

CS4102 Algorithms

Spring 2020

Today's Keywords

- Reductions
- Bipartite Matching
- Vertex Cover
- Independent Set
- Decision problems, verification problems
- NP, NP-Hard, NP-Complete

CLRS Readings

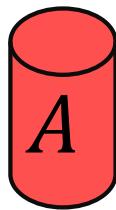
- Chapter 34

Reductions

- Algorithm technique of supreme ultimate power
- Convert instance of problem A to an instance of Problem B
- Convert solution of problem B back to a solution of problem A
- Why? (You might be asking. ☺)
 - As you've seen, might be a useful way to develop solution to A
 - Also, lower-bounds proofs
 - We can't find polynomial solutions to some problems.
We want to know if they are really exponential!

MacGyver's Reduction

Problem we don't know how to solve

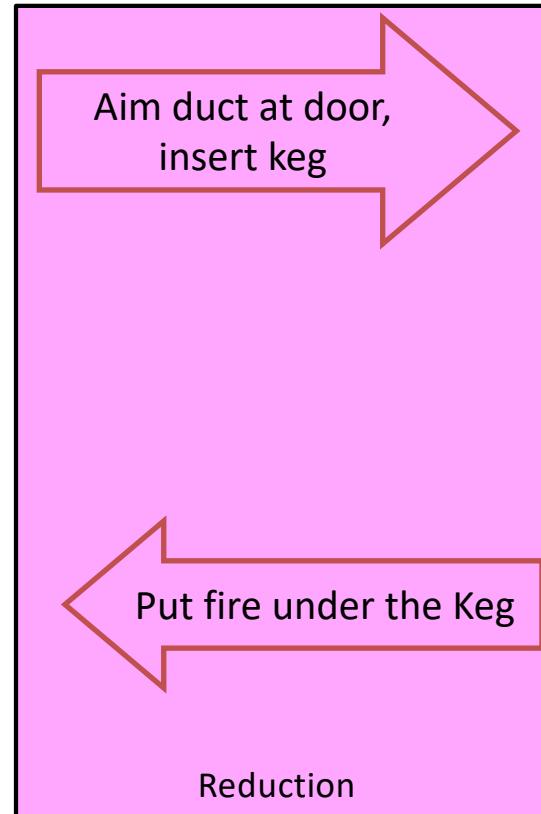


Opening a door



Solution for **A**

Keg cannon
battering ram



Problem we do know how to solve



Lighting a fire



How?

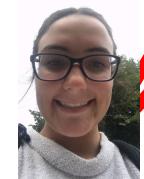
Solution for **B**

Alcohol, wood,
matches



Maximum Bipartite Matching

Dog Lovers



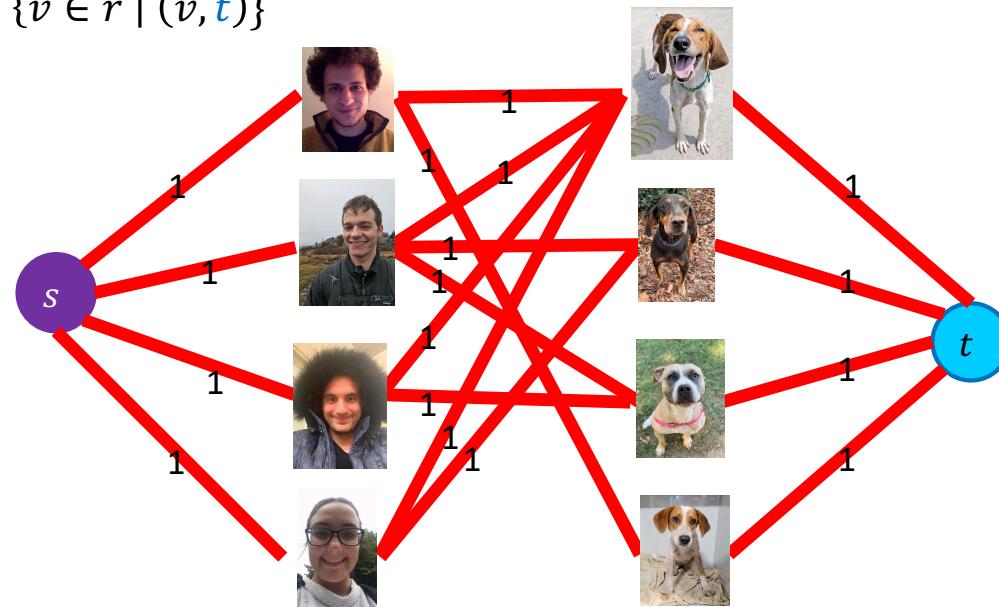
Dogs



Maximum Bipartite Matching Using Max Flow

Make $G = (L, R, E)$ a flow network $G' = (V', E')$ by:

- Adding in a **source** and **sink** to the set of nodes:
 - $V' = L \cup R \cup \{s, t\}$
- Adding an edge from **source** to L and from R to **sink**:
 - $E' = E \cup \{u \in L \mid (s, u)\} \cup \{v \in R \mid (v, t)\}$
- Make each edge capacity 1:
 - $\forall e \in E', c(e) = 1$



Remember: need to show

1. How to map instance of MBM to MF (and back) - construction
2. A valid solution to MF instance is a valid solution to MBM instance

Bipartite Matching Reduction

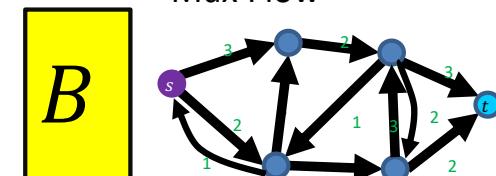
Problem we don't know how to solve

Bipartite Matching



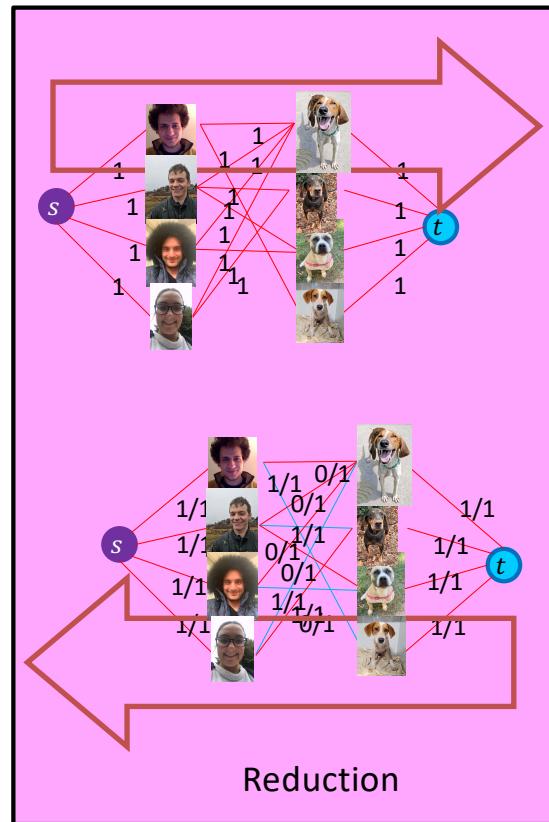
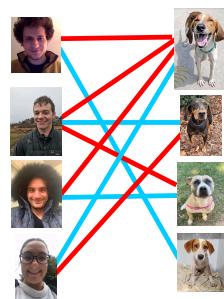
Problem we do know how to solve

Max Flow

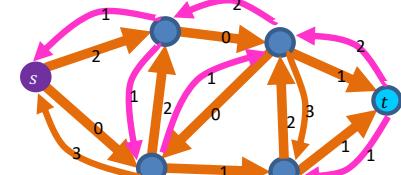


Ford Fulkerson

Solution for **A**



Solution for **B**



In General: Reduction

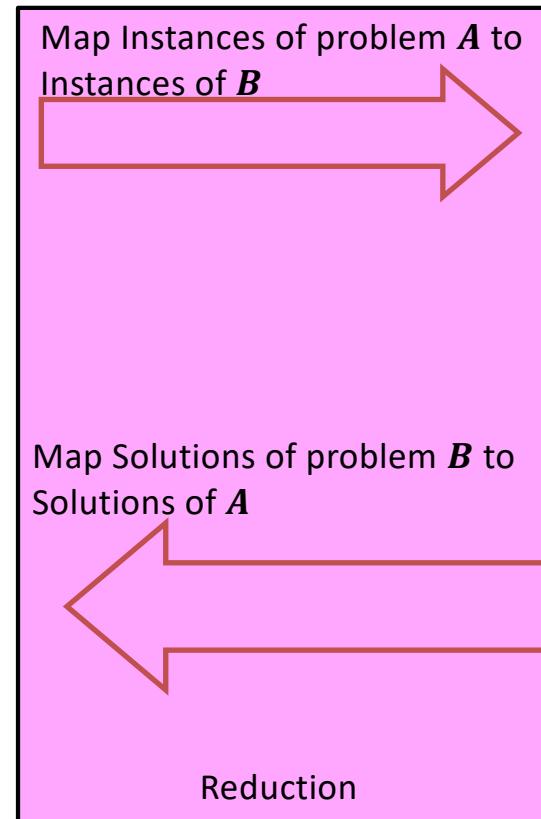
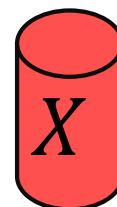
Problem we don't know how to solve



