



Department of Computer Science and Engineering

Data Structures and Object-Oriented Design

(CSE – 2050)

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CSE-2050 – Data Structures and Object-Oriented Design

Review

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Module 5 – Recursion and Dynamic Programming

- Recursion
→ Solving a problem using function that recursively call itself.

```
1 def sum(k):
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
5 print(sum(5))
```

```
1 def sum(k):           k=0
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
1 def sum(k):           k=1
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
1 def sum(k):           k=2
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
1 def sum(k):           k=3
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
1 def sum(k):           k=4
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
1 def sum(k):           k=5
2     if k > 0:
3         return sum(k - 1) + k
4     return 0
```

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Module 5 – Recursion and Dynamic Programming

- FunctionCall Stack
 - Used Stack (LIFO) structure to hold recursive calls
 - In Python, infinite recursive calls are restricted to 1000 (by default)
 - Can be modified using `sys.setrecursionlimit(N)`

```

1 def sum(k):
2     if k > 0:
3         return sum(k - 1) + k
4     return 0

```

•

```

1 def sum(k):
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```

```

1 def sum(k):
2     if k > 0:
3         return sum(k - 1) + k
4     return 0

```

FunctionCall Stack



Module 5 – Recursion and Dynamic Programming

1. Have a Base Case:

```

1 def sum(k):
2     if k > 0:
3         return sum(k - 1) + k
4     return 0

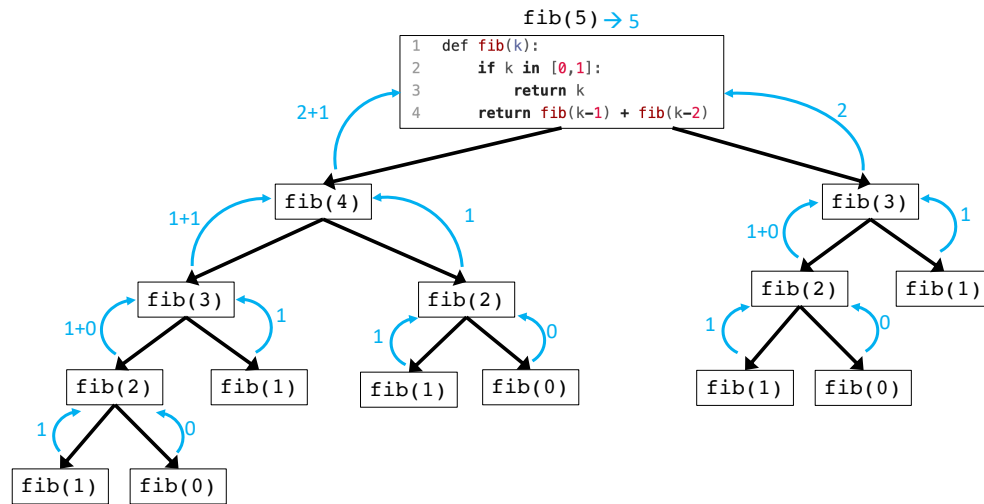
```

2. Recursion calls should move towards the base case



Module 5 – Recursion and Dynamic Programming

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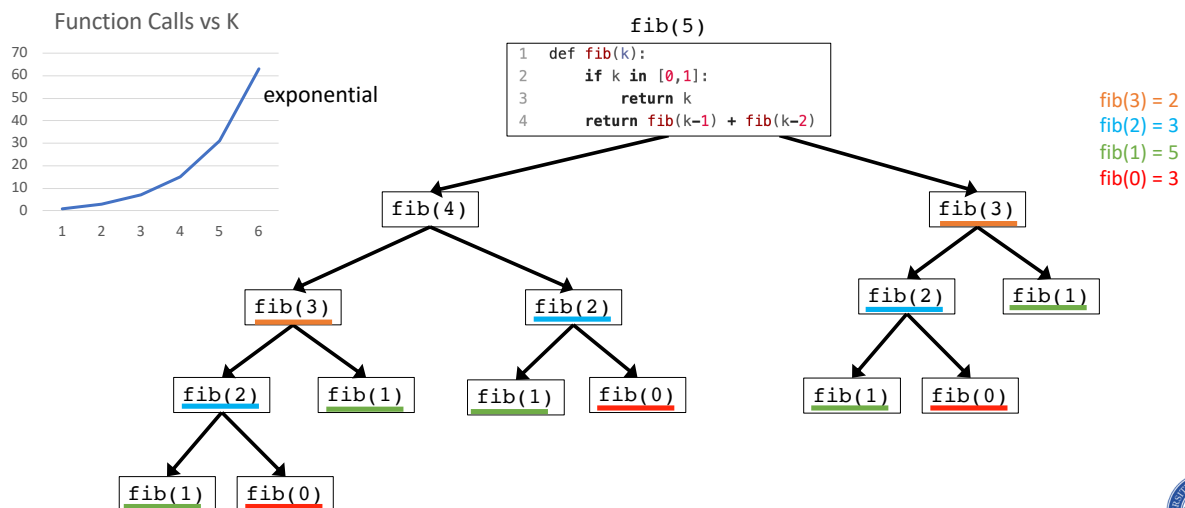
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Module 5 – Recursion and Dynamic Programming

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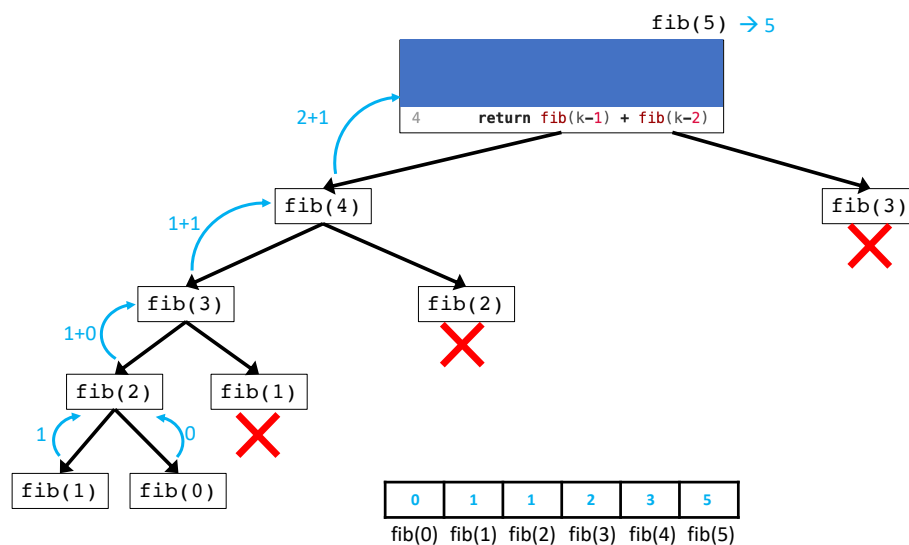
Module 5 – Recursion and Dynamic Programming

Memoization

- Avoid making a function call again which has already been executed by:
- storing the intermediate solution of subproblems and use them later wherever needed
- Also called Top-down approach which uses recursion + caching

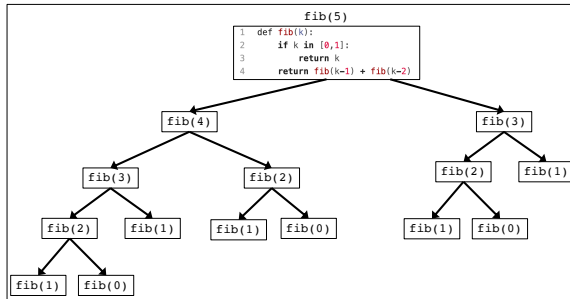


Module 5 – Recursion and Dynamic Programming

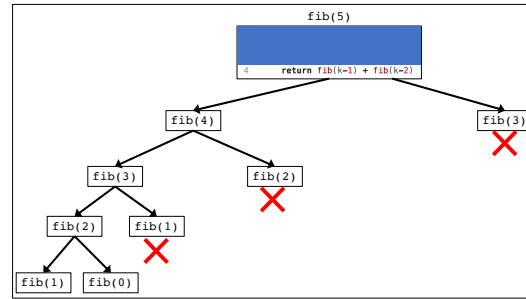


Module 5 – Recursion and Dynamic Programming

Total number of function calls



15



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$$\text{fib}(n) = n + 3$$

$$O(n)$$

The complexity reduces from exponential to polynomial or linear

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Module 5 – Recursion and Dynamic Programming

Dynamic Programming –

Bottom up Approach using tabulation (iterative) method

0	1	1	2	3	5
---	---	---	---	---	---

