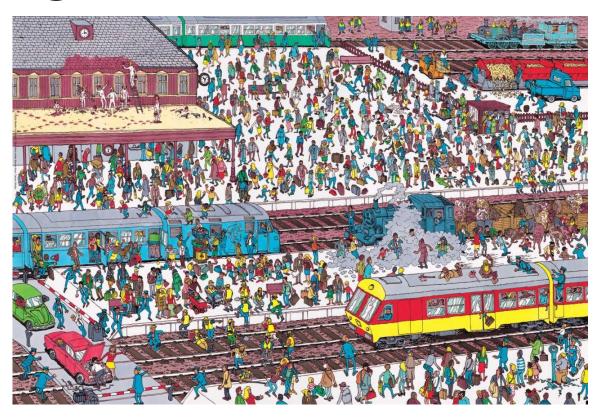
Searching, Sorting, and Algorithm Analysis

Chapter 9

Searching

Searching





Often called sequential searching

Often called sequential searching

The list is searched one element at a time in order

Often called sequential searching

- The list is searched one element at a time in order
 - Example: Line up every character in the Where's Waldo picture and check if each one is Waldo
 - Always remember to stop searching when the result is found. i.e. break, return, ...

How efficient is a linear search?

- How efficient is a linear search?
 - Where 'N' represents the number of elements,
 the algorithm may have to search all 'N' elements
 - What if the list is an ordered array of numbers 1 through 999 billion and we are searching for 999000000000?

 On average the search will find an item in N/2 attempts

Binary Search

More efficient than linear searching

- Requirements
 - The array must be sorted

Where x is the desired item

- 1. Check the middle item, y
 - 1.1. If x == y, then return
 - 1.2. if x < y, go to step 1 with the smaller half of the array
 - 1.3. if x > y, go to step 1 with the larger half of the array

Search for x = 6

2 3 4 5 6 7

Search for x = 6

y = 4

1	2	3	4	5	6	7

Search for x = 6

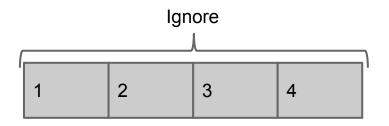
y = 4

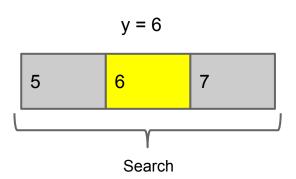
1	2	3	4	5	6	7	

Problem: 6 != 4, continue

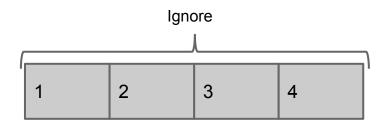
6 > 4, so choose the larger half to search

Search for x = 6

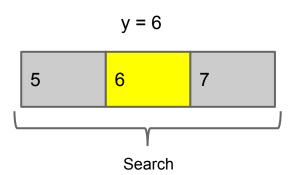




Search for x = 6



Problem: 6 == 6, return



 Is the Binary Search more efficient than the Linear search?

- Every iteration of the Binary Search
 Algorithm eliminates half of the list
 - List of 20,000 items
 - Linear: 20,000 possible comparisons
 - Binary search:

- Every iteration of the Binary Search
 Algorithm eliminates half of the list
 - List of 20,000 items
 - Linear: 20,000 possible comparisons
 - Binary search: 15 possible comparisons

- Every iteration of the Binary Search
 Algorithm eliminates half of the list
 - List of 20,000 items
 - Linear: 20,000 possible comparisons
 - Binary search: 15 possible comparisons
 - Thats a power of 2, ie Log2(20000) or log N

Searching

 Are there uses for both Linear and Binary searching?

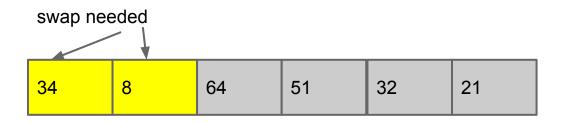
Searching

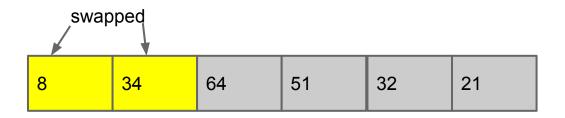
 Are there uses for both Linear and Binary searching?

