

# Tutorial 06: Complexity Analysis

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CSCI2520 - DATA STRUCTURES AND APPLICATIONS

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# Outlines

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- Big-O Notation
- Big-O Simplification Rules
- Big-O Formal Definition
- Determining Computational Complexity from Code Structure
- Binary Search

# Big-O Notation

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- The relationship between the problem size  $N$  and the performance of an algorithm as  $N$  becomes large is called the ***computational complexity*** (or time complexity) of the algorithm.
- The ***big-O notation*** is used to provide a *quantitative insight* as to how changes in the problem size  $N$  affect the algorithmic performance as  $N$  becomes *large*.

# Big-O Simplification Rules

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- *Eliminate any term whose contribution to the total becomes **insignificant** as N becomes large.*
  - When a formula involves a summation of several terms, the *fastest growing term* alone will *control* the running time of the algorithm for large N.
  - Example:  $N^3 + 1000N^2 + N = O(N^3)$
- *Eliminate any **constant factors**.*
  - What we want to capture in computational complexity is how changes in N affect the algorithmic performance.
  - Example:  $0.1N^3 = O(N^3)$

# Exercise 1

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- Order the following functions by ascending order of growth rates.
  1.  $0.0001N^3 + N$
  2.  $2N^{1/2} + \log N + 1$
  3.  $3N^{3/2} - 2N^{1/2}$
  4.  $N + 5\log N + 100$
  5.  $2N^2 + 9N\log N + 1$

# Exercise 1 Solution

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- Order the following functions by ascending order of growth rates.
    - $2N^{1/2} + \log N + 1 = O(N^{1/2})$
    - $N + 5\log N + 100 = O(N)$
    - $3N^{3/2} - 2N^{1/2} = O(N^{3/2})$
    - $2N^2 + 9N\log N + 1 = O(N^2)$
    - $0.0001N^3 + N = O(N^3)$
- The answer is 2 4 3 5 1

# Big-O Formal Definition

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- Definition:  $t(N) = O(f(N))$  if and only if
  - there are positive constants  $n_0$  and  $c$
  - for every value of  $N \geq n_0$ , the following condition holds:
$$t(N) \leq c \times f(N)$$
- As long as  $N$  is “*large enough*”,  $t(N)$  is always bounded by a constant multiple of  $f(N)$ .

# Exercise 2

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- What are the  $c$  to make  $t(N) \leq c \times f(N)$  when  $n_0 = 1$ ?
  - $2N^{1/2} + \log N + 1 = O(N^{1/2})$
  - $N + 5\log N + 100 = O(N)$
  - $3N^{3/2} - 2N^{1/2} = O(N^{3/2})$
  - $2N^2 + 9N\log N + 1 = O(N^2)$
  - $0.0001N^3 + N = O(N^3)$



# Exercise 2 Solution

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- What are the  $c$  to make  $t(N) \leq c \times f(N)$  when  $n_0 = 1$ ?
  - $2N^{1/2} + \log N + 1 \leq 2N^{1/2} + N^{1/2} + N^{1/2} = 4 \times N^{1/2}$
  - $N + 5\log N + 100 \leq N + 5N + 100N = 106 \times N$
  - $3N^{3/2} - 2N^{1/2} \leq 3 \times N^{3/2}$
  - $2N^2 + 9N\log N + 1 \leq 2N^2 + 9N^2 + N^2 = 12 \times N^2$
  - $0.0001N^3 + N \leq 0.0001N^3 + N^3 = 1.0001 \times N^3$
- Do you still remember big-O simplification rules?
- Using larger  $n_0$ ,  $c$  can be reduced. The requirement of  $n_0 = 1$  here is just for illustration.

# Determining Computational Complexity from Code Structure

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- In general, we can determine the time complexity simply by finding the piece of the code that is executed **most often**.
- However, if an expression or statement involves ***function calls***, it must be accounted *separately*.

