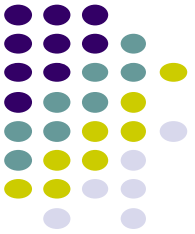




What makes an algorithm good?
Why would an algorithm be better than another?



Goodness of an algorithm?

- #1 criteria: correct
- Correct results

if we do not obtain the needed is it useless?

It depends on the problem...

It may be all we can do given the problem...

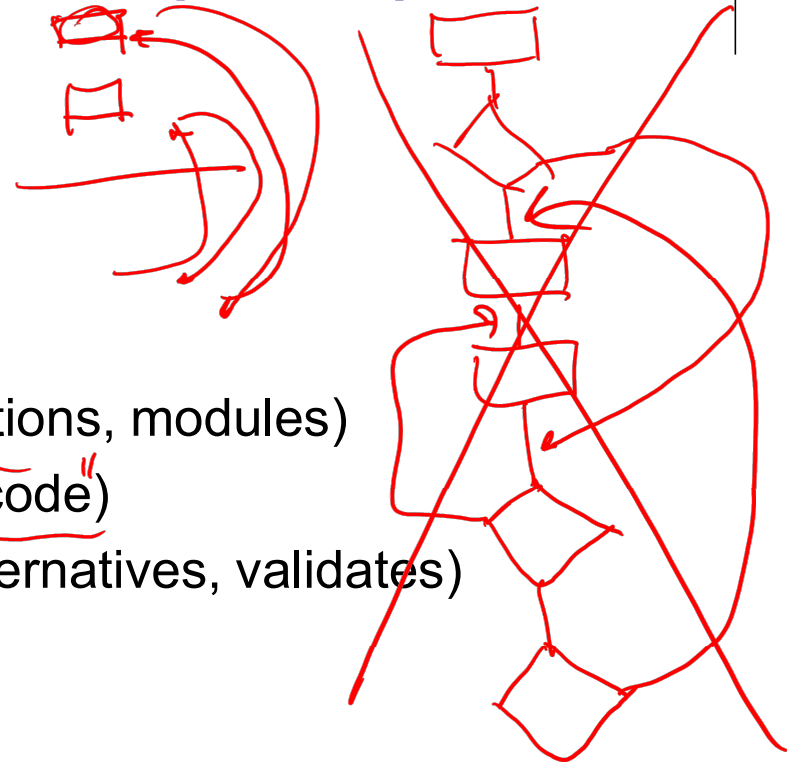
- (some cases may need to compromise w/ an approximate solution)

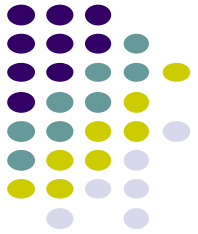


Goodness of an algorithm? (cont.)

- #2: Desirable qualities:

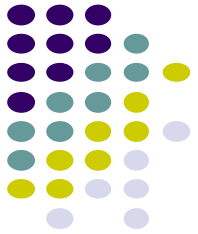
- Clear to read code and to debug
- Code easy to understand
- Concise code
- Modular organization (top level, functions, modules)
- Structured (no breaks, no "spaghetti code")
- Robust (does not crash, previews alternatives, validates)
- Easy to maintain and revise
- Clear/informative user interaction






Goodness of an algorithm? (cont.)

- Efficient time wise:
 - executes efficiently (with a realistic response time): TIME COMPLEXITY
- Efficient use of space (memory):
 - SPACE COMPLEXITY
- Efficiency: essential qualities to consider for large size problems!

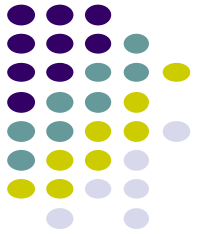


Measuring Algorithm Time Complexity

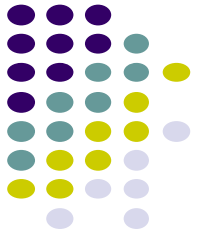
- Empirically: Seconds, milliseconds.... good idea if controlling the environment. BUT it depends on so many variables!
- Exact count of operations that get executed 

 **Order of an algorithm** → Time complexity of the algorithm

Order of an algorithm

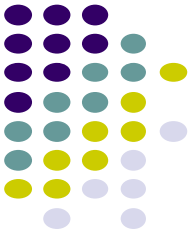


- Gives a notion of the Time complexity of the algorithm
- This is a math based theory that is most relevant for problems and algorithms involving large numbers of data (large size of problems)



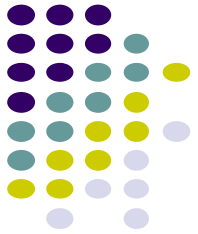
Order of an algorithm (cont.)