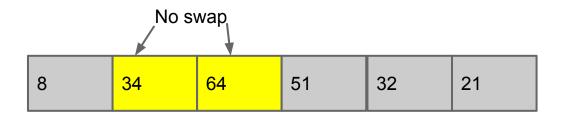
 What has to happen to take advantage of the binary search?

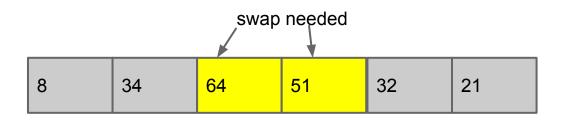
Sorting

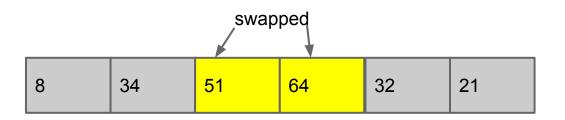
- Loop over the data N 1 times
 - o if current item is bigger than next item
 - swap

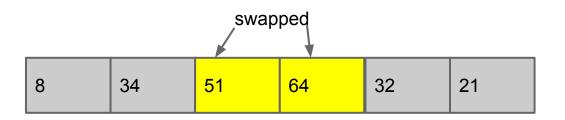


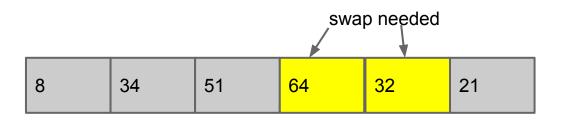




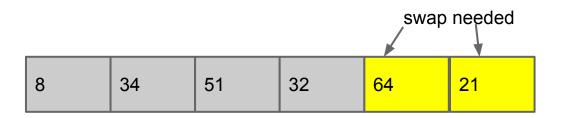


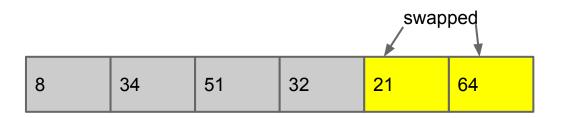




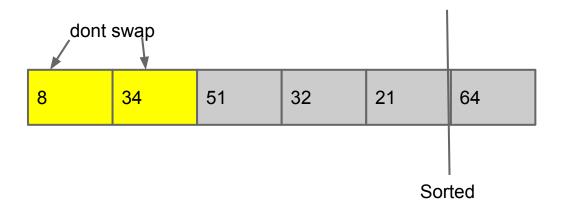


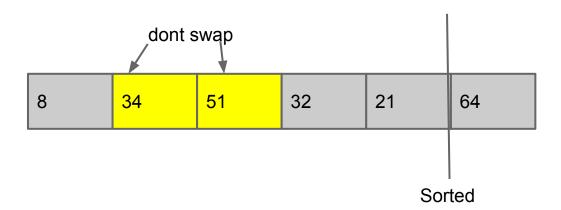


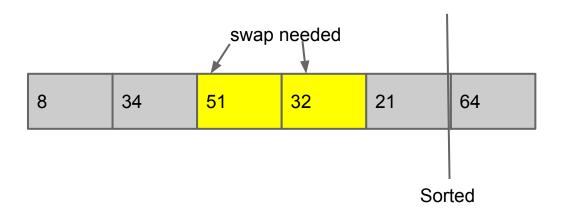


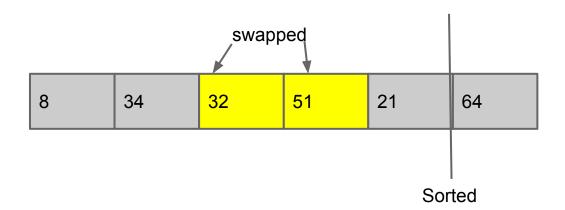


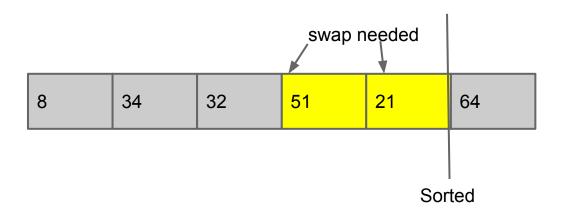
What does the first pass guarantee?

