Report on Sorting Algorithms

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0.1 Introduction

This report provides the introduction, algorithm, running time and graphs of sorting algorithms.

0.2 Insertion Sort

Insertion sort is an efficient algorithm for sorting a small number of elements. We present the pseudocode called INSERTION-SORT. It takes as a parameter an array /textitA[1..n] containing a sequence of length n that is to be sorted.

0.2.1 Algorithm

```
INSERTION - SORT(A)
1 for j = 2 to A.length
2  key = A[j]
3  i = j - 1
4  while i > 0 and A[i] > key
5  A[i + 1] = A[i]
6  i = i - 1
7  A[i + 1] = key
```

0.2.2 Time Complexity of Insertion Sort

$$INSERTION - SORT(A)$$
 cost times
1 for $j = 2$ to $A.length$ c_1 n

$$2 key = A[j] c_2 n - 1$$

$$3 i = j - 1 c_3 n - 1$$

$$4 while i > 0 and A[i] > key c_4 \Sigma_{j=2}^n t_j$$

$$5 A[i+1] = A[i] c_5 \Sigma_{j=2}^n (t_j - 1)$$

$$6 i = i - 1 c_6 \Sigma_{j=2}^n (t_j - 1)$$

$$7 A[i+1] = key c_7 n - 1$$

The running time of the algorithm is the sum of running times for each statement executed. To compute T(n), the running time of INSERTION-SORT on an input of n values, we sum the products of the cost and times columns, obtaining

$$T(n) = c_1 n + c_2 (n-1) + c_3 (n-1) + c_4 \sum_{j=2}^{n} t_j + c_5 \sum_{j=2}^{n} (t_j - 1) + c_6 \sum_{j=2}^{n} (t_j - 1) + c_7 (n-1)$$

1.Best case:

In INSERTION - SORT, the best case occurs if the array is already sorted. For each j = 2, 3, ..., n, we then find that $A[i] \le key$ in line 5 when i has its initial value of j - 1. Thus $t_j = 1$ for j = 2, 3, ..., n, and the best-case running time is

$$T(n) = c_1 n + c_2 (n-1) + c_3 (n-1) + c_4 (n-1) + c_7 (n-1)$$
$$= (c_1 + c_2 + c_3 + c_4 + c_7)n - (c_2 + c_3 + c_4 + c_7).$$

$$=an+b$$

where, $a = (c_1 + c_2 + c_3 + c_4 + c_7)$ and $b = -(c_2 + c_3 + c_4 + c_7)$, which depend on the statement costs c_i ; it is thus a **linear function** of n.

2. Worst case:

If the array is in reverse sorted order-that is, in decreasing order- the worst case results. We must compare each element A[j] with each element in the entire sorted subarray A[1..j-1], and so $t_j = j$ for j = 2, 3, ..., n. Noting that

$$\sum_{j=2}^{n} j = n(n+1)/2 - 1$$

and

$$\sum_{j=2}^{n} (j-1) = n(n-1)/2$$

Therefore, in worst case, the running time of INSERTION-SORT is

$$T(n) = c_1 n + c_2 (n-1) + c_3 (n-1) + c_4 (n(n+1)/2 - 1) + c_5 (n(n-1)/2 - 1) + c_5 (n(n-1$$

which is a **quadratic equation**. Here, $an^2 + bn + c$ for constants a,b and c that again depend on the statement costs c_i ; it is thus a **quadratic function** of n.

3. Average case:

The **average case** is often roughly as bad as the worst case. On average half the elements in A[1..j-1] are less than A[j], and half the elements are greater. On average, therefore, we check half of the subarray A[1..j-1], and so t_j is about j/2. The resulting average case running time turns out to be a quadratic function of the input size, just like the worst-case running time. Here is the chart of time complexity of insertion sort.

