time
  - $O(n^k)$  – polynomial-time
  - $O(k^n)$  – exponential-time
  - $O(n!)$  – factorial-time



# NP - Naming convention

- NP-hard: Class of problems are at least as hard as the hardest problems in NP.
- NP-hard problems do not have to be in NP; means NP hard problem may not even be decidable.
- NP-complete: Class of decision problems which contains the hardest problems in NP. Each NP-complete problem has to be in NP.
- NP: Class of computational decision problems for which any given yes-solution can be verified as a solution in polynomial time
- NP-easy: At most as hard as NP, but not necessarily in NP.

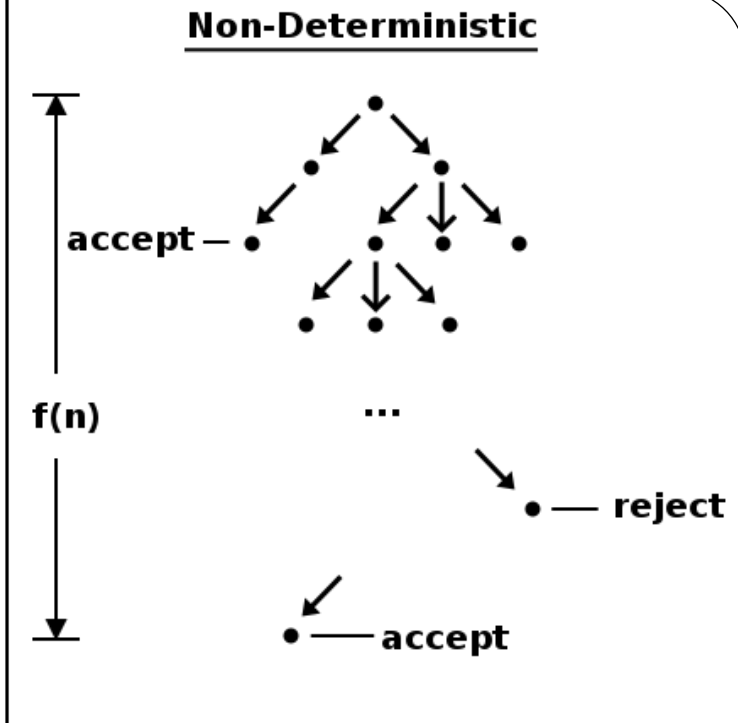
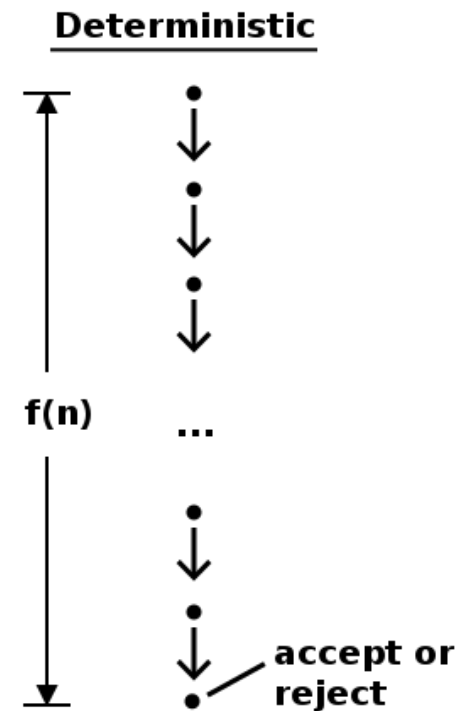


<https://en.wikipedia.org/wiki/NP-hardness>

<https://www.baeldung.com/cs/p-np-np-complete-np-hard>

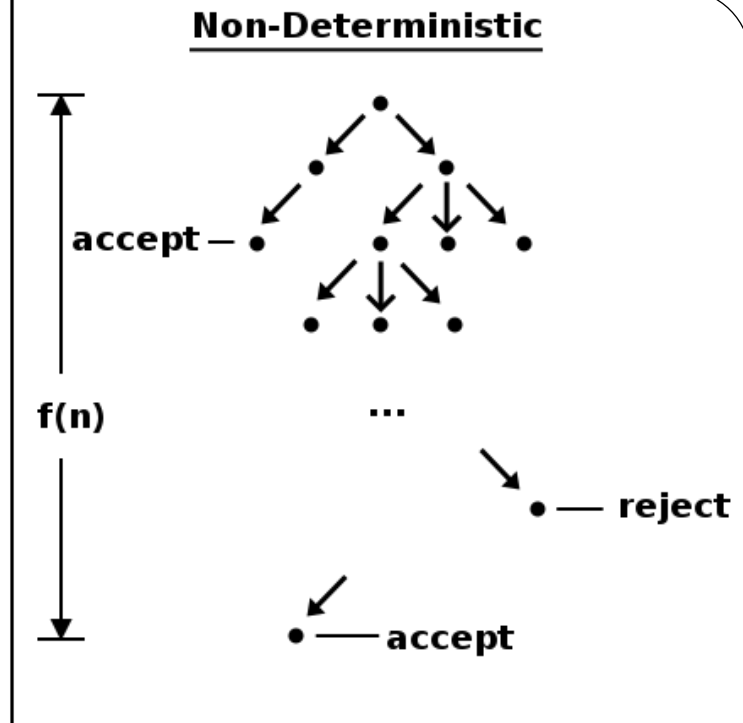
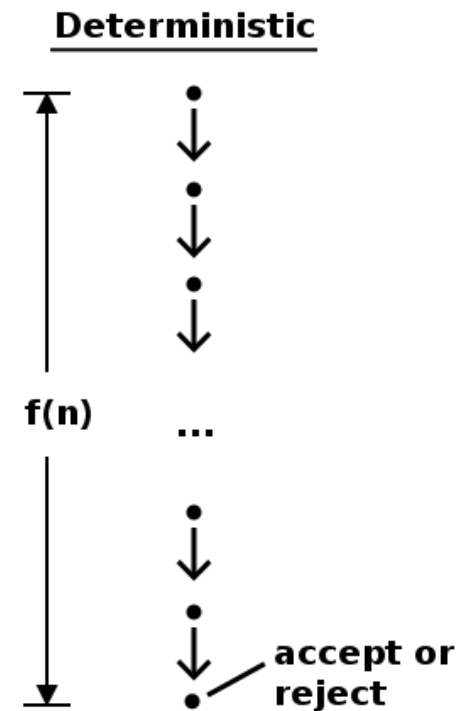
# P and NP Problems

- Nondeterministic Polynomial-time
- “Nondeterministic” refers to
  - “luckiest possible guesser”
- "Complete" refers to
  - “in the same complexity class”
- **P** versus **NP** determine
  - whether a problem can be verified in polynomial time
  - whether the problem can also be solved in polynomial time.
- If it turned out that **P**  $\neq$  **NP**, (widely accepted/believed),
- There are problems in **NP** that are harder to compute than to verify:
- NP problems could not be solved in polynomial time, but the answer could be verified in polynomial time.



# NP Complete

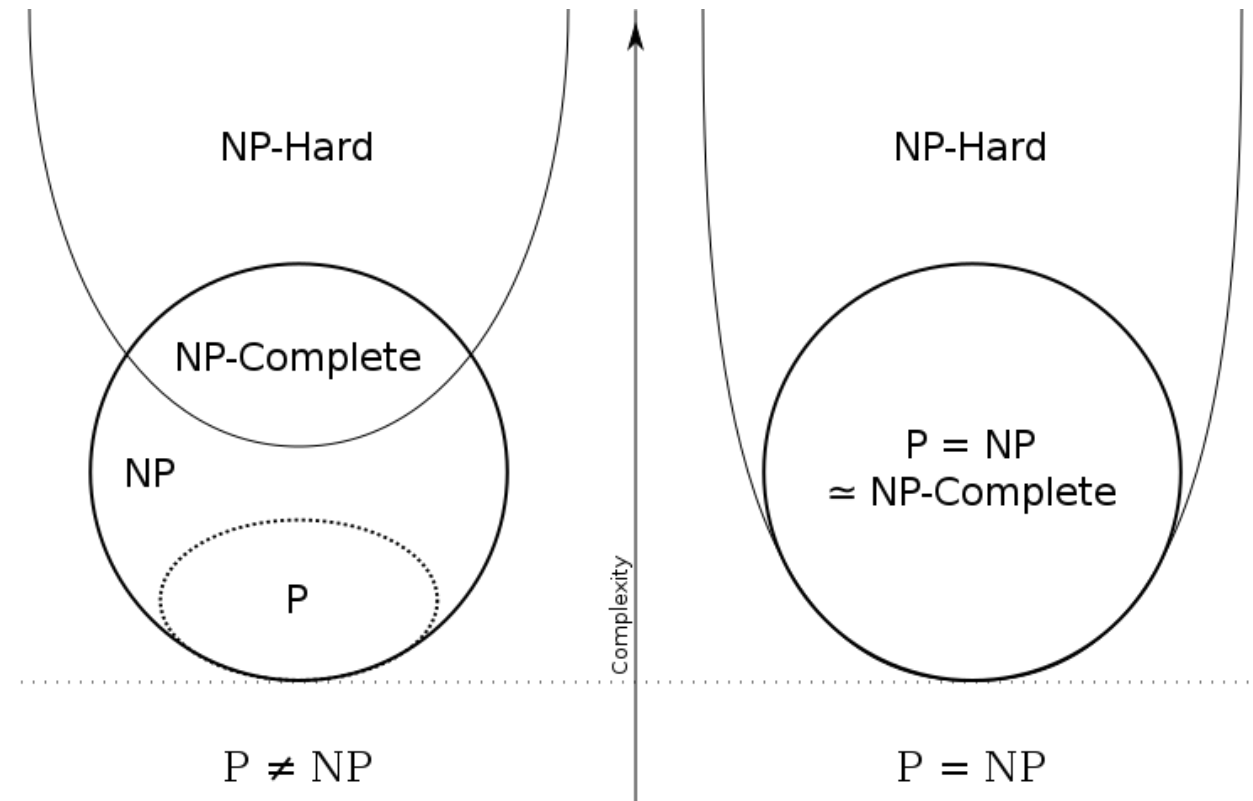
- Nondeterministic Polynomial-time Complete



- A problem is NP-complete when:
  - a brute-force search algorithm can solve it, and the correctness of each solution can be verified quickly, and
  - the problem can be used to simulate any other problem with similar solvability.
- NP-complete problem can be *verified* "quickly",
- There is no known way to *find* a solution quickly.

# NP - Hard Problems

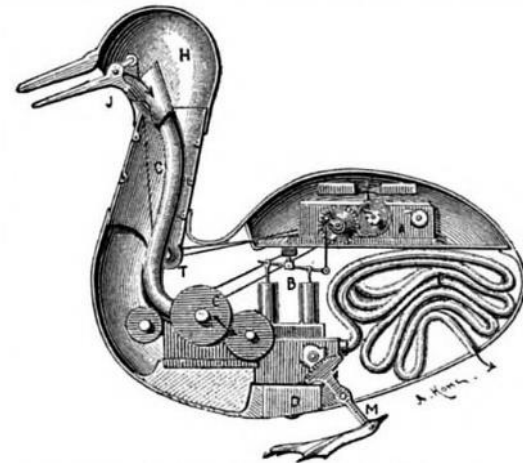
- Non-Deterministic Polynomial-time hardness
- At least as hard as the hardest problems in NP
- There might be some polynomial-time algorithms for NP-hard problems but might not have been discovered yet
- NP-hard but not NP-complete
  - halting problem: "given a program and its input, will it run forever?"
  - traveling salesman problem



# Reductionism in Algorithms

# Reductionism

- Reductionism is old domain since 16<sup>th</sup> century
- explains system in terms of parts and their interactions
  - Define a domain of possible parts
  - Generate inputs over the interaction between parts
  - Perform a deterministic computation on the input data
  - Aggregate the results
- interprets a complex system as the sum of its parts





# General Problems, Input Size and Time Complexity

- Time complexity of algorithms :
  - polynomial time algorithm ("efficient algorithm") v.s.
  - exponential time algorithm ("inefficient algorithm")
  - Find the **shortest path** in a graph from X to Y. (**easy**) polynomial time.
  - Find the **longest path** in a graph from X to Y. (with no cycles) (**hard**) exponential time
  - Is there a simple path from X to Y with weight  $\leq M$ ? (**easy**) polynomial time.
  - Is there a simple path from X to Y with weight  $\geq M$ ? (**hard**) exponential time

<b>f(n)</b>	<b>10</b>	<b>30</b>	<b>50</b>
n	0.00001 sec	0.00003 sec	0.00005 sec
$n^5$	0.1 sec	24.3 sec	5.2 mins
$2^n$	0.001 sec	17.9 mins	35.7 yrs

# Decision problems

- The solution to the problem is "yes" or "no". Most optimization problems can be phrased as decision problems (still have the same time complexity).

- Given a decision problem  $X$ .

If there is a polynomial time Non-deterministic Turing machine (or computer) program that solves  $X$ , then  $X$  belongs to NP

*Given a decision problem  $X$ .*

*For every instance  $I$  of  $X$ ,*

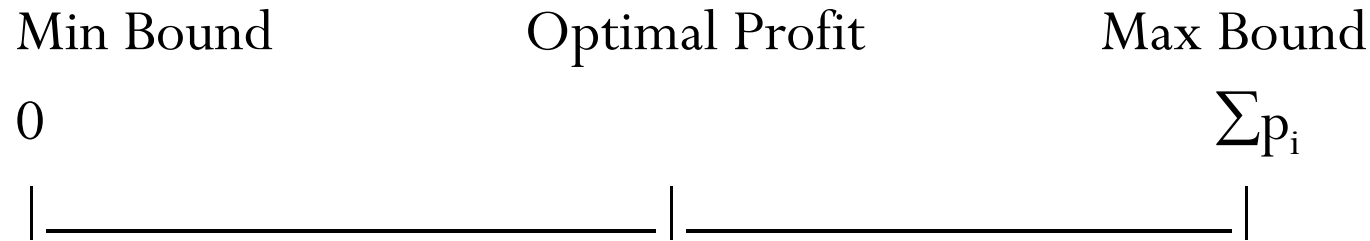
*(a) guess solution  $S$  for  $I$ , and*

*(b) check “is  $S$  a solution to  $I$ ?”*

*If (a) and (b) can be done in polynomial time, then  $X$  belongs to NP.*

# Optimization problems

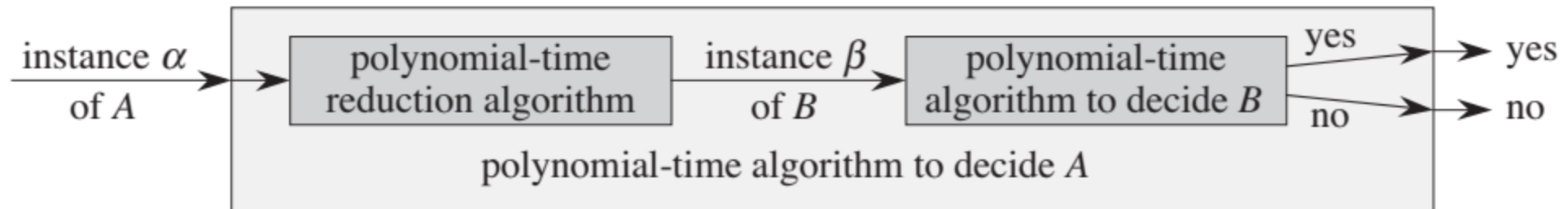
- We can repeatedly run algorithm X for various profits(P values) to find an optimal solution.
- Example : Use binary search to get the optimal profit, maximum of  $\lg \sum p_i$  runs, (where M is the capacity of the knapsack optimization problem)



