

MA2252 Introduction to computing

lectures 13-14

More on complexity and

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Complexity: visually inspecting

The following is a useful code for inspecting the average complexity of an algorithm

```
N = 20; K = 30;
T = zeros(N,K);
for k=1:K
    for n = 1:N
        arr = rand(1,n)
        tic,
        myFunction(arr);
        T(n,k) = toc;
    end
end
T = 1/K*sum(T,2);
plot(T(1:end), ' * ')
```

It gives an average of K samples of the time it takes to run the function for each input of size 1 to N.

Complexity: complexity of the binary search algorithm

Best case: When the element is the middle pointer, so the element is found in one iteration $\Rightarrow O(1)$

Average case:

- ▶ **Case 1:** the key is present in the array $\rightarrow n$ possible cases
- ▶ **Case 2:** the key is not in the array $\rightarrow 1$ possible case
- ▶ A key at index $n/2$ is found in 1 comparison
- ▶ A key at index $n/4$ and $3n/4$ is found in 2 comparisons
- ▶ A key at index $n/8$, $3n/8$, $5n/8$, and $7n/8$ is found in 3 comparisons, etc

$$\begin{aligned}\Rightarrow \frac{\text{Tot. comparisons}}{\text{numb. cases}} &= \frac{1 \times 1 + 2 \times 2 + 3 \times 4 + \dots + \log n \times 2^{\log n - 1}}{n + 1} \\ &= \frac{n(\log n - 1) + 1}{n + 1} \\ &= O(\log n)\end{aligned}$$

Worst case: The key is not in the array (or it is in the first element). Following the same analysis as before we need to do $O(\log n)$ comparisons.

Complexity: complexity of the quick-sort algorithm

Best case: We select the pivot as the mean

$$\begin{aligned}T(n) &= 2 \times T\left(\frac{n}{2}\right) + n \times \text{constant} \\&= 2 \times \left(2 \times T\left(\frac{n}{4}\right) + \frac{n}{2} \times \text{constant}\right) + n \times \text{constant} \\&= 4 \times T\left(\frac{n}{4}\right) + 2 \times \text{constant} \times n. \\&= 2^k \times T\left(\frac{n}{2^k}\right) + k \times \text{constant} \times n \\&= n \times T(1) + n \times \log n.\end{aligned}$$

So in the best case the complexity is $O(n \log n)$

Complexity: complexity of the quick-sort algorithm

Worst case: The array gets divided into one part consisting of $n - 1$ elements and that one into $n - 2$ elements, so on and so forth.

$$\begin{aligned}T(n) &= T(n - 1) + n \times \text{constant} \\&= T(n - 2) + (n - 1) \times \text{constant} + n \times \text{constant} \\&= T(n - k) + k \times n \times \text{constant} - \text{constant} \times (k \times (k - 1))/2\end{aligned}$$

For $k = n$ we have

$$T(n) = T(0) + n \times n \times \text{constant} - \text{constant} \times (n \times (n - 1))/2 = O(n^2).$$