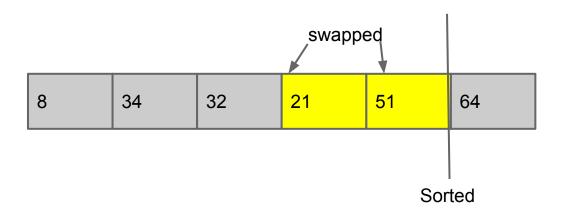


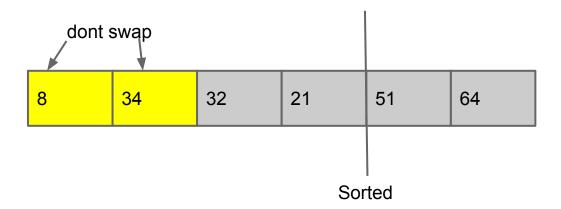


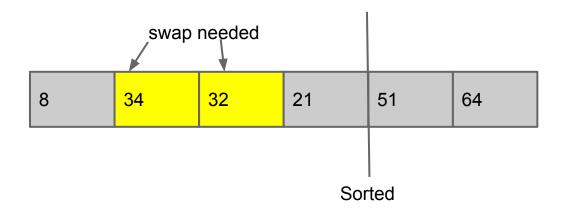


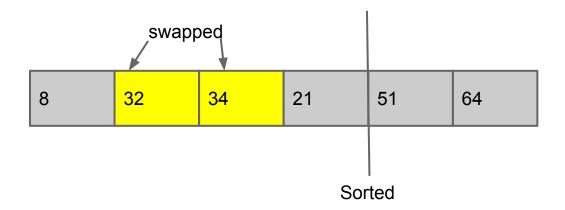


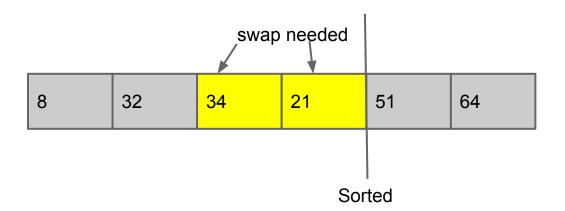


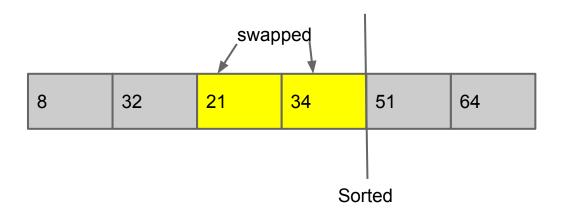


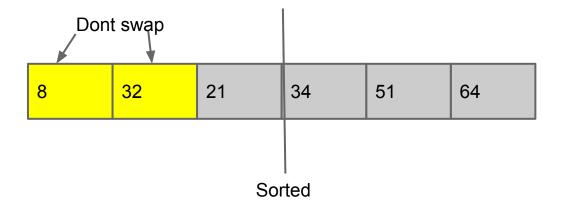


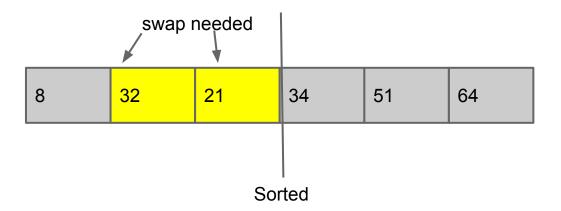


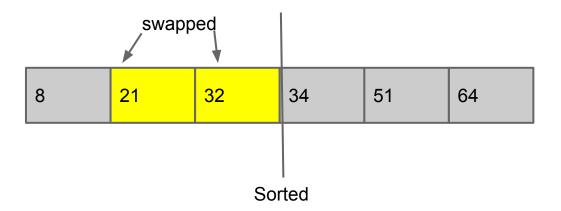


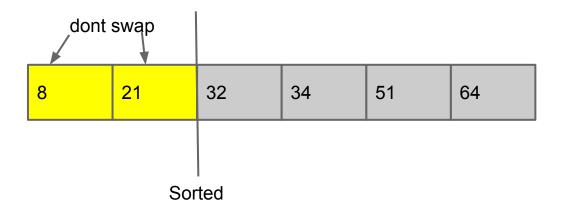


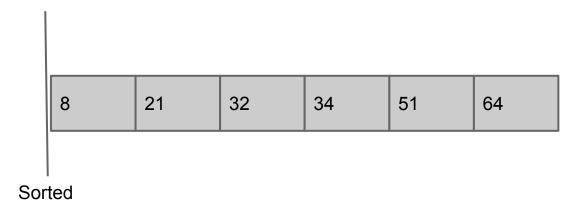












What is the complexity

- What is the complexity
 - o Best N
 - Worst N²

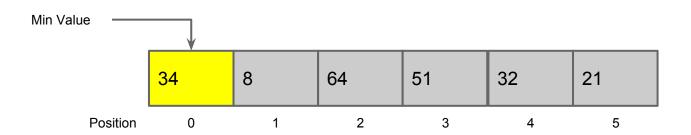
 Is this a good algorithm for very large sets of data?

http://www.youtube.com/watch?v=lyZQPjUT5B4&feature=relmfu

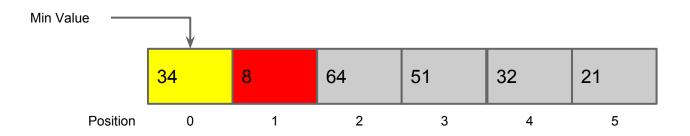
 The selection sort tries to minimize the number of swaps

- For every position in the list, 0 through N-1 as 'i'
 - loop over the list and find the smallest item
 - If item is the smallest, swap it into the current location (i) and continue outer loop