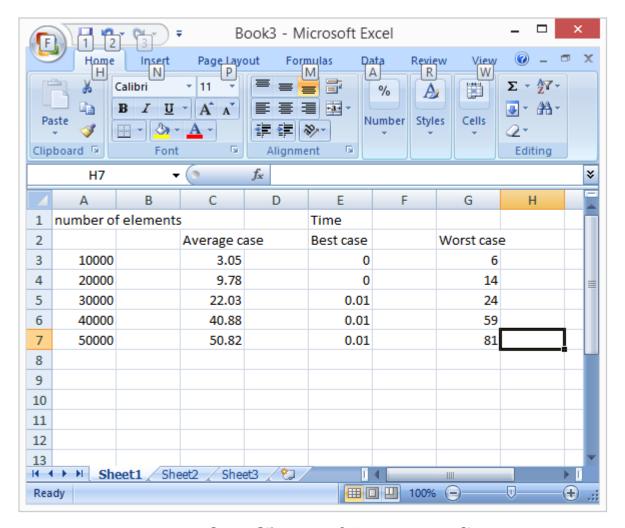


fig: Chart of Insertion Sort.

0.2.3 Graph of Time Complexity of Insertion Sort

Here is the graph of comparison of running times of average case, best case and worst case.

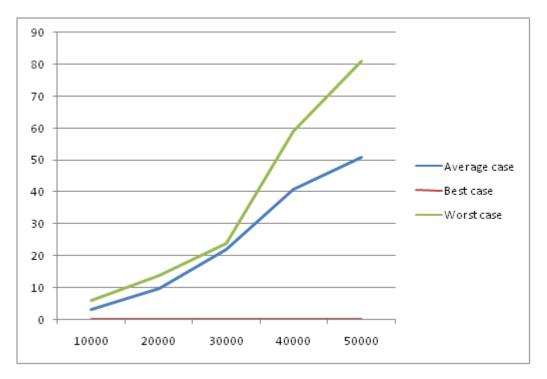


fig: time vs n graph for insertion sort.

0.3 Merge Sort

The **Merge Sort** algorithm closely follows the divide and conquer paradigm. Intuitively, it operates as follows,

Divide:

Divide the n- element sequence to be sorted into two subsequences of n/2 elements each.

Conquer:

Sort the two subsequences recursively using merge sort.

Combine:

Merge the two sorted subsequences to produce the sorted answer.

The key operation of the merge sort algorithm is the merging of two sorted sequences in the **combine** step. We merge by calling an auxiliary procedure MERGE(A, p, q, r), where A is an array and p,q and r are indices into the array such that $p \leq q < r$. The procedure assumes that the subarray A[p..q] and A[q+1..r] are in sorted order. It **merges** them to form a single sorted subarray that replaces the current subarray A[p..r].

Our MERGE procedure takes time $\Theta(n)$, where n = r - p + 1 is the total number of elements being merged.

0.3.1 Algorithm

The following pseudocode implements MERGE(A, p, q, r) procedure, which merges the two subsequences. And another pseudocode MERGE - SORT(A, p, r) sorts the elements in the subarray A[p..r].

$$MERGE(A, p, q, r)$$

1 $n_1 = q - p + 1$
2 $n_2 = r - q$
3 //create arrays $L[1...n_1 + 1]$ and $R[1...n_2 + 1]$.
4 **for** $i = 1$ to n_1
5 $do L[i] = A[p + i - 1]$

```
6 for j = 1 \text{ to } n_2
7 do R[j] = A[q+j]
8 \quad L[n_1+1] = \infty
9 R[n_2+1] = \infty
10 \ i = 1
11 \ j = 1
12 for k = p \ to \ r
13 do if L[i] \leq R[j]
14  then A[k] = L[i]
15 i = i + 1
16 else A[k] = R[j]
17 j = j + 1
MERGE - SORT(A, p, r)
1 if p < r
2 	 q = [(p+r)/2]
3 \quad MERGE - SORT(A, p, q)
4 MERGE - SORT(A, q + 1, r)
5 \quad MERGE(A, p, q, r)
```

0.3.2 Time Complexity of MERGE-SORT

There are no such differences among the best, worst and average case of time complexity in merge sort.

Best case: In best case, the running time of merge sort is, T(n) = O(nlog(n)).

Average case: In average case, the running time of merge sort is, T(n) = O(nlog(n)).