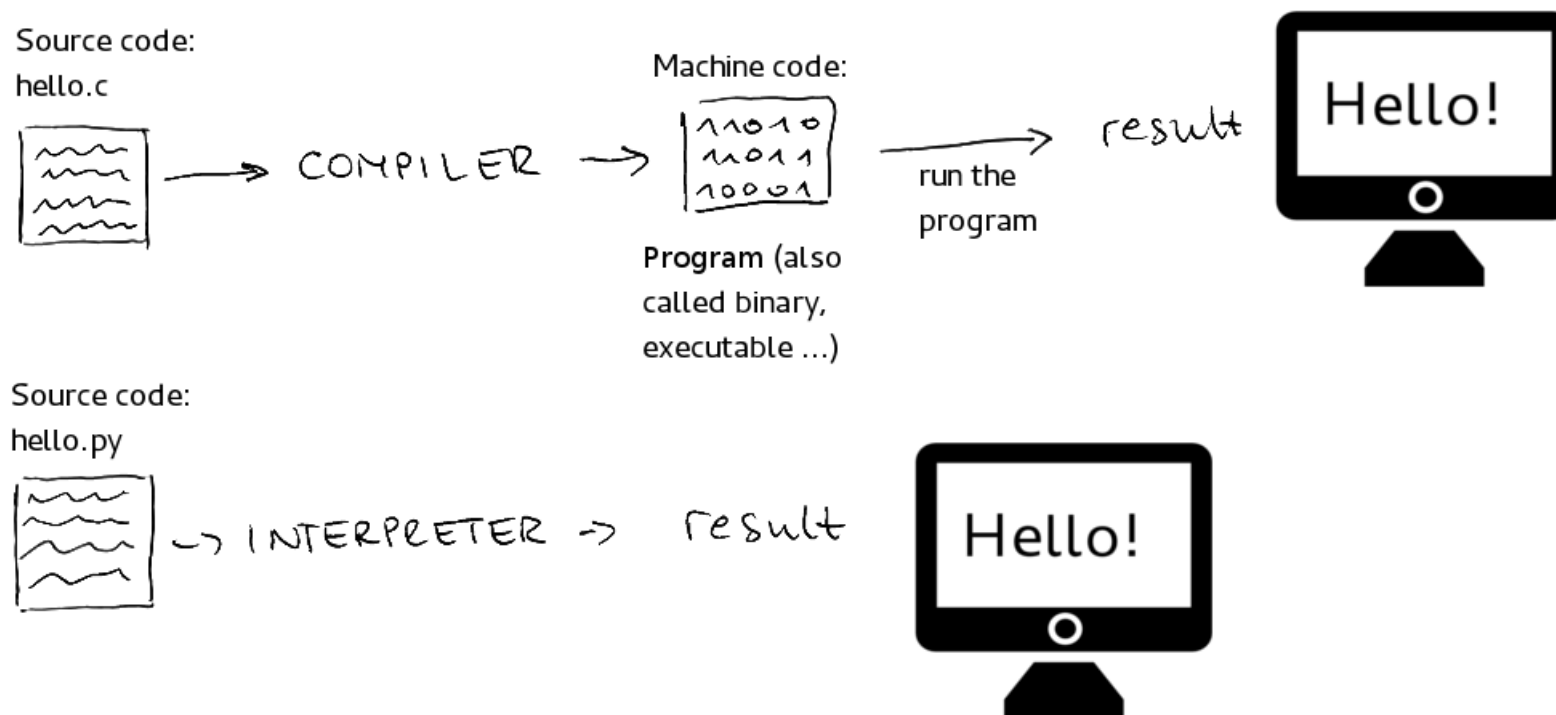
Average case: It can be shown that in the average case the complexity is $O(n \log n)$.

Complexity: a good online reference

`https://www.geeksforgeeks.org/
analysis-algorithms-big-o-analysis/`

Efficient coding in MATLAB: interpreted v/s compiled languages

There are (roughly) two types of programming languages:
interpreted and **compiled**



MATLAB is an interpreted programming language, but it has built-in pre-compiled functions \Rightarrow try to use them as much as you can.

Representation of Numbers

Base-N numbers

Base: Number of unique digits to represent a number.

Example:

Decimal number (Base-10 number)

- ① Uses digits from 0 to 9. So, base is 10.
- ② Decimal number can be expanded in powers of 10.

Examples:

$$582 = 5 \times 10^2 + 8 \times 10 + 2 \times 10^0,$$

$$25.36 = 2 \times 10^1 + 5 \times 10^0 + 3 \times 10^{-1} + 6 \times 10^{-2}.$$

Other examples:

- **Binary number:** Uses only two digits: 0 and 1.
- **Octal number:** Uses digits from 0 to 7.
- **Hexadecimal number:** Uses 16 digits: 0-9 followed by A-F or a-f.

Binary numbers

- Binary numbers are represented by only 0s and 1s. So, the base is 2.
- A digit in binary number is called a **bit**.
- **Example:** The number 1001000110.

This number can be converted to a decimal number as follows:

$$1 \times 2^9 + 0 \times 2^8 + 0 \times 2^7 + 1 \times 2^6 + 0 \times 2^5 + 0 \times 2^4 + 0 \times 2^3 + 1 \times 2^2 + 1 \times 2^1 + 0 \times 2^0 = 582$$

Binary numbers (contd.)

Pros

- Easier for a computer to store and work with.
- A computer can do all arithmetic operations with them.
- A 32-bit computer can represent upto 2^{23} binary numbers.

Cons

- Harder for humans to do binary algebra.
- Limited **range** and **precision** to perform all mathematical calculations.

Floating Point Numbers

MATLAB uses **floating point** numbers or **floats** to achieve the desired range and precision required to perform mathematical calculations.

Note: Floating point numbers are always **rational**. MATLAB approximates irrational numbers (e.g. π) by rational numbers.

Types of float:

- single precision (32 bits)
- double precision (64 bits)

Single precision float

In the **IEEE754** standard for single precision, a float which is represented as

$$n = -1^s 2^{e-127} (1 + f)$$

Here, s , e and f are sign indicator, exponent and fraction respectively.