Remember: need to show

1. How to map instance of A to B (and back)
2. Why solution to B was a valid solution to A

Solution for A



Problem we do know how to solve

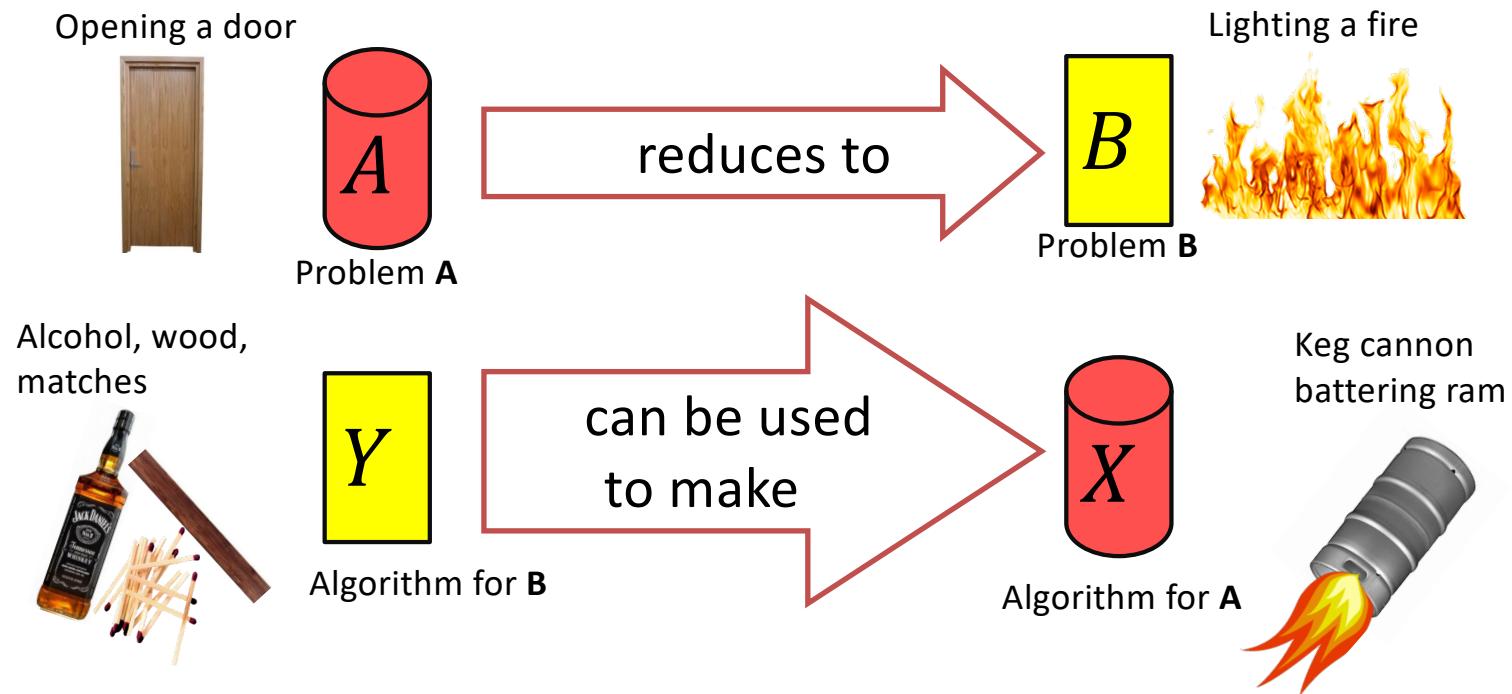


Using any Algorithm for B



Solution for B

Worst-case lower-bound Proofs



A is not a harder problem than B

$$A \leq B$$

The name “reduces” is confusing: it is in the *opposite* direction of the making

Bipartite Matching Reduction

Problem we don't know how to solve

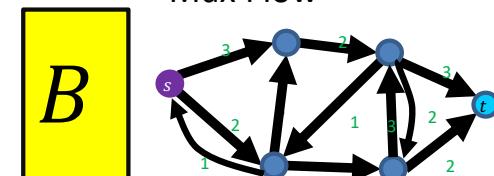
Bipartite Matching



Then this is fast

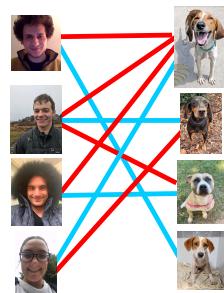
Problem we do know how to solve

Max Flow

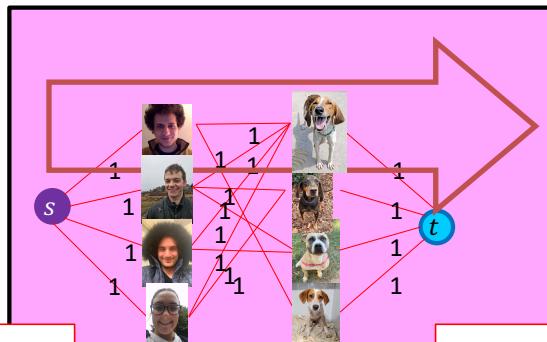


Ford Fulkerson

Solution for **A**

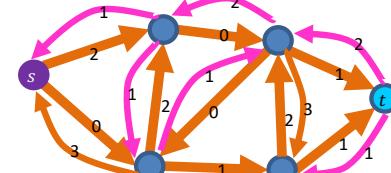


If this is fast



Reduction

Solution for **B**

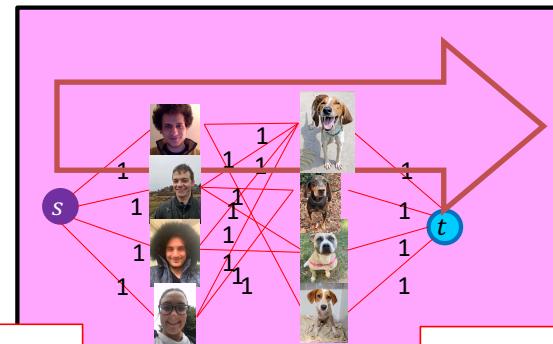


Bipartite Matching Reduction

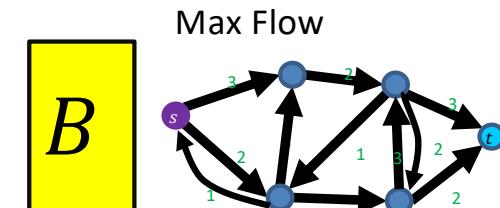
Problem we don't know how to solve



If this is slow

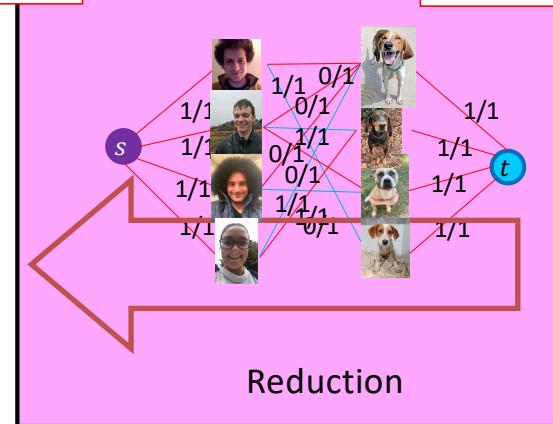
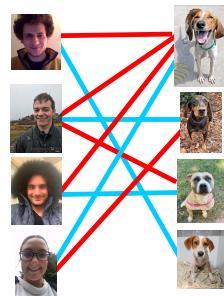


Problem we do know how to solve

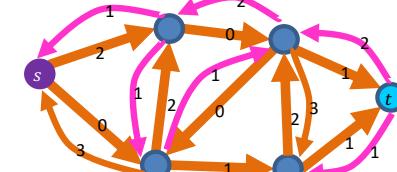


Ford Fulkerson

Solution for **A**

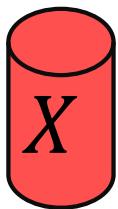


Solution for **B**



Proof of Lower Bound by Reduction

To Show: Y is slow



1. We know X is slow (by a proof)
(e.g., X = some way to open the door)



2. Assume Y is quick [toward contradiction]
(Y = some way to light a fire)



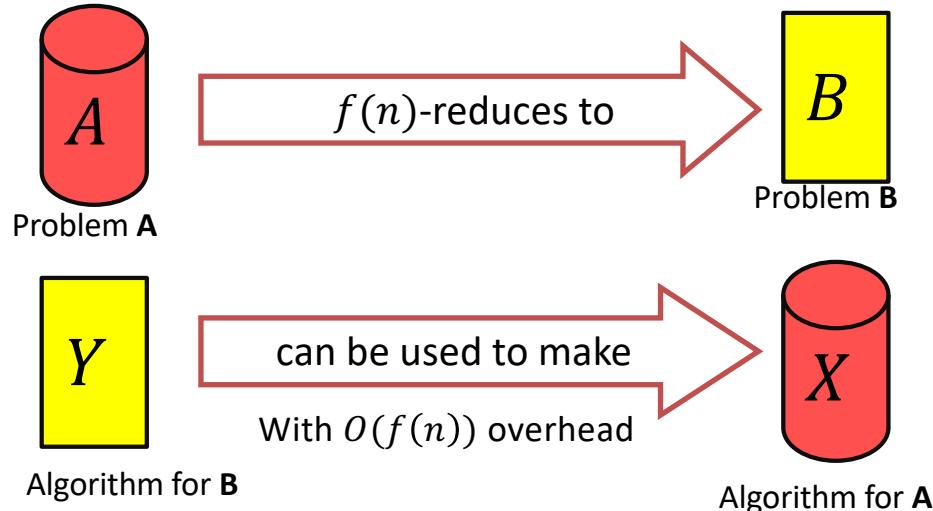
3. Show how to use Y to perform X quickly

4. X is slow, but Y could be used to perform X quickly
conclusion: Y must not actually be quick

Same Again, Different Explanation

- Say we know these two things about problems A and B:
 - First, $A \leq B$
 - Second, we've proven solving A is "slow" (using some lower-bounds proof)
- What can we say about B?
 - Solving B must be "slow". Why?
- Argument:
 - Assume solving B could be "fast"
 - We can solve A using B
 - That's a fast solution for A
 - But one of our givens: it's been proved A has no fast solutions. Contradiction!
 - Therefore assumption that B is "fast" is wrong. Solving B must be "slow".
 - Remember we said: A is no harder than B
- **Big point:** We can use known "slow" problems to show other problems are "slow"

Reduction Proof Notation



A is not a harder problem than B

$$A \leq B$$

If A requires time $\Omega(f(n))$ time then B also requires $\Omega(f(n))$ time

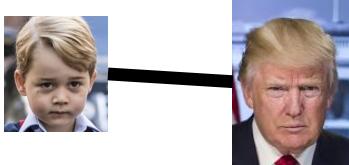
$$A \leq_{f(n)} B$$

Or we could have solved A faster using B's solver!

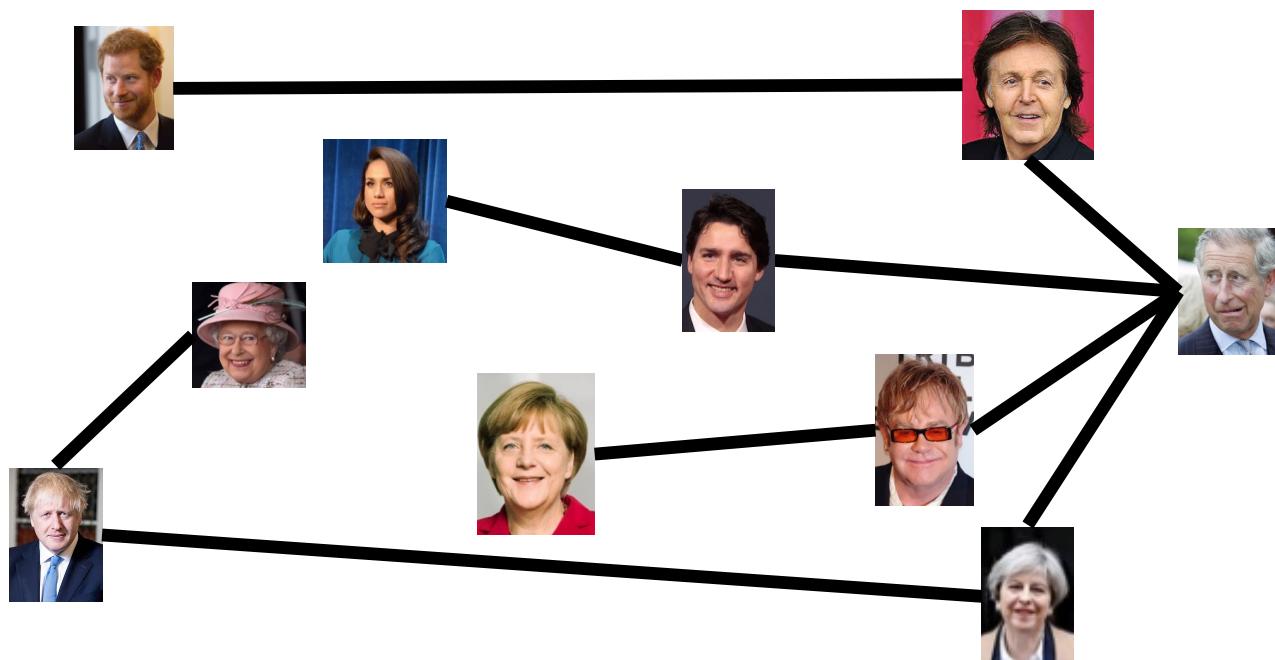
Peek Ahead to Where We're Going

- We're going to start looking at a set of *intractable* problems
 - No known polynomial solutions have been found
 - But none have proven to require exponential time either!
- We've found polynomial reductions between a group of these (called NP-C), and we'll see that
 - None of them are “harder” than any of the others.
 - If one has a polynomial solution, they all do.
 - If there's an exponential lower-bound proof for one, all are exponential.
 - And there's more to say about these ideas later!
- Important note about discussions that follow:
 - Not showing how to solve any of these problems directly.
 - Only showing how to reduce one problem to another!

Party Problem



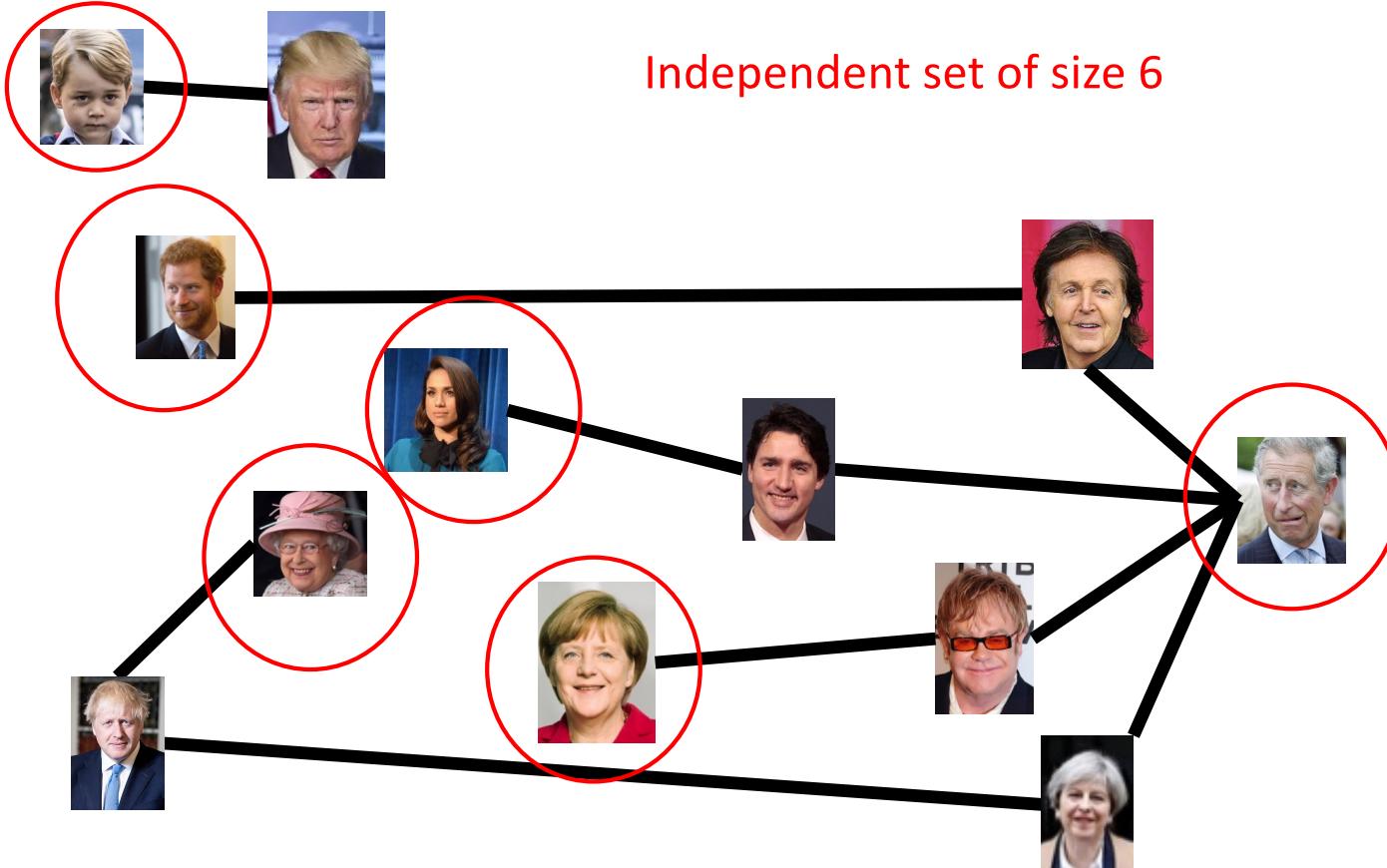
Draw Edges between people who don't get along
Find the maximum number of people who get along



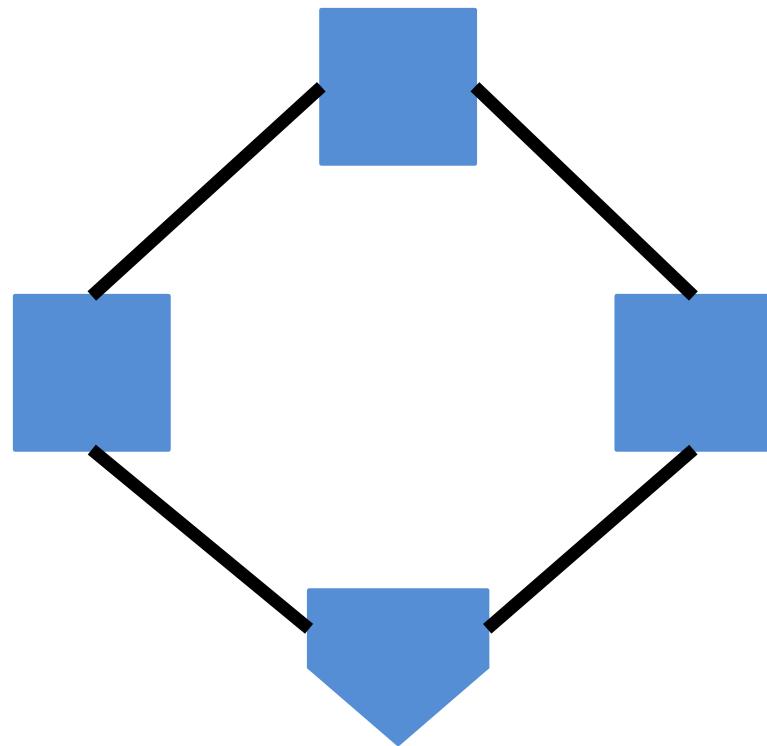
Maximum Independent Set

- Independent set: $S \subseteq V$ is an independent set if no two nodes in S share an edge
- Maximum Independent Set Problem: Given a graph $G = (V, E)$ find the maximum independent set S