```

1 def fibo_dyn(k):
2     if k <= 1:
3         return k
4
5     F = [0, 1]
6
7     for i in range(2, k+1, 1):
8         F.append( F[i - 1] + F[i - 2] )
9     return F[k]
  
```

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Module 6 – Searching and Sorting

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Binary Search

Search for an element in a sorted array by dividing the search interval in half.

1. Start by examining the middle item → return if it is the required item
2. If the required item is less than the middle item, disregard the upper half of the search space. If it is greater than the middle item, disregard the lower half of the search space.
3. Repeat until the required item is found or until the search space becomes empty



Module 6 – Searching and Sorting

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```
student_ids = [1432, 1433, 1434, 1435, 1436, 1437, 1438, 1439, 1440, 1441, 1442, 1443, 1444, 1445, 1446, 1447,
1448, 1449, 1450, 1451, 1452, 1453, 1454, 1455, 1456, 1457, 1458, 1459, 1460, 1461, 1462, 1463,
1464, 1465, 1466, 1467, 1468, 1469, 1470, 1471, 1472, 1473, 1474, 1475, 1476, 1477, 1478, 1479,
1480, 1481, 1482, 1483, 1484, 1485, 1486, 1487, 1488, 1489, 1490, 1491, 1492, 1493, 1494, 1495,
1496, 1497, 1498, 1499, 1500, 1501, 1502, 1503, 1504, 1505, 1506, 1507, 1508, 1509, 1510, 1511,
1512, 1513, 1514, 1515, 1516, 1517, 1518, 1519, 1520, 1521, 1522, 1523, 1524, 1525, 1526, 1527,
1528, 1529, 1530, 1531, 1532, 1533, 1534, 1535, 1536, 1537, 1538, 1539, 1540, 1541, 1542, 1543,
1544, 1545, 1546, 1547, 1548, 1549, 1550, 1551, 1552, 1553, 1554, 1555, 1556, 1557, 1558, 1559,
1560, 1561, 1562, 1563, 1564, 1565, 1566, 1567, 1568, 1569, 1570, 1571, 1572, 1573, 1574, 1575,
1576, 1577, 1578, 1579, 1580, 1581, 1582, 1583, 1584, 1585, 1586, 1587, 1588, 1589, 1590, 1591,
1592, 1593, 1594, 1595, 1596, 1597, 1598, 1599, 1600, 1601, 1602, 1603, 1604, 1605, 1606, 1607,
1608, 1609, 1610, 1611, 1612, 1613, 1614, 1615, 1616, 1617, 1618, 1619, 1620, 1621, 1622, 1623,
1624, 1625, 1626, 1627, 1628, 1629, 1630, 1631, 1632, 1633, 1634, 1635, 1636, 1637, 1638, 1639,
1640, 1641, 1642, 1643, 1644, 1645, 1646, 1647, 1648, 1649, 1650, 1651, 1652, 1653, 1654, 1655,
1656, 1657, 1658, 1659, 1660, 1661, 1662, 1663, 1664, 1665, 1666, 1667, 1668, 1669, 1670, 1671,
1672, 1673, 1674, 1675, 1676, 1677, 1678, 1679, 1680, 1681, 1682, 1683, 1684, 1685, 1686, 1687,
1688, 1689, 1690, 1691, 1692, 1693, 1694, 1695, 1696, 1697, 1698, 1699]
```



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Review

Module 6 – Searching and Sorting
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student_ids =

1649

[1634, 1635, 1636, 1637, 1638, 1639, 1640, 1641, 1642, 1643, 1644, 1645, 1646, 1647, 1648, 1649, 1650, 1651, 1652, 1653, 1654, 1655, 1656, 1657, 1658, 1659, 1660, 1661, 1662, 1663, 1664, 1665]

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Review

Module 6 – Searching and Sorting
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student_ids =

[... 1651 ...]

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Module 6 – Searching and Sorting

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```
def BS(L, item):
    if len(L) == 0:
        return False
    mid_index = len(L) // 2
    if item == L[mid_index]:
        return True
    elif item < L[mid_index]:
        return BS(L[:mid_index], item)
    else:
        return BS(L[mid_index + 1:], item)
```

Using Slicing → O(n)



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```
def BS_improved(L, item, lower, upper):
    if lower > upper:
        return False
    else:
        mid_index = (lower + upper) // 2
        if item == L[mid_index]:
            return True
        elif item < L[mid_index]:
            return BS_improved(L, item, lower, mid_index - 1)
        else:
            return BS_improved(L, item, mid_index + 1, upper)
```

Slicing removed



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0th Iteration:Data size = n 1st Iteration:Data size = $n/2$ or $n/2^1$ 2nd Iteration:Data size = $(n/2)/2 \rightarrow n/4$ or $n/2^2$ 

....

....

At K^{th} Iteration, the data size becomes 1:

$$n/2^k = 1$$

$$n = 2^k$$

$$\log n = \log 2^k$$

$$\rightarrow O(\log n)$$

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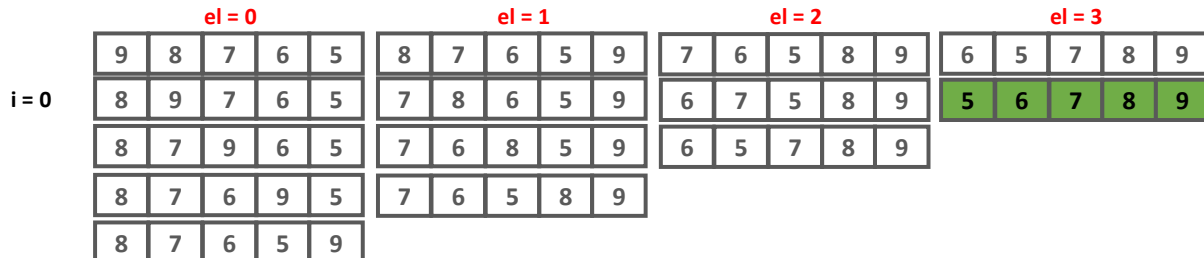
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Bubble Sort Algorithm

```

1 def bubble_sort(L):
2     for el in range(len(L) - 1):
3         for i in range(len(L) - 1 - el):
4             if L[i] > L[i+1]:           #If two items are out of order
5                 L[i], L[i+1] = L[i+1], L[i]   #Switch them

