

Programming and Data Structures - I

Lecture 17

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SORTING



BUBBLE SORT

Bubble Sort

Key steps

- Multiple passes
- In each pass,
 - an element (say $a[j]$) is compared with its successor ($a[j + 1]$)
 - $a[j]$ and $a[j + 1]$ are interchanged if not in the desired order (for ascending order, interchange if $a[j] > a[j + 1]$)
- After the first pass, for ascending order sort, the largest element is at the last position (In general, after the i th iteration, $a[n - i + 1]$ is placed in its proper position)

Bubble Sort: Implementation

Code

```
for(i=0; i<n - 1; i++)
{
    for(j=0; j<n - 1 - i; j++)
    {
        if(arr[j] > arr[j+1])
        {
            temp = arr[j];
            arr[j] = arr[j+1];
            arr[j+1] = temp;
        }
    }
}
```

Bubble sort: illustration

Pass 1

A

5	4	3	2	1
---	---	---	---	---



4	5	3	2	1
---	---	---	---	---

Bubble Sort

Exchange

4	3	5	2	1
---	---	---	---	---

Exchange

4	3	2	5	1
---	---	---	---	---

Exchange

4	3	2	1	5
---	---	---	---	---

Exchange

Bubble sort: illustration (contd.)

Bubble Sort

A

5	4	3	2	1
---	---	---	---	---

Before sort

4	3	2	1	5
---	---	---	---	---

After Pass 1

3	2	1	4	5
---	---	---	---	---

After Pass 2

2	1	3	4	5
---	---	---	---	---

After Pass 3

1	2	3	4	5
---	---	---	---	---

After Pass 4 (Sorted)

Bubble Sort: Time Complexity

Intuition

- Number of passes = $n - 1$
- Number of steps per pass $\propto n - i$ ($i = 1, 2, \dots$, no of passes)
- Total number of operations $\propto (n - 1) + (n - 2) + \dots + 1 = \frac{(n-1)(n-1+1)}{2} = \frac{(n-1)n}{2}$

Asymptotic worst case time complexity

$$\mathcal{O}(n^2)$$

Bubble Sort: Conclusions

Advantages

- Easy to understand
- Introductory sorting algorithm
- In-place sort: No extra space requirement

Disadvantages

- Worst performing sorting algorithm: best, worst and average case complexity is $\mathcal{O}(n^2)$
- Impractical for large data

SELECTION SORT

Selection Sort

Key steps

For an array $a[0..n - 1]$

- In step 1, the smallest element in $a[0..n - 1]$ is placed in $a[0]$; In step 2, the smallest element in $a[1..n - 1]$ is placed in $a[1]$,..., in step $n - 1$, the smallest element in $a[n - 2..n - 1]$ is placed in $a[n - 2]$
- In general, in step i , the smallest element in $a[i - 1..n - 1]$ is placed in $a[i - 1]$
- At each step, the original element in $a[i - 1]$ is swapped with the smallest element

Selection Sort: illustration

A

0	1	2	3	4
5	4	3	2	1

min

Searching in A[0...4]

Placing min at A[0]

1	4	3	2	5
---	---	---	---	---

min

Searching in A[1..4]

Placing 2nd min at A[1]

1	2	3	4	5
---	---	---	---	---

min

Searching in A[2..4]

Placing 3rd min at A[2]

1	2	3	4	5
---	---	---	---	---

min

Searching in A[3..4]

Placing 4th min at A[3]

1	2	3	4	5
---	---	---	---	---

SORTING COMPLETE!

Selection Sort

Code

```
for(i=0; i<n-1; i++)
{
    /*Find the index of smallest element*/
    min=i;
    for(j=i+1; j<n ; j++)
    {
        if(arr[min] > arr[j])
            min=j ;
    }
    if(i!=min) Swap smallest element with arr[i]
    {
        temp = arr[i];
        arr[i] = arr[min];
        arr[min] = temp ;
    }
}
```

Selection Sort: Time Complexity

Intuition

- Number of passes = $n - 1$
- Number of steps per pass $\propto (n - i) + 1$ ($i = 1, 2, \dots$, no of passes)
- Total number of operations $\approx ((n - 1) + (n - 2) + \dots + 1 + n - 1) = \frac{(n-1)(n-1+1)}{2} + (n - 1) = \frac{(n-1)n}{2} + n - 1 = \frac{n^2}{2} + \frac{n}{2} - 1$

Asymptotic worst case time complexity

$$\mathcal{O}(n^2)$$

Selection Sort: Conclusions

Advantages

- Easy to understand and implement
- In-place sort: No extra space requirement
- Considered to be faster than Bubble sort as it takes lesser (i.e. $\mathcal{O}(n)$) swaps