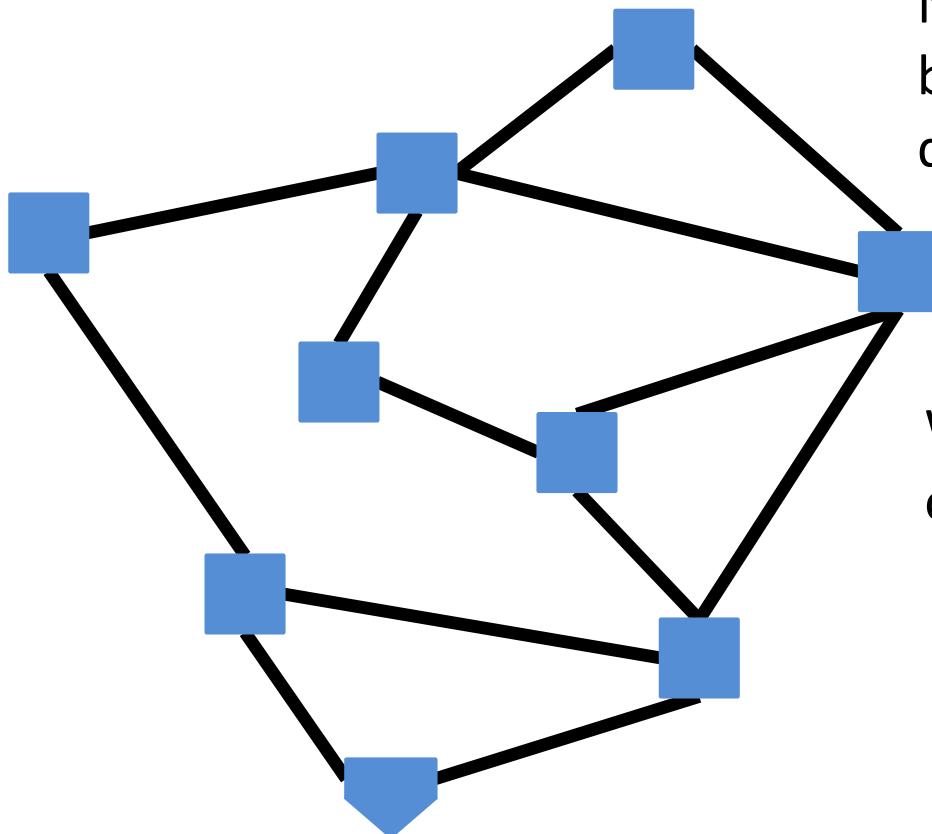
Example



Generalized Baseball



Generalized Baseball



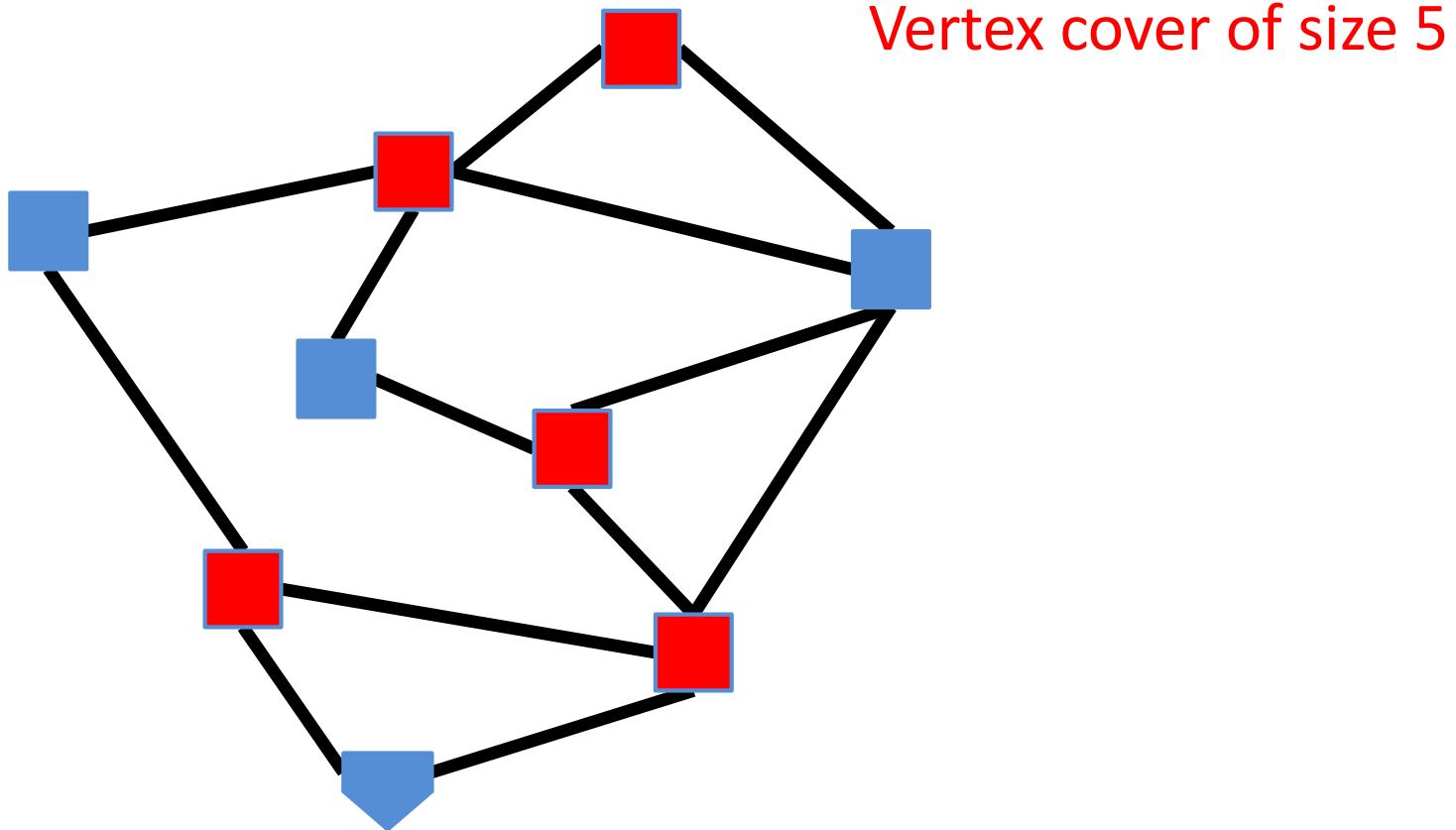
Need to place defenders on bases such that every edge is defended

What's the fewest number of defenders needed?

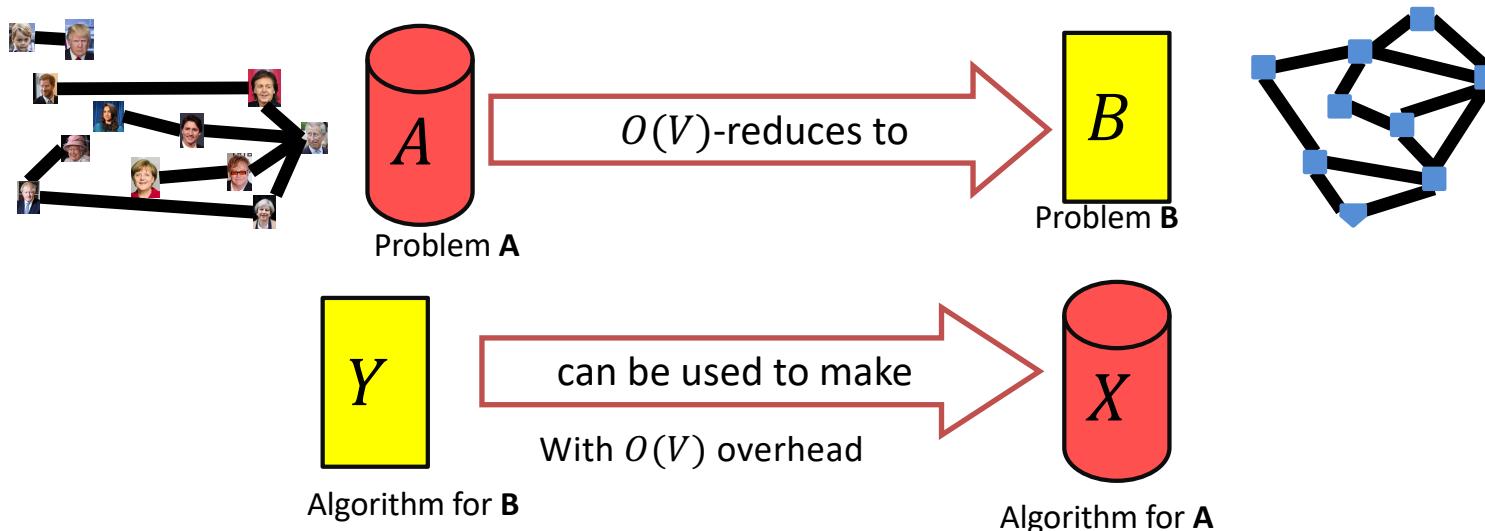
Minimum Vertex Cover

- Vertex Cover: $C \subseteq V$ is a vertex cover if every edge in E has one of its endpoints in C
- Minimum Vertex Cover: Given a graph $G = (V, E)$ find the minimum vertex cover C

Example



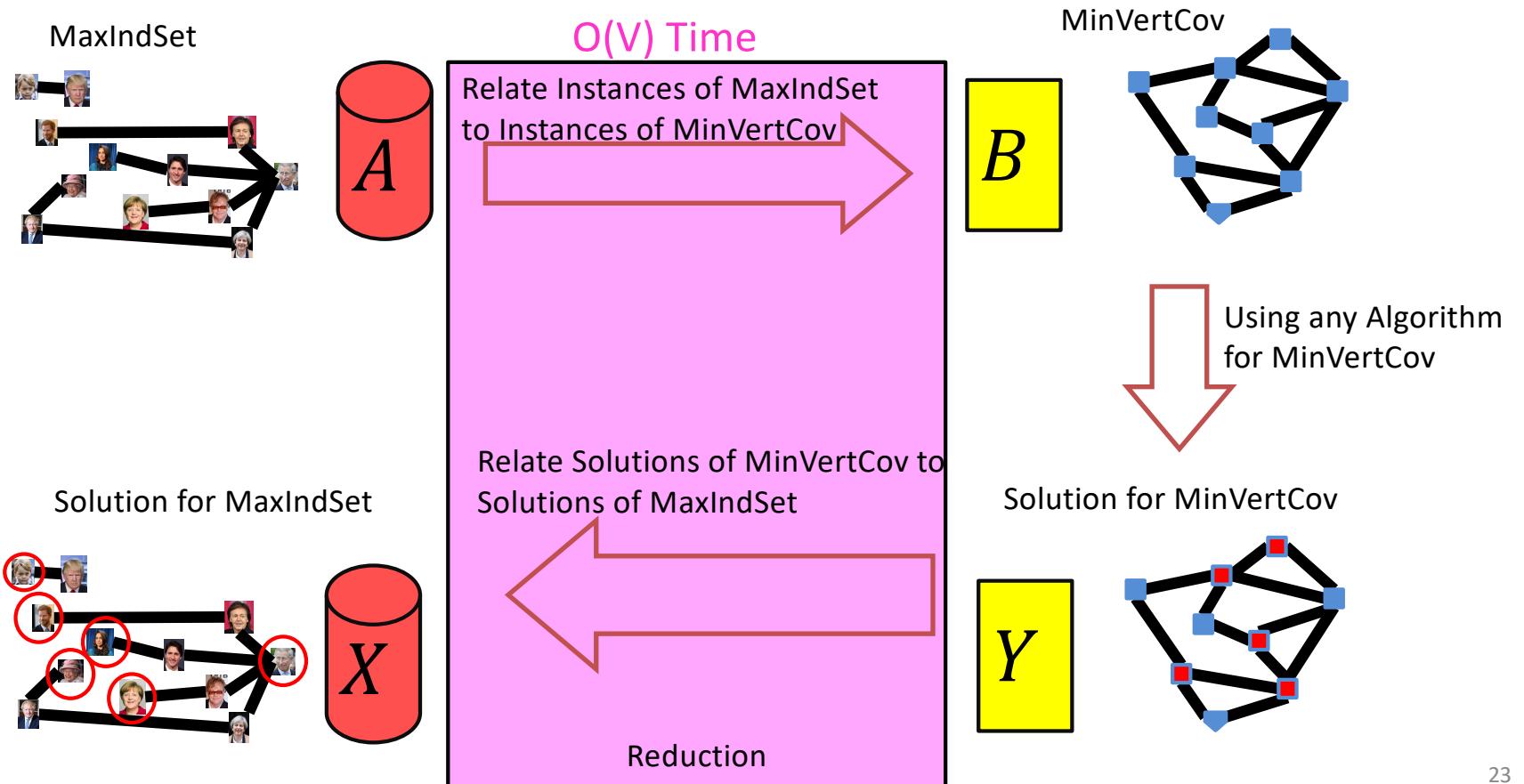
$\text{MaxIndSet} \leq_V \text{MinVertCov}$