Search for the optimal solution

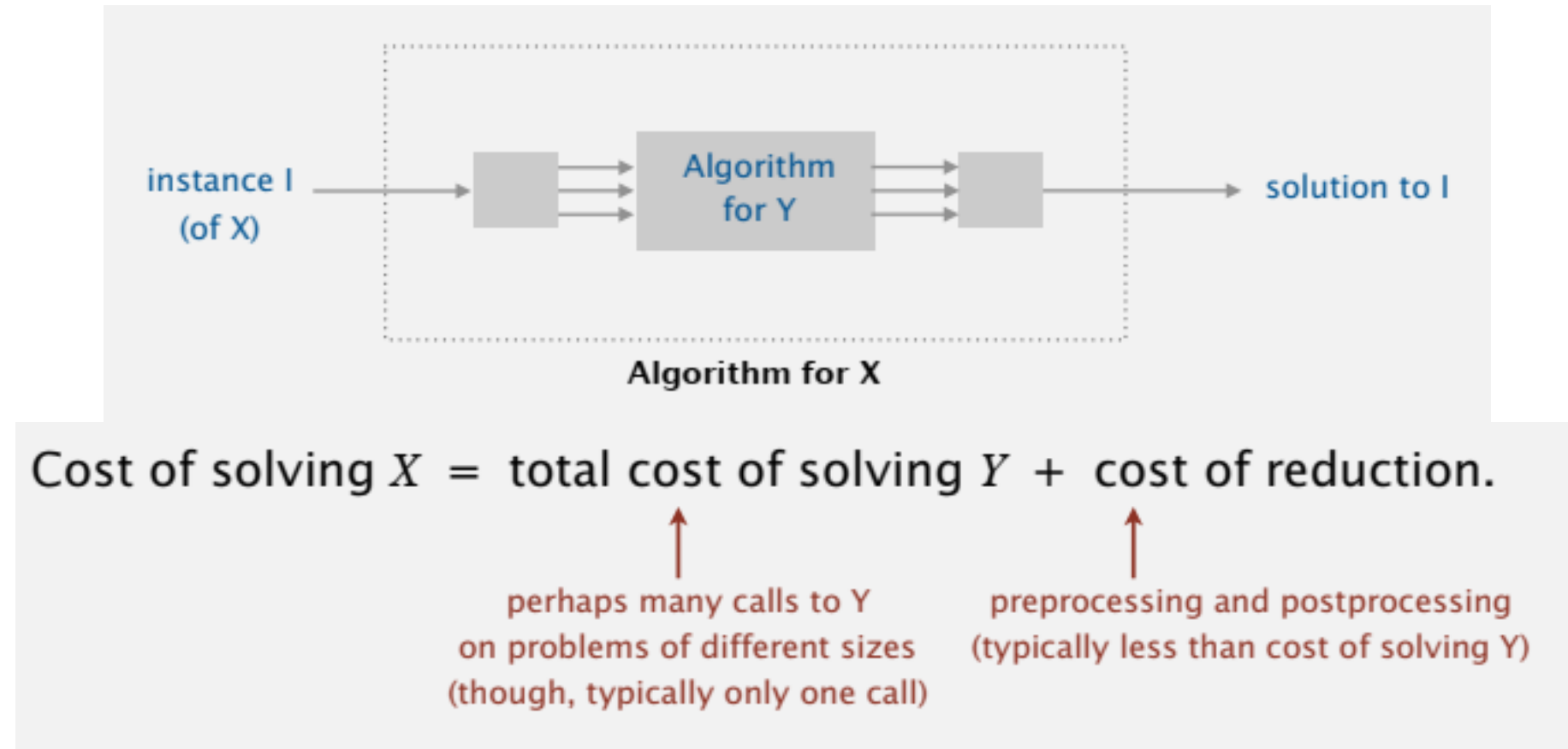
# Polynomial-time Reduction Algorithm

- Input to a particular problem an *instance* of that problem
- Suppose that we have a procedure that transforms any instance  $\alpha$  of A into some instance  $\beta$  of B
- Given an instance  $\alpha$  of problem A, use a polynomial-time reduction algorithm to transform it to an instance  $\beta$  of problem B.
- Run the polynomial-time decision algorithm for B on the instance  $\beta$
- Use the answer for  $\beta$  as the answer for  $\alpha$ .



# Algorithm Reduction

- Def. Problem  $X$  reduces to problem  $Y$  if you can use an algorithm that solves  $Y$  to help solve  $X$ .



# Algorithm Reduction

Ex 1. [finding the median reduces to sorting]

To find the median of  $N$  items:

- Sort  $N$  items.
- Return item in the middle.

Cost of solving finding the median.  $N \log N + 1$ .

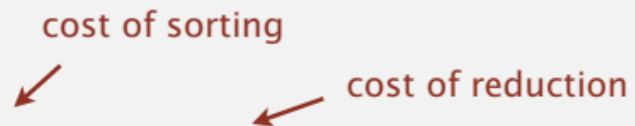


Ex 2. [element distinctness reduces to sorting]

To solve element distinctness on  $N$  items:

- Sort  $N$  items.
- Check adjacent pairs for equality.

Cost of solving element distinctness.  $N \log N + N$ .



# Algorithm Reduction

- Design algorithm. Given algorithm for  $Y$ , can also solve  $X$ .
  - CPM reduces to topological sort.
  - Arbitrage reduces to negative cycles.
  - Bipartite matching reduces to maxflow.
  - Seam carving reduces to shortest paths in a DAG.
  - Burrows-Wheeler transform reduces to suffix sort

Since I know how to solve  $Y$ , can I use that algorithm to solve  $X$ ?



programmer's version: I have code for  $Y$ . Can I use it for  $X$ ?

ON COMPUTABLE NUMBERS, WITH AN APPLICATION TO  
THE ENTSCHEIDUNGSPROBLEM

*By* A. M. TURING.

[Received 28 May, 1936.—Read 12 November, 1936.]



# Algorithm Reduction – Language Transformation

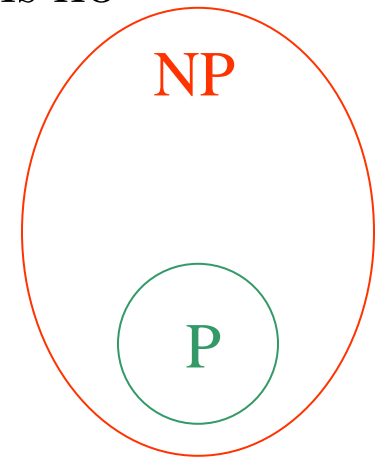
- The class P and Deterministic Turing Machine
  - Given a decision problem X, if there is a polynomial time Deterministic Turing Machine (or computer) program that solves X, then X belongs to P
  - *Informally, there is a polynomial time algorithm to solve the problem*
  - Polynomial Transformation (" $\propto$ ")
  - **$L1 \propto L2$**  : There is a polynomial time transformation that transforms arbitrary instance of L1 to some instance of L2.
  - **If  $L1 \propto L2$  then  $L2$  is in P implies  $L1$  is in P**  
(or  $L1$  is not in P implies  $L2$  is not in P)
  - **If  $L1 \propto L2$  and  $L2 \propto L3$  then  $L1 \propto L3$**



# NP-Completeness and Cooks Theorem

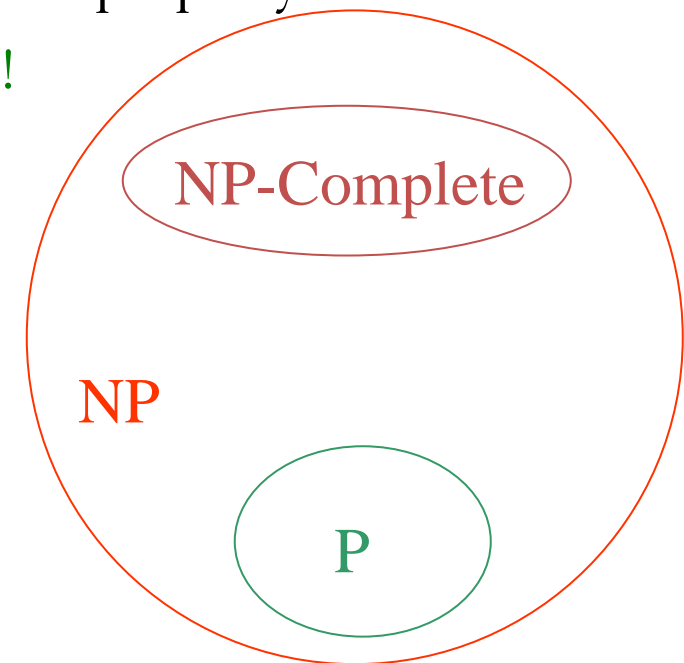
# P and NP

- Obvious :  $P \subseteq NP$ , i.e. A (decision) problem in P does not need “guess solution”.  
The correct solution can be computed in polynomial time.
- One of the most important open problem in theoretical compute science :  
**Is  $P=NP$  ?**
- Most likely “No”.
- Currently, there are many known (decision) problems in NP, and there is no solution to show anyone of them in P.



# P, NP, and NP-Complete

- P: (Decision) problems solvable by deterministic algorithms in polynomial time
- NP: (Decision) problems solved by non-deterministic algorithms in polynomial time
- A group of (decision) problems, including all of the ones we have discussed ([Satisfiability](#), [0/1 Knapsack](#), [Longest Path](#), [Partition](#)) have an additional important property:  
If any of them can be solved in polynomial time, then they all can!
- These problems are called NP-complete problems.



# Cooks Theorem

- Stephen Cook introduced the notion of NP-Complete Problems.
  - This makes the problem “ $P = NP$  ?” much more interesting to study.
  - $L$  is **NP-Complete** if  $L \in NP$  then for all other  $L' \in NP, L' \propto L$
  - $L$  is **NP-Hard** if for all other  $L' \in NP, L' \propto L$
  - If an NP-complete problem can be solved in polynomial time then all problems in NP can be solved in polynomial time.
  - If a problem in NP cannot be solved in polynomial time then all problems in NP-complete cannot be solved in polynomial time.
  - Note that an NP-complete problem is one of those hardest problems in NP.
- Lemma :  
If  $L1$  and  $L2$  belong to NP,  
 $L1$  is NP-complete, and  $L1 \propto L2$   
then  $L2$  is NP-complete.  
*i.e.  $L1, L2 \in NP$  and for all other  $L' \in NP, L' \propto L1$  and  $L1 \propto L2 \rightarrow L' \propto L2$*

# Cooks Theorem and Satisfiability

- SATISFIABILITY is NP-Complete. (The first NP-Complete problem)
  - Instance : Given a set of variables,  $U$ , and a collection of clauses,  $C$ , over  $U$ .
  - Question : Is there a truth assignment for  $U$  that satisfies all clauses in  $C$ ?
- CNF-Satisfiability (CNF – Conjunctive Normal Form)
  - because the expression is in (the product of sums).

- Example:

**“ $\neg x_i$ ” = “not  $x_i$ ”    “OR” = “logical or”    “AND” = “logical and”     $U = \{x_1, x_2\}$**

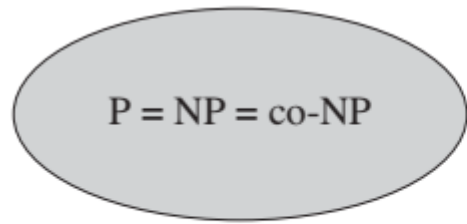
$$\begin{aligned} C_1 &= \{(x_1, \neg x_2), (\neg x_1, x_2)\} \\ &= (x_1 \text{ OR } \neg x_2) \text{ AND } (\neg x_1 \text{ OR } x_2) \\ &\quad \text{if } x_1 = x_2 = \text{True} \rightarrow C_1 = \text{True} \end{aligned}$$

$$C_2 = (x_1, x_2) (x_1, \neg x_2) (\neg x_1) \rightarrow \text{not satisfiable}$$

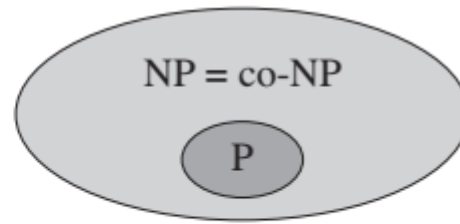
# Polynomial-Time Verifier

- NP = decision problems for which there exists a polynomial-time verifier
- Algorithm  $A$  with two inputs
  - input to the problem:  $x$
  - certificate:  $y$
- $A$  is polynomial-time verifier: for any  $x$  there exists certificate  $y$  such that
- $A(x, y)$  outputs
  - “yes” iff  $x$  is “yes”- instance, and  $A$  runs in polynomial time for such instances.
  - “No” - instances do not have to be verifiable in polynomial time.

# Verification algorithm

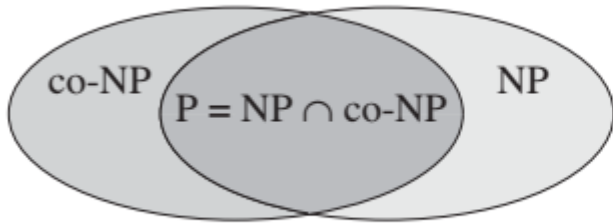


(a)



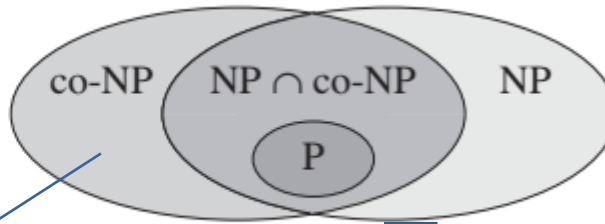
(b)

complexity class of NP co-NP

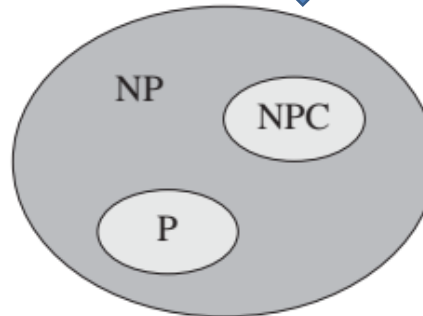


(c)

NP-Hard



(d)



**(a)** Most researchers regard this possibility as the most unlikely.

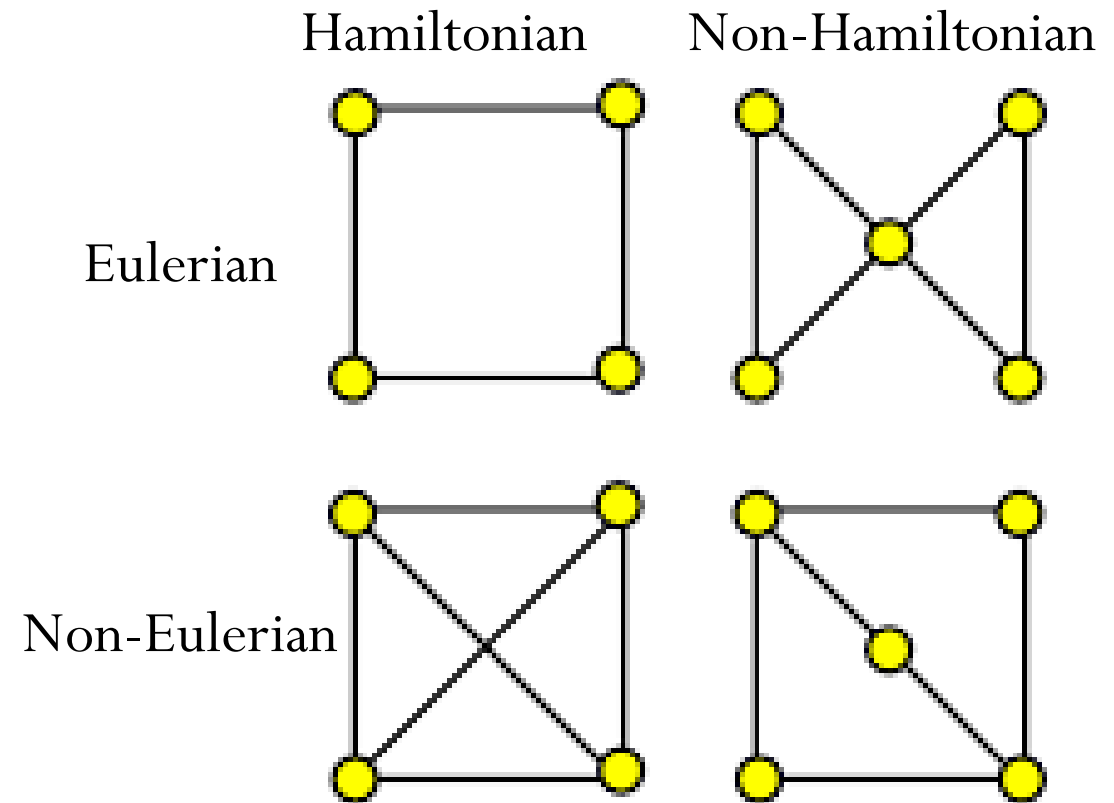
**(b)** but it need not be the case that  $P = NP$ .