Disadvantages

- Best, worst and average case complexity is $\mathcal{O}(n^2)$
- Impractical for large data

QUICK SORT

Quick Sort (partition exchange sort)

Key steps

In an array arr of length n

- choose an element $pivot$ (e.g. the first element)
- **Partitioning**
 - place $pivot$ in the expected position in the ascending sorted permutation of arr
 - the sub-array $arr[0 \dots pivot - 1]$ contains elements in arr which are smaller than (or equal to) $pivot$
 - the sub-array $arr[pivot + 1 \dots n - 1]$ contains elements in arr which are greater than (or equal to) $pivot$
- Partitioning continues in the subarrays until all the $pivots$ are correctly placed in arr

Developed by British computer scientist **Tony Hoare** (Turing award winner).

Quick Sort: Partitioning

Code

```
int partition(int arr[], int low, int up)
{
    int temp, i, j, pivot;
    i = low + 1;
    j = up;
    pivot = arr[low];

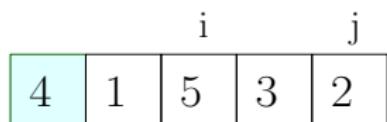
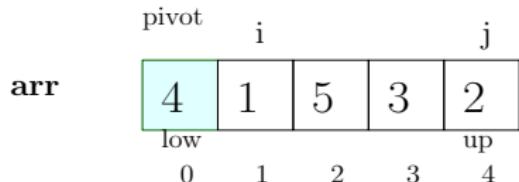
    while(i <= j)
    {
        while(arr[i] < pivot && i<up)
            i++;

        while( arr[j] > pivot )
            j--;

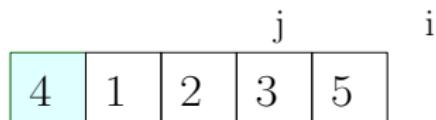
        if(i < j)
        {
            temp=arr[i];
            arr[i]=arr[j];
            arr[j]=temp;
            i++;
            j--;
        }
        else
            i++;
    }

    arr[low] = arr[j];
    arr[j] = pivot;
    return j;
}
```

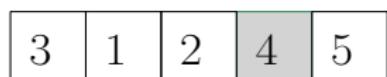
Quick sort partitioning: illustration



```
while(arr[i] < pivot && i < up)
    i++;
while(arr[j] > pivot )
    j--;
if(i < j)
    swap(arr[i], arr[j])
else i++;
```



Swapping and continuation of the loop



left subarray

```
arr[low]=arr[j];
arr[j]=pivot;
```

right subarray

Partitioning at 4

Quick Sort: Subroutine

Code

```
void quick(int arr[],int low,int up)
{
    int pivloc;
    if(low >= up)
        return;
    pivloc = partition(arr, low, up);
    quick(arr,low, pivloc - 1); /*recursive call to left subarray*/
    quick(arr, pivloc + 1, up); /*recursive call to right subarray*/
}
```

Quick Sort: main

Code

```
main()
{
    int array[MAX], n, i;
    printf("Enter the number of elements : ");
    scanf("%d", &n);
    for(i=0;i<n;i++)
    {
        printf("Enter element %d : ", i+1);
        scanf("%d", &array[i]);
    }
    quick(array, 0, n-1);
    printf("Sorted list is :\n");
    for( i=0; i<n; i++ )
        printf("%d ", array[i]);
    printf("\n");
}
```

Quick Sort: Time Complexity

Best and average case

- If partitioning divides the array (of length n) into near equal halves every time, the number of partition steps be k
- $\frac{n}{2^k} = 1 \Rightarrow k \approx \log_2 n$ (from n we end when the sub-arrays contain single elements)
- Each partition step has maximum of n operations (approx)
- Total number of steps $\approx n * \log_2 n = O(n \log n)$

Quick Sort: Time Complexity

Worst case

- If partitioning divides the array (of length n) into subarrays of size 0 and $n - 1$ (e.g. if the array is in descending order), the number of partition steps be k
- $k = n - 1$ (from n we end when the sub-arrays contain single elements)
- Each partition step has maximum of n operations (approx)
- Total number of steps $\approx n * (n - 1) = \mathcal{O}(n^2)$

Quick Sort: Conclusions

Worst case

- Worst case rarely happens
- Even for highly unbalanced partitions (e.g. 9:1 or even 99:1) the complexity remains $\mathcal{O}(n \log_2 n)$
- Preferred algorithm for large data

