

# *CMSC 451: Reductions & NP-completeness*

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Based on Section 8.1 of *Algorithm Design* by Kleinberg & Tardos.

# Reductions as tool for hardness

We want prove some problems are computationally difficult.

As a first step, we settle for relative judgements:

Problem  $X$  is at least as hard as problem  $Y$

To prove such a statement, we **reduce** problem  $Y$  to problem  $X$ :

*If you had a black box that can solve instances of problem  $X$ , how can you solve any instance of  $Y$  using polynomial number of steps, plus a polynomial number of calls to the black box that solves  $X$ ?*

# Polynomial Reductions

- If problem  $Y$  can be reduced to problem  $X$ , we denote this by  $Y \leq_P X$ .
- This means “ $Y$  is polynomial-time reducible to  $X$ .”
- It also means that  $X$  is at least as hard as  $Y$  because if you can solve  $X$ , you can solve  $Y$ .
- Note: We reduce *to* the problem we want to show is the harder problem.

# Polynomial Problems

Suppose:

- $Y \leq_P X$ , and
- there is an polynomial time algorithm for  $X$ .

Then, there is a polynomial time algorithm for  $Y$ .

Why?

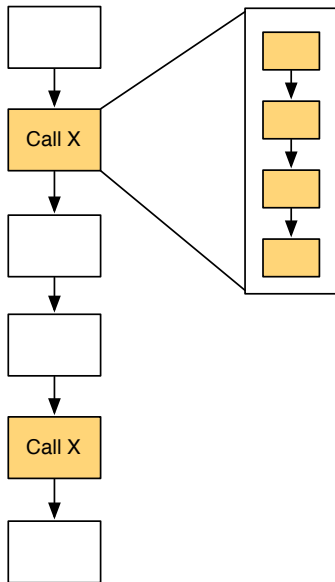
# Polynomial Problems

Suppose:

- $Y \leq_P X$ , and
- there is a polynomial time algorithm for  $X$ .

Then, there is a polynomial time algorithm for  $Y$ .

**Why?** Because polynomials compose.



# We've Seen Reductions Before

Examples of Reductions:

- MAX BIPARTITE MATCHING  $\leq_P$  MAX NETWORK FLOW.
- IMAGE SEGMENTATION  $\leq_P$  MIN-CUT.
- SURVEY DESIGN  $\leq_P$  MAX NETWORK FLOW.
- DISJOINT PATHS  $\leq_P$  MAX NETWORK FLOW.

# Reductions for Hardness

## Theorem

*If  $Y \leq_P X$  and  $Y$  cannot be solved in polynomial time, then  $X$  cannot be solved in polynomial time.*

Why? If we *could* solve  $X$  in polynomial time, then we'd be able to solve  $Y$  in polynomial time using the reduction, contradicting the assumption.

So: If we could find one hard problem  $Y$ , we could prove that another problem  $X$  is hard by reducing  $Y$  to  $X$ .

# Vertex Cover

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

In other words, we try to “cover” each of the edges by choosing at least one of its vertices.

## Vertex Cover

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



# Independent Set to Vertex Cover

## Independent Set

Given graph  $G$  and a number  $k$ , does  $G$  contain a set of at least  $k$  independent vertices?

Can we reduce independent set to vertex cover?

## Vertex Cover

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

# Relation btw Vertex Cover and Indep. Set

## Theorem

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

*Proof.*  $\implies$  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  $u, v$  is in  $V - S$ . So,  $V - S$  is a vertex cover.

$\impliedby$  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.  $\square$

# Independent Set $\leq_P$ Vertex Cover

## Independent Set $\leq_P$ Vertex Cover

To show this, we change any instance of Independent Set into an instance of Vertex Cover:

- Given an instance of Independent Set  $\langle G, k \rangle$ ,
- We ask our Vertex Cover black box if there is a vertex cover  $V - S$  of size  $\leq |V| - k$ .

By our previous theorem,  $S$  is an independent set iff  $V - S$  is a vertex cover. If the Vertex Cover black box said:

*yes: then  $S$  must be an independent set of size  $\geq k$ .*

*no: then there is no vertex cover  $V - S$  of size  $\leq |V| - k$ , hence there is no independent set of size  $\geq k$ .*

# Vertex Cover $\leq_P$ Independent Set

Actually, we also have:

Vertex Cover  $\leq_P$  Independent Set

*Proof.* To decide if  $G$  has an vertex cover of size  $k$ , we ask if it has an independent set of size  $n - k$ .  $\square$

So: VERTEX COVER and INDEPENDENT SET are equivalently difficult.

