# CSE 332: Data Structures \& Parallelism Lecture 22: P, NP, NP-Complete 



Arthur Liu
Summer 2022

## Announcements

- Reminder your final is on two days, Section 10/18, Lecture 10/19
- Make sure to be in your correct quiz section on Thursday for pt1. of the exam! We will take attendance, so bring student ID to section
- Final Review Session: MOR 220 Wed 10/17 from 3:00-4:00pm
- Exam Topics and Practice Exams on the website!
- Make sure to look at some past finals to practice!


## Class Survey!

- You should have received an email for a survey for this class! - It closes this Friday!
- I will give out 1 extra credit point to everyone who fills it out
- It is anonymous, I will know if you filled it out but not what you said


## Outline

- A Few Problems:
- Euler Circuits
- Hamiltonian Circuits
- $P$ and NP
- NP-Complete
- What now?



## Try it



Which of these can you draw (trace all edges) without lifting your pencil, drawing each line only once?
Can you start and end at the same point?

## Your First Task

- Your company has to inspect a set of roads between cities by driving over each of them.
- Driving over the roads costs money (fuel), and there are a lot of roads.
- Your boss wants you to figure out how to drive over each road exactly once, returning to your starting point.



## Euler Circuits

- Euler circuit: a path through a graph that visits each edge exactly once and starts and ends at the same vertex
- Named after Leonhard Euler (1707-1783), who cracked this problem and founded graph theory in 1736
- A Euler circuit exists iff
- the graph is connected and
- each vertex has even degree (= \# of edges on the vertex)


## Euler Circuit Example



Euler(A) :

## Euler Circuit Example



Euler(A) :
ABGEDGCA

## Euler Circuit Example



Euler(A) :
ABGEDGCA


Euler(B)

## Euler Circuit Example



Euler(A) :
ABGEDGCA


Euler(B): BDFECB

## Euler Circuit Example



Euler(A) :
ABGEDGCA
Splice

## The Road Inspector: Finding Euler Circuits

Given a connected, undirected graph $G=(V, E)$, find an Euler circuit in $G$

Can check if one exists:

- Check if all vertices have even degree


## Basic Euler Circuit Algorithm:

1. Do an edge walk from a start vertex until you are back to the start vertex.

- You never get stuck because of the even degree property.

2. "Remove" the walk, leaving several components each with the even degree property.

- Recursively find Euler circuits for these.

3. Splice all these circuits into a Euler circuit

> Running time?


## The Road Inspector: Finding Euler Circuits

Given a connected, undirected graph $G=(V, E)$, find an Euler circuit in $G$

Can check if one exists: (in $\mathrm{O}(|\mathrm{V}|+|\mathrm{E}|)$ )

- Check if all vertices have even degree


## Basic Euler Circuit Algorithm:

1. Do an edge walk from a start vertex until you are back to the start vertex.

- You never get stuck because of the even degree property.

2. "Remove" the walk, leaving several components each with the even degree property.

- Recursively find Euler circuits for these.

3. Splice all these circuits into a Euler circuit

$$
\text { Running time? } \mathrm{O}(|\mathrm{~V}|+|\mathrm{E}|)
$$



## Your Second Task

- Your boss is pleased...and assigns you a new task.
- Your company has to send someone by car to a set of cities.
- The primary cost is the exorbitant toll going into each city.
- Your boss wants you to figure out how to drive to each city exactly once, returning in the end to the city of origin.


## Hamiltonian Circuits

- Euler circuit: A cycle that goes through each edge exactly once
- Hamiltonian circuit: A cycle that goes through each vertex exactly once
- Does graph I have:
- An Euler circuit?

- A Hamiltonian circuit?
- Does graph II have:
- An Euler circuit?
- A Hamiltonian circuit?
- Which problem sounds harder?



## Finding Hamiltonian Circuits

- Problem: Find a Hamiltonian circuit in a connected, undirected graph G
- One solution: Search through all paths to find one that visits each vertex exactly once
- Can use your favorite graph search algorithm to find paths
- This is an exhaustive search ("brute force") algorithm
- Worst case: need to search all paths
- How many paths??


## Analysis of Exhaustive Search Algorithm

Worst case: need to search all paths

- How many paths?

Can depict these paths as a search tree:


Search tree of paths from B

## Analysis of Exhaustive Search Algorithm

Let the average branching factor of each node in this tree be b
$|V|$ vertices, each with $\approx b$ branches

Total number of paths $\approx b \cdot b \cdot b . . . \cdot b$

Worst case $\rightarrow$


Search tree of paths from B

## Analysis of Exhaustive Search Algorithm

Let the average branching factor of each node in this tree be b
$|V|$ vertices, each with $\approx b$ branches

Total number of paths $\approx b \cdot b \cdot b . . . \cdot b$ $O\left(b^{|V|}\right)$

Worst case $\rightarrow$ Exponential time!


