



"the toolbox"

- Algorithm Analysis:
DEF: To investigate algorithm efficiency with respect to time and space.

- Analytical not Empirical

Suppose:

A

B

which one is better?

Analytical:

we will model performance using a mathematical function the "runtime" function

$R()$: input :- the size of the input data.
output :- the number of "operations" performed by the program

operations: $\leftarrow + * =$
 $[\] < >$

$R_A(n)$ vs $R_B(n)$

to compare performance we compare functions

E.G.

```

FUNCTION Sum(A[0..n-1])
  s ← 0  -1
  i ← 0  -1
  while i < n do
    { s ← s + A[i]  -3
      i ← i + 1    -2 } n(3+2+1) = 6n
  return s - 1
    
```

$$R_{\text{sum}}(n-1) = 6n + 3$$

BUT: there are many inputs of size n

- BUT: there are many inputs of size n
which one to choose?

E.G

FUNCTION Find ($A[0..n-1], x$)

```

FOR i ← 0 TO n-1 DO
  IF  $x = A[i]$  THEN
    RETURN i
RETURN -1

```

Find($\boxed{1 \mid 2 \mid 3}, 2$)

Find($\boxed{1 \mid 2 \mid 3}, 4$)

$$R_{\text{worst Find}}(n) = 4 \cdot n + 1$$

★ - Worst input :- the one that takes most steps.

- best input :- optimistic underestimate.

- all inputs, and then average.

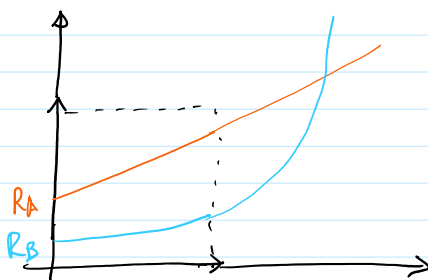
- amortize cost over all inputs

- BUT: which size n to use?

- no specific values !!!

we want to know what happens to the runtime function as n becomes really large.

- Graphs are decietfull



- we are going to care about a function's
"rate-of-growth" or "order-of-growth"
what happens $n \rightarrow \infty$

- When you work with "rate-of-growth"
you ignore constant factors.

$$R_{\text{sum}}(n-1) = \cancel{6n + 3}$$

$$R_{\text{worst Find}}(n) = \cancel{4 \cdot n + 1}$$

• ASYMPTOTIC NOTATION

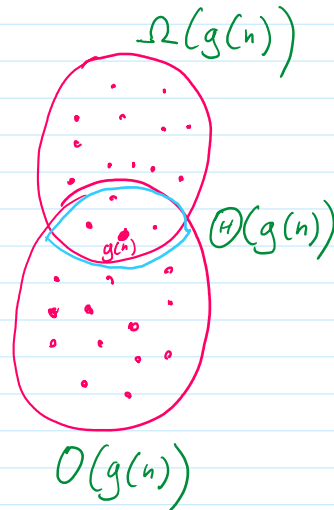
(informal definition)

Let $g(n)$ be a function:

$O(g(n))$ is the set of all functions with lower or same "rate-of-growth"

$\Omega(g(n))$ is the set of all functions with larger or same "rate-of-growth"

$\Theta(g(n))$ is the set of functions with the same "rate-of-growth"



E.g. functions in $O(n^2)$

- $n \in O(n^2)$
- $27n+3 \in O(n^2)$
- $3n^2 \in O(n^2)$
- $\cancel{327}n^2 + \underline{1024}n + \underline{10,972} \in O(n^2)$
- note: ignore constant factors.

more.

- $n^3 \in \Omega(n^2)$
- $3n^2 \in \Omega(n^2)$
- $2^n \in \Omega(n^2)$
- $n! \in \Omega(n^2)$
- $3n^2 \in \Theta(n^2)$

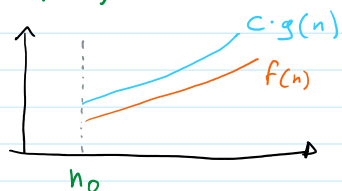
DEF:

We say that $f(n) \in O(g(n))$ if there exist constants C and n_0 such that for all $n > n_0$

$$f(n) \leq C \cdot g(n)$$

- no restriction on C and n_0
- C is what allows us to ignore constant factors.

in English: $f(n)$ is bound from above by a multiple of $g(n)$



E.G.