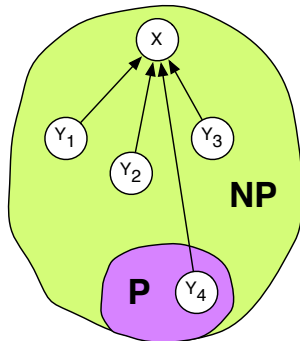
# NP-completeness

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

- $X \in \mathbf{NP}$
- for all  $Y \in \mathbf{NP}$ ,  $Y \leq_P X$ .

If these hold, then  $X$  can be used to solve every problem in **NP**.

Therefore,  $X$  is definitely at least as hard as every problem in **NP**.



# NP-completeness and $P=NP$

## Theorem

*If  $X$  is NP-complete, then  $X$  is solvable in polynomial time if and only if  $P = NP$ .*

*Proof.* If  $P = NP$ , then  $X$  can be solved in polytime.

Suppose  $X$  is solvable in polytime, and let  $Y$  be any problem in  $NP$ . We can solve  $Y$  in polynomial time: reduce it to  $X$ .

Therefore, every problem in  $NP$  has a polytime algorithm and  $P = NP$ .

# Reductions and NP-completeness

## Theorem

*If  $Y$  is NP-complete, and*

①  *$X$  is in NP*

②  *$Y \leq_P X$*

*then  $X$  is NP-complete.*

In other words, we can prove a new problem is NP-complete by reducing some other NP-complete problem to it.

*Proof.* Let  $Z$  be any problem in **NP**. Since  $Y$  is NP-complete,  $Z \leq_P Y$ . By assumption,  $Y \leq_P X$ . Therefore:  $Z \leq_P Y \leq_P X$ .  $\square$

# Some First NP-complete problem

We need to find some first NP-complete problem.

Finding the first NP-complete problem was the result of the Cook-Levin theorem.

We'll deal with this later. For now, trust me that:

- Independent Set is a *packing problem* and is NP-complete.
- Vertex Cover is a *covering problem* and is NP-complete.



# Set Cover

Another very general and useful covering problem:

## Set Cover

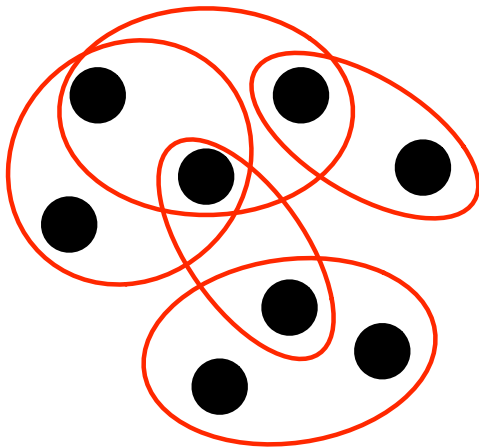
Given a set  $U$  of elements and a collection  $S_1, \dots, S_m$  of subsets of  $U$ , is there a collection of at most  $k$  of these sets whose union equals  $U$ ?

We will show that

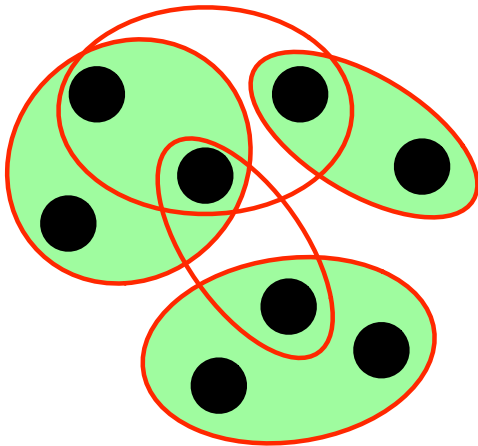
$$\begin{aligned} \text{SET COVER} &\in NP \\ \text{VERTEX COVER} &\leq_P \text{SET COVER} \end{aligned}$$

And therefore that SET COVER is NP-complete.

## Set Cover, Figure



## Set Cover, Figure



# Vertex Cover $\leq_P$ Set Cover

**Thm.** Vertex Cover  $\leq_P$  Set Cover

*Proof.* Let  $G = (V, E)$  and  $k$  be an instance of VERTEX COVER.  
Create an instance of SET COVER:

- $U = E$
- Create a  $S_u$  for each  $u \in V$ , where  $S_u$  contains the edges adjacent to  $u$ .

$U$  can be covered by  $\leq k$  sets iff  $G$  has a vertex cover of size  $\leq k$ .

