

# CS 341: ALGORITHMS

**Lecture 24: intractability VI – Decidability, more NPC transformations**

Readings: see website

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# COMPLEXITY CLASS **EXPTIME**

A very brief overview

**(Non-core material)**



**EXPTIME** is the set of all **decision problems** that can be solved in **exponential time**.  
I.e., in time  $O(2^{poly(n)})$  where  $poly(n)$  is a polynomial in the input size.

Observe that  $NP \subseteq EXPTIME$

The idea is to generate all possible certificates of an appropriate length and check them for correctness using the given certificate verification algorithm. An example, for **Hamiltonian Cycle**, we could generate all  $n!$  certificates and check each one in turn.

$$O(n!) \subseteq O(n^n) = O(2^{n \log n}) \text{ time}$$

We do not know if there are problems in **NP** that cannot be solved in polynomial time (because the **P = NP?** conjecture is not yet resolved). However, it is possible to prove that there exist problems in **EXPTIME \ P**.

One such problem is the **Bounded Halting** problem. Here an instance  $I = (A, x, t)$ , where  $A$  is a program,  $x$  is an input to  $A$ , and  $t$  is a positive integer (in binary). The question to be solved is if  $A(x)$  halts after at most  $t$  computation steps.

The **Bounded Halting** problem can be solved in time  $O(t)$ , but this is not a polynomial time algorithm because  $\text{size}(I) = |A| + |x| + \log_2 t$ .

Actually, it can be proven that **Bounded Halting** is EXPTIME-complete. This implies that it is in **EXPTIME**  $\setminus$  **P**, since it is known that **EXPTIME**  $\neq$  **P**.

**$t$  is exponential in  $\log t$ .**

(And  **$\log t$**  might be the largest term in the input size, in which case  $O(t)$  would be **exponential in the input size.**)



# UNDECIDABILITY

Problems that are *impossible* to solve

# DECIDABLE VS UNDECIDABLE PROBLEMS

We say an algorithm  $A$  “solves” a decision problem if, for **every** instance  $I$ ,  $A(I)$  has **finite** runtime and returns the correct answer

If an algorithm  $A$  **solves** decision problem  $\Pi$ ,  
then we say  $\Pi$  is **decidable**.

Formally,  $\Pi$  is **decidable IFF** there exists some algorithm  $A$  such that,  
for **every** instance  $I$ ,  $A(I)$  returns the correct answer in **finite** time.

If it is **not possible** to design an algorithm  $A$  that **solves** decision problem  $\Pi$ ,  
then we say  $\Pi$  is **undecidable**.

Formally,  $\Pi$  is **undecidable IFF** there **cannot exist** an algorithm  $A$  such that,  
for every instance  $I$ ,  $A(I)$  returns the correct answer in **finite time**.

Equivalently,  $\Pi$  is **undecidable IFF**, for every algorithm  $A$ , there exists some input  $I$   
such that  $A(I)$  **does not** return the correct answer in finite time.

I.e., for some input,  $A(I)$  either runs forever or returns the wrong answer



# HALTING: AN UNDECIDABLE PROBLEM

## Problem 7.19

### Halting

**Instance:** *A computer program  $A$  and input  $x$  for the program  $A$ .*

**Question:** *When program  $A$  is executed with input  $x$ , will it halt in finite time?*

For example, you could run  $Halt(BFS, G)$  to determine whether,  $BFS(G)$  will halt in finite time, which it will, so  $Halt(BFS, G)$  returns yes.

The **Halting** problem is **decidable** IFF there **exists an algorithm**  $Halt(I)$  that, for **every** instance  $I = (A, x)$ ,  $Halt(I)$  has **finite** runtime and correctly answers the question: "would a call to  $A(x)$  halt in finite time?"

# UNDECIDABILITY OF THE HALTING PROBLEM

Suppose that *Halt* is a program that solves the **Halting Problem**.

We suppose *Halt* **exists**, to obtain a **contradiction**...

The statement “*Halt* solves the Halting problem” means that *Halt* runs in finite time, and:

$$Halt(A, x) = \begin{cases} \text{true} & \text{if } A(x) \text{ halts} \\ \text{false} & \text{if } A(x) \text{ doesn't halt.} \end{cases}$$

Note that *A* (the “algorithm”) and *x* (the “input” to *A*) are both strings over some finite alphabet.

Since ***A* is a string** (of code), and its input ***x* is also a string**...  
we **could** pass *A* as an argument to itself: *A*(*A*)

Then we could ask **if *A*(*A*) halts**, by running *Halt*(*A*, *A*)...

Weird... Let's try to obtain a contradiction by doing this...



Consider the following algorithm *Strange*.

**Algorithm:** *Strange*(*A*)

external *Halt*

if not *Halt*(*A*, *A*)

then return (!)

else  $\left\{ \begin{array}{l} i \leftarrow 1 \\ \text{while } i \neq 0 \text{ do } i \leftarrow i + 1 \end{array} \right.$

If not *Halt*(*A*, *A*), then *A*(*A*) will run forever

but *Strange*(*A*) terminates in finite time

Else if *Halt*(*A*, *A*), then *A*(*A*) will terminate in finite time

But *Strange*(*A*) will run forever

What happens when we run *Strange*(*Strange*)?

Two cases: *Strange*(*Strange*) either halts or does not halt

**Suppose *Strange*(*Strange*) halts.** Then, it must return.  
This means it sees *not Halt*(*A*, *A*) just before returning.

But *A* = *Strange*, so it sees *not Halt*(*Strange*, *Strange*).

**So, *Strange*(*Strange*) does not halt --- contradiction!**

**Suppose *Strange*(*Strange*) does not halt.** Then, it must spin in the while loop forever. This means *Halt*(*A*, *A*) = *true*.

