

CS 341: ALGORITHMS

Lecture 20: intractability II – complexity class NP

Readings: see website

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THIS TIME

- **Finishing TSP reductions**
- Complexity class **NP**
 - Oracles, certificates, polytime verification algorithms

RECALL

- So far we know
 - TSP-Dec \leq_p^T TSP-Optimal Value
 - TSP-Dec \leq_p^T TSP-Optimization
- **In progress**
 - **TSP-Optimal Value \leq_p^T TSP-Dec**

Travelling Salesperson Problems

Problem 7.5

TSP-Optimization

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

Find: A hamiltonian cycle H in G such that $w(H) = \sum_{e \in H} w(e)$ is minimized.

Problem 7.6

TSP-Optimal Value

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

Find: The minimum T such that there exists a hamiltonian cycle H in G with $w(H) = T$.

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 $w(H) \leq T$?

What's the size of the input $I = (G, w)$?

$$Size(I) = Size(G) + Size(w)$$

So, suppose G is represented as an **array of adjacency lists** (one list for each vertex), with each list containing **edges** to neighbouring vertices, and an edge is represented by a **weight** and the **name** of the target vertex

Bits to store **weight** of the edge
(storing $w(e)$ takes $\log w(e) + 1$ bits)

Bits to store the **name** of the target vertex (in $1..|V|$)

$$Size(I) = |V| + \underbrace{\sum_{e \in E} (\log w(e) + 1 + \log|V| + 1)}_{\text{For all edges}}$$

Array of empty lists for all vertices v

Let's relate this to runtime... what's the runtime?

TSP-Optimal Value \leq_{P}^{T} TSP-Dec

Let's assume $O(1)$ time for operations on weights

Technically not needed to show polytime.. But simplifies

Algorithm: *TSP-OptimalValue-Solver*(G, w)

external *TSP-Dec-Solver*

$hi \leftarrow \sum_{e \in E} w(e)$

$O(|E|)$

$lo \leftarrow 0$

$O(1)$

if not *TSP-Dec-Solver*(G, w, hi) **then return** (∞)

$O(1)$ for the oracle

while $hi > lo$

iterations: $O(\log(hi - lo))$

do {
 $mid \leftarrow \lfloor \frac{hi+lo}{2} \rfloor$
 if *TSP-Dec-Solver*(G, w, mid)
 then $hi \leftarrow mid$
 else $lo \leftarrow mid + 1$

$= \log \sum_{e \in E} w(e)$

$O(1)$

return (hi)

Runtime $T(I) \in O(|E| + \log \sum_{e \in E} w(e))$

COMPARING $T(I)$ AND $Size(I)$

- $T(I) \in O(|E| + \log \sum_{e \in E} w(e))$
- $Size(I) = |V| + \sum_{e \in E} (\log w(e) + 1 + \log|V| + 1)$
 $= |V| + \sum_{e \in E} (\log w(e) + 1) + \sum_{e \in E} (\log|V| + 1)$
 $= |V| + \sum_{e \in E} (\log w(e) + 1) + \sum_{e \in E} (\log|V|) + |E|$
- Want to show $T(I) \in O(Size(I)^c)$ for some constant c (we show $c=1$)
 $O(|E| + \log \sum_{e \in E} w(e)) \stackrel{?}{\subseteq} O(|V| + \sum_{e \in E} (\log w(e) + 1) + \sum_{e \in E} \log|V| + |E|)$
 $\Leftrightarrow O(\log \sum_{e \in E} w(e)) \stackrel{?}{\subseteq} O(|V| + \sum_{e \in E} (\log w(e) + 1) + \sum_{e \in E} \log|V|)$
How to compare $\log \sum_{e \in E} w(e)$ and $\sum_{e \in E} (\log w(e) + 1)$?

COMPARING $T(I)$ AND $Size(I)$

- **How to compare $\log \sum_{e \in E} w(e)$ and $\sum_{e \in E} (\log w(e) + 1)$?**
- $\sum_{e \in E} (\log w(e) + 1) = (\log w(e_1) + 1) + (\log w(e_2) + 1) + \dots + (\log (w(e_{|E|})) + 1)$
- Can we combine these terms into one log using $\log x + \log y = \log xy$?
- $\sum_{e \in E} (\log w(e) + 1) = (\log w(e_1) + \log 2) + \dots + (\log (w(e_{|E|})) + \log 2)$
- $\sum_{e \in E} (\log w(e) + 1) = \log 2w(e_1) 2w(e_2) \dots 2w(e_{|E|}) = \log \prod_{e \in E} 2w(e)$
- So how to compare $\log \prod_{e \in E} 2w(e)$ and $\log \sum_{e \in E} w(e)$?
 - **All $w(e)$ are positive integers, so $\prod_{e \in E} 2w(e) \geq \sum_{e \in E} w(e)$**
 - Since log is increasing on \mathbb{Z}^+ , **$\log \prod_{e \in E} 2w(e) \geq \log \sum_{e \in E} w(e)$**