Why? If  $k$  sets  $S_{u_1}, \dots, S_{u_k}$  cover  $U$  then every edge is adjacent to at least one of the vertices  $u_1, \dots, u_k$ , yielding a vertex cover of size  $k$ .

If  $u_1, \dots, u_k$  is a vertex cover, then sets  $S_{u_1}, \dots, S_{u_k}$  cover  $U$ .  $\square$

## Last Step:

We still have to show that Set Cover is in **NP**!

The certificate is a list of  $k$  sets from the given collection.

We can check in polytime whether they cover all of  $U$ .

Since we have a certificate that can be checked in polynomial time, Set Cover is in **NP**.

# Summary

You can prove a problem is NP-complete by reducing a known NP-complete problem to it.

We know the following problems are NP-complete:

- Vertex Cover
- Independent Set
- Set Cover

Warning: You should reduce the *known* NP-complete problem to the problem you are interested in. (You *will* mistakenly do this backwards sometimes.)

# *CMSC 451: SAT, Coloring, Hamiltonian Cycle, TSP*

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Based on Sects. 8.2, 8.7, 8.5 of *Algorithm Design* by Kleinberg & Tardos.

# Boolean Formulas

## Boolean Formulas:

**Variables:**  $x_1, x_2, x_3$  (can be either **true** or **false**)

**Terms:**  $t_1, t_2, \dots, t_\ell$ :  $t_j$  is either  $x_i$  or  $\bar{x}_i$   
(meaning either  $x_i$  or **not**  $x_i$ ).

**Clauses:**  $t_1 \vee t_2 \vee \dots \vee t_\ell$  ( $\vee$  stands for “OR”)  
A clause is **true** if any term in it is **true**.

**Example 1:**  $(x_1 \vee \bar{x}_2), (\bar{x}_1 \vee \bar{x}_3), (x_2 \vee \bar{x}_3)$

**Example 2:**  $(x_1 \vee x_2 \vee \bar{x}_3), (\bar{x}_2 \vee x_1)$



# Boolean Formulas

**Def.** A **truth assignment** is a choice of **true** or **false** for each variable, ie, a function  $v : X \rightarrow \{\mathbf{true}, \mathbf{false}\}$ .

**Def.** A CNF formula is a conjunction of clauses:

$$C_1 \wedge C_2 \wedge \cdots \wedge C_k$$

**Example:**  $(x_1 \vee \bar{x}_2) \wedge (\bar{x}_1 \vee \bar{x}_3) \wedge (x_2 \vee \bar{v}_3)$

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

# SAT and 3-SAT

## 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?

## Satisfiability (3-SAT)

Given a set of clauses  $C_1, \dots, C_k$ , **each of length 3**, over variables  $X = \{x_1, \dots, x_n\}$  is there a satisfying assignment?

# Cook-Levin Theorem

## Theorem (Cook-Levin)

*3-SAT is NP-complete.*

Proven in early 1970s by Cook. Slightly different proof by Levin independently.

**Idea of the proof:** encode the workings of a Nondeterministic Turing machine for an instance  $I$  of problem  $X \in \mathbf{NP}$  as a SAT formula so that the formula is satisfiable if and only if the nondeterministic Turing machine would accept instance  $I$ .

We won't have time to prove this, but it gives us our first hard problem.

# Reducing 3-SAT to Independent Set

**Thm.**  $3\text{-SAT} \leq_P \text{Independent Set}$

*Proof.* Suppose we have an algorithm to solve Independent Set, how can we use it to solve 3-SAT?

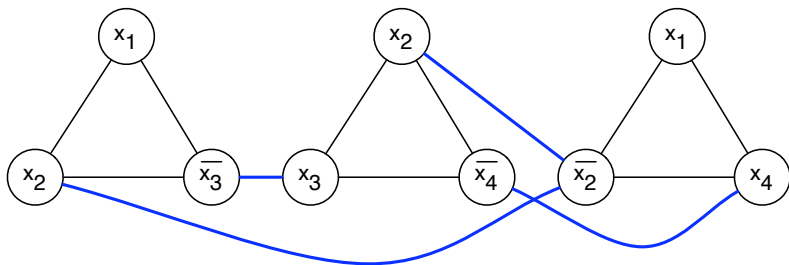
To solve 3-SAT:

- you have to choose a term from each clause to set to **true**,
- but you can't set both  $x_i$  and  $\bar{x}_i$  to **true**.

How do we do the reduction?