**(c)** but  $NP$  is not closed under complement.

**(d)** Most researchers regard this possibility as the most likely.

# Euler tour is in P

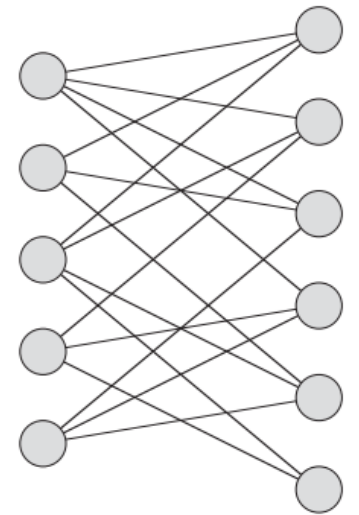
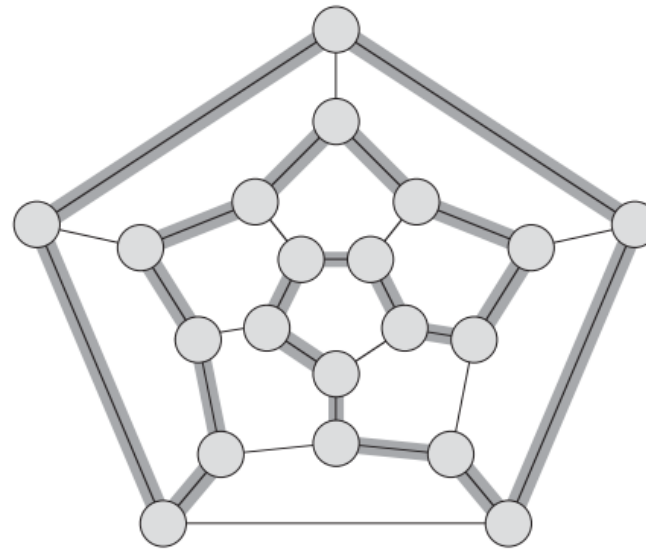
- **Euler tour:** An *Euler tour* of a connected, directed graph is a cycle that traverses each *edge* of  $G$  exactly once, although it is allowed to visit each vertex more than once.
- We can determine whether a graph has an Euler tour in only  $O(E)$  time and, in fact, we can find the edges of the Euler tour in  $O(E)$  time.





# Hamiltonian-cycle Problem is NP-Complete

- “Does a graph  $G$  have a Hamiltonian cycle?” **NP-Complete**
- A graph representing the vertices, edges, and faces of a dodecahedron, with a Hamiltonian cycle shown by shaded edges.
- A bipartite graph with an odd number of vertices. Any such graph is non-Hamiltonian.



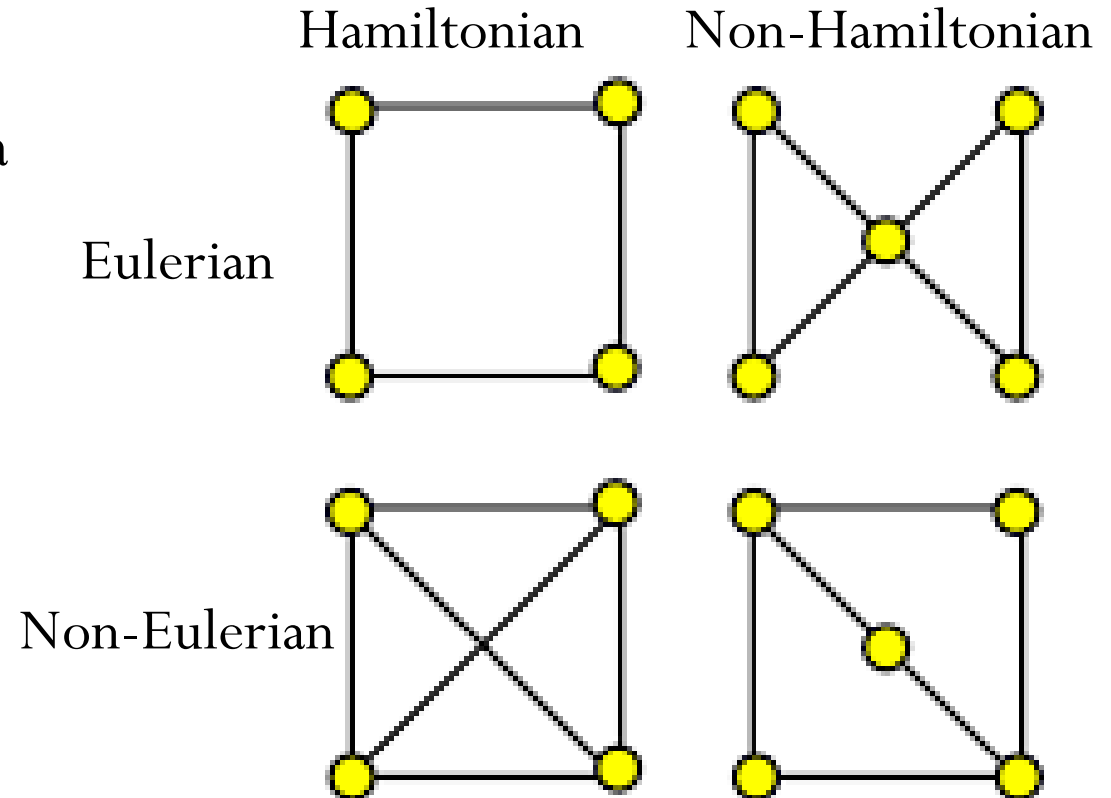
# Hamiltonian-cycle Problem is NP-Complete

- “Does a graph  $G$  have a Hamiltonian cycle?”

HAM-CYCLE

$$= \{ \langle G \rangle : G \text{ is a Hamiltonian graph} \}$$

- A Hamiltonian cycle of a directed graph is a simple cycle that contains each vertex in  $V$
- Determining whether a directed graph has a Hamiltonian cycle is NP-complete.
- Determining whether an undirected graph has a Hamiltonian cycle is NP-complete.



# NP-Complete and NP-Hard Problems

# P, NP, and NP-Hard

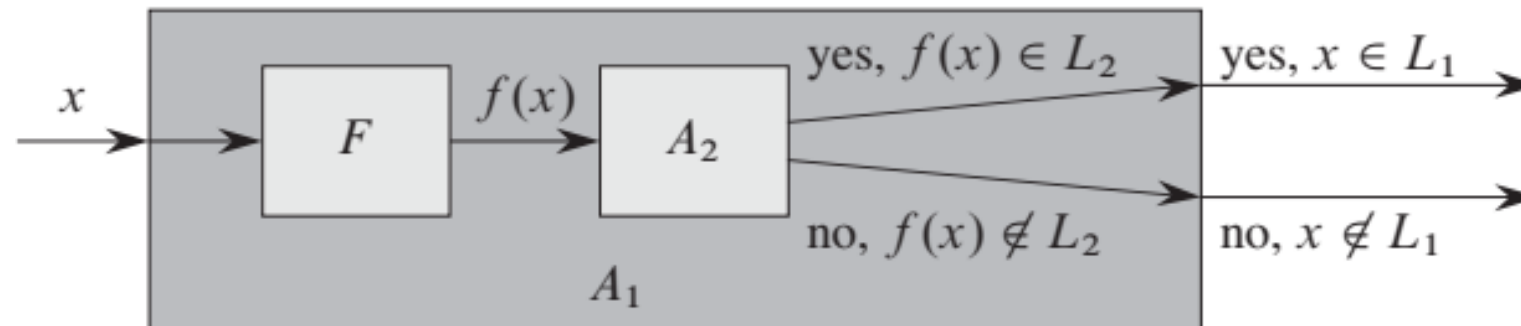
- $P$  = class of all problems for which we can **compute** a solution in polynomial time
- $P$  = problems solvable in polynomial time on Turing machine (or computer)
- $NP$  = class of problems for which we can **verify** a given solution in polynomial time
- $NP$  = problems solvable in polynomial time on non-deterministic Turing machine (or computer)
- anything that is computable is computable by a Turing machine
  - (*Church-Turing Thesis*)
- “polynomial time” on Turing machine  $\cong$  “polynomial time” on a Computer

# NP-Complete and NP-Hard

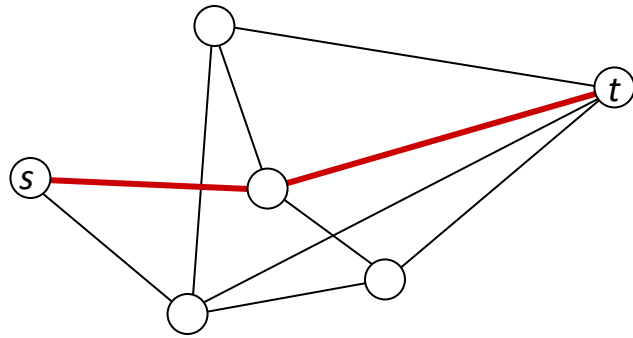
- $L_1 \leq_p L_2$ , then  $L_1$  is not more than a polynomial factor harder than  $L_2$ ,
- A language  $L \subseteq \{0, 1\}^*$  is **NP-complete** if
  1.  $L \in \text{NP}$ , and
  2.  $L' \leq_p L$  for every  $L' \in \text{NP}$ .
- If a language  $L$  satisfies property 2, but not necessarily property 1, we say that  $L$  is **NP-hard**.
- If  $L$  is a language such that  $L' \leq_p L$  for some  $L' \in \text{NPC}$ , then  $L$  is NP-hard. If, in addition,  $L \in \text{NP}$ , then  $L \in \text{NPC}$ .

# NP-Complete and NP-Hard

- Algorithm  $F$  is a reduction algorithm that computes the reduction function  $f$  from  $L_1$  to  $L_2$  in polynomial time, and
- $A_2$  is a polynomial-time algorithm that decides  $L_2$ .
- Algorithm  $A_1$  decides whether  $x \in L_1$  by using  $F$  to transform any input  $x$  into  $f(x)$  and then using  $A_2$  to decide whether  $f(x) \in L_2$ .



# Two similar graph problems

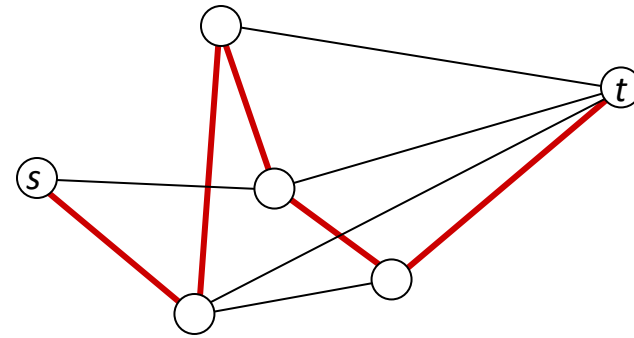


## Shortest Path

**Input:** graph, nodes  $s$  and  $t$

**Output:** simple  $s$ -to- $t$  path with **minimum** number of edges

BFS:  $O(|V| + |E|)$



## Longest Path

**Input:** graph, nodes  $s$  and  $t$

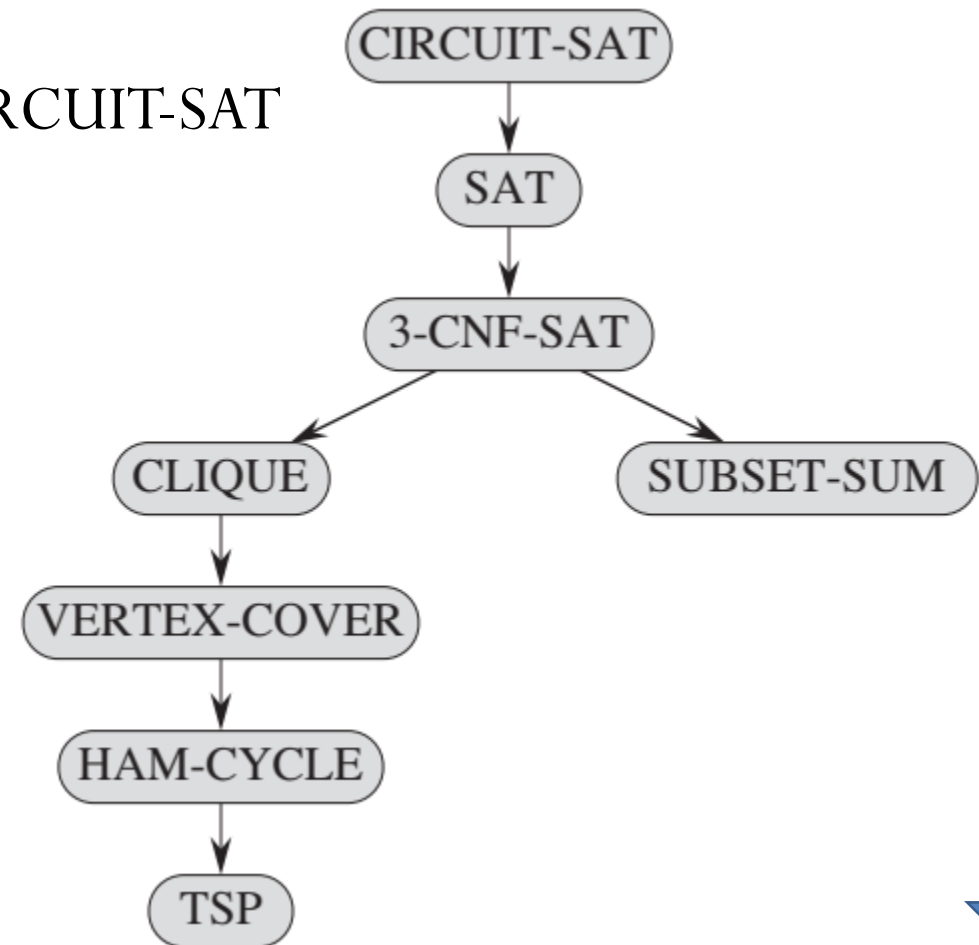
**Output:** simple  $s$ -to- $t$  path with **maximum** number of edges

no polynomial-time algorithm known

$O(n^c)$  for some constant  $c$

# NP-Complete problems

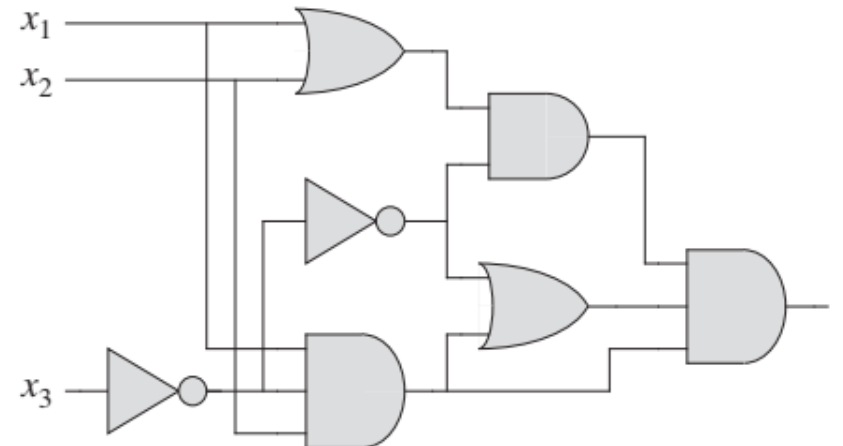
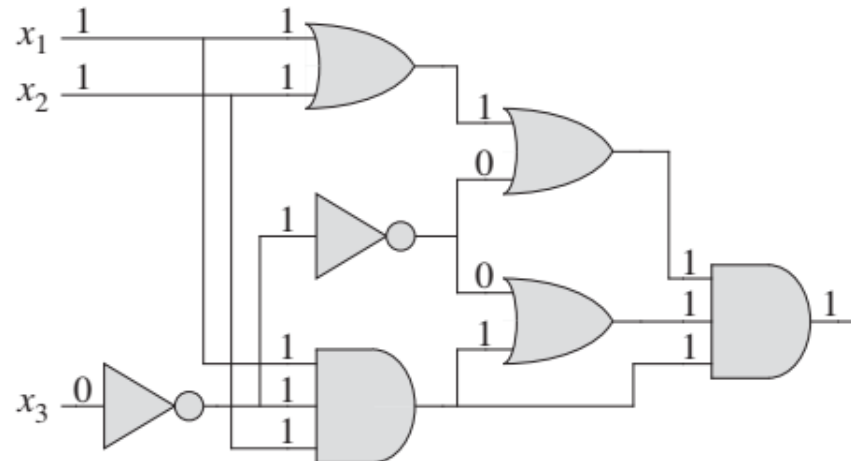
- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT





# Circuit-Satisfiability problem is NP-Complete

- “Given a Boolean combinational circuit composed of AND, OR, and NOT gates, is it satisfiable?”  $O(n^k)$
- **(a)** The assignment  $\langle x_1 = 1, x_2 = 1, x_3 = 0 \rangle$  to the inputs of this circuit causes the output of the circuit to be 1. The circuit is therefore satisfiable.
- **(b)** No assignment to the inputs of this circuit can cause the output of the circuit to be 1. The circuit is therefore unsatisfiable



# Two similar problems on Boolean

$$(x_1 \vee \neg x_3) \wedge (x_2 \vee x_5) \wedge (x_3 \vee \neg x_4)$$

## 2-SAT

Input: 2-CNF formula

Output: “yes” if formula can be satisfied, “no” otherwise

$O(\text{\#clauses})$

$$(x_1 \vee \neg x_2 \vee x_3) \wedge (x_2 \vee x_3 \vee \neg x_5) \wedge (x_1 \vee x_3 \vee x_4)$$

## 3-SAT

Input: 3-CNF formula

Output: “yes” if formula can be satisfied, “no” otherwise

no polynomial-time algorithm known

# Boolean Formula Satisfiability (SAT) is NPC

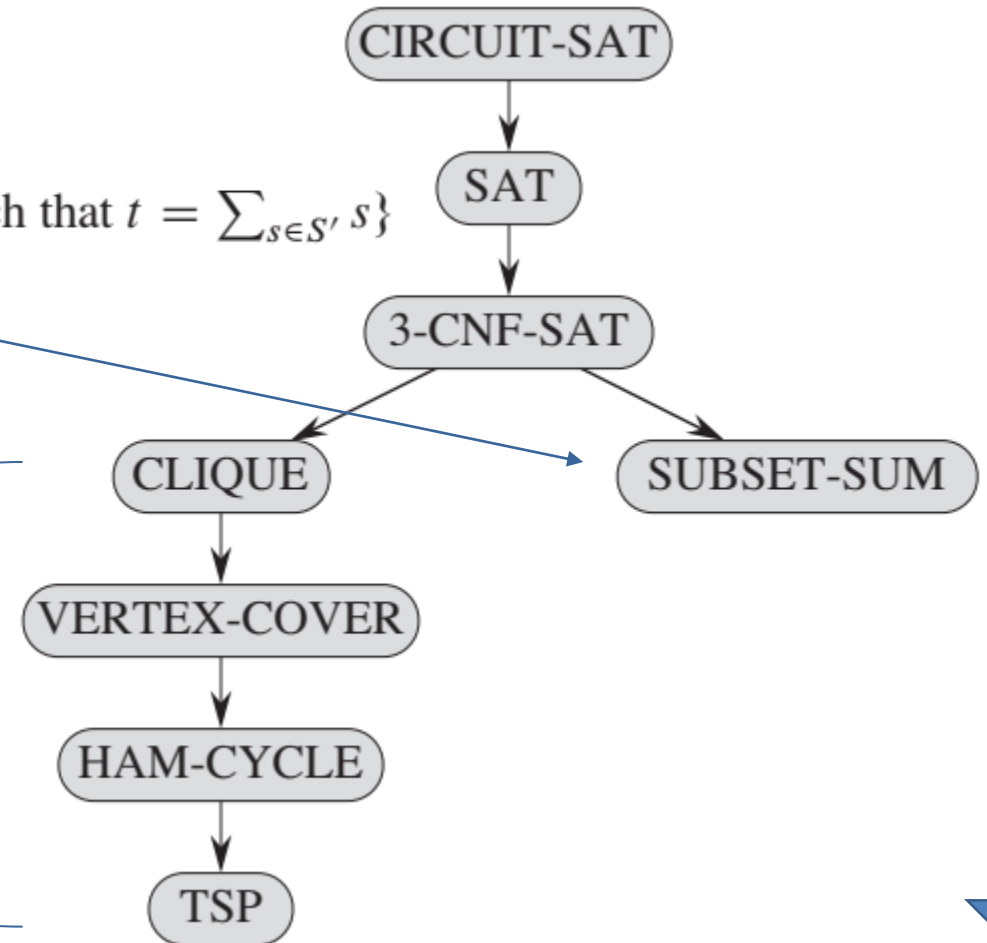
- Satisfiability of boolean formulas is NP-Complete.
  1.  $n$  boolean variables:  $x_1, x_2, \dots, x_n$
  2.  $m$  boolean connectives
  3. parentheses
- $\Omega(2^n)$  asymptotic lower bound is  $2^n$
- $\text{CIRCUIT-SAT} \leq_p \text{SAT}$
- Satisfiability of boolean formulas in 3-conjunctive normal form is NP-Complete.

# Subset-Sum Problem is NP-Complete

$3\text{-CNF-SAT} \leq_p \text{SUBSET-SUM}$

$\text{SUBSET-SUM} = \{ \langle S, t \rangle : \text{there exists a subset } S' \subseteq S \text{ such that } t = \sum_{s \in S'} s \}$

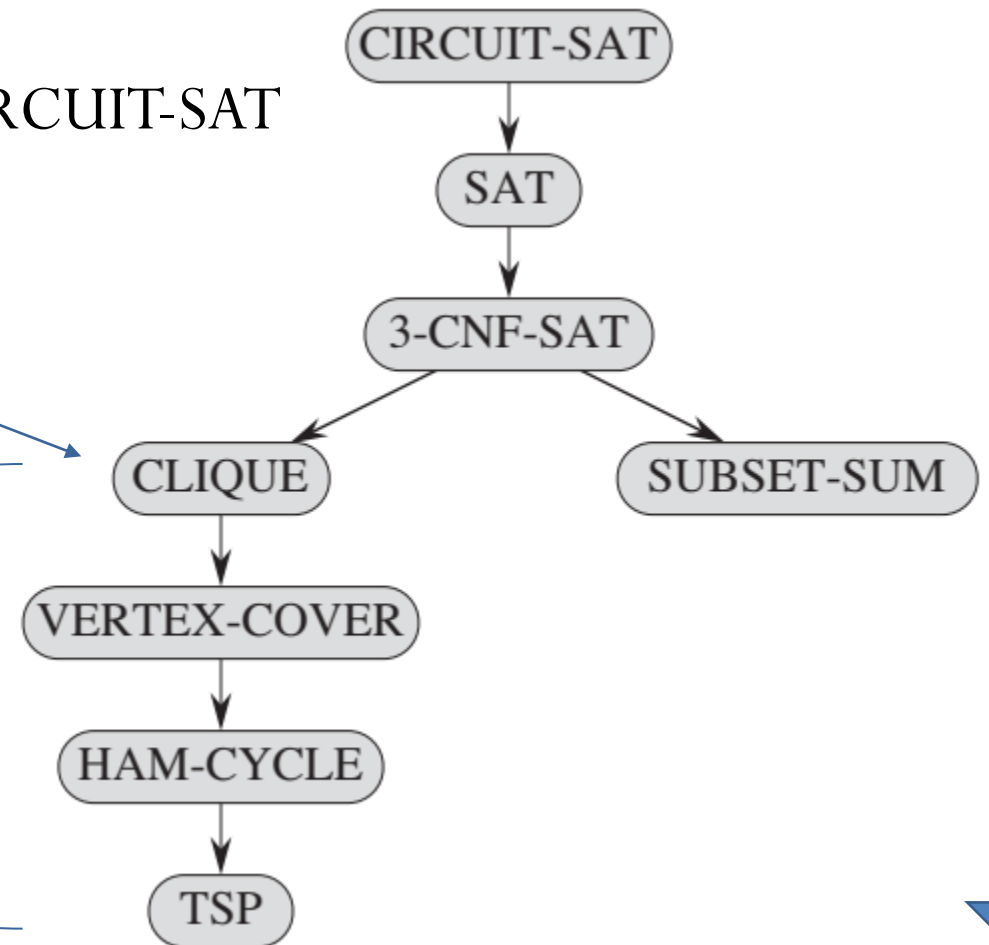
We will only explore Graph Problems!



# NP-Complete problems

- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT
- $3\text{-CNF-SAT} \leq_p \text{CLIQUE}$

We will only explore Graph Problems!

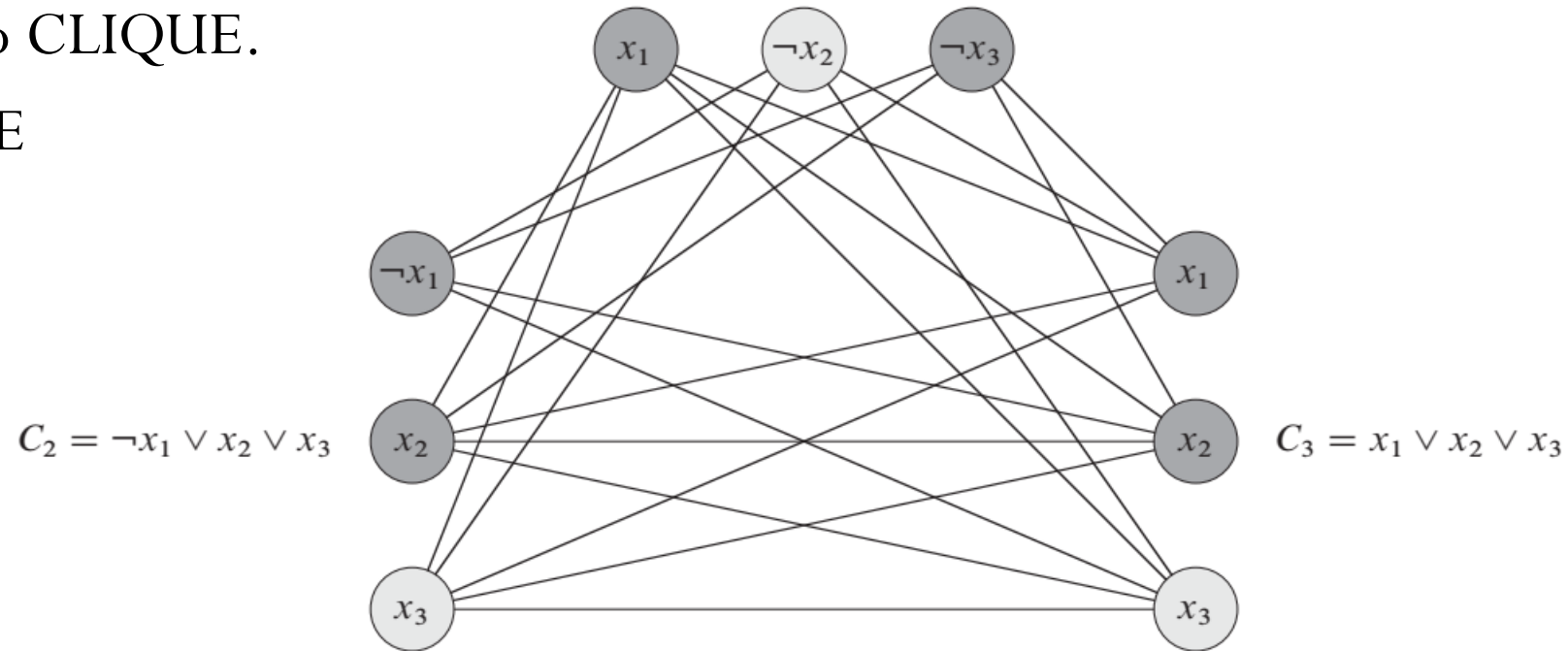


# Clique Problem is NP-Complete

- A clique in an undirected graph is a subset  $V' \subseteq V$  of vertices, each pair of which is connected by an edge in  $E$ .
- In other words, a clique is a complete subgraph of  $G$ . The size of a clique is the number of vertices it contains.
- The clique problem is the optimization problem of finding a clique of maximum size in a graph. As a decision problem, we ask simply
- “Whether a clique of a given size  $k$  exists in the graph”.

# Clique Problem is NP-Complete

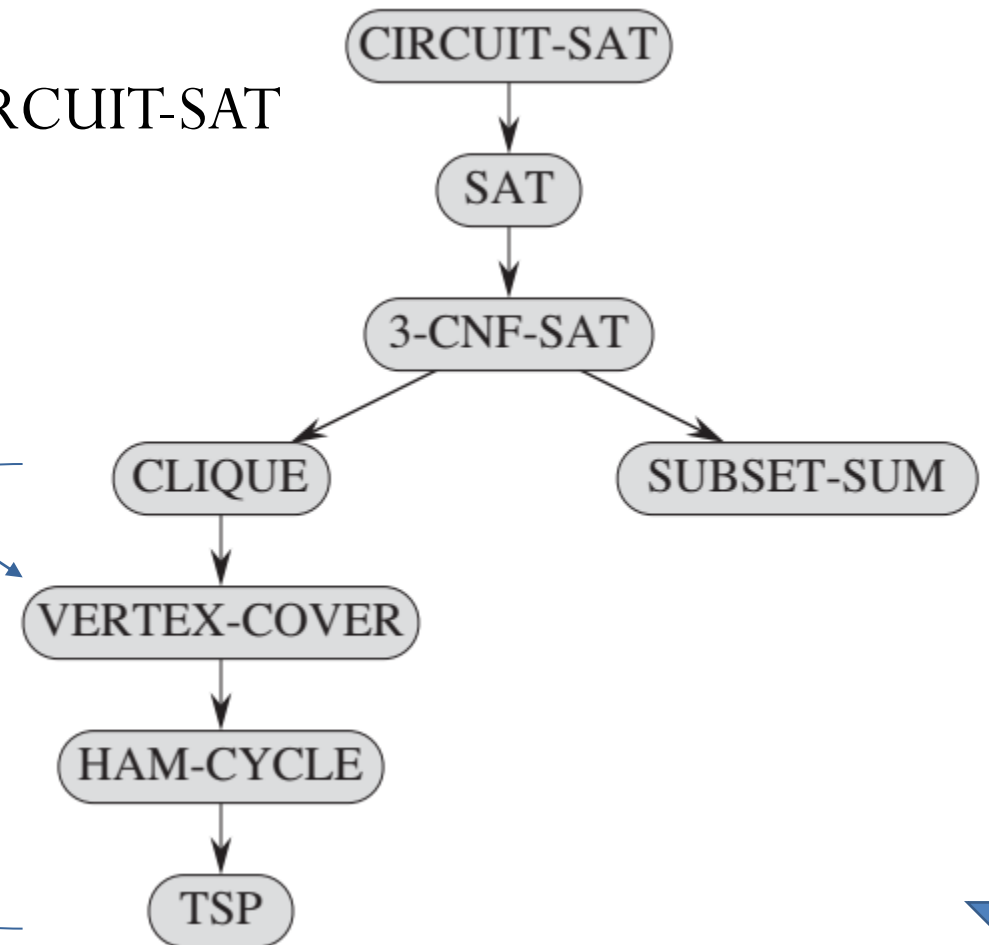
- “Whether a clique of a given size  $k$  exists in the graph”.
- $\text{CLIQUE} = \{ \langle G, k \rangle : G \text{ is a graph containing a clique of size } k \}$
- The graph  $G$  derived from the 3-CNF formula,  $C_1 = x_1 \vee \neg x_2 \vee \neg x_3$
- reducing 3-CNF-SAT to CLIQUE.
- $3\text{-CNF-SAT} \leq_p \text{CLIQUE}$



# NP-Complete problems

- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT
- $\text{CLIQUE} \leq_p \text{VERTEX-COVER}$

We will only explore Graph Problems!



