CSE 332 Summer 2024 Lecture 23: P & NP

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Running Times



Input Size

Running times we've seen:

- Θ(1)
- $\Theta(\log n)$
- $\Theta(n)$
- $\Theta(n \log n)$
- $\Theta(n^2)$
- $\Theta(2^n)$

Running Times

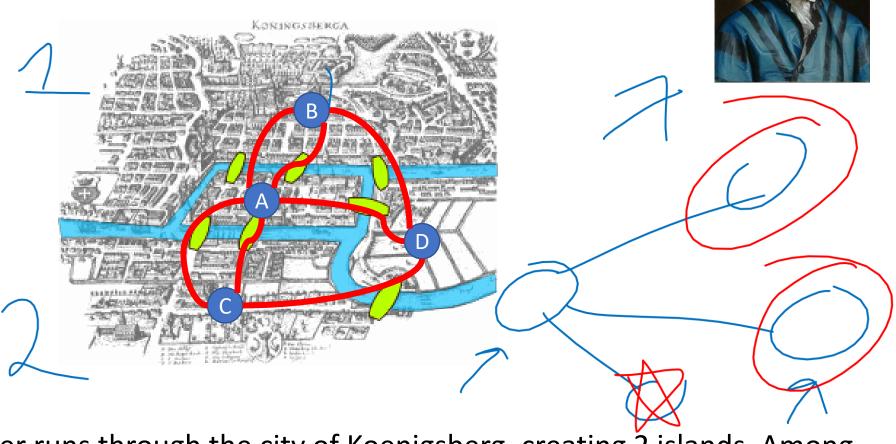
Table 2.1 The running times (rounded up) of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second. In cases where the running time exceeds 10^{25} years, we simply record the algorithm as taking a very long time.

	п	$n \log_2 n$	n^2	n^3	1.5^{n}	2^n	n!
	_		_		_		
n = 10	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
n = 30	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	10 ²⁵ years
n = 50	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
n = 100	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	10^{17} years	very long
n = 1,000	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
n = 10,000	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
n = 100,000	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
n = 1,000,000	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

Tractability

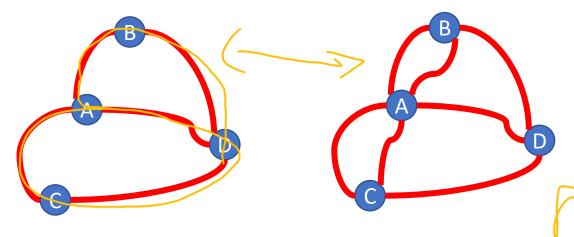
- Tractable:
 - Feasible to solve in the "real world"
- ✓ Intractable?
 - Infeasible to solve in the "real world"
- Whether a problem is considered "tractable" or "intractable" depends on the use case
 - For machine learning, big data, etc. tractable might mean O(n) or even $O(\log n)$
 - For most applications it's more like $\mathcal{O}(n^3)$ or $\mathcal{O}(n^2)$
- A strange pattern:
 - Most "natural" problems are either done in small-degree polynomial (e.g. n^2) or else exponential time (e.g. 2^n)/
 - It's rare to have problems which require a running time of n^5 , for example

7 Bridges of Königsberg



The Pregel River runs through the city of Koenigsberg, creating 2 islands. Among these 2 islands and the 2 sides of the river, there are 7 bridges. Is there any path starting at one landmass which crosses each bridge exactly once?

Euler Path Problem



• Path:

• A sequence of nodes $v_1, v_2, ...$ such that for every consecutive pair are connected by an edge (i.e. (v_i, v_{i+1}) is an edge for each i in the path)

• Euler Path:

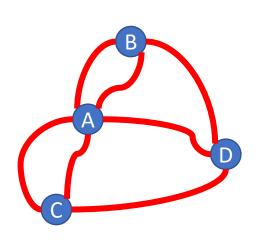
- A path such that every edge in the graph appears exactly once
 - If the graph is not simple then some pairs need to appear multiple times!

• Euler path problem:

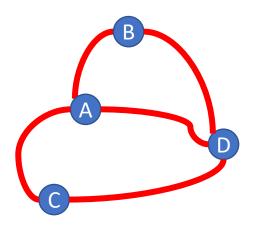
• Given an undirected graph G = (V, E), does there exist an Euler path for G?