```


 $O(n^2)$

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Selection Sort Algorithm

- Selection sort algorithm is another approach of sorting data in $O(n^2)$ quadratic running time
- We can either find the smallest item and place it in the beginning
OR
- Find the biggest item and move it to end

```

1 def SS_min(L):
2     for i in range(len(L) - 1):
3         min = i
4         for j in range(i + 1, len(L)):
5             if L[j] < L[min]:
6                 min = j
7         #swap
8         L[i], L[min] = L[min], L[i]

```

```

def SS_max(L):
    for i in range(len(L) - 1):
        max = 0
        for j in range(1, len(L)-i):
            if L[j] > L[max]:
                max = j
        #swap
        L[-1 - i], L[max] = L[max], L[-1 - i]

```

- Selection sort is better for applications where less number of write operations are required

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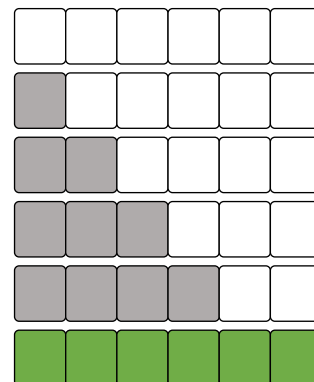
Insertion Sort Algorithm

- Online algorithm – sort array as it receives data (example from web)

```

1 def insertion_sort(L):
2     for i in range(1, len(L)):
3         j = i
4         while j > 0 and L[j] < L[j-1]:
5             L[j-1], L[j] = L[j], L[j-1]
6             j -= 1

```



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Module 6 – Searching and Sorting

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- **Bubble sort**
 - Iterates over every pair in collection, swaps out of order pairs
 - After x iterations, the last x items are in their final (sorted) place
- **Selection sort**
 - Iterates over every unsorted item in collection, selects the next smallest/biggest
 - After x iterations, the last x items are in their final (sorted) place
- **Insertion sort**
 - Iterates over a progressively growing sorted section of the list
 - Bubbles the next un-sorted item into place
 - After x iterations, the first x items are sorted but may not be in their final place.
- **Cocktail sort**
 - Invariant of bubble sort
 - After x iterations, smallest and largest elements are placed at the right place

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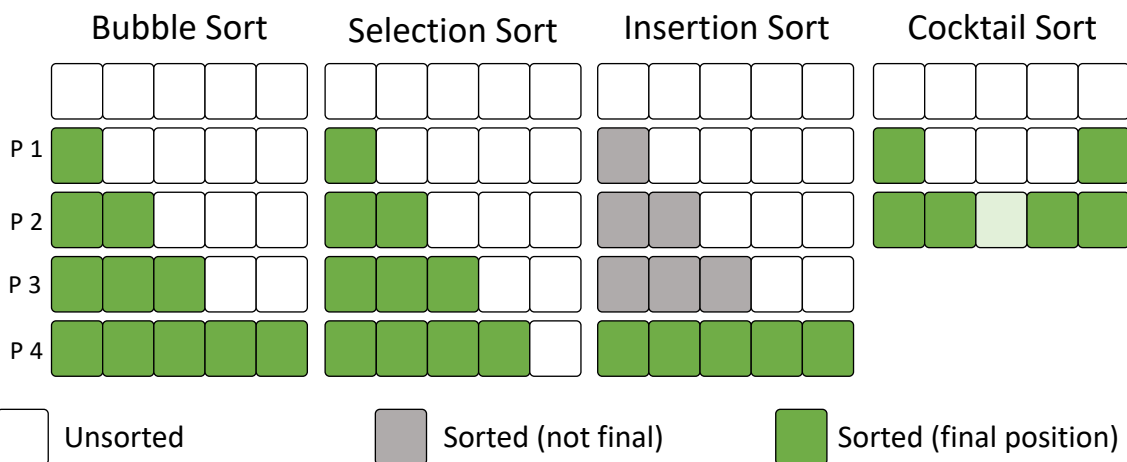


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Module 6 – Searching and Sorting

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Summary



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Module 7 – Divide and Conquer

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- Divide and Conquer is a paradigm for algorithm design and consists of the following three steps:
 1. **Divide:** Divide the input data D into two or more disjoint subsets, D_1 and D_2 .
 2. **Conquer:** Recursively solve the subproblems associated with the subsets, D_1 and D_2 .
 3. **Combine:** Take the solutions to the subproblems, D_1 and D_2 , and merge them into a solution to the original problem D .

Base case: Base case for the recursion are subproblems of size 0 or 1.