- The Order gives an “approximate” measure of an algorithm in terms of number of “critical operations” that are executed in the algorithm...
- “approximate” is in fact very precisely defined mathematically.



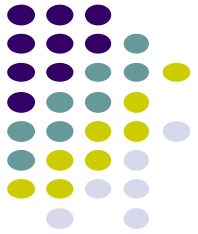
Order of an algorithm (cont.)

- Critical operations can be:
 - additions
 - comparisons (if statements)
 - transfer operations (assignments, swaps)
 - ...



Order of an algorithm (cont.)

- The **order** is expressed as the **number of critical operations** that the algorithm executes expressed as **a function of the problem input size**
- problems **input sizes** may be:
 - dimensions of lists/ matrices,
 - the number of values to be added,
 - the number of values among which we search
 - the number of values to be sorted...



Big-O

big Ω
 \ominus

- An algorithm is rated in terms of some reference function. It is said to be in the order (big-O) of some reference function:
 - $O(n)$, $O(n^2)$, $O(\log n)$, etc.
- Intuitively, that means that for sufficiently large n , the time that the algorithm will take (to execute the critical operation) will be proportional to n , n^2 , $\log n$, etc, where n is the size of the problem.



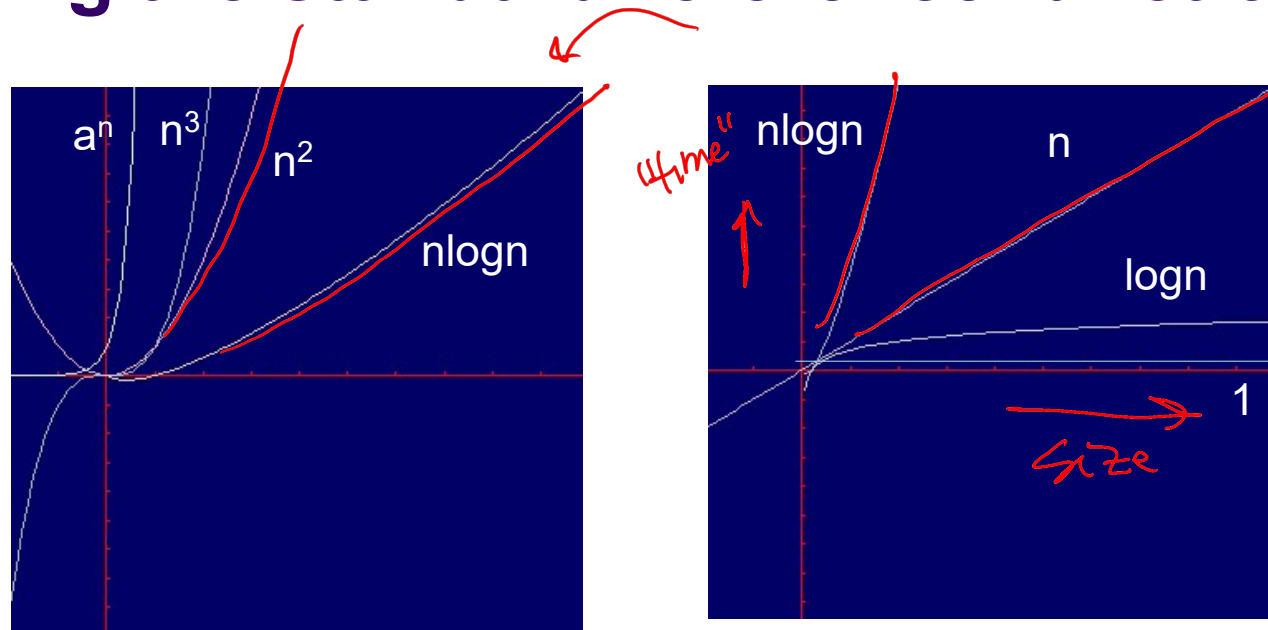
Standard Reference functions

Category	Reference Function
Constant	1
Logarithmic	$\log_2(n)$
Linear	n
nlogn	$n \log_2(n)$
Quadratic	n^2
Cubic	n^3
Exponential	$a^n, a > 1$

$\log_2 8 = 3$ because $2^3 = 8$
 $\log_2 n = x$ when $2^x = n$

$n * \log_2 n$
↑
multiplied

Comparing the standard reference functions





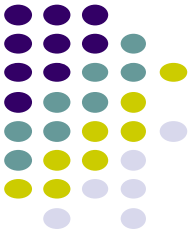
What could be an algorithm of order...

