

# Complexity Theory

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## Introduction

### Reading:

- ▶ “NP-Completeness” 34 CLRS
- ▶ “NP-Completeness” 13 GT

Focus was on designing efficient algorithms.

- ▶ Upperbounded space and time requirements.

Complexity theory tells us

- ▶ only inefficient algorithms are possible, or
- ▶ no algorithm is possible.

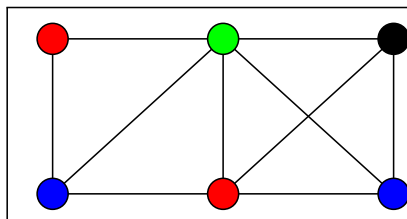
# Asymptotic hierarchy

How “hard” is a problem?

- ▶  $\Theta(1)$ : find the minimum element in a min-heap, hash-table look-up
- ▶  $\Theta(\log n)$ : binary search, range tree computation, 2-3-4 tree operations
- ▶  $\Theta(n)$ : order statistics, bucket sort
- ▶  $\Theta(n \log n)$ : sorting, weighted activity selection, closest pair of points
- ▶  $\Theta(n^2)$ : longest common subsequence
- ▶  $\Theta(n^2 \log n)$ : Prim's and Dijkstra's algorithms
- ▶  $\Theta(n^3)$ : Bellman-Ford algorithm
- ▶  $\vdots$

## Complexity hierarchy

- ▶ NP-hard: No efficient (polynomially bounded time) algorithm exists.



Colour vertices with the *least* number of colours so that adjacent vertices have different colours.

$\vdots$

- ▶ PSPACE-hard: Space efficient (polynomially bounded) solutions, but takes exponential time.

$\vdots$

- ▶ Undecidable: No (complete) algorithm exists

Will a given program eventually stop on a given input input?

## Decision vs. optimization

- ▶ Decision problem: The answer is either YES or NO.
- ▶ Optimization problem: Find a maximal or minimal solution.

Typically, a poly-time solution to one variant gives a poly-time solution to the other variant.

### Graph colouring

Given a graph  $G = (V, E)$ , a  $k$ -colouring of a  $G$  is a function  $\chi: V \rightarrow \{1, \dots, k\}$  such that  $\{u, v\} \in E$  implies  $\chi(u) \neq \chi(v)$ .

Decision problem: Is a given graph  $k$ -colourable?

Optimization problem: Find the smallest  $k^*$  such that a given graph is  $k^*$ -colourable.

## Poly-time

### Definition

A problem is poly-time solvable if some algorithm  $A$  solves every instance of the problem in  $O(n^k)$ , where  $k$  is a constant and  $n$  is the # of bits used to represent the input.

### Note: Depends on the input representation

**Problem:** Print the letter 'A'  $m$  times.

If  $m$  is represented in binary, then  $n = \lg m$  and size of the output is  $O(2^n)$ .

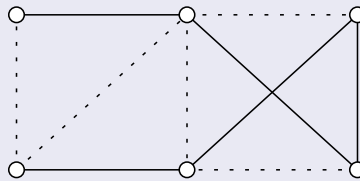
If  $m$  is written as a sequence of 1s, where the number of 1s is  $m$  (unary), then  $n = m$  and the size of the output is  $O(n)$ .

## Problem class: P vs NP

- ▶ P: all decision problems that are poly-time.
- ▶ NP: all decision problems where a YES can be poly-time verified given an appropriate certificate

### Hamiltonian path problem

Given a graph, does some path visit every vertex exactly once?



Finding a such a path is hard. But verifying that a given path (certificate) is Hamiltonian is easy.

## P vs. NP (cont'd)

- ▶  $P \subseteq NP$  because correct algorithms are certificates.
- ▶ Is  $NP \subseteq P$ ? \$1,000,000 if you find out  
[http://www.claymath.org/millennium/P\\_vs\\_NP](http://www.claymath.org/millennium/P_vs_NP)

### Satisfiability problem (SAT):

Given a set of boolean variables  $x_1, \dots, x_n$  and a set of clauses  $C_1, \dots, C_m$ , where each clause is a disjunction of variables and their complements

$$\begin{aligned} \text{e.g.} \quad C_1 &= x_1 \vee x_2 \vee \overline{x_4} \\ C_2 &= \overline{x_1} \vee x_3 \vee x_4 \\ C_3 &= x_2 \vee \overline{x_3} \vee \overline{x_5} \vee x_7, \end{aligned}$$

can we assign true/false values to each variable so that every clause is satisfied?

# NP-Completeness

SAT is clearly in NP (what's the certificate?). But Cook (from the U of Toronto) showed the following:

## Theorem

*If SAT is poly-time, every problem in NP is poly-time (i.e.  $NP = P$ ).*

## Definition

Problems with the property of Cook's theorem are called *NP-hard*. If a NP-hard problem is also in NP, it is called *NP-complete*.

We strongly suspect that a *NP-hard* problem has no polynomial time solution.

How do we prove that a problem is *NP-hard*?

## Reductions

### Definition

We reduce a problem  $A$  to another problem  $B$  by providing a transformation that

- ▶ takes any instance  $I_A$  of  $A$  and
- ▶ returns an instance  $I_B$  of problem  $B$
- ▶ such that an answer to  $I_B$  gives an answer to  $I_A$ .

### Example

Multiply every value by  $-1$  to reduce finding the maximum to finding the minimum.

e.g.  $\{1, 0, -3, 4, 2, \mathbf{6}\} \Rightarrow \{-1, 0, 3, -4, -2, -\mathbf{6}\}$

If the transformation is easy to compute, problem  $A$  is easier than problem  $B$  because solving  $B$  indirectly solves  $A$ .

## Poly-time reductions

To prove that problem  $B$  is  $NP$ -hard:

- ▶ Find a known  $NP$ -hard problem  $A$  and
- ▶ reduce  $A$  to  $B$  with a poly-time reduction.

Let us be precise.

1. Pick a known  $NP$ -hard decision problem  $A$ .
2. Give a transformation  $T$  that makes any instance  $I_A$  of  $A$  an instance  $I_B$  of  $B$  in time polynomially bounded in the size of  $I_A$
3. The transformation must be such that  $I_A$  is YES if and only if  $I_B$  is YES.

Then a poly-time solution to  $B$  gives a poly-time solution to  $A$  via  $T$ .

## Why should you care?

Because you can

- ▶ stop looking for an efficient algorithm and graduate!
- ▶ justify heuristic methods that give good answers most of the time.
- ▶ justify approximation algorithms that give sub-optimal answers all of the time.
- ▶ prove that other problems are  $NP$ -hard more easily.

# Graph colouring

## Theorem

*Graph 3-coloring is NP-complete.*

## Proof

Clearly 3-colouring is in  $NP$  because given a colouring we can verify that a colouring is valid (i.e. no two adjacent vertices have the same colour) and only uses three colours in poly-time.

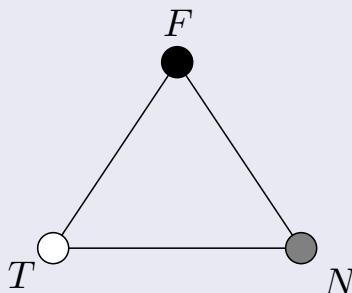
To prove that the problem is  $NP$ -hard, we reduce SAT to graph 3-colouring. The reduction uses colours to represent truth assignments and groups of vertices to represent clauses and variables.

We start with a triangle of labelled vertices  $T, F, N$ . The colour of  $T$  will represent true, the colour  $F$  will represent false, and the colour of  $N$  is neutral.

## Truth assignments

### Proof (cont'd)

We can swap colours of a valid colouring to get another valid colouring. So assume without loss of generality that  $T$  is white,  $F$  is black, and  $N$  is gray in every colouring.



Before we describe the construction in detail, let's analyse one subgraph (widget) that we use over and over again.

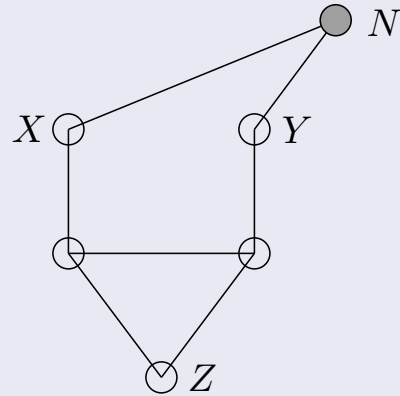
## Widget colouring

### Lemma

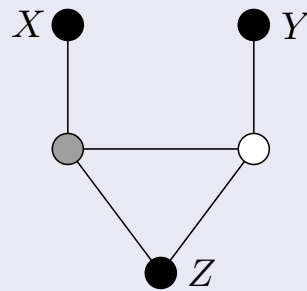
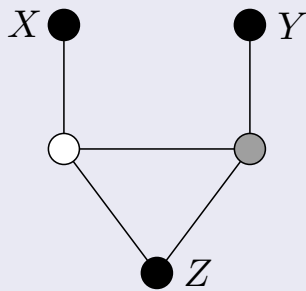
Note that  $X, Y$  are either black or white.

- ▶ If  $Z$  is white, at least one of  $X, Y$  is white.
- ▶ If both  $X, Y$  are black, then  $Z$  is black.

These statements are equivalent.



