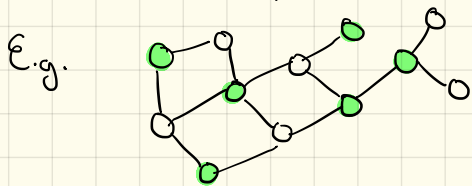


## Reduction Example: Independent Set $\rightarrow$ Vertex Cover

Def: A vertex cover of a graph is a set  $S$  of nodes such that each edge has at least one endpoint in  $S$

Intuitively, we try to cover all edges of the graph by choosing some set of endpoints



Here, the green nodes constitute a vertex cover of the graph



Problem: Given a graph  $G$  and a number  $k$ , does  $G$  contain a vertex cover of size at most  $k$ ?

$\uparrow$  smaller covers are harder to find

Problem (Independent Set): Given a graph  $G$  and a number  $k$ , does  $G$  contain a set of at least  $k$  independent vertices?

$\uparrow$  larger independent sets are harder to find

Can we reduce Independent Set to Vertex Cover?

# Relationship between Independent Set and vertex cover

Theorem: If  $G=(V,E)$  is a graph, then  $S \subseteq V$  is an independent set  $\iff V-S$  is a vertex cover.

Proof:  $\Rightarrow$  Suppose  $S$  is an independent set, and let  $e=(u,v)$  be some edge. Only one of  $u,v$  can be in  $S$ , hence at least one of  $u,v \in V-S$ . So  $V-S$  is a vertex cover.

Proof:  $\Leftarrow$  Suppose  $V-S$  is a vertex cover and let  $u,v \in S$ . There can't be an edge between  $u$  and  $v$  (otherwise that edge wouldn't be covered in  $V-S$ ). So,  $S$  is an independent set.

## Independent Set $\leq_p$ Vertex Cover

Given an arbitrary instance of Independent Set  $\langle G, k \rangle$

- Ask vertex cover algorithm if there is a vertex cover  $V-S$  of size  $\leq |V| - k$

By  $S$  is an IS iff  $V-S$  is a VC

If the VC algo said yes:  $S$  must be IS  $\geq k$   
no: There is no VC of size  $\leq |V| - k$ , hence no IS of size  $\geq k$ .

Actually, we also have  $VC \leq_p IS$

Reduction: To decide if  $G$  has a VC of size  $k$ , ask if it has an IS of size  $n - k$ .

So VC and IS are equivalently difficult.

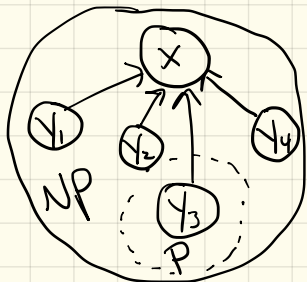
NP-completeness: Now, we can define what it means for a problem to be NP-complete.

Def: We say  $X$  is NP-complete if:

- 1)  $X \in NP$
- 2) For all  $Y \in NP$ ,  $Y \leq_p X$

If these hold, an algorithm for  $X$  could be used to solve all problems in NP.

So,  $X$  is at least as hard as any problem in NP.



Theorem: If  $X$  is NP-complete, then  $X$  is solvable in polynomial time  
iff  $P=NP$

Proof: IF  $P=NP$ ,  $X$  is solvable in polynomial time

Suppose  $X$  is solvable in poly-time, let  $Y$  be any problem in NP.  
We can solve  $Y$  in poly-time by reduction to  $X$ .

Therefore, every problem in NP would have a poly-time algo  
and we would have  $P=NP$ .

All of this relies on having some first NPC problem. Finding that  
first problem is the result of the Cook-Levin theorem  
(will mention briefly later).

For now, let's look at another reduction.

Problem (Set Cover): Given a universe (set)  $U$  of elements and a collection  $S_1, \dots, S_m$  of subsets of  $U$ , is there a collection of at most  $k$  of these subsets whose union equals  $U$ ?

Goal: Show that Set Cover is NP-complete. To show that we need to show

1) Set Cover  $\in$  NP

2) Some NP-complete problem reduces to Set Cover (we'll use vertex cover)

For 1, consider the collection of  $\leq k$  subsets as the certificate. Clearly, we can verify a set cover instance in poly-time.

Theorem: Vertex Cover  $\leq_p$  Set Cover

Proof: Let  $\langle G = (V, E), k \rangle$  be an arbitrary instance of Vertex Cover. Create the following instance of Set Cover.

-  $U = E$

- Create a subset  $S_i$  for all  $i \in V$  where  $S_i$  contains the edges adjacent to vertex  $i$ .

$U$  can be covered by  $\leq k$  sets iff  $G$  has a VC of size  $\leq k$ . Why?

$\Rightarrow$  Let  $S_1, \dots, S_j$  be a set cover of size  $\leq k$ . Then select vertices  $v_1, \dots, v_j$  in  $G$ . By the construction of our set cover instance, they constitute a VC of  $G$ . Since every  $u \in U$  is covered and  $U = E$  then every  $e \in E$  must be adjacent to a chosen vertex.

$\Leftarrow$  Let  $G$  have a vertex cover of size  $\leq k$ , and let the set of vertices be given by  $C^*$ . Since  $C^*$  is a VC, every  $e \in E$  is adjacent to some  $v \in C^*$ . However, by our reduction, we have  $U = E$  and we also have that for all  $e$  adjacent to  $i \in V$  then  $e \in S_i$  in our set cover instance. Hence  $\bigcup_{i \in C^*} S_i = U$ .

