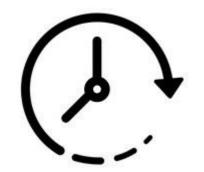
ANALYSIS & DESIGN OF ALGORITHMS

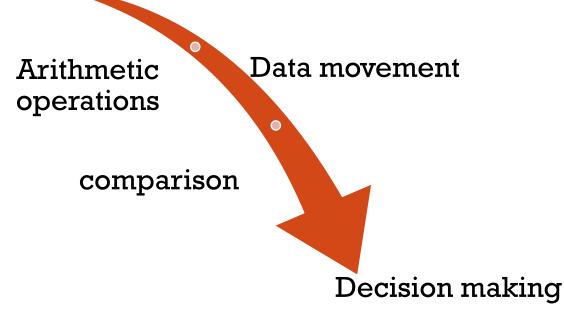
Dr. Sondos Fadl





• The running time of an algorithm on a particular input is the number of primitive operations or "steps" executed before termination.

Approximate not exact!





PRIMITIVE OPERATIONS

• Ex: Find min value of an array?



EFFICIENCY VS. SPEED

Array size 10^6

• Ex: sorting numbers:

Ahmed's computer=10⁹ operations /sec

Ahmed's Algorithm= $2n^2$ operations

Time=
$$\frac{Algorithm\ complexity}{device\ speed} = \frac{2 \times (10^6)^2}{10^9} = 2000\ sec$$

Mohamed's computer=10⁷ operations /sec

Mohamed's Algorithm= $50n \log n$ operations

20 times better !!!

Time=
$$\frac{Algorithm\ complexity}{device\ speed} = \frac{50 \times 10^6\ \log 10^6}{10^7} = 100\ sec$$

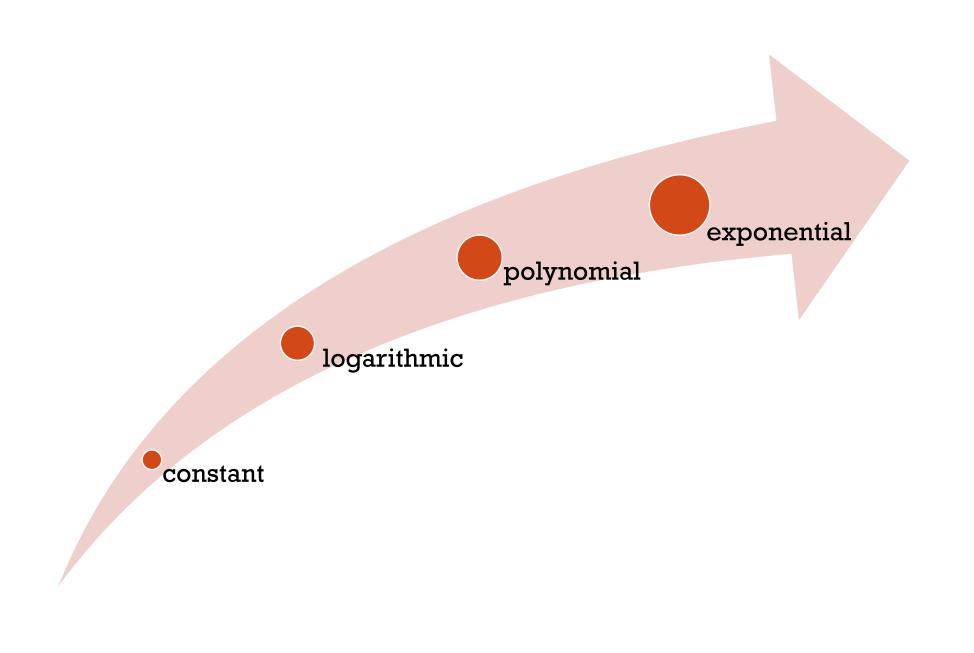


TYPICAL RUNNING TIME

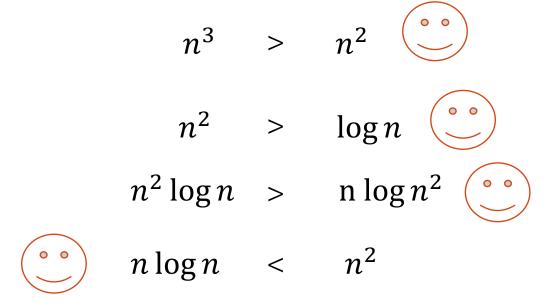
- 1 (constant)
- $\log n$ (logarithmic)
- n^k (polynomial)
 - *n* linear
 - n^2 quadratic
 - n^3 cubic
- 2^n (exponential)



