

CSE 332 Autumn 2023

Lecture 2: Algorithm Analysis and Priority Queues

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Warm Up

Give the worst case running time for the following code

```
doSomething(List myList){
    n = myList.size();
    x = 0;
    for (i=0; i < n; i++){
        for (j=0; j < i; j++){
            x++;
        }
    }
    return x;
}
```

Questions to ask:

- What are the units of the input size?
- What are the operations we're counting?
- For each line:
 - How many times will it run?
 - How long does it take to run?
 - Does this change with the input size?

Goals for Algorithm Analysis

- Identify a *function* which maps the algorithm's input size to a measure of resources used
 - Domain of the function: **sizes** of the input
 - Number of characters in a string, number of items in a list, number of pixels in an image
 - Codomain of the function: **counts** of resources used
 - Number of times the algorithm adds two numbers together, number times the algorithm does a $>$ or $<$ comparison, maximum number of bytes of memory the algorithm uses at any time
- Important note: Make sure you know the “units” of your domain and codomain!
 - Domain = inputs to the function
 - Codomain = outputs to the function

Worst Case Running Time Analysis

- If an algorithm has a worst case running time of $f(n)$
 - Among all possible size- n inputs, the “worst” one will do $f(n)$ “operations”
 - I.e. $f(n)$ gives the maximum operation count from among all inputs of size n

Worst Case Running Time Analysis

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Worst Case Running Time – General Guide

- Add together the time of consecutive statements
- Loops: Sum up the time required through each iteration of the loop
 - If each takes the same time, then [time per loop \times number of iterations]
- Conditionals: Sum together the time to check the condition and time of the slowest branch
- Function Calls: Time of the function's body
- Recursion: Solve a **recurrence relation**

