

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.

$O(n!) \subseteq O(n^n) = O(2^{n \log_2 n})$ 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 $EXPTIME \setminus P$.

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 $size(I) = |A| + |x| + \log_2 t$.

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 Π , if, for **every** instance I , $A(I)$ has **finite** runtime and returns the correct answer

If an algorithm A solves decision problem Π , then we say Π is **decidable**.

Formally, Π 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 Π , then we say Π is **undecidable**.

Formally, Π 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, Π 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?"

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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
  { i ← 1
  while i ≠ 0 do
    { 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
    }
    i ← i + 1
  }

```

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**.

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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 $Halt \leq^T Halt-All$, 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

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

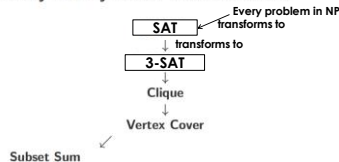
so we can solve the halting problem.

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

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FINISHING NPC TRANSFORMATIONS/REDUCTIONS

Summary of Polynomial Transformations



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Summary of Polynomial Transformations



REDUCE TARGET SUBSET SUM TO 0-1 KNAPSACK

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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 $\leq M$?

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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 $\sum_{i \in J} s_i = T$?

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 $\sum p_i x_i \geq T$?

How should we poly-transform (Target) Subset-Sum input into (Target) 0-1 Knapsack input
 Such that: I contains a subset that sums to T IFF $(\geq T)$ profit can be obtained in knapsack input $f(I)$

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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

$$p_i = s_i, 1 \leq i \leq n$$

$$w_i = s_i, 1 \leq i \leq n$$

$$M = T$$

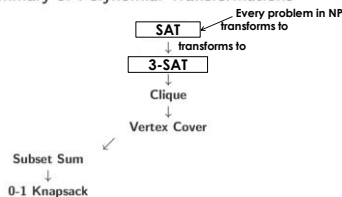
Then define $f(I)$ to be the instance of **0-1 Knapsack** consisting of profits $[p_1, \dots, p_n]$, weights $[w_1, \dots, w_n]$, capacity M and target profit T .

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 $f(I)$

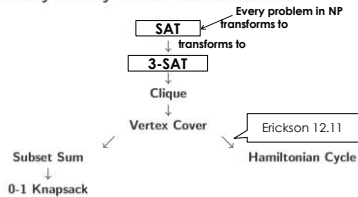
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Summary of Polynomial Transformations



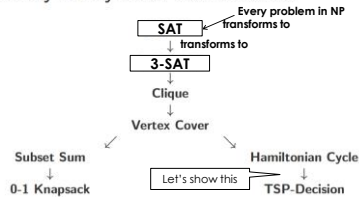
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Summary of Polynomial Transformations



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Summary of Polynomial Transformations



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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

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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:

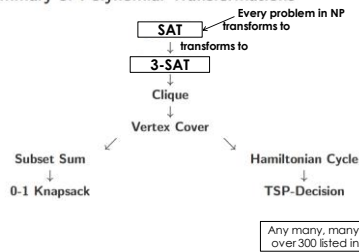
$$w(uv) = \begin{cases} 1 & \text{if } uv \in E \\ 2 & \text{if } uv \notin E. \end{cases}$$

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

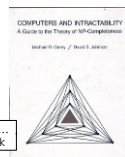
Exercise: Prove the correctness of this transformation.

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Summary of Polynomial Transformations



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



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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.

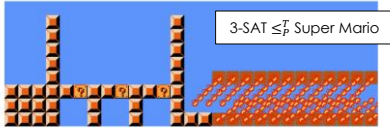


Figure 11: Clause gadget for Super Mario Bros.

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(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...

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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 **poly(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 **poly(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 Π in poly-time
 - NP-hard** (at least as hard as NPC) All of NPC, and e.g., TSP-optimization, TSP-optimal value
 - problems Π s.t. every $\Pi' \in NP$ can be **reduced** to Π in poly-time
- Note: P, NP and NPC problems are **decidable**

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POLYTIME 2-SAT
(IF WE HAVE TIME)

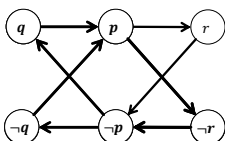
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2-SAT EXAMPLES

- $(p \vee q) \wedge (\neg p \vee r) \wedge (\neg r \vee \neg p)$
 - 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 $\neg q \Rightarrow p$

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



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!

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

- 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 $X \cup \bar{X}$.
- Determine the strongly connected components of this directed graph.
- 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 $x = \text{false}$ else set $x = \text{true}$.

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