Search tree of paths from B

## Running Times




## More Running Times

Table 2.1 The running times (rounded up) of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second. In cases where the running time exceeds $10^{25}$ years, we simply record the algorithm as taking a very long time.

|  | $n$ | $n \log _{2} n$ | $n^{2}$ | $n^{3}$ | $1.5^{n}$ | $2^{n}$ | $n!$ |
| :---: | ---: | ---: | :---: | :---: | :---: | :---: | ---: |
| $n=10$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | 4 sec |
| $n=30$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | 18 min | $10^{25}$ years |
| $n=50$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | 11 min | 36 years | very long |
| $n=100$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | 1 sec | 12,892 years | $10^{17}$ years | very long |
| $n=1,000$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | 1 sec | 18 min | very long | very long | very long |
| $n=10,000$ | $<1 \mathrm{sec}$ | $<1 \mathrm{sec}$ | 2 min | 12 days | very long | very long | very long |
| $n=100,000$ | $<1 \mathrm{sec}$ | 2 sec | 3 hours | 32 years | very long | very long | very long |
| $n=1,000,000$ | 1 sec | 20 sec | 12 days | 31,710 years | very long | very long | very long |

[^0]
## Polynomial vs. Exponential Time

- All of the algorithms we have discussed in this class have been polynomial time algorithms:
- Examples: $\mathrm{O}(\log \mathrm{N}), \mathrm{O}(\mathrm{N}), \mathrm{O}(\mathrm{N} \log \mathrm{N}), \mathrm{O}\left(\mathrm{N}^{2}\right)$
- Algorithms whose running time is $0\left(\mathrm{~N}^{k}\right)$ for some $\mathrm{k}>0$
- Exponential time $b^{N}$ is asymptotically worse than any polynomial function $\mathrm{N}^{\mathrm{k}}$ for any k



## The Complexity Class P

Definition: P is the set of all problems that can be solved in polynomial worst-case time

All problems that have some algorithm whose running time is $\mathrm{O}\left(\mathrm{N}^{k}\right)$ for some $k$

Examples of problems in P :
sorting, shortest path, Euler circuit, etc.



Hamiltonian Circuit


Hamiltonian Circuit
Satisfiability (SAT)
Vertex Cover
Travelling Salesman

## Satisfiability

Input: a logic formula of size $m$ containing $n$ variables
Output: An assignment of Boolean values to the variables in the formula such that the formula is true

Algorithm: Try every variable assignment

$$
\left(\neg x_{1} \vee x_{2} \vee x_{4}\right) \wedge\left(x_{1} \vee \neg x_{3} \vee x_{4}\right) \wedge\left(x_{2} \vee \neg x_{4} \vee \neg x_{5}\right)
$$

## Vertex Cover

Input: A graph (V,E) and a number m
Output: A subset S of V such that for every edge (u,v) in E, at least one of $\mathbf{u}$ or $\mathbf{v}$ is in $\mathbf{S}$ and $|S|=m$ (if such an $S$ exists)

Algorithm: Try every subset of vertices of size m


## Traveling Salesman

Input: A complete weighted graph (V,E) and a number m
Output: A circuit that visits each vertex exactly once and has total cost < m if one exists

Algorithm: Try every path, stop if find cheap enough one

## A Glimmer of Hope

If given a candidate solution to a problem, we can check if that solution is correct in polynomial-time, then maybe a polynomial-time solution exists?

Can we do this with Hamiltonian Circuit?
Given a candidate path, is it a Hamiltonian Circuit?

## A Glimmer of Hope

If given a candidate solution to a problem, we can check if that solution is correct in polynomial-time, then maybe a polynomial-time solution exists?

Can we do this with Hamiltonian Circuit?
Given a candidate path, is it a Hamiltonian Circuit? just check if all vertices are visited exactly once in the candidate path

## The Complexity Class NP

Definition: NP is the set of all problems for which a given candidate solution can be tested in polynomial time

## Examples of problems in NP:

Hamiltonian circuit: Given a candidate path, can test in linear time if it is a Hamiltonian circuit

Vertex Cover: Given a subset of vertices, do they cover all edges?
All problems that are in $P$ (why?)


## Why do we call it "NP"?

- NP stands for Nondeterministic Polynomial time
- Why "nondeterministic"? Corresponds to algorithms that can guess a solution (if it exists), the solution is then verified to be correct in polynomial time
- Can also think of as allowing a special operation that allows the algorithm to magically guess the right choice at each branch point.
- Nondeterministic algorithms don't exist - purely theoretical idea invented to understand how hard a problem could be


## It does NOT stand for "non-polynomial"

## Reductions

- Let's say we want to make some claim about NP problems, we would want to pick the "most difficult" NP problem as our representative.
- What does it mean for one problem to be harder than another?