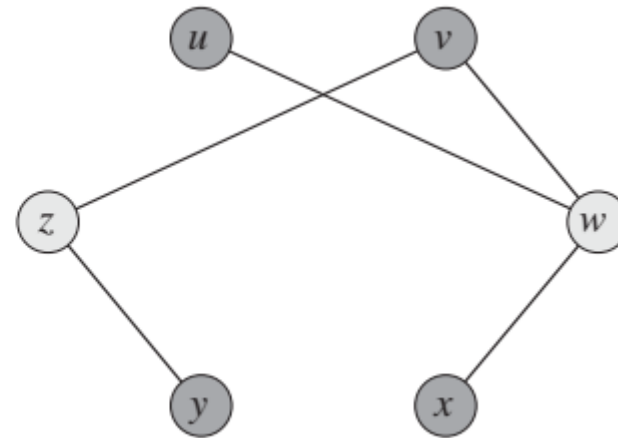
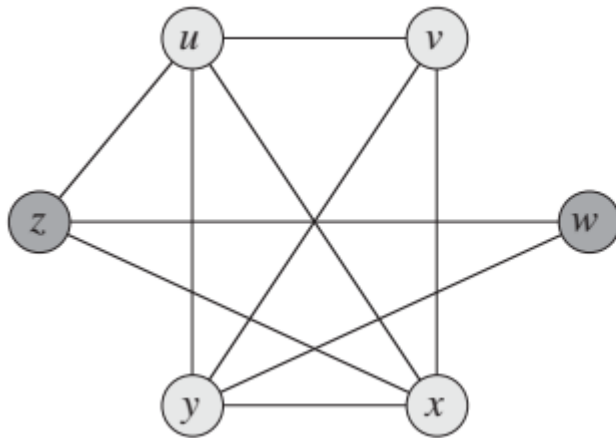


# Vertex cover problem is NP-Complete

- A *vertex cover* of an undirected graph is a subset  $V' \subseteq V$  such that if  $(u, v) \in E$ , then  $u \in V'$  or  $v \in V'$  (or both). That is, each vertex “covers” its incident edges, and a vertex cover for  $G$  is a set of vertices that covers all the edges in  $E$ .
- The *size* of a vertex cover is the number of vertices in it.
- The *vertex-cover problem* is to find a vertex cover of minimum size in a given graph.
- “Determine whether a graph has a vertex cover of a given size  $k$ .”
- $\text{VERTEX-COVER} = \{ \langle G, k \rangle : \text{graph } G \text{ has a vertex cover of size } k \}.$

# Vertex cover problem is NP-Complete

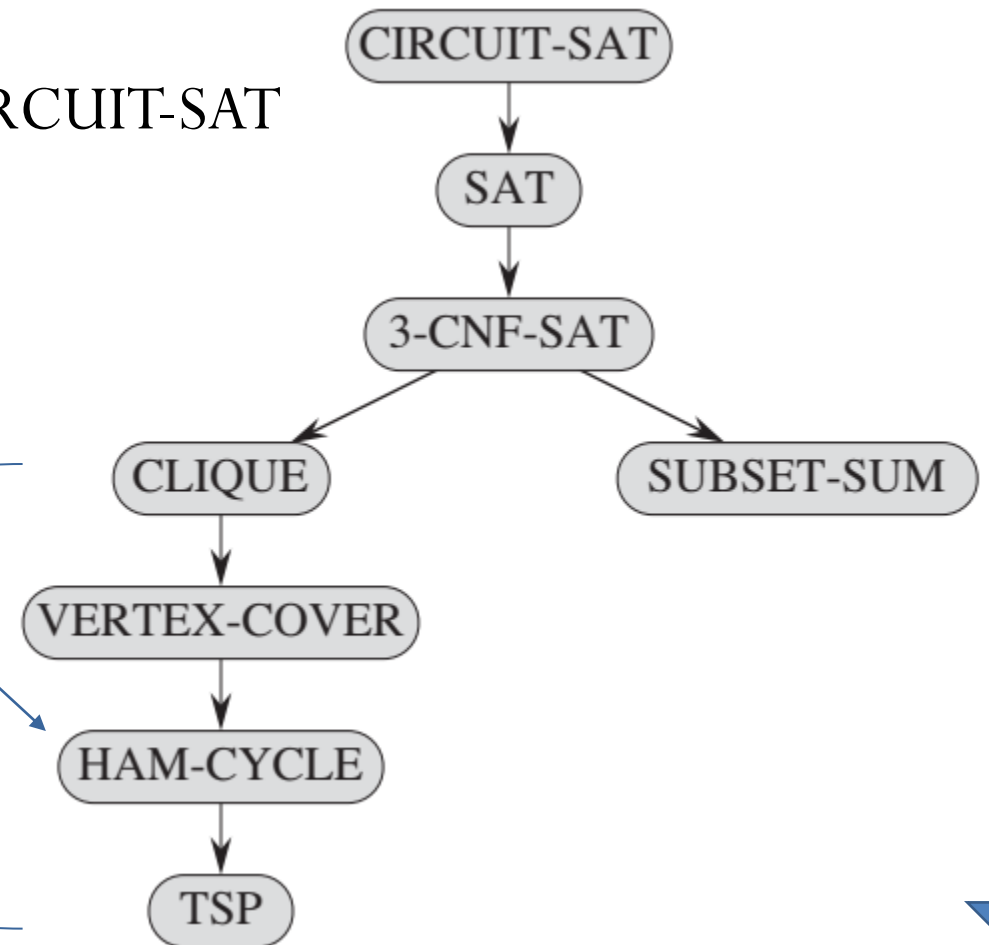
- The *vertex-cover problem* is to find a vertex cover of minimum size in a given graph.
- “Determine whether a graph has a vertex cover of a given size  $k$ .”
- Reducing CLIQUE to VERTEX-COVER
- $\text{CLIQUE} \leq_p \text{VERTEX-COVER}$



# NP-Complete problems

- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT
- $\text{VERTEX-COVER} \leq_p \text{HAM-CYCLE}$

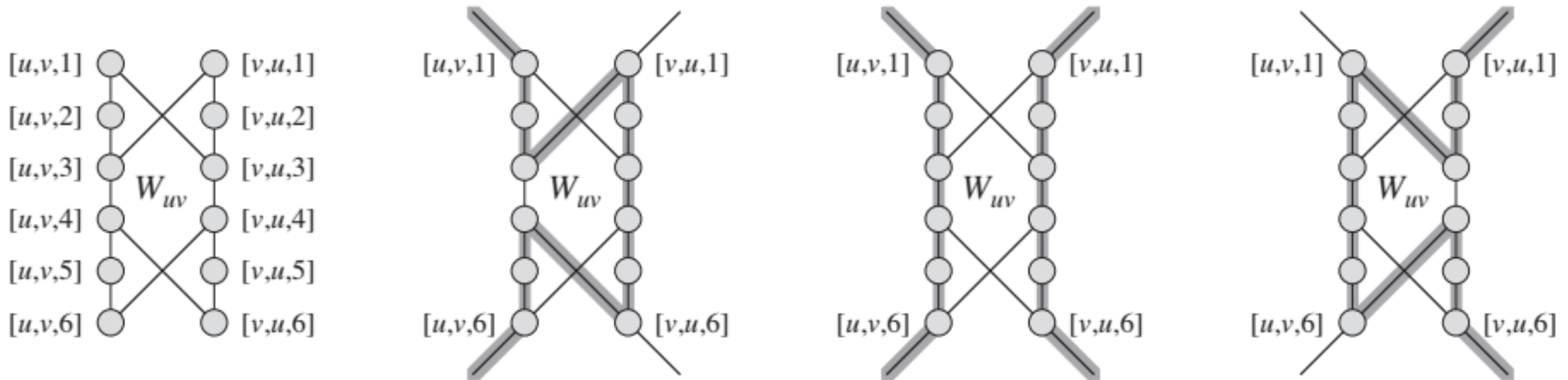
We will only explore Graph Problems!



Hardness Increase

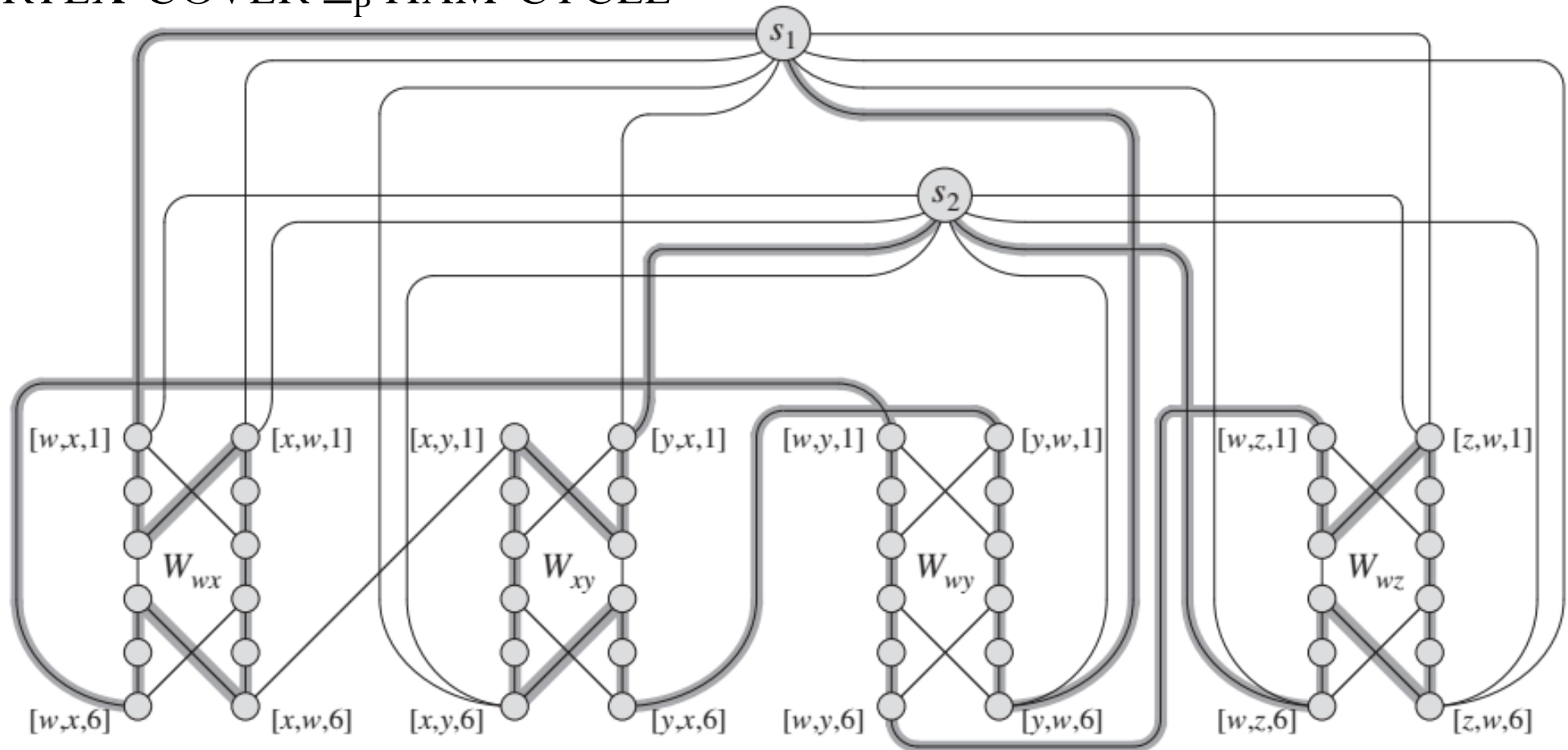
# Hamiltonian cycle problem is NP-Complete

- $\text{VERTEX-COVER} \leq_p \text{HAM-CYCLE}$



# Hamiltonian cycle problem is NP-Complete

- VERTEX-COVER  $\leq_p$  HAM-CYCLE

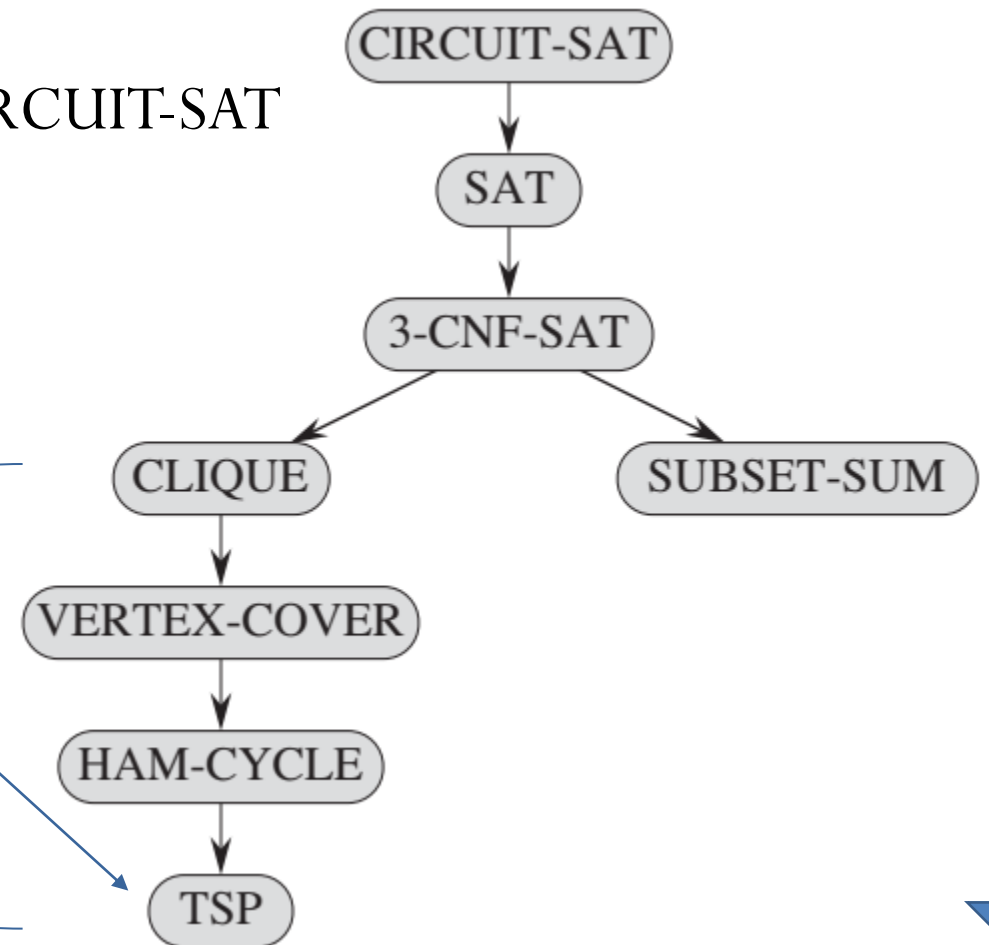


Cormen, T. H., Leiserson, C. E., Rivest, R. L., & Stein, C. (2009). *Introduction to Algorithms* (Vol. 3, pp. 624-642). Cambridge: MIT press.

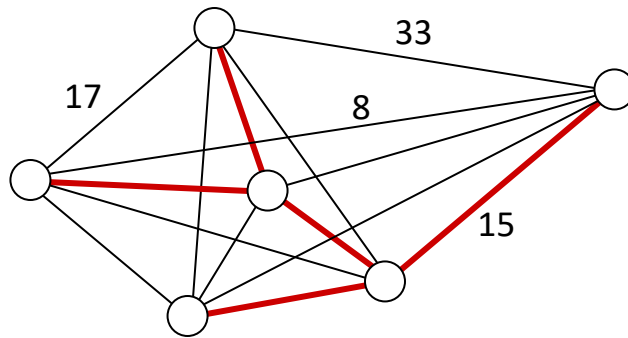
# NP-Complete problems

- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT
- $\text{HAM-CYCLE} \leq_p \text{TSP}$

We will only explore Graph Problems!



# Two similar graph problems

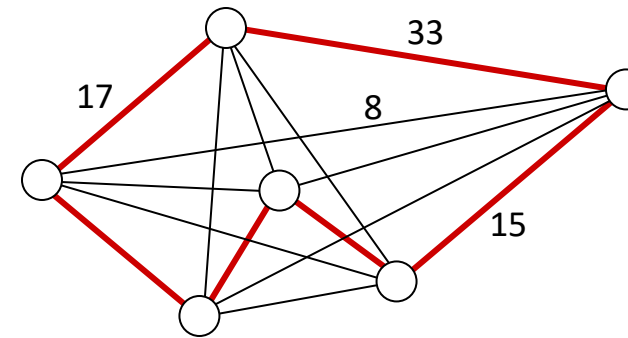


## Min Spanning Tree

**Input:** weighted graph

**Output:** minimum-weight **tree** connecting all nodes

greedy:  $O(|E| + |V| \log |V|)$



## Traveling Salesman (TSP)

**Input:** complete weighted graph

**Output:** minimum-weight **tour** connecting all nodes

backtracking:  $O(|V|!)$

# Traveling-Salesman Problem is NP-Complete

- "Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city exactly once and returns to the origin city?"

