

## Lecture 2: Structure of Computation

### Review

- Problem, Algorithm, Efficiency, Model of Computation, Data Structure
- How to Solve an Algorithms Problem
  - Reduce to a problem you already know (use data structure or algorithm)
  - Design your own (recursive) algorithm

| Class                | Graph | Visited |
|----------------------|-------|---------|
| Brute Force          | Star  | All     |
| Decrease & Conquer   | Chain | All     |
| Divide & Conquer     | Tree  | All     |
| Dynamic Programming  | DAG   | All     |
| Greedy / Incremental | DAG   | Some    |

### Contains

- **Problem:** Given an array  $A$  of  $n$  integers, does it contain integer  $v$ ?
- **Example:** Seven element array.

#### Brute Force $O(n)$

```

1 def contains(A, v):
2     for i in range(len(A)):
3         if A[i] == v: return True
4     return False

```

#### Decrease & Conquer $O(n)$

```

1 def contains(A, v, i = 0):
2     if i == len(A): return False
3     if A[i] == v: return True
4     return contains(A, v, i + 1)

```

#### Divide & Conquer $O(n)$

```

1 def contains(A, v, i = 0, j = None):
2     if j is None: j = len(A)
3     if j - i == 0: return False
4     c = (i + j) // 2
5     if A[c] == v: return True
6     left = contains(A, v, i, c)
7     right = contains(A, v, c + 1, j)
8     return left or right

```

#### Greedy, if sorted $A$ , $O(\log n)$

```

1 def contains(A, v, i = 0, j = None):
2     if j is None: j = len(A)
3     if j - i == 0: return False
4     c = (i + j) // 2
5     if A[c] == v: return True
6     if A[c] > v:
7         return contains(A, v, i, c)
8     return contains(A, v, c + 1, j)

```

## Max Subarray Sum

- **Problem:** Given an array  $A$  of  $n$  integers, what is the largest sum of any nonempty subarray? (in this class, **subarray** always means a contiguous sequence of elements)

- **Example:**  $A = [-9, 1, -5, 4, 3, -6, 7, 8, -2]$ , largest sum is 16.

- Brute Force:

- # subarrays:  $\binom{n}{2} + \binom{n}{1} = O(n^2)$
- Can compute subarray sum of  $k$  elements in  $O(k)$  time
- $n$  subarrays have 1 element,  $n - 1$  have 2, ... , 1 has  $n$  elements
- Work is  $c \sum_{k=1}^n (n - k + 1)k = cn(n + 1)(n + 2)/6 = O(n^3)$
- Graph: single node, or quadratic branching star, each with linear work

|   |  |
|---|--|
| <pre> 1 def MSS(A): 2     m = A[0] 3     for j in range(1, len(A) + 1): 4         for i in range(0, j): 5             m = max(m, SS(A, i, j)) 6     return m </pre> | <pre> 1 def SS(A, i, j): 2     s = 0 3     for k in range(i, j): 4         s += A[k] 5     return s 6 . </pre> |
|---|--|

- Divide & Conquer

- Max subarray is either:
  1. fully in left half,
  2. fully in right half,
  3. or contains elements from both halves.
- Third case = (MSS\_EA, ending at middle) + (MSS\_SA, starting at middle)
- Combine step takes linear time
- Graph is binary tree with linear work at each vertex
- $T(n) = 2T(n/2) + O(n) \implies T(n) = O(n \log n)$
- (Draw tree, Master Theorem will also be discussed in recitation)

|  |   |   |
|--|---|---|
| <pre> 1 def MSS(A, i = 0, j = None): 2     if j is None: j = len(A) 3     if j - i == 1: return A[i] 4     c = (i + j) // 2 5     return max(MSS(A, i, c), MSS(A, c, j), 6               MSS_EA(A, i, c) + MSS_SA(A, c, j)) </pre> | <pre> 1 def MSS_SA(A, i, j): 2     s = m = A[i] 3     for k in range(1, j - i): 4         s += A[i + k] 5         m = max(s, m) 6     return m </pre> | <pre> 1 def MSS_EA(A, i, j): 2     s = m = A[j - 1] 3     for k in range(1, j - i): 4         s += A[j - 1 - k] 5         m = max(s, m) 6     return m </pre> |
|--|---|---|

- Dynamic Programming

- $\text{MSS\_EA}(A, 0, k)$  finds largest subarray in  $A$  ending at  $k$
- MSS must end somewhere, so check  $\text{MSS\_EA}(A, 0, k)$  for all  $k$ . (Brute Force)

```

1 def MSS(A):
2     m = A[0]
3     for k in range(len(A)):
4         s = MSS_EA(A, 0, k + 1)
5         m = max(m, s)
6     return m

```

- But takes  $c \sum_{k=1}^n k = cn(n+1)/2 = O(n^2)$  time.
- Computing a lot of subarray sums; can we reuse any work?
- Let's rewrite  $\text{MSS\_EA}$  recursively

```

1 def MSS_EA(A, i, j):
2     if j - i == 1: return A[i]
3     return A[j - 1] + max(0, MSS_EA(A, i, j - 1))

```

- Graph of function calls is a tree with  $O(n^2)$  nodes
- Same function called many times!
- Redraw call graph as a DAG of overlapping problems.
- Only  $O(n)$  nodes!
- Dynamic programming: remember work done before, or compute from bottom up

```

1 def MSS(A):
2     m = mss_ea = A[0]
3     for i in range(1, len(A)):
4         mss_ea = A[i] + max(0, mss_ea)
5         m = max(m, mss_ea)
6     return m

```