But *A* = *Strange*, so *Halt*(*Strange*, *Strange*) = *true*.

**So, *Strange*(*Strange*) halts --- contradiction!**

Both cases lead to a contradiction.  
So, our only assumption, **that *Halt* exists**,  
must be false!

Therefore, the Halting  
problem is **undecidable**.

# Another Undecidable Problem

Here is another example of an undecidable problem. The problem **Halt-All** takes a program  $A$  as input and asks if  $A$  halts on all inputs  $x$ .

We describe a Turing reduction **Halting**  $\leq^T$  **Halt-All**, which proves that **Halt-All** is undecidable.

Assume we have a program *HaltAllSolver*.

For a fixed program  $A$  and input  $x$ , let  $B_x()$  be the program that executes  $A(x)$  (so  $B_x$  has no input).

Here is the reduction:

Given  $A$  and  $x$  (an instance of **Halting**), construct the program  $B_x$ .  
Run *HaltAllSolver*( $B_x$ ),

We have

$$\text{HaltAllSolver}(B_x) = \mathbf{true} \iff A(x) \text{ halts,}$$

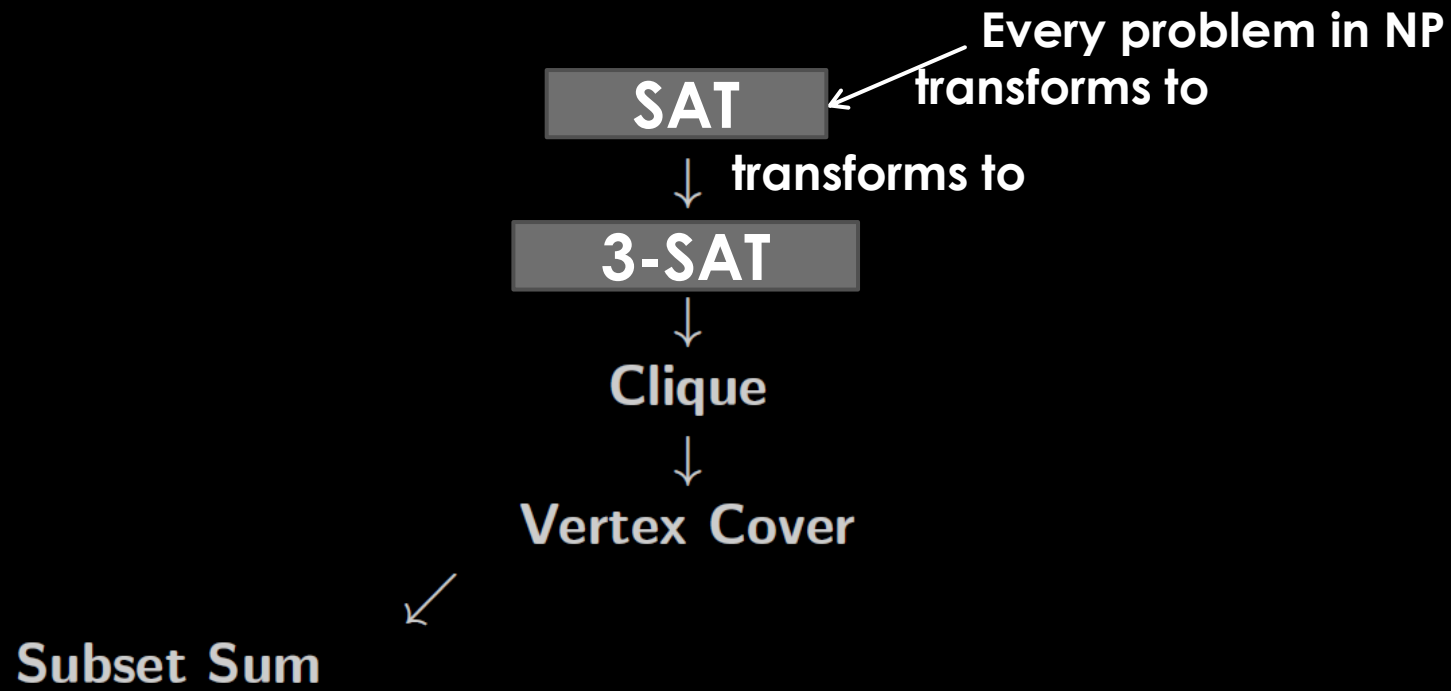
so we can solve the halting problem.

If we have *HaltAllSolver* then we have *Halt*, but this is impossible, so *HaltAllSolver* cannot exist, so **the Halt-All problem is undecidable!**