COMPARING $T(I)$ AND $Size(I)$

- We in fact show $T(I) \in O(Size(I))$

$$O(\log \sum_{e \in E} w(e)) \stackrel{?}{\subseteq} O(|V| + \sum_{e \in E} (\log w(e) + 1) + \sum_{e \in E} \log |V|)$$

How to compare $\log \sum_{e \in E} w(e)$ and $\sum_{e \in E} (\log w(e) + 1)$?

We just saw $\sum_{e \in E} (\log w(e) + 1) = \log \prod_{e \in E} 2w(e) \geq \log \sum_{e \in E} w(e)$

So $T(I) \in O(Size(I)^c)$ where $c = 1$

So this reduction has runtime that is polynomial in the input size!

TSP-Optimal Value $\leq_{T_P}^T$ **TSP-Dec**

Algorithm: *TSP-OptimalValue-Solver*(G, w)
external *TSP-Dec-Solver*
 $hi \leftarrow \sum_{e \in E} w(e)$
 $lo \leftarrow 0$
if not *TSP-Dec-Solver*(G, w, hi) **then return** (∞)
while $hi > lo$
 do $\left\{ \begin{array}{l} mid \leftarrow \lfloor \frac{hi+lo}{2} \rfloor \\ \text{if } TSP-Dec-Solver(G, w, mid) \\ \text{then } hi \leftarrow mid \\ \text{else } lo \leftarrow mid + 1 \end{array} \right.$
return (hi)

So TSP-OptimalValue-Solver is polytime... But is it a **correct reduction from TSP-Optimal Value to TSP-Dec?**

Need to prove:
TSP-OptimalValue-Solver(G, w)
returns the weight **W**
of the **shortest Hamiltonian Cycle (HC)** in **G**

Sketch: We return ∞ iff there is **no HC**.
Key loop invariant: $W \in [lo, hi]$.
So, at termination when $hi = lo$,
we return exactly $hi = W$.

TSP-Optimal Value \leq_P^T **TSP-Dec**

Algorithm: *TSP-OptimalValue-Solver*(G, w)

external *TSP-Dec-Solver*

$hi \leftarrow \sum_{e \in E} w(e)$

$lo \leftarrow 0$

if not *TSP-Dec-Solver*(G, w, hi) **then return** (∞)

while $hi > lo$

do $\left\{ \begin{array}{l} mid \leftarrow \lfloor \frac{hi+lo}{2} \rfloor \\ \mathbf{if} \ iTSP-Dec-Solver(G, w, mid) \\ \quad \mathbf{then} \ hi \leftarrow mid \\ \quad \mathbf{else} \ lo \leftarrow mid + 1 \end{array} \right.$

return (hi)

So, TSP-OptimalValue-Solver is **polytime**,
and is a **correct** reduction.

We have therefore shown:
**TSP-Optimal Value is polytime
reducible to TSP-Dec**

So, if an **$O(1)$** implementation of TSP-Dec-Solver
exists, then we have a **polytime** implementation of
TSP-Optimal-Value-Solver!

In fact, TSP-OptimalValue-Solver remains
polytime even if the implementation of the
oracle runs in polytime instead of $O(1)$! (bonus slides)

PROVING REDUCTIONS CORRECT

- **In more complex reductions** where we **transform the input** before calling the oracle, we will need a **more complex proof**:
- (A) If there is a(n optimal) solution in the input, our transformation will preserve that solution so the oracle can find it, and
- (B) Our transformation doesn't introduce new solutions that are **not** present in the original input
 - *(i.e., if we find a solution in the transformed input, there was a corresponding solution in the original input)*

More on this later...

INPUT SIZE CHEAT SHEET

Exponentially larger than optimal representation!