Input size	n	n^2	$\log_{10} n$	
10	10	100	1	
100	100	10000	2	
10^{6}	10 ⁶	$(10^6)^2$	6	









LOOP

```
For i=1 to n

print 'Hi'
```

```
For i=1 to n

if i <11

print 'Hi'

else

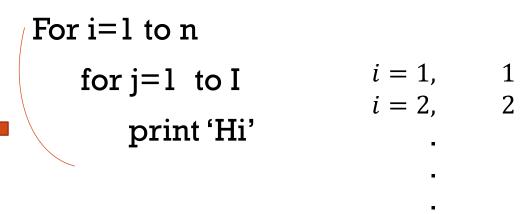
break
```

LOOP

Nested loop:

For i=1 to n
$$\sum_{i=1}^{n} n \leftarrow n^{2}$$
for j=1 to n
$$n \text{ print 'Hi}$$

Dependent nested loop:



i = n,

n



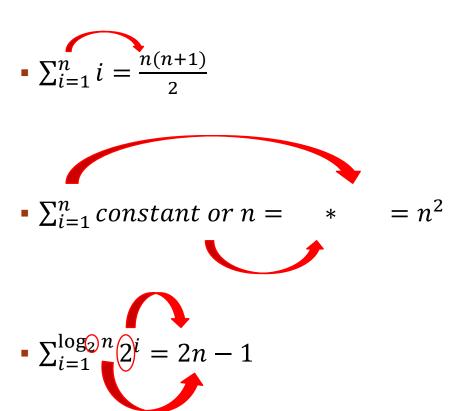
SUM EQUATIONS

Why study summations?

- 1. In general, the running time of a *loop* can be expressed as the <u>sum</u> of the running time of each iteration.
- 2. Summations come up in solving recurrences.



SUM EQUATIONS





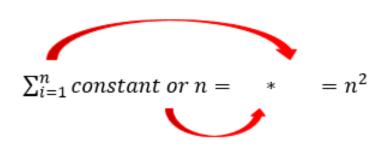
$$\sum_{i=j}^{n} 1 = n - j + 1$$

$$\sum_{i=1}^{n} x^{i} = \frac{x^{n+1}-1}{x-1}$$

 $x \rightarrow constant$

For
$$(i=1, i \le n, i++)$$

$$\sum_{i=1}^{n} 1$$
Print i
end



$$T(N) = \sum_{i=1}^{n} 1 = 1 * n = n$$

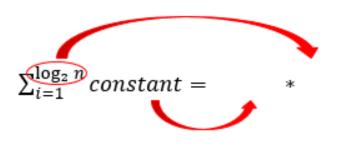


For (i=n-1, i
$$\geq$$
 1,i--)
$$\sum_{i=1}^{n-1} 1 \quad \begin{cases} \text{Print i} \\ \text{end} \end{cases}$$

$$T(N) = \sum_{i=1}^{n-1} 1 = 1 * (n-1) = n-1$$



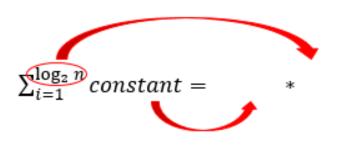
$$\sum_{i=1}^{\log_2 n} 1 \begin{cases} \text{For (i=1, i \le n, i*=2)} \\ \text{Print i} \end{cases}$$
end



$$T(N) = \sum_{i=1}^{\log_2 n} 1 = 1 * \log_2 n = \log_2 n$$



$$\sum_{i=1}^{\log_2 n} 1 \begin{cases} For (i=1, i \le n, i/=2) \\ Print i \end{cases}$$
end



$$T(N) = \sum_{i=1}^{\log_2 n} 1 = 1 * \log_2 n = \log_2 n$$



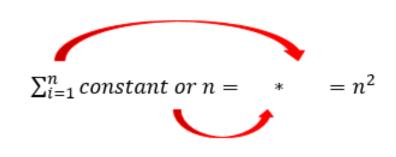
For
$$(i=1, i \le n, i++)$$

For $(j=1, j \le n, j++)$

print "hi"