Start at the beginning of the list. Start with i as index 0 and minValue as 34

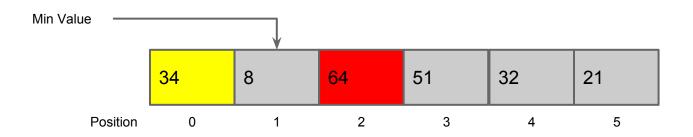


Loop over list to find the smallest item and place it at position 0



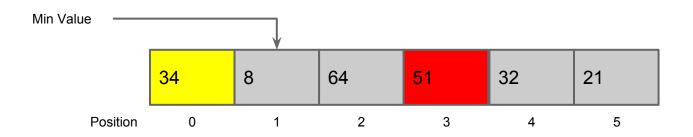
Is 34 < 8?
No, correct MinValue to 8

Loop over list to find the smallest item and place it at position 0



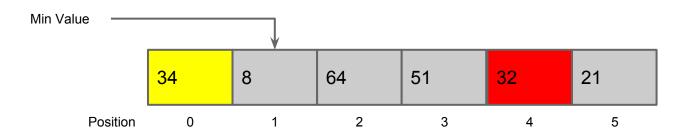
Is 8 < 64? Yes, continue

Loop over list to find the smallest item and place it at position 0



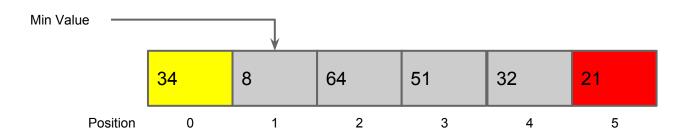
Is 8 < 51? Yes, continue

Loop over list to find the smallest item and place it at position 0



Is 8 < 32? Yes, continue

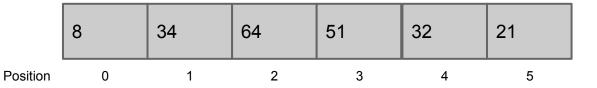
Loop over list to find the smallest item and place it at position 0



Is 8 < 21?
Yes. Since we are at the end of the list, 8 must be the min value. Swap position 0 with position 1

Loop over list to find the smallest item and place it at position 0

Min Value

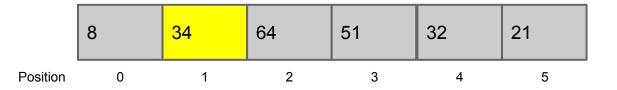


What is guaranteed after the first pass?

