

# Algorithm Analysis

Class 5

# Big-Theta

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Big-Theta is a two-part if and only if, so to prove it requires two subproofs:

1.  $T(n) \in O(n^2)$
2.  $T(n) \in \Omega(n^2)$

## Subproof 1: Big-O

Show that  $T(n) \in O(n^2)$  by finding  $c_1$  and  $n_0$  such that  $T(n) \leq c_1 n^2$  when  $n \geq n_0$

$$n^2 + 5n - 3 \leq c_1 n^2$$

or equivalently

$$c_1 \geq 1 + \frac{5}{n} - \frac{3}{n^2}$$

what is the maximum value of  $1 + \frac{5}{n} - \frac{3}{n^2}$ ?

$c_1$  must be greater than or equal to that value

## Big-O

$$\frac{d}{dn} \left( 1 + \frac{5}{n} - \frac{3}{n^2} \right) = \frac{6 - 5n}{n^3}$$

this has real zero at  $n = \frac{6}{5}$ , 2nd derivative shows it's a maximum

the value of  $1 + \frac{5}{n} - \frac{3}{n^2}$  at  $n = \frac{6}{5}$  is  $\frac{37}{12}$

thus we choose  $c_1 = 4$  and  $n_0 = 2$

and we can write

$$n^2 + 5n - 3 \leq 4n^2 \text{ when } n \geq 2$$

and big-Oh is proved by the definition.

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## Subproof 2: Big-Omega

Show that  $T(n) \in \Omega(n^2)$  by finding  $c_2$  and  $n_0$  s.t.  $T(n) \geq c_2 n^2$  when  $n \geq n_0$

$$n^2 + 5n - 3 \geq c_2 n^2$$

or equivalently

$$c_2 \leq 1 + \frac{5}{n} - \frac{3}{n^2}$$

what is the minimum value of  $1 + \frac{5}{n} - \frac{3}{n^2}$ ?

$c_2$  must be equal to or smaller than that value

## Big-Omega

$$\lim_{n \rightarrow \infty} \left( 1 + \frac{5}{n} - \frac{3}{n^2} \right) = 1$$

thus we choose  $c_2 = \frac{1}{2}$  and,  $n_0 = 2$  still works,

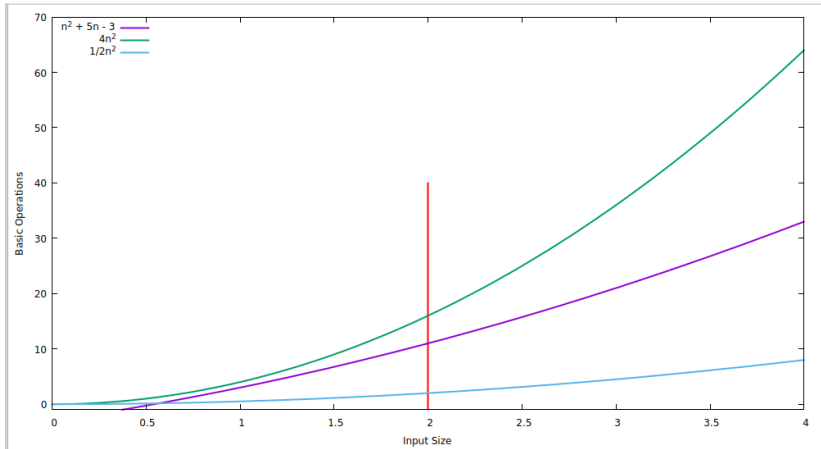
and so we can write

$$n^2 + 5n - 3 \geq \frac{1}{2}n^2 \text{ when } n \geq 2$$

and big-Omega is proved by the definition.



# Graphic Illustration



## Rules of Thumb

- lower-order terms are irrelevant — we can ignore them
- the highest-order term corresponds to the basic operations that occur most frequently — these are what we count, e.g.,

$$f(n) \leq n^3 - \frac{27}{3}n^2 + 16n - 4 \in O(n^3)$$

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- leading coefficients indicate how many most-frequently occurring basic operations are performed
- since we know we can ignore leading coefficients, the **exact count** of basic operations does not affect the final analysis, e.g.,

$$f(n) \geq 3n^2 \in \Omega(n^2)$$

## Analyze an Algorithm

- in the previous set of slides we analyzed the continuous function  $n^2 + 5n - 3$
- in CS, we do not wish to analyze continuous functions
- we wish to analyze discrete algorithms
- analyze the algorithm `find_max` (only the function)  
(this means find a big-Theta set or a pair of big-Oh and big-Omega sets for the algorithm)

we must determine and state several things:

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we must determine and state several things:

1. what is the input and how do we measure its size?  
the input is the array; the size is the number of elements
2. does the input arrangement matter?  
no; the algorithm cannot end early even if the max element is the very first element

## Counting Basic Operations

3. what basic operations are we counting? when `find_max` runs, what statements are executed, and how many times?

let  $n$  be the size of the array (line 45)



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let  $n$  be the size of the array (line 45)

- line 44: one assignment (regardless of  $n$ )
- line 45: one assignment (regardless of  $n$ )
- line 47: one assignment or increment, and one comparison  $\times n = 2n$  operations
- line 49: one array access, one comparison, and one assignment  $\times n - 1 = 3(n - 1)$  operations
- line 51: we do **not** count return statements

if  $n$  is 3, exactly how many operations are performed?

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if  $n$  is 3, exactly how many operations are performed? **14**

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how many for  $n = 5$ ?

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what is a formula for  $T(n)$ , and what is its analysis?

$$T(n) = 5n - 1$$

$$T(n) \in \Theta(n)$$

# Number of Operations

- some algorithms execute an **exact** number of operations as a function of input size:  $T(n) = f(n)$



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- some algorithms execute an **exact** number of operations as a function of input size:  $T(n) = f(n)$
- some algorithms execute **at most** a number of operations based on input size:  $T(n) \leq f(n)$  (worst case)
- and execute **at least** different number of operations:  $T(n) \geq g(n)$  (best case)
- in other words:  $g(n) \leq T(n) \leq f(n)$

# Efficiency Classes

## big-Theta

- if we can determine an algorithm's **exact** number of operations, we can find  $c_1$ ,  $c_2$ , and  $n_0$  for some **standard function**  $f(n)$  and thus we have  $T(n) \in \Theta(f(n))$

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## big-O

- if we can determine that an algorithm executes at most a certain number of operations, we can find a  $c$  and  $n_0$  for some **standard function**  $f(n)$  and thus we have  $T(n) \in O(f(n))$

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## big-Omega

- if we can determine that an algorithm executes at least a certain number of operations, we can find a  $c$  and  $n_0$  for some **standard function**  $f(n)$  and thus we have  $T(n) \in \Omega(f(n))$

# Big-Theta

to analyze an algorithm, we must find an upper and a lower bound for its behavior

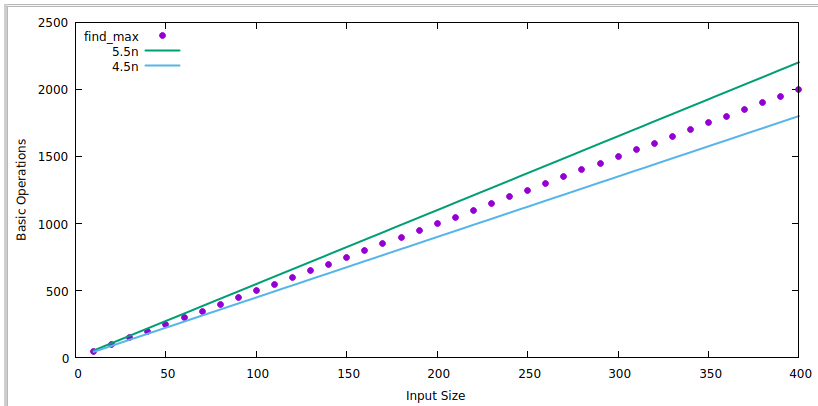
- if we find a big-Theta set for an algorithm
- that function is **both** an upper and a lower bound for the algorithm
- in this case we know exactly how the algorithm scales

# Big-Oh and Big-Omega

- if we cannot find a single function and thus a big-Theta set
- we must find a big-Oh set for its upper bound
- and a big-Omega set for its lower bound

# Big-Theta

- we saw an illustration of big-Theta earlier with the function we analyzed
- `find_max` has a similar picture



## Big-Oh and Big-Omega

if there is not a big-Theta set, we have one big-Oh set and a different big-Omega set