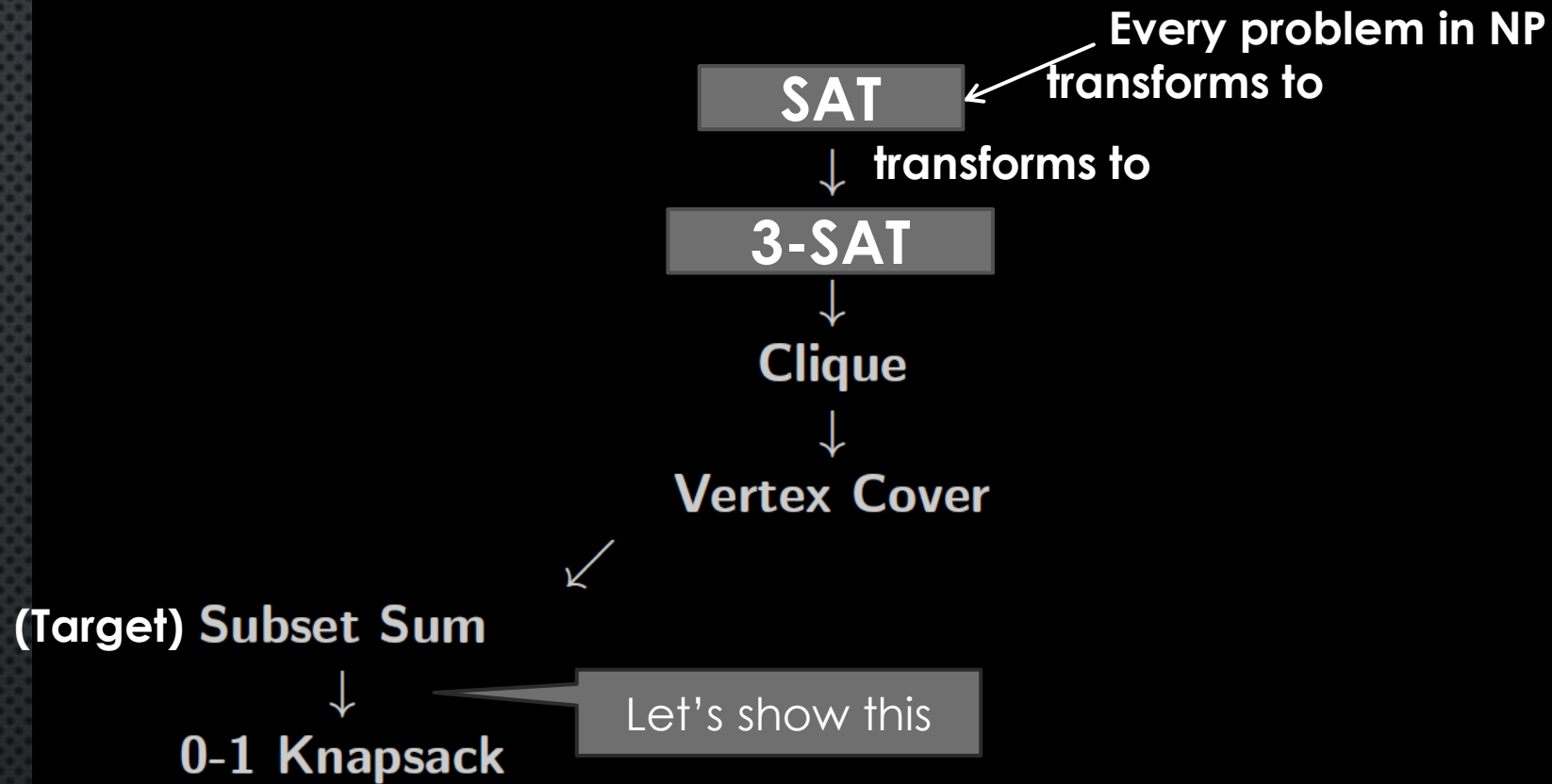
# FINISHING NPC TRANSFORMATIONS/REDUCTIONS

# Summary of Polynomial Transformations





# Summary of Polynomial Transformations



REDUCE TARGET SUBSET SUM  
TO 0-1 KNAPSACK



# RECALL: 0-1 KNAPSACK PROBLEM

## Problem 7.3

### 0-1 Knapsack-Dec

**Instance:** a list of profits,  $P = [p_1, \dots, p_n]$ ; a list of weights,  $W = [w_1, \dots, w_n]$ ; a capacity,  $M$ ; and a target profit,  $T$ .

**Question:** Is there an  $n$ -tuple  $[x_1, x_2, \dots, x_n] \in \{0, 1\}^n$  such that  $\sum w_i x_i \leq M$  and  $\sum p_i x_i \geq T$ ?

Can I obtain profit  $T$  (or better) by taking (whole) items with total weight  $\leq M$ ?

# TARGET SUBSET SUM $\leq_P$ 0-1 KNAPSACK

## Problem 7.18

### Subset Sum

**Instance:** A list of sizes  $S = [s_1, \dots, s_n]$ ; and a target sum,  $T$ . These are all positive integers.

**Question:** Does there exist a subset  $J \subseteq \{1, \dots, n\}$  such that  $\sum_{i \in J} s_i = T$ ?

How should we poly-transform (Target) Subset-Sum input into (Target) 0-1 Knapsack input

## Problem 7.3

### 0-1 Knapsack-Dec

**Instance:** a list of profits,  $P = [p_1, \dots, p_n]$ ; a list of weights,  $W = [w_1, \dots, w_n]$ ; a capacity,  $M$ ; and a target profit,  $T$ .

**Question:** Is there an  $n$ -tuple  $[x_1, x_2, \dots, x_n] \in \{0, 1\}^n$  such that  $\sum w_i x_i \leq M$  and  $\sum p_i x_i \geq T$ ?

Such that:  $I$  contains a subset that sums to  $T$  IFF ( $\geq T$ ) profit can be obtained in knapsack input  $f(I)$



# Subset Sum $\leq_P$ 0-1 Knapsack

Let  $I$  be an instance of **Subset Sum** consisting of ints  $[s_1, \dots, s_n]$  and target sum  $T$ .

