

Math 251, Fri 8-Oct-2021 -- Fri 8-Oct-2021
Discrete Mathematics
Fall 2021

Friday, October 8th 2021

Wk 6, Fr
Topic:: Algorithmic complexity
Read:: Rosen 3.3

for - - -

break

Algorithmic Complexity

Basic idea: relate size n of input to, for instance

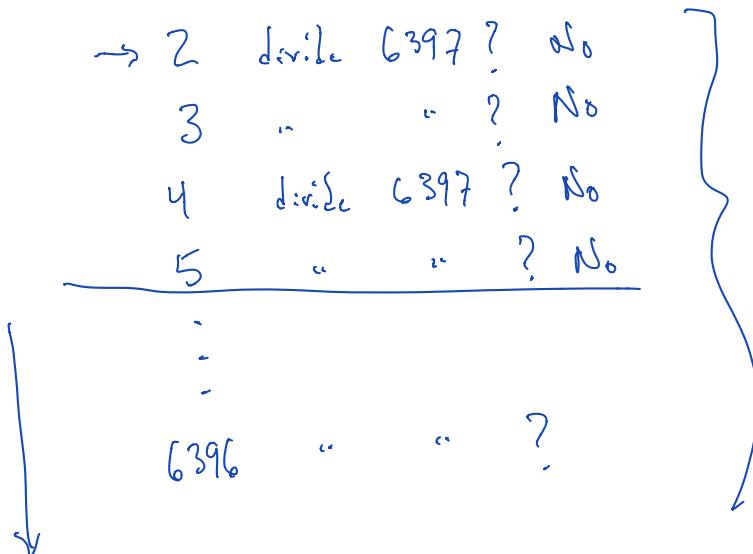
- time complexity (often assessed by number of comparisons, flops, etc.)
 - worst-case analysis
 - average-case analysis
- space complexity
- terms like
 - linear complexity $\Theta(n)$
 - quadratic complexity $\Theta(n^2)$
 - polynomial complexity $\Theta(n^r)$ some $r \in \mathbb{Z}^+$
 - * linear, quadratic are special cases
 - * call these problems "tractable", and are of **Class P**
 - exponential complexity
 - * call these problems "intractable"
 - may still be of **Class P** if a polynomial-time algorithm exists
 - say it is of **Class NP** if no alg. of polynomial time is known for solving it, but there is a polynomial time alg. for checking a solution
 - * **NP-complete** problems, and the **P vs. NP problem**

Algorithm:

1. Seek divisor of $n \in \mathbb{Z}^+$
Similar to analysis of linear search algorithm }
2. binary search - sorted list
3. bubble sort
- 4. matrix multiplication
5. evaluating an n^{th} -degree polynomial

1. Given positive integer n . Task: Find a divisor

Ex. $\boxed{6397 = n}$



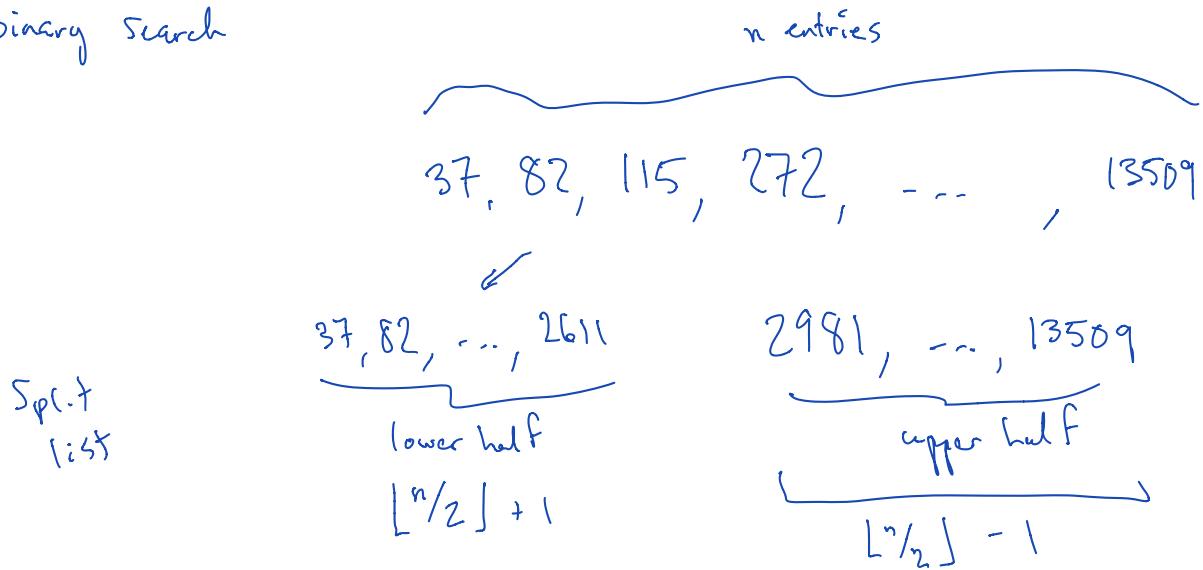
If no short-circuit,
then # of ~~attempted~~ divisions

