

# NP-Completeness

CS 4102: Algorithms

Spring 2022

Robbie Hott and Tom Horton

# Today's Keywords

- Reductions
- P vs NP
- NP Hard, NP Completeness
- k-Independent Set
- k-Vertex Cover
- 3SAT
- k-Clique

# Reductions

(We're about to get interested in problems that seem to require exponential time...)

# Max-flow vs. Edge-disjoint paths

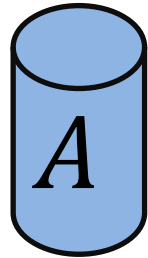
- These two problems are “related”
  - Here we’re saying: if you can solve **Max-flow**, you can solve **Edge-disjoint path**
- Alternatively, we can say that one problem **reduces** to the other
  - The problem of finding Edge-disjoint paths **reduces to** the problem of finding max-flow
  - Maybe this **reduction** requires some work to “convert”
    - Could be nothing or minimal
  - For these problems, the cost of the conversion is **hopefully small (more on this in a moment)**.

# Reduction

- A **reduction** is a transformation of one problem into another problem
  - Edge-disjoint paths is reducible to max-flow because we can use max-flow to solve it
  - Formally, problem A is **reducible** to problem B if we can use a solution to B to solve A
- We're particularly interested in reductions that happen fast!
  - *Meaning the work to do the conversion is fast (how fast?)*
- If A is **polynomial-time reducible** to B, we denote this as:  
 $A \leq_p B$
- **If the conversion takes linear time, we might say  $A \leq_n B$**

# In General: A Reduction

A problem we don't know how to solve

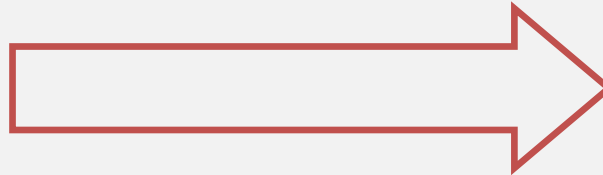


Solution for  $A$

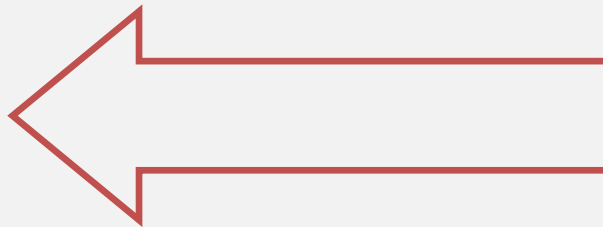


## Reduction

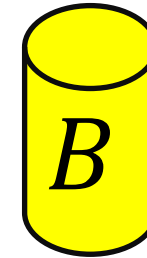
Map Instances of problem  $A$  to Instances of  $B$



Map Solutions of problem  $B$  to Solutions of  $A$



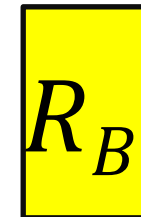
A problem we do know how to solve



Using any Algorithm for  $B$

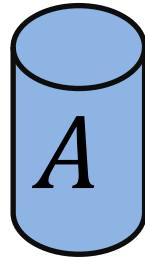


Solution for  $B$



# In General: Reduction

Problem we don't  
know how to solve

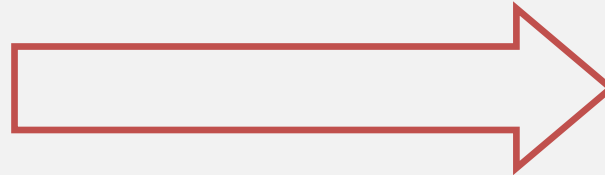


Solution for  $A$

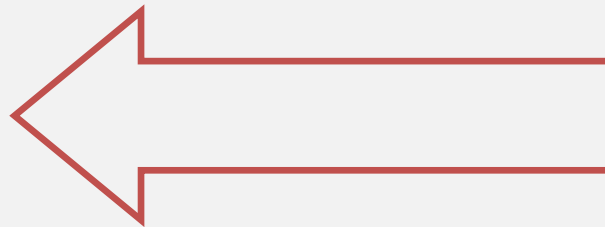


## Reduction

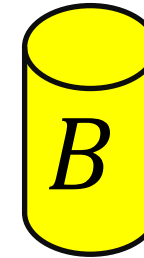
Map Instances of problem  
 $A$  to Instances of  $B$



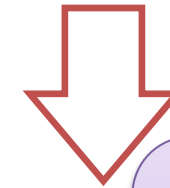
Map Solutions of problem  
 $B$  to Solutions of  $A$



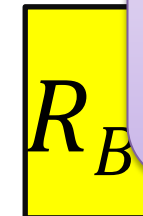
Problem we do know  
how to solve



Using any Algorithm for  $B$



Solution



For now: we are **NOT**  
focusing on an algorithm  
to solve one of these,  
just on the reduction!

# Total Runtime

- The total runtime to solve A is:
  - ***Convert(A→B)***: Time it takes to convert problem A into problem B
  - ***Execute(B)***: Time it takes to execute algorithm to solve B
  - ***Solve(B→A)***: Time to convert solution of B back to solution of A
- ***Total Runtime: Convert(A→B) + Execute(B) + Solve(B→A)***
- Do you see why we want convert() and solve() to be FAST. We want the slowest part to be the actual algorithm that solves B if possible.



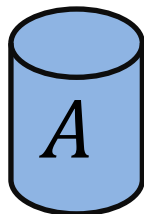
# Relative Hardness

***Total Runtime: Convert(A->B) + Execute(B) + Solve(B->A)***

- Generally, we want execute() to be the slowest term, we then can say  $A \leq_{\alpha} B$  where alpha is the runtime of convert() and solve()
- If Execute() IS the slowest part, then what does that mean about the relative difficulty of solving A versus B??
  - It means that B is as hard or harder than A. Why?
  - Because B provides an algorithm that solves A, so if algorithm to solve B gets faster, so does the algorithm that solves A.

# Reductions and Hardness (again)

Problem we don't know how to solve



Solution for  $A$



Solve  $A$  using  $B$ 's solver.

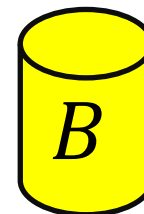
Cost: Cost of  $B$ 's solver plus  $O(f(n))$  for reduction.

