

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?

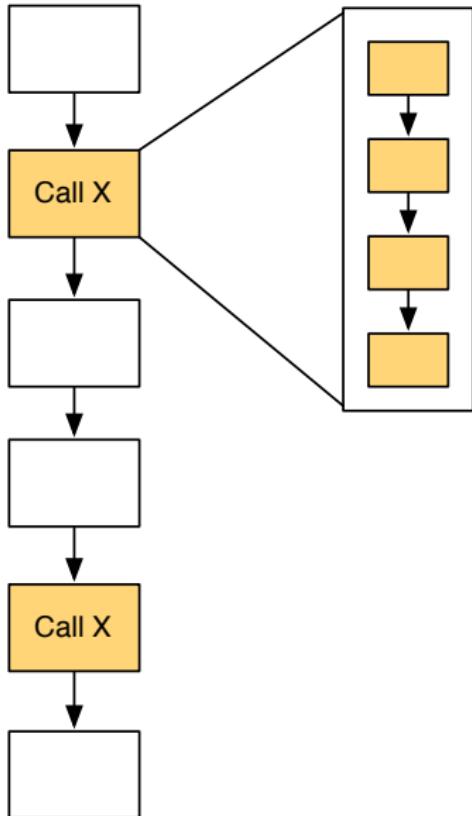
Polynomial Problems

Suppose:

- $Y \leq_P X$, and
- there is an 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 a 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.

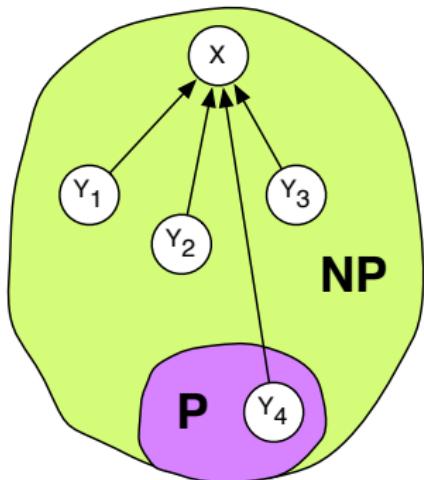
NP-completeness

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

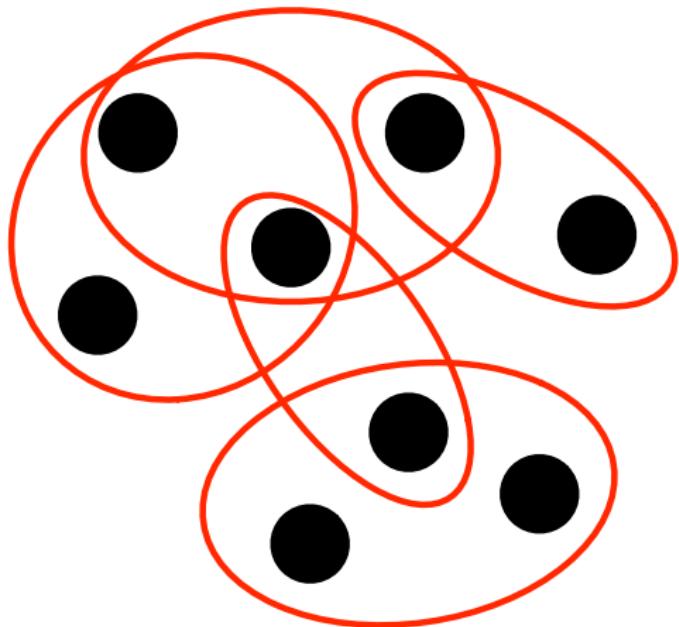
Given a set U of elements and a collection S_1, \dots, S_m of subsets of U , is there a collection of at most k of these sets whose union equals U ?