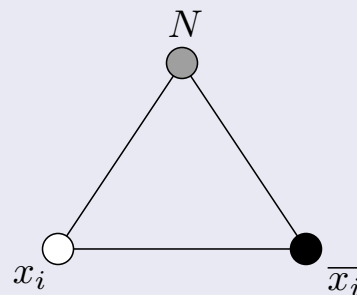
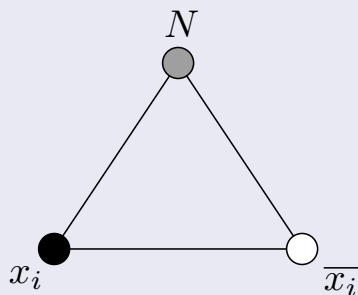
### Proof.



## Variable triangles

### Proof of NP-hardness (cont'd)

For each variable  $x_i$  we introduce two vertices labelled  $x_i$  and  $\overline{x_i}$  that form a triangle with  $N$ . Colouring  $x_i$  white corresponds to assigning the variable  $x_i$  the value of true; colouring it black corresponds to assigning the variable the value of false.

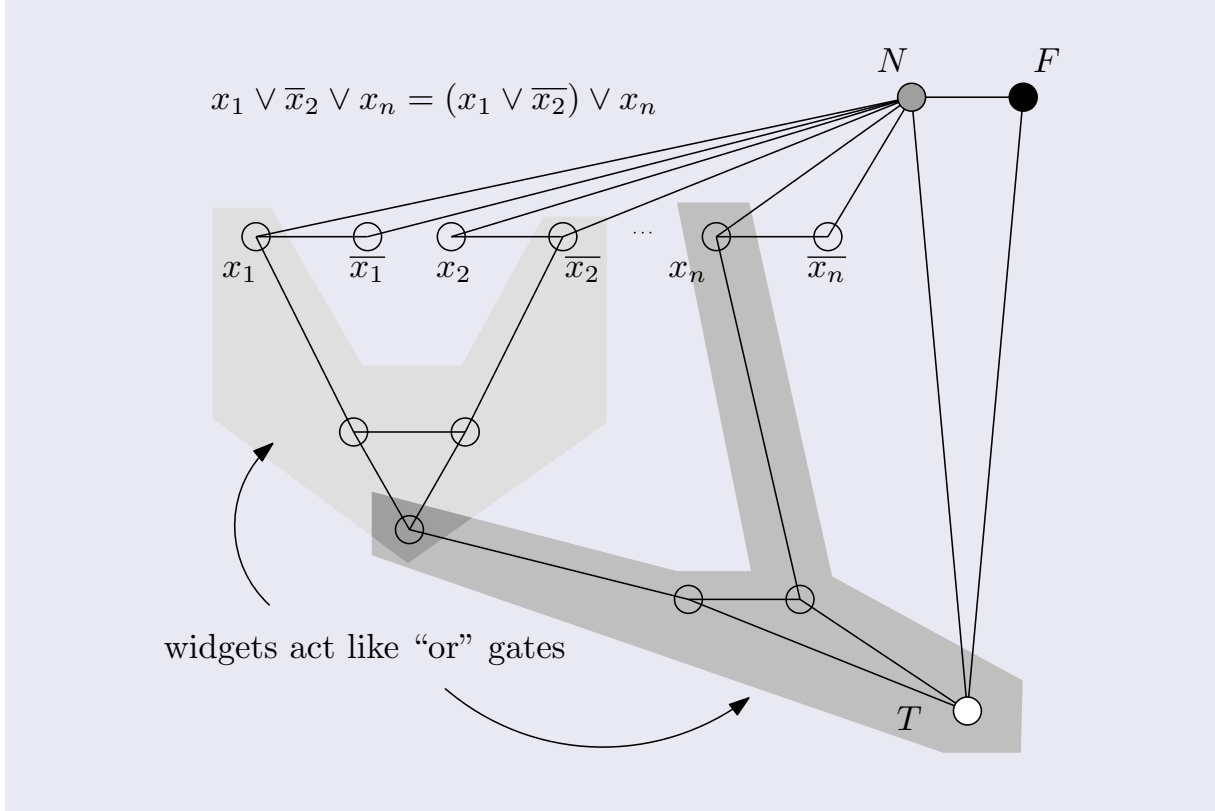


The triangle implies that  $x_i$  is coloured white if and only if  $\overline{x_i}$  is coloured black.



# Clause construction

## Proof (cont'd)



## Analysis

- ▶ The number of vertices and edges is linear in # of variables + # of clauses.
- ▶ The graph structure closely follows SAT instance, so it's a polynomial time construction.
- ▶ A satisfying truth-value-assignment to the SAT instance gives a colouring.
- ▶ That a colouring gives a truth-value-assignment is best seen with an example...