Input I	Perfectly fine choices of $Size(I)$
int x	1 or $\lceil \log(x) \rceil + 1$ (can simplify to $\log(x) + 1$ or $\log x$)
Graph (V, E) with weights W :	$ V $ or $ E $ or $ V ^2$ or $ V + E $ or $\sum_{e \in E} (\log(w(e)) + 1)$ or $\sum_{u, v \in V} (\log(w(u, v)) + 1)$ or any sum of terms above
$A[1..n]$ of int	n or $\sum_i (\log(A[i]) + 1)$
$n \times n$ matrix m	n^2 or $\sum_{i, j} (\log(m_{ij}) + 1)$

To write down $x=1$, need $\log(1)+1=1$ bit.
For $x=2$ this is 2 bits.
For $x=4$, 3 bits.

Pick any expression that makes your analysis easy

Input I	Examples of BAD choices of $Size(I)$
int x	x
Graph (V, E)	$2^{ V }$ or $ V ^{ E }$ or $\sum_{e \in E} w(e)$
$A[1..n]$ of int	2^n or $\sum_i A[i]$

Pseudo-polynomial \sim no exponentiation of non-constant terms

Technically any **pseudo-polynomial combination** of these terms is fine.
For example, the following is fine:
 $(|E|^{100} + |V|^2) \cdot \sum_{e \in E} (\log(w(e)) + 1)$

- So far we know
 - $\text{TSP-Dec} \leq \frac{T}{P} \text{TSP-Optimal Value}$
 - $\text{TSP-Dec} \leq \frac{T}{P} \text{TSP-Optimization}$
 - $\text{TSP-Optimal Value} \leq \frac{T}{P} \text{TSP-Dec}$
- Let's show
 - $\text{TSP-Optimization} \leq \frac{T}{P} \text{TSP-Dec}$

WHAT ABOUT REDUCING TSP-OPTIMIZATION TO TSP-DEC?

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 $w(H) \leq T$?

Need to return the **actual** minimum Hamiltonian Cycle!

We already know how to get the **weight T^*** of the minimum HC...

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 $w(H) \leq T$?

Given only a **single bit** of information **per call** to the oracle

Idea: Use **T^*** along with calls to the oracle to somehow figure out **which edges** are involved in the minimum HC?

TSP-Optimization \leq_P^T TSP-Dec

```
Algorithm: TSP-Optimization-Solver( $G = (V, E), w$ )
external TSP-OptimalValue-Solver, TSP-Dec-Solver
 $T^* \leftarrow \text{TSP-OptimalValue-Solver}(G, w)$ 
if  $T^* = \infty$  then return ("no hamiltonian cycle exists")
 $w_0 \leftarrow w$ 
 $H \leftarrow \emptyset$ 
for all  $e \in E$ 
do {
   $w_0[e] \leftarrow \infty$ 
  if not TSP-Dec-Solver( $G, w_0, T^*$ )
  then {
     $w_0[e] \leftarrow w[e]$ 
     $H \leftarrow H \cup \{e\}$ 
  }
}
return ( $H$ )
```

To remove any dependence on this "other oracle," simply replace this call with the reduction **code** we showed

Already know this call is poly-time reducible to TSP-Dec!

If removing edge e removes **every** Hamiltonian cycle of minimum weight

then e is part of **every** minimum Hamiltonian cycle, and we add it to H (and add it back into the graph)

At the end, the graph contains precisely the edges that are needed to produce a minimum HC

[Correctness] **Loop invariant:** there exists a HC of weight T^* in w_0

By the end of the loop, H contains all finite edges in w_0

So some HC C of weight T^* **is contained in H**

At the end of the algorithm, there is a Hamiltonian Cycle \mathcal{C} of optimal weight T^* **contained in** H

If H is **precisely** \mathcal{C} , then we are done.
Suppose not to obtain a contradiction.