We will show that

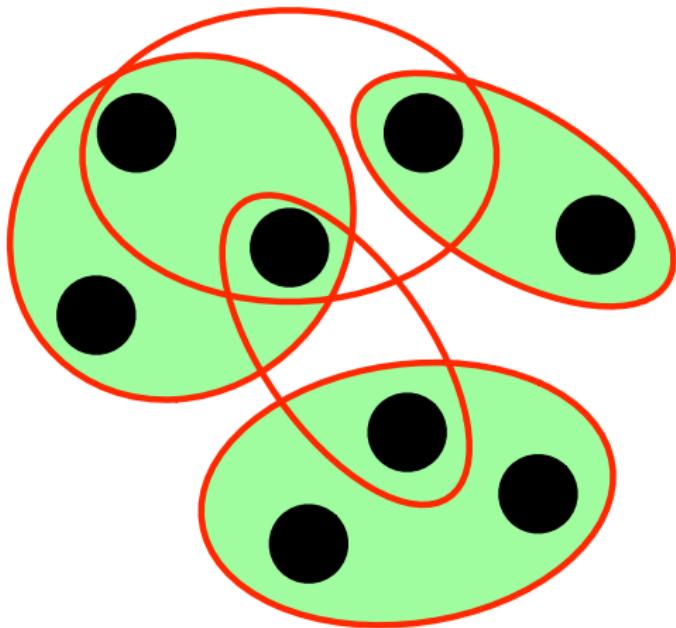
$$\begin{aligned} \text{SET COVER} &\in NP \\ \text{VERTEX COVER} &\leq_P \text{SET COVER} \end{aligned}$$

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_ℓ : 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: $(x_1 \vee \bar{x}_2), (\bar{x}_1 \vee \bar{x}_3), (x_2 \vee \bar{v}_3)$

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

Def. A CNF formula is a conjunction of clauses:

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

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

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 $X \in \textbf{NP}$ 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. $\text{3-SAT} \leq_P \text{Independent Set}$

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

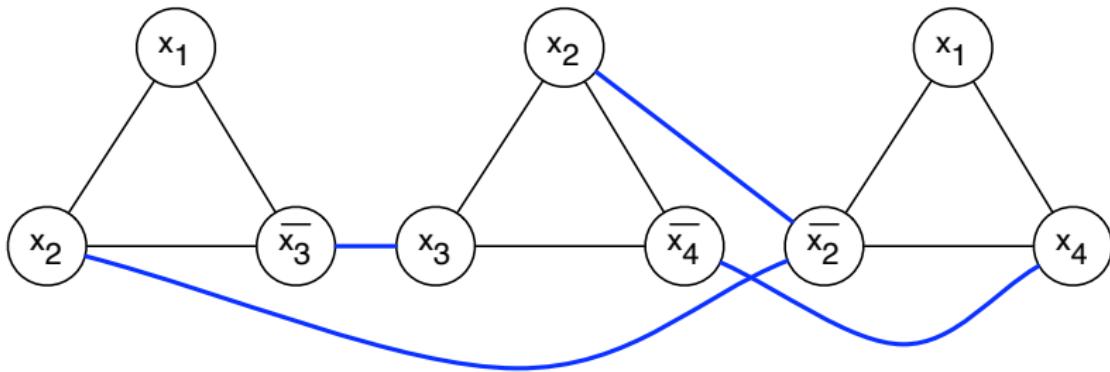
To solve 3-SAT:

- 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

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



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

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 $\in \textbf{NP}$: 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.

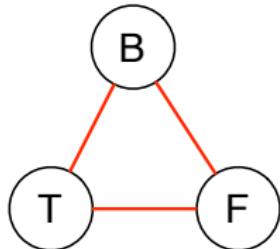
Reduction from 3-SAT

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:

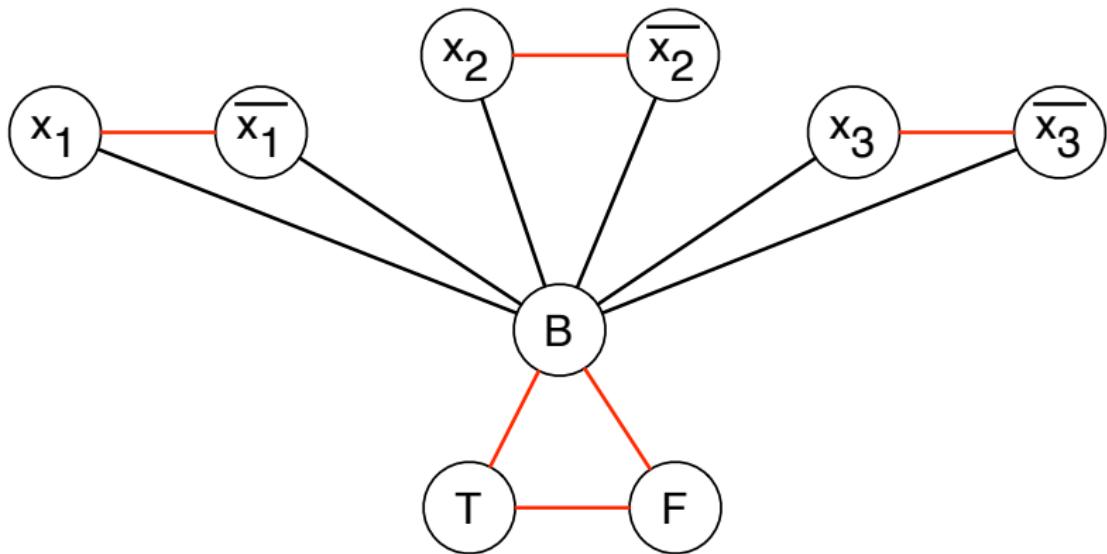


Create 3 *special nodes* T, F, and B, joined in a triangle:



Connecting them up

Connect every variable node to B:



Properties

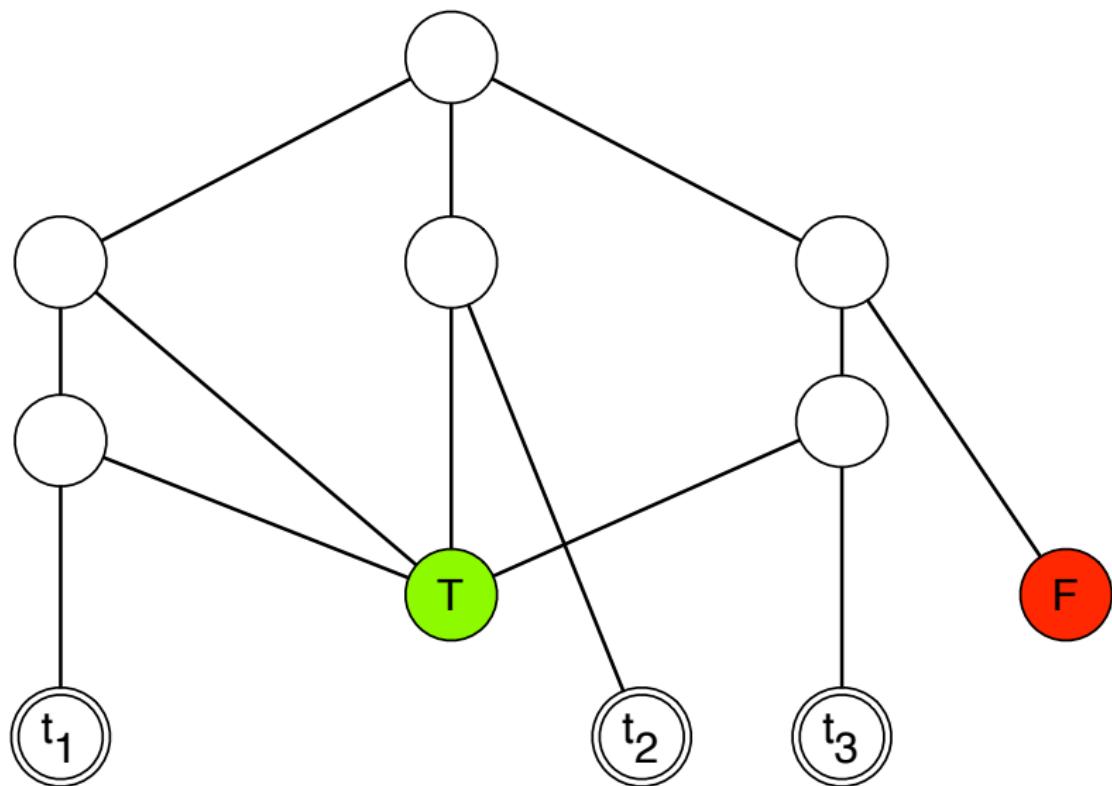
Properties:

- 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.