$$6395 = n - 2$$

$O(n)$

Could stop once reached \sqrt{n} : Applying this as a stopping criterion makes it $O(\sqrt{n})$.

2. Binary Search



Q: How many times splitting list in half (size n) before I have singleton lists

A: $\lceil \log_2 n \rceil + 1$ is $O(\log_2 n)$

3. Bubble sort

of comparisons: $1 + 2 + 3 + \dots + (n-1) = \frac{1}{2}n(n-1)$
 for list size n

$$= \frac{1}{2}n^2 - \frac{1}{2}n, \quad \Theta(n^2).$$

4. Matrix multiplication

$$\begin{bmatrix} 3 & -1 \\ 2 & 7 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 1 & 5 \\ -1 & -3 \end{bmatrix}_{2 \times 2} = \begin{bmatrix} (3)(1) + (-1)(-1) \\ (2)(1) + (7)(-1) \end{bmatrix}_{2 \times 2}$$

Same process applied to $n \times n$ matrices

$$\begin{array}{c} \text{row } i \rightarrow \begin{bmatrix} a_{i1} & a_{i2} & a_{i3} & \dots & a_{in} \end{bmatrix}_{n \times 1} \\ \text{column } j \leftarrow \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix}_{n \times 1} \end{array} = \begin{bmatrix} c_{ij} \end{bmatrix}_{1 \times 1} \quad \begin{array}{l} \text{results} \\ \{ \text{i-th row,} \\ \text{j-th column} \} \end{array}$$

$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj}$

← n multiplications
 $(n-1)$ additions
 to fill one slot in product

$n + (n-1)$, or $2n-1$ flops

$$\underbrace{n^2(2n-1)}_{\text{flops}} \text{ fills one slot} = 2n^3 - n^2 \text{ flops} \quad \Theta(n^3).$$

5. Evaluating n^{th} -degree polynomial

$$p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

at $x=c$.

$$\text{Compute } p(c) = \underbrace{a_n \cdot c \cdot c \cdot c \cdots c}_{\substack{n \text{ factors} \\ \rightarrow n \text{ mults.}}} + \underbrace{a_{n-1} \cdot c \cdot c \cdots c}_{(n-1) \text{ mults.}} + \dots$$

$$+ \underbrace{a_1 c}_{1 \text{ mult.}} + a_0$$

$$\begin{aligned} \text{flops} &= (\# \text{ of mults.}) + (\# \text{ of addition}) \\ &= \underbrace{(1+2+3+\dots+n)}_{\text{ }} + n \\ &= \frac{n}{2}(n+1) + n = \frac{1}{2}n^2 + \frac{3}{2}n \quad \left(\mathcal{O}(n^2) \right) \end{aligned}$$

Better algorithm, labeled Horner's Algorithm, for evaluating polynomials

Comes up in the textbook, Exercise 14 of Section 3.3.