Module 7 – Divide and Conquer

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Merge-Sort Algorithm

85	24	63	45	17	31	96	50
----	----	----	----	----	----	----	----

```

Algorithm mergeSort( $D$ )
  Input sequence  $D$  with  $n$ 
           elements
  Output sequence  $D$  sorted

  if  $D.size() > 1$ 
     $(D_1, D_2) \leftarrow partition(D, n/2)$ 
    mergeSort( $D_1$ )
    mergeSort( $D_2$ )
     $D \leftarrow merge(D_1, D_2)$ 

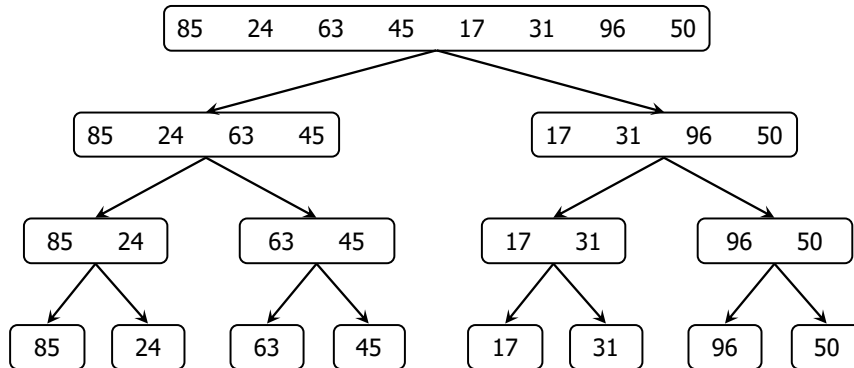
```



Module 7 – Divide and Conquer

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Merge-Sort Algorithm

**Algorithm** *mergeSort(D)***Input** sequence *D* with *n* elements**Output** sequence *D* sorted

```

if D.size() > 1
  (D1, D2) ← partition(D, n/2)
  mergeSort(D1)
  mergeSort(D2)
  D ← merge(D1, D2)

```

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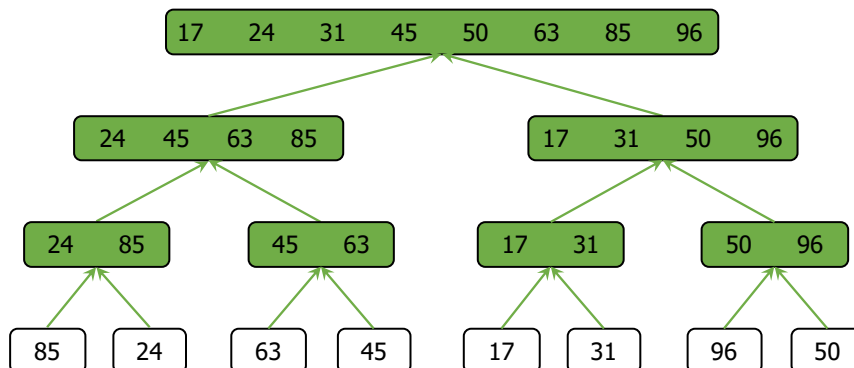


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Module 7 – Divide and Conquer

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Merge-Sort Algorithm

**Algorithm** *mergeSort(D)***Input** sequence *D* with *n* elements**Output** sequence *D* sorted

```

if D.size() > 1
  (D1, D2) ← partition(D, n/2)
  mergeSort(D1)
  mergeSort(D2)
  D ← merge(D1, D2)

```

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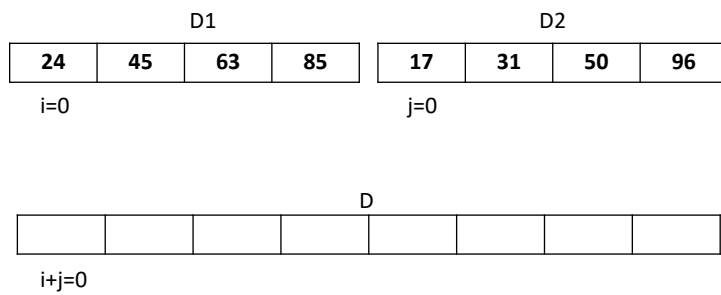


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Module 7 – Divide and Conquer

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Merge-Sort Algorithm



```

merge(D1, D2, D):
    i=j=0
    while i < len(D1) and while j < len(D1)
        if D1[i] < D2[j]:
            D[i+j] = D1[i]
            i += 1
        else:
            D[i+j] = D2[j]
            j += 1
    D[i+j:] = D1[i : ] + D2[j : ]

```

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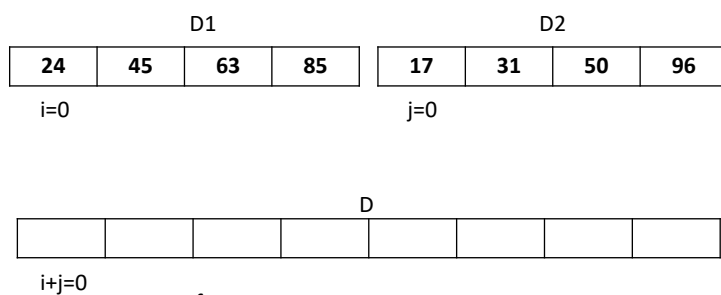


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Module 7 – Divide and Conquer