## Reduction

Map Instances of problem  $A$  to Instances of  $B$

Reduction cost:  $O(f(n))$

Map Solutions of problem  $B$  to Solutions of  $A$



Problem we do know how to solve

Using any Algorithm for  $B$



Solution for  $B$



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

$A$  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

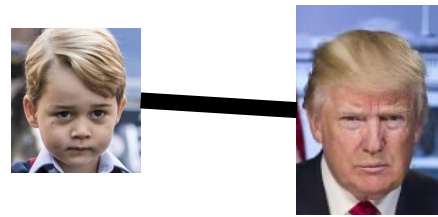
$$A \leq_{f(n)} B$$

# Reduction for Bipartite Matching

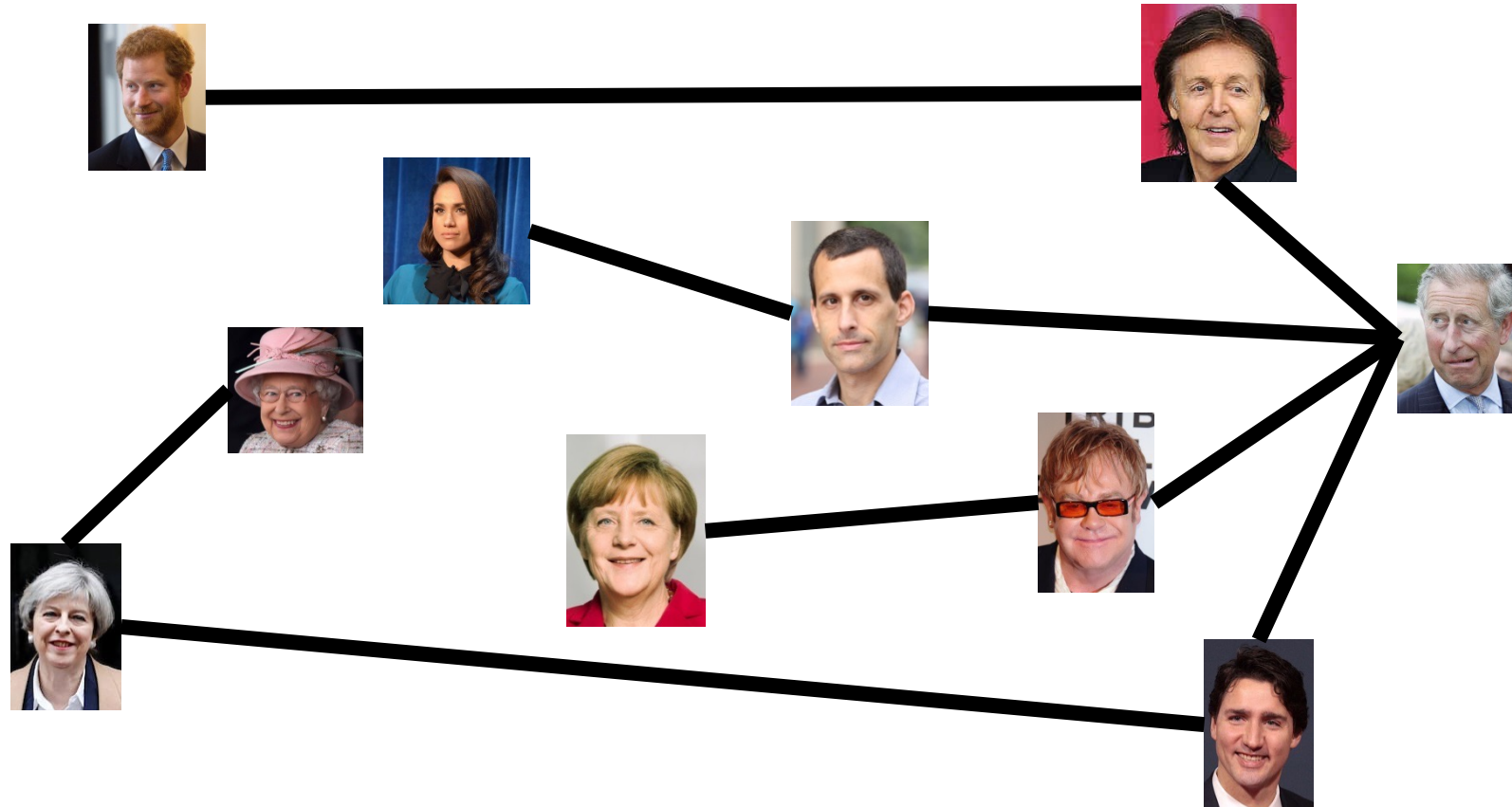
- We have transformed (in polynomial time) a **bipartite matching** problem into a **max flow** problem
- Specifically, bipartite-matching  $\leq_p$  max-flow
  - Because we can transform bipartite matching to max-flow in polynomial time
- But is it the case that max-flow  $\leq_p$  bipartite-matching?
  - Not so much: a solution to bipartite matching does not help us with a non-bipartite graph



# Party Problem



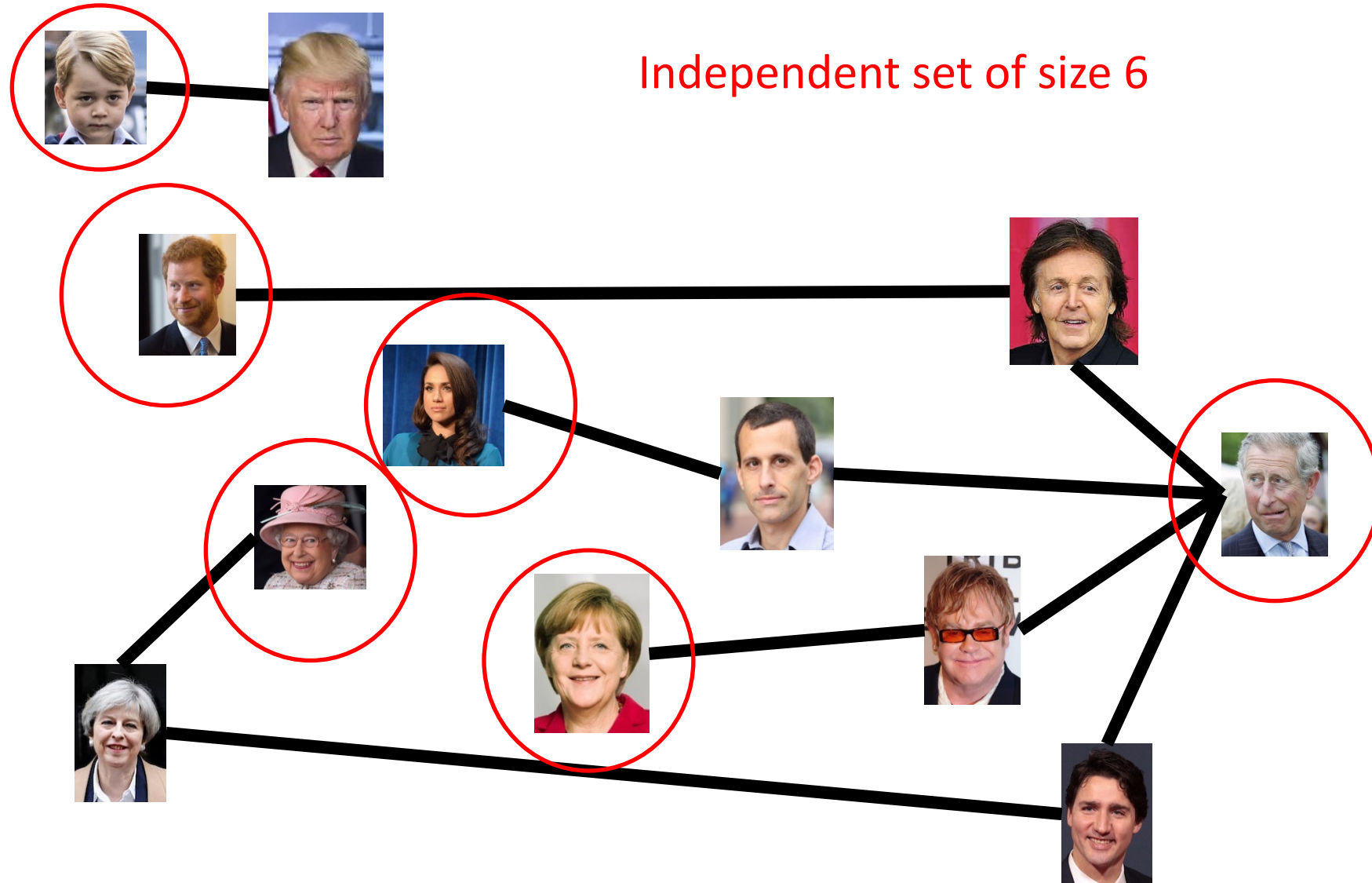
Draw edges between people who don't get along.  
Find the maximum number of people who get along.