Worst case: In worst case, the running time of worst case is, $T(n) = O(n\log(n))$.

Here, is the chart of time complexity in merge sort-

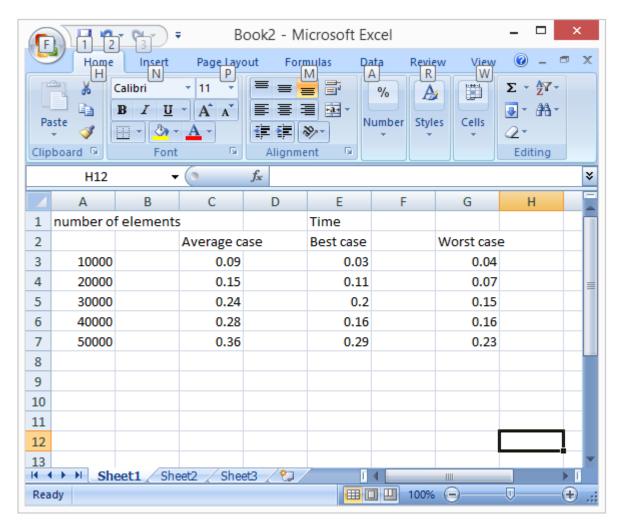


fig: Time Complexity chart.

0.3.3 Graph of Time Complexity of MERGE-SORT

We can represent time complexity of MERGE-SORT by the following graph.

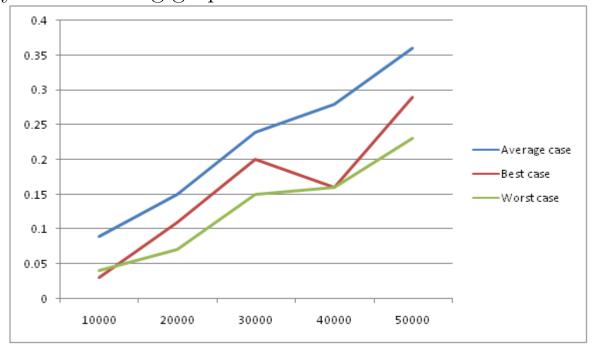


fig: Time complexity graph of merge

0.4 Bubble Sort

sort.

Bubble Sort is a popular, but inefficient, sorting algorithm. It works by repeatedly swapping adjacent elements that are out of order.

0.4.1 Algorithm of BUBBLE SORT

```
BUBBLE - SORT(A)
1 for i = 1 to length[A]
2 do for j = length[A] downto i + 1
3 do if A[j] < A[j - 1]
4 then exchange A[j] < -> A[j - 1]
```

0.4.2 Time Complexity of BUBBLE SORT

In average and worst case, the running time of BUBBLE-SORT are same, $T(n) = O(n^2)$. But in best case, the running time is O(n).

Here is the table of time complexity-

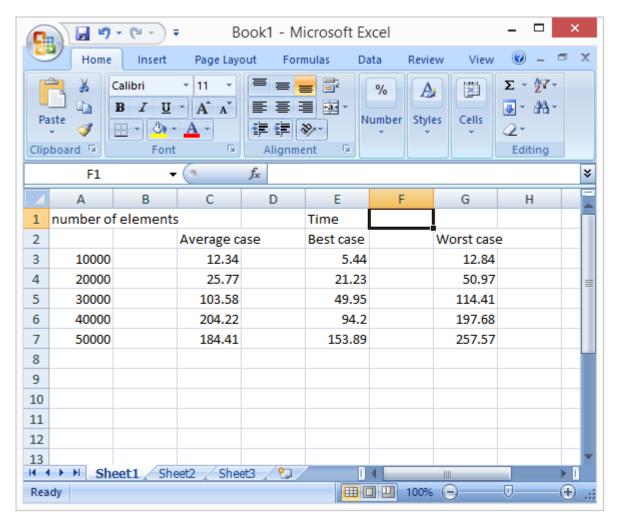
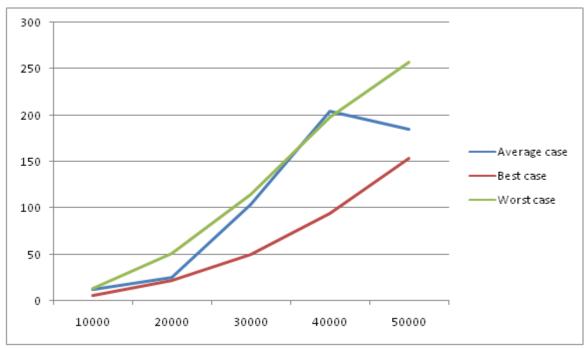


