

CS 740 – Computational Complexity and Algorithm Analysis

Spring Quarter 2010

Slides 1

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1. **A motivating example**
2. What is *Computational Complexity* all about?
3. More examples
4. A computational complexity success story
5. Organizational matters

A motivating example

Input: A connected graph (undirected)

Output: “yes” if the graph has a path which

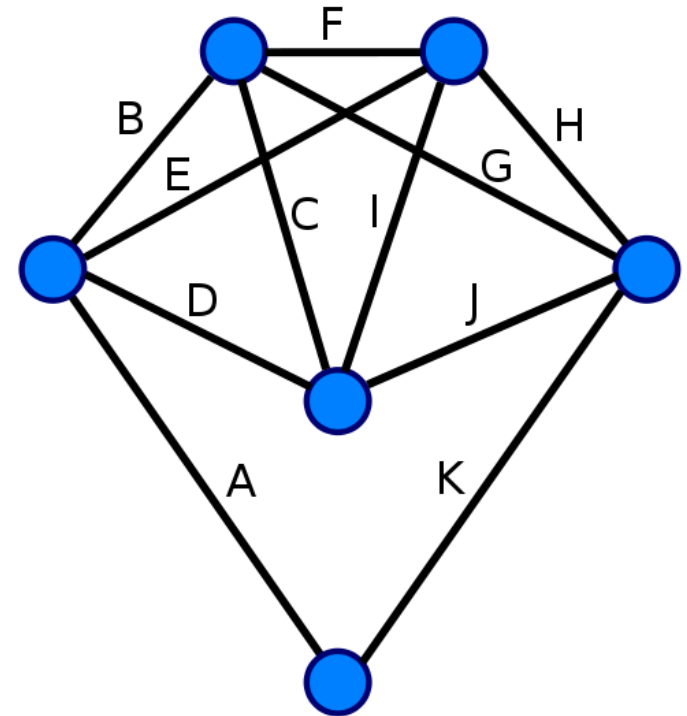
- visits each edge exactly once and
- starts and ends on the same vertex.

Output: “no” otherwise

Find an algorithm for this problem.

[Such graphs are called *Eulerian*.]

[Such paths are called *Eulerian cycles*.]



- Graph consists of
 - m vertices: $1, 2, \dots, m$
 - n edges, written as (x, y) [edge between vertex x and vertex y]

for each permutation P of the n edges

[i.e., $P = ((x_1, y_1), \dots, (x_n, y_n))$]

output "yes" if P constitutes an Eulerian cycle

output "no" if no Eulerian cycle was found

Is this a good algorithm?

How to improve?

- Graph consists of
 - m vertices: $1, 2, \dots, m$
 - n edges, written as (x, y) [edge between vertex x and vertex y]

for each permutation P of the n edges

[i.e., $P = ((x_1, y_1), \dots, (x_n, y_n))$]

output "yes" if P constitutes an Eulerian cycle

output "no" if no Eulerian cycle was found

How costly is this? (roughly, order of magnitude)

If no Eulerian cycle is found, we have to check $n!$ permutations
(that's the *worst case*).

n! is quite a lot!

n	n!
5	120
10	3,628,800
15	$\approx 1.3 \cdot 10^{12}$
20	$\approx 2.4 \cdot 10^{18}$
50	$\approx 3 \cdot 10^{64}$
70	$\approx 10^{100}$

10^{100} – That's more than there are particles in the universe

n! is quite a lot!

- 10^1
- 10^2
- 10^3 **Number of students in the college of engineering**
- 10^4 **Number of students enrolled at WSU**
- 10^6 **Number of people in Dayton**
- 10^7 **Number of people in Ohio**
Number of seconds in a year
- 10^8 **Number of people in Germany**
- 10^{10} **Number of stars in the galaxy**
Number of people on earth
Number of milliseconds per year
- 10^{20} **Number of stars in the universe**
- 10^{80} **Number of particles in the universe**

- Graph consists of
 - m vertices: $1, 2, \dots, m$
 - n edges, written as (x, y) [edge between vertex x and vertex y]

Fix the first vertex, say, x_1 .

Make a systematic depth-first search on the graph edges.

For each resulting maximal path P , if P is an Eulerian cycle, output "yes".

If no Eulerian cycle is found, output "no".

Algorithm is better – but is it *significantly* better?

In the worst case, fully connected graph, we have to check $m!$ paths.

When do we know we have *the best* algorithm?

Theorem:

A connected undirected graph is Eulerian if and only if every vertex has an even number of edges (counting loops twice).

For each vertex x

if number of edges of x is odd, output "No" and stop.

Output "Yes".

In the worst case, we have to make m checks, each of which consists of counting at most n edges.

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What is Comp. Complexity all about?

- **Some problems seem hard but are not. Identify them.**
- **Some problems seem easy but are not. Identify them.**

- **Know when to stop searching for a smarter algorithm.
[And instead turn to optimizations and heuristics.]**

- **What does “computationally hard problem” mean exactly?**
- **In what sense can we really say that some problem is computationally harder than some other problem?**

What is Comp. Complexity all about?

- It's a part of *theoretical* computer science.
- It's a formal theory of the analysis of computational hardness of problems.
- It's probably rarely going to help you directly in practice.
- But indirectly, in form of having a systematic understanding of problem hardness, it is indispensable.