If A requires time $\Omega(f(n))$ time then B also requires $\Omega(f(n))$ time

$$A \leq_V B$$

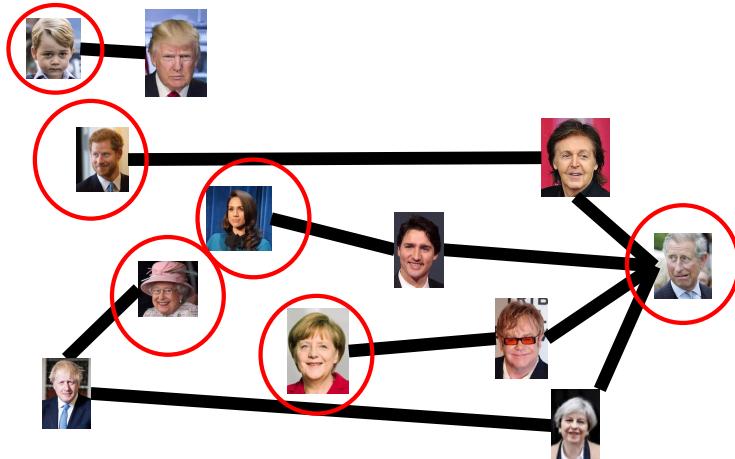
We need to build this Reduction



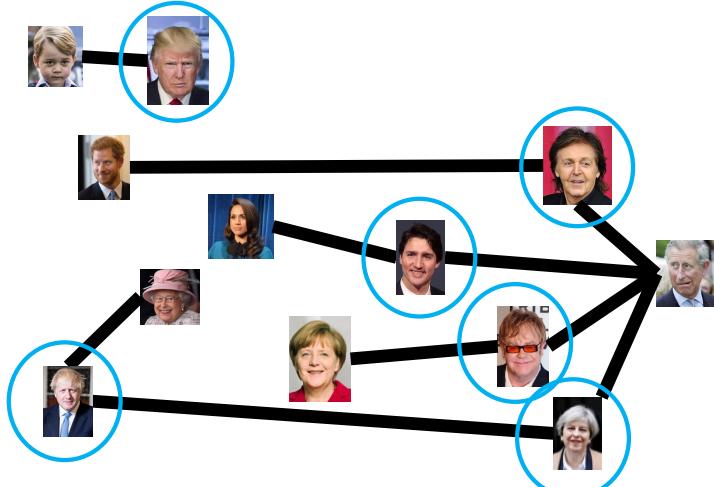
Reduction Idea

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

Independent Set

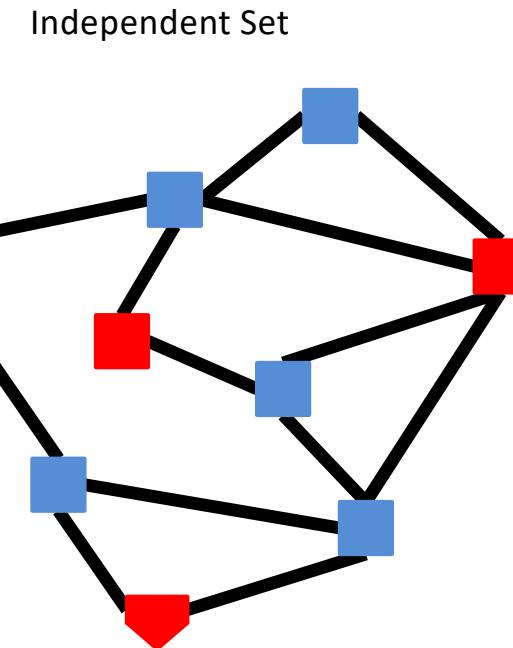
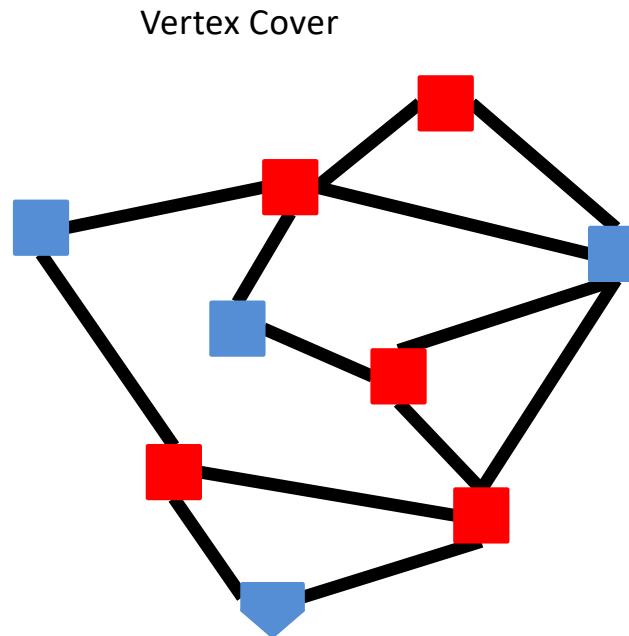


Vertex Cover

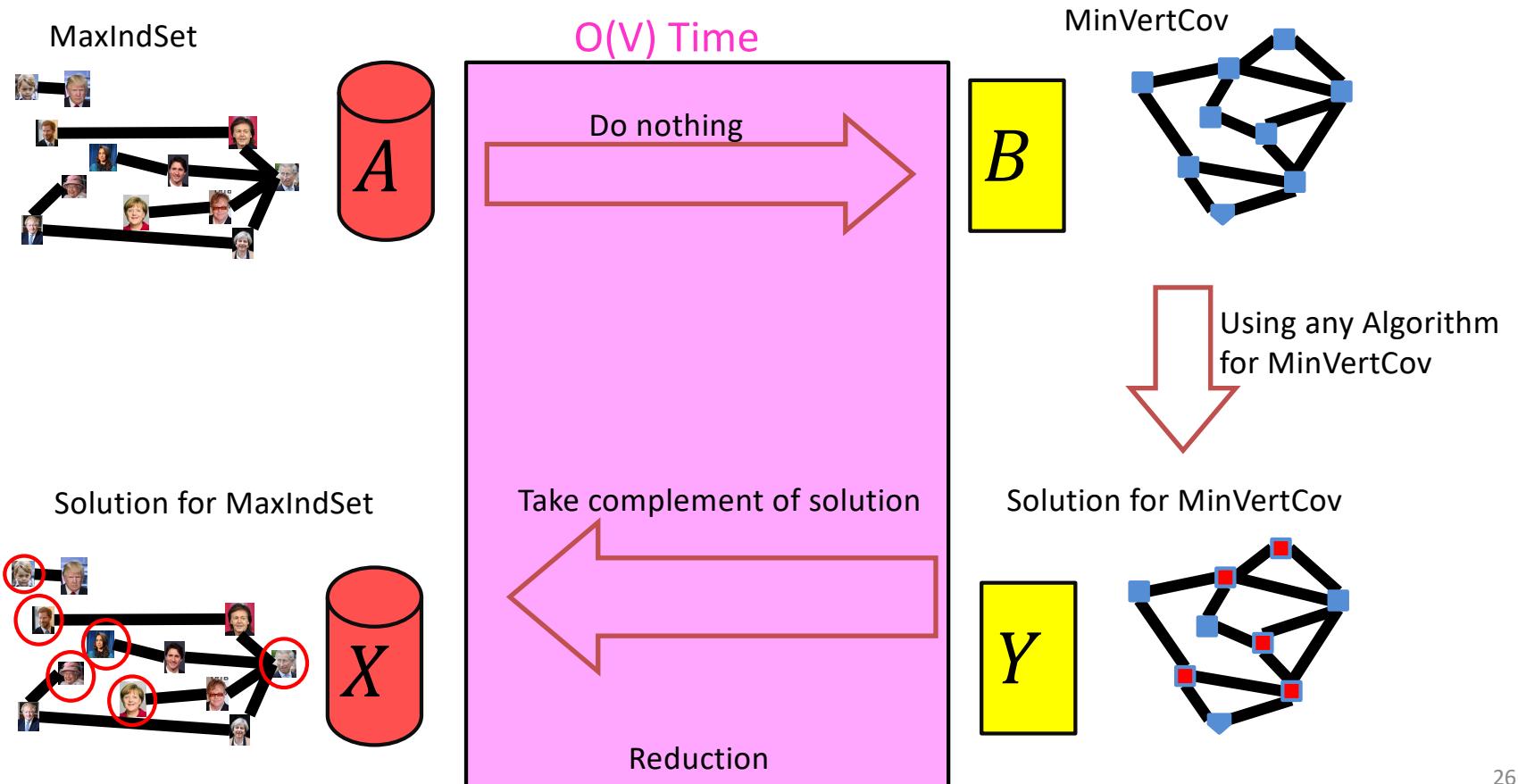


Reduction Idea

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



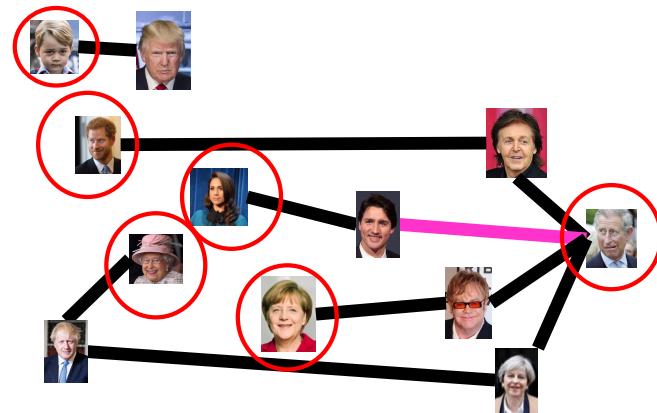
MaxVertCov V -Time Reducible to MinIndSet



Proof: \Rightarrow

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

Let S be an independent set



Consider any edge $(x, y) \in E$

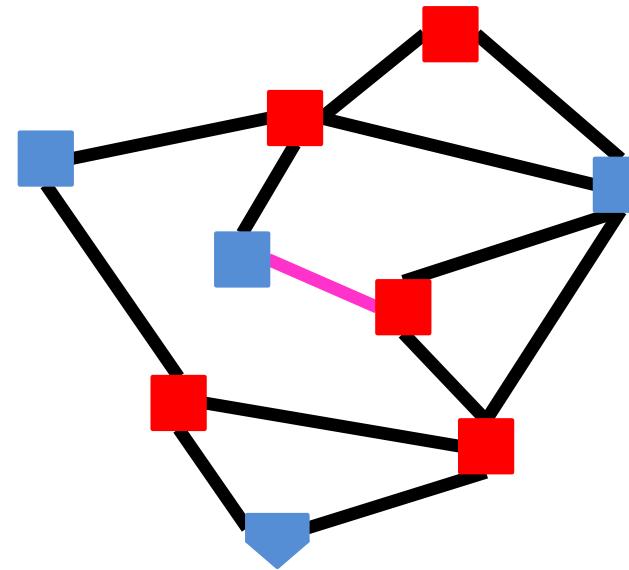
If $x \in S$ then $y \notin S$, because otherwise S would not be an independent set

Therefore $y \in V - S$, so edge (x, y) is covered by $V - S$

Proof: \Leftarrow

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

Let $V - S$ be a vertex cover



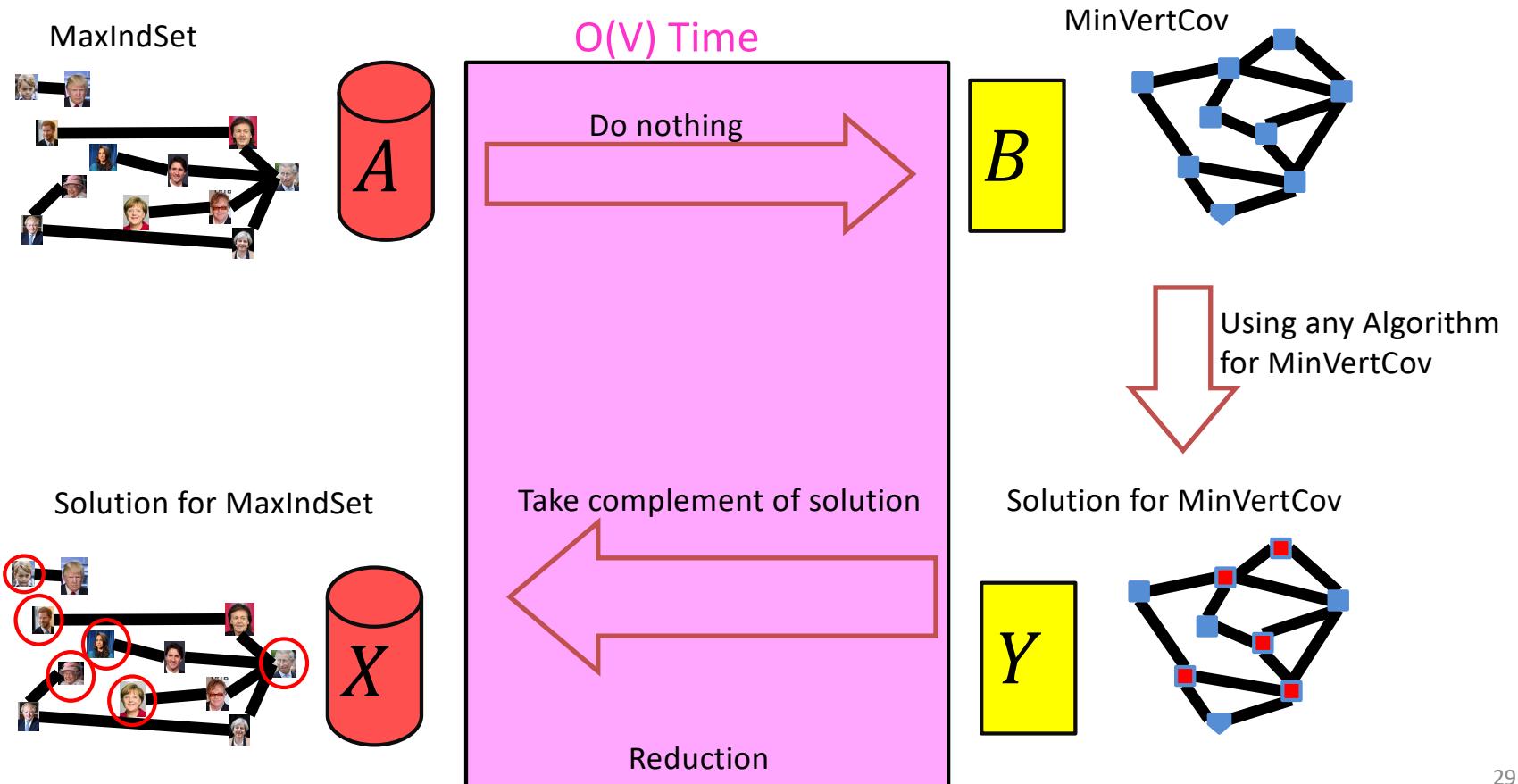
Consider any edge $(x, y) \in E$

At least one of x and y belong to $V - S$, because $V - S$ is a vertex cover

Therefore x and y are not both in S ,

No edge has both end-nodes in S , thus S is an independent set

MaxVertCov V -Time Reducible to MinIndSet



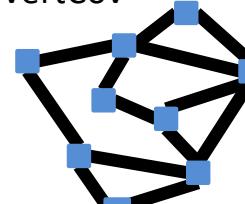
MaxVertCov V -Time Reducible to MinIndSet