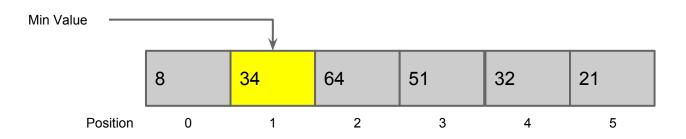
What is guaranteed after any single pass?

Loop over list to find the next smallest item and place it at position 1

Min Value

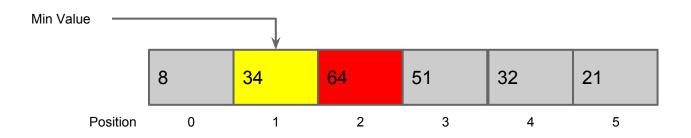


Loop over list to find the next smallest item and place it at position 1



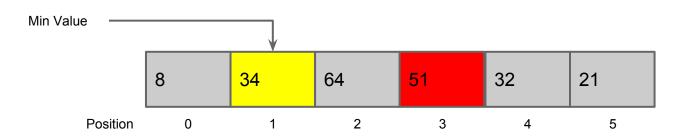
Set Min Value to 34 and begin looping again

Loop over list to find the next smallest item and place it at position 1



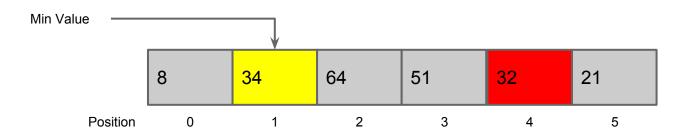
is 34 < 64? Yes, continue

Loop over list to find the next smallest item and place it at position 1



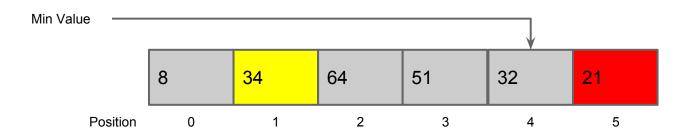
is 34 < 51? Yes, continue

Loop over list to find the next smallest item and place it at position 1



is 34 < 32? No, set Min Value to 32

Loop over list to find the next smallest item and place it at position 1



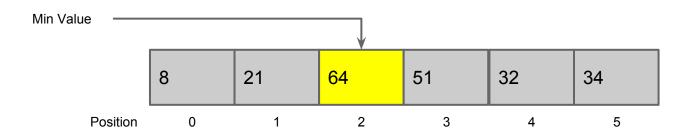
is 32 < 21?

No, set Min Value to 21. There are no more elements to search, so swap 21 into position 1.

Loop over list to find the next smallest item and place it at position 1

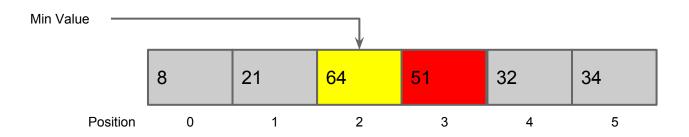
Min Value		Sorted				
	8	21	64	51	32	34
Position	0	1	2	3	4	5

Loop over list to find the next smallest item and place it at position 2



Set Min Value to 64.

Loop over list to find the next smallest item and place it at position 2



is 64 < 51? No, set Min Value to 51.

How many swaps happen per main loop?

How many swaps happen per main loop?

- How many comparisons are performed?
 - \circ N^2

Analysis

 How can we compare one algorithm against another?

- How can we compare one algorithm against another?
 - Define the computational problem
 - Look at the requirements
 - Space
 - Time

Space-Time...



- Space
 - The amount of memory

- Time
 - Length of execution

How could we analyse an algorithm?

- How could we analyse an algorithm?
 - Run the algorithm and time it

Problems?