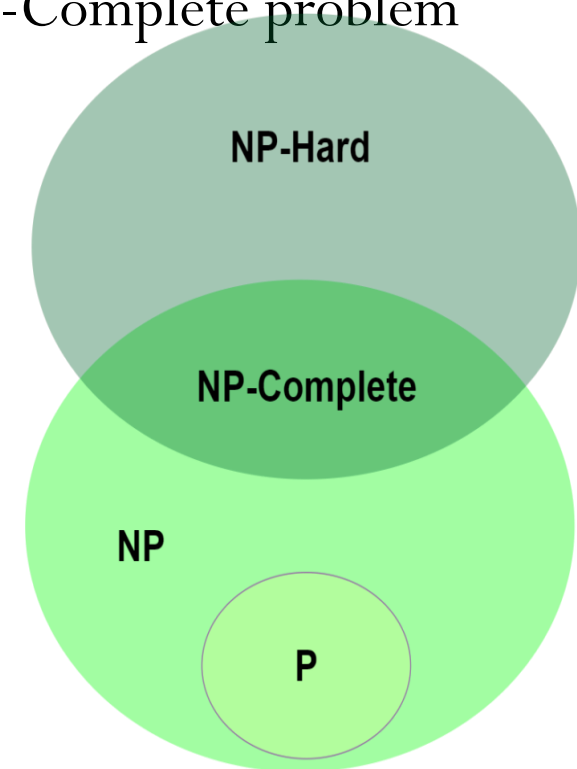
$$\text{TSP} = \{ \langle G, c, k \rangle : G = (V, E) \text{ is a complete graph,} \\ c \text{ is a function from } V \times V \rightarrow \mathbb{Z}, \\ k \in \mathbb{Z}, \text{ and} \\ G \text{ has a traveling-salesman tour with cost at most } k \} .$$

- $\text{HAM-CYCLE} \leq_p \text{TSP}$



# NP-Hard Problem

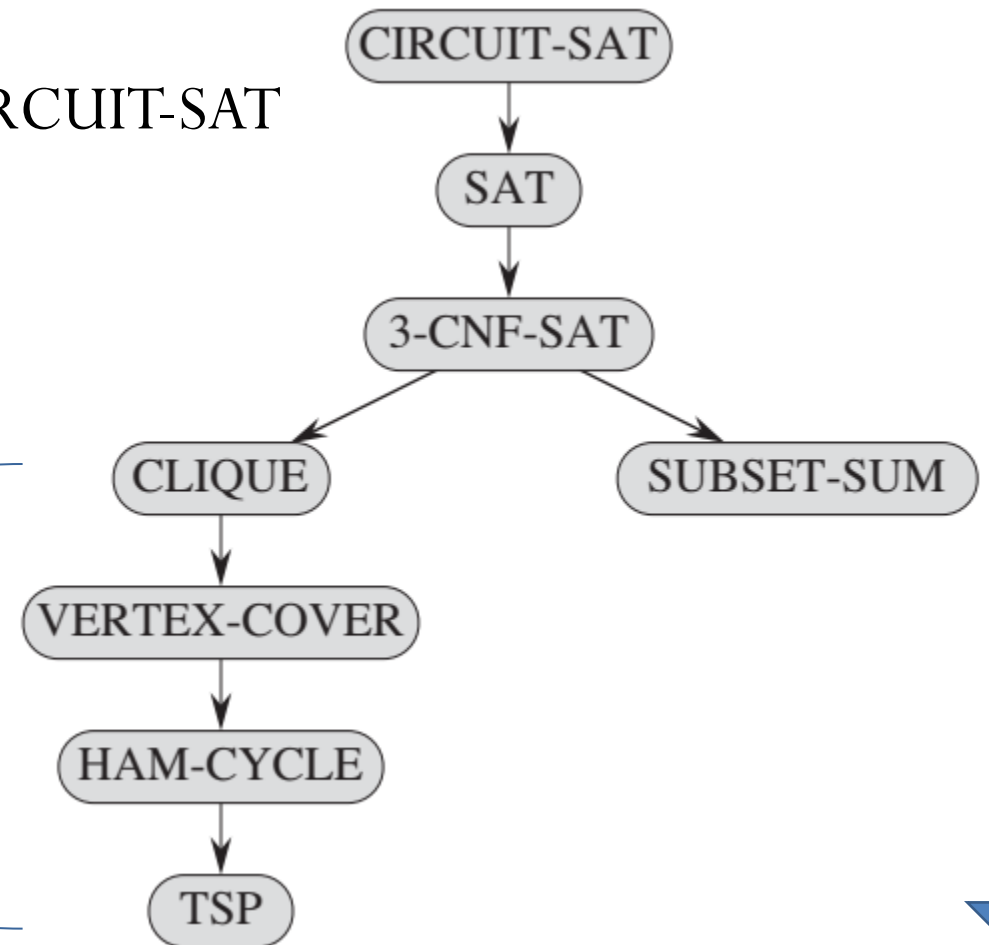
- L is **NP-Complete** *if  $L \in NP$  then for all other  $L' \in NP, L' \propto L$*
- L is **NP-Hard** if *for all other  $L' \in NP, L' \propto L$*
- Note that an NP-Hard problem is a problem which is as hard as an NP-Complete problem and it's not necessary a decision problem.
- So, if an NP-complete problem is in P then  $P=NP$
- if  $P \neq NP$  then all NP-complete problems are in NP-P
- For 3-SAT, TSP, Longest Path
  - Are we capable to write fast (polynomial-time) algorithms  
OR are these problems unsolvable ? *we don't know:  $P \neq NP$  ?*
  - 3-SAT, TSP, Longest Path are so-called **NP-hard** problems  
if  $P \neq NP$  (which most researchers believe to be the case)  
then NP-hard problems cannot be solved in polynomial time



# NP-Complete problems

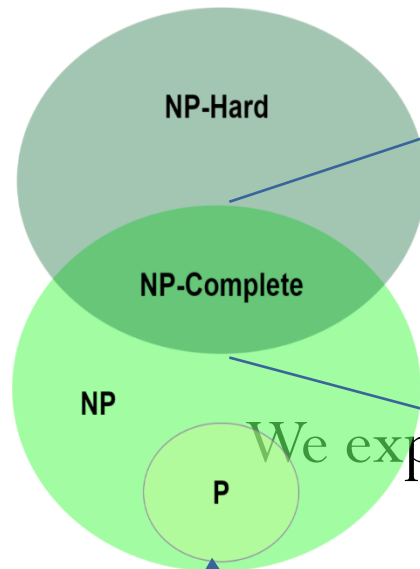
- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT
- Done!

We explored all Graph Problems!

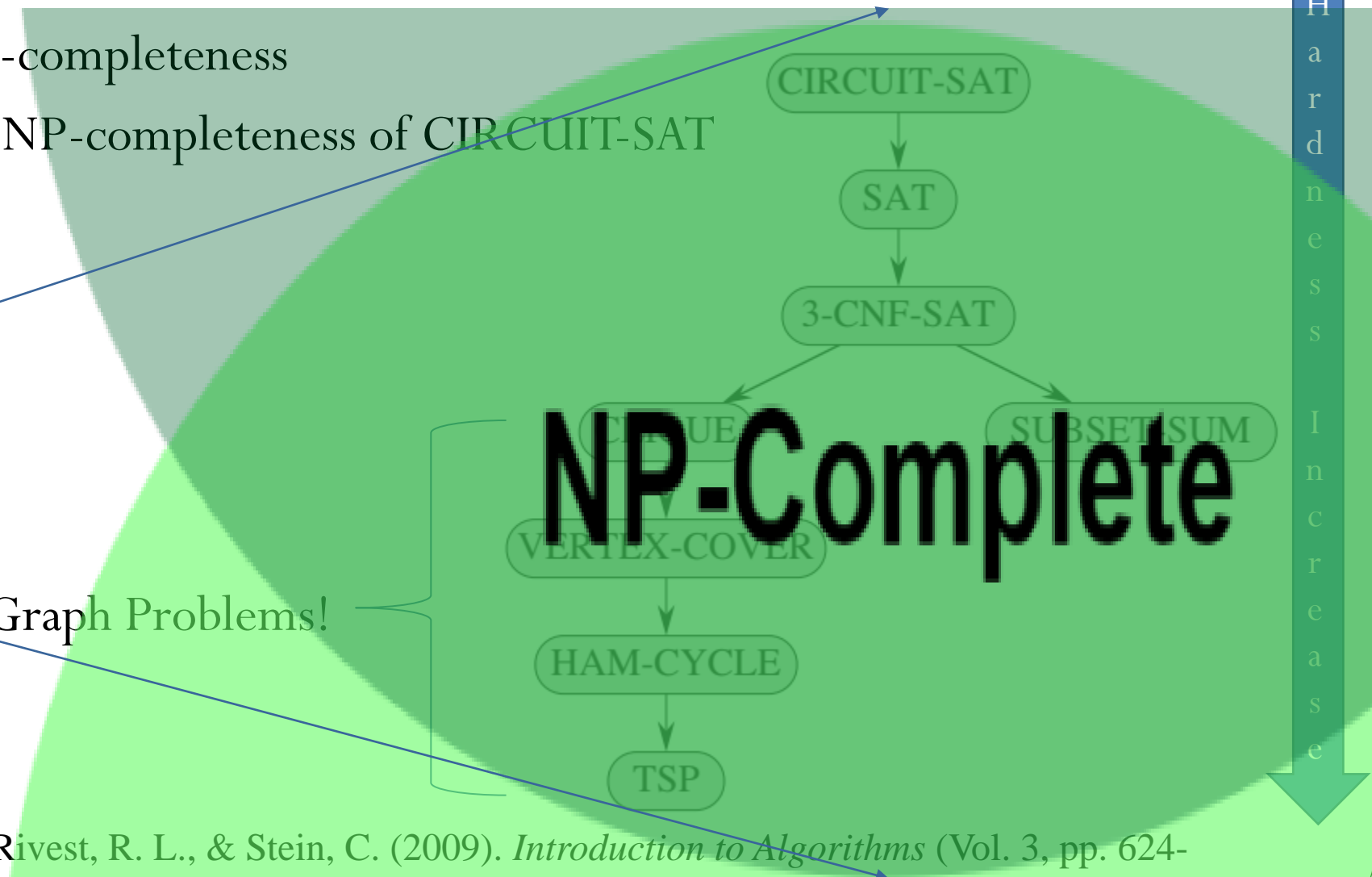


# NP-Complete problems

- The structure of NP-completeness
- Reduction from the NP-completeness of CIRCUIT-SAT
- Done!



We explored all Graph Problems!



# Travelling Salesman Problem (TSP)

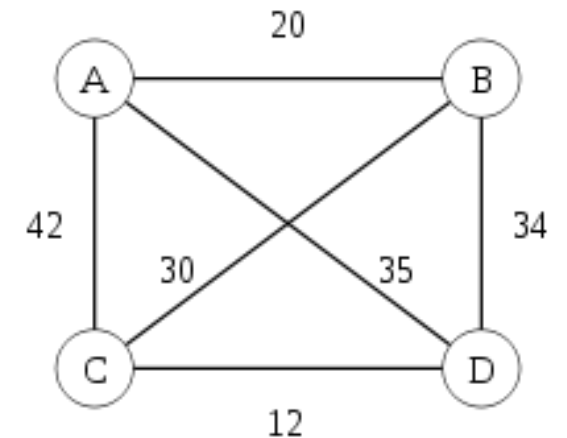
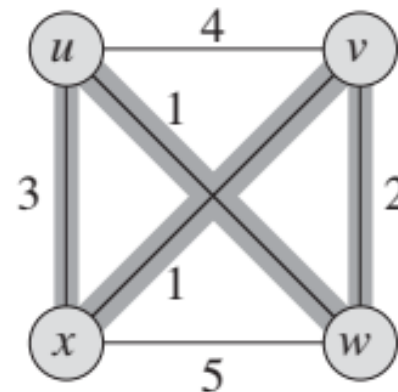
# Travelling Salesman Problem (TSP)

- "Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city exactly once and returns to the origin city?"
- **Routing problem:** Is there a route of at most 2000 kilometers passing through all of Germany's 15 largest cities?
- **DNA sequencing:**
  - to determine the order of  
Adenine, Thymine, Guanine, Cytosine (ATGC)
  - slightly modified Travelling Salesman Problem (TSP),
  - it appears as a sub-problem in DNA sequencing



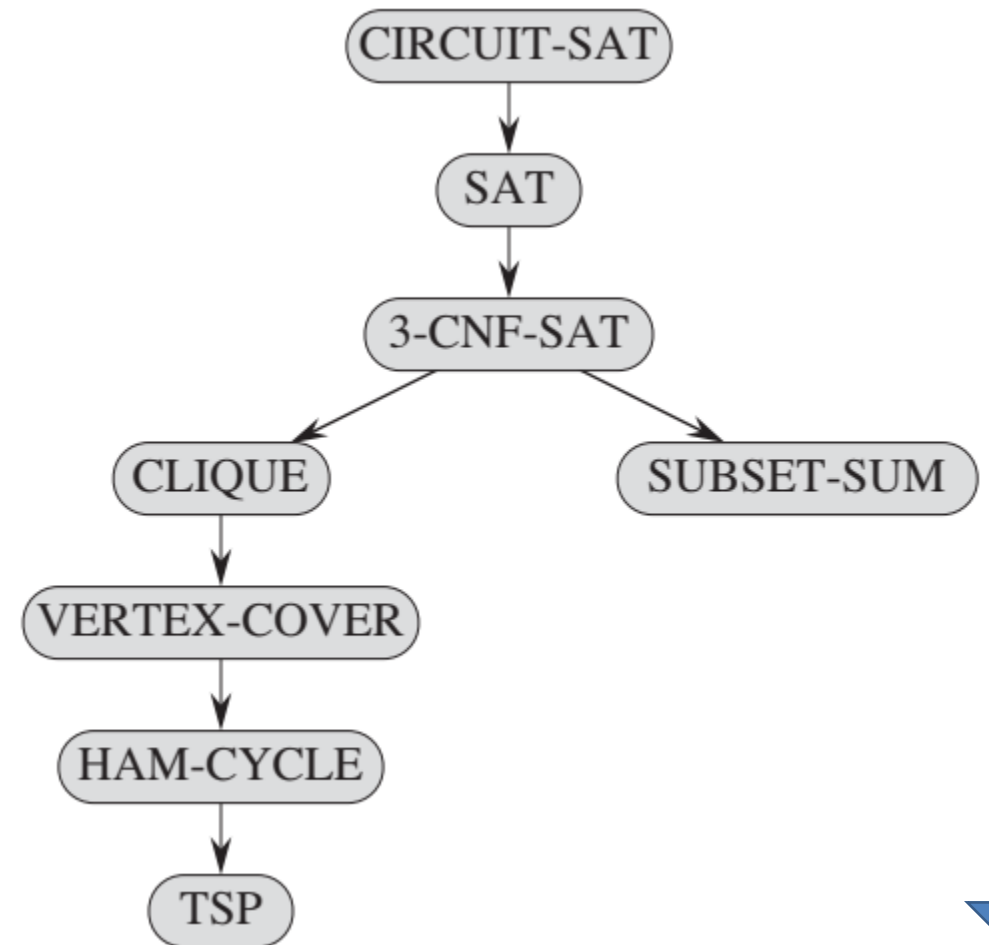
# Travelling Salesman Problem (TSP)

- TSP can be modelled as an undirected weighted graph, such that
  - cities are the graph's vertices,
  - paths are the graph's edges, and
  - a path's distance is the edge's weight.
- **Symmetric TSP:** the distance between two cities is the same in each opposite direction, forming an undirected graph.
- **Asymmetric TSP:** paths may not exist in both directions, or the distances might be different, forming a directed graph.



# Travelling Salesman Problem (TSP)

- Hamiltonian-Cycle Problem is not more than a polynomial factor harder than Traveling Salesman Problem
  - $\text{HAM-CYCLE} \leq_p \text{TSP}$
  - TSP is NP-Complete
- **An NP-hard problem:**
  - **Brute-Force Search:** Try all permutations (ordered combinations) to find shortest route.
  - Running time is  $O(n!)$ , the factorial of the number of cities,
  - This solution becomes impractical even for 20 cities.



# TSP - Approximation Algorithms

- TSP is an NP-hard problem
- TSP is solvable for few vertices problem about  $V = 15$ , depending upon computational power
- Even polynomial algorithms also fails on Big Graph for large value of  $V$  and  $E$ 
  - Minimum Spanning Tree: Kruskal's  $O(E \lg V)$  and Prim's  $O(E + V \lg V)$  algorithm
  - Shortest Path : Dijkstra's  $O(E \lg V)$  and Bellman-Ford's  $O(EV)$  algorithm
  - Limitation of computing power of single machine may make the polynomial algorithm fail on Big Graphs.
  - Parallel Computing environment might be the solution
    - for example, **Apache Spark GraphX**.
- TSP – Approximation Algorithm provides approximate solution to given Graph.