## 3-SAT $\leq_P$ Independent Set

$$(x_1 \vee x_2 \vee \bar{x}_3) \wedge (x_2 \vee x_3 \vee \bar{x}_4) \wedge (x_1 \vee \bar{x}_2 \vee x_4)$$



# Proof

## Theorem

*This graph has an independent set of size  $k$  iff the formula is satisfiable.*

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

$\implies$  If the graph has an independent set  $S$  of size  $k$ , we know that it has one node from each “clause triangle.” Set those terms to **true**. This is possible because no 2 are negations of each other.  $\square$

## Graph Coloring

# Graph Coloring Problem

## Graph Coloring Problem

Given a graph  $G$ , can you color the nodes with  $\leq k$  colors such that the endpoints of every edge are colored differently?

Notation: A  $k$ -coloring is a function  $f : V \rightarrow \{1, \dots, k\}$  such that for every edge  $\{u, v\}$  we have  $f(u) \neq f(v)$ .

If such a function exists for a given graph  $G$ , then  $G$  is  **$k$ -colorable**.



## Special case of $k = 2$

How can we test if a graph has a 2-coloring?

## Special case of $k = 2$

How can we test if a graph has a 2-coloring?

Check if the graph is bipartite.

Unfortunately, for  $k \geq 3$ , the problem is NP-complete.

### Theorem

*3-Coloring is NP-complete.*

# Graph Coloring is NP-complete

3-Coloring  $\in$  **NP**: A valid coloring gives a certificate.

We will show that:

$$3\text{-SAT} \leq_P 3\text{-Coloring}$$

Let  $x_1, \dots, x_n, C_1, \dots, C_k$  be an instance of 3-SAT.

We show how to use 3-Coloring to solve it.

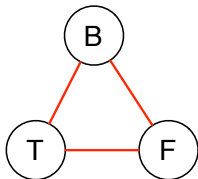
# Reduction from 3-SAT

We construct a graph  $G$  that will be 3-colorable iff the 3-SAT instance is satisfiable.

For every variable  $x_i$ , create 2 nodes in  $G$ , one for  $x_i$  and one for  $\bar{x}_i$ . Connect these nodes by an edge:

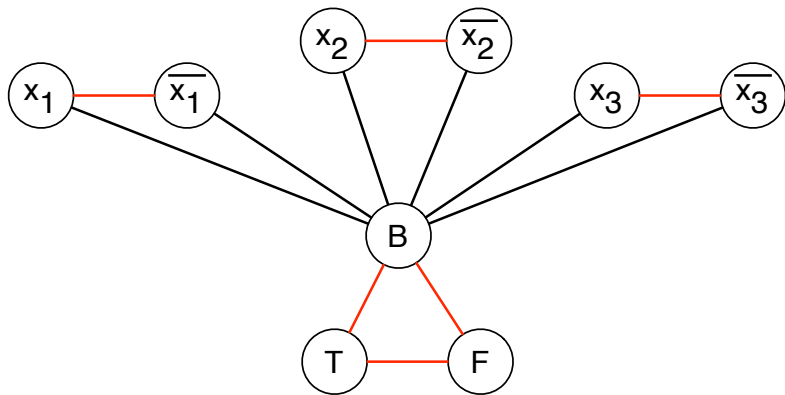


Create 3 *special nodes* T, F, and B, joined in a triangle:



# Connecting them up

Connect every variable node to B:



# Properties

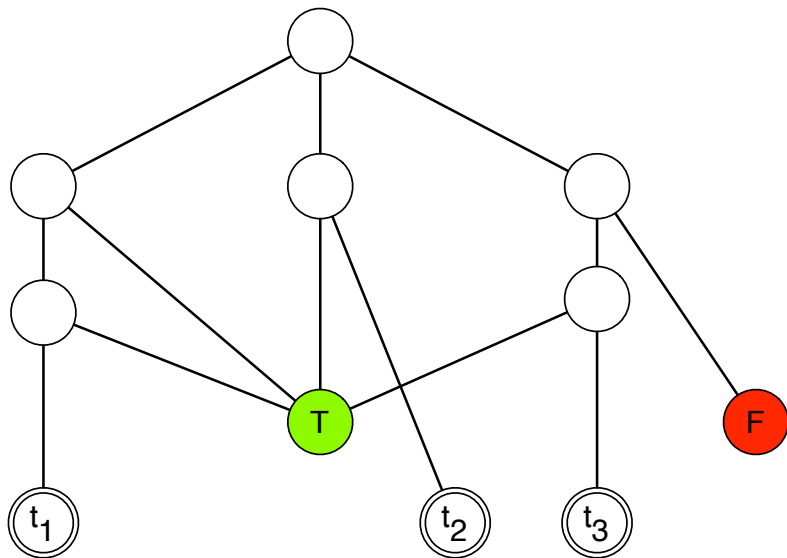
## Properties:

- Each of  $x_i$  and  $\bar{x}_i$  must get different colors
- Each must be different than the color of B.
- B, T, and F must get different colors.