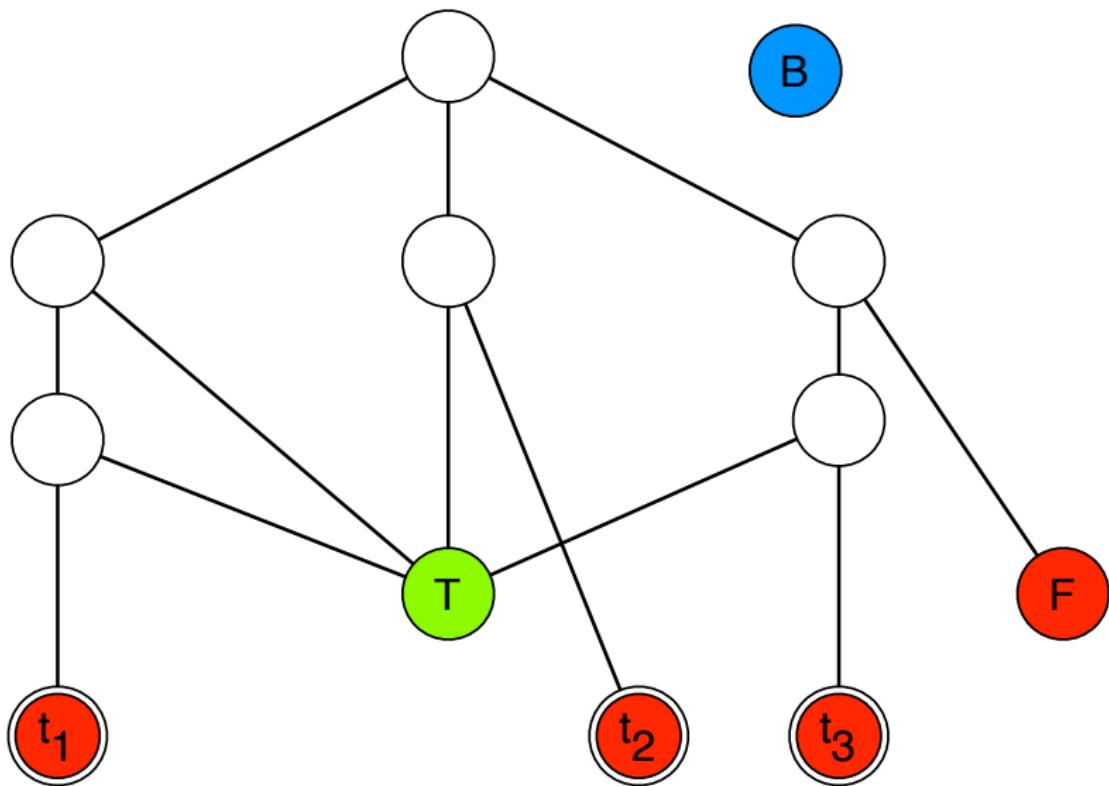
Connect Clause (t_1, t_2, t_3) up like this:



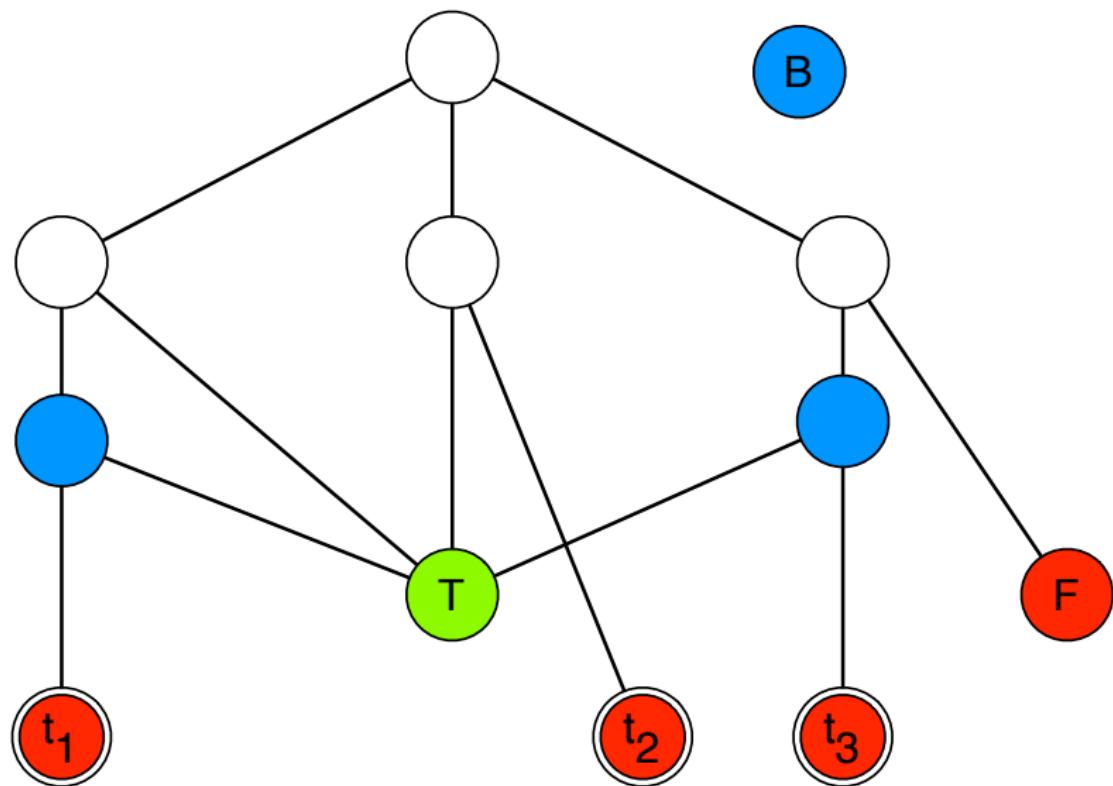
Suppose Every Term Was False

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

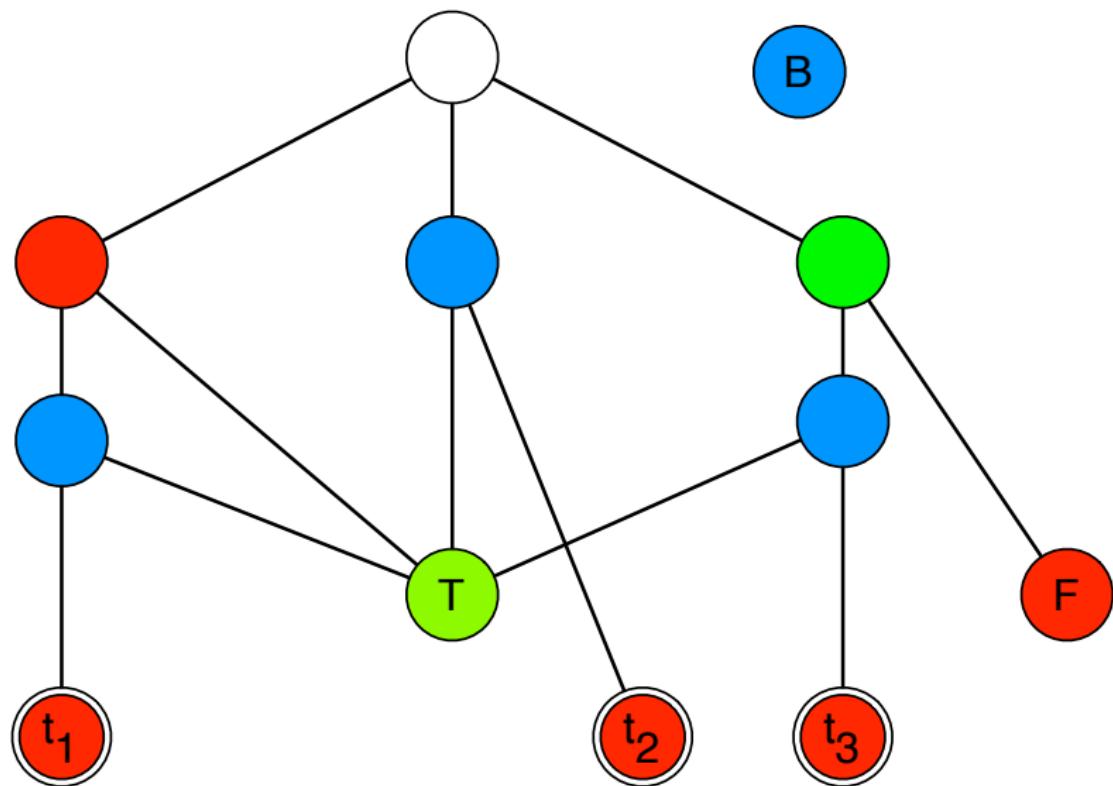
Connect Clause (t_1, t_2, t_3) up like this:



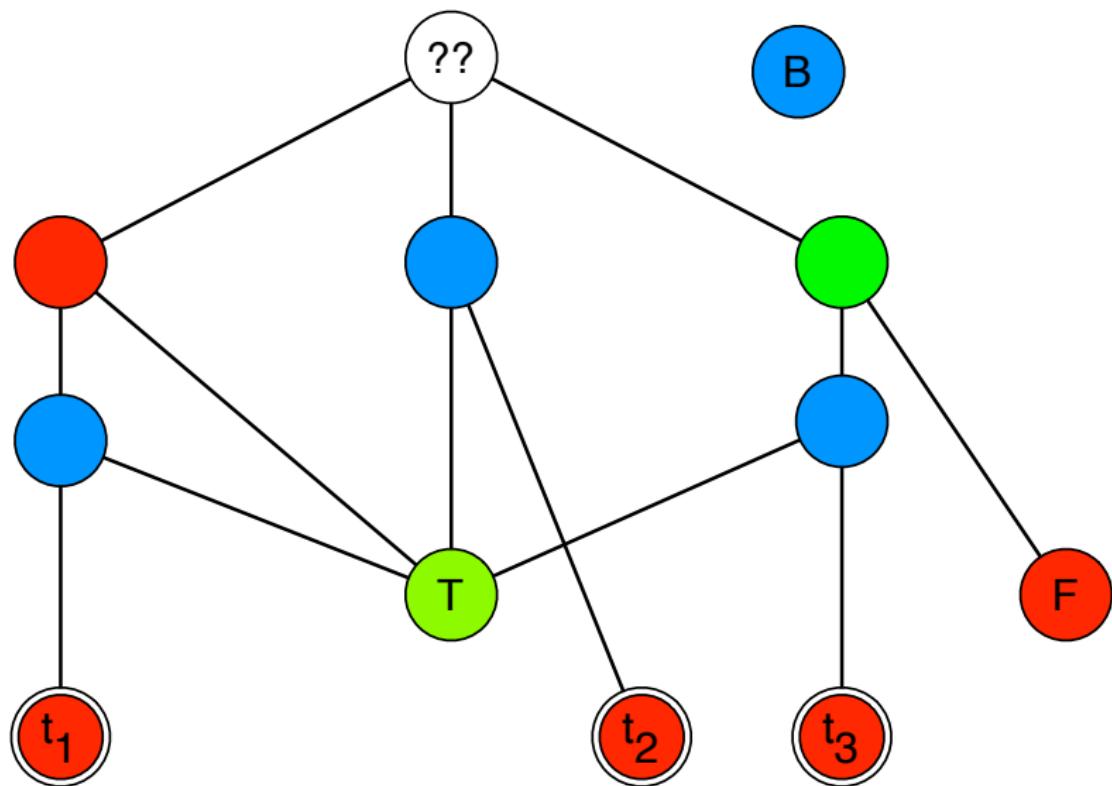
Connect Clause (t_1, t_2, t_3) up like this:



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 $x_i = \text{true}$ 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 $x_i = \text{true}$, color v_i with the same color as T and \bar{v}_i with the color of F.
- If $x_i = \text{false}$, 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 $X \in \text{NP}$. 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

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.