32 bits are allocated as follows:

- s has 1 bit so it takes values of 0 or 1.
- e has 8 bits so it can take 2^8 values.
- f has 23 bits so it can take 2^{23} values. $0 \leq f < 1$.

Floating Point Numbers (contd.)

Example:

Convert the number

1 10000011 1100000000000000000000 (IEEE754)

into decimal (base 10) number.

Floating Point Numbers (contd.)

Solution:

$$\begin{aligned}s &= 1, \\ e &= 1 \times 2^7 + 0 + 0 + 0 + 0 + 0 + 1 \times 2^1 + 1 \times 2^0 = 131, \\ f &= 1 \times 2^{-1} + 1 \times 2^{-2} + \dots = 0.75.\end{aligned}$$

$$n = -1^1 2^{131-127} (1 + 0.75) = -2^4 \times 1.75 = -28.$$

Gaps between numbers

Not every decimal (base 10) number can be represented by double precision float. This causes gaps between numbers in MATLAB.

To find the gap, use MATLAB's `eps` function. `eps(x)` gives the gap between number `x` and next representable number.

Example:

`eps(5)=8.881784197001252e-16.`

The gap increases as numbers get large because the factor 2^{e-127} grows in size.

Special cases

① $e=0$ (00000000(base 2))

In this case, the float is calculated using

$$n = -1^s 2^{-126} f$$

② $e=255$ (11111111(base 2))

- $f \neq 0$: $n=\text{NaN}$ (Not a Number)
- $f=0$ and $s=0$: $n=\text{Inf}$
- $f=0$ and $s=1$: $n=-\text{Inf}$

Floating Point Numbers (contd.)

Single precision float interesting facts

- Largest defined number

0 11111110 111111111111111111111111
(3.402823466385289e+38)

- Smallest defined positive 'normal' number

0 00000001 000000000000000000000000
(1.175494350822288e-38)
(2^{-126})

Note: Use MATLAB functions `realmax('single')` and `realmin('single')` to find the above results.

Floating Point Numbers (contd.)

- Smallest defined subnormal number
0 00000000 000000000000000000000001
(1.401298464324817e-45)
(2^{-149})

Note: Use MATLAB `single()` function to represent a number in single precision.

Double precision float

MATLAB uses default double precision of **IEEE754** standard. A double precision is represented as

$$n = -1^s 2^{e-1023} (1 + f)$$

Again, s , e and f are sign indicator, exponent and fraction respectively.

64 bits are allocated as follows:

- s has 1 bit so it takes values of 0 or 1.
- e has 11 bits so it can take 2^{11} values.
- f has 52 bits so it can take 2^{52} values. $0 \leq f < 1$.

Floating Point Numbers (contd.)

The special cases for double precision float are similar to single-precision float.

Double precision float interesting facts

- Largest defined number
1.797693134862316e+308 (type `realmax` in command window)
- Smallest defined positive 'normal' number
2.225073858507201e-308 (type `realmin` in command window)
- Smallest defined subnormal number
4.940656458412465e-324
(2^{-1074})

Loss of precision

- ▶ Precision cannot be gained but can be lost.
- ▶ If you add the two quantities 0.171717 and 0.51, then the result should only have two significant digits.
- ▶ **Subtractive cancellation:** A loss of significance can be incurred if two nearly equal quantities are subtracted from one another!

Example:

$$0.177241 - 0.177589 = 0.348 \times 10^{-3}$$

So three digits of accuracy have been lost.

- ▶ Can often be avoided by rewriting the expression.

Loss of precision

Example 1: Consider the stability of $\sqrt{x+1} - 1$ when x is near zero. Say, $x = 1.2345678 \times 10^{-5}$, then $\sqrt{x+1} \approx 1.000006173$. In a 8 significant digits computer we'd have $\sqrt{x+1} - 1 = 6.2 \times 10^{-6}$, so 6 digits of accuracy have been lost.

To fix this:

$$\sqrt{x+1} - 1 = \sqrt{x+1} - 1 \frac{\sqrt{x+1} + 1}{\sqrt{x+1} + 1} = \frac{x}{\sqrt{x+1} + 1}$$

No subtractions so no subtractive cancellation!

$$\frac{1.2345678 \times 10^{-5}}{2.0000062} \approx 6.17281995 \times 10^{-6}$$

Loss of precision

Example 2: Rewrite the roots of the quadratic equation $x^2 + bx + c = 0$, for when $b \gg c > 0$.

Loss of precision

Example 2: Rewrite $e^x - \cos x$ to be stable when x is near zero.