- $O(1)$? Color shoes first person in line
- $O(n)$? Linear search (in the worse case)
- $O(n^2)$? Selection sort

As n gets larger... (as the size of the problem is larger...)



n	Ex: Get first element in list always constant $O(1)$	Ex: Linear <u>search</u> $O(n)$	Ex: Binary <u>search</u> $O(\log n)$	Ex: Selection <u>sort</u> $O(n^2)$	Ex: Merge <u>sort</u> $O(n \log n)$
10	1	10	3.32	100	33.2
100	1	100	6.64	10,000	664
1,000	1	1000	9.96	1,000,000	9,960
10,000	1	10,000	13.28	100,000,000	132,800



Comparing algorithms:

- Which algorithm is preferred for large n if we are choosing based on time complexity?

We prefer an algorithm growing slowest as the size of the problem increases

Time needed to process n items for 6 algorithms

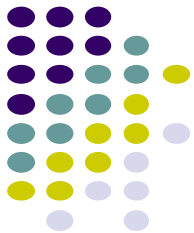
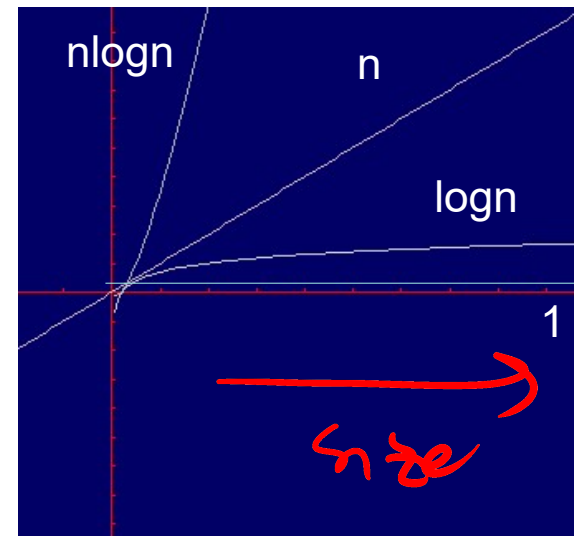
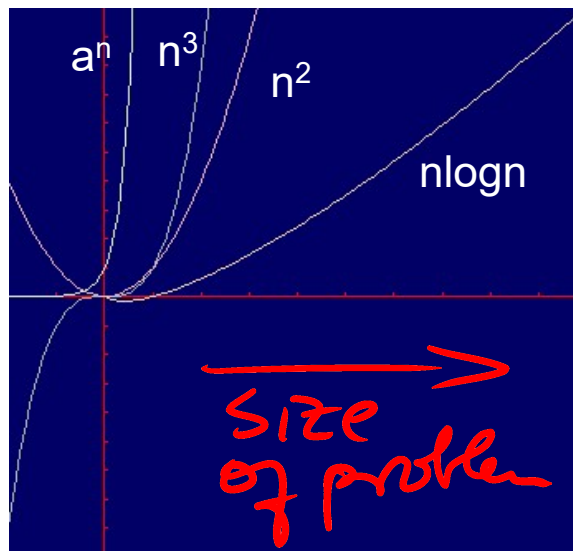
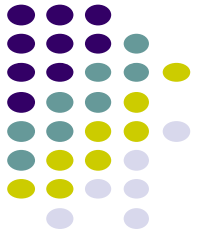


Table 4.2, Gossett

$g_i(n)$	<u>$n=10,000$</u>	<u>$n=100,000$</u>	$n=1,000,000$	<u>$n=250,000,000$</u>
$\log_2(n)$	13 nanosecs	17 nanosecs	20 nanosecs	28 nanosecs
<u>n</u>	<u>0.00001 secs</u>	0.0001 ses	0.001 secs	<u>0.25 secs</u>
$n\log_2(n)$	0.00013 secs	0.00166 secs	0.01993 secs	6.97434 secs
<u>n^2</u>	<u>0.1 secs</u>	10.0 secs	16 mins, 40 secs	<u>≈ 1.98 years</u>
<u>n^3</u>	<u>16 mins, 40 secs</u>	≈ 11 days, 14 hs	32 years	<u>≈ 500 million years</u>
<u>2^n</u>	<u>$\approx 6 \times 10^{2993}$ years</u>	-----Too long to contemplate -----		

Comparing the standard reference functions



- (A) algorithm in $O(1)$ is preferred as n gets larger
- (B) algorithm in $O(n^2)$ is preferred as n gets larger