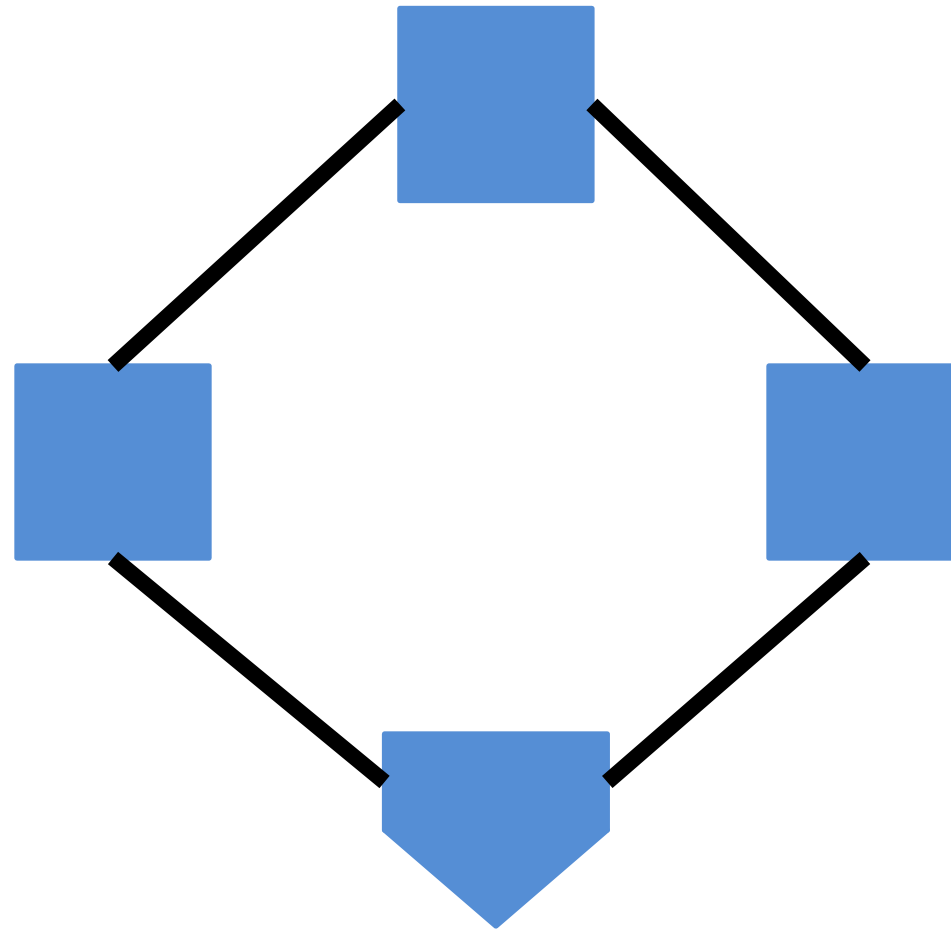
# Maximum Independent Set

- 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

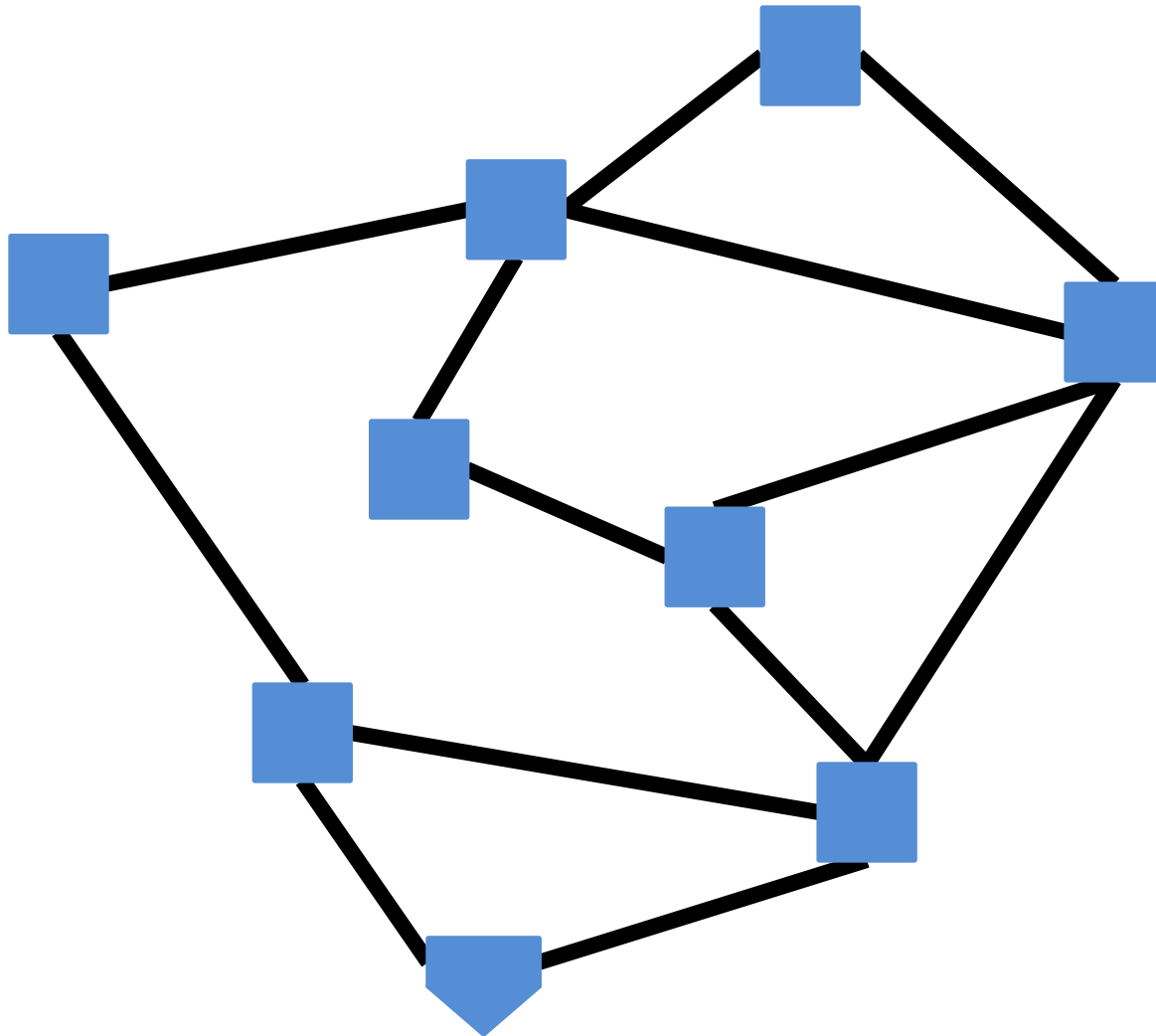


