

Asymptotic notations are used to compare algorithms by providing a mathematical boundary for their runtime functions. These notations define an upper bound, a lower bound, or both, helping to analyze the efficiency of algorithms as input size grows.

Big-O: upper bound on runtime growth denoted $O(f(n))$ which is
 $g(n) : 0 < g(n) \leq cf(n)$ whenever $n \geq n_0$ for some $c > 0, n_0 > 0$

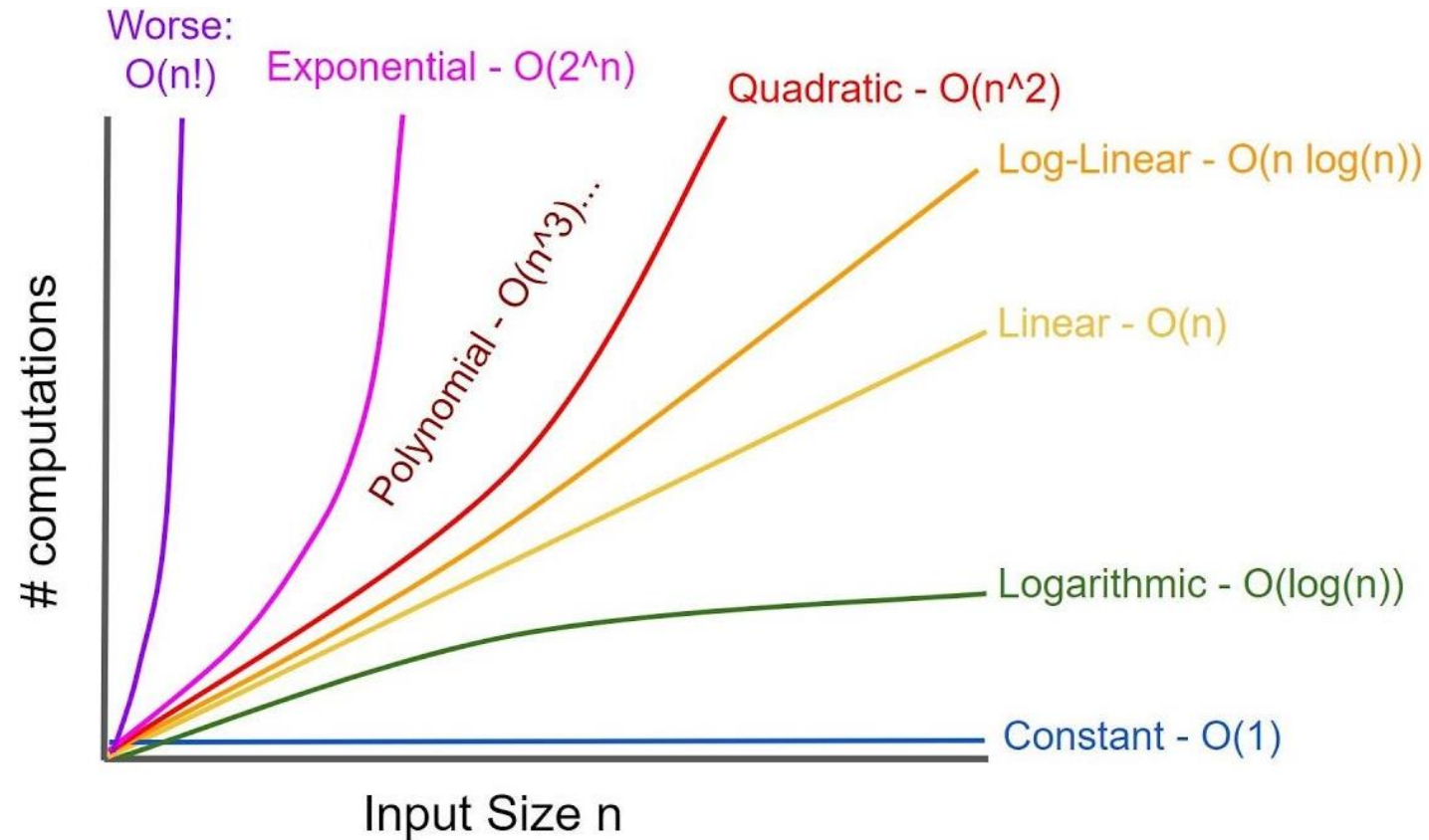
Big-Omega: lower bound on runtime growth denoted $\Omega(f(n))$ which is
 $g(n) : 0 < cf(n) \leq g(n)$ whenever $n \geq n_0$ for some $c > 0, n_0 > 0$