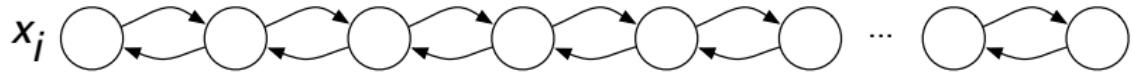
Consider an instance I of 3-SAT, with variables x_1, \dots, x_n and clauses C_1, \dots, C_k .

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

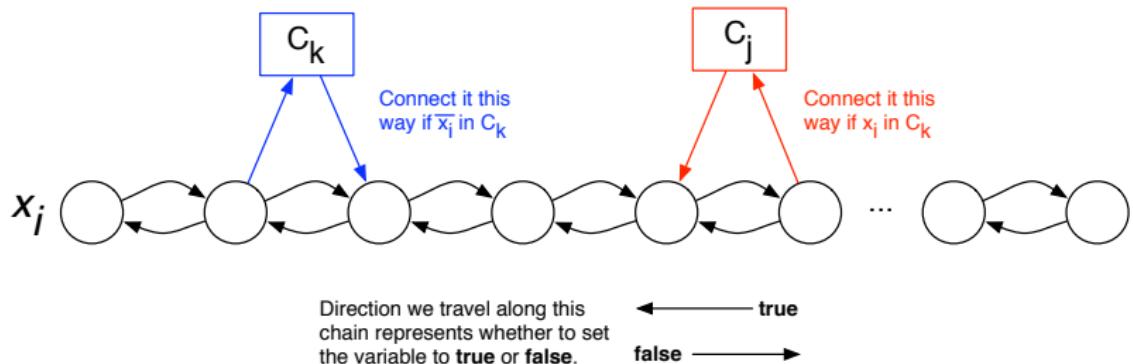


Direction we travel along this
chain represents whether to set
the variable to **true** or **false**.