- How could we analyse an algorithm?
 - Run the algorithm and time it

- Problems?
 - Results depend on...
 - Hardware
 - Data sets
 - OS
 - **....**

- Break the algorithms up into 'steps'
 - A 'step' is a basic operation
 - Execution is time bounded by some constant regardless of input size
 - Swap vs find

- Define a generic size of the array or data set
 - Usually 'N'

- Constants are ignored
 - One machine may execute the operation in one step, yet another takes multiple

- Measuring Complexity
 - Determine the number of steps required to run the algorithm

- Measuring Complexity
 - Determine the number of steps required to run the algorithm

Running Time Calculations: The Rules

- 1. FOR loops
 - 1. running time of the statements multiplied by the size of the loop
- 2. Nested loops
 - 1. running time of the statements multiplied by the product of the sizes of all loops
- 3. Consecutive Statements
 - 1. add them
- 4. If/Else
 - 1. never more than the running time of the test plus the larger of the running times

```
int sum(int n)
  int partialSum;
  partialSum = 0;
  for (int i = 1; i \le n; i++)
     partialSum += i * i * i;
  return partialSum;
```

```
int sum( int n)
  int partialSum;
                                   //0
  partialSum = 0;
                                   //1
  for (int i = 1; i \le n; i++)
     partialSum += i * i * i;
  return partialSum;
                                   //1
```

```
int sum( int n)
  int partialSum;
                                   //0
  partialSum = 0;
                                  //1
  for (int i = 1; i \le n; i++)
     partialSum += i * i * i;  //1 + 1 + 1 + 1 = 4
  return partialSum;
                                  //1
```

```
int sum( int n)
  int partialSum;
                                 //0
  partialSum = 0;
                                 //1
  for (int i = 1; i \le n; i++)
                                //1 (init i) + N+1(compare) +
                                 //N (for i++) = 2N+2
                            //1 + 1 + 1 + 1 = 4N
     partialSum += i * i * i;
  return partialSum;
                                //1
```

Total: 6N + 4

```
int sum( int n)
  int partialSum;
                                 //0
  partialSum = 0;
                                //1
                                //1 (init i) + N+1(compare) +
  for (int i = 1; i \le n; i++)
                                 //N (for i++) = 2N+2
                            //1 + 1 + 1 + 1 = 4N
     partialSum += i * i * i;
  return partialSum;
                                //1
```

 Since inner sum is constant despite the size of the array, we can count it as a single step

```
int sum( int n)
  int partialSum;
                                  //0
  partialSum = 0;
                                 //1
  for (int i = 1; i <= n; i++)
                                 //1 (init i) + N+1(compare) +
                                  //N (for i++) = 2N+2
     partialSum += i * i * i;
                                 //1
  return partialSum;
                                 //1
```

Total: 3N + 4 // a little simpler

- To compare algorithms, we must compare apples to apples. Thus, functions are usually found for:
 - Worse-Case Complexity
 - The number of steps performed on an input that requires the most steps
 - Average-Case Complexity
 - The number of steps performed on an input with an average number of steps (not the best indicator)
 - Best-Case Complexity
 - The number of steps performed on an input with the least number of steps (not the best indicator either)

- The worst case can be shown as a function of f(n)
 - o f(n) = 6N + 4 // Previous algorithm

- The worst case can be shown as a function of f(n) or g(n) or ...
 - o f(n) = 6N + 4 // Previous algorithm
- The worst case complexity of two algorithms can be compared by looking at the ration of f(n)/g(n) when n gets large

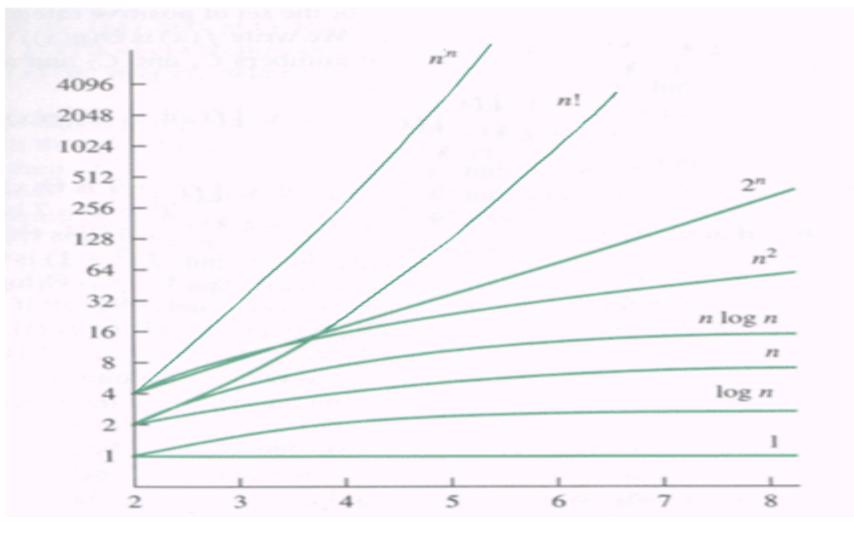
- Examples
 - \circ f(n) = n
 - \circ g(n) = n²

