Sorting - 2

Insertion sort Selection sort Bubble sort The efficiency of handling data can be substantially improved if the data is sorted according to some criteria of order. In a telephone directory we are able to locate a phone number, only because the names are alphabetically ordered. Same thing holds true for listing of directories created by us on the computer. Retrieval of a data item would be very time consuming if we don't follow some order to store book indexes, payrolls, bank accounts, customer records, items inventory records, especially when the number of records is pretty large.

We want to keep information in a sensible order. It could be one of the following schemes:

- alphabetical order
- ascending/descending order
- order according to name, ID, year, department etc.

The aim of sorting algorithms is to organize the available information in an ordered form.

There are dozens of sorting algorithms. The more popular ones are listed below:

- Selection Sort
- Bubble Sort
- Insertion Sort
- Merge Sort
- Quick Sort

As we have been doing throughout the course, we are interested in finding out as to which algorithms are best suited for a particular situation. The efficiency of a sorting algorithm can be worked out by counting the number of comparisons and the number of data movements involved in each of the algorithms. The order of magnitude can vary depending on the initial ordering of data.

How much time does a computer spend on data ordering if the data is already ordered? We often try to compute the data movements, and comparisons for the following three cases:

best case (often, data is already in order), worst case(sometimes, the data is in reverse order), and average case(data in random order).

Some sorting methods perform the same operations regardless of the initial ordering of data. Why should we consider both comparisons and data movements?

If simple keys are compared, such as integers or characters, then the comparisons are relatively fast and inexpensive. If strings or arrays of numbers are compared, then the cost of comparisons goes up substantially.

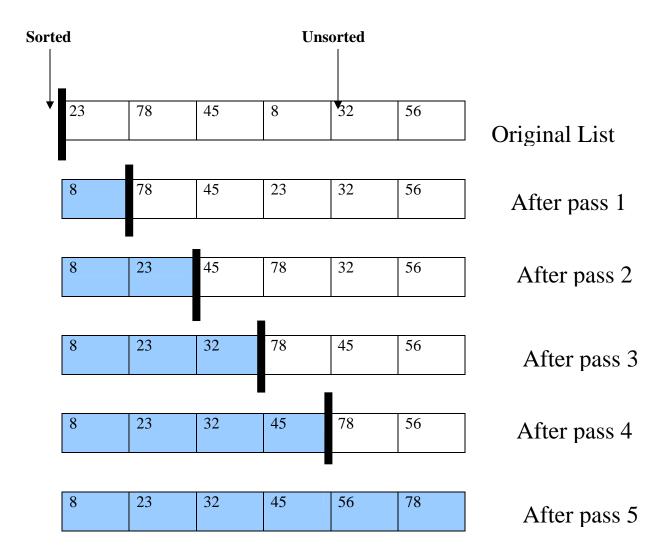
If on the other hand, the data items moved are large, such as structures, then the movement measure may stand out as the determining factor in efficiency considerations. In this section we shall discuss two efficient sorting algorithms – the merge sort and the quick sort procedures.

Selection Sort

Selection sort is an attempt to localize the exchanges of array elements by finding a misplaced element first and putting it in its final place.

- ⇒ The list is divided into two sublists, *sorted* and *unsorted*, which are divided by an imaginary wall.
- ⇒ We select the smallest element from the unsorted sublist and swap it with the element at the beginning of the unsorted data.
- ⇒ Then the smallest value among the remaining elements is selected and put in the second position and so on.
- After each selection and swapping, the imaginary wall between the two sublists move one element ahead, increasing the number of sorted elements and decreasing the number of unsorted ones.
- ⇒ Each time we move one element from the unsorted sublist to the sorted sublist, we say that we have completed a sort pass.
- \Rightarrow A list of *n* elements requires *n-1* passes to completely rearrange the data.

Selection Sort Example



Selection Sort Algorithm

```
/* Sorts by selecting smallest element in unsorted
   portion of array and exchanging it with element
   at the beginning of the unsorted list.
void selectionSort(int list[], int n)
   int cur, j, smallest, tempData;
   for (cur = 0; cur <= n; cur ++){
      smallest = cur;
          for ( j = cur +1; j <= n ; j++)
          if(list[ j ] < list[smallest])</pre>
             smallest = j ;
      // Smallest selected; swap with current element
         tempData = list[cur];
      list[cur] = list[smallest];
      list[smallest] = tempData;
   }
}
The outer loop is executed n times, and the inner loop iterates
i = (n - 1 - cur) times.
```

So it is a n times summation of j. Thus the order turns out to be $O(n^2)$.

The number of comparisons remain the same in all cases.

There may be some saving in terms of data movements (swappings).

Insertion Sort

The key idea is to pick up a data element and insert it into its proper place in the partial data considered so far. An outline of the algorithm is as follows:

Let the n data elements be stored in a array list[]. Then,

```
for (cur = 1; cur < n; cur++)
```

move all elements list [j] greater than data [cur] by one position;

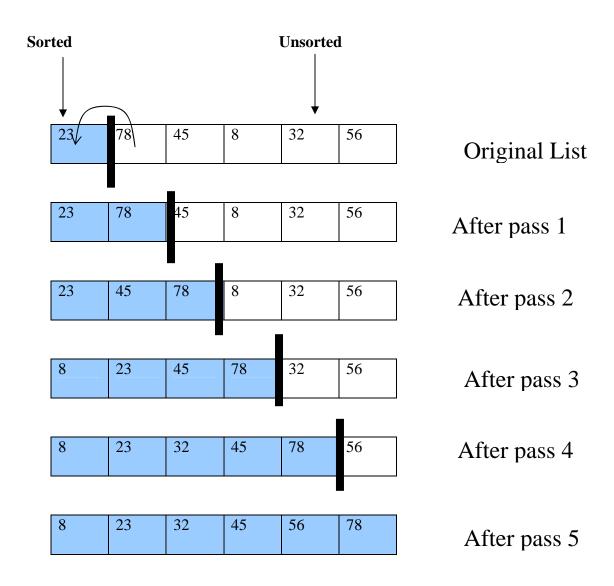
place data [cur] in its proper position.

Note that the sorting is restricted only to fraction of the array in each iteraion.

The list is divided into two parts: sorted and unsorted.

 \Rightarrow In each pass, the first element of the unsorted part is picked up, transferred to the sorted sublist, and inserted at the appropriate place. A list of n elements will take at most n-l passes to sort the data.

Insertion Sort Example



Insertion Sort Algorithm

```
/* With each pass, first element in unsorted
   sublist is inserted into sorted sublist.
*/
void insertionSort(int list[], int n)
   int cur, located, temp, j;
   for (cur = 1; cur <= n; cur++){
     located = 0;
     temp = list[cur];
     for ( j = cur - 1; j >= 0 && !located;)
        if(temp < list[ j ]){</pre>
           list[j + 1] = list[ j ];
           j--;
        else
           located = 1;
     list[j + 1] = temp;
  }
   return;
}
```

An advantage of this method is that it sorts the array only when it is really necessary.

If the array is already in order, no moves are performed.

However, it overlooks the fact that the elements may already be in their proper positions . When an item is to be inserted, all elements greater than this have to be moved. There may be large number of redundant moves, as an element (properly located) may be moved, but later brought back to its position.

The best case is when the data are already in order. Only one comparison is made for each position, so the comparison complexity is O(n), and the data movement is 2n - 1, i.e. it is O(n).

The worst case is when the data are in reverse order.

Each data element is to be moved to new position and for that each of the other elements have to be shifted. This works out to be complexity of $O(n^2)$.

What happens when the elements are in **random order**? Does the over complexity is nearer to the best case, or nearer to the worst case? It turns out that both number of comparisons and movements turn out to be closer to the worst case.

Bubble Sort

Imagine all elements are objects of various sizes and are placed in a vertical column. If the objects are allowed to float, then the smallest element will bubble its way to the top. This is how the algorithm gets its name.

⇒ The list scanned from the bottom up, and two adjacent elements are interchanged if they are found to be out of order with respect to each other.