# Generalized Baseball





# Generalized Baseball



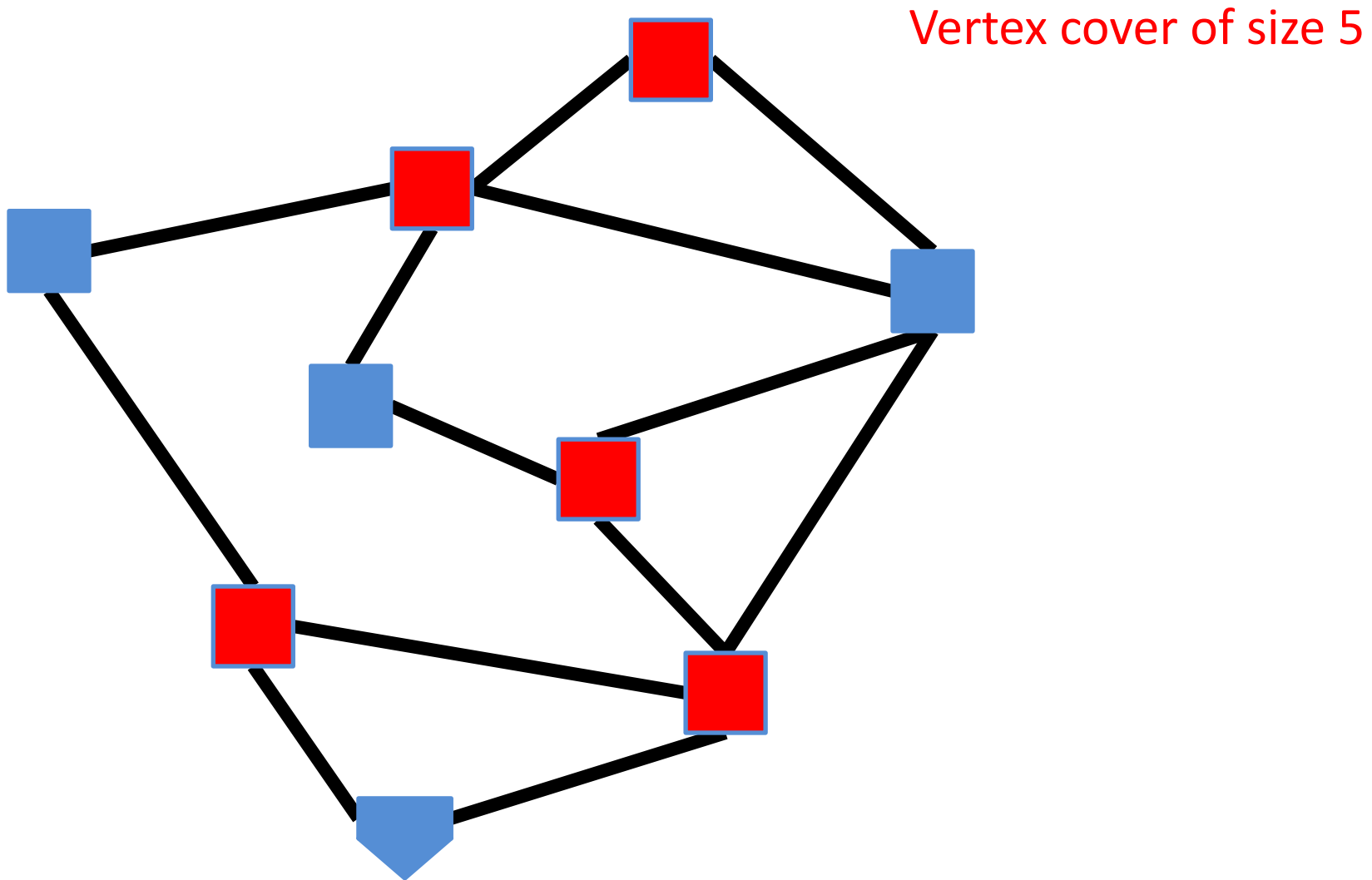
Need to place defenders on bases such that every edge is defended

What's the fewest number of defenders needed?