Define

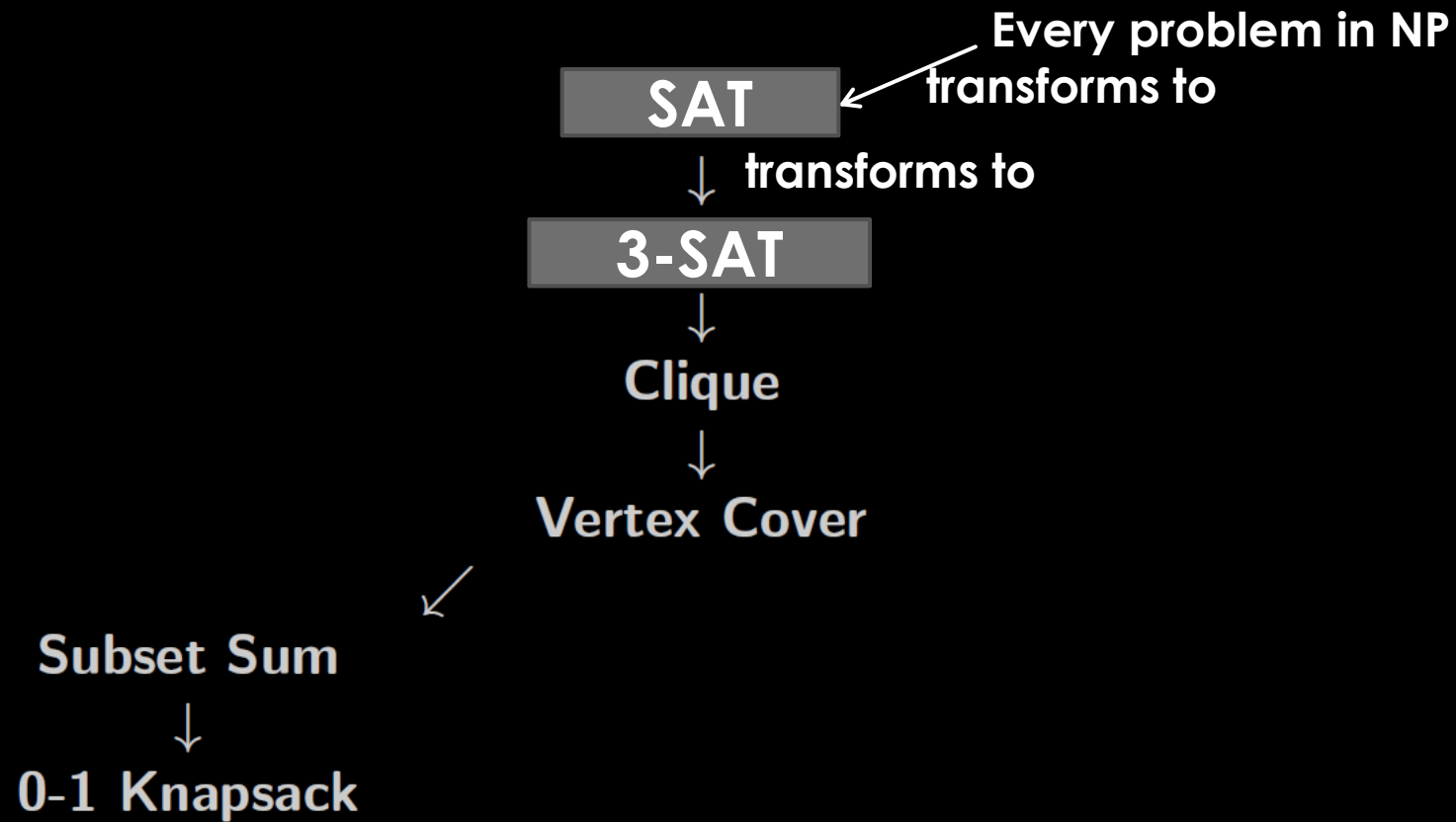
$$\begin{aligned}p_i &= s_i, 1 \leq i \leq n \\w_i &= s_i, 1 \leq i \leq n \\M &= T\end{aligned}$$

Then define  $f(I)$  to be the instance of **0-1 Knapsack** consisting of profits  $[p_1, \dots, p_n]$ , weights  $[w_1, \dots, w_n]$ , capacity  $M$  and target profit  $T$ .

Exercise: Prove the correctness of this transformation.