We will certainly also learn about the

P = NP?

problem.

What it is.

Why it is important.

Why some people make such a fuzz about it.

- **Problem:**
A mapping from input to output.

- **Algorithm:**
A method or a process followed to solve a problem.

- A problem can have many algorithms.

- **Problem:**
A mapping from input to output.

- **Algorithm:**
A method or a process followed to solve a problem.

- **We focus on problems. Algorithm analysis is also interesting, but not as foundationally important.**

- **Problem:**
A mapping from input to output.
 - We use the *order of magnitude* of the number of steps needed to solve a problem.
 - measured as a number which depends on the *input size*.
 - We are really interested in the *worst case* scenario.
 - i.e., how many steps do we need if the input is as unfavorable as possible?

Can also be studied:

best case (usually not that interesting)

average case (of practical interest for concrete algorithms)

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- **Input:** An array A of integers, and a value v .
- **Output:** “Yes” if v is an element of A ;
“No” otherwise.

$A =$

3	12	7	25	7	32	11	56	28	43	6	87	68	91	2
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$v = 28$

Algorithms: Exhaustive search; random search; sort and linear search; sort and binary search

Not all inputs of a given size take the same time to run.

Sequential search for v in an array of n integers:

- **Begin at first element in array and look at each element in turn until v is found**

Best case:

Worst case:

Average case?

Linear Search – Average case

Case: i	Time: $T(i)$	Probability $P(i)$	Cost: $T(i) * P(i)$
1	1	$1/n$	$1/n$
2	2	$1/n$	$2/n$
3	3	$1/n$	$3/n$
...
n	n	$1/n$	1

$$\begin{aligned}\sum Cost &= \frac{1}{n} + \frac{2}{n} + \dots + \frac{n}{n} \\ &= \frac{1}{n} \times \sum_{i=1}^n i \\ &= \frac{1}{n} \times \frac{n(n+1)}{2} = \frac{n+1}{2}\end{aligned}$$

- **Algorithm 1:**

```
for pass = 0...n-1 {  
  for position = 0...n-1 {  
    if (array[position] > array[position+1]) {  
      swap (array[position], array[position+1])  
    }  
  }  
}
```

Best, worst, average: $\approx (n^2)$

- **Algorithm 2:**

```
for pass = 0...n-1 {  
  for position = 0...n-pass-1 {  
    if (array[position] > array[position+1]) {  
      swap (array[position], array[position+1])  
    }  
  }  
}
```

Best, worst, average: $\approx (n^2)$
(Within a constant factor)

- **Algorithm 3:**

```
for pass = 0...n-1 {  
    swaps = 0;  
    for position = 0...n-pass-1 {  
        if (array[position] > array[position+1]) {  
            swap (array[position], array[position+1]);  
            swaps++;  
        }  
    }  
    if (swaps == 0) return;  
}
```

Best: $\approx n$,
Worst: $\approx (n^2)$
Average = ??

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- **Research Area: Semantic Web**
Aimed at endowing information on the World Wide Web with “machine-processable meaning” (semantics).
- This is done using languages for representing knowledge.
E.g., the knowledge on a website.
- These languages can also be used for querying this knowledge.
- These languages are also able to represent problems.
Knowledge: A graph.
Query: Does it have an Eulerian cycle?
- These languages differ in how “complex” the problems representable in them can be.

- **Web Ontology Language OWL**
Recommended standard by the World Wide Web Consortium W3C.
Established 2004, revised 2009.
- **Research which led to OWL was driven by computational complexity analysis.**
- **Complexity used as a priori measure for runtime.**
- **Goal was finding a language which allows maximum freedom in specifying knowledge (problems), while being of minimal complexity.**
- **This approach paid off extremely well:
Currently e.g. substantial commercial interest generated.**

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- **Office Hours: Wed 3-4, Joshi 389.**
Email contact preferred.
- **Textbook (required):**
Thomas A. Sudkamp, Languages and Machines, Third Edition,
Addison Wesley, 2006.
- **Textbook (supplementary):**
Michael R. Garey and David S. Johnson, Computers and
Intractability, Freeman, 1979
- **Grading:**
Midterm exam: 30%
Final exam: 50%
Exercises: 20%

- I will be absent May 3rd and 5th.
 - How do we make up for this?
 - Have 7.5 sessions 20 minutes longer (agreed).
 - Find alternative dates for meeting.
 - Give you extra hand-in homework.
- We will frequently make exercise sessions.
You will get exercises, to be done at home and graded by me, and discussed afterwards in class.
Each exercise usually counts 4 points [exceptions are marked].
Exercises are due *one week after I pose them*.
- Webpage/slides.
I prefer to use a public website:
<http://knoesis.wright.edu/faculty/pascal/teaching/s10/complexity.html>

Tentative

We recap most of chapter 8

We cover most of chapters 14 and 15

We cover parts of chapters 16 and 17, tbd what/how much.

- **March 29/31: Introduction. Big-O notation.**
- **April 5/7: Recap Turing Machines. Exercise session.**
- **April 12/14: Time complexity.**
- **April 19/21:**
- **April 26/28: mid-term exam**
- **May 3/5: no classes**
- **May 10/12:**
- **May 17/19:**
- **May 24/26:**
- **June 2:**