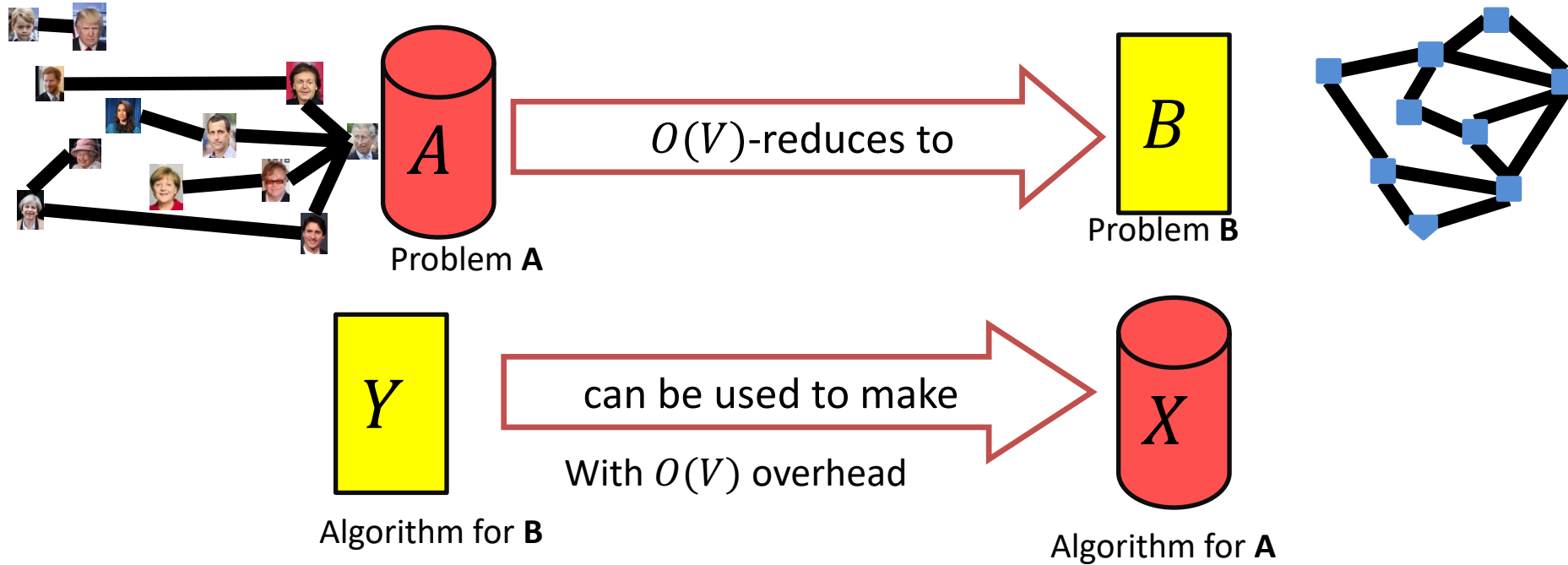
# Minimum Vertex Cover

- 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



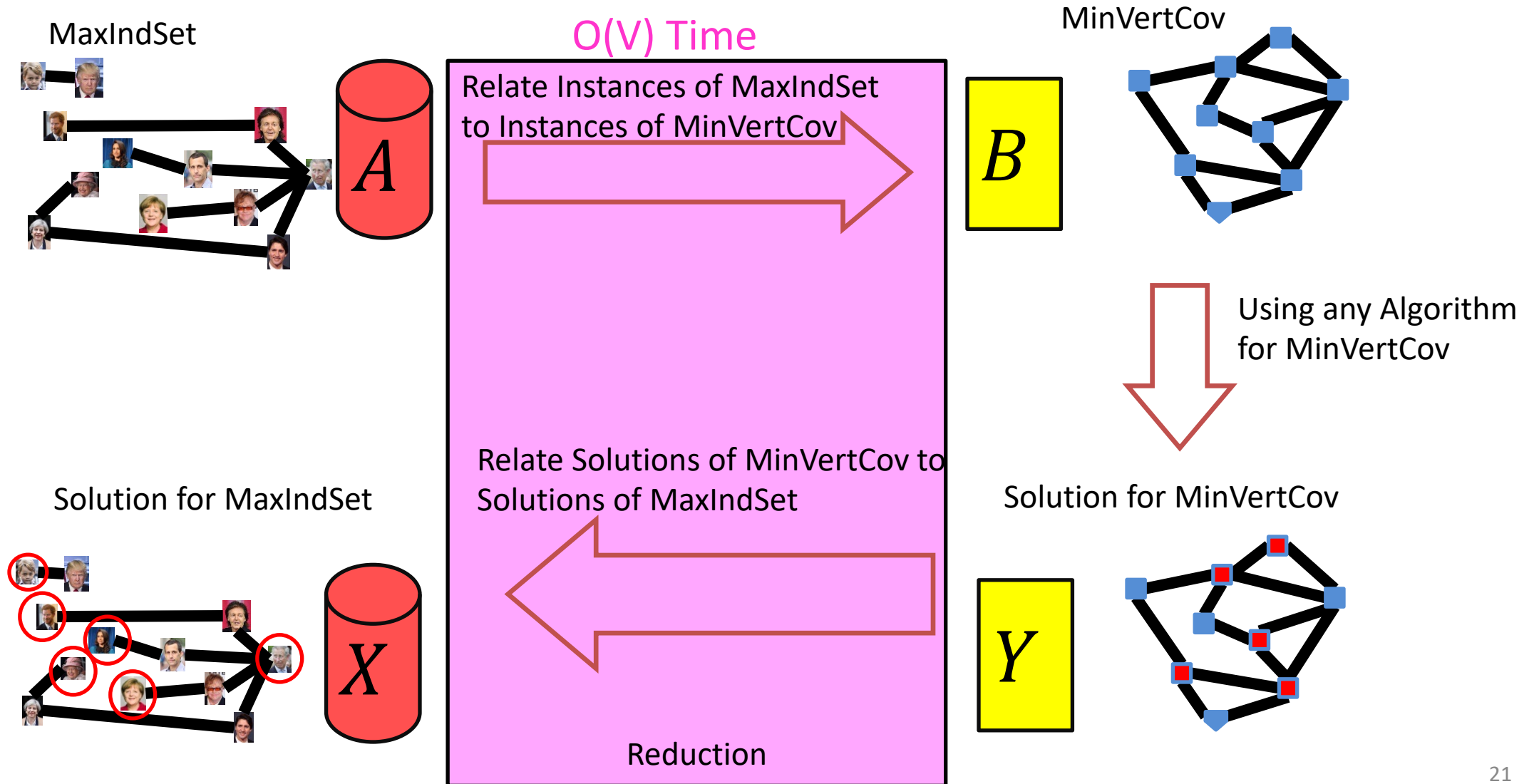
# MaxIndSet $\leq_V$ MinVertCov



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

$$A \leq_V B$$

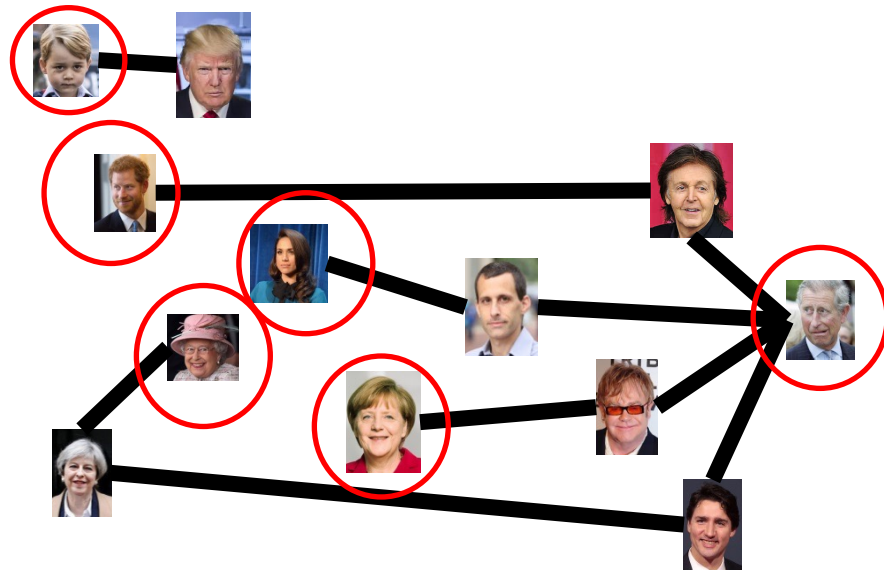
# We need to build this Reduction



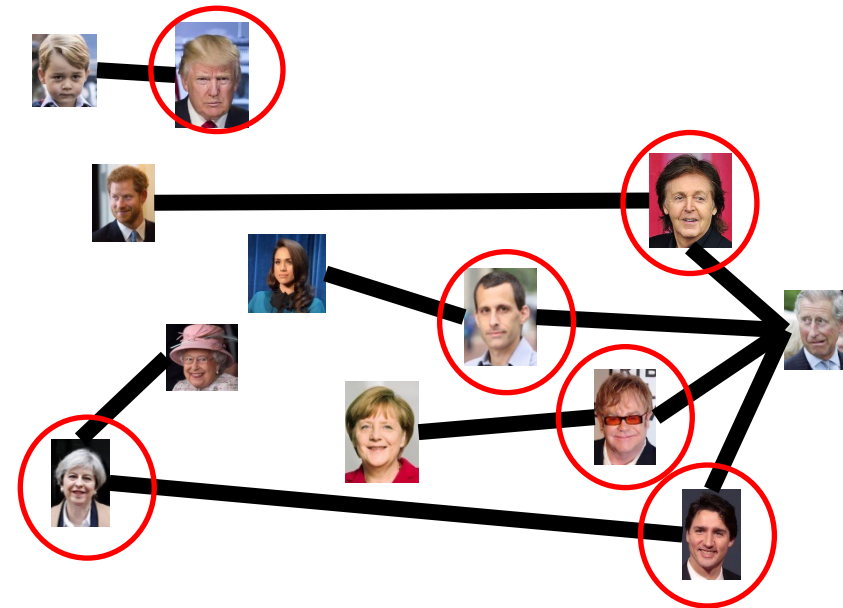
# Reduction Idea

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