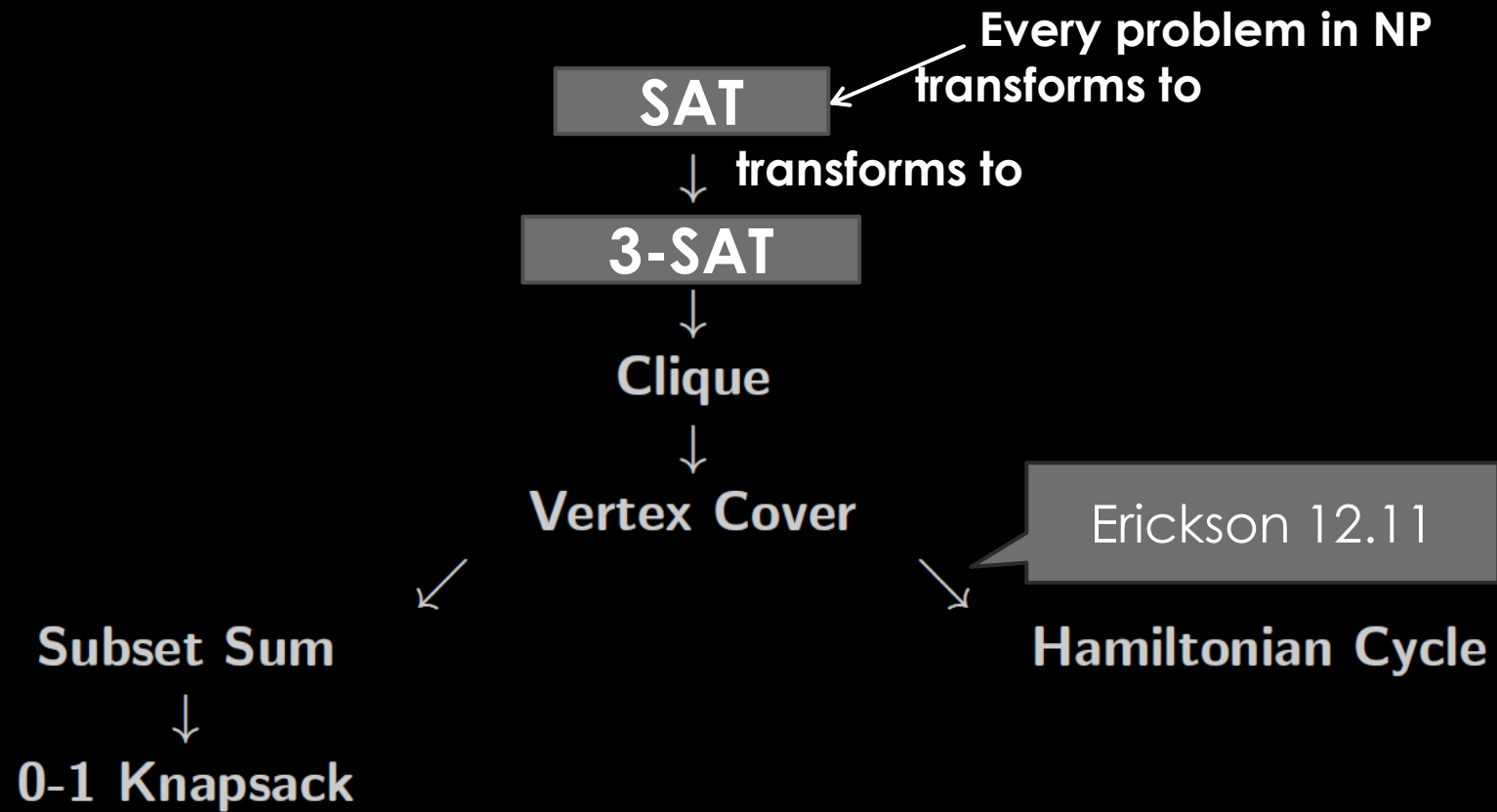
Claim:  $I$  contains a subset that sums to  $T$  **IFF** ( $\geq T$ ) profit can be obtained in knapsack input  $f(I)$