MERGE

Merge

Purpose

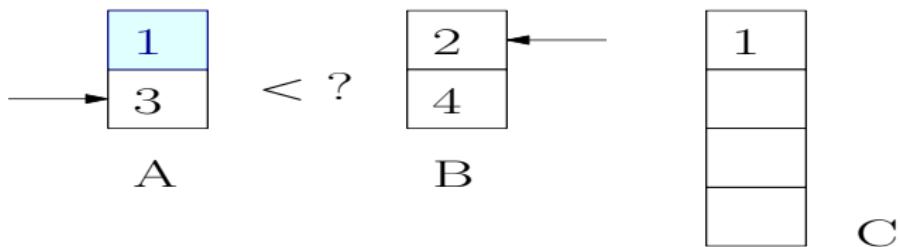
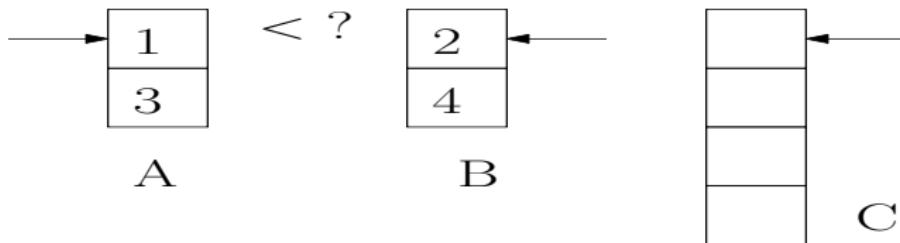
Produce a sorted list from sorted sublists

Features

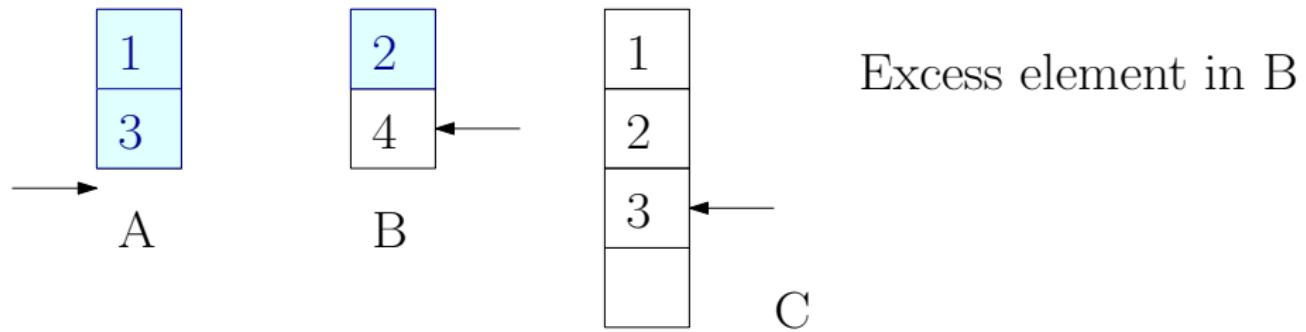
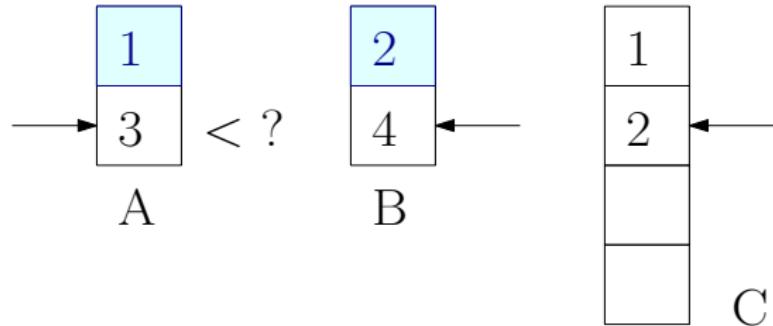
For two sorted lists $A[0 \dots n - 1]$ and $B[0 \dots m - 1]$ and an empty list C

- ① Compare $A[0]$ and $B[0]$ and put the smaller in C ; consider the element next to which is put in C
- ② In general, put the smaller of $A[i]$ and $B[j]$ in C ; increment i or j according as whether $A[i]$ or $B[j]$ is copied to C
- ③ If all elements in A (or B) are copied to C , copy the remaining elements of B (or A) to C

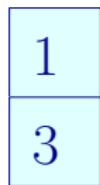
Merge: illustration



Merge: illustration



Merge: illustration



A



B



C

MERGE COMPLETE!

