# Big $\mathcal O$ notation

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# A little problem: What kind of algorithm is that?

```
/* uli is an abbreviation for the type:
   unsigned long int */
function do_something_weird(uli $x$, uli y): uli
    uli result = 0:
    while x = 0 do:
        if \times \& 1 = 1 then:
             result = result + y;
        end
        x = x >> 1:
        v = v << 1:
    end
    return result:
end
```

- Big  $\mathcal{O}$  time/space is the lenguage and metric we use to describe the efficiency of algorithms.
- Imagine the following scenario:
  - You got a file on a hard drive and you need to send it to a friend who lives across the country. You need to get the file to your friend as fast as possible. How should you send it?
  - E-mail
  - FTP, HTTP, SCP, . . .
  - Dropbox, Google Drive, Sky Drive, ...
  - Airplane
  - What if the file was really large?

- The ammount of elemental operations or stored information to solve a problem we know as "Computational Complexity"
- The are two kinds of complexity:
- Runtime (temporal complexity)
- Memory (spatial complexity)
- ullet Big  ${\mathcal O}$  notation allow us classify our algorithms in categories based-on input size.

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# Notation $\mathcal{O}(\cdot)$ : categories

- Those categories are:
  - $\mathcal{O}(1)$ ,  $\mathcal{O}(n)$ ,  $\mathcal{O}(n^2)$ ,  $\mathcal{O}(n^3)$ ,  $\mathcal{O}(n^4)$ , ...  $\mathcal{O}(\log n)$ ,  $\mathcal{O}(n \log n)$ ,  $\mathcal{O}(n^2 \log n)$ , ...

  - $\circ$   $\mathcal{O}(n!)$ ,  $\mathcal{O}(2^n)$ , ...
  - What the hell does this mean? O.o.o.O.0-0

### Definitions and notation

- For example, to send a file to a friend the run time is:
  - By e-mail, FTP, Dropbox is  $\mathcal{O}(s)$  where s is the file size.
  - By airplane is  $\mathcal{O}(1)$ .
  - But, maybe is  $O(2^n)$  in money, where  $n \dots$ ? WTF! Why?
- Rule: Drop the constants
  - What?
  - Why?
  - Example: seems that to send by e-mail is indeed at least 2s in time where s is the file size.

# Best, Worst and Average/Expected cases

- Think about search an element in an unsorted simply linked list:
  - Best case: the element is at the start of the list. (Maybe you think this is  $\mathcal{O}(1)$ ).
  - Worst case: the element is at the end of the list. (Maybe you think this is  $\mathcal{O}(n)$ ).
  - Average/Expected case: the element is in any position at the list. (What do you think about this?)

## Examples: Print elements in arrays A and B

```
\mathcal{O}\left(\alpha+b\right): where \alpha and b are the size of array A and B respectively. 

for each x in A:
    print(x);
end
for each y in B:
    print(y);
end
```

## Examples: Print elements in arrays A and B

```
\mathcal{O}\left(\alpha b\right): where \alpha and b are the size of array A and B respectively. 
 for each \times in A:
   for each y in B:
    print(\times, y);
   end
end
```

# Examples: Print elements in arrays A

```
\mathcal{O}(n^2): where n is the size of array A.
for i = 0 to n - 1 do:
     for j = i to n - 1 do:
         print(A[j]);
    end
end
A[0], A[1], A[2], ..., A[n-1]
A[1], A[2], \dots, A[n-1]
A[2], ..., A[n-1]
A[n-1]
```

# What does following function? What is the run time?

```
\begin{array}{ll} \textbf{function} & \textbf{something} \, (\, \textbf{n} \colon \, \textbf{integer} \,) \colon \, \textbf{integer} \\ & \times = \, 0 \, ; \\ & \textbf{while} \, \times \, * \, \times \, < \, \textbf{n} \, \, \textbf{do} \, ; \\ & \times \, = \, \times \, + \, 1 \, ; \\ & \textbf{end} \\ & \textbf{return} \, \, \times \, ; \\ & \textbf{end} \end{array}
```

### References

- Gayle Laakmann Cracking the Coding Interview
- Robert Sedgewick Algorithms C++
- Thomas H. Cormen Introduction to Algorithms
- Donald E. Knuth The Art of Computer Programming
- Wikipedia
- Quora