Examples

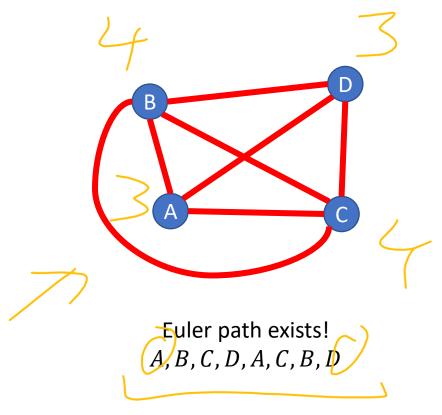
Which of the graphs below have an Euler path?



No Euler path exists!

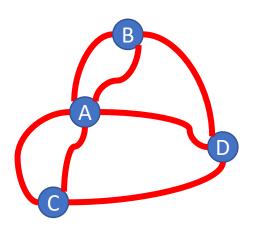


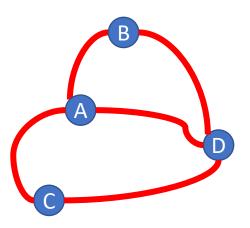
Euler path exists! A, B, D, A, C, D

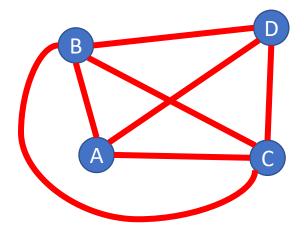


Euler's Theorem

 A graph has an Euler Path if and only if it is connected and has exactly 0 or 2 nodes with odd degree.







Algorithm for the Euler Path Problem

• Given an undirected graph G = (V, E), does there exist an Euler path for G?

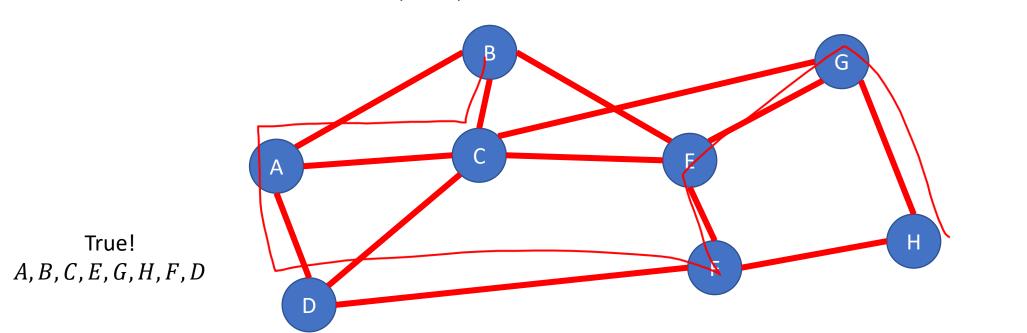
• Algorithm:

- Check if the graph is connected
- Check the degree of each node
- If the number of nodes with odd degree is 0 or 2, return true
- Otherwise return false
- Running time?



A Seemingly Similar Problem

- Hamiltonian Path:
 - A path that includes every node in the graph exactly once
- Hamiltonian Path Problem:
 - Given a graph G = (V, E), does that graph have a Hamiltonian Path?



Algorithms for the Hamiltonian Path Problem

• Option 1:

- Explore all possible simple paths through the graph
- Check to see if any of those are length V

Option 2:

- Write down every sequence of nodes
- Check to see if any of those are a path



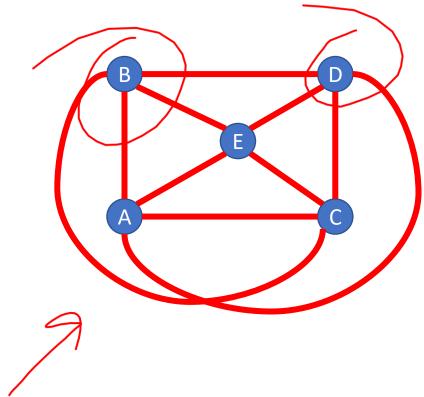


Option 2: List all sequences, look for a path

- Running time:
 - G = (V, E)
 - Number of permutations of V is |V|!
 - $n! \equiv n \cdot (n-1) \cdot (n-2) \cdot \dots \cdot 2 \cdot 1$ How does n! compare with 2^n ?
 - $n! \in \Omega(2^n)$
 - Exponential running time!

Option 1: Explore all simple paths, check for one of length V

- Running time:
 - G = (V, E)
 - Number of paths
 - Pick a first node (|V| choices)
 - Pick a neighbor (up to |V| 1 choices)
 - Pick a neighbor (up to |V| 2 choices)
 - Repeat |V| 1 total times
 - Overall: |V|! paths
 - Exponential running time



Complexity Classes

- A Complexity Class is a set of problems (e.g. sorting, Euler path, Hamiltonian path)
 - The problems included in a complexity class are those whose most efficient algorithm has a specific upper bound on its running time (or memory use, or...)

Examples:

- The set of all problems that can be solved by an algorithm with running time O(n)
 - Contains: Finding the minimum of a list, finding the maximum of a list, buildheap, summing a list, etc.
- The set of all problems that can be solved by an algorithm with running time $O(n^2)$
 - Contains: everything above as well as sorting, Euler path
- The set of all problems that can be solved by an algorithm with running time O(n!)
 - Contains: everything we've seen in this class so far



Complexity Classes and Tractability

- To explore what problems are and are not tractable, we give some complexity classes special names:
- Complexity Class *P*:
 - Stands for "Polynomial"
 - The set of problems which have an algorithm whose running time is $ho(n^p)$ for some choice of $p \in \mathbb{R}$.
 - We say all problems belonging to P are "Tractable"
- Complexity Class EXP:
 Stands for "Exponential"

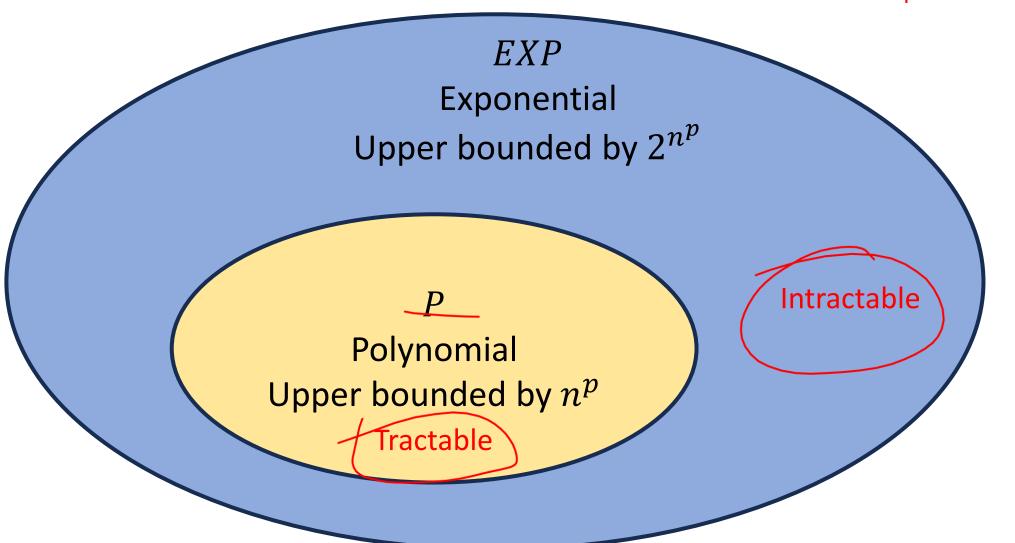
 - The set of problems which have an algorithm whose running time is $O(2^{n^{\nu}})$ for some choice of $p \in \mathbb{R}$
 - We say all problems belonging to EXP P are "Intractable"
 - Disclaimer: Really it's all problems outside of P, and there are problems which do not belong to EXP, but we're not going to worry about those in this class

Important!

 $P \subset EXP$

EXP and P

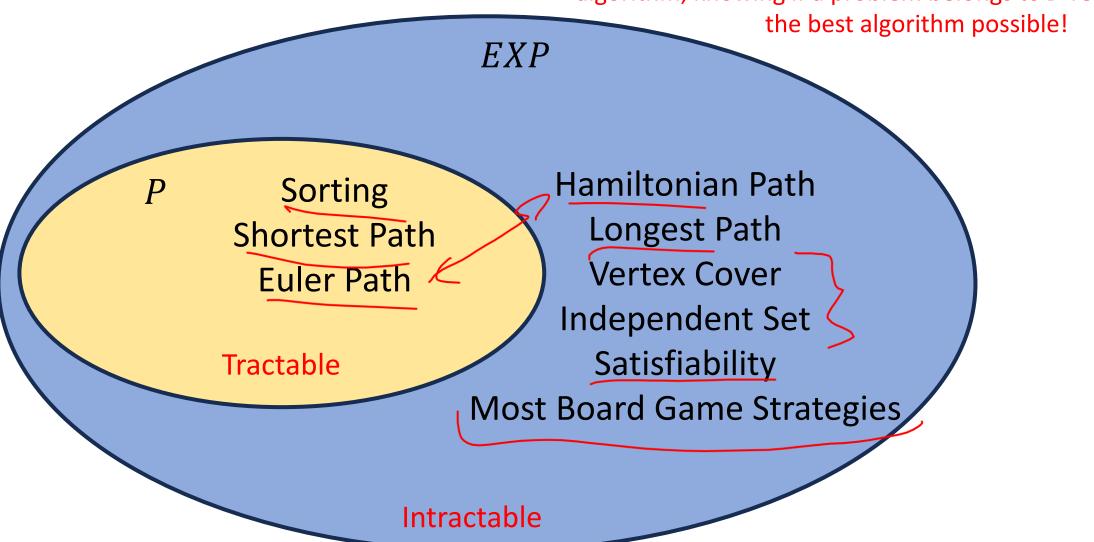
Every problem within P is also within EXP The intractable ones are the problems within EXP but NOT P