- $n^2 \in O(3n^2 + n)$

need to show that $n^2 \leq C \cdot (3n^2 + n)$ for $n > n_0$

$C = 1$

$n_0 = 1$

$n^2 \leq 3n^2 + n$

$n^2 \leq 3n^2$

for $n > 1$

$$C = 1$$

$$n_0 = 1$$

$$n^2 \leq 3n^2 + n$$

$$n^2 \leq 3n^2 \quad \text{②}$$

for $n > 1$

E.G.

$$3n^2 + n \in O(n^2)$$

need to show that

$$3n^2 + n \leq C \cdot (n^2) \text{ for } n > n_0$$

$$C = 4$$

$$n_0 = 1$$

$$3n^2 + n \leq 4n^2$$

$$\cancel{3n^2} + n \leq \cancel{3n^2} + 1n^2$$

$$n \leq n^2$$

②

DEF:

we say that $f(n) \in \Omega(g(n))$
if there exists constants C and n_0
such that for every $n > n_0$

$$f(n) \geq C \cdot (g(n))$$

In English: $f(n)$ is bounded from below by a multiple of $g(n)$

DEF:

we say that $f(n) \in \Theta(g(n))$
if there exists constants C_1, C_2 and n_0
such that for every $n > n_0$

$$C_1 \cdot (g(n)) \leq f(n) \leq C_2 \cdot (g(n))$$

In English: $f(n)$ is bound from above and below
by multiples of $g(n)$



In English: $f(n)$ and $g(n)$ have the same "order-of-growth"

Note: IF $f(n) \in O(g(n))$ AND $f(n) \in \Omega(g(n))$ THEN $f(n) \in \Theta(g(n))$

BASIC EFFICIENCY CLASSES:

$$\begin{matrix} n^2 & 3n^2 + n & 4n^2 \\ 2n^2 - n & 5n^2 + 48n + 12 & n^2 \end{matrix}$$

When comparing functions:

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Bill

Kyle

↓ which one is better?

$$R_{\text{Bill}}(n) = 5n^3 + 2n^2$$

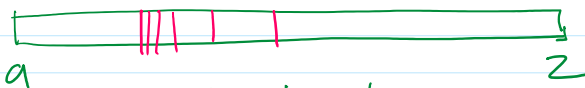
$$R_{\text{Kyle}}(n) = \frac{1}{2}n^2 + 3$$

we place functions in their class, and then compare classes.

The Complexity Hierarchy

$n!$	"factorial" :- generate all permutations of a collection of size n .
2^n	"exponential" :- generate all subsets of a collection of size n .
\vdots	
n^5	
n^4	
n^3	"cubic" :- 3 nested loops.
n^2	"quadratic" :- 2 nested for loops.
$n \cdot \log n$	"linear-logarithmic" :- "divide and conquer" algorithms.
n	"linear" :- do one thing for every element in input
$\log n$	"logarithmic" :- split input in half, and continue on one half.
1	"constant" :- programs without loops.

E.g. searching in a sorted array



n = number of elements

- How many times can you cut n elements in half?
- Same as $2^k = n$? $k = \log_2 n$

• MORE MATH : USEFUL PROPERTIES OF ORDER-OF-GROWTH

what about running Bill after Kyle.

$$R_{\text{BK}}(n) = R_{\text{Bill}}(n) + R_{\text{Kyle}}(n)$$

- if you have functions $f_1(n)$ $f_2(n)$ $g_1(n)$ $g_2(n)$

and you know that

- $f_1(n) \in O(g_1(n))$
- and
- $f_2(n) \in O(g_2(n))$

then $f_1(n) + f_2(n) \in O(\max\{g_1(n), g_2(n)\})$

e.g. $R_{Bill}(n) \in O(n^3)$ $R_{K&L} \in O(n^2)$

$$R_{BK}(n) \in O(\max\{n^3, n^2\})$$

$$R_{BK}(n) \in O(n^3)$$

- A polynomial of degree k is $O(n^k)$

- Limits and the order-of-growth

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \begin{cases} 0 & \text{:- } f(n) \text{ has a smaller order-of-growth than } g(n) \\ C & \text{:- } f(n) \text{ has the same order-of-growth as } g(n) \\ \infty & \text{:- } f(n) \text{ has a larger order-of-growth than } g(n) \end{cases}$$

these allows us to use convenient results from calculus

- L'Hopital rule:

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{f'(n)}{g'(n)}$$

- Stirling's formula:

$$n! \approx \sqrt{2 \cdot \pi \cdot n} \cdot \left(\frac{n}{e}\right)^n \quad \text{for large values of } n$$

—o— EOF