Searching in a Sorted List

5	8	13	42	75	79	88	90	95	99
0	1	2	3	4	5	6	7	8	9

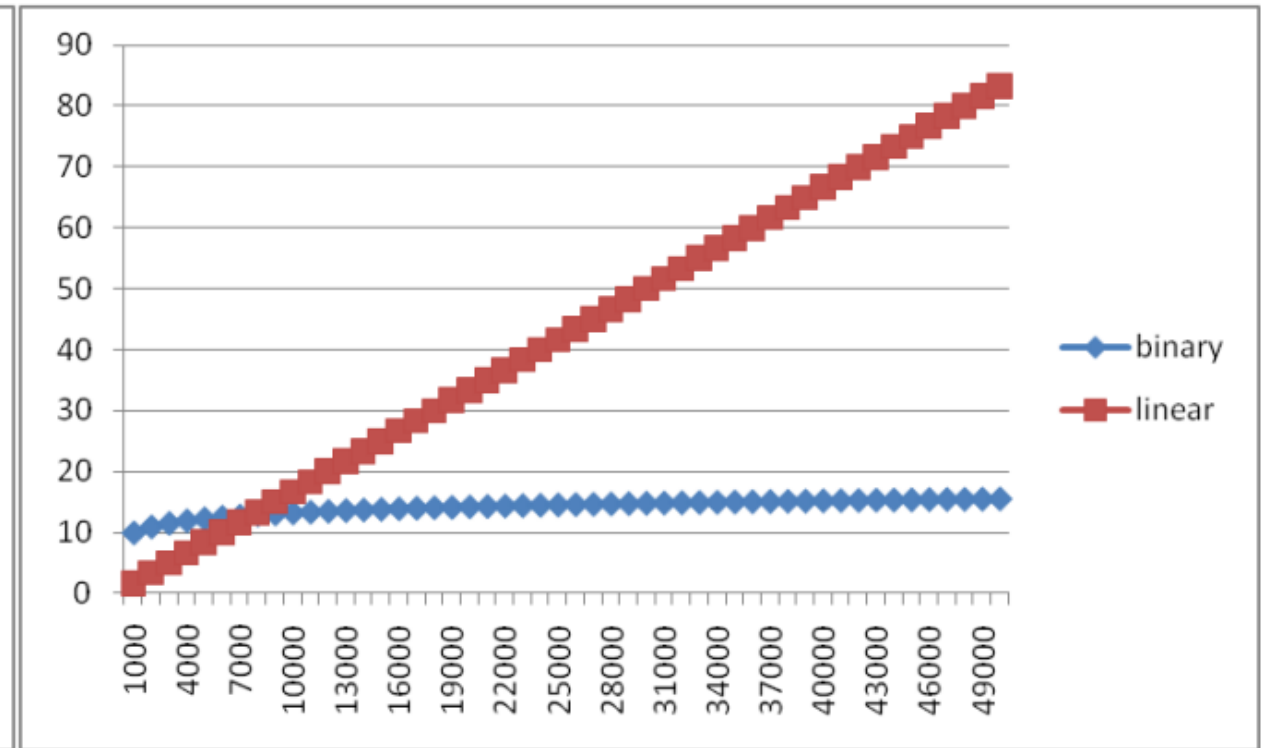
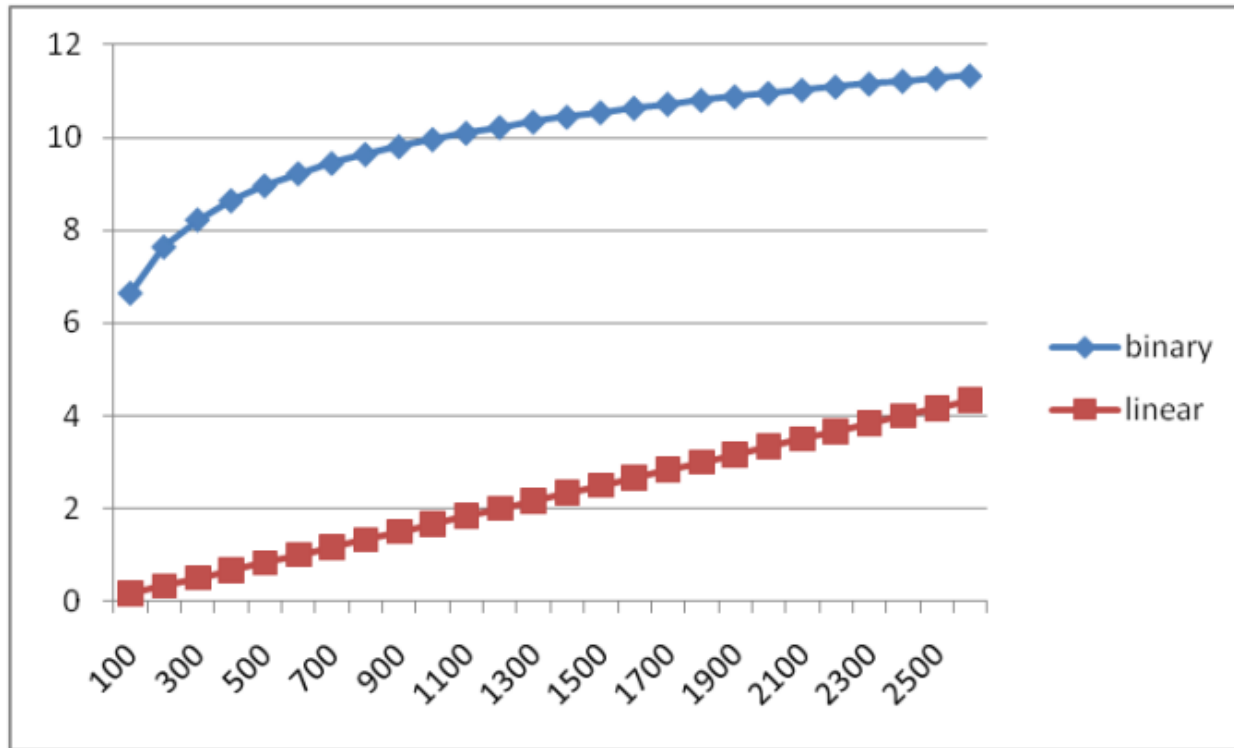
```
boolean linearSearch(array a, integer k){  
    for(i=0; i< a.length; i++){  
        if (a[i] == k){  
            return true;  
        }  
    }  
    return false;  
}
```

Faster way?