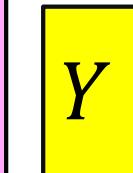
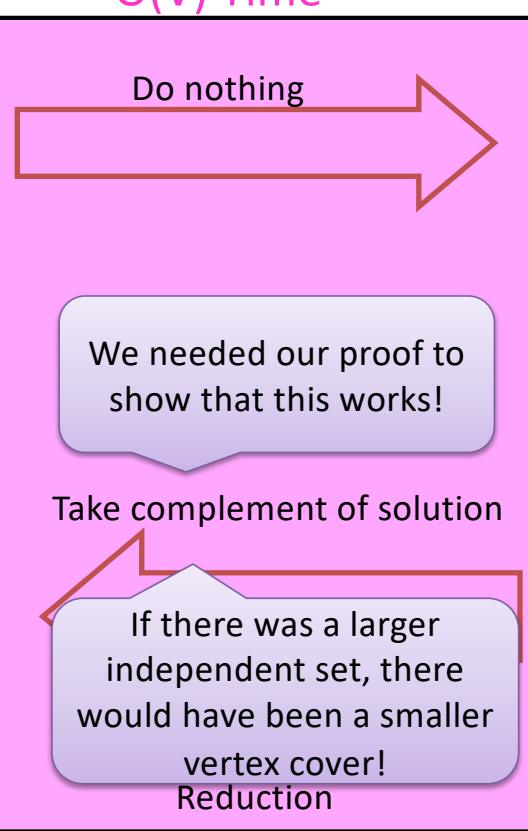
$O(V)$ Time

Do nothing

MinVertCov



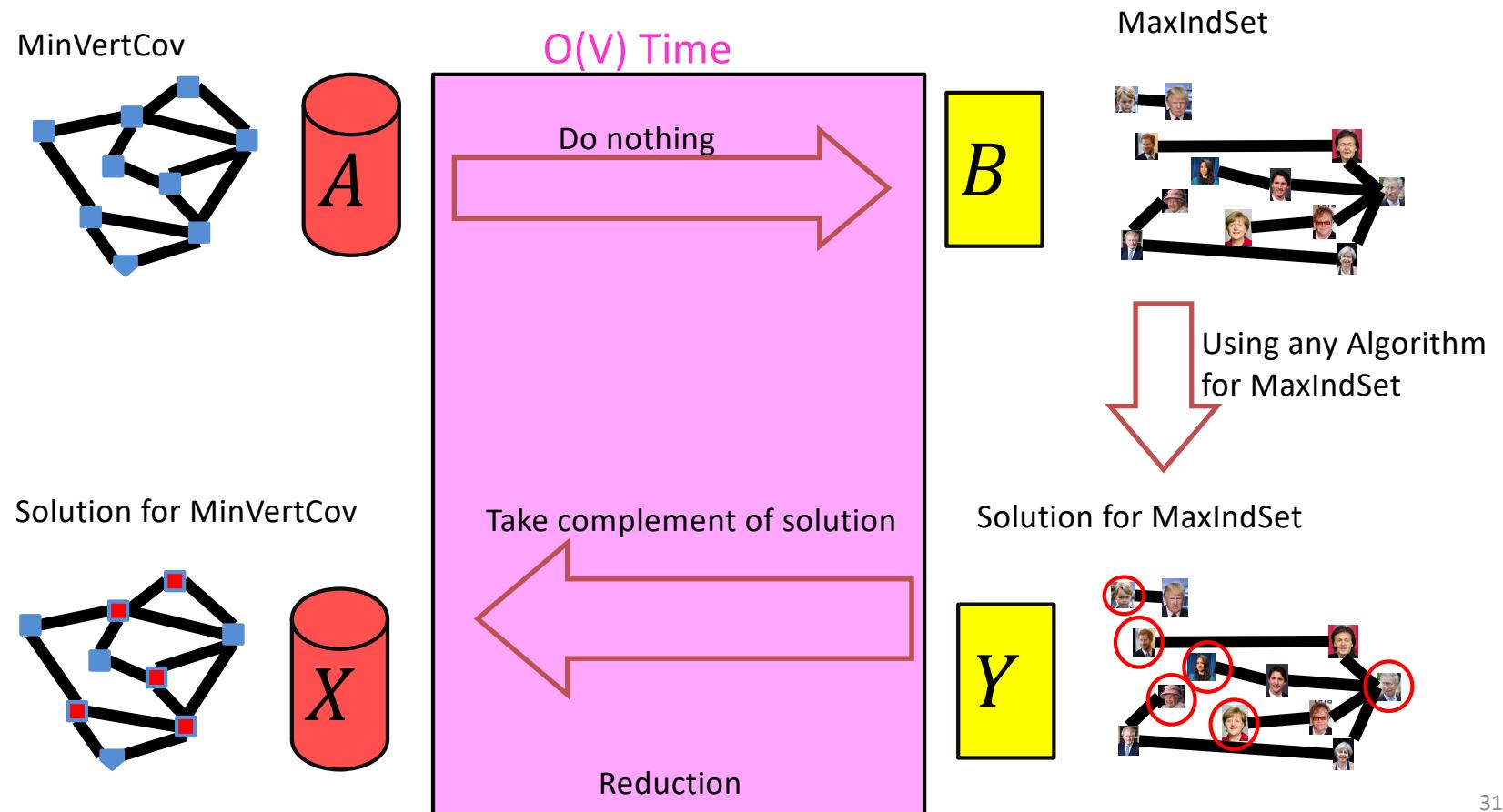
Solution for MaxIndSet



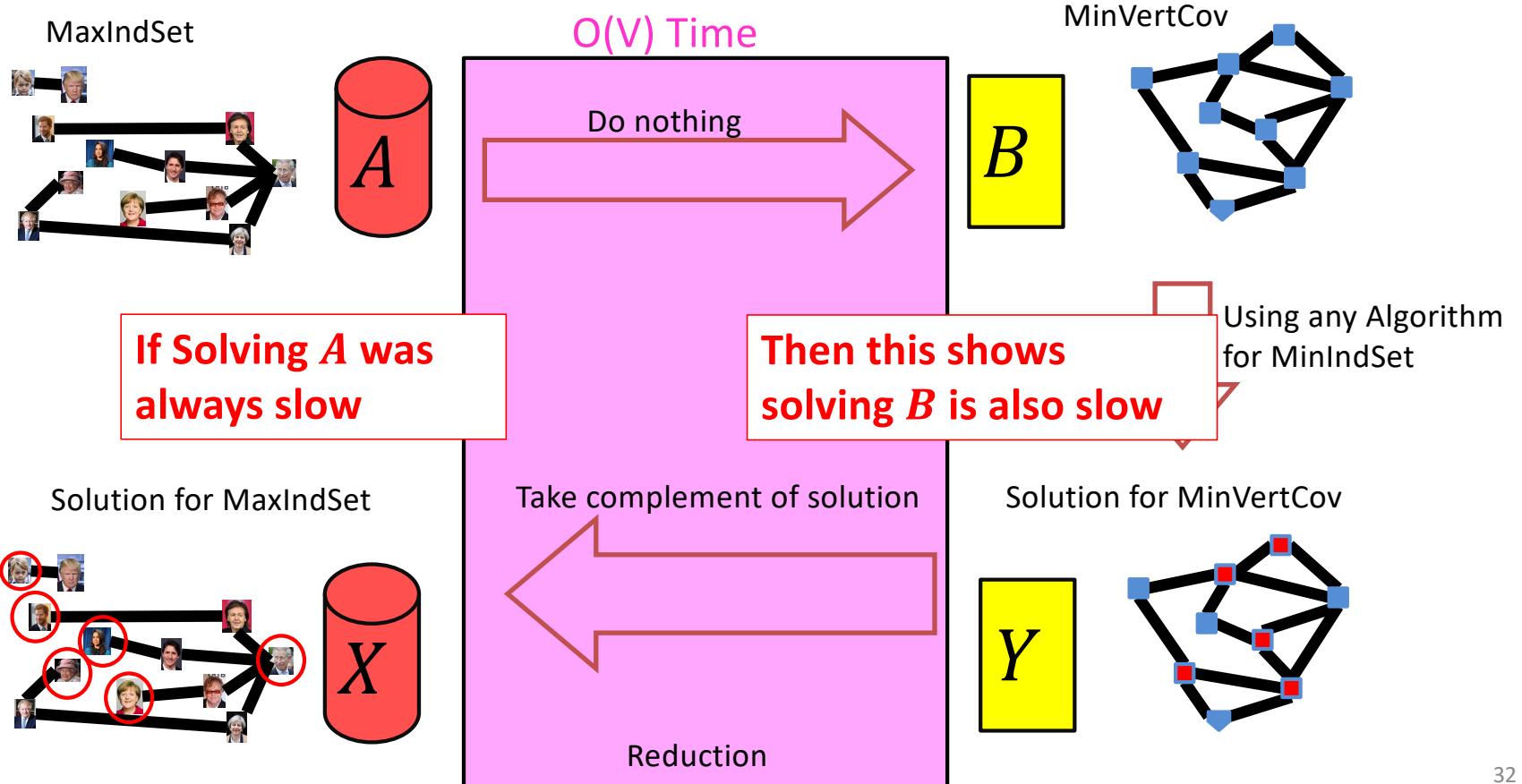
Using any Algorithm for MinVertCov

Solution for MinVertCov

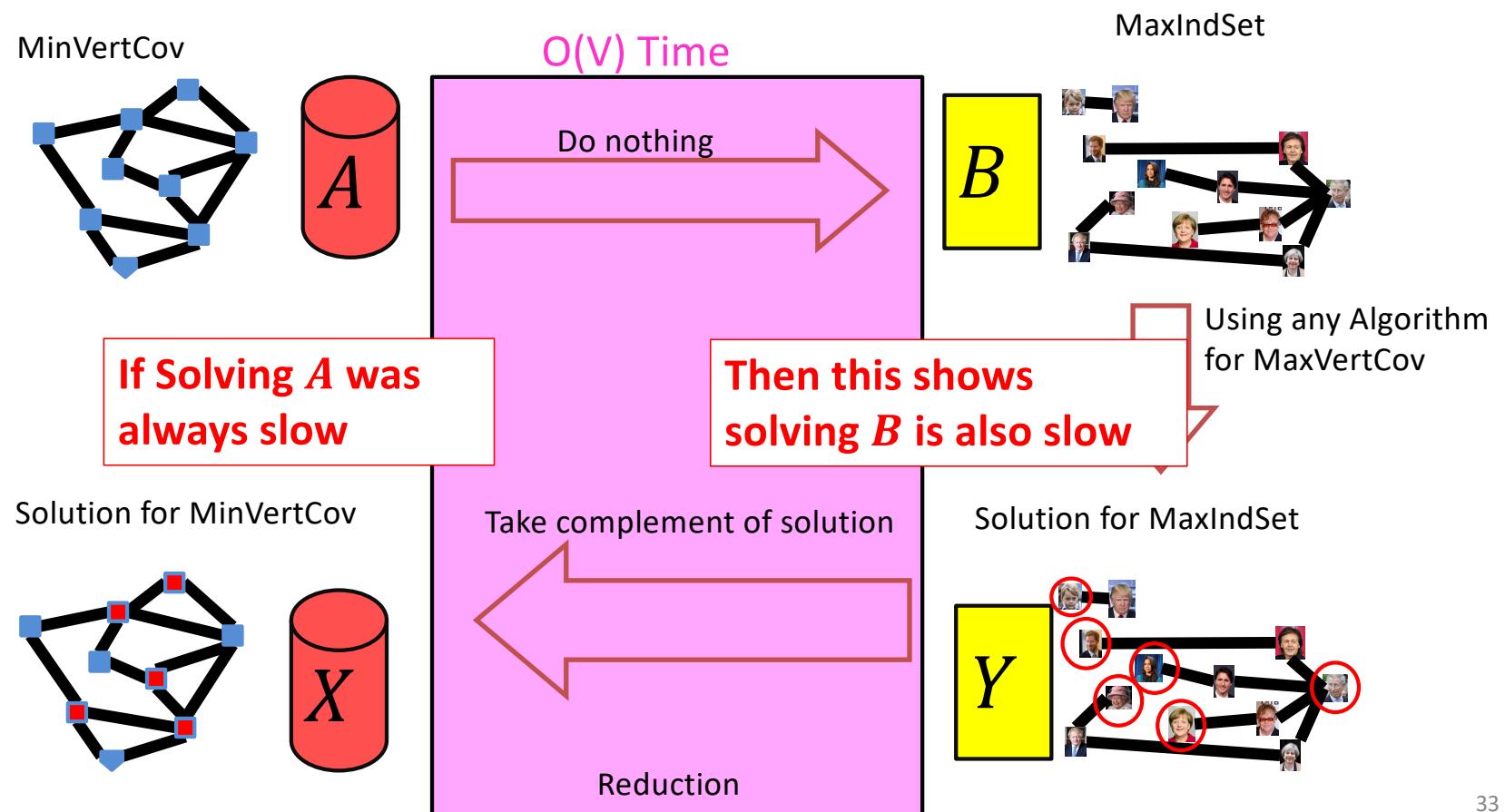
MaxIndSet V -Time Reducible to MinVertCov



Corollary



Corollary



Conclusion

- MaxIndSet and MinVertCov are either both fast, or both slow
 - Spoiler alert: We don't know which!
 - (But we think they're both slow)
 - Both problems are NP-Complete

Mid-class warm up:

What is a Decision Problem?

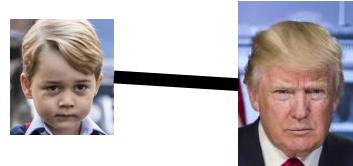
Your response is maybe:

Groan! Do we really need to know?