Summary: To show a problem is NP-complete, you must show it is in NP, and must reduce a known NP-complete problem to your new problem.

Some more problems

Boolean Formulas:

Variables:  $x_1, x_2, \dots$  (can be either true or false)

Terms:  $t_1, t_2, \dots, t_j$  is either  $x_j$  or  $\bar{x}_j$  (i.e. either  $x_j$  or not  $x_j$ )

Clauses:  $t_1 \vee t_2 \vee \dots \vee t_k$ : ( $\vee$  stands for "or"). a clause is true if any of its terms are true

E.g.  $(x_1 \vee \bar{x}_2)$ ,  $(\bar{x}_1 \vee \bar{x}_3)$ ,  $(x_2 \vee \bar{x}_3)$ ,  $(x_1 \vee x_2 \vee \bar{x}_3)$

Def: A truth assignment is a choice of true or false for each variable i.e. a function

$$v: X \rightarrow \{\text{true}, \text{false}\}$$

Def: A Conjunctive Normal Form (CNF) formula is a conjunction (and-ing) of clauses:

$$C_1 \wedge C_2 \wedge \dots \wedge C_k$$



Def: A truth assignment is a satisfying assignment for such a formula if it makes every clause true.

SAT and 3-SAT

Problem [Satisfiability (SAT)]: Given a set of clauses  $C_1, \dots, C_k$  over variables  $X = \{X_1, \dots, X_n\}$ , is there a satisfying assignment?

Problem [3-SAT]: Given a set of clauses  $C_1, \dots, C_k$ , each of length 3 (i.e. containing 3 terms), over variables  $X = \{X_1, \dots, X_n\}$ , is there a satisfying assignment?

Cook-Levin Theorem shows that SAT is NP-complete.

Richard Karp showed (1972) that  $SAT \leq_P 3-SAT$ . He, in fact, shows via reduction, the NP-completeness of 21 different problems.

The Garey and Johnson text "Computers and Intractability" shows > 300 NP-complete problems.

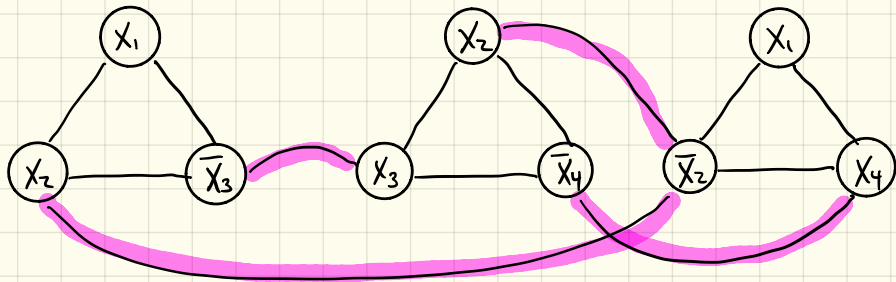
The CL theorem gives us the first "hook" on which to hang new NPC proofs (reductions).


Another (non-covering) reduction.

Theorem: 3-SAT  $\leq_p$  Independent Set

Proof: Consider the following mapping from clauses in a 3-SAT instance to a graph

$$(X_1 \vee X_2 \vee \bar{X}_3) \wedge (X_2 \vee X_3 \vee \bar{X}_4) \wedge (X_1 \vee \bar{X}_2 \vee X_4)$$




The  edges are called "conflict links". We just draw them like this; in  $G$  they are regular edges.

That is, we create a triangle for each clause where the vertices are labeled with the terms and there are edges between all terms in a clause. Additionally, we add an edge between each vertex labeled with a term and each instance of a vertex labeled with the negation of that term (e.g.  $X_2 - \bar{X}_2$ )

Claim: This graph has an IS of size  $\geq k$  iff the formula is satisfiable.

Proof:  $\Rightarrow$  If formula is satisfiable, there is at least 1 true literal in each clause. Let  $S$  be the set of one such literal from each clause.  $|S|=k$  and no two nodes in  $S$  are connected by an edge.

$\Leftarrow$  If the graph has an IS  $S$  of size  $k$ , we know that it has 1 node from each "clause triangle" (since we can have at most 1 node chosen from each fully connected triangle in  $S$ ). Set those terms to true. This is possible because no two terms in  $S$  are negations of each other (because of conflict links). 

General Proof Strategy for showing a problem is NP-complete:

- 1) Show  $X \in NP$  by finding an efficient certifier
- 2) look for some known NP-complete problem (there are many)  $Y$  that seems "similar" to your problem  $X$  in some way.

3) Show that  $Y \leq_p X$

One way to show  $Y \leq_p X$ :

1) Let  $I_Y$  be an arbitrary instance of problem  $Y$

2) Show how to construct an instance  $I_X$  of your problem  $X$  in polynomial time such that:

- if  $I_Y \in Y$  then  $I_X \in X$
  - if  $I_X \in X$  then  $I_Y \in Y$
- } Need both

Striking dichotomy between which problems are NP-complete vs in P

NP-Complete	in P
3-SAT	2-SAT
TSP	MST
Longest Path	Shortest Path
3D matching	Bipartite Matching
Knapsack	Unary Knapsack
Independent Set	Independent Set on trees
Integer Linear Programming	Linear Programming
Hamiltonian Path	Eulerian Path
Balanced Cut	Minimum Cut