5	8	13	42	75	79	88	90	95	99
0	1	2	3	4	5	6	7	8	9

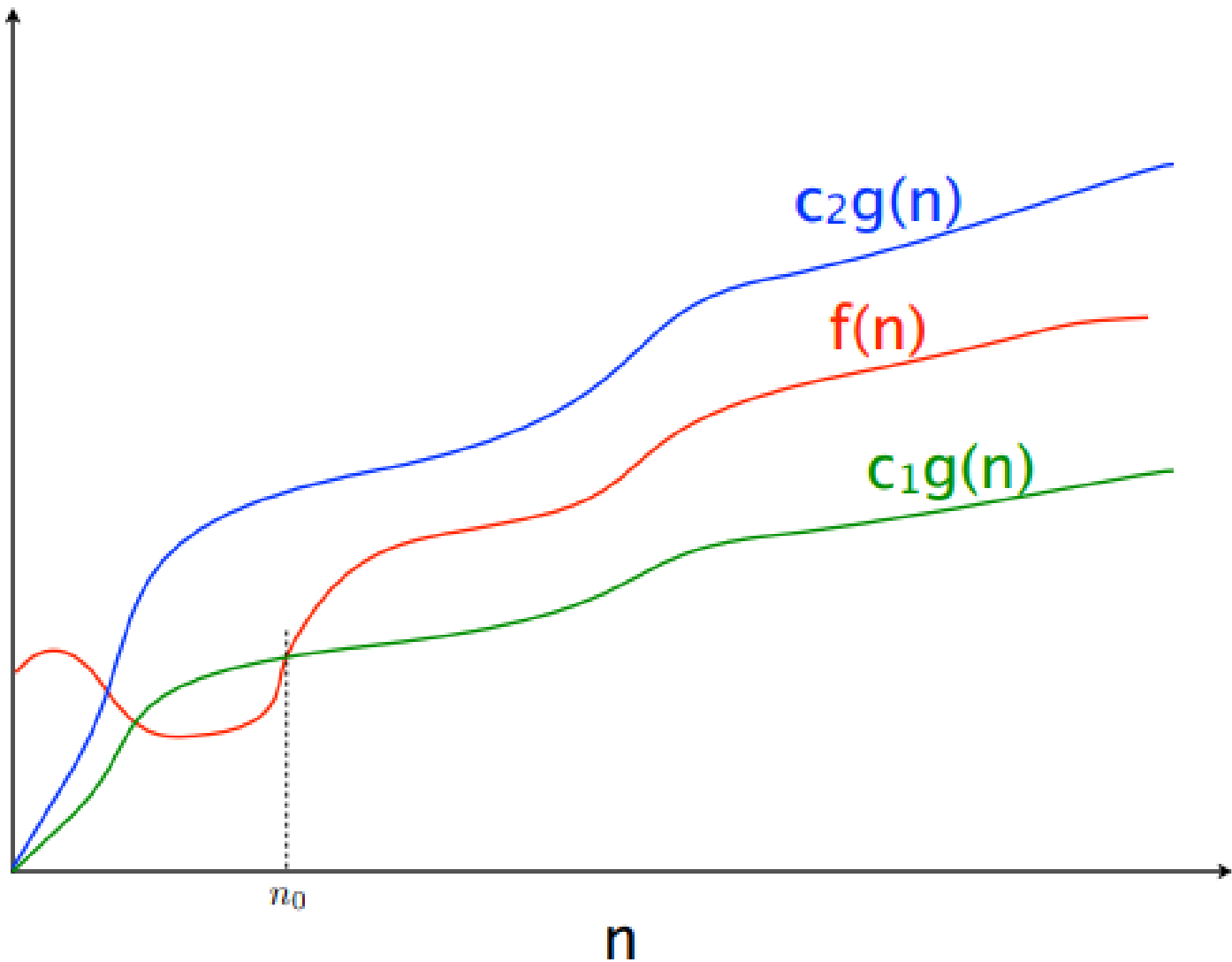
Can you think of a faster algorithm to solve this problem?