Hence, any 3-coloring of this graph defines a valid truth assignment!

Still have to constrain the truth assignments to satisfy the given clauses, however.

Connect Clause  $(t_1, t_2, t_3)$  up like this:

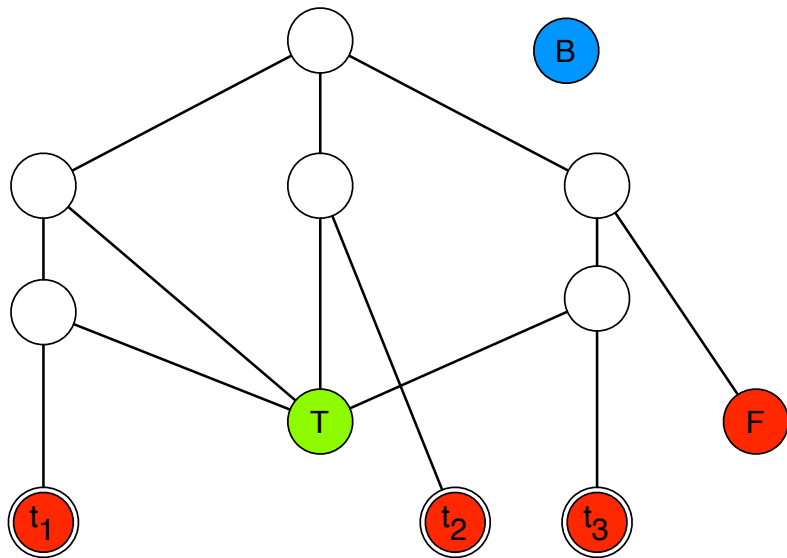


# Suppose Every Term Was False

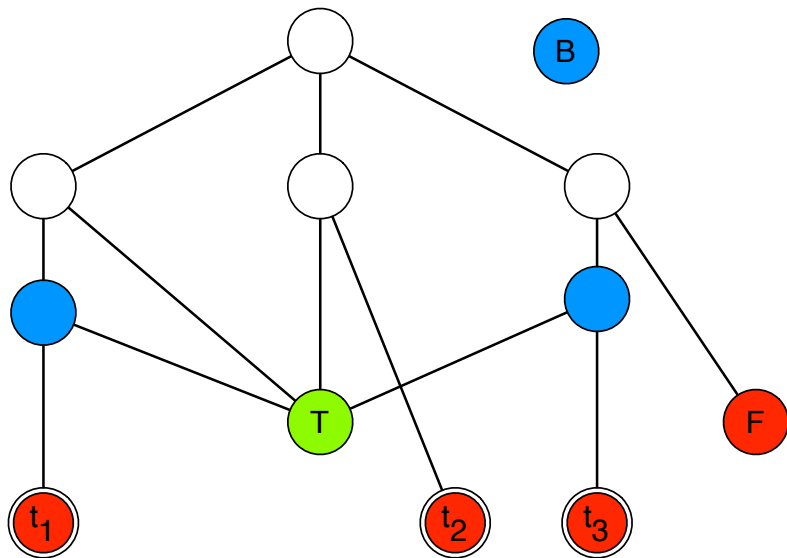
What if every term in the clause was assigned the **false** color?



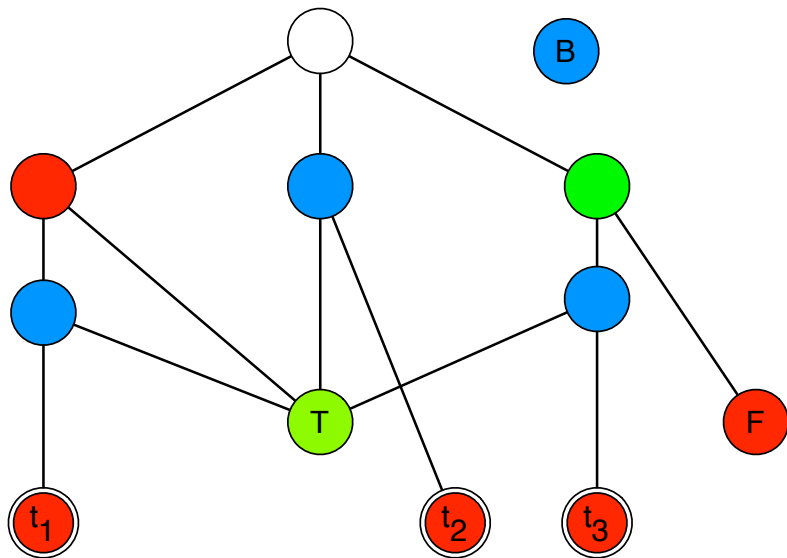
Connect Clause  $(t_1, t_2, t_3)$  up like this:



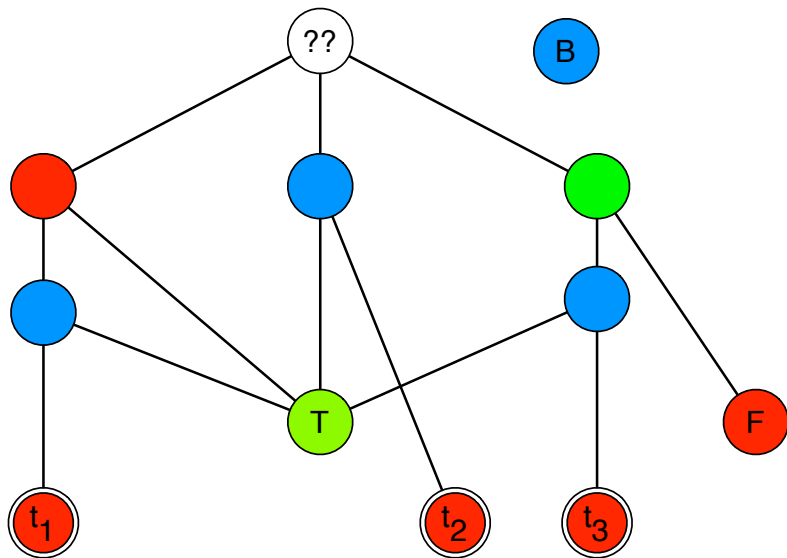
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Connect Clause  $(t_1, t_2, t_3)$  up like this:



Connect Clause  $(t_1, t_2, t_3)$  up like this:



# Suppose there is a 3-coloring

Top node is colorable iff one of its terms gets the **true** color.

Suppose there is a 3-coloring.

We get a satisfying assignment by:

- Setting  $x_i = \mathbf{true}$  iff  $v_i$  is colored the same as T

Let  $C$  be any clause in the formula. At least 1 of its terms must be true, because if they were all false, we couldn't complete the coloring (as shown above).

# Suppose there is a satisfying assignment

Suppose there is a satisfying assignment.

We get a 3-coloring of  $G$  by:

- Coloring  $T$ ,  $F$ ,  $B$  arbitrarily with 3 different colors
- If  $x_i = \mathbf{true}$ , color  $v_i$  with the same color as  $T$  and  $\bar{v}_i$  with the color of  $F$ .
- If  $x_i = \mathbf{false}$ , do the opposite.
- Extend this coloring into the clause gadgets.

Hence: the graph is 3-colorable iff the formula it is derived from is satisfiable.

# General Proof Strategy

## General Strategy for Proving Something is NP-complete:

- 1 Must show that  $X \in \mathbf{NP}$ . Do this by showing there is an certificate that can be efficiently checked.
- 2 Look at some problems that are known to be NP-complete (there are thousands), and choose one  $Y$  that seems “similar” to your problem in some way.
- 3 Show that  $Y \leq_P X$ .

# Strategy for Showing $Y \leq_P X$

One strategy for showing that  $Y \leq_P X$  often works:

- 1 Let  $I_Y$  be any instance of problem  $Y$ .
- 2 Show how to construct an instance  $I_X$  of problem  $X$  in polynomial time such that:
  - If  $I_Y \in Y$ , then  $I_X \in X$
  - If  $I_X \in X$ , then  $I_Y \in Y$



## Hamiltonian Cycle

# Hamiltonian Cycle Problem

## Hamiltonian Cycle

Given a directed graph  $G$ , is there a cycle that visits every vertex exactly once?

Such a cycle is called a **Hamiltonian cycle**.

# Hamiltonian Cycle is NP-complete

## Theorem

*Hamiltonian Cycle is NP-complete.*

*Proof.* First,  $\text{HamCycle} \in \text{NP}$ . Why?

Second, we show  $3\text{-SAT} \leq_P \text{Hamiltonian Cycle}$ .