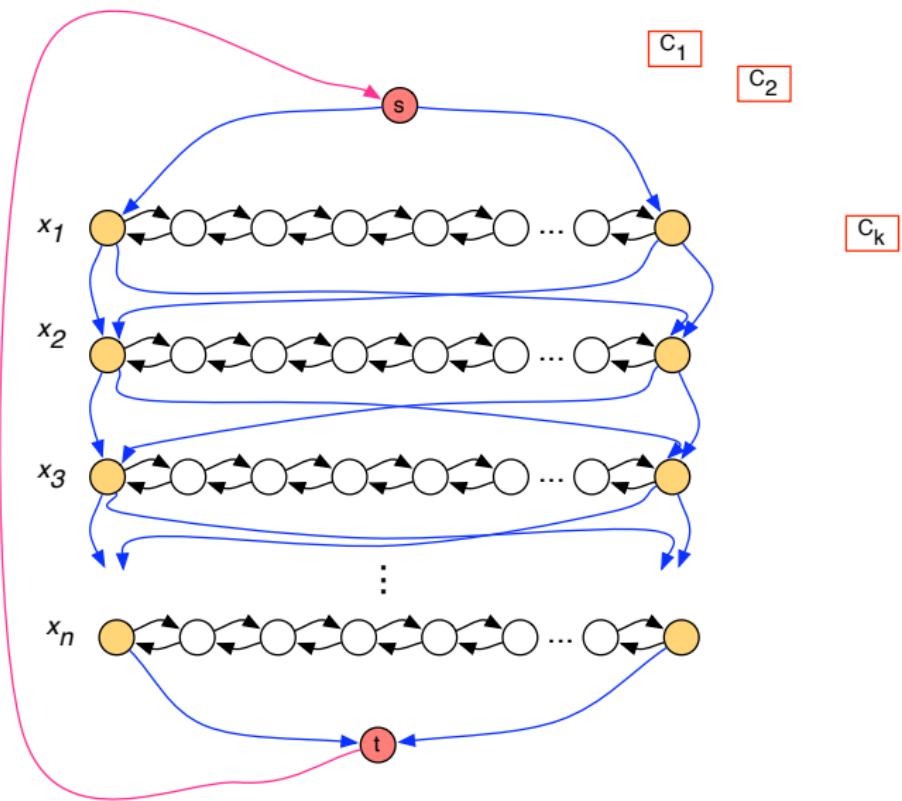
← true
false →

Hooking in the Clauses

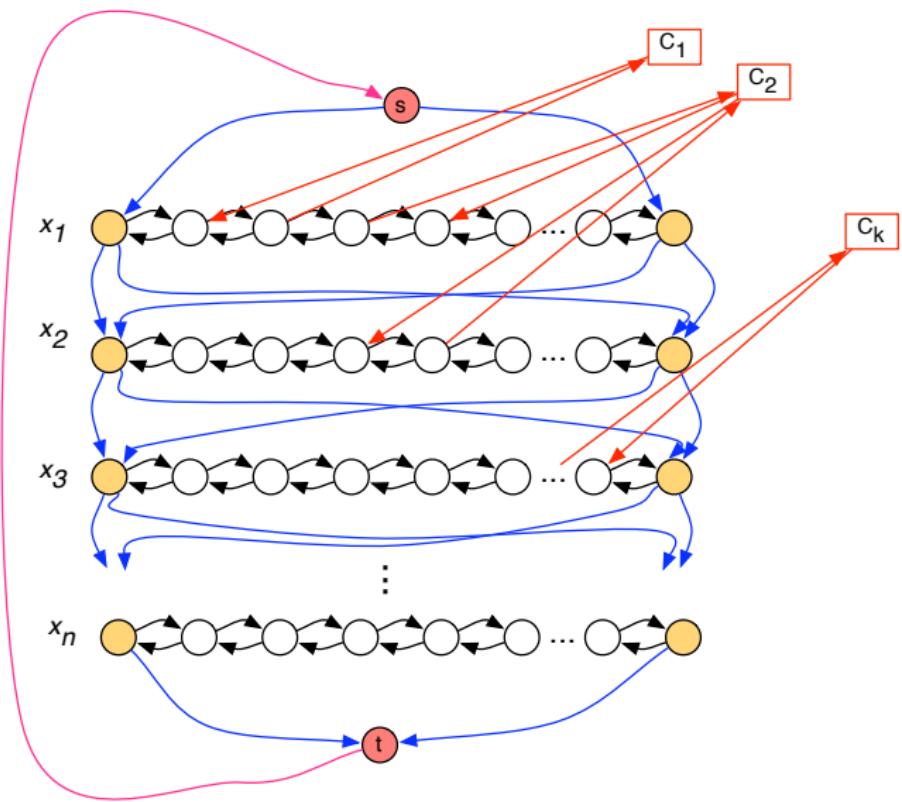
Add a new node for each clause:



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?

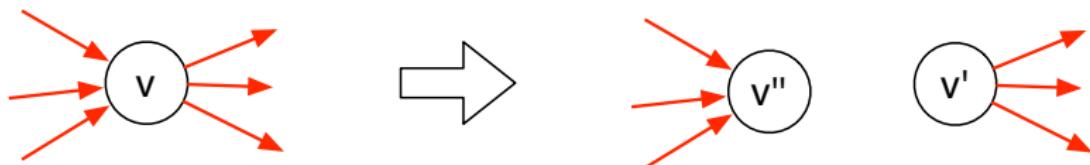
Hamiltonian Path

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 .

\iff 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

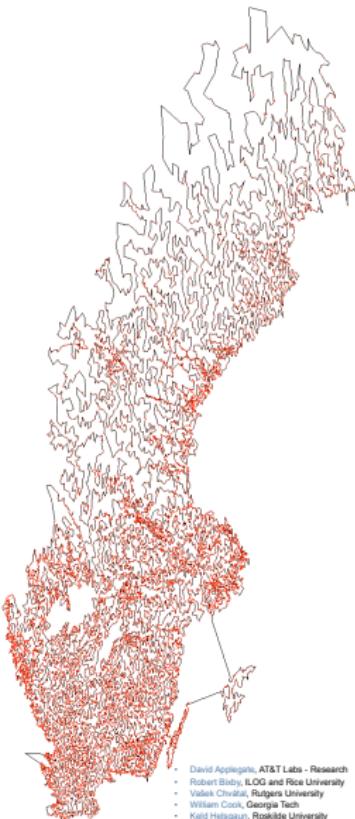
Traveling Salesman Problem

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, Vašek 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

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

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 $\in \textbf{NP}$: 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.