Comparing



Comparing Running Times

- Suppose I have these algorithms, all of which have the same input/output behavior:
 - Algorithm A's worst case running time is $10n + 900$
 - Algorithm B's worst case running time is $100n - 50$
 - Algorithm C's worst case running time is $\frac{n^2}{100}$
- Which algorithm is best?



$$f(n) = O(g(n))$$

$$f(n) = \Theta(g(n))$$

$$f(n) = \Omega(g(n))$$

Asymptotic Notation

- $O(g(n))$
 - The **set of functions** with asymptotic behavior less than or equal to $g(n)$
 - **Upper-bounded** by a constant times g for large enough values n
 - $f \in O(g(n)) \equiv \exists c > 0. \exists n_0 > 0. \forall n \geq n_0. f(n) \leq c \cdot g(n)$
- $\Omega(g(n))$
 - the **set of functions** with asymptotic behavior greater than or equal to $g(n)$
 - **Lower-bounded** by a constant times g for large enough values n
 - $f \in \Omega(g(n)) \equiv \exists c > 0. \exists n_0 > 0. \forall n \geq n_0. f(n) \geq c \cdot g(n)$
- $\Theta(g(n))$
 - “**Tightly**” within constant of g for large n
 - $\Omega(g(n)) \cap O(g(n))$

Asymptotic Notation Example

- Show: $10n + 100 \in O(n^2)$
 - **Technique:** find values $c > 0$ and $n_0 > 0$ such that $\forall n > n_0. 10n + 100 \leq c \cdot n^2$
 - **Proof:**

Asymptotic Notation Example

- Show: $10n + 100 \in O(n^2)$
 - **Technique:** find values $c > 0$ and $n_0 > 0$ such that $\forall n \geq n_0. 10n + 100 \leq c \cdot n^2$
 - **Proof:** Let $c = 10$ and $n_0 = 6$. Show $\forall n \geq 6. 10n + 100 \leq 10n^2$
 - $10n + 100 \leq 10n^2$
 - $\equiv n + 10 \leq n^2$
 - $\equiv 10 \leq n^2 - n$
 - $\equiv 10 \leq n(n - 1)$
 - This is True because $n(n - 1)$ is strictly increasing and $6(6 - 1) > 10$

Asymptotic Notation Example

- Show: $13n^2 - 50n \in \Omega(n^2)$
 - **Technique:** find values $c > 0$ and $n_0 > 0$ such that $\forall n \geq n_0. 13n^2 - 50n \geq c \cdot n^2$
 - **Proof:**

Asymptotic Notation Example

- Show: $13n^2 - 50n \in \Omega(n^2)$
 - **Technique:** find values $c > 0$ and $n_0 > 0$ such that $\forall n \geq n_0. 13n^2 - 50n \geq c \cdot n^2$
 - **Proof:** let $c = 12$ and $n_0 = 50$. Show $\forall n \geq 50. 13n^2 - 50n \geq 12n^2$
 - $13n^2 - 50n \geq 12n^2$
 - $\equiv n^2 - 50n \geq 0$
 - $\equiv n^2 \geq 50n$
 - $\equiv n \geq 50$
- This is certainly true $\forall n \geq 50$.

Asymptotic Notation Example

- Show: $n^2 \notin O(n)$

Asymptotic Notation Example

Proof by
Contradiction!