Big-Theta: tight bound (both upper and lower bound) denoted $\Theta(f(n))$ which is
 $g(n) : c_1f(n) \leq g(n) \leq c_2f(n)$ whenever $n \geq n_0$ for some $c_2 \geq c_1 > 0, n_0 > 0$

Common Big-O Functions

- **$O(1)$** : constant time (independent of input size).
- **$O(\log_2 n)$** : logarithmic time (e.g., binary search).
- **$O(n)$** : linear time (e.g., iterating over an array).
- **$O(n^2)$** : quadratic time (e.g., nested loops).
- **$O(2^n)$** : exponential time (e.g., brute-force recursive algorithms).

Visualizing Big-O



```
1  template <typename T>
2  void Swap(T& a, T& b)
3  {
4      T temp = a;
5      a = b;
6      b = temp;
7  }
```

Steps	Code	Cost (c _i)	Time (t _i)
1	<code>T temp = a;</code>	c ₁	1
2	<code>a = b;</code>	c ₂	1
3	<code>b = temp;</code>	c ₃	1

$$T(n) = \sum_{i=1}^m c_i \cdot t_i$$

$$T(n) = c_1(1) + c_2(1) + c_3(1)$$

$$T(n) = 1 + 1 + 1$$

$$T(n) = 3$$

$$O(1)$$

```
1  template <typename T>
2  T Maximum(T arr[], int size)
3  {
4      T max = arr[0];
5      int i = 1;
6      while (i < size)
7      {
8          if (arr[i] > max)
9          {
10             max = arr[i];
11         }
12         i++;
13     }
14     return max;
15 }
```

Steps	Code	Cost (c _i)	Time (t _i)
1	<code>T max = arr[0];</code>	c ₁	1
2	<code>int i = 1;</code>	c ₂	1
3	<code>while (i < size)</code>	c ₃	n - 1 + 1
4	<code>if (arr[i] > max)</code>	c ₄	n - 1
5	<code>max = arr[0];</code>	c ₅	n - 1
6	<code>i++;</code>	c ₆	n - 1
7	<code>return max;</code>	c ₇	1

$$T(n) = c_1(1) + c_2(1) + c_3(n) + c_4(n - 1) + c_5(n - 1) + c_6(n - 1) + c_7(1)$$

$$T(n) = 1 + 1 + n + n - 1 + n - 1 + n - 1 + 1$$

$$T(n) = 4n$$

$$O(n)$$

```
5  string Square(int length)
6  {
7      string result;
8      for (int i = 0; i < length; i++)
9      {
10         for (int j = 0; j < length; j++)
11         {
12             result += "*";
13         }
14         result += "\n";
15     }
16     return result;
17 }
```


Steps	Code	Cost (c _i)	Time (t _i)
1	<code>string result;</code>	c ₁	1
2	<code>for (int i = 0; i < length; i++)</code>	c ₂	1 + n + n + 1
3	<code>for (int j = 0; j < length; j++)</code>	c ₃	n + n ² + n ² + n
4	<code>result += "*";</code>	c ₄	n ²
5	<code>result += "\n";</code>	c ₅	n
6	<code>return result;</code>	c ₆	1

$$T(n) = c_1(1) + c_2(2n + 2) + c_3(2n^2 + 2n) + c_4(n^2) + c_5(n) + c_6(1)$$

$$T(n) = 1 + 2n + 2 + 2n^2 + 2n + n^2 + n + 1$$

$$T(n) = 3n^2 + 5n + 4$$

$$O(n^2)$$