Important!

Members

Some of the problems listed in EXP could also be members of P Since membership is determined by a problem's most efficient algorithm, knowing if a problem belongs to P requires knowing the best algorithm possible!



Studying Complexity and Tractability

- Organizing problems into complexity classes helps us to reason more carefully and flexibly about tractability
- The goal for each problem is to either
 - Find an efficient algorithm if it exists
 - i.e. show it belongs to *P*
 - Prove that no efficient algorithm exists
 - i.e. show it does not belong to P
- Complexity classes allow us to reason about sets of problems at a time, rather than each problem individually
 - If we can find more precise classes to organize problems into, we might be able to draw conclusions about the entire class
 - It may be easier to show a problem belongs to class C than to P, so it may help to show that $C \subseteq P$

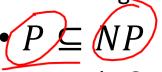
Some problems in *EXP* seem "easier"

- There are some problems that we do not have polynomial time algorithms to solve, but provided answers are easy to check
- Hamiltonian Path:
 - It's "hard" to look at a graph and determine whether it has a Hamiltonian Path
 - It's "easy" to look at a graph and a candidate path together and determine whether THAT path is a Hamiltonian Path
 - It's easy to verify whether a given path is a Hamiltonian path

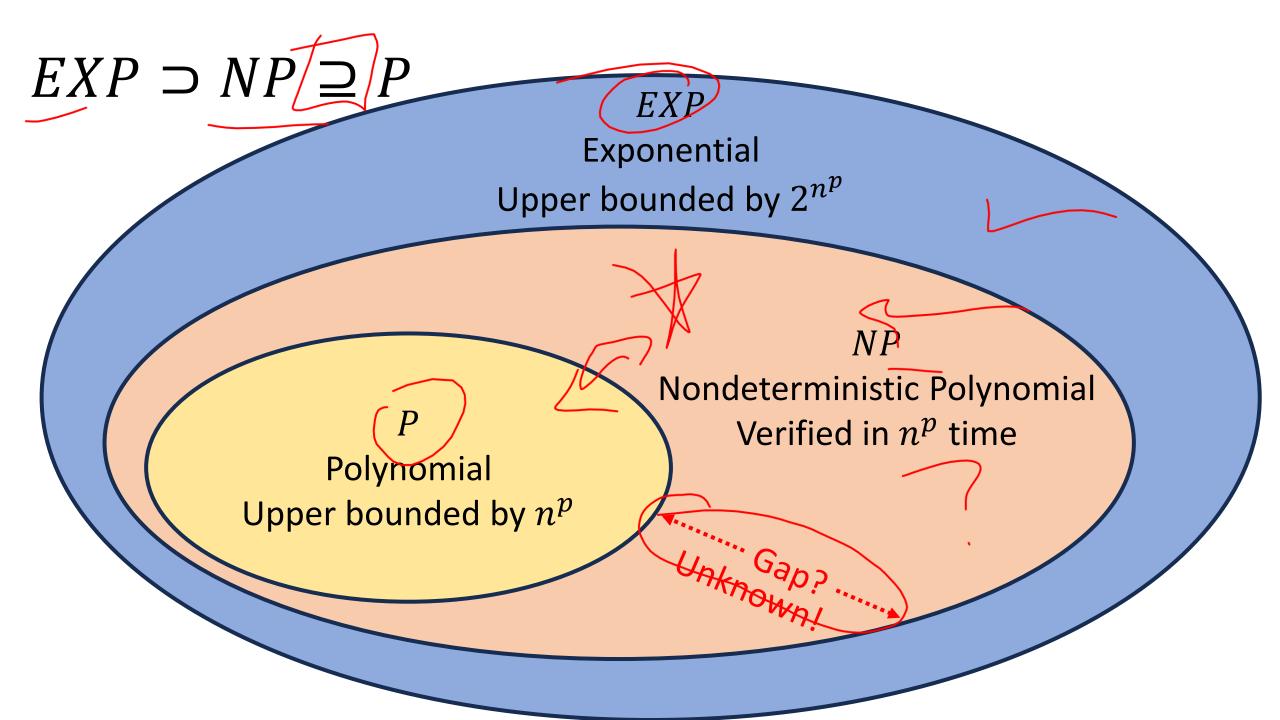
Class NP

• *NP*