- To Show: $n^2 \notin O(n)$

- **Technique: Contradiction**

- **Proof:** Assume $n^2 \in O(n)$. Then $\exists c, n_0 > 0$ s. t. $\forall n > n_0, n^2 \leq cn$

Let us derive constant c . For all $n > n_0 > 0$, we know:

$$cn \geq n^2,$$

$$c \geq n.$$

Since c is lower bounded by n , c cannot be a constant and make this True.

Contradiction. Therefore $n^2 \notin O(n)$.

Gaining Intuition

- When doing asymptotic analysis of functions:
 - If multiple expressions are added together, ignore all but the “biggest”
 - If $f(n)$ grows asymptotically faster than $g(n)$, then $f(n) + g(n) \in \Theta(f(n))$
 - Ignore all multiplicative constants
 - $f(n) + c \in \Theta(f(n))$ for any constant $c \in \mathbb{R}$
 - Ignore bases of logarithms
 - Do NOT ignore:
 - Non-multiplicative and non-additive constants (e.g. in exponents, bases of exponents)
 - Logarithms themselves
- Examples:
 - $4n + 5$
 - $0.5n \log n + 2n + 7$
 - $n^3 + 2^n + 3n$
 - $n \log(10n^2)$

More Examples

- Is each of the following True or False?
 - $4 + 3n \in O(n)$
 - $n + 2 \log n \in O(\log n)$
 - $\log n + 2 \in O(1)$
 - $n^{50} \in O(1.1^n)$
 - $3^n \in \Theta(2^n)$

Common Categories

- $O(1)$ “constant”
- $O(\log n)$ “logarithmic”
- $O(n)$ “linear”
- $O(n \log n)$ “log-linear”
- $O(n^2)$ “quadratic”
- $O(n^3)$ “cubic”
- $O(n^k)$ “polynomial”
- $O(k^n)$ “exponential”

Defining your running time function

- Worst-case complexity:
 - max number of steps algorithm takes on “most challenging” input
- Best-case complexity:
 - min number of steps algorithm takes on “easiest” input
- Average/expected complexity:
 - avg number of steps algorithm takes on random inputs (context-dependent)
- Amortized complexity:
 - max total number of steps algorithm takes on M “most challenging” consecutive inputs, divided by M (i.e., divide the max total sum by M).

ADT: Queue

- What is it?
 - A “First In First Out” (FIFO) collection of items
- What Operations do we need?
 - Enqueue
 - Add a new item to the queue
 - Dequeue
 - Remove the “oldest” item from the queue
 - Is_empty
 - Indicate whether or not there are items still on the queue

ADT: Priority Queue

- What is it?
 - A collection of items and their “priorities”
 - Allows quick access/removal to the “top priority” thing
- What Operations do we need?
 - insert(item, priority)
 - Add a new item to the PQ with indicated priority
 - Usually, smaller priority value means more important
 - deleteMin
 - Remove and return the “top priority” item from the queue
 - Is_empty
 - Indicate whether or not there are items still on the queue
- Note: the “priority” value can be any type/class so long as it’s comparable (i.e. you can use “<” or “compareTo” with it)

Priority Queue, example

```
PriorityQueue PQ = new PriorityQueue();
```

```
PQ.insert(5,5)
```

```
PQ.insert(6,6)
```

```
PQ.insert(1,1)
```

```
PQ.insert(3,3)
```

```
PQ.insert(8,8)
```

```
Print(PQ.deleteMin)
```

```
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PQ.insert(1,1)
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```
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```

```
PQ.insert(3,3)
```

```
Print(PQ.deleteMin)
```

```
Print(PQ.deleteMin)
```

```
PQ.insert(8,8)
```

```
Print(PQ.deleteMin)
```

```
Print(PQ.deleteMin)
```

Applications?

Thinking through implementations

Data Structure	Worst case time to insert	Worst case time to deleteMin
Unsorted Array		
Unsorted Linked List		
Sorted Circular Array		
Sorted Linked List		
Binary Search Tree		

Note: Assume we know the maximum size of the PQ in advance