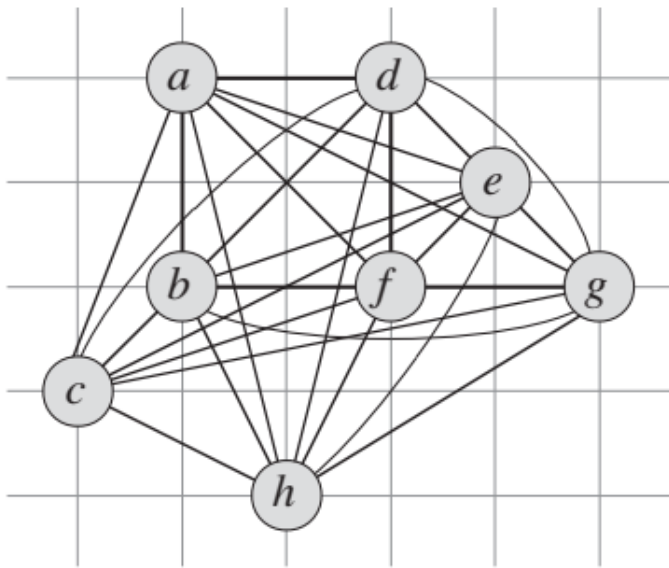
# Travelling Salesman Problem (TSP)

## Approximation Algorithms

# Travelling Salesman Problem: Approximation

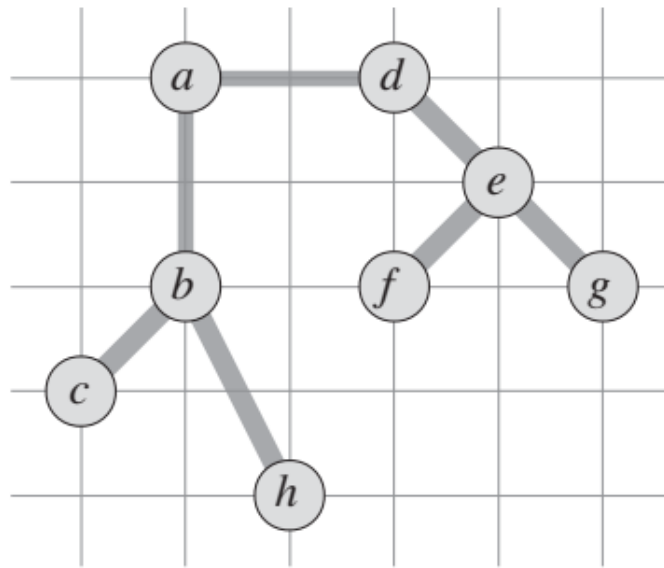
APPROX-TSP-TOUR( $G, c$ )

- 1 select a vertex  $r \in G.V$  to be a “root” vertex
- 2 compute a minimum spanning tree  $T$  for  $G$  from root  $r$   
using MST-PRIM( $G, c, r$ )
- 3 let  $H$  be a list of vertices, ordered according to when they are first visited  
in a preorder tree walk of  $T$
- 4 **return** the hamiltonian cycle  $H$



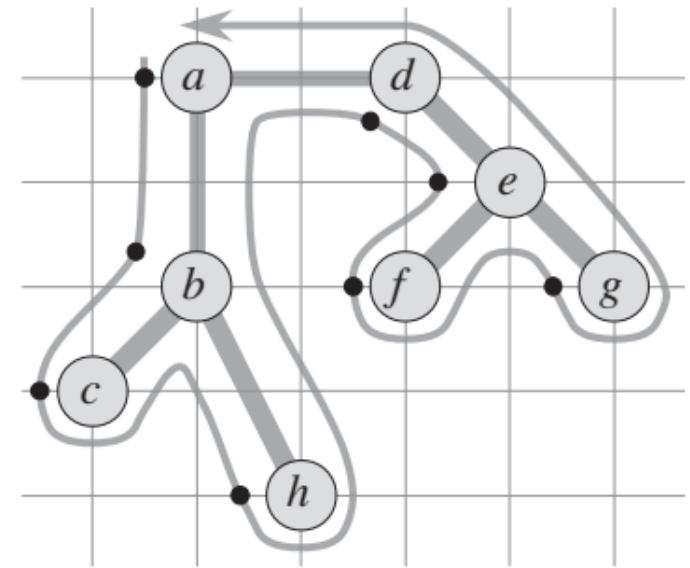
(a)

A complete undirected graph with ordinary Euclidean distance.



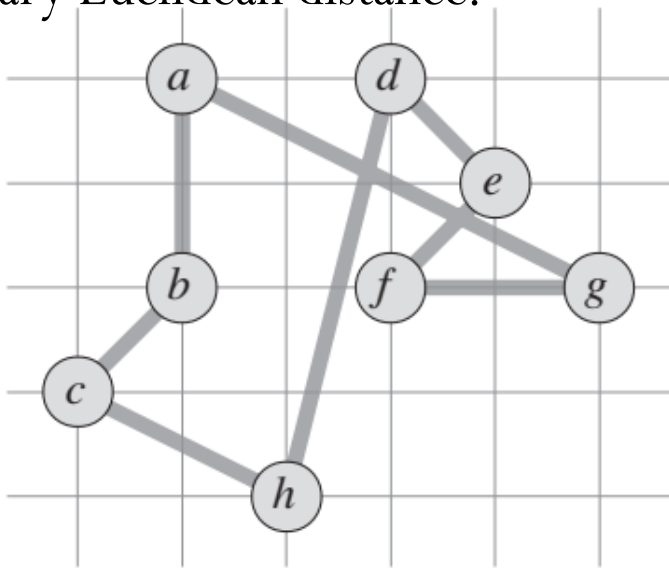
(b)

MST computed by MST-PRIM  
'a' is the root vertex

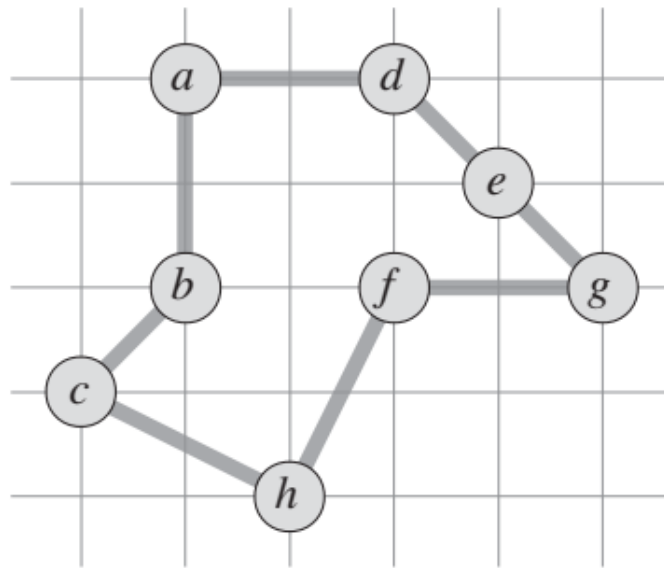


(c)

A walk of  $T$ , starting at  $a$ .  
Yielding the ordering  $a; b; c; h; d; e; f; g$ .



(d)



(e)

(d) Tour  $H$  by APPROX-TSP-TOUR  
approximately 19.074.

(e) An optimal tour approximately  
14.715.

APPROX-TSP-TOUR is  $\theta(V^2)$

# Travelling Salesman Problem: Approximation

## **Christofides-Serdyukov algorithm**

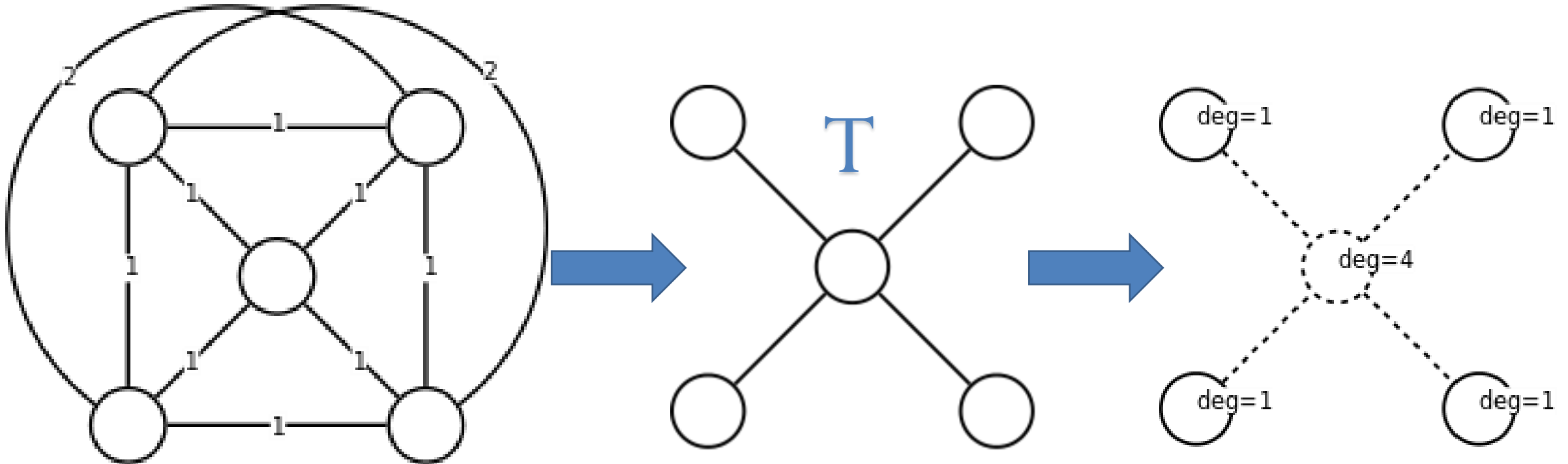
- In the worst case, is at most 1.5 times longer than the optimal solution.
- This remains the method with the best worst-case scenario.
- This gives a TSP tour which is at most 1.5 times the optimal.
- It was one of the first approximation algorithms
- Drawn attention to approximation algorithms as a practical approach to intractable problems.

# Christofides-Serdyukov algorithm

- Create a minimum spanning tree  $T$  of  $G$ .
- Let  $O$  be the set of vertices with odd degree in  $T$ . By the handshaking lemma,  $O$  has an even number of vertices.
- Find a minimum-weight perfect matching  $M$  in the induced subgraph given by the vertices from  $O$ .
- Combine the edges of  $M$  and  $T$  to form a connected multigraph  $H$  in which each vertex has even degree.
- Form an Eulerian circuit in  $H$ .
- Make the circuit found in previous step into a Hamiltonian circuit by skipping repeated vertices (shortcutting).

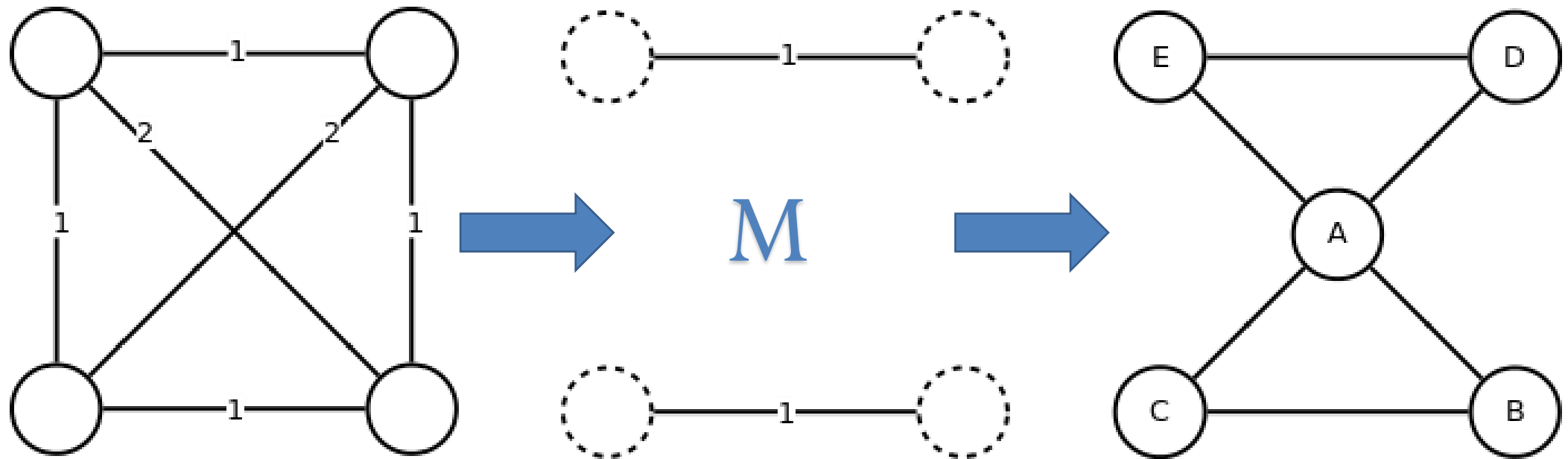
# Christofides-Serdyukov algorithm

- Given: complete graph whose edge weights obey the triangle inequality
- Calculate minimum spanning tree  $T$
- Calculate the set of vertices  $O$  with odd degree in  $T$



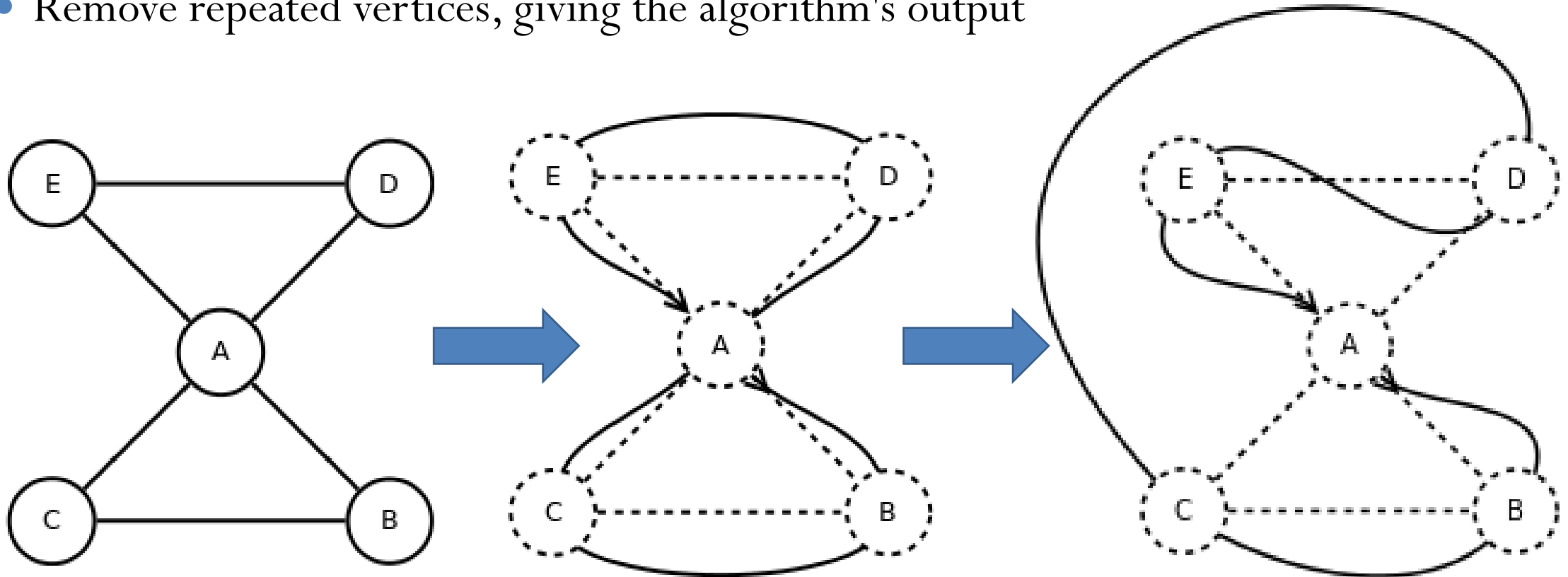
# Christofides-Serdyukov algorithm

- Form the subgraph of  $G$  using only the vertices of  $O$
- Construct a minimum-weight perfect matching  $M$  in this subgraph
- Unite matching and spanning tree  $T \cup M$  to form an Eulerian multigraph



# Christofides-Serdyukov algorithm

- Unite matching and spanning tree  $T \cup M$  to form an Eulerian multigraph
- Calculate Euler tour
- Remove repeated vertices, giving the algorithm's output





PRIMES is in P  
(A hope for NP problems in P)

# Basics on Prime number

- A **prime number** (or a prime) is a natural number greater than 1 that is not a product of two smaller natural numbers.
- A natural number greater than 1 that is not prime is called a **composite number**.
- For example,
  - 5 is prime because the only ways of writing it as a product,  $1 \times 5$  or  $5 \times 1$ , involve 5 itself.
  - 4 is composite because it is a product ( $2 \times 2$ ) in which both numbers are smaller than 4.
- The first 25 prime numbers (all the prime numbers less than 100) are: 2, 3, 5, 7, 11, 13, 17, 19, 23, 29, 31, 37, 41, 43, 47, 53, 59, 61, 67, 71, 73, 79, 83, 89, 97
- Primes are used in several routines in information technology,
  - such as public-key cryptography, which relies on the difficulty of factoring large numbers into their prime factors.