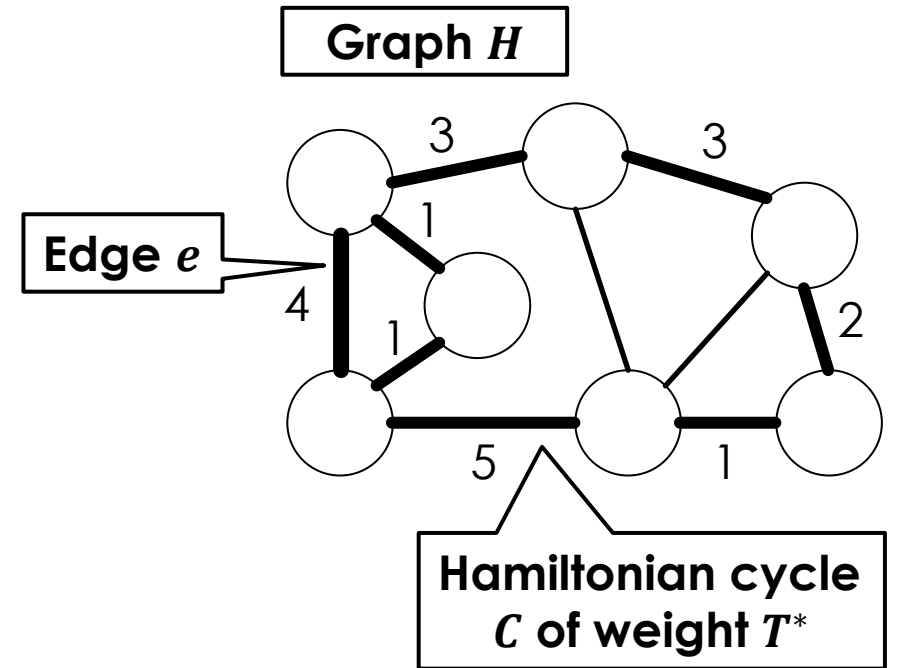
In this case, there are some **other edges** in H as well.

Let e be one such edge.

Consider the iteration when e was processed.
Note e was **not removed** in this iteration!

Doing so would remove **all** Hamiltonian Cycles of weight T^* ,
including \mathcal{C} .

This means the edge must be part of \mathcal{C} ---contradiction!



TSP-Optimization \leq_P^T TSP-Dec

Algorithm: *TSP-Optimization-Solver*($G = (V, E), w$)

external *TSP-OptimalValue-Solver*, *TSP-Dec-Solver*

$T^* \leftarrow \text{TSP-OptimalValue-Solver}(G, w)$ $\text{poly}(\text{Size}(I))$

if $T^* = \infty$ then return ("no hamiltonian cycle exists") $O(1)$

$w_0 \leftarrow w$ $O(m)$ to copy matrix

$H \leftarrow \emptyset$ $O(1)$ to create list

for all $e \in E$

do $O(m)$ iterations $O(1)$ per iteration

$\left\{ \begin{array}{l} w_0[e] \leftarrow \infty \\ \text{if not } \text{TSP-Dec-Solver}(G, w_0, T^*) \\ \text{then } \left\{ \begin{array}{l} w_0[e] \leftarrow w[e] \\ H \leftarrow H \cup \{e\} \end{array} \right. \end{array} \right.$

return (H)

So this is a **correct** reduction.
Is it a **polytime reduction**?

What's the runtime?

Let's assume unit costs for simplicity

Runtime = $\text{poly}(\text{Size}(I)) + O(m)$

What's $\text{Size}(I)$?

(What's a "useful" lower bound?)

$\text{Size}(I) = \Omega(|E|) = \Omega(m)$

Clearly $O(m) \in O(\text{Size}(I)^1)$
So runtime is in $\text{poly}(\text{Size}(I))$

So **yes**, this is a **polytime reduction**

What would change if we precisely counted the number of bits in each edge, weight, etc., in $\text{Size}(I)$?

What if **operations on weight w** took $O(\log w)$ time? (bonus slides)

RECAP

- Showed three flavours of TSP are **polytime-equivalent** (i.e., if you can solve one flavour in polytime, you can solve all three flavours in polytime)
 - One of these was a decision problem (yes/no), and the other two were not (total weight, actual cycle)
- **Decision and non-decision flavours** of a problem are often polytime-equivalent
- Proofs for a **polytime Turing reduction**
 - **Correctness** (return value is correct for every possible input)
 - **Polytime** (runtime is polynomial in the input size)
[or poly(some lower bound on the input size)]

Note: only one of my
sections got here

COMPLEXITY CLASS NP

NP: Non-deterministic polynomial time

EXAMPLE: SUBSET-SUM PROBLEM

- Suppose we are given some integers, -7, -3, -2, 5, 8
- Does **some** subset of these **sum to zero**?
 - In this case, yes: $(-3) + (-2) + 5 = 0$

Finding such a subset can be extremely difficult

Suppose I give you a **certificate** consisting of an array of numbers, and **claim** it represents such a subset

Of course, I might lie and give you a subset that does **not sum to zero**...

If I'm telling the truth, then we call this a **yes-certificate**. It is essentially a **proof** that "yes" is the correct output.

I could even give you numbers that are **not in the input**...

Can you use a yes-certificate to solve the problem efficiently?

Can you determine whether I am lying in polynomial time?