end

end



$$T(N) = \sum_{i=1}^{n} \sum_{j=1}^{n} 1 = \sum_{i=1}^{n} n = n^{2}$$



INSERTION SORT

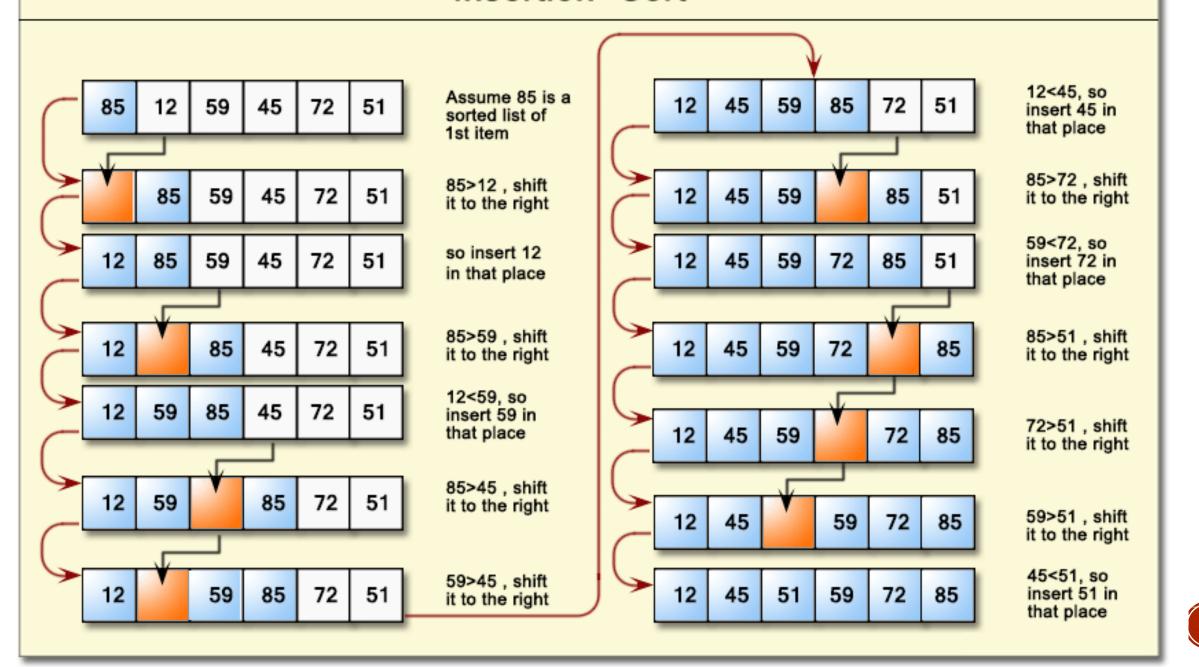
• Our first algorithm, insertion sort, solves the sorting problem.

Input: A sequence of *n* numbers $\langle a_1, a_2, \dots, a_n \rangle$.

Output: A permutation (reordering) $\langle a'_1, a'_2, \dots, a'_n \rangle$ of the input sequence such that $a'_1 \leq a'_2 \leq \dots \leq a'_n$.



Insertion Sort



INSERTION SORT

```
1: for j = 2 to A.length do
2: key = A[j]
3: i = j - 1
4: while i > 0 and A[i] > key do
       A[i+1] = A[i]
5:
6: i = i - 1
7: end while
8: A[i+1] = key
9: end for
```

$$T(N) = \sum_{j=2}^{n} t$$

The best case occurs

• If the array is already sorted, Thus t_j = 1 for j = 2;3; ...; n, and the best-case running time is

$$T(N) = \sum_{j=2}^{n} 1 = n$$

The worst case occurs

• If the array is in reverse sorted order that is, in decreasing order—the worst case results. We must compare each element A[j] with each element in the entire sorted subarray A[1..j-1], and so t_j = j for j = 2;3; ...;n

$$T(N) = \sum_{j=2}^{n} j = \frac{n(n+1)}{2} = \underbrace{\binom{n^2 + n}{2}}_{2}$$

$$O(n^2)$$