- The set of problems for which a candidate solution can be verified in polynomial time
- Stands for "Non-deterministic Polynomial"
 - Corresponds to algorithms that can guess a solution (if it exists), that solution is then verified to be correct in polynomial time
 - Can also think of as allowing a special operation that allows the algorithm to magically guess the right choice at each step of an exhaustive search



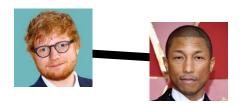
• Why?



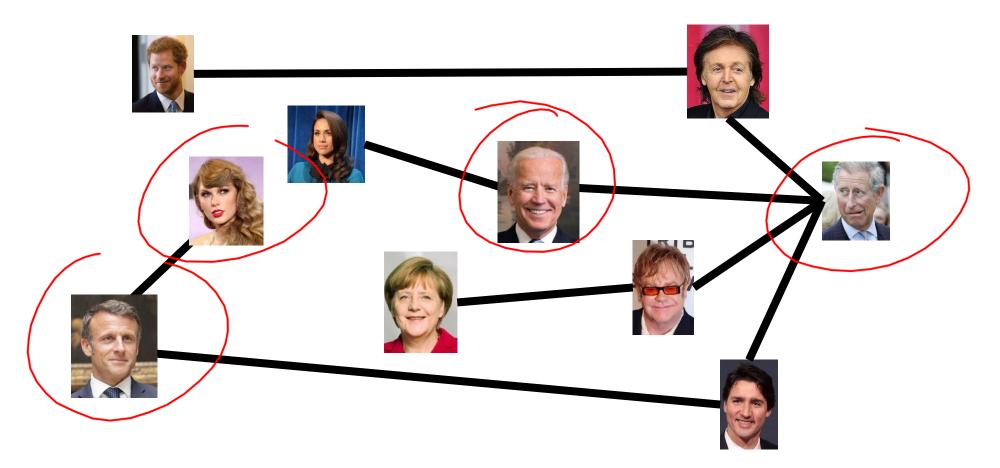
Solving and Verifying Hamiltonian Path

- Give an algorithm to solve Hamiltonian Path
 - Input: G = (V, E)
 - Output: True if G has a Hamiltonian Path
 - Algorithm: Check whether each permutation of V is a path.
 - Running time: |V|!, so does not show whether it belongs to P
- Give an algorithm to verify Hamiltonian Path
 - Input: G = (V, E) and a sequence of nodes
 - Output: True if that sequence of nodes is a Hamiltonian Path
 - Algorithm:
 - Check that each node appears in the sequence exactly once
 - Check that the sequence is a path
 - Running time: $Q(V \cdot E)$ so it belongs to NP

Party Problem



Draw Edges between people who don't get along How many people can I invite to a party if everyone must get along?



Independent Set

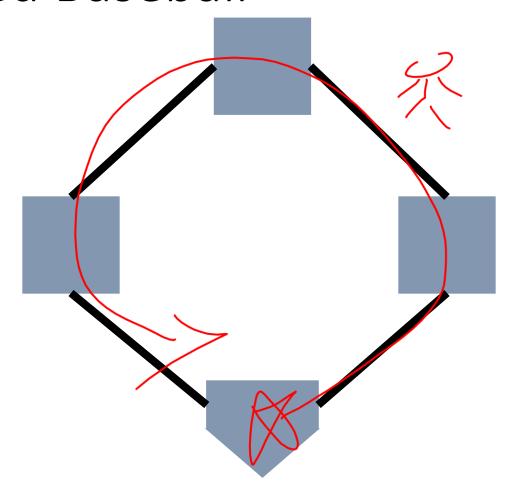
- Independent set:
 - $S \subseteq V$ is an independent set if no two nodes in S share an edge
- Independent Set Problem:
 - Given a graph G=(V,E) and a number k, determine whether there is an independent set S of size k