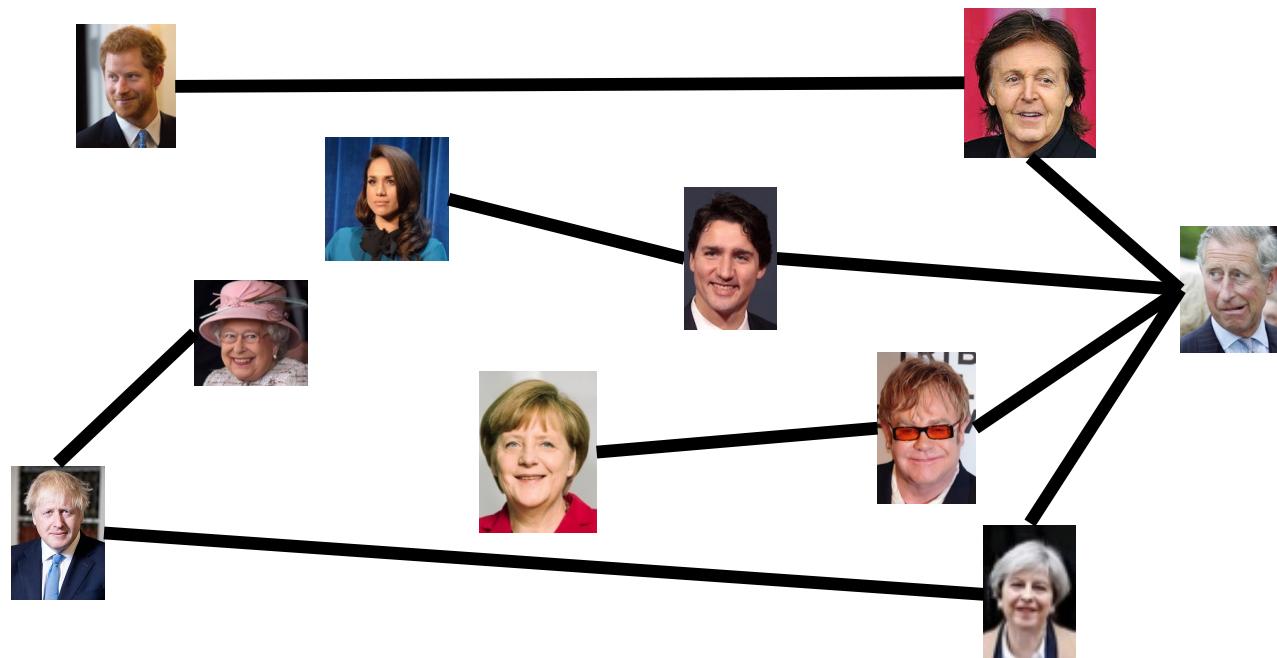
Why do we need to care?

Turns out that the math and theory
on NP-complete problems starts with
decision problems.

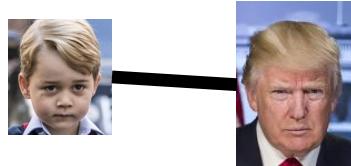
Max Independent Set



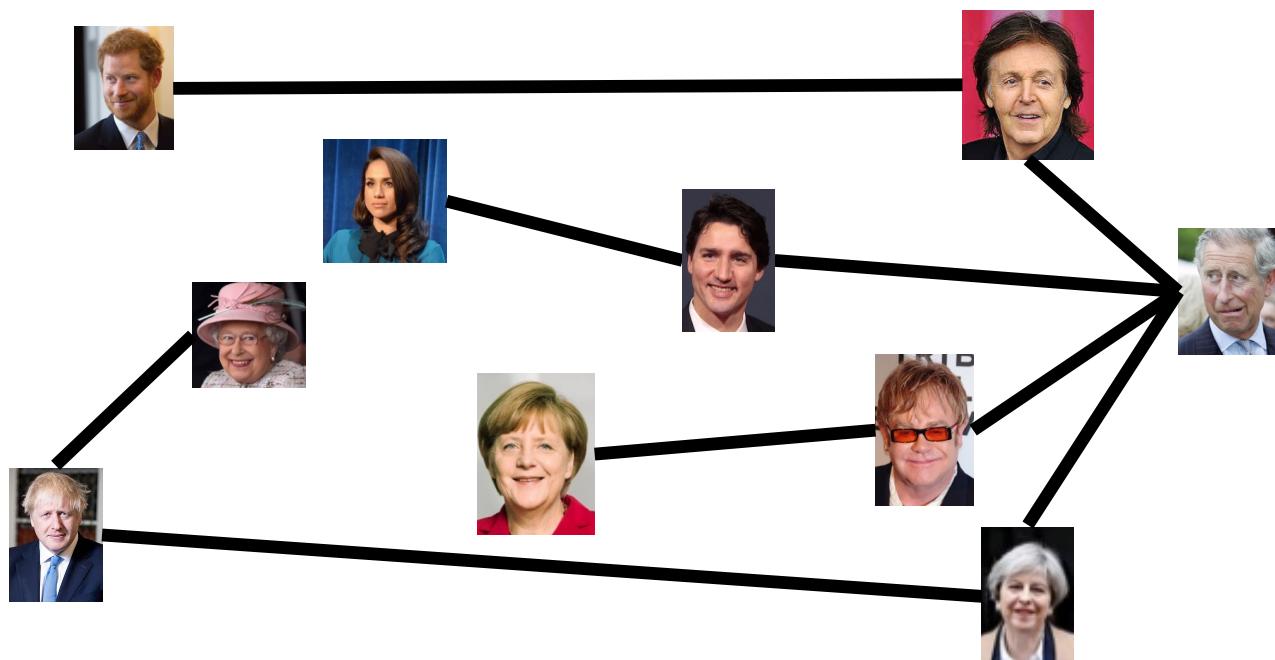
Find the largest set of non-adjacent nodes



k Independent Set



Is there a set of non-adjacent nodes of size k ?



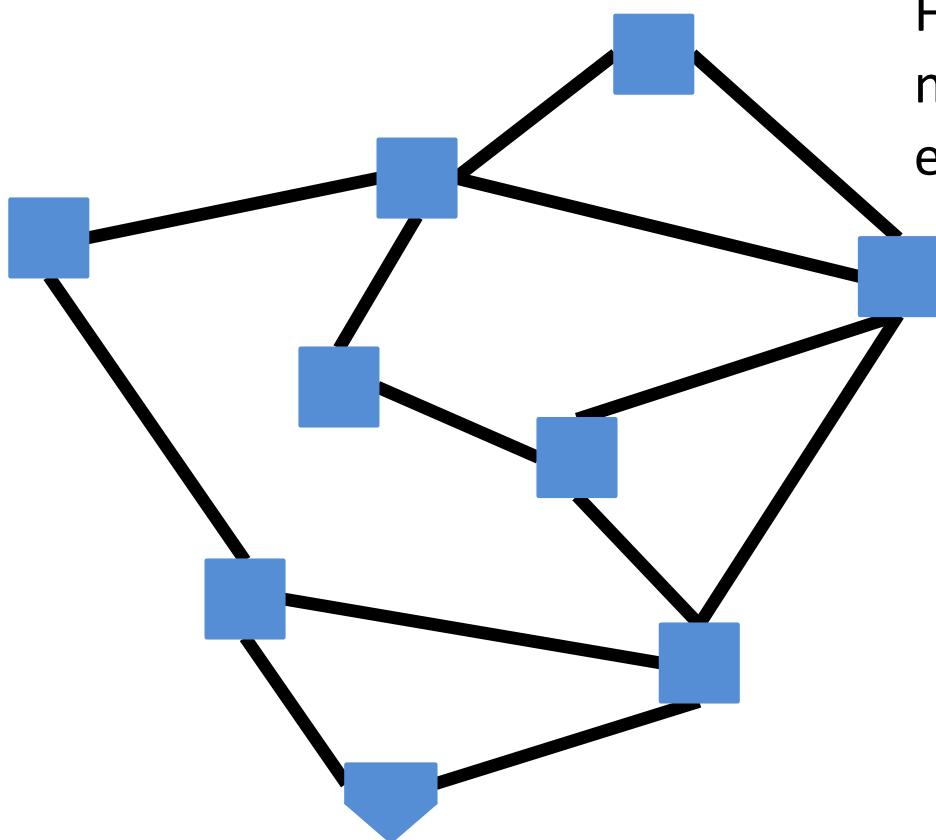
Maximum Independent Set

- Independent set: $S \subseteq V$ is an independent set if no two nodes in S share an edge
- Maximum Independent Set Problem: Given a graph $G = (V, E)$ find the maximum independent set S

k Independent Set

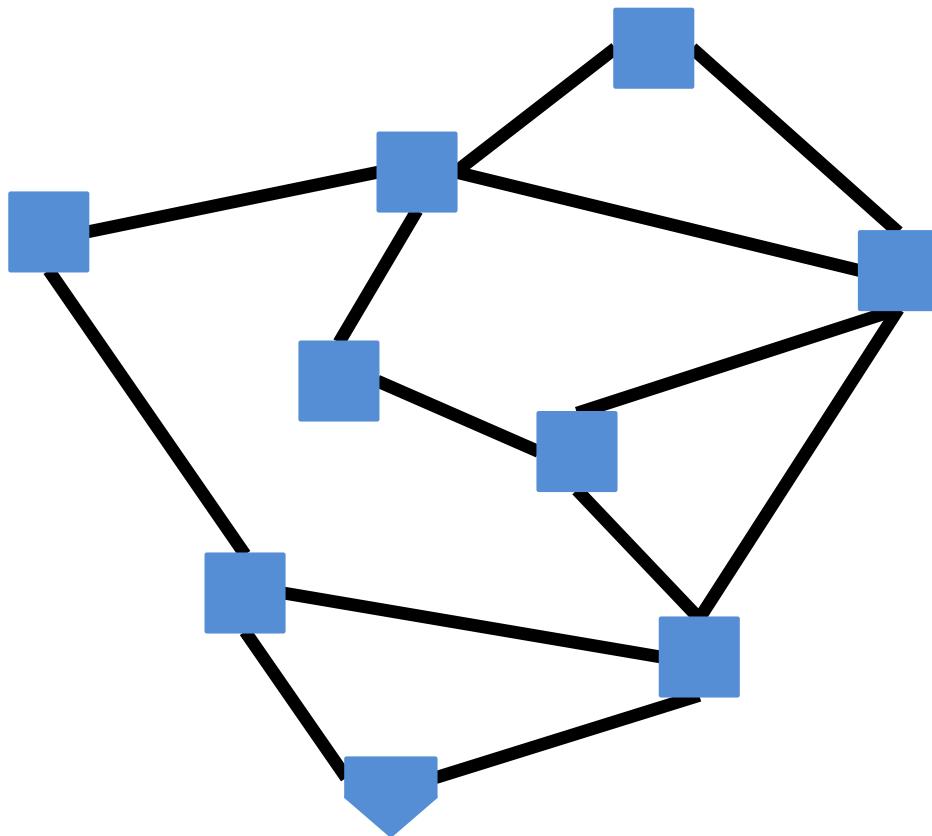
- Independent set: $S \subseteq V$ is an independent set if no two nodes in S share an edge
- k Independent Set Problem: Given a graph $G = (V, E)$ and a number k , **determine whether there is an independent set S of size k**

Min Vertex Cover



Find the smallest set of nodes which covers every edge

k Vertex Cover



Is there a set of nodes of size k which covers every edge?

Minimum Vertex Cover

- Vertex Cover: $C \subseteq V$ is a vertex cover if every edge in E has one of its endpoints in C
- Minimum Vertex Cover: Given a graph $G = (V, E)$ find the minimum vertex cover C

k Vertex Cover

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

Problem Types

- Decision Problems:
 - Is there a solution?
 - Output is True/False
 - Is there a vertex cover of size k ?
- Optimal Value Problems
 - E.g. What's the min k for k -vertex cover problem?
- Search Problems:
 - Find a solution
 - Output is complex
 - Give a vertex cover of size k
- Verification Problems:
 - Given a potential solution, is it valid?
 - Output is True/False
 - Is **this** a vertex cover of size k ?

If we can solve this

Then we can solve this

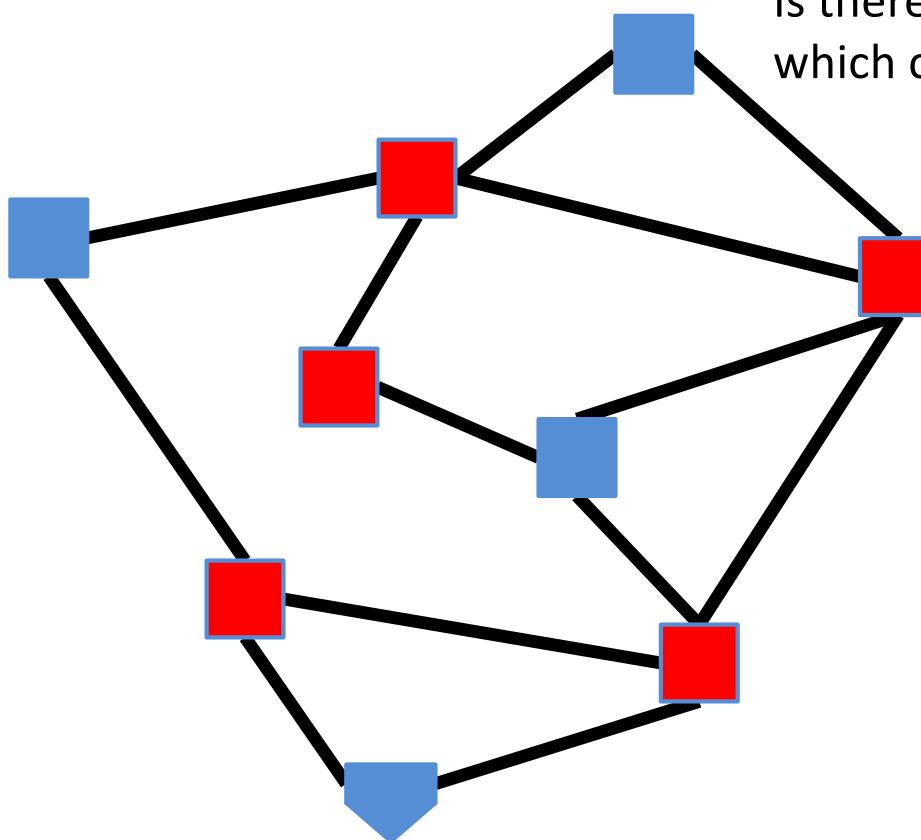
and this

Using a k -VertexCover decider to build a searcher

- Set $i = k - 1$
- Remove nodes (and incident edges) one at a time
- Check if there is a vertex cover of size i
 - If so, then that removed node was part of the k vertex cover, set $i = i - 1$
 - Else, it wasn't

Did I need this node to cover its edges to have a vertex cover of size k ?

