

Computational Complexity

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Outline

1. **A motivating example**
2. What is *Computational Complexity* all about?
3. More examples

A motivating example

Input: A connected graph (undirected)

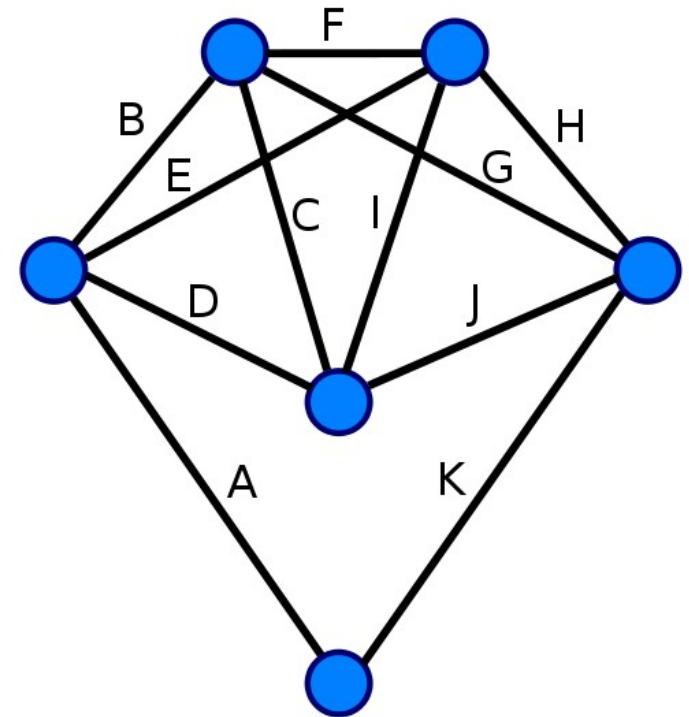
Output: “yes” if the graph has a path which
 ~ visits each edge exactly once and
 ~ starts and ends on the same vertex.

Output: “no” otherwise

Find an algorithm for this problem.

[Such graphs are called *Eulerian*.]

[Such paths are called *Eulerian cycles*.]



Naive algorithm 1 – brute force

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?

Naive algorithm 1 – brute force

Graph consists of

- m vertices: $1, 2, \dots, m$

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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	$\frac{1}{4} 1.3 \times 10^{12}$
20	$\frac{1}{4} 2.4 \times 10^{18}$
50	$\frac{1}{4} 3 \times 10^{64}$
70	$\frac{1}{4} 10^{100}$

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

n! is quite a lot!

┐┌ **10¹**

┐┌ **10²**

┐┌ **10³** **Number of students in the college of engineering**

┐┌ **10⁴** **Number of students enrolled at WSU**

┐┌ **10⁶** **Number of people in Dayton Metro Area**

┐┌ **10⁷** **Number of people in Ohio**

Number of seconds in a year

┐┌ **10⁸** **Number of people in the Midwest**

┐┌ **10¹⁰** **Number of stars in the galaxy**

Number of people on earth

Number of milliseconds per year

┐┌ **10²⁰** **Number of stars in the universe**

┐┌ **10⁸⁰** **Number of particles in the universe**

Naive algorithm 2 - backtracking

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?

Smart 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.

What is Comp. Complexity all about?

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.

Some basic notions

⌋⌋ **Problem:**

A mapping from input to output.

⌋⌋ **Algorithm:**

A method or a process followed to solve a problem.

⌋⌋ **A problem can have many algorithms.**

Some basic notions

⌋⌋ **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.**

Some basic notions

⌋⌋ 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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Linear Search

- 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
---	----	---	----	---	----	----	----	----	----	---	----	----	----	---

$v = 28$

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

Linear 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

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

Bubble Sort

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: $\Theta(n^2)$

Bubble Sort

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: $\Theta(n^2)$
(Within a constant factor)

Bubble Sort

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