# Summary of Polynomial Transformations

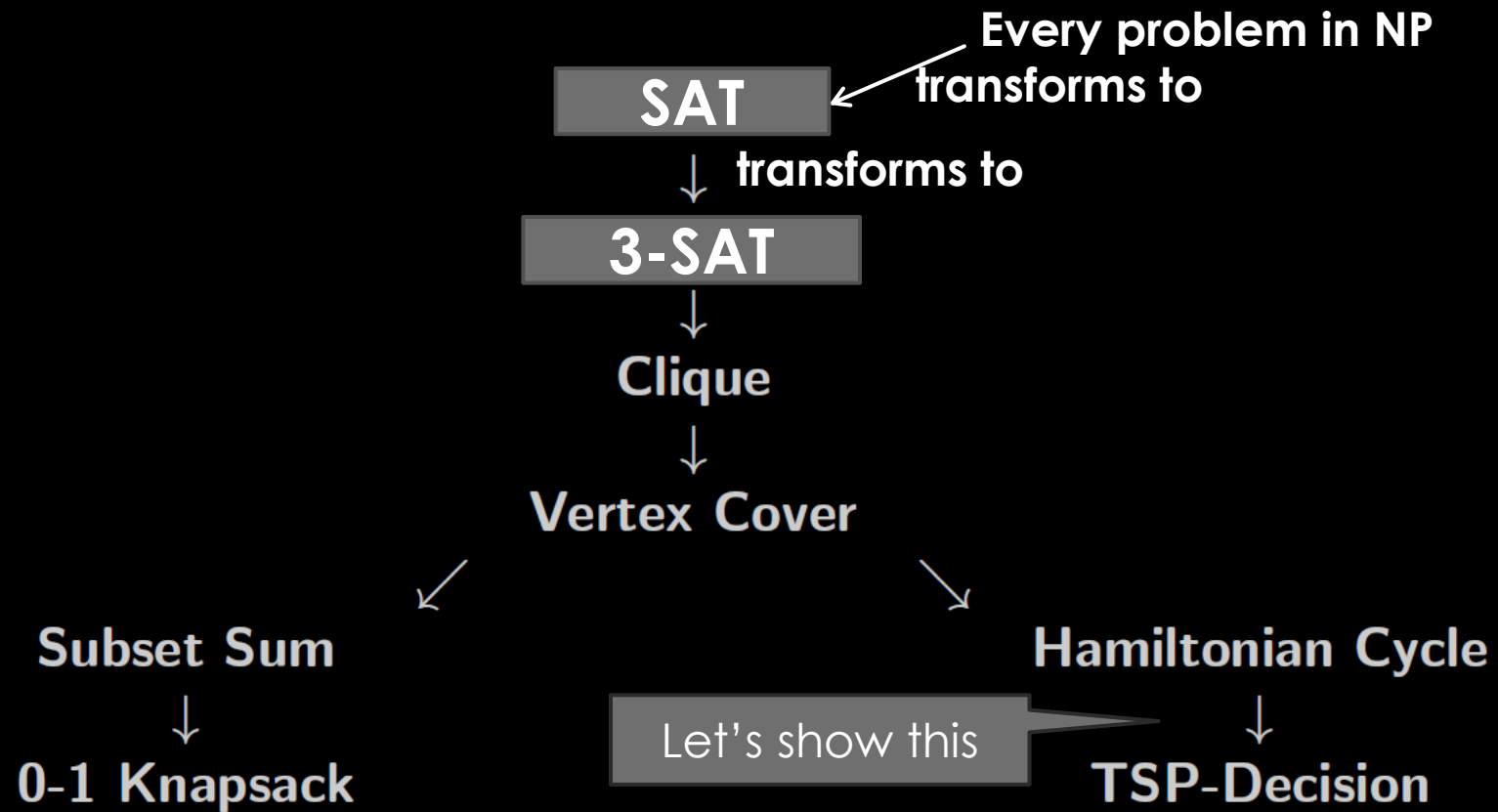




# Summary of Polynomial Transformations



# Summary of Polynomial Transformations





**REDUCE HAMILTONIAN CYCLE TO TSP-DECISION**

# EXERCISE: GIVE A POLY-TRANSFORMATION

## Problem 7.2

### Hamiltonian Cycle

**Instance:** *An undirected graph  $G = (V, E)$ .*

**Question:** *Does  $G$  contain a hamiltonian cycle?*

**This exercise:** Show how to transform Hamiltonian Cycle input into TSP-Decision input (in poly time).

A **hamiltonian cycle** is a cycle that passes through every vertex in  $V$  exactly once.

## Problem 7.7

### TSP-Decision

**Instance:** *A graph  $G$ , edge weights  $w : E \rightarrow \mathbb{Z}^+$ , and a target  $T$ .*

**Question:** *Does there exist a hamiltonian cycle  $H$  in  $G$  with  $w(H) \leq T$ ?*

**Such that:**  $I$  contains a Ham Cycle  
**IFF**  $f(I)$  contains a Ham Cycle of  
weight at most  $T$

# Hamiltonian Cycle $\leq_P$ TSP-Dec

Let  $I$  be an instance of **Hamiltonian Cycle** consisting of a graph  $G = (V, E)$ .

For the complete graph  $K_n$ , where  $n = |V|$ , define edge weights as follows:

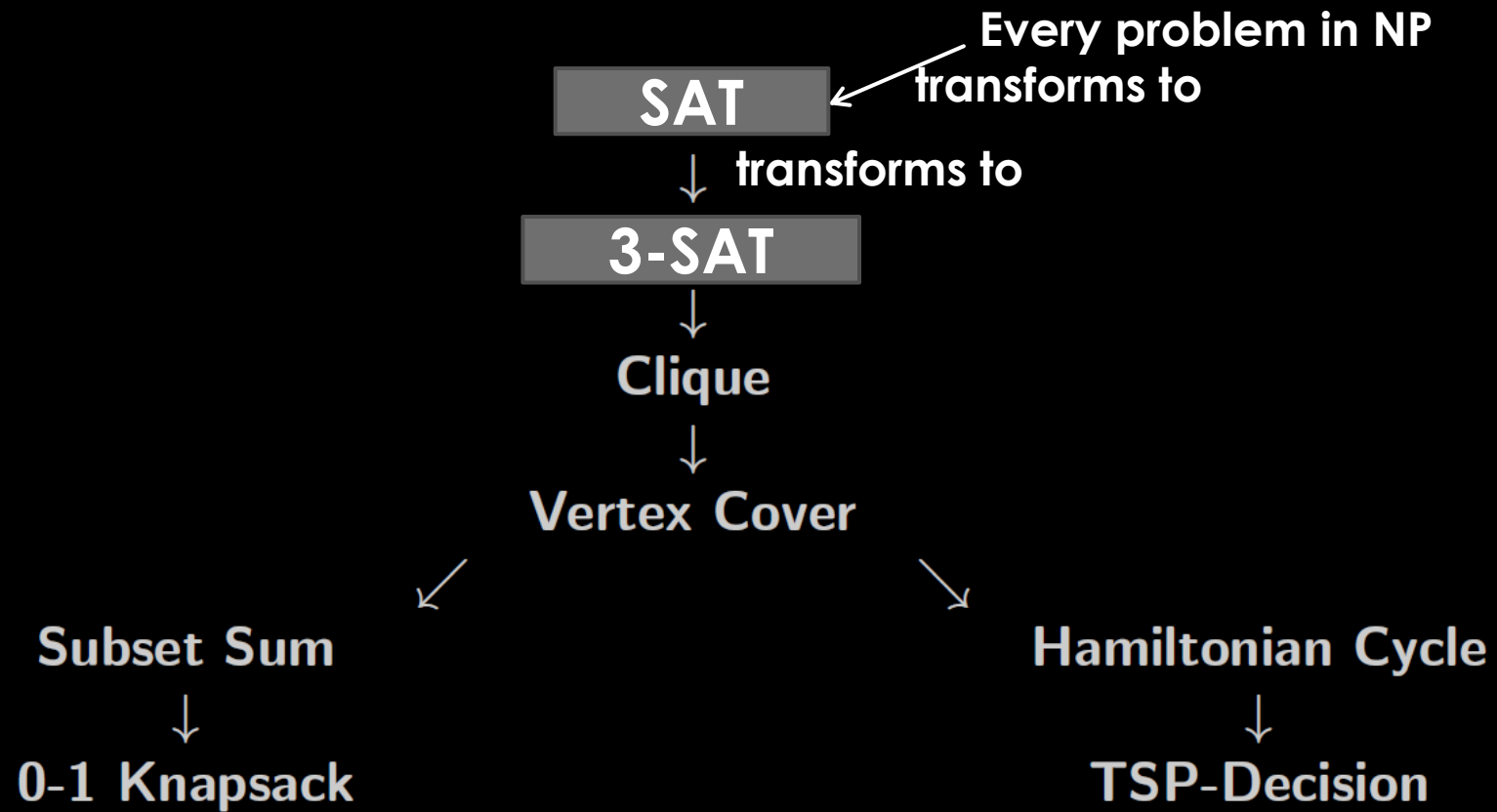
$$w(uv) = \begin{cases} 1 & \text{if } uv \in E \\ 2 & \text{if } uv \notin E. \end{cases}$$

Then define  $f(I)$  to be the instance of **TSP-Dec** consisting of the graph  $K_n$ , edge weights  $w$  and target  $T = n$ .