Suppose we have a black box to solve Hamiltonian Cycle, how do we solve 3-SAT?

In other words: how do we encode an instance  $I$  of 3-SAT as a graph  $G$  such that  $I$  is satisfiable exactly when  $G$  has a Hamiltonian cycle.

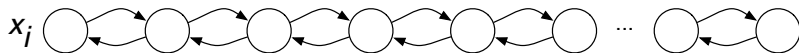
Consider an instance  $I$  of 3-SAT, with variables  $x_1, \dots, x_n$  and clauses  $C_1, \dots, C_k$ .

# Reduction Idea

## Reduction Idea (very high level):

- Create some graph structure (a “gadget”) that represents the variables
- And some graph structure that represents the clauses
- Hook them up in some way that encodes the formula
- Show that this graph has a Ham. cycle iff the formula is satisfiable.

# Gadget Representing the Variables



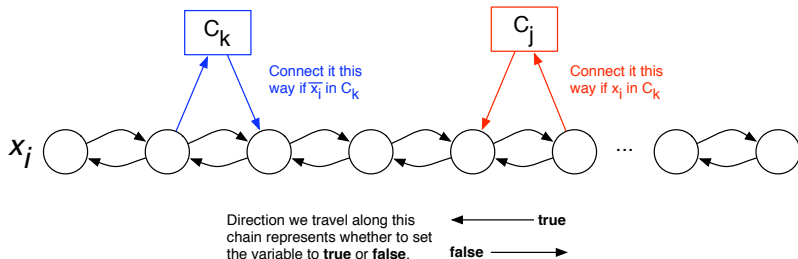
Direction we travel along this chain represents whether to set the variable to **true** or **false**.

← **true**

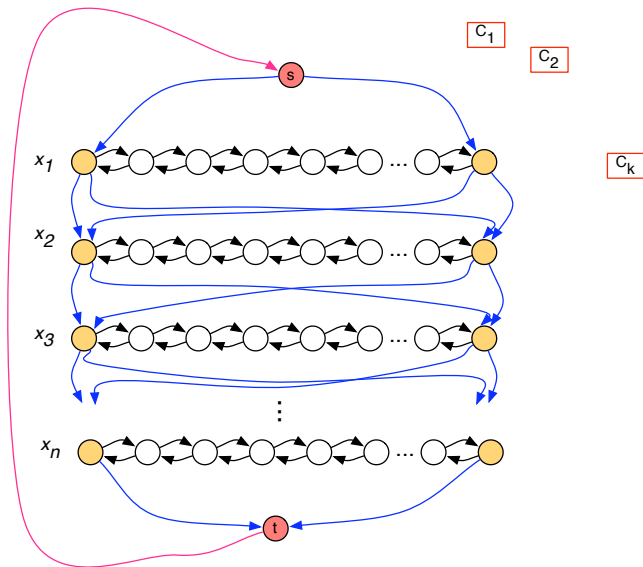
**false** →

# Hooking in the Clauses

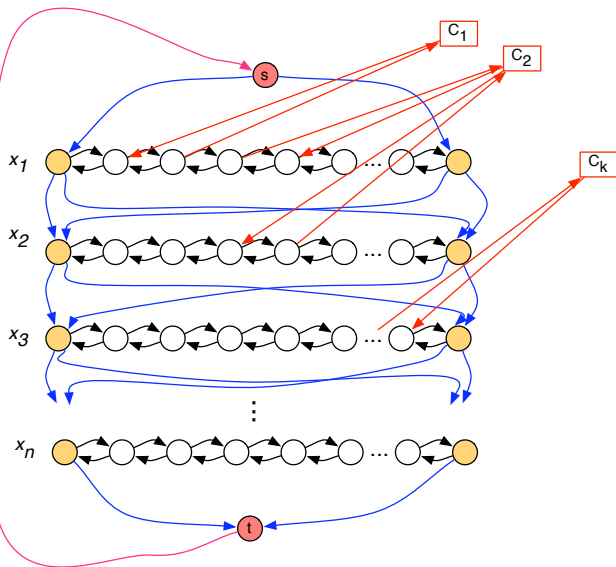
Add a new node for each clause:



# Connecting up the paths



# Connecting up the paths





# Hamiltonian Cycle is NP-complete

- A Hamiltonian path encodes a truth assignment for the variables (depending on which direction each chain is traversed)
- For there to be a Hamiltonian cycle, we have to visit every clause node
- We can only visit a clause if we satisfy it (by setting one of its terms to true)
- Hence, if there is a Hamiltonian cycle, there is a satisfying assignment

# Hamiltonian Path

**Hamiltonian Path:** Does  $G$  contain a **path** that visits every node exactly once?

How could you prove this problem is NP-complete?