Independent Set



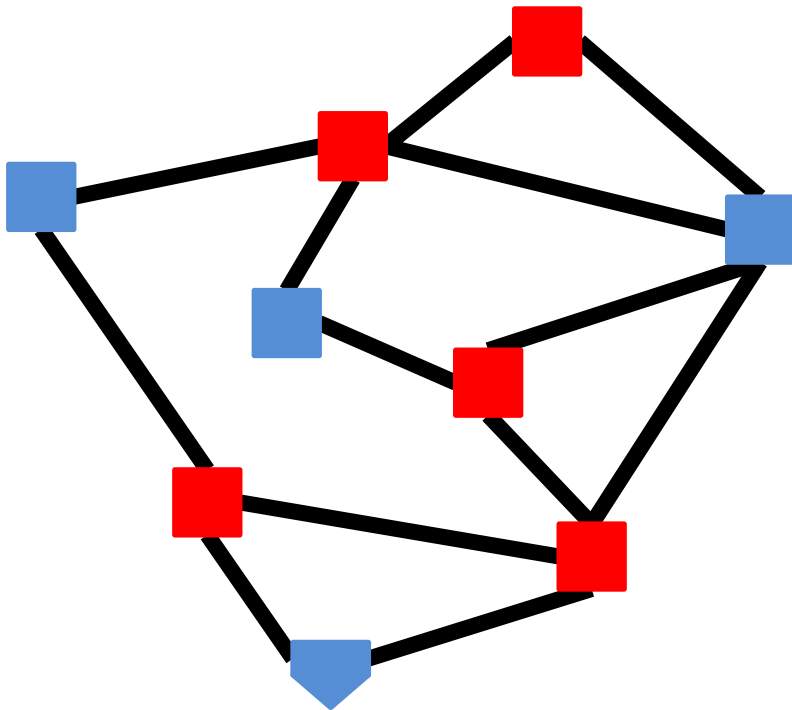
Vertex Cover



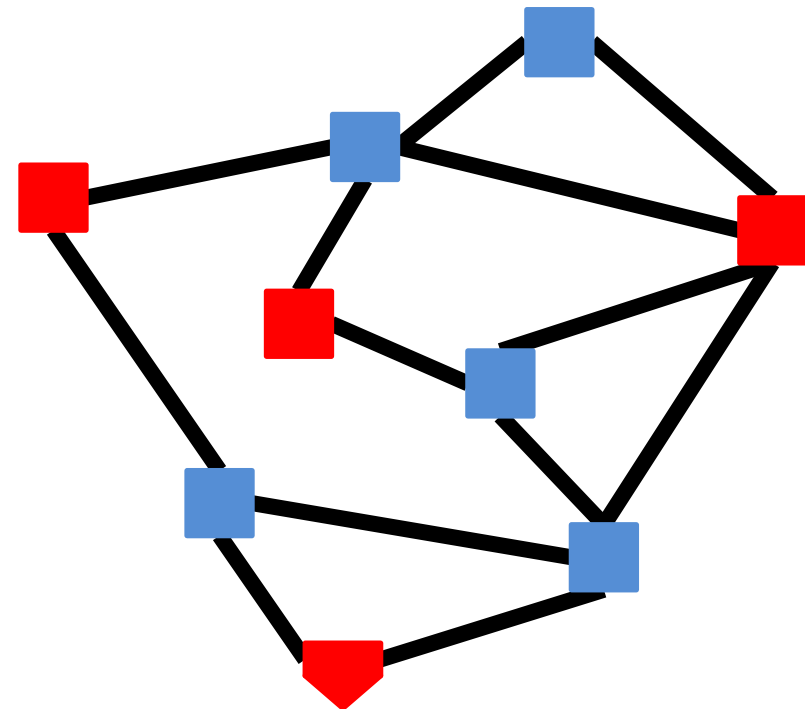
# Reduction Idea

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

Vertex Cover



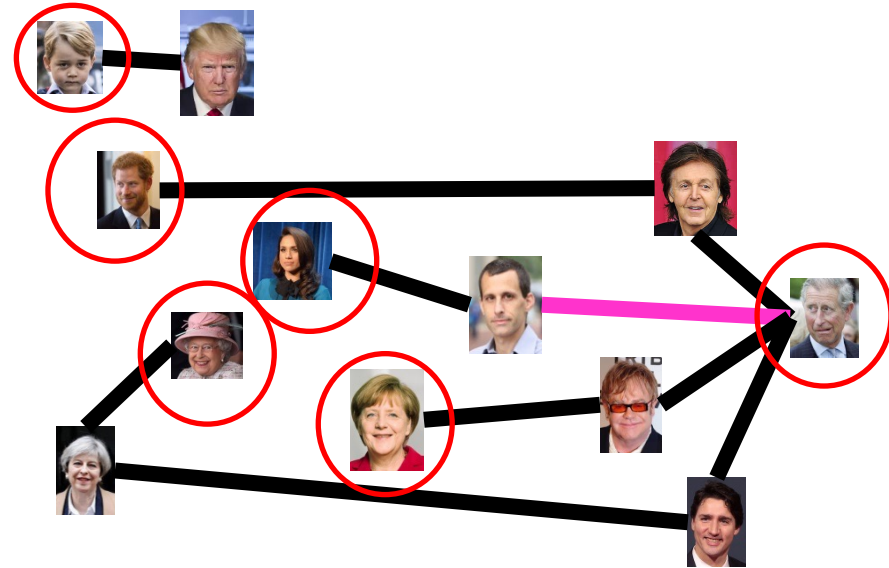
Independent Set



# 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 o.w.  $S$  would not be an independent set

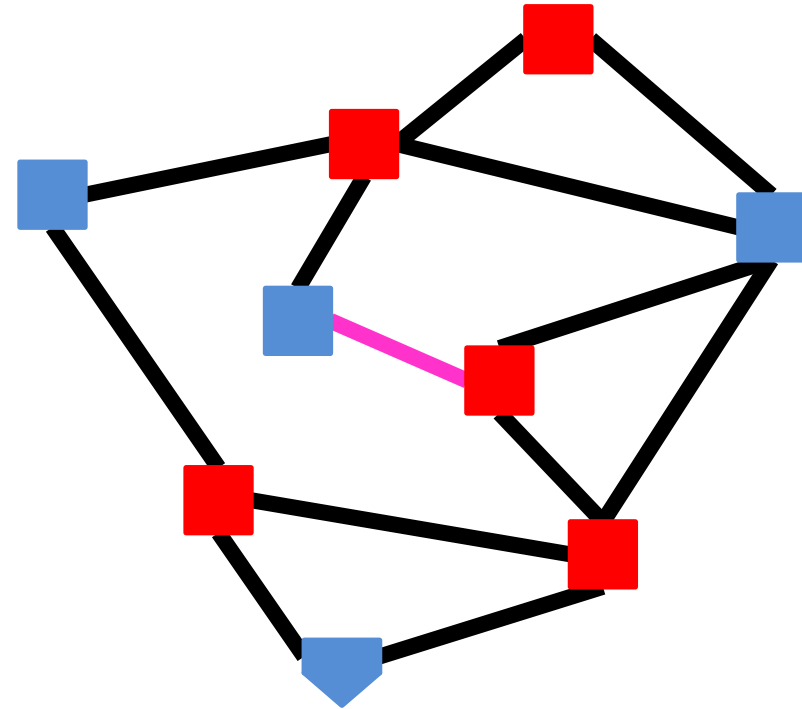