Example Independent set of size 6

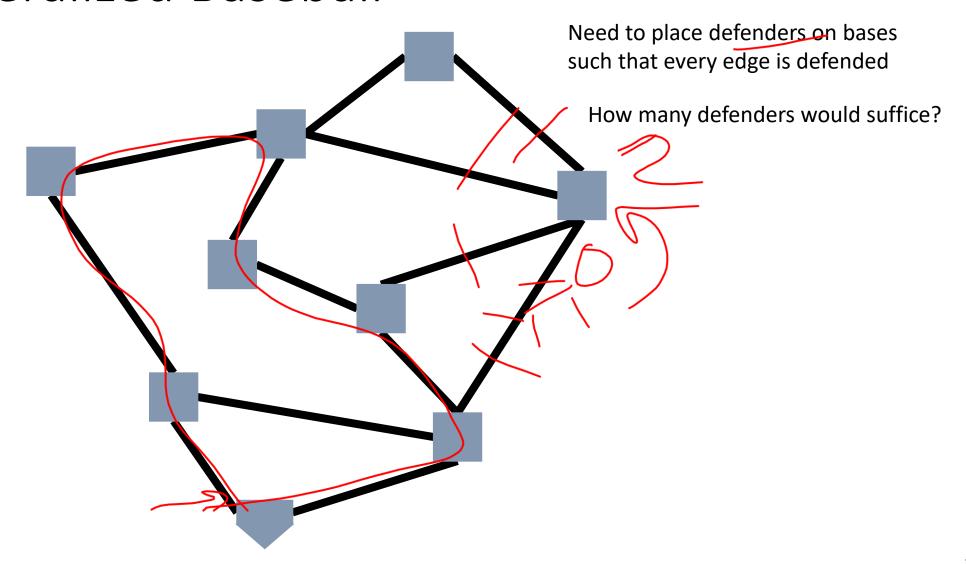
Solving and Verifying Independent Set

- Give an algorithm to solve independent set
 - Input: G = (V, E) and a number k
 - Output: True if G has an independent set of size k
- Give an algorithm to verify independent set
 - Input: G = (V, E), a number k, and a set $S \subseteq V$
 - Output: True if S is an independent set of size k

Generalized Baseball



Generalized Baseball



Vertex Cover

- Vertex Cover:
 - $C \subseteq V$ is a vertex cover if every edge in E has one of its endpoints in C
- Vertex Cover Problem:
 - Given a graph G = (V, E) and a number k, determine if there is a vertex cover C of size k

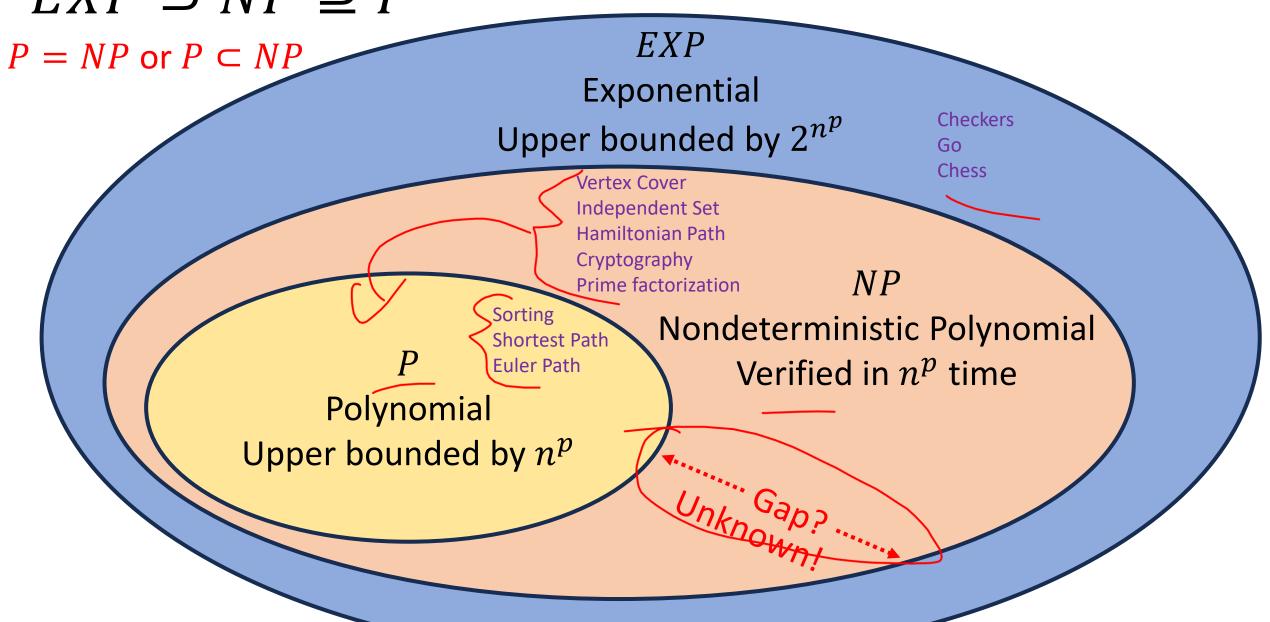
Example Vertex cover of size 5

Solving and Verifying Vertex Cover

- Give an algorithm to solve vertex cover
 - Input: G = (V, E) and a number k
 - Output: True if G has a vertex cover of size k
- Give an algorithm to verify vertex cover
 - Input: G = (V, E), a number k, and a set $S \subseteq E$
 - Output: True if S is a vertex cover of size k

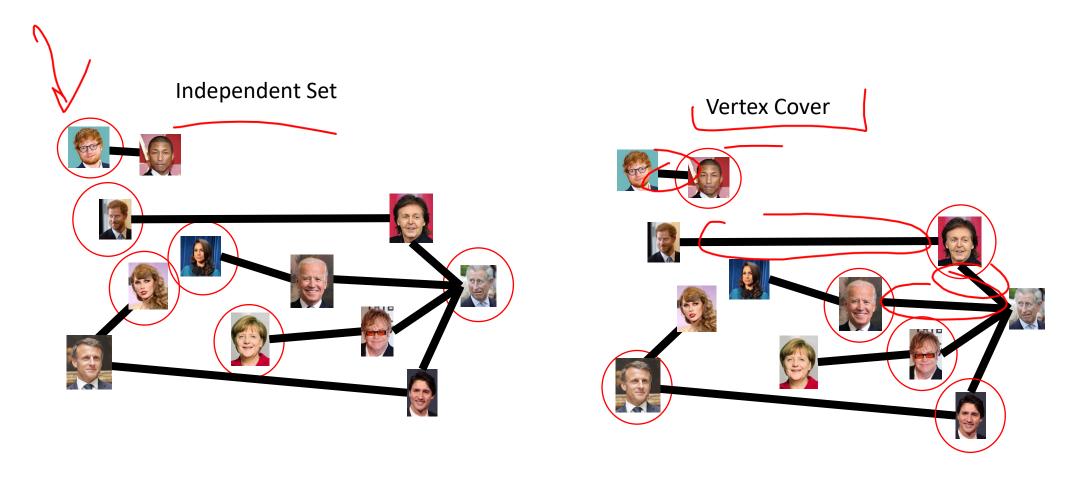


 $EXP \supset NP \supseteq P$



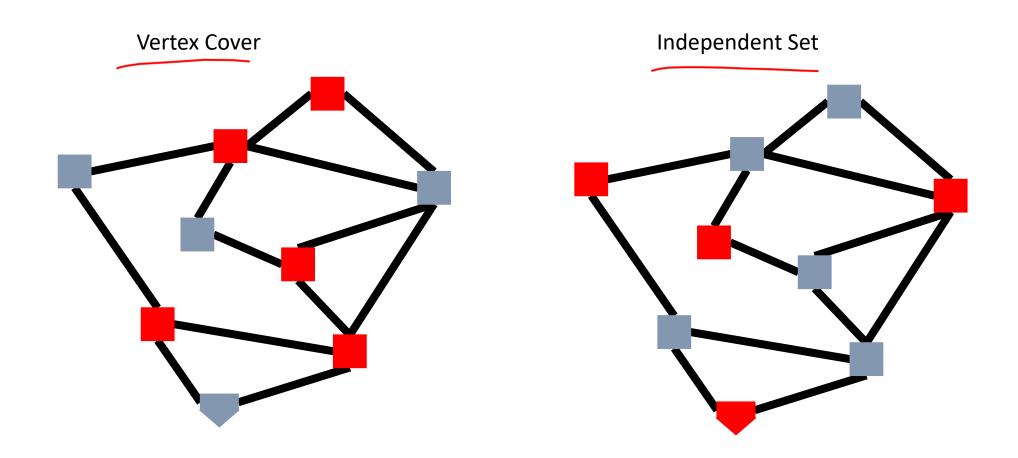
Way Cool!

S is an independent set of G iff V-S is a vertex cover of G



Way Cool!

S is an independent set of G iff V - S is a vertex cover of G



Solving Vertex Cover and Independent Set

- Algorithm to solve vertex cover
 - Input, G = (V, E) and a number k
 - Output: True if G has a vertex cover of size k
 - Check if there is an Independent Set of G of size |V|-k
- Algorithm to solve independent set
 - Input: G = (V, E) and a number k
 - Output: True if G has an independent set of size k
 - Check if there is a Vertex Cover of G of size $|V|-k_I$

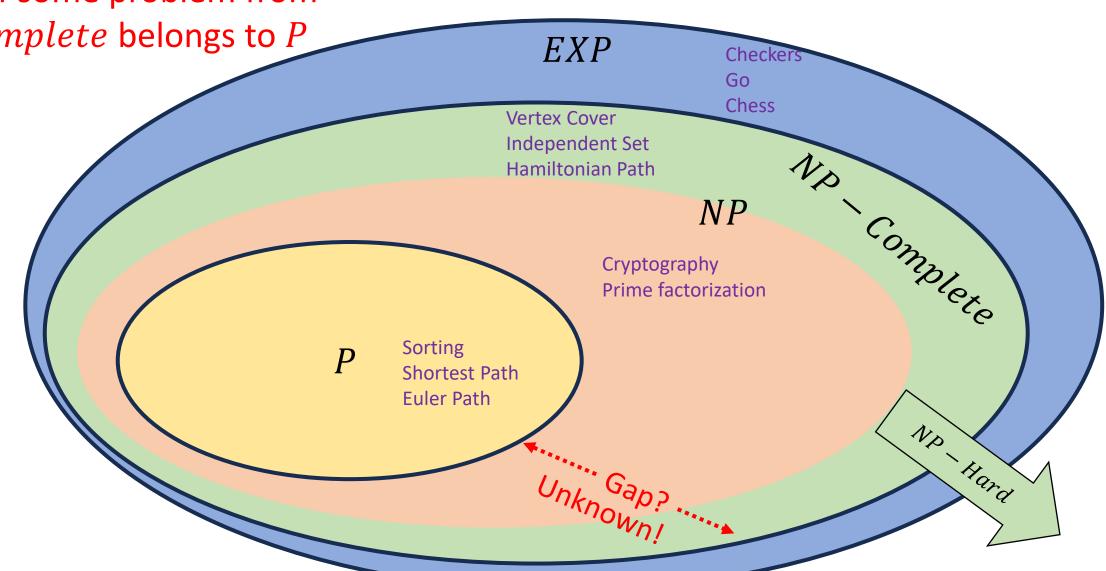
Either both problems belong to *P*, or else neither does!

NP-Complete

- A set of "together they stand, together they fall" problems
- The problems in this set either all belong to P, or none of them do
- Intuitively, the "hardest" problems in NP
- Collection of problems from NP that can all be "transformed" into each other in polynomial time
 - Like we could transform independent set to vertex cover, and vice-versa
 - We can also transform vertex cover into Hamiltonian path, and Hamiltonian path into independent set, and ...

$EXP \supset NP - Complete \supseteq NP \supseteq P$

P = NP iff some problem from NP - Complete belongs to P



Overview

- Problems not belonging to P are considered intractable
- The problems within *NP* have some properties that make them seem like they might be tractable, but we've been unsuccessful with finding polynomial time algorithms for many
- The class NP-Complete contains problems with the properties:
 - All members are also members of NP
 - All members of NP can be transformed into every member of NP Complete
 - Therefore if any one member of NP-Complete belongs to P, then P=NP

Why should YOU care?

- If you can find a polynomial time algorithm for any NP Complete problem then:
 - You will win \$1million
 - You will win a Turing Award
 - You will be world famous
 - You will have done something that no one else on Earth has been able to do in spite of the above!
- If you are told to write an algorithm a problem that is NP-Complete
 - You can tell that person everything above to set expectations
 - Change the requirements!
 - **Approximate the solution**: Instead of finding a path that visits every node, find a path that visits at least 75% of the nodes
 - Add Assumptions: problem might be tractable if we can assume the graph is acyclic, a tree
 - Use Heuristics: Write an algorithm that's "good enough" for small inputs, ignore edge cases