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Merge-Sort Algorithm



If:
 D1 contains $m1$ elements
 D2 contains $m2$ elements

→ $O(m1 + m2)$

```

merge(D1, D2, D):
    i=j=0
    while i < len(D1) and while j < len(D1)
        if D1[i] < D2[j]:
            D[i+j] = D1[i]
            i += 1
        else:
            D[i+j] = D2[j]
            j += 1
    D[i+j:] = D1[i : ] + D2[j : ]

```

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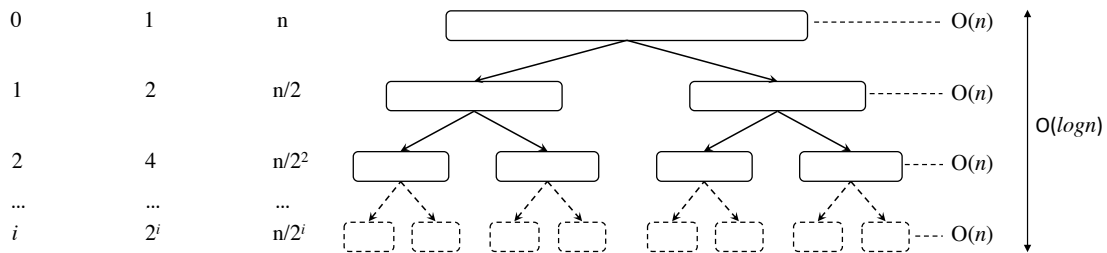
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Merge-Sort Algorithm

Depth | # nodes/sequences | size



- The amount of work done at each node is merge + partition
 - Total work done at depth i is: number of nodes \times size of nodes = $2^i \times n/2^i \rightarrow O(n)$
- What is the stopping condition of recursion? $\rightarrow O(\log n)$

\rightarrow Total time complexity = $O(n \log n)$

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Module 7 – Divide and Conquer

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Quick-Sort Algorithm

85	24	63	45	17	31	96	50
----	----	----	----	----	----	----	----

```

Algorithm quickSort(D)
Input sequence D with n
elements
Output sequence D sorted

if D.size() > 1
    pivot ← pick x from D
    L ← elements less than x
    E ← element equal to x
    G ← elements greater than x
    L = quickSort(L)
    G = quickSort(G)
    return L + E + G
else:
    return D
  
```

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Module 7 – Divide and Conquer

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Quick-Sort Algorithm

- Picking last element in the list as pivot

85 24 63 45 17 31 96 **50**

```

Algorithm quickSort(D)
  Input sequence D with n
        elements
  Output sequence D sorted

  if D.size() > 1
    pivot ← pick x from D
    L ← elements less than x
    E ← element equal to x
    G ← elements greater than x
    L = quickSort(L)
    G = quickSort(G)
    return L + E + G
  else:
    return D

```

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Module 7 – Divide and Conquer

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Quick-Sort Algorithm

- Picking last element in the list as pivot

24 45 17 31 **50** 85 63 96

```

Algorithm quickSort(D)
  Input sequence D with n
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  Output sequence D sorted

  if D.size() > 1
    pivot ← pick x from D
    L ← elements less than x
    E ← element equal to x
    G ← elements greater than x
    L = quickSort(L)
    G = quickSort(G)
    return L + E + G
  else:
    return D

```

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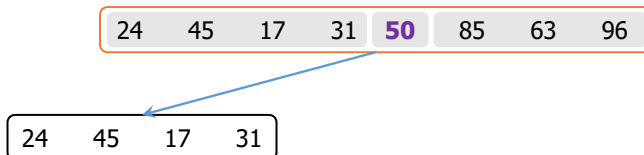


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Module 7 – Divide and Conquer

Quick-Sort Algorithm

- Picking last element in the list as pivot



```

Algorithm quickSort(D)
  Input sequence D with n
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  if D.size() > 1
    pivot ← pick x from D
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    G ← elements greater than x
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    G = quickSort(G)
    return L + E + G
  else:
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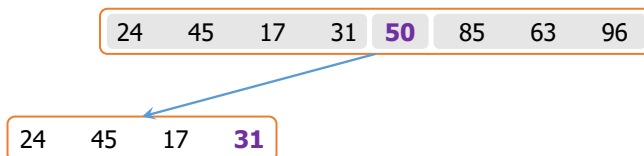
```



Module 7 – Divide and Conquer

Quick-Sort Algorithm

- Picking last element in the list as pivot



```

Algorithm quickSort(D)
  Input sequence D with n
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    G ← elements greater than x
    L = quickSort(L)
    G = quickSort(G)
    return L + E + G
  else:
    return D

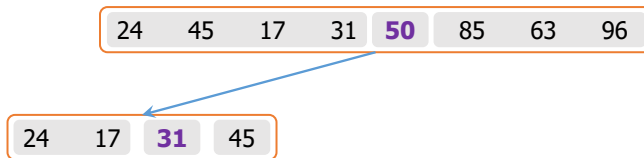
```