# Exercise 3 Binary Search

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- A binary search algorithm finds the position of a target value within a *sorted* array.
- Compare the target value to the value of the *middle* element of the sorted array.
  - If the target value is equal to the middle element's value, then the position is returned and the search is finished.
  - If the target value is less than the middle element's value, then the search continues on the lower half of the array
  - If the target value is greater than the middle element's value, then the search continues on the upper half of the array.
  - In each iteration, half of the elements are eliminated.

# Exercise 3 Binary Search

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- Input
  - target value: 10
  - input array: ( 1 2 3 4 5 6 7 8 10 11 12 13 24 25 26 )
- First iteration:
  - ( 1 2 3 4 5 6 7 **8** 10 11 12 13 24 25 26 )
  - go to upper half since  $10 > 8$

# Exercise 3 Binary Search

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- Second iteration:
  - ( 10 11 12 **13** 24 25 26 )
  - go to lower half since  $10 < 13$
- Third iteration:
  - ( 10 **11** 12 )
  - go to lower half since  $10 < 11$
- Fourth iteration:
  - ( **10** ) get it!

# Exercise 3 Binary Search

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- Finish the implementation of the following function.

```
int BinarySearch(int array[], int n, int tarVal)
```

- Hints
  - You may try to implement the following function first.
  - What if the array is of even length (e.g. 2)? What should be the end case?

```
int BinarySearch(int array[], int tarVal, int iMin, int iMax)
```

- Can you think of this problem recursively?
- Question: what happens if there is no target value?

# Exercise 3 Binary Search

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```
int BinarySearch(int array[], int tarVal, int iMin, int iMax){  
    if (iMin > iMax)  
        return -1;  
    int iMid = (iMin + iMax) / 2;  
    if (array[iMid] == tarVal)  
        return iMid;  
    else{  
        if (array[iMid] > tarVal) iMax = iMid - 1;  
        else iMin = iMid + 1;  
        return BinarySearch(array, tarVal, iMin, iMax);  
    }  
}  
int BinarySearch(int array[], int n, int tarVal){  
    return BinarySearch(array, tarVal, 0, n - 1);  
}
```

Two *tests* to stop or continue

Two *end cases* to terminate the recursion

A *recursive call* to continue recursion

# Exercise 3 Binary Search

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- What is the complexity of binary search in worst case?
  - Hint: what are worst cases?

```
int BinarySearch(int array[], int tarVal, int iMin, int iMax){  
    if (iMin > iMax) return -1;  
    int iMid = (iMin + iMax) / 2;  
    if (array[iMid] == tarVal) return iMid;  
    else{  
        if (array[iMid] > tarVal) iMax = iMid - 1;  
        else iMin = iMid + 1;  
        return BinarySearch(array, tarVal, iMin, iMax);  
    }  
}
```

$O(1)$

$O(?)$



# Exercise 3 Binary Search

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- The time spend at each level is  $O(1)$ .
- There are totally  $k$  levels where  $2^k = N$ , i.e.,

$$k = \log_2 N$$

- Hence, the time complexity of binary search in worst case is

$$O(\log_2 N)$$

- We usually omit the logarithmic base in writing complexities. That is, we usually write:

$$O(\log N)$$

# Exercise 4

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- What is the time complexity of following functions?

```
void function1(String s) {  
    for (int i = 0; i < StringLength(s); ++i) {  
        ChangeToDigit(s[i]);  
    }  
}  
// the time complexity of ChangeToDigit() is O(1).  
// the time complexity of StringLength() is O(n), where n is the length of the String.
```

$$O(N^2)$$

# Exercise 4

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- What is the time complexity of following functions?

```
int function2(int x) {  
    if (x == 1 || x == 0) return 1;  
    else  
        return function2(x / 2) + function2(x / 2);  
}
```

$$T(n) = 2T\left(\frac{n}{2}\right) = 2^2T\left(\frac{n}{2^2}\right) = \dots = 2^kT(1) \quad (k = \log_2 n)$$
$$O(n)$$