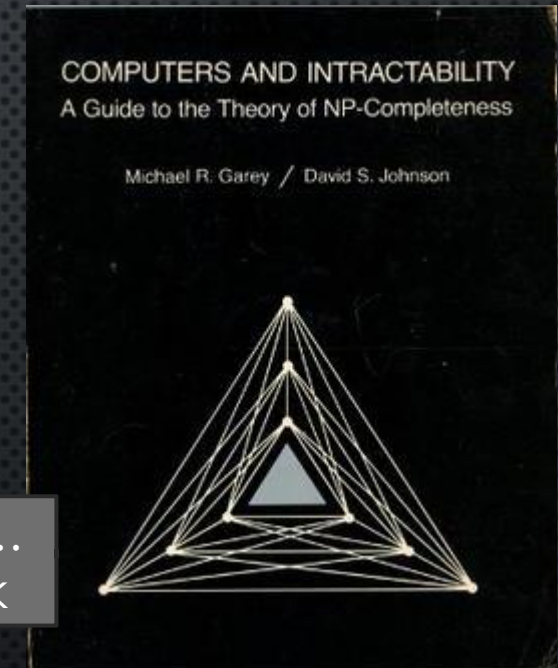
Exercise: Prove the correctness of this transformation.



# Summary of Polynomial Transformations



Any many, many more ☺...  
over 300 listed in this book

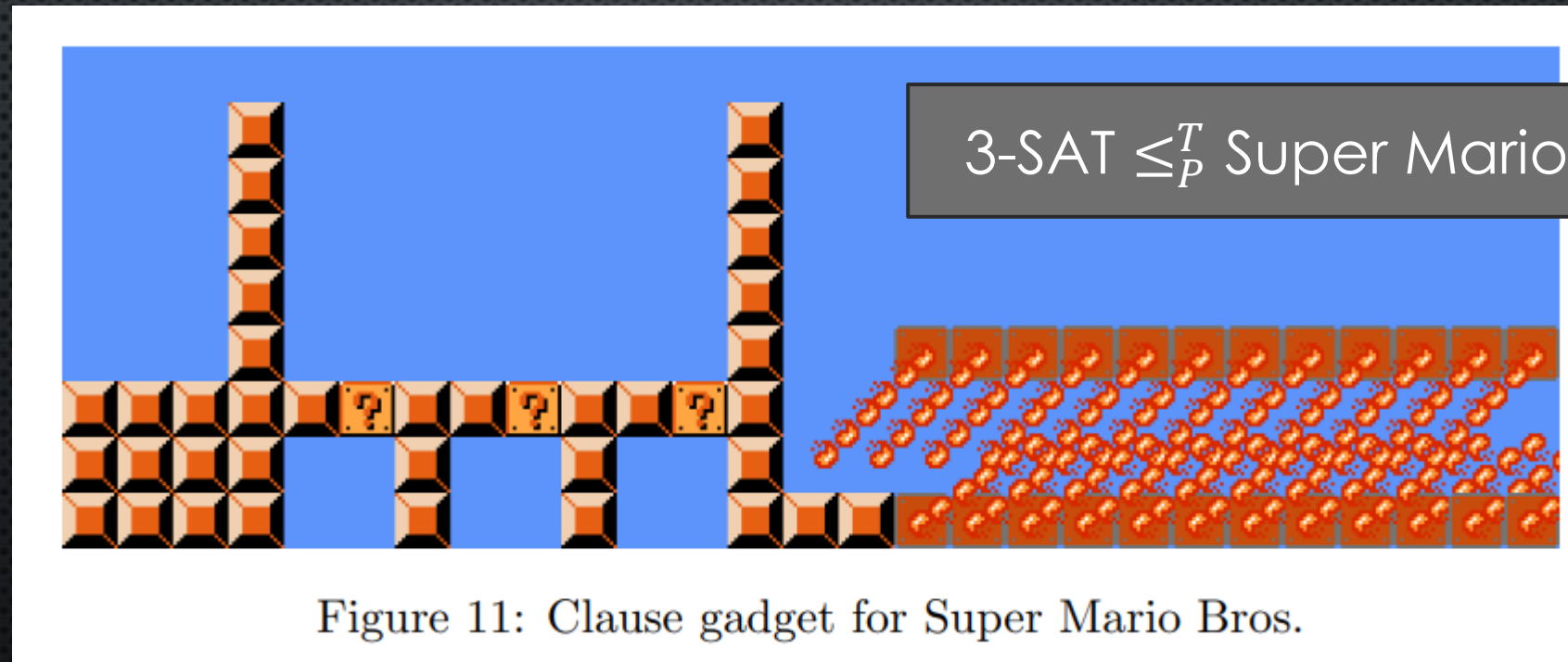


# FUN AND GAMES

<https://arxiv.org/pdf/1203.1895.pdf>

## Abstract

We prove NP-hardness results for five of Nintendo's largest video game franchises: Mario, Donkey Kong, Legend of Zelda, Metroid, and Pokémon. Our results apply to generalized versions of Super Mario Bros. 1–3, The Lost Levels, and Super Mario World; Donkey Kong Country 1–3; all Legend of Zelda games; all Metroid games; and all Pokémon role-playing games. In addition, we prove PSPACE-completeness of the Donkey Kong Country games and several Legend of Zelda games.