Module 7 – Divide and Conquer

Quick-Sort Algorithm

- Picking last element in the list as pivot

**Algorithm** *quickSort(D)*

Input sequence *D* with *n* elements

Output sequence *D* sorted

if *D.size()* > 1

pivot ← pick *x* from *D*

L ← elements less than *x*

E ← element equal to *x*

G ← elements greater than *x*

L = *quickSort*(*L*)

G = *quickSort*(*G*)

 return *L* + *E* + *G*

else:

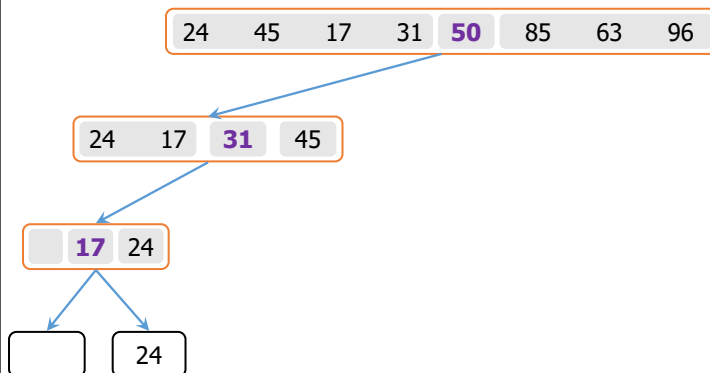
 return *D*



Module 7 – Divide and Conquer

Quick-Sort Algorithm

- Picking last element in the list as pivot

**Algorithm** *quickSort(D)*

Input sequence *D* with *n* elements

Output sequence *D* sorted

if *D.size()* > 1

pivot ← pick *x* from *D*

L ← elements less than *x*

E ← element equal to *x*

G ← elements greater than *x*

L = *quickSort*(*L*)

G = *quickSort*(*G*)

 return *L* + *E* + *G*

else:

 return *D*

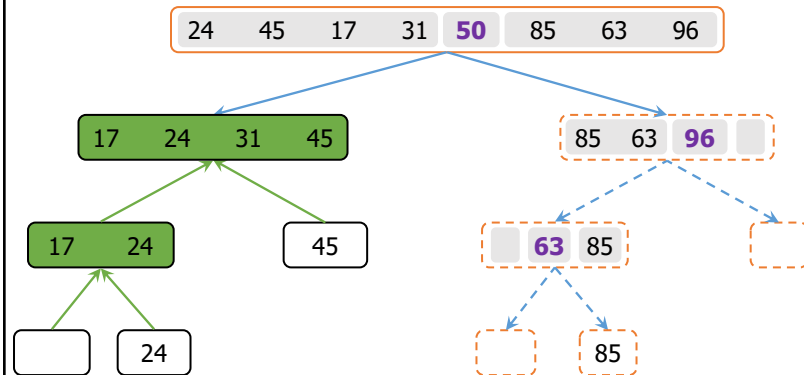


Module 7 – Divide and Conquer

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Quick-Sort Algorithm

- Picking last element in the list as pivot

**Algorithm** *quickSort(D)***Input** sequence *D* with *n* elements**Output** sequence *D* sorted**if** *D.size()* > 1 *pivot* ← pick *x* from *D* *L* ← elements less than *x* *E* ← element equal to *x* *G* ← elements greater than *x* *L* = *quickSort(L)* *G* = *quickSort(G)* return *L* + *E* + *G***else:** return *D*

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Module 7 – Divide and Conquer

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Quick-Sort Algorithm (in-place) implementation

- Data items are manipulated within the same container for sorting → thus saves memory

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Module 7 – Divide and Conquer

Quick-Sort Algorithm (in-place) implementation

```

quickSort(D, L_idx, H_idx)
  if L_idx < H_idx:
    pivot = partition(D, L_idx, H_idx)
    quickSort(D, L_idx, pivot-1)
    quickSort(D, pivot+1, H_idx)
  end

```

D

23	6	4	-1	0	12	8	3	1
----	---	---	----	---	----	---	---	---

L_idx=0

H_idx=8

```

partition(D, low, high)
  pivotindex = (low+high) // 2
  swap(pivotindex, high)

  i = low

  for j in range (low, high+1)
    if D[j] <= D[high]
      swap(i, j)
      i += 1
  return i-1

```

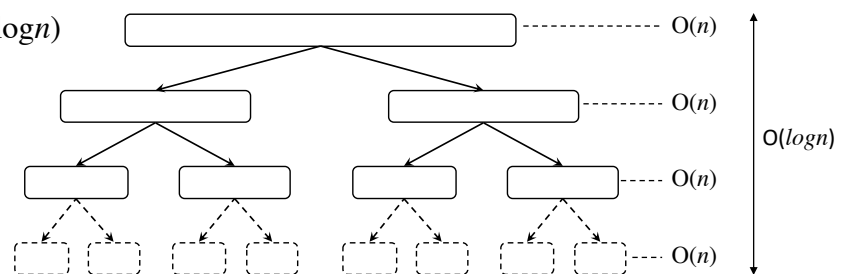
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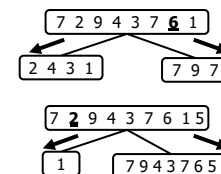
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Module 7 – Divide and Conquer