Merge: implementation

Code

```
void merge(int arr1[], int arr2[], int temp[], int n1, int n2 ) //Merge arr1 and arr2 to temp
{
    int i = 0;
    int j = 0;
    int k = 0;

    while(i <= n1 && j <= n2)
    {
        if( arr1[i] <= arr2[j])
            temp[k++] = arr1[i++];
        else
            temp[k++] = arr2[j++];
    }
    while(i <= n1) //Copy excess elements of arr1 to temp
    {
        temp[k++] = arr1[i++];
    }
    while(j<=n2) //Copy excess elements of arr2 to temp
    {
        temp[k++] = arr2[j++];
    }
}
```

Merge: Conclusions

- Time complexity = $\mathcal{O}(m + n)$ (m, n being lengths of the two sublists)
- Vital component of efficient sorting algorithms like **Merge Sort**

MERGE SORT

Merge Sort

Key steps

For a list (array) $A[0 \dots n - 1]$

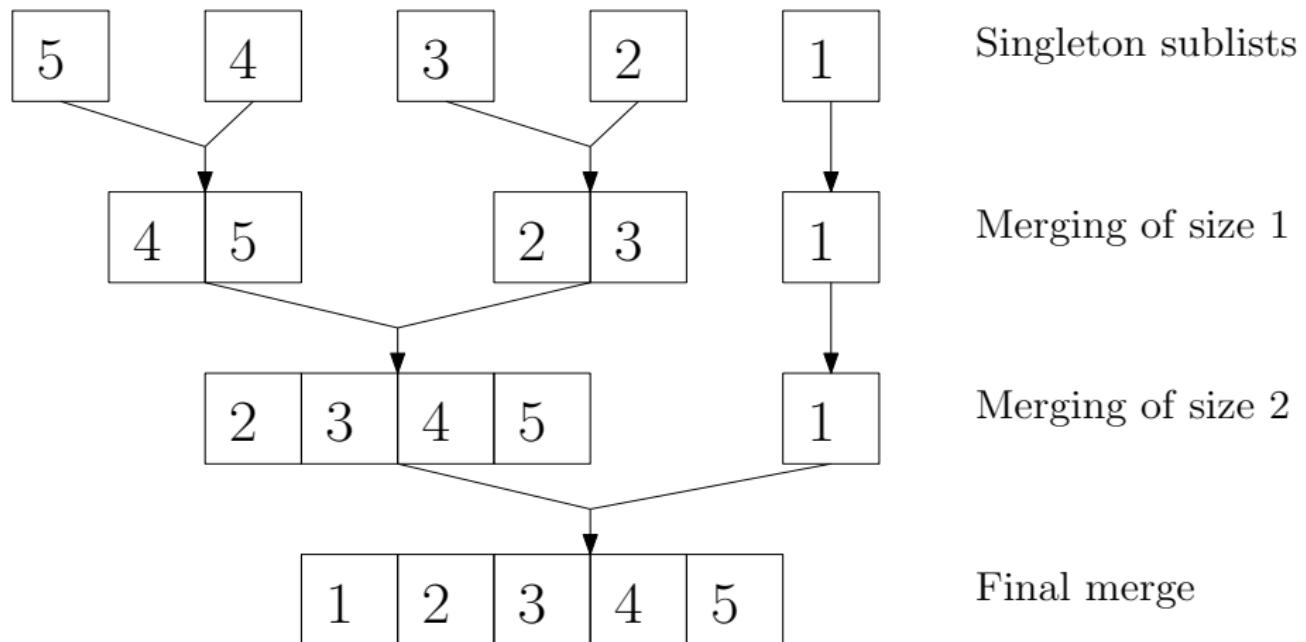
- ① Split into singleton sublists $A[0], A[1], \dots, A[n - 1]$
- ② Use **merge** algorithm to merge these sublists of sizes 1, 2, ...
- ③ Finally a sorted list is produced

Approaches

- **Bottom Up:** Merging sublists of size 1 iteratively to produce the sorted list.
- **Top down:** Recursively splitting the list until each sublist is of size 1 and merging them to produce the sorted list.

Developed by **Herman Heine Goldstine** and **John von Neumann** in 1948.

Merge Sort: illustration of merging



Merge Sort: Time Complexity

- No of passes (merging of sizes 1, 2, ...) $\propto \lceil \log_2 n \rceil$
- No of operations per pass $\propto n$ (maximum size of each sublist is $\frac{n}{2}$; merge operation has complexity $\mathcal{O}(n)$)
- Total number of steps $\propto n \lceil \log_2 n \rceil = \mathcal{O}(n \log_2 n)$

Merge Sort: Conclusions

Advantage

Time complexity is always $\mathcal{O}(n \log_2 n)$

Disadvantage

Requires $\mathcal{O}(n)$ extra space (an extra array of length n)