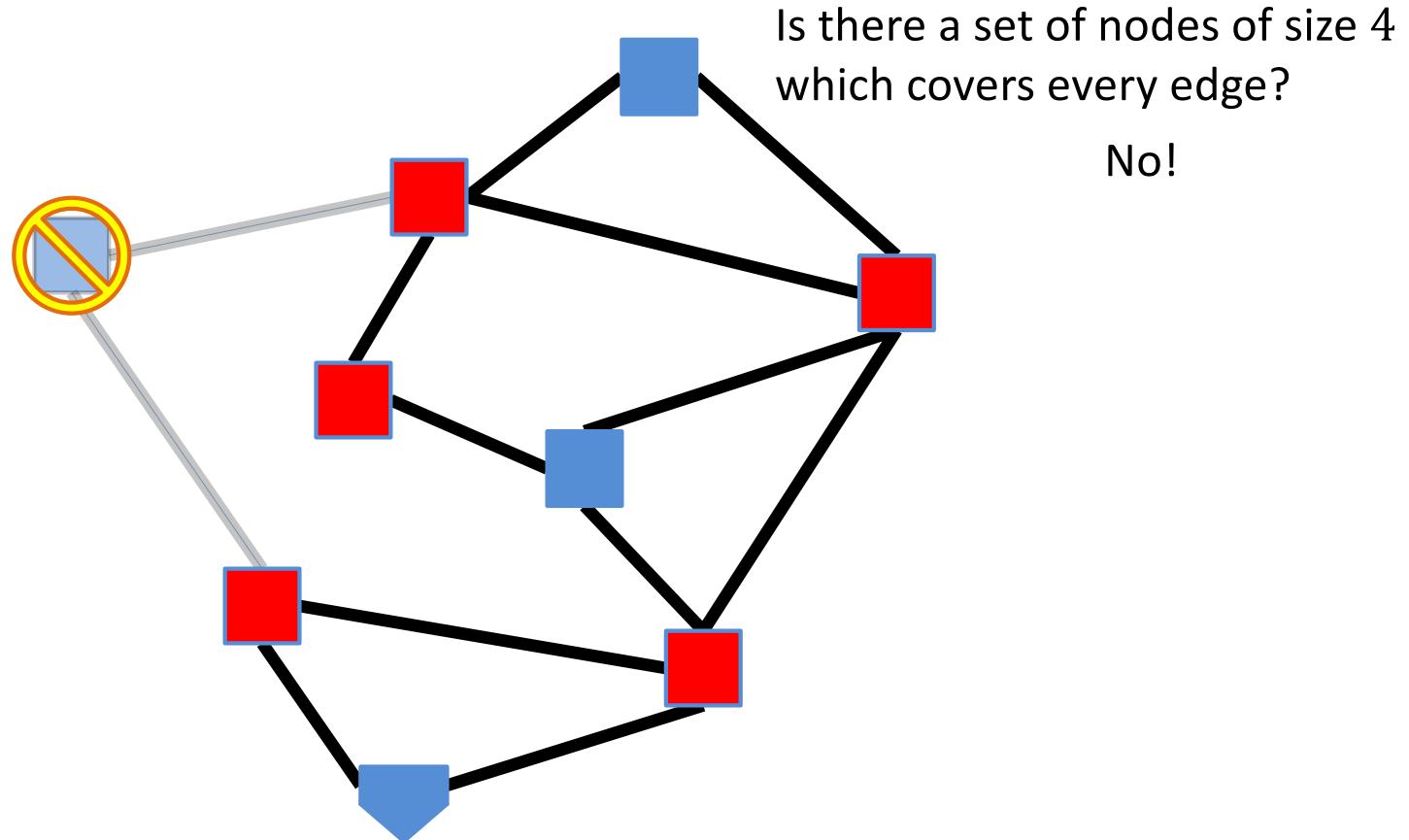
5 Vertex Cover (Decision)



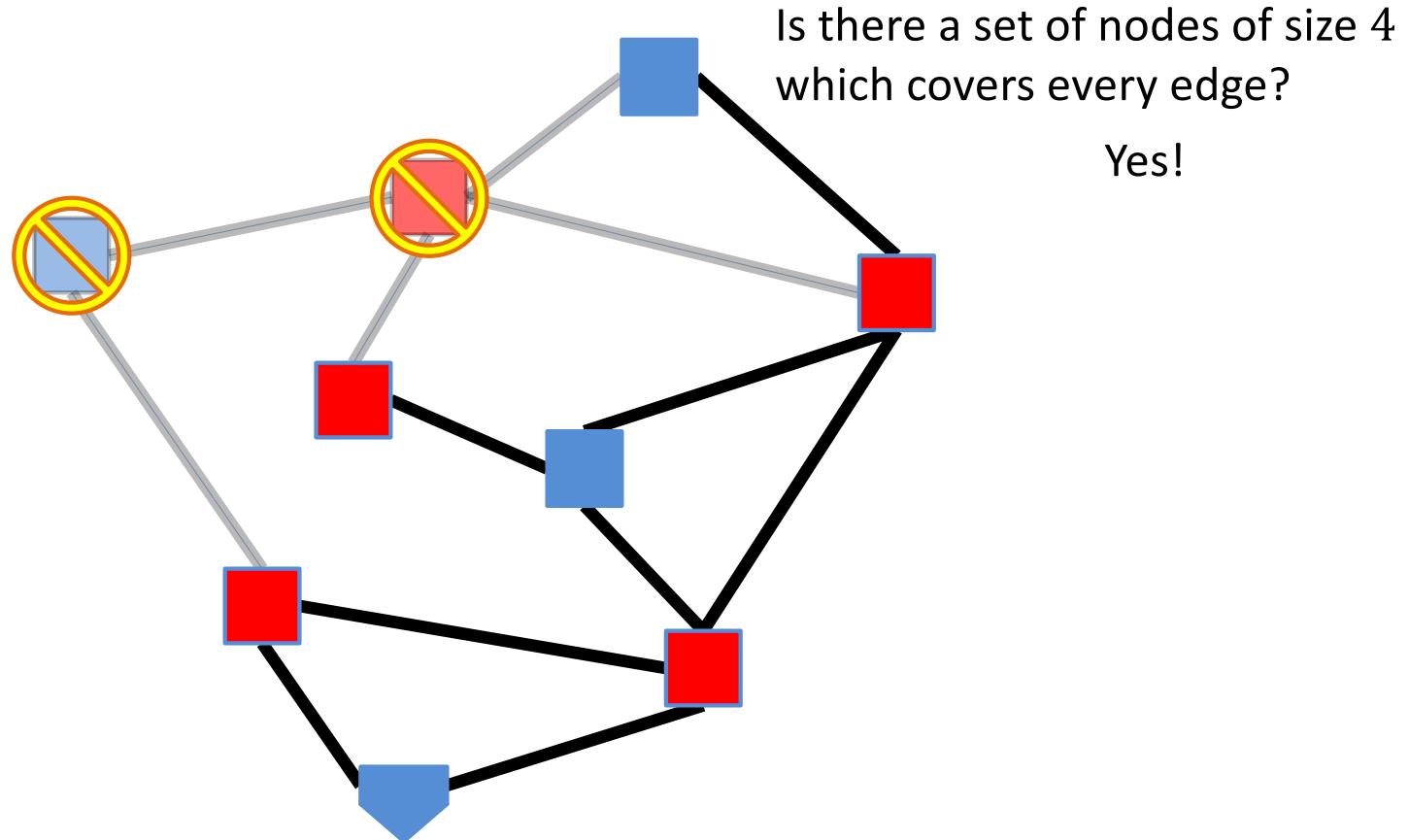
Is there a set of nodes of size 5
which covers every edge?

Yes!

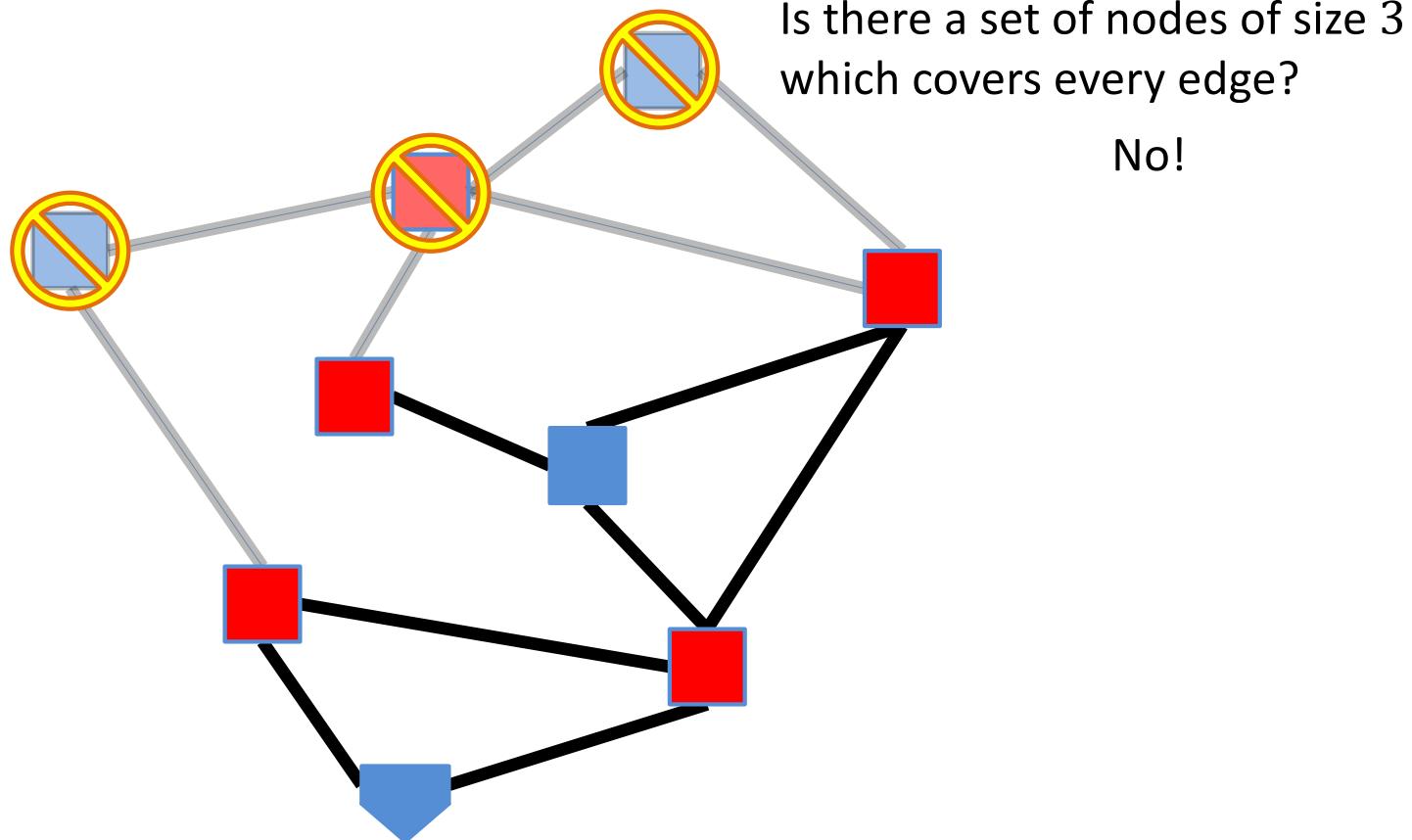
4 Vertex Cover (Decision)



4 Vertex Cover (Decision)



3 Vertex Cover (Decision)

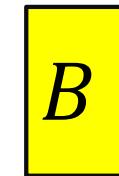


Reduction

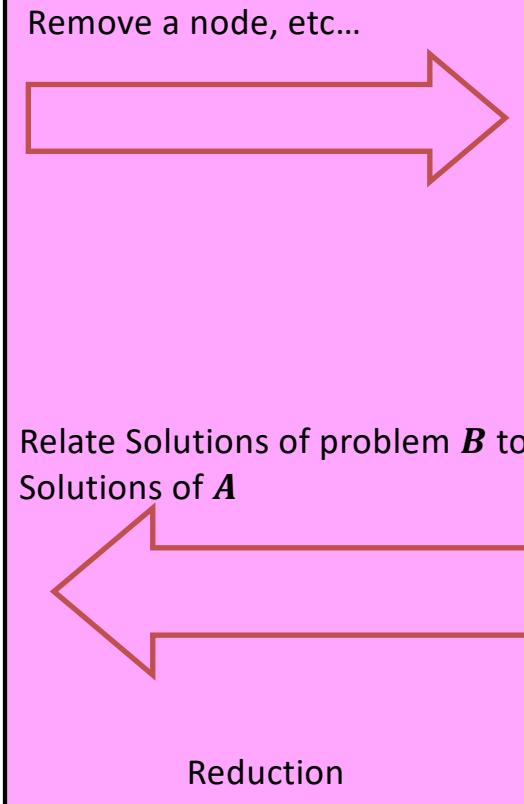
k -VertexCover Solver



k -VertexCover Decider



Solution for A



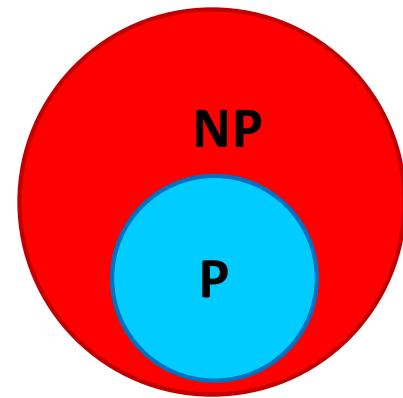
Using any Algorithm
for B

Solution for B



P vs NP

- P
 - Deterministic Polynomial Time
 - Problems solvable in polynomial time
 - $O(n^p)$ for some number p
- NP
 - Non-Deterministic Polynomial Time
 - Problems verifiable in polynomial time
 - $O(n^p)$ for some number p
- Open Problem: Does $P=NP$?
 - Certainly $P \subseteq NP$



k -Independent Set is NP

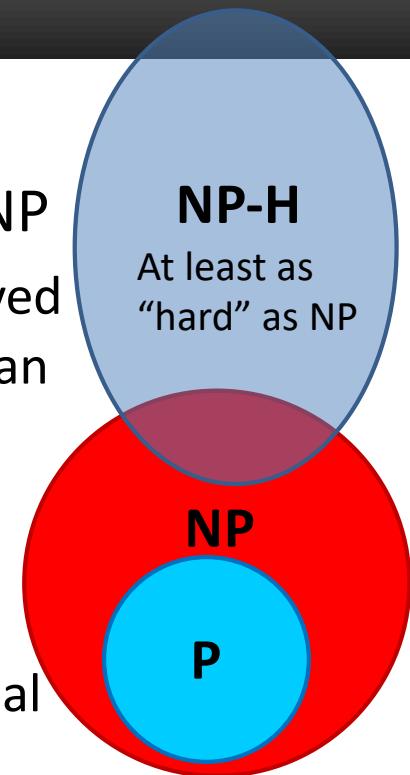
- To show: Given a potential solution, can we **verify** it in $O(n^p)$? [$n = V + E$]

How can we verify it?