Therefore  $y \in V - S$ , so edge  $(x, y)$  is covered by  $V - S$



# 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



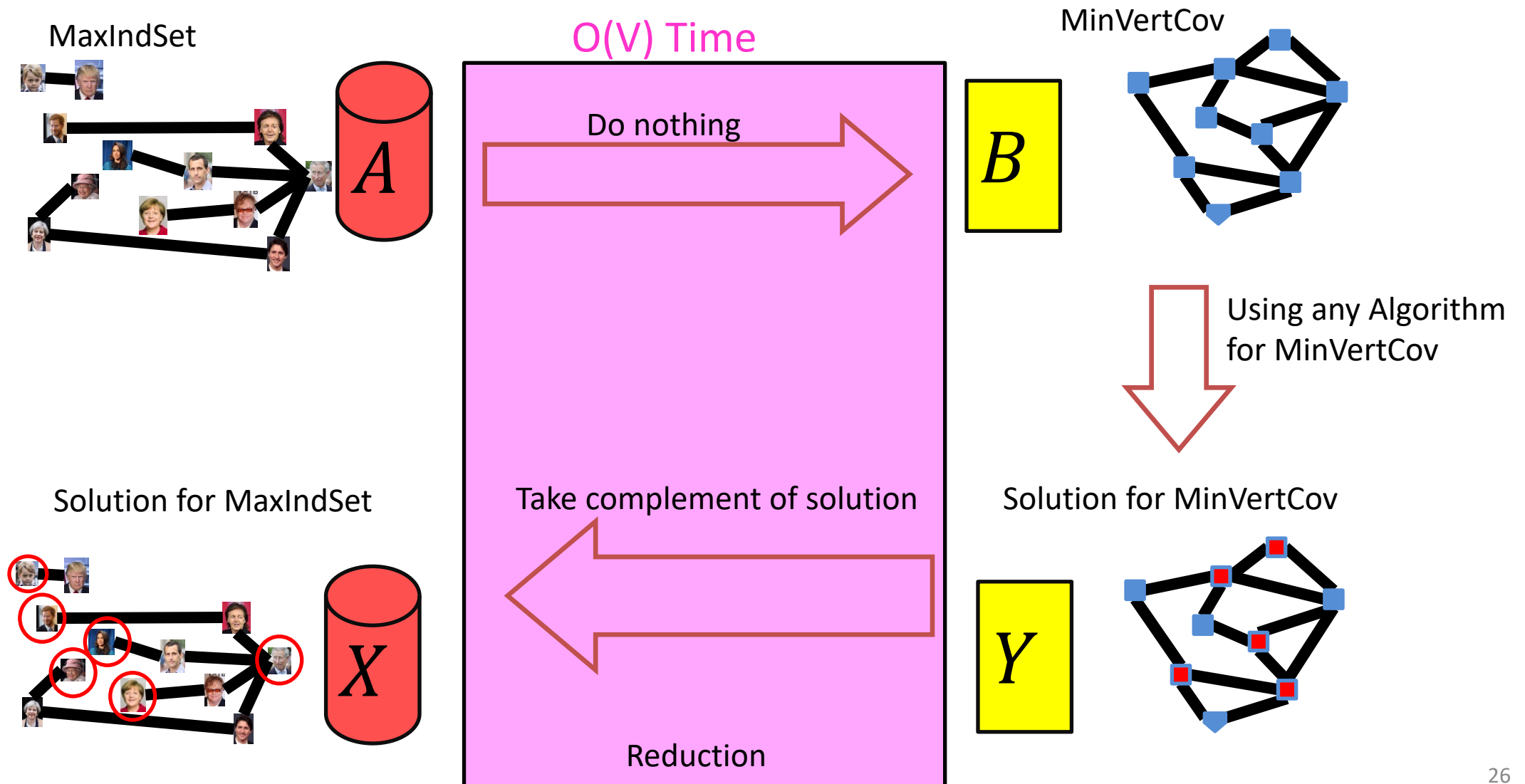
Consider any edge  $(x, y) \in E$

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  $S$ ,

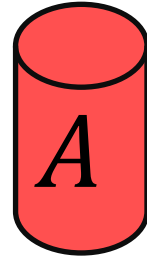
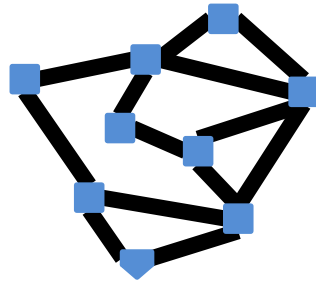
No edge has both end-nodes in  $S$ , thus  $S$  is an independent set