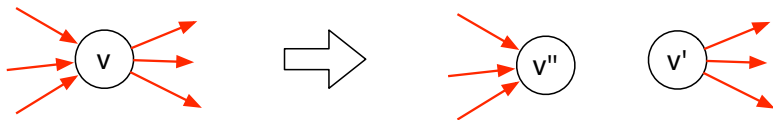
# Hamiltonian Path

**Hamiltonian Path:** Does  $G$  contain a **path** that visits every node exactly once?

How could you prove this problem is NP-complete?

Reduce Hamiltonian Cycle to Hamiltonian Path.

Given instance of Hamiltonian Cycle  $G$ , choose an arbitrary node  $v$  and split it into two nodes to get graph  $G'$ :



Now any Hamiltonian Path must start at  $v'$  and end at  $v''$ .

# Hamiltonian Path

$G''$  has a Hamiltonian Path  $\iff G$  has a Hamiltonian Cycle.

$\implies$  If  $G''$  has a Hamiltonian Path, then the same ordering of nodes (after we glue  $v'$  and  $v''$  back together) is a Hamiltonian cycle in  $G$ .

$\impliedby$  If  $G$  has a Hamiltonian Cycle, then the same ordering of nodes is a Hamiltonian path of  $G'$  if we split up  $v$  into  $v'$  and  $v''$ .  $\square$

Hence, Hamiltonian Path is NP-complete.

# Traveling Salesman Problem

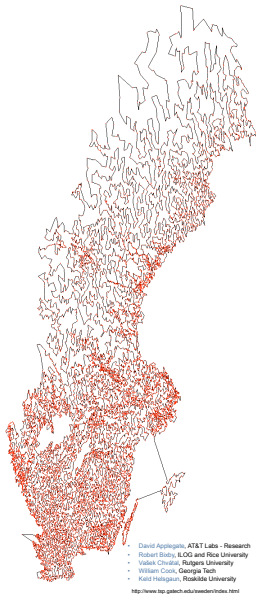
## Traveling Salesman Problem

Given  $n$  cities, and distances  $d(i, j)$  between each pair of cities, does there exist a path of length  $\leq k$  that visits each city?

### Notes:

- We have a distance between every pair of cities.
- In this version,  $d(i, j)$  doesn't have to equal  $d(j, i)$ .
- And the distances don't have to obey the triangle inequality ( $d(i, j) \leq d(i, k) + d(k, j)$  for all  $i, j, k$ ).

# TSP large instance



- TSP visiting 24,978 (all) cities in Sweden.
- Solved by David Applegate, Robert Bixby, Václav Chvátal, William Cook, and Keld Helsgaun
- <http://www.tsp.gatech.edu/sweden/index.html>
- Lots more cool TSP at <http://www.tsp.gatech.edu/>

# Traveling Salesman is NP-complete

**Thm.** Traveling Salesman is NP-complete.

TSP seems a lot like Hamiltonian Cycle. We will show that

$$\text{HAMILTONIAN CYCLE} \leq_P \text{TSP}$$

To do that:

**Given:** a graph  $G = (V, E)$  that we want to test for a Hamiltonian cycle,

**Create:** an instance of TSP.

# Creating a TSP instance

A TSP instance  $D$  consists of  $n$  cities, and  $n(n - 1)$  distances.

**Cities** We have a city  $c_i$  for every node  $v_i$ .

**Distances** Let  $d(c_i, c_j) = \begin{cases} 1 & \text{if edge } (v_i, v_j) \in E \\ 2 & \text{otherwise} \end{cases}$



# TSP Reduction

## Theorem

*$G$  has a Hamiltonian cycle  $\iff D$  has a tour of length  $\leq n$ .*

*Proof.* If  $G$  has a Ham. Cycle, then this ordering of cities gives a tour of length  $\leq n$  in  $D$  (only distances of length 1 are used).

Suppose  $D$  has a tour of length  $\leq n$ . The tour length is the sum of  $n$  terms, meaning each term must equal 1, and hence cities that are visited consecutively must be connected by an edge in  $G$ .  $\square$

Also, TSP  $\in$  **NP**: a certificate is simply an ordering of the  $n$  cities.

# TSP is NP-complete

Hence, TSP is NP-complete.

Even TSP restricted to the case when the  $d(i, j)$  values come from actual distances on a map is NP-complete.