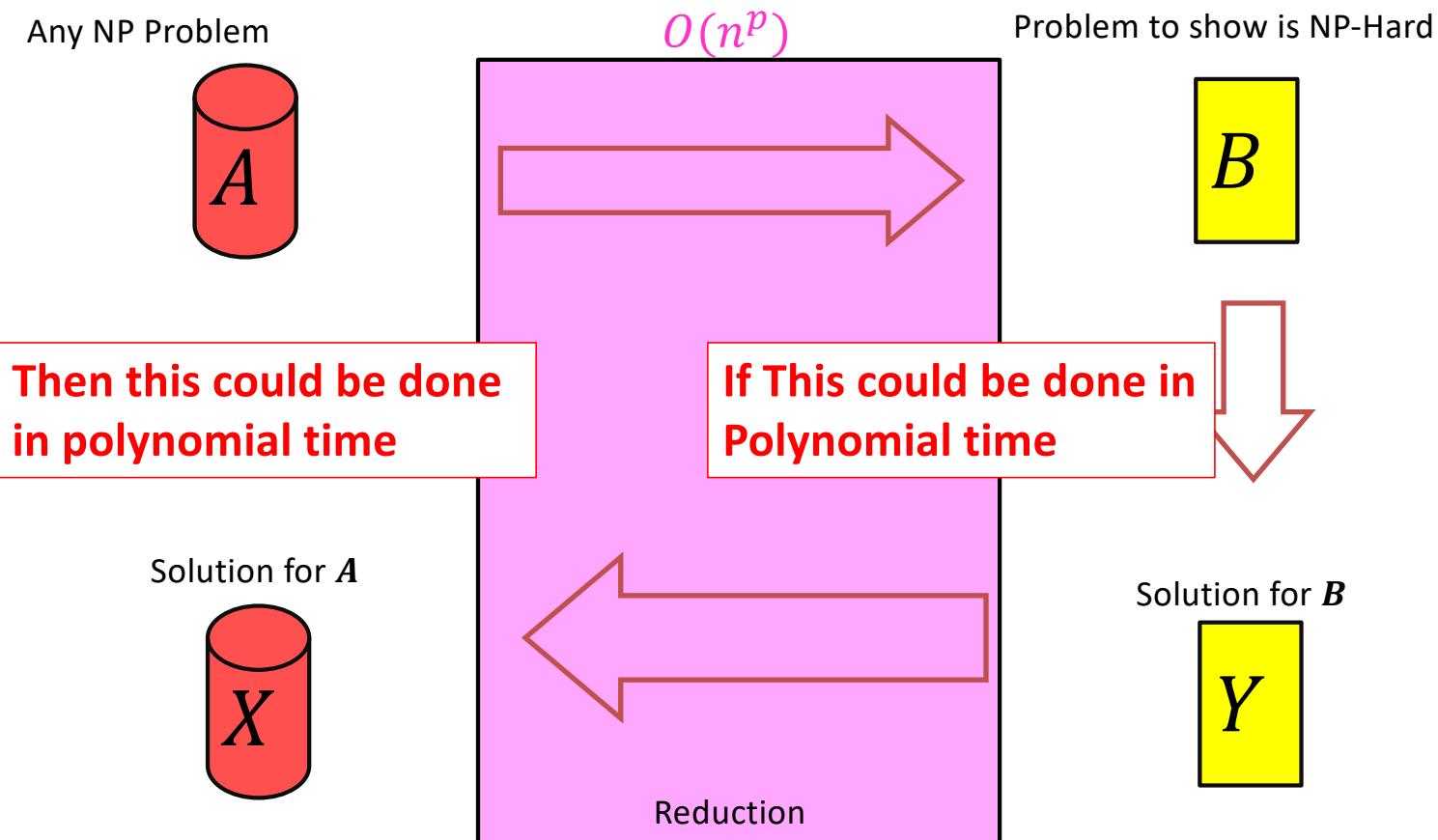
1. Check that it's of size k ? Takes $O(V)$
2. Check that it's an independent set? Takes $O(V^2)$

NP-Hard

- How can we try to figure out if $P=NP$?
- Identify problems at least as “hard” as NP
 - If any of these “hard” problems can be solved in polynomial time, then all NP problems can be solved in polynomial time.
- Definition: NP-Hard:
 - B is NP-Hard if $\forall A \in NP, A \leq_p B$
 - $A \leq_p B$ means A reduces to B in polynomial time

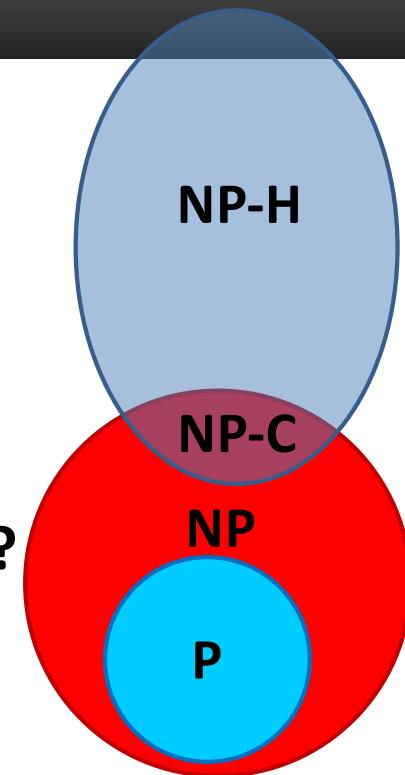


NP-Hardness Reduction

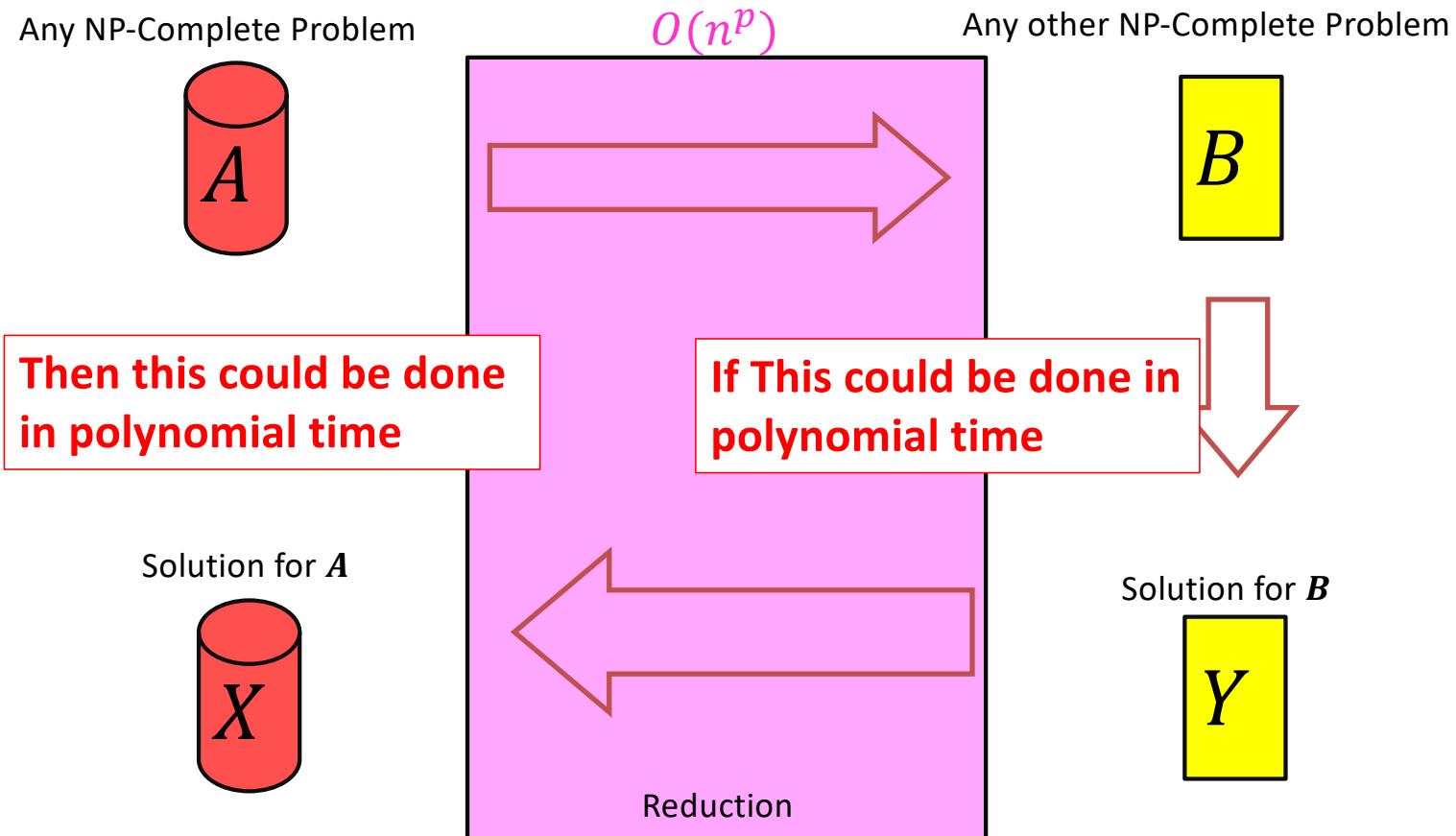


NP-Complete

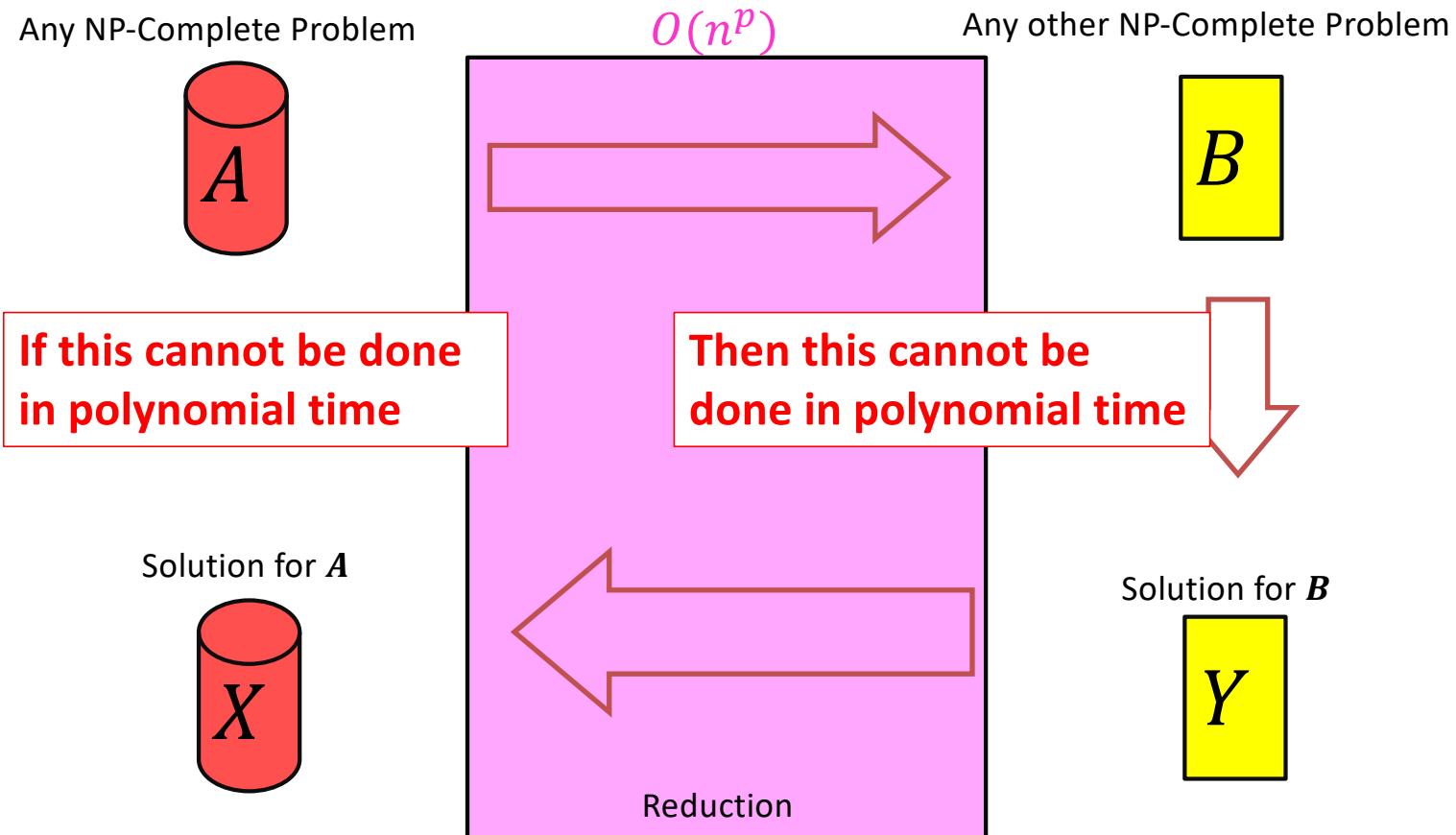
- “Together they stand, together they fall”
 - Problems solvable in polynomial time iff ALL NP problems are
 - $\text{NP-Complete} = \text{NP} \cap \text{NP-Hard}$
 - **How to show a problem is NP-Complete?**
 - Show it belongs to NP
 - Give a polynomial time verifier
 - Show it is NP-Hard
 - Give a reduction from another NP-H problem
- We now just need a FIRST NP-Hard problem



NP-Completeness



NP-Completeness



Wrap Up

- Reductions used to show “hardness” relationships between problems
- Intractable problems often reduce to each other
- Starting to define “classes” of problems based on complexity issues
 - P are problems that can be solved in polynomial time
 - NP are problems where a solution can be verified in polynomial time
 - NP-hard are problems that are at least as hard as anything in NP
 - NP-complete are NP-hard problems that “stand or fall together”