## Polynomial Time Reducible

We say $A$ reduces to $B$ in polynomial time, if there is an algorithm that, using a (hypothetical) polynomial-time algorithm for $B$, solves problem $A$ in polynomial-time.

## Another way of thinking about it

- Problem A reduces to Problem B
- Problem A "can be converted" to Problem B
- Problem B is the "broader, harder" problem.
- If we can solve problem B, we can solve problem A.

Problem A: I want to solve a math equation with only addition


Problem B: I want to solve a math equation with any operators

## NP-complete

- Let's say we want to make some claim about NP problems, we would want to pick the "most difficult" NP problem as our representative.
- What does it mean for one problem to be harder than another?

> | NP-complete |
| :--- |
| a problem $B$ is NP-complete if $B$ is in NP and |
| for all problems $A$ in NP, $A$ reduces to $B$ in polynomial time. |

Interesting fact: If any one NP-complete problem could be solved in polynomial time, then all NP-complete problems could be solved in polynomial time...
...and all NP problems can be solved in polynomial time

## What The World Looks Like (We Think)



## One more class - NP-Hard

```
NP-complete
a problem \(B\) is NP-complete if \(B\) is in NP and for all problems \(A\) in NP, \(A\) reduces to \(B\) in polynomial time.
```

```
NP-Hard
a problem B is NP-hard if
for all problems A in NP, A reduces to B in polynomial time.
```


## What The World Looks Like (We Think)



## Your Chance to Win a Turing Award!

It is generally believed that $P \neq N P$,
i.e. prove there are problems in NP that are not in $P$

- But no one has been able to show even one such problem!
- This is the fundamental open problem in theoretical computer science
- Nearly everyone has given up trying to prove it. Instead, theoreticians prove theorems about what follows once we assume $P \neq N P$ !


## Saving Your Job

Try as you might, every solution you come up with for the Hamiltonian Circuit problem runs in exponential time.....

- You have to report back to your boss.
- Your options:
- Keep working
- Come up with an alternative plan...


## In general, what to do with a Hard Problem

- Your problem seems really hard.
- If you can transform a known NP-complete problem into the one you're trying to solve, then you can stop working on your problem!


## Your Third Task

- Your boss buys your story that others couldn't solve the last problem.
- Again, your company has to send someone by car to a set of cities. There is a road between every pair of cities.
- The primary cost is distance traveled (which translates to fuel costs).
- Your boss wants you to figure out how to drive to each city exactly once, then return to the first city, while staying within a fixed mileage budget $k$.


## Travelling Salesman Problem (TSP)

- Your third task is basically TSP:
- Given complete weighted graph G, integer k.
- Is there a cycle that visits all vertices with cost $<=k$ ?
- One of the canonical problems.
- Note difference from Hamiltonian cycle:
- graph is complete
- we care about weight.


## Transforming Hamiltonian Cycle to TSP

- We can "reduce" Hamiltonian Cycle to TSP.
- Given graph $\mathrm{G}=(\mathrm{V}, \mathrm{E})$ :
- Assign weight of 1 to each edge
- Augment the graph with edges until it is a complete graph G'=(V, E')
- Assign weights of 2 to the new edges
- Let $\mathrm{k}=|\mathrm{V}|$.

Notes:

- The transformation must take polynomial time
- You reduce the known NP-complete problem into your problem (not the other way around)
- In this case we are assuming Hamiltonian Cycle is our known NP-complete problem (in reality, both are known NP-complete)


## Example



G
Input to Hamiltonian
Circuit Problem

## Example



G
Input to Hamiltonian
Circuit Problem


Polynomial time transformation

Input to Traveling Salesman Problem

## Polynomial-time transformation

- G' has a TSP tour of weight |V| iff G has a Hamiltonian Cycle.
- What was the cost of transforming HC into TSP?
- In the end, because there is a polynomial time transformation from HC to TSP, we say TSP is "at least as hard as" Hamiltonian cycle.


## NP-Complete Problems

But Wait! There's more!
By 1979, at least 300 problems had been proven NP-complete.

Garey and Johnson put a list of all the NPcomplete problems they could find in this textbook.

Took them almost 100 pages to just list them all.

## COMPUTERS AND INTRACTABILITY

A Guide to the Theory of NP-Completeness

Michael R. Garey / David S. Johnson


## What do we do about it?

- Approximation Algorithm:
- Can we get an efficient algorithm that guarantees something close to optimal? (e.g. Answer is guaranteed to be within $1.5 x$ of Optimal, but solved in polynomial time).
- Restrictions:
- Many hard problems are easy for restricted inputs (e.g. graph is always a tree, degree of vertices is always 3 or less).
- Heuristics:
- Can we get something that seems to work well (good approximation/fast enough) most of the time? (e.g. In practice, n is small-ish)


[^0]:    Somewhat old, from Rosen