SUBSET-SUM VIA NON-DETERMINISTIC ORACLE

- Suppose there is a **non-deterministic oracle**, which returns **a subset that sums to 0 if one exists** and otherwise can return **anything** (even garbage)
- We call the oracle's output a **certificate**
- Given a **certificate**, can you **verify in polytime** whether it describes a solution to the problem?

```
1 SubsetSumWithOracle(I)
2   C = Oracle(I)
3   return verify(I, C)
4
5 verify(I, C)
6   if C not subset of I then return false
7   return (sum(C) == 0)
```

Otherwise, either C is not a subset of the input (return false), or C sums to a non-zero value (return false)

If there **exists** a subset that sums to 0, then **C** is one such subset, and we return **true**

Given such an oracle, this algorithm would **solve** subset-sum

“Non-deterministic” is the **N** in **NP**, and it is so named because of oracles

Here **“non-deterministic”** just means the oracle is magically guaranteed to return a yes-certificate if one exists

BONUS SLIDES

TSP-Optimal Value \leq_P^T TSP-Dec

Algorithm: *TSP-OptimalValue-Solver*(G, w)

external *TSP-Dec-Solver*

$hi \leftarrow \sum_{e \in E} w(e)$

$lo \leftarrow 0$

if not *TSP-Dec-Solver*(G, w, hi) **then return** (∞)

while $hi > lo$

do $\left\{ \begin{array}{l} mid \leftarrow \lfloor \frac{hi+lo}{2} \rfloor \\ \text{if } TSP-Dec-Solver(G, w, mid) \\ \text{then } hi \leftarrow mid \\ \text{else } lo \leftarrow mid + 1 \end{array} \right.$

return (hi)

TSP-OptimalValue-Solver remains **polytime** even if the **oracle runs in polytime** instead of $O(1)$!

The key idea is: Consider polynomials $P_R(s)$ and $P_O(s)$ representing the runtime of a reduction and its oracle, respectively, on an input of size s .

Worst possible runtime happens if **every step** in the reduction is a call to the oracle.

This is $P_R(s)P_O(s)$ --- multiplication of polynomials.

But **multiplying polynomials** of degrees d_1, d_2 results in a **polynomial** of degree $\leq d_1 + d_2$. **Example:**

$$P_1(x) = 5x^2 + 10x + 100$$

$$P_2(x) = 20x^3 + 20$$

$$\begin{aligned} P_1(x)P_2(x) &= (5x^2 + 10x + 100)(20x^3 + 20) \\ &= 100x^5 + 200x^4 + 2000x^3 + 100x^2 + 200x + 2000 \end{aligned}$$

Let's assume $O(\log w)$ time for reading/writing/arithmetic operations on each weight w (and $O(\log w)$ space).

So this is a **correct** reduction.
Is it a **polytime reduction**?

What's the runtime on such an input?

$$\text{Runtime} = \text{poly}(\text{Size}(I)) + O(m + \sum_{u,v \in V} \log w(u,v))$$

Algorithm: *TSP-Optimization-Solver*($G = (V, E), w$)

external *TSP-OptimalValue-Solver*, *TSP-Dec-Solver*

$T^* \leftarrow \text{TSP-OptimalValue-Solver}(G, w)$

Suppose we show this is $\text{poly}(\text{Size}(I))$

if $T^* = \infty$ then return ("no hamiltonian cycle exists")

$w_0 \leftarrow w$ $O(\sum_{u,v \in V} \log w(u,v))$ to copy matrix

$O(1)$

$H \leftarrow \emptyset$ $O(1)$ to create list

for all $e \in E$ $O(m)$ iterations: for all u, v

$w_0[e] \leftarrow \infty$ $O(\log w(u,v))$

do if not *TSP-Dec-Solver*(G, w_0, T^*) $O(1)$

then $\begin{cases} w_0[e] \leftarrow w[e] \\ H \leftarrow H \cup \{e\} \end{cases}$ $O(\log w(u,v))$

return (H) $O(1)$

What's $\text{Size}(I)$?

(or a useful lower bound on it)

$$\text{Size}(I) = O(|E| + \sum_{u,v \in V} \log w(u,v))$$

Clearly $O(m + \sum_{u,v \in V} \log w(u,v)) \in \text{poly}(\text{Size}(I))$

So, this is **still** a **polytime reduction**

Unit cost vs non-unit cost assumptions usually do **not** usually make a difference...

This should not be surprising, since the same $O(\log w)$ terms are introduced into both space and time complexities...