$$\frac{f(n)}{g(n)} = \frac{N}{N^2}$$

- Examples
 - \circ f(n) = n
 - \circ g(n) = n²

$$\frac{f(n)}{g(n)} = \frac{N}{N^2} = \frac{1}{N}$$

g(n) grows a lot faster than f(n)



- What does this mean for us?
 - We really only care about the growth rate of algorithms
 - Welcome to Big O notation!

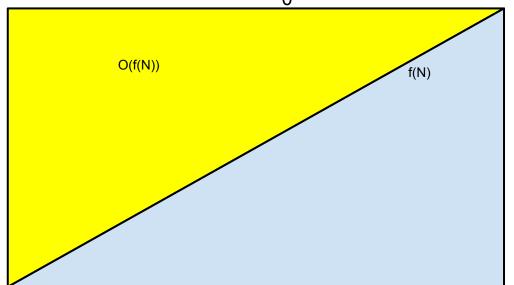
Algorithm Analysis: Big-Oh

- \bullet T(N) = O(f(N))
 - o if there are positive constants c and n_0 such that $T(N) \le cf(N)$ when $N \ge n_0$

What does that mean?

Algorithm Analysis: Big-Oh

- \bullet T(N) = O(f(N))
 - o if there are positive constants c and n_0 such that $T(N) \le cf(N)$ when $N \ge n_0$



We only need to make sure that the true complexity function is above the Big O (With the biggest Big O possible)

Algorithm Analysis: Big-Oh

- T(N) = O(f(N))
 if there are positive constants c and n₀ such that
 T(N) <= cf(N) when N >= n₀
- The growth rate of T(N) is <= the growth of f(N)

- T(N) = 1000N○ f(N) = ?
- $T(N) = 100N^2$
- \bullet T(N) = 100N² + 3N + 300000000
- Would it be correct to say that f(N) = 2^N if T(N) = 3N?

```
void foo(int* paMyArray, int length)
  int sum;
  for (int i = 0; i < 7; i++)
     sum += paMyArray[i];
```

```
void foo(int* paMyArray, int length)
  int sum;
  for (int i = 0; i < 7; i++)
     sum += paMyArray[i];
```

Total: O(1)

```
void bar(int* paMyArray, int len)
  for (int i = 0; i < len; i++)
     paMyArray[i] = 0;
  for (int i = 0; i < len; i++)
     for (int j = 0; j < len; j++)
        paMyArray[i] = i + j;
```

```
void bar(int* paMyArray, int len)
  for (int i = 0; i < len; i++)
     paMyArray[i] = 0;
                                           Total: O(N^2)
  for (int i = 0; i < len; i++)
     for (int j = 0; j < len; j++)
        paMyArray[i] = i + j;
```

```
int binarySearch(int * paArray, int len, const int x)
  int low = 0, high = len - 1;
  while (low <= high)
     int mid = (low + high) / 2;
     if (paArray[mid] < x)
       low = mid + 1:
     else if (paArray[mid] > x)
       high = mid - 1;
     else
       return mid;
  return -1;
```

```
int binarySearch(int * paArray, int len, const int x)
  int low = 0, high = len - 1;
  while (low <= high)
     int mid = (low + high) / 2;
     if (paArray[mid] < x)
        low = mid + 1:
     else if (paArray[mid] > x)
       high = mid - 1;
     else
        return mid;
  return -1;
```

Total: O(log N)