fig: Table of Time Complexity of Bubble Sort.

0.4.3 Time Complexity Graph of BUBBLE-SORT

The following graph represents time complexity of BUBBLE-SORT.



sort.png

fig: Time complexity of BUBBLE-

SORT.

0.5 Quick Sort

Quick Sort, like Merge Sort, applies the divide and conquer paradigm. Here is the three-step divide and conquer process for sorting a typical subarray A[p..r].

Divide: Partition the array A[p..r] into two (possible empty) subarrays A[p..q-1] and A[q+1..r] such that each element of A[p..q-1] is less than or equal to A[q], which is, in turn, less than or equal to each element of A[q+1..r]. Compute the index q as part of this partitioning procedure.

Conquer: Sort the two subarrays A[p..q-1] and A[q+1..r] by recursive calls to quick sort.

Combine: Because the subarrays are already sorted,

no work is needed to combine them: the entire array A[p..r] is now sorted.

0.5.1 Algorithm of Quick Sort

The following procedure implements quicksort:

```
\begin{array}{ll} QUICKSORT(A,p,r) \\ 1 & \textbf{if } p < r \\ 2 & q = PARTITION(A,p,r) \\ 3 & QUICKSORT(A,p,q-1) \\ 4 & QUICKSORT(A,q+1,r) \end{array}
```

Partitioning the array

The key to the algorithm is the PARTITIONING procedure, which rearranges the subarray A[p..r] in place.

0.5.2 Time Complexity of Quick Sort

The running time of quick sort is similar at best and average case, but it is larger at worst case.

Average case: The running time is O(nlgn).

Best case: The running time is O(nlgn). Worst case: The running time is $O(n^2)$.

Here is the Time Complexity Chart of Quick Sort:

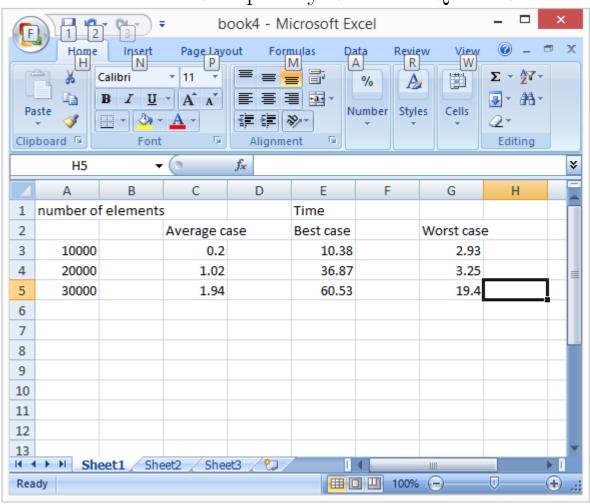


fig: Time Complexity chart of Quick

Sort.

0.5.3 Time Complexity Graph of Quick Sort

From chart, we draw a graph that represents the efficiency of Quick Sort algorithm.

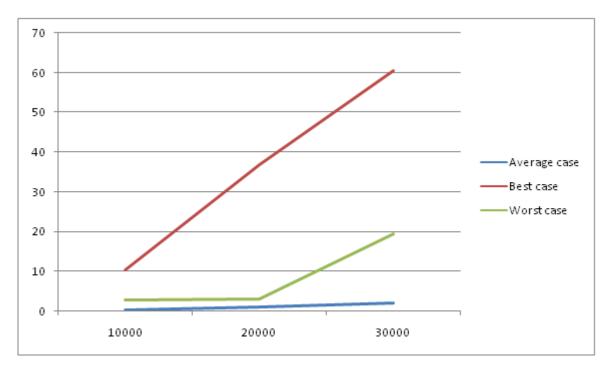


fig: Time Complexity graph of quick

0.6 Heap Sort

sort.

Like merge sort, but unlike insertion sort, heapsort's running time is O(nlgn). Like insertion sort, but unlike merge sort, heapsort sorts in place: only a constant number of array elements are stored outside the input array at any time.

Heapsort also introduces another algorithm design technique: using a data structure, in this case one we call a

heap, to manage information.

0.6.1 Algorithm of Heap Sort

```
MAX - HEAPIFY(A, i)
 1 \quad l = LEFT(i)
 2 r = RIGHT(i)
 3 if l \leq A.heap - size and A[l] > A[i]
 4 \quad largest = l
 5 else largest = i
 6 if r \leq A.heap - size and A[r] > A[largest]
 7 \quad largest = r
 8 if largest \neq i
     exchange A[i] with A[largest]
 9
      MAX - HEAPIFY(A, largest)
 10
 BUILD - MAX - HEAP(A)
```

- $1 \quad A.heap\text{-}size = A.length$
- 2 for i = [A.length/2] downto 1
- MAX HEAPIFY(A, i)

HEAPSORT(A)

- BUILD MAX HEAP(A)
- 2 for i = A.length downto 2
- exchange A[1] with A[i]3
- $4 \quad A.heap\text{-}size=A.heap\text{-}size\text{-}1$
- 5 MAX - HEAPIFY(A, 1)

0.6.2 Time Complexity of Heap Sort

The Time complexity of Heap sort is not so complex. The running time of Heap sort is O(nlgn). This is same for the best, average and worst case.

Chart for Time Complexity

Here is the table that contains the data of running time and number of elements.

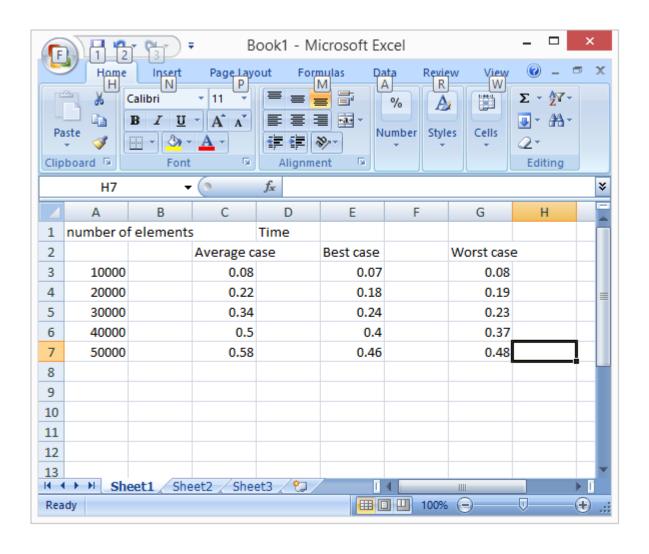


fig: Time Complexity chart.

0.6.3 Time Complexity Graph

Here is the graph that represents the time complexity graph in Heap-Sort.

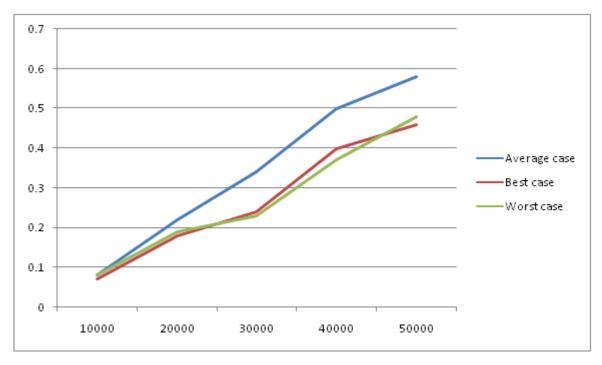


fig: Time Complexity graph of Heap-

Sort.