# (FACTUALLY INCORRECT) MEMES

- There's also an old video meme about proving that Super Mario Bros is NP complete
  - (Long before it was legitimately proved NP hard 😊)
- Whereas the stuff on the **previous slide is real math**, the stuff in [this video](#) is just a meme, and is **very wrong**. but you may find it funny...



# SUMMARY OF COMPLEXITY CLASSES

See this slide's notes

- **P** (Poly-time) E.g., (**decision** problem variants of:) BFS, Dijkstra's, some DP algorithms
  - **Decision** problems that can be solved by algorithms with runtime  $\text{poly}(\text{input size})$
- **NP** (Non-deterministic poly-time) **All of P**, and e.g., vertex cover, clique, SAT, subset sum
  - **Decision** problems for which **certificates** can be **verified** in time  $\text{poly}(\text{input size})$
  - Equivalently: decision problems that can be solved in poly-time if you have access to a non-deterministic oracle that returns a yes-certificate if one exists
- **NPC** (NP-complete) E.g., vertex cover, clique, SAT, subset sum, TSP-decision
  - **Decision** problems  $\Pi \in NP$  s.t. every  $\Pi' \in NP$  can be **transformed** to  $\Pi$  in poly-time
- **NP-hard** (at least as hard as NPC) **All of NPC**, and e.g., TSP-optimization, TSP-optimal value
  - problems  $\Pi$  s.t. every  $\Pi' \in NP$  can be **reduced** to  $\Pi$  in poly-time
- Note: P, NP and NPC problems are **decidable**

# POLYTIME 2-SAT

(IF WE HAVE TIME)

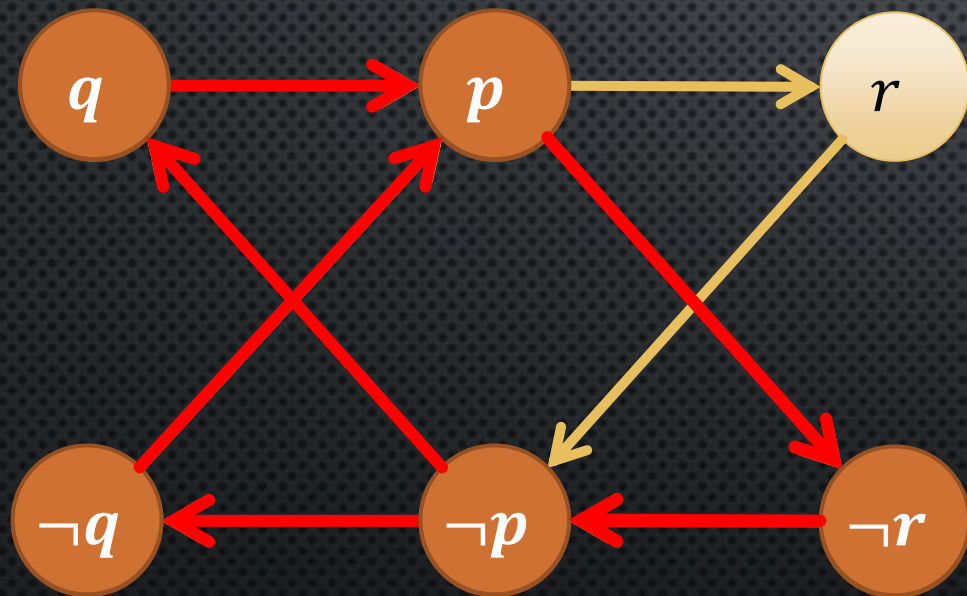


# 2-SAT EXAMPLES

- $(p \vee q) \wedge (\neg p \vee r) \wedge (\neg r \vee \neg p)$ 
  - Satisfiable:  $p = 0, q = 1, r \in \{0,1\}$
- $(p \vee q) \wedge (\neg p \vee r) \wedge (\neg r \vee \neg p) \wedge (p \vee \neg q)$

Logical refresher:  
 $p \Rightarrow q$  is **equivalent** to  
 $\neg p \vee q$ .

Therefore,  $p \vee q$  is  
**equivalent** to  $\neg p \Rightarrow q$  and  
**equivalent** to  $\neg q \Rightarrow p$



Edges (implications of clauses)...

$\neg p \Rightarrow q$	$p \Rightarrow r$	$r \Rightarrow \neg p$	$\neg p \Rightarrow \neg q$
$\neg q \Rightarrow p$	$\neg r \Rightarrow \neg p$	$p \Rightarrow \neg r$	$q \Rightarrow p$

$q \Rightarrow p \Rightarrow \neg r \Rightarrow \neg p \Rightarrow \neg q \dots$  so  $q$  cannot be *true*

$\neg q \Rightarrow p \Rightarrow \neg r \Rightarrow \neg p \Rightarrow q \dots$  so  $q$  cannot be *false*

Therefore the formula **cannot** be satisfied!



(variable names are integers in  $1..|X|$ )

**2-SAT** can be solved in polynomial time. Suppose we are given an instance  $I$  of **2-SAT** on a set of boolean variables  $X = \{1..|X|\}$

- (1) For every clause  $x \vee y$  (where  $x$  and  $y$  are literals), construct two directed edges  $\bar{x}y$  and  $\bar{y}x$ . We get a directed graph on vertex set  $X \cup \bar{X}$ .
- (2) Determine the strongly connected components of this directed graph.
- (3)  $I$  is a yes-instance if and only if there is no strongly connected component containing  $x$  and  $\bar{x}$ , for any  $x \in X$ .

Suppose no variable  $x$  is in the same SCC as  $\bar{x}$ , then to get a satisfying assignment do the following:

For each  $x$ , if  $\exists$  path from  $x$  to  $\bar{x}$ , then set  $x = false$  else set  $x = true$ .