- Depends on pivot
- **Best case** runtime $\rightarrow O(n \log n)$



- For a sequence of size D
 - Good Pivot \rightarrow generates L and G each of size less than $\frac{3}{4}D$
 - Bad Pivot \rightarrow generates either L or G of size greater than $\frac{3}{4}D$



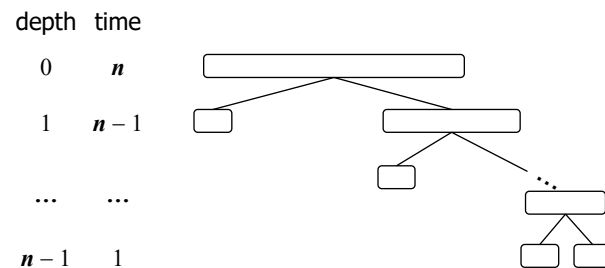
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Module 7 – Divide and Conquer

- The worst case for quick-sort occurs when the pivot is the unique minimum or maximum element

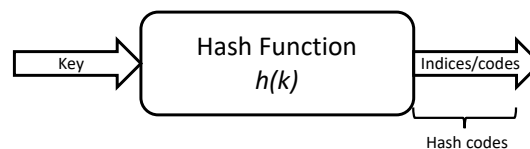


- Worst case runtime $\rightarrow O(n^2)$



Module 8 – Mapping and Hash Tables

- Mapping \rightarrow association between two objects \rightarrow key-value pairs
 - Python has dictionary data structure to hold key-value pairs
- To map keys, hash function is used to generate the index location

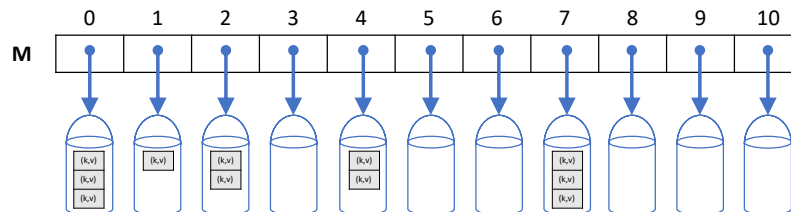


- Hash Collision \rightarrow hash function generate the same code for different keys to map at the same location



Collision Handling scheme

- Separate Chaining
 - Instead of having a single object at each location in hash table, we conceptualized to have buckets



- In Worst case:
 - Time to search for the bucket is $O(1)$
 - Time to search a key in the bucket depends on the size of bucket:
For size $n \rightarrow O(n)$

**Collision Handling scheme**

- Open addressing – Linear Probing
 - Inserting an element (k,v) at $M[j]$
 - j is the index generated by hash function
 - If j^{th} place is occupied, we try $M[(j+1) \% N]$
 - If this place is also occupied, we next try $M[(j+2) \% N]$, and so on

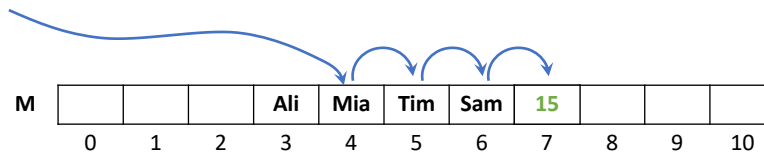


Module 8 – Mapping and Hash Tables

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Collision Handling scheme

- Inserting a new element with key $k = 15 \rightarrow k \bmod N \rightarrow 15 \bmod 11 = 4$
- This new item should be placed at location 4



- This requires additional implementation to search for an existing key
- Accessing cell array is analogous to probing the bucket to find its content

Hasan Baig

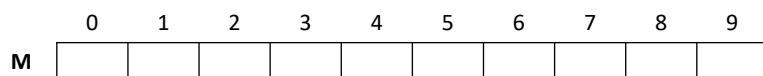


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Module 8 – Mapping and Hash Tables

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- A good hash function index “n” items of a map in a bucket array of capacity “N”
 \rightarrow The expected size of a bucket is n/N
- The ration $\lambda = n/N$ is called “**Load Factor**” of the hash table
 - Bounded by a small constant (preferably below 1)
- Example:
 - $n = 15$
 - $N = 10$
 - $\rightarrow \lambda = 1.5 \rightarrow$ collision



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Announcements

- Exam Duration – 50 minutes
 - Total points: 30
 - Total 4 sections → 6 MCQs in each
 - All questions carry 1 point except one that carry 2.5 points
 - Question # 25 is a bonus question which carries 5 points
- No Labs this week



if D has n elements
then looping through D:

for i in range($\text{len}(D)$)

...

will have $O(n)$ time complexity!

This is very simple!

Now, if we say D has $n \times n$ elements
i.e. n^2 elements, then looping through
D will have $O(n^2)$ time complexity.