# MaxVertCov $V$ -Time Reducible to MinIndSet



# MaxVertCov $V$ -Time Reducible to MinIndSet

MinVertCov

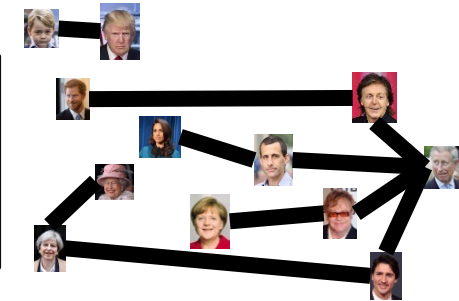


$O(V)$  Time

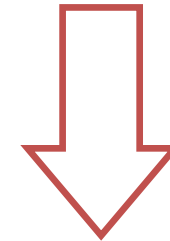
Do nothing



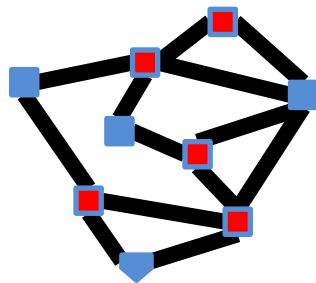
MaxIndSet



Using any Algorithm for MaxIndSet



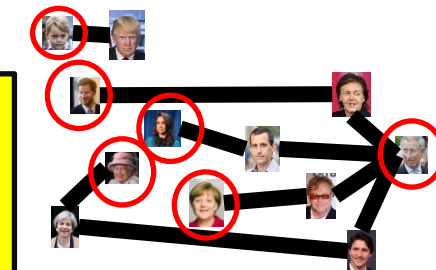
Solution for MinVertCov



Take complement of solution

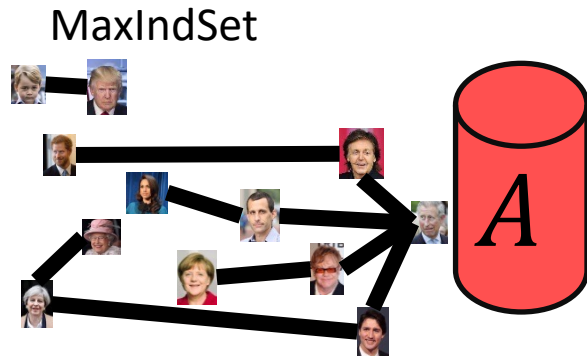


Solution for MaxIndSet



Reduction

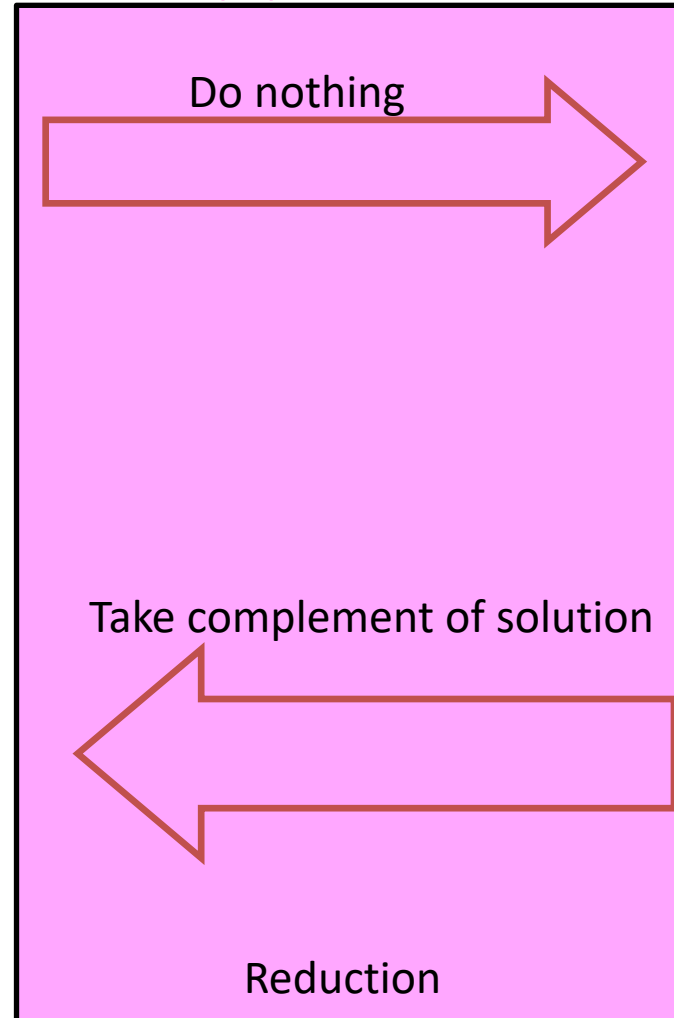
# Corollary



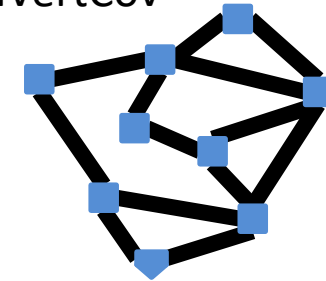
**If Solving  $A$  was always slow**



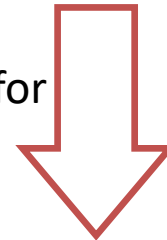
$O(V)$  Time



MinVertCov

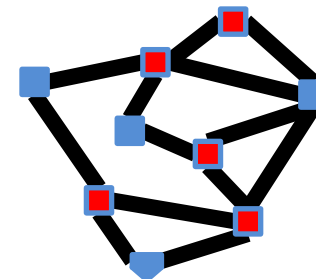


Using any Algorithm for MinIndSet



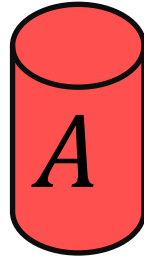
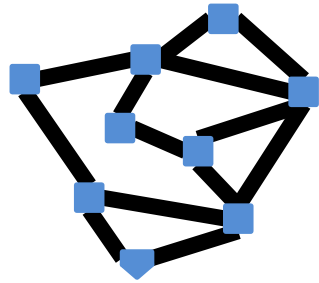
**Then this shows solving  $B$  is also slow**

Solution for MinVertCov



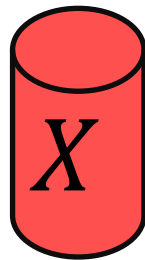
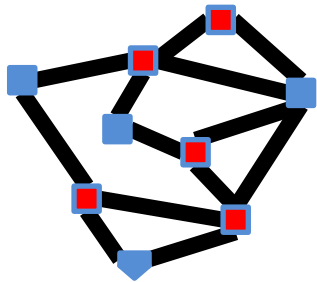
# Corollary

MinVertCov



**If Solving *A* was always slow**

Solution for MinVertCov

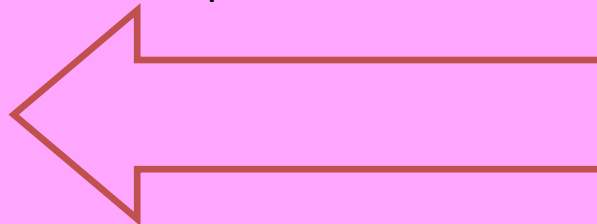


$O(V)$  Time

Do nothing

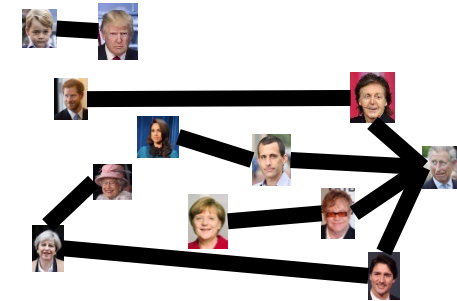
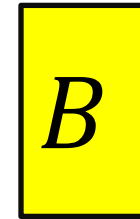


Take complement of solution



Reduction

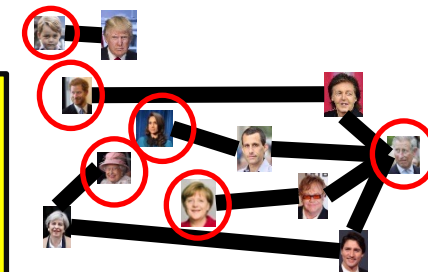
MaxIndSet



Using any Algorithm for MaxVertCov

**Then this shows solving *B* is also slow**

Solution for MaxIndSet



# 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
    - Next time!