[https://en.wikipedia.org/wiki/Prime\\_number](https://en.wikipedia.org/wiki/Prime_number)

[https://en.wikipedia.org/wiki/Prime\\_\(disambiguation\)](https://en.wikipedia.org/wiki/Prime_(disambiguation))

# PRIMES is in P

- In 2002, it was shown that the problem of determining if a number is prime is in P.
- AKS (Agrawal–Kayal–Saxena) primality test, which is a famous research of IIT Kanpur, and authors received the 2006 Gödel Prize and the 2006 Fulkerson Prize
- AKS primality test: “an unconditional deterministic polynomial-time algorithm that determines whether an input number is prime or composite”
- The key idea is to find the coefficient of  $x^i$  in  $((x + a)^n - (x^n + a))$ 
  - if all coefficients are multiple of  $n$ , then  $n$  is prime
  - else composite number
- **We work out it for  $a = -1$ .**
  - **What are the coefficient of  $x^i$  in  $((x - 1)^n - (x^n - 1))$ ?**

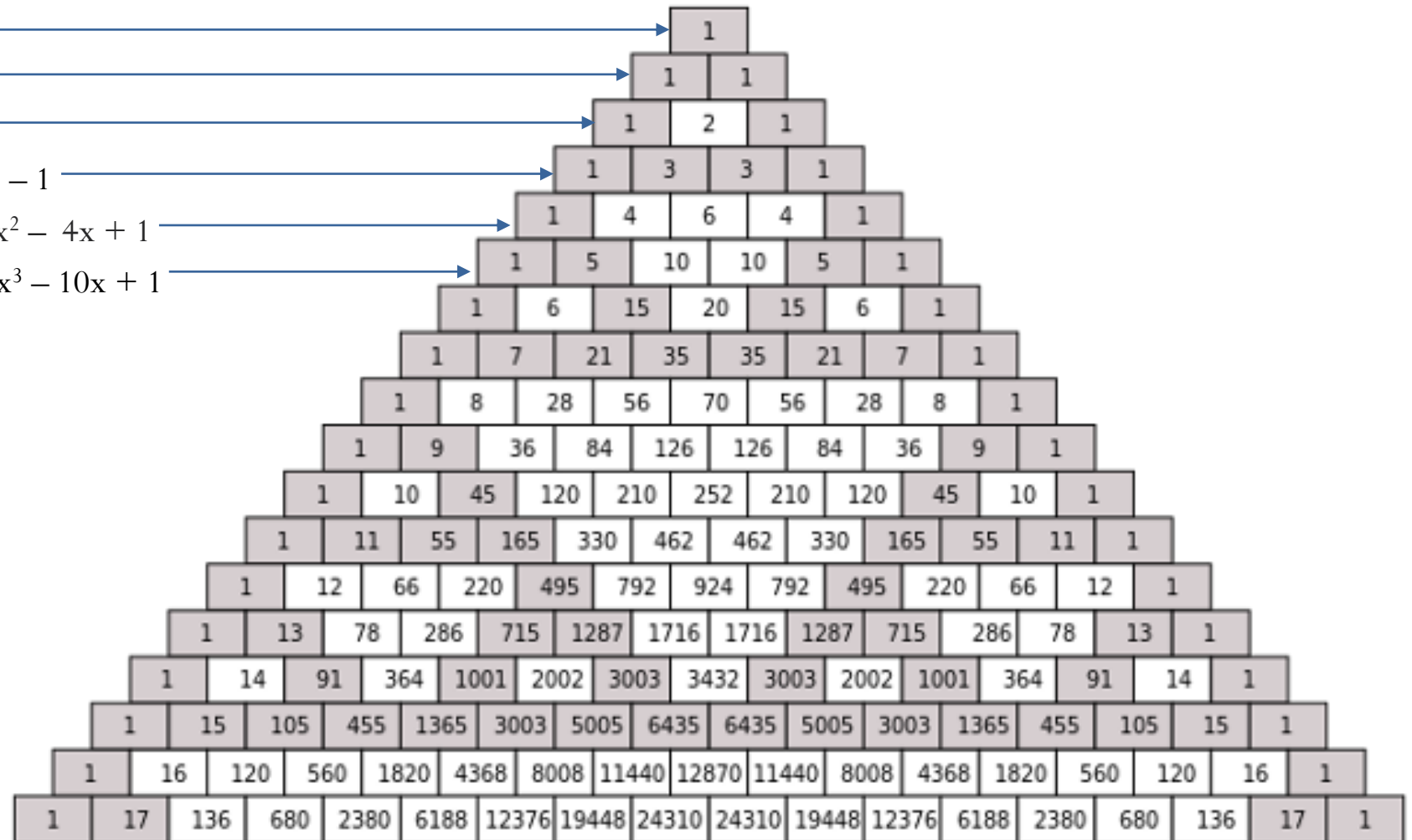
Agrawal, Manindra, Neeraj Kayal, and Nitin Saxena. "PRIMES is in P." *Annals of mathematics* (2004): 781-793.

[https://en.wikipedia.org/wiki/AKS\\_primality\\_test](https://en.wikipedia.org/wiki/AKS_primality_test)

# Pascal Triangle

- Coefficients of  $(x - 1)^n$

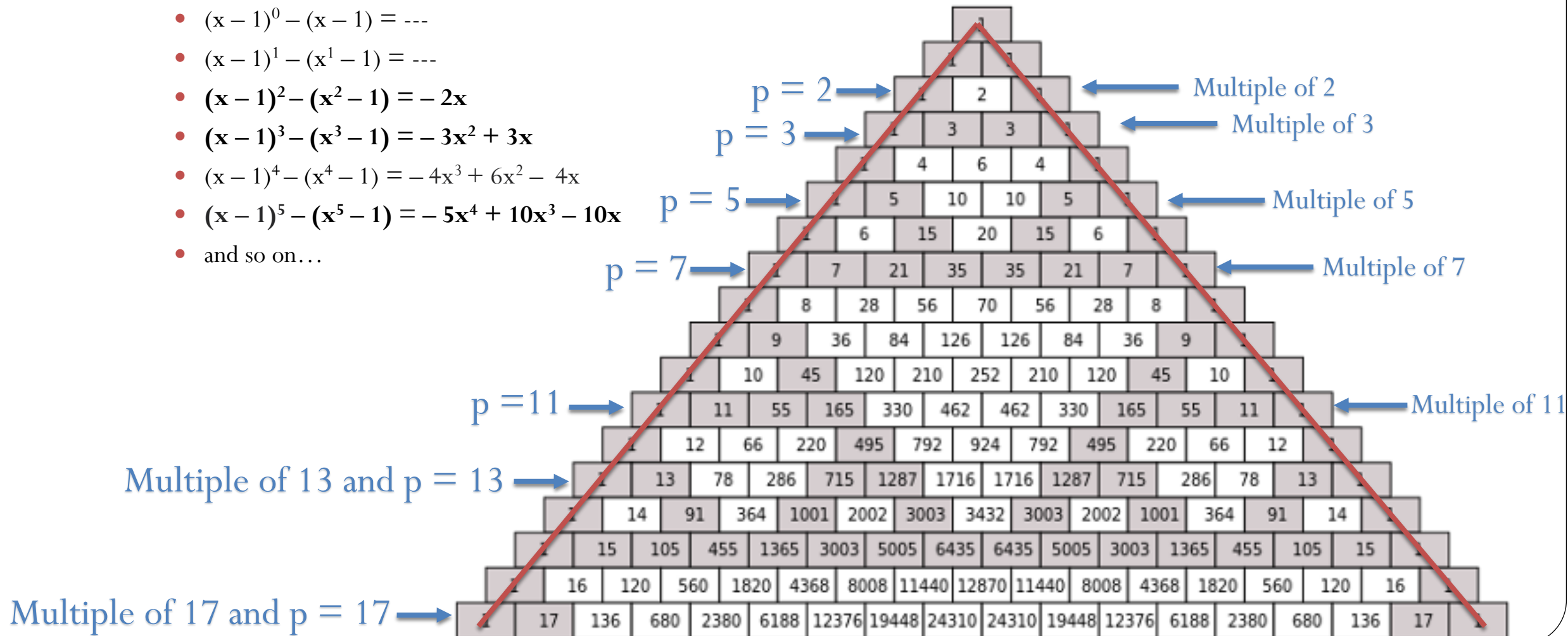
- $(x - 1)^0 = 1$
- $(x - 1)^1 = x - 1$
- $(x - 1)^2 = (x^2 - 2x + 1)$
- $(x - 1)^3 = x^3 - 3x^2 + 3x - 1$
- $(x - 1)^4 = x^4 - 4x^3 + 6x^2 - 4x + 1$
- $(x - 1)^5 = x^5 - 5x^4 + 10x^3 - 10x^2 + 5x - 1$
- and so on...



# Prime and Pascals Triangle

- Coefficients of  $(x - 1)^n - (x^n - 1)$

- $(x - 1)^0 - (x - 1) = \dots$
- $(x - 1)^1 - (x^1 - 1) = \dots$
- $(x - 1)^2 - (x^2 - 1) = -2x$
- $(x - 1)^3 - (x^3 - 1) = -3x^2 + 3x$
- $(x - 1)^4 - (x^4 - 1) = -4x^3 + 6x^2 - 4x$
- $(x - 1)^5 - (x^5 - 1) = -5x^4 + 10x^3 - 10x^2 + 5x$
- and so on...



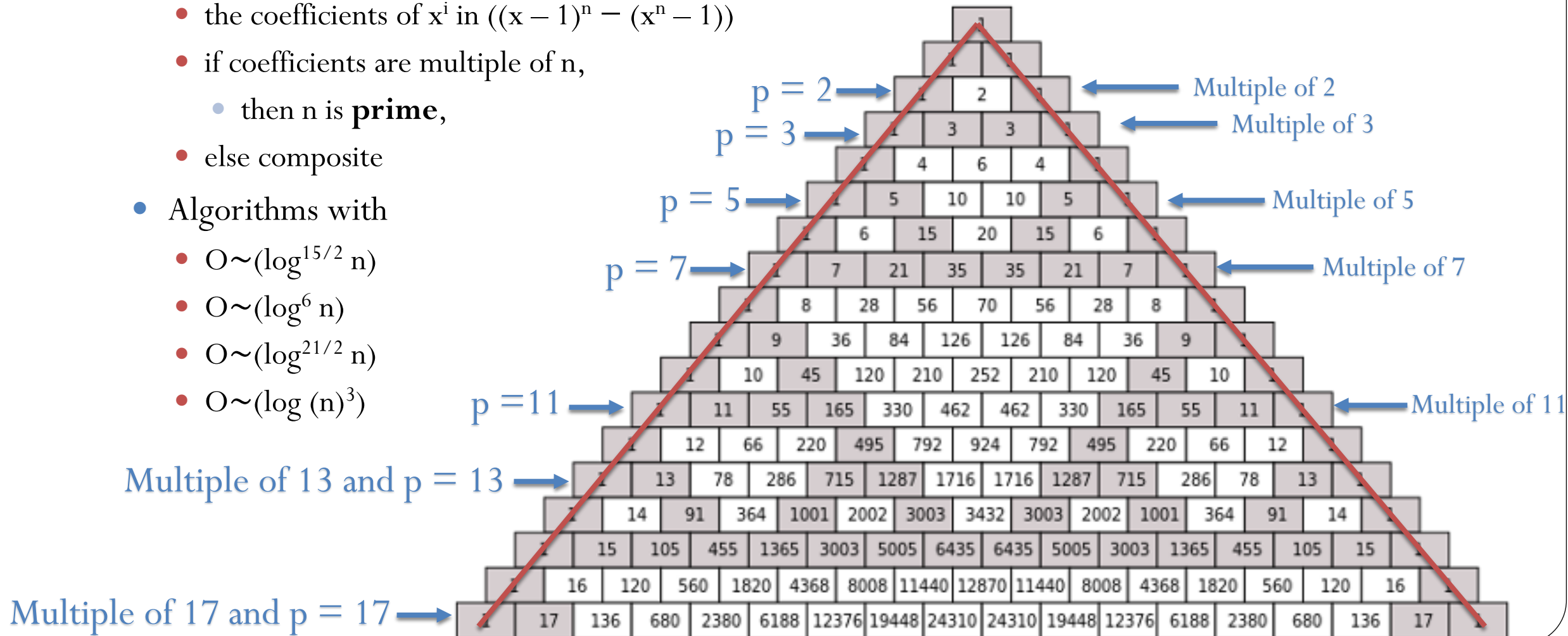
# Prime and Pascals Triangle

- It is possible to write a polynomial time algorithm to find

- the coefficients of  $x^i$  in  $((x - 1)^n - (x^n - 1))$
- if coefficients are multiple of  $n$ ,
  - then  $n$  is **prime**,
- else composite

- Algorithms with

- $O(\log^{15/2} n)$
- $O(\log^6 n)$
- $O(\log^{21/2} n)$
- $O(\log(n)^3)$



# More on Prime number

- Fast methods for primality test are available, such as Mersenne numbers.
- As of December 2018, the largest known prime number is a Mersenne prime with 24,862,048 decimal digits.

[https://en.wikipedia.org/wiki/Prime\\_number](https://en.wikipedia.org/wiki/Prime_number)

[https://en.wikipedia.org/wiki/Prime\\_\(disambiguation\)](https://en.wikipedia.org/wiki/Prime_(disambiguation))

# Millennium Problems



# Millennium Problems

- The Millennium Prize Problems are seven problems in mathematics that were stated by the Clay Mathematics Institute on May 24, 2000.
- One of 7 Millennium Problems for which Clay Math Institute awards \$1,000,000 i.e., US\$1 million prize
- One-million dollar (\*) question:  $P = NP$  ?
- almost all researchers think  $P \neq NP$

<https://www.claymath.org/millennium-problems>

<https://www.claymath.org/millennium-problems/millennium-prize-problems>

[https://en.wikipedia.org/wiki/Millennium\\_Prize\\_Problems](https://en.wikipedia.org/wiki/Millennium_Prize_Problems)

# Millennium Problems

- [Yang–Mills and Mass Gap](#)
- [Riemann Hypothesis](#)
- [P vs NP Problem](#): If it is easy to check that a solution to a problem is correct, is it also easy to solve the problem? This is the essence of the P vs NP question. Typical of the NP problems is that of the Hamiltonian Path Problem: given  $N$  cities to visit, how can one do this without visiting a city twice? If you give me a solution, I can easily check that it is correct. But I cannot so easily find a solution.
- [Navier–Stokes Equation](#)
- [Hodge Conjecture](#)
- [Poincaré Conjecture](#)
- [Birch and Swinnerton-Dyer Conjecture](#)

# Millennium Problems

- To date, the only Millennium Prize problem to have been solved is the Poincaré conjecture,
- A century passed between its formulation in 1904 by Henri Poincaré and its solution by Grigoriy Perelman, announced in preprints posted on ArXiv.org in 2002 and 2003.
- Grigoriy Perelman is the Russian mathematician Grigori Perelman.
- He declined the prize money.
- Perelman was selected to receive the Fields Medal for his solution, but he declined the award.

<https://www.claymath.org/millennium-problems/poincar%C3%A9-conjecture>

[https://en.wikipedia.org/wiki/Millennium\\_Prize\\_Problems](https://en.wikipedia.org/wiki/Millennium_Prize_Problems)

# Conclusions

# Conclusions

- This presentation is
  - more related to Computing Algorithms.
  - less related to Theory of Computation (Language, Turing Machine, etc.).
  - more focused toward modern Graph Analytics.
  - more focused toward applied mathematics i.e., Computer Science and Engineering.
  - less focused toward core Mathematics.

ขอบคุณ

Thai

Grazie  
Italian

תודה רבה  
Hebrew

Gracias

Spanish

Спасибо

Russian

English

*Thank You*

Obrigado

Portuguese

شكراً

Arabic

多謝

Traditional  
Chinese

<https://sites.google.com/site/animeshchaturvedi07>

Merci

French

Danke

German

धन्यवाद

Hindi

多谢

Simplified  
Chinese

நன்றி

Tamil

Tamil

ありがとうございました

Japanese

감사합니다

Korean