

## Computational Complexity (Algorithm Analysis).

- + need method to evaluate the runtime of an algorithm (i.e. how fast is the algorithm).
- + often care about worst-case analysis (upper bound on performance)
- + want to understand how an algorithm will perform as the amount of input data increases
- + sometimes it might be useful to also examine the average case execution, if an "average case" can be well defined.

### Big-O Notation:

If  $g(n)$  is an upper bound of  $f(n)$ , then for some constant  $c$ , it is possible to find a value of  $n, n_0$ , for which any value of  $n \geq n_0$  will result in  $f(n) \leq cg(n)$

$n$  is typically a variable used to represent the size of data.

- + For big-O notation, we want to capture the growth rate, so we do not have to be extremely precise.

### Rules:

$$O(c) = O(1) \quad \equiv \text{constant runtime not dependent on } n$$

$$O(cT) = cO(T) = O(T) \quad \equiv \text{executing tasks with a constant multiplicative number have the same growth rate}$$

$$O(T_1) + O(T_2) = O(T_1 + T_2) = \max(O(T_1), O(T_2)) \quad \equiv \text{if one algorithm has a dominant growth rate, we only need to capture this growth rate}$$

$$O(T_1)O(T_2) = O(T_1T_2)$$

Example:  $T(n) = 3n^2 + 10n + 10$

$$O(T(n)) = O(3n^2 + 10n + 10) = O(3n^2) = \underline{\underline{O(n^2)}}$$

### Common Algorithm Complexities:

