

# SCC.121: ALGORITHMS AND COMPLEXITY

## Sentinel and Binary Search Algorithms

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# Today's Lecture

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**Aim:** Look at searching algorithms sentinel search and binary search and their time complexity. Introduce asymptotic analysis and the growth of functions.

## Learning objectives:

- Know how sentinel and binary search work and be able to estimate their time complexity
- To know what is meant by the growth rate of functions and be able to determine the order of growth of simple functions

# Outline

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- Linear Search vs Sentinel Search
- Binary Search
- Introduction to growth of functions

# Outline

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- **Linear Search vs Sentinel Search**
- Binary Search
- Introduction to growth of functions

# Linear Search

The overall program time (**Worst case**)?

- $T(N) = 3N+3$

```
int isInArray(int theArray[], int N, int iSearch)
{
    o1 → 1 o2 → N+1 o3 → N
    for (int i = 0; i < N; i++)
        o4 → N
        if (theArray[i] == iSearch)
            return 1;    o5 → 0
    return 0;    o6 → 1
}
```

# Sentinel Search

- 
- When a linear search is performed on an array of size  $N$  then in the worst case a total of  $(N + 1)$  comparisons are made for the index of the element to be compared so that the **index is not out of bounds of the array**.
  - Sentinel search is a type of Linear search where **the number of comparisons is reduced** as compared to the linear search.

# Sentinel Search

```
int isInSentinel(int theArray[], int N, int iSearch)
{
    // Sentinel Search code
}
```

	0	1	2	3	4	5	6	7	8	9
theArray	5	17	6	13	28	1	3	6	44	32

iSearch 6      The number we are looking for in the array

- ❑ Check if iSearch is not at the end of theArray. Then, replace the number at the end of array with iSearch

	0	1	2	3	4	5	6	7	8	9
theArray	5	17	6	13	28	1	3	6	44	6

- ❑ Perform Linear Search

iSearch

# Sentinel Search

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```
int isInSentinel(int theArray[], int N, int iSearch)
{
    if (theArray[N-1] == iSearch)
        return 1;

    theArray[N-1] = iSearch;

    for (int i=0; ; i++) {
        if (theArray[i] == iSearch)
            return (i < (N-1));
    }
}
```





**How many times do we execute `i++` in sentinel search in the worst case?**

# Sentinel Search (worst case)

```
int isInSentinel(int theArray[], int N, int iSearch)
{
    if (theArray[N-1] == iSearch) o1 → 1
        return 1; o2 → 0

    theArray[N-1] = iSearch; o3 → 1
    o4 → 1 o4 → N-1
    for (int i=0; ; i++) {

        if (theArray[i] == iSearch) o4 → N
            return (i < (N-1)); o5 → 1
    }
}
```

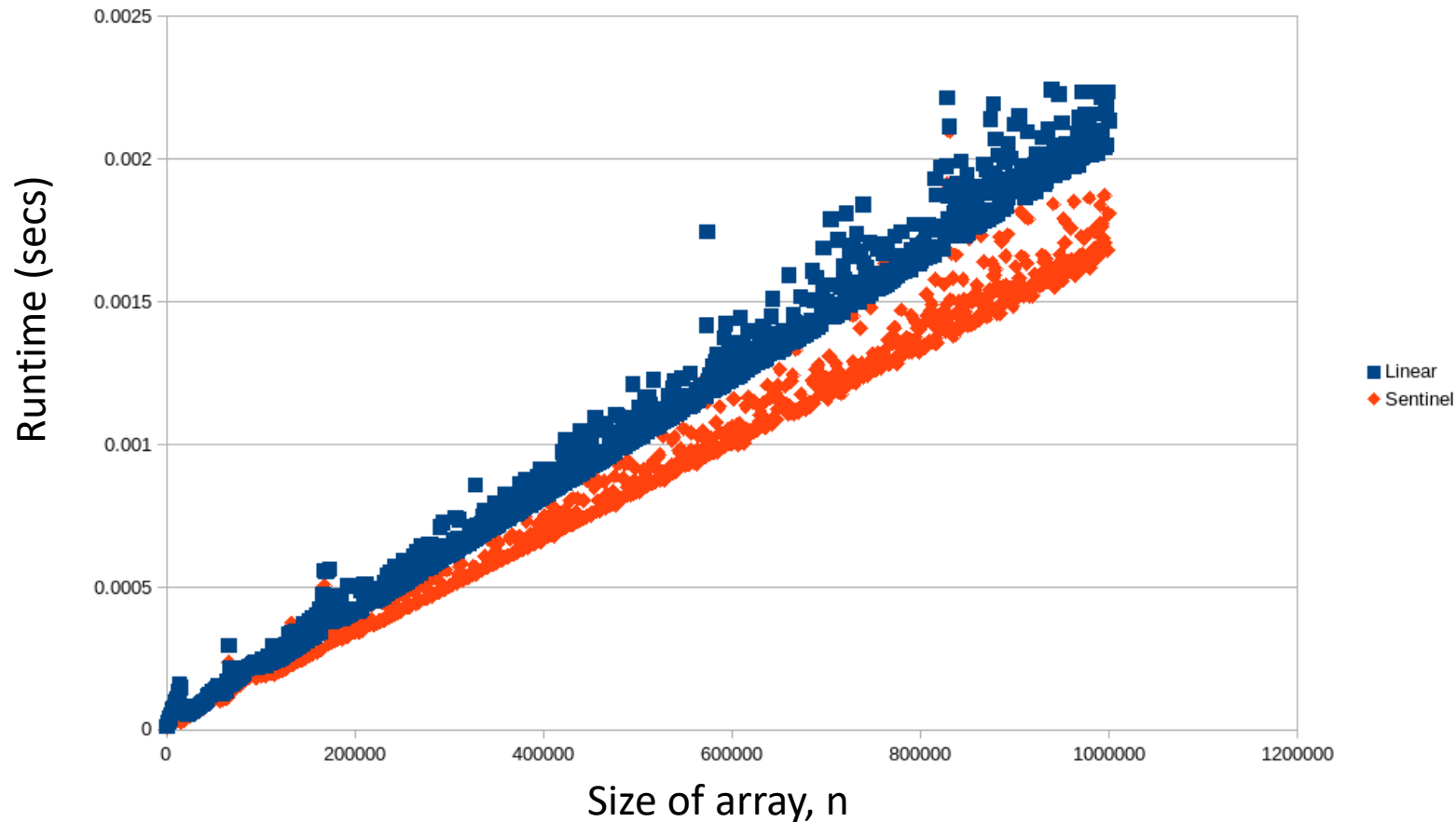
# Compare Linear vs. Sentinel Search

```
int isInArray(int theArray[], int N,
int iSearch)
{
    for (int i=0; i<N; i++)
        if (theArray[i] == iSearch)
            return 1;
    return 0;
}
```

```
int isInSentinel(int theArray[], int N,
int iSearch)
{
    if (theArray[N-1] == iSearch)
        return 1;
    theArray[N-1] = iSearch;
    for (int i=0;  ; i++)
    {
        if (theArray[i] == iSearch)
            return (i < (N-1));
    }
}
```

- Which one is better?
  - Count the number of instructions
    - Linear:  $3N+3$  (Worst-case)
    - Sentinel:  $2N+3$  (Worst-case)
  - Both searches have **linear** time complexity

# Linear vs Sentinel Search: Actual Runtimes (worst case)



# Outline

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- Linear Search vs Sentinel Search
- **Binary Search**
- Growth of functions

# Binary Search

- Binary search works on sorted arrays
- Locates a target value in a sorted array by successively eliminating half of the array from consideration
- Let's assume you're given the below array and  $iSearch = 33$

[illegible]

# Binary Search

- Locates a target value in a sorted array by successively eliminating half of the array from consideration.
- Let's assume you're given the below array and *search* = 33

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
↑							↑							↑
lo							mid							hi

$$mid = \frac{(hi + lo)}{2}$$

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				↑	↑	↑								
				lo	mid	hi								

$$mid = \frac{(hi + lo)}{2}$$

# Binary Search

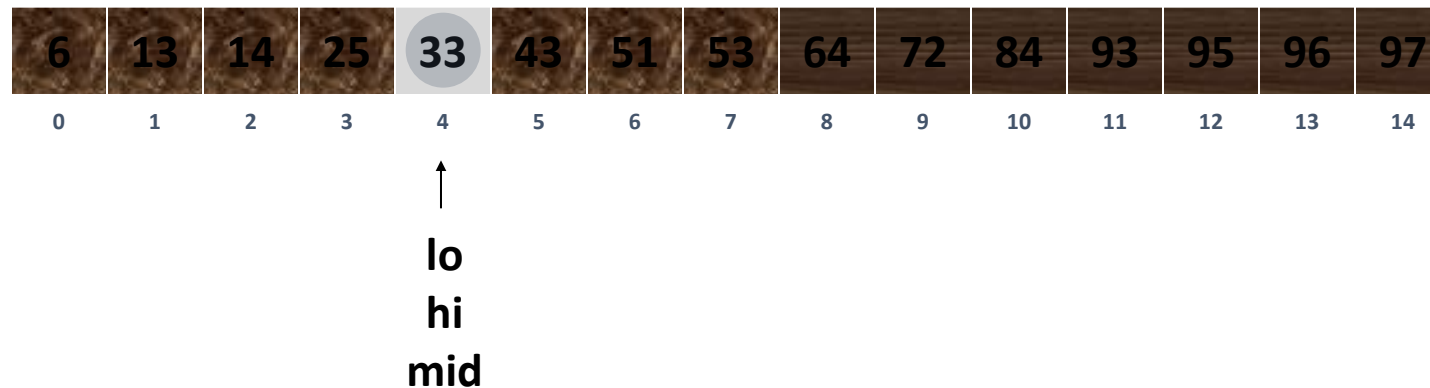
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hi

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0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

↑  
lo  
hi  
mid

# Binary Search Algorithm (an iterative implementation)

```
boolean isInBinary(int[] theArray, int N, int iSearch)
{
    int lo = 0;
    int hi = N - 1;
    int mid = 0;

    while (hi >= lo) {
        mid = (lo + hi)/2; //round to higher integer
        if (theArray[mid] == iSearch)
            return true;
        else if (theArray[mid] < iSearch)
            lo = mid + 1;
        else
            hi = mid - 1;
    }
    return false;
}
```



# Binary Search Questions

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- What is the time complexity in the Worst-case?
- What is the time complexity in the Best-case?

# Binary Search Algorithm (an iterative implementation)

```
boolean isInBinary(int[] theArray, int N, int iSearch)
{
    int lo = 0; o1
    int hi = N - 1; o2
    int mid = 0; o3

    while (hi >= lo) o4
    {
        mid = (lo + hi)/2; //round to higher integer o5
        if (theArray[mid] == iSearch) o6
            return true; o7
        else if (theArray[mid] < iSearch) o8
            lo = mid + 1; o9
        else
            hi = mid - 1; o10
    }
    return false; o11
}
```

## Worst Case:

- iSearch is not in the array
- o6 always false
- Assume o8 always false

# Binary Search (Worst Case)

Try out for an array size: here try  $N = 2^6 = 64$

	lo	hi	mid
Initially	0	63	0
After 1 <sup>st</sup> search	0	31	32
After 2 <sup>nd</sup> search	0	15	16
After 3 <sup>rd</sup> search	0	7	8
After 4 <sup>th</sup> search	0	3	4
After 5 <sup>th</sup> search	0	1	2
After 6 <sup>th</sup> search	0	0	1
After 7 <sup>th</sup> search	0	-1	0

- Here  $N = 2^6 = 64$
- Or equivalently  $7 = \log_2 64 + 1 = \log_2 N + 1$
- So, we go around the loop  $\log_2 N + 1$  times
- Each time around the loop o4, o5, o6, o8 and o10 are implemented
- o4 is implemented one additional time
- o1, o2, o3 and o11 are implemented once

Total time complexity in worst case

- $T(N) = o1 + o2 + o3 + o4 + o5 + o6 + o8 + o10 + o11$
- $T(N) = 1 + 1 + 1 + 5 * (1 + \log_2 N) + 1 + 1$
- $T(N) = 10 + 5 \log_2 N$

**Worst Case time complexity is logarithmic!  $T(N) = C_1 + C_2 \log N$**

# Binary Search Algorithm

```
boolean isInBinary(int[] theArray, int N, int iSearch)
{
    int lo = 0;
    int hi = N - 1;
    int mid = 0;

    while (hi >= lo) {
        mid = (lo + hi)/2; //round to higher integer
        if (theArray[mid] == iSearch)
            return true;
        else if (theArray[mid] < iSearch)
            lo = mid + 1;
        else
            hi = mid - 1;
    }
    return false;
}
```



**What is the time complexity class of Binary Search Algorithm in the best case?**

# Binary Search Questions

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- What is the time complexity in the Worst-case?
  - **Logarithmic:**  $T(N) = C_1 + C_2 \log N$
- What is the time complexity in the Best-case?
  - **Constant:**  $T(N) = C_1$

# Linear, Sentinel and Binary Search Summary

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- Sentinel search and Linear search algorithms are both linear in worst-case scenario, but Sentinel search requires executing fewer operations
- Binary search algorithm is more efficient comparing to Linear and Sentinel Search algorithms (Binary logarithmic in worst case), but the input array should be sorted
- Actual values of constants generally unimportant (except in specific circumstances)
- What we really care about is behaviour as the size of our input increases
  - **asymptotic behaviour**

# Outline

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- Linear Search vs Sentinel Search
- Binary Search
- **Introduction to growth of functions**



# The Growth of Functions

- 
- For example, let us assume two algorithms **A** and **B** that solve the same class of problems
  - The time complexity of **A** is  $T(n) = 5000n$ , the one for **B** is  $T(n) = 1.1^n$  for an input with  $n$  elements
  - Which algorithm is better?

# The Growth of Functions

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**Comparison:** time complexity of algorithms A and B

Input Size	Algorithm A	Algorithm B
n	$T(n) = 5,000n$	$T(n) = 1.1^n$
10	50,000	3
100	500,000	13,781
1,000	5,000,000	$2.5 \times 10^{41}$
1,000,000	$5 \times 10^9$	$4.8 \times 10^{41398}$



The time complexity of A is  $T(n) = 5000n$ , the one for B is  $T(n) = 1.1^n$  for an input with  $n$  elements, Which algorithm is better?

# The Growth of Functions

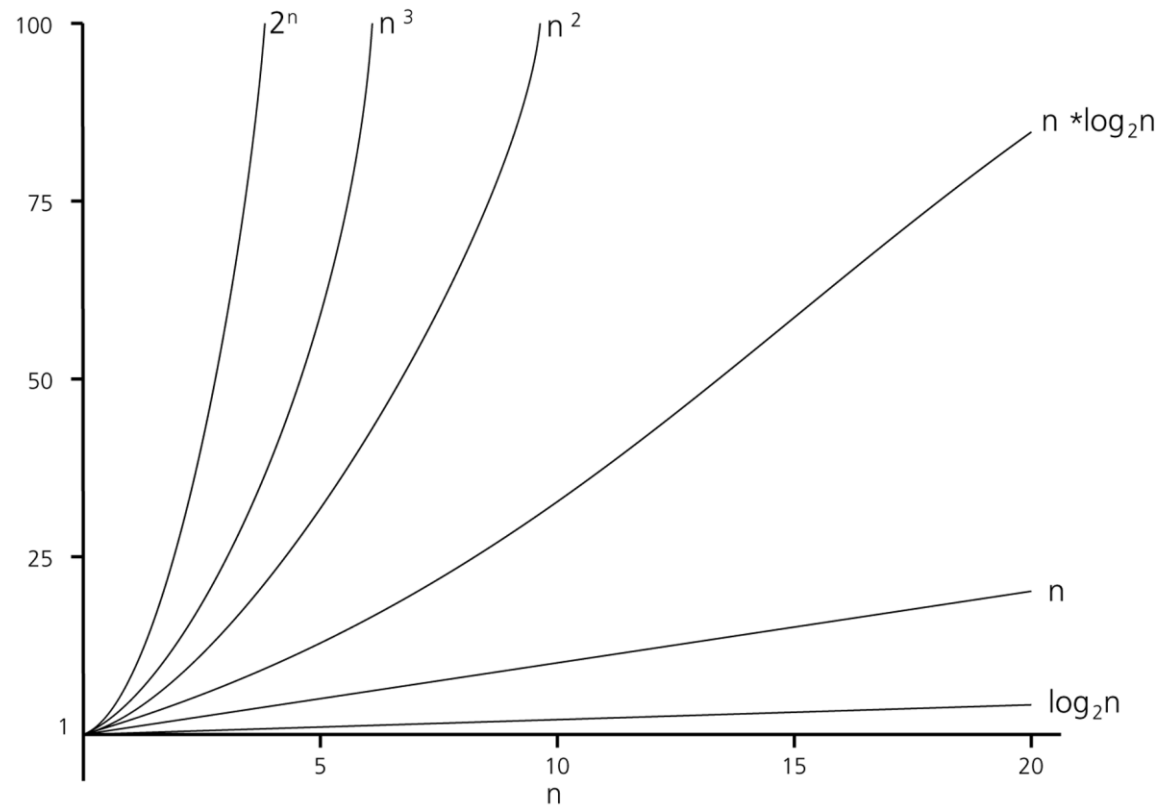
- 
- This means that algorithm B cannot be used for large inputs, while algorithm A is still feasible
  - So what is important is the **growth** of the time complexity functions
  - The growth of time complexity with increasing input size 'n' is a suitable measure for the comparison of algorithms

# The Growth of Functions (Table)

Order of Growth ↓

Function	n					
	10	100	1,000	10,000	100,000	1,000,000
1	1	1	1	1	1	1
$\log_2 n$	3	6	9	13	16	19
$n$	10	$10^2$	$10^3$	$10^4$	$10^5$	$10^6$
$n * \log_2 n$	30	664	9,965	$10^5$	$10^6$	$10^7$
$n^2$	$10^2$	$10^4$	$10^6$	$10^8$	$10^{10}$	$10^{12}$
$n^3$	$10^3$	$10^6$	$10^9$	$10^{12}$	$10^{15}$	$10^{18}$
$2^n$	$10^3$	$10^{30}$	$10^{301}$	$10^{3,010}$	$10^{30,103}$	$10^{301,030}$

# The Growth of Functions (Plot)



# The Growth of Functions (Summary)

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$c$  is a constant  $0 < c < 1$

$$\log n < n^c < n < n \log n < n^2 < n^3 \dots < 2^n < 3^n < \dots < n!$$

For example:  $\sqrt{n} = n^{\frac{1}{2}}$

# Growth rate of functions

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Listed from slowest to fastest growth:

- $1 \rightarrow$  Constant growth
- $\log n \rightarrow$  Logarithmic growth
- $n^c \rightarrow$  where  $0 < c < 1$
- $n \rightarrow$  Linear growth
- $n \log n$
- $n^2 \rightarrow$  Quadratic growth
- $n^2 \log n$
- $n^3 \rightarrow$  Cubic growth
- $n^c \rightarrow$  Polynomial growth ( $c$  is a constant number)
- $2^n \rightarrow$  Exponential growth
- $3^n \rightarrow$  Exponential growth
- $c^n \rightarrow$  Exponential growth ( $c$  is a constant number)
- $n! \rightarrow$  Factorial growth



Constant  $\prec \log n \prec n^c$  ( $0 < c < 1$ )  $\prec n \prec n \log n \prec n^2 \prec n^3 \dots \prec 2^n \prec 3^n \prec \dots \prec n!$

- 
- $T_1(n) = (1.5)^n$
  - $T_2(n) = 8n^3 + 17n^2 + 11$
  - $T_3(n) = \log(n)$
  - $T_4(n) = 2^n$
  - $T_5(n) = \log(\log(n))$
  - $T_6(n) = n^2 \log(n)$
  - $T_7(n) = 2^n(n^2 + 1)$
  - $T_8(n) = 100000$
  - $T_9(n) = n!$
  - $T_8(n) = 100000$
  - $T_5(n) = \log(\log(n))$
  - $T_3(n) = \log(n)$
  - $T_6(n) = n^2 \log(n)$
  - $T_2(n) = 8n^3 + 17n^2 + 11$
  - $T_1(n) = (1.5)^n$
  - $T_4(n) = 2^n$
  - $T_7(n) = 2^n(n^2 + 1)$
  - $T_9(n) = n!$

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## Audience Q&A

① Start presenting to display the audience questions on this slide.

# Summary

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**Today's lecture:** focused on sentinel and binary search algorithms and their time complexity. Introduced the growth of functions

- Sentinel search and Linear search algorithms are both linear in worst-case scenario, but Sentinel search requires executing fewer operations
- Binary search algorithm is more efficient comparing to Linear and Sentinel Search algorithms, but the input array should be sorted
- In most cases - the growth of time complexity with increasing input size 'n' is a suitable measure for the comparison of algorithms. Useful to know the order in which simple functions grow!

**Next Lecture:** Asymptotic analysis and Big-O notation