

LECTURE 4

Which g(n)- Inherent imprecision of the big-O

The big-O notation is **inherently imprecise** as there can be infinitely many functions g for a **given function** f.

- For example, the f(n)=2 n^2+3n+1 is big-O not only of n^2 , but also of n^3, \ldots, n^k, \ldots for any $k \ge 2$.
- To avoid this embarrassment of riches, the smallest function g is chosen, n^2 in this case.

Real World Example:

- If a child, ask what comes after class 1
 - Answer could be class 2nd, 3rd, 4th, 5th, ... but one should say 2nd



EXAMPLES BASIC LOOP ORDERS

Simple Loop Orders

Example 0

Loop will run approximately **n** times ... O(n)

Example 1

```
for (i=0;i< n;i=i+k)
```

Loop will run approximately **n/k** times ... O(n)

Example 2

```
for (i=n;i>0;i=i-k)
```

Loop will run approximately **n/k** times ... O(n)

Example 3

Loop will run approximately $\log_k n$ times ... $O(\log_k n)$

Example

```
for(i=0;i<n; i++)
for (j=i; j<n;++j)
```

Nested loop approximately run n(n+1)/2 times. $O(n^2)$

Example

```
for (i=0;i<n;++i)
for (j=0;j<m;++j)
```

Nested loop approximately run **n*m** times. ...

 $O(n^2)$ given $n \ge m$

Example 5

for(
$$i=1$$
; $i <= n$; $++i$)
for ($j=1$; $j <= i$; $++j$)

$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2} = O(n^2)$$

Nested loop approximately runs n(n+1)/2 times. (Arithmetic Series) $O(n^2)$

Example

for(
$$i = 1$$
; $i < n$; $++i$)
for($j = 1$; $j < n * n$; $++j$)

Approximately runs n^3 times. $O(n^3)$

	No of times loop runs
for(i = 1; i <= n; ++i)	$\sum_{i=1}^{n} 1 = n$
for($j = 1; j < i * i; ++j$)	$\sum_{i=1}^{n} i^2$

O(n³) ... Arithmetic Series

```
for(i=1; i<=n; i=i*2)
    for (j=1;j<=i;++j)
        sum+=1;</pre>
```

Nested loop run appr.
2n-1 times.

- Outer loop runs lgn times
- Inner loop runs 1 2 4 8 16 32 64 ... times
- We need to sum up 1+2+4+8+16+32+64
- This forms a **Geometric series** sum up to lgn
- $1+2^1+2^2+2^3+2^4+2^5+2^6 \cdots 2^{\lg n}$

$$\sum_{k=0}^{n} x^{k} = 1 + x + x^{2} + \dots + x^{n} = \frac{x^{n+1} - 1}{x - 1} (x \neq 1)$$

$$\sum_{i=1}^{lgn} 2^{x} = 1 + 2^{i} + 2^{2} + 2^{3} + \dots 2^{lgn} = \frac{2^{lgn+1} - 1}{2 - 1} = O(n)$$

$$\frac{2^{lg_2^n}-1}{2-1} = n-1$$

Example

for(i=1;i<=n; i=i*2)
for (j=1;j<=
$$\bf{n}$$
;++ \bf{j})

Outer loop runs O(lgn) times Inner loop runs n times for each i ... n+n+n+...+n ... lgn times Nested loop approximately run O(nlgn) times.

Outer loop runs n time and inner lg1+lg2+lg3+...lgn times ..this is arithmetic series of lg

$$\sum_{k=1}^{n} \lg k = n \lg_2 n$$

Nested Loop approx. runs O(nlgn) times.

Example

```
for(i=1;i<=n; i*2)
for (j=1;j<=n; j*2)
```

Nested Loop approximately runs $O(\lg_2 n)^2$ times. Outer loop runs $\lg n$ time and inner $\lg n$ for each i

Linear Search

```
int LinearSearch(const int a[], int key, int n) {
    for (int i = 0; i < n && a[i]!= key; i++);
    if (i == n)
        return -1;
    return i;
}</pre>
```

Unsuccessful Search: \rightarrow O(n)

Successful Search:

Best-Case: *item* is in the first location of the array \rightarrow O(1)

Worst-Case: *item* is in the last location of the array \rightarrow O(n)

Average-Case: The number of key comparisons 1, 2, ..., n

Types of Analysis

Worst case

- Provides an upper bound on running time (maximum number of steps)
- An absolute guarantee that the algorithm would not run longer, no matter what the inputs are

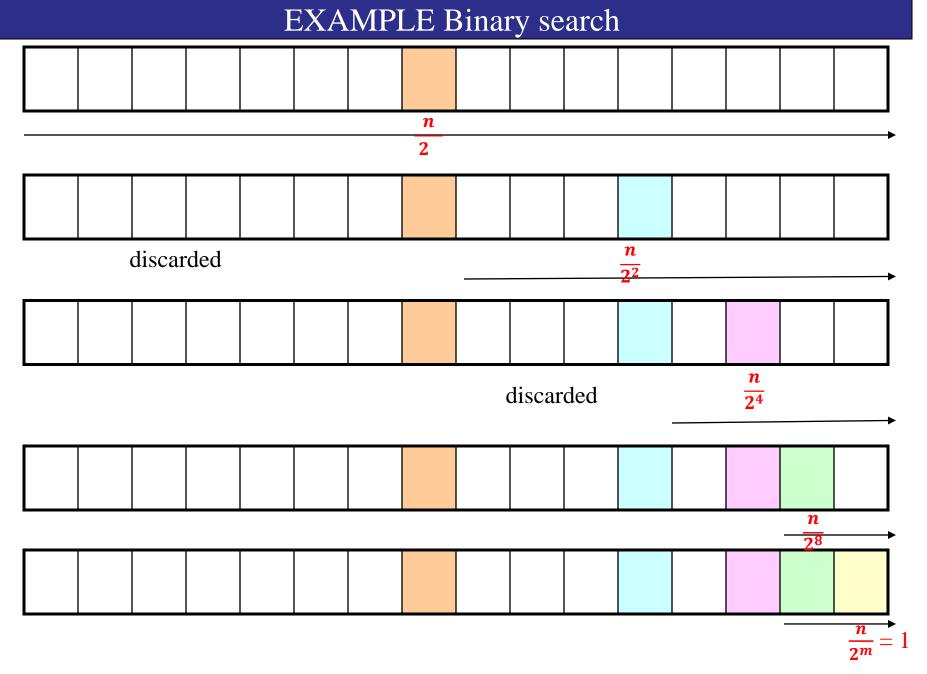
Best case

- Provides a lower bound on running time (number of steps is the smallest)
- Input is the one for which the algorithm runs the fastest

$Lower\ Bound \le Running\ Time \le Upper\ Bound$

Average case

- Provides a prediction about the running time
- Assumes that the input is random

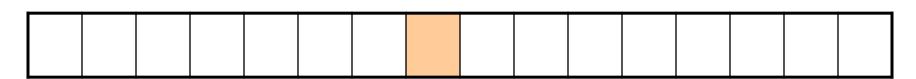


Binary Search – Analysis

- For an unsuccessful search:
 - The number of iterations in the loop is $\lfloor \log_2 n \rfloor + 1$
 - \rightarrow O(log₂n)
- For a successful search:
 - **Best-Case:** The number of iterations is 1.

- \rightarrow O(1)
- *Worst-Case:* The number of iterations is $\lfloor \log_2 n \rfloor + 1$ \rightarrow $O(\log_2 n)$
- Average-Case: The avg. # of iterations $< \log_2 n$
 - \rightarrow O(log₂n)
- 1 2 3 4 5 6 7 \leftarrow an array with size 8
- 3 2 3 1 3 2 3 4 **←** # of iterations

The average # of iterations = $21/8 < \log_2 8$



BUBBLE SORT

```
void BubbleSort(int arr[], int n) {
   bool done = false;
   for(i = 1; i < n; i++) //repeat a pass of bubble sort
      for (j=0; j < n-i; j++)//inner loop swaps consecutive items
         if (arr[j+1] < arr[j]){
            swap(arr[j+1],arr[j])
```

BUBBLE SORT

```
void BubbleSort(int arr[], int n) {
    bool done = false;
   for(i = 1; (i < n) \&\& !done; i++){//repeat a pass of bubble sort}
       done = true;
      for (j=0; j < n-i; j++){ //inner loop swaps consecutive items
          if (arr[j+1] < arr[j])
              swap(arr[j+1],arr[j])
              done = false; //a swap is made and so sorting continues
                                         Best case O(n)
                                         Worst case O(n<sup>2</sup>)
```

SELECTION SORT

```
void SelectionSort(int arr[], int n) {
  for (i=0;i<n;++i){
    maxIndex = FindMaxIndex(arr,i,n-1);
    swap(arr[i],arr[maxIndex]);
}</pre>
```

```
//finds the maximum item in the partial array startIndex to endIndex
int FindMaxIndex(int arr[], int startIndex, int endIndex){
    int max = startIndex
    for (i= startIndex; i<= endIndex;++i)
        if (a[max] < arr[i]) max =i;
    return max;
}</pre>
```

SPACE COMPLEXITY

- Space complexity is the amount of memory a program needs to run to completion
 - If program uses array of size n → O(n) Space
 - IF program uses 2D array of size $n*n -> O(n^2)$ Space

• Time complexity is the amount of computer time a program needs to run to completion

• <u>...\2. Lists.pptx</u>



Practice Questions

Euclids GCD

```
• 1 long gcd( long m, long n )
   -2 {
   -3 while (n!=0)
   -4 {
       • 5 long rem = m % n;
       • 6 \text{ m} = \text{n};
       • 7 \text{ n} = \text{rem};
   -8
   -9 return m;
```

• 10 }

EXAMPLES

- Summing an array of size n: O(n)
- Printing a matrix of size nxn: $O(n^2)$
- Summing two matrices of size nxn: $O(n^2)$
- Product of two matrices of size nxn: $O(n^3)$
- Linear search in array of size n: O(n)
- Binary search in array of size n: O(log n)
- Printing all numbers that can be represented by n bits: $O(2^n)$
- Printing all subsets of numbers in an array of size n: O(2ⁿ)
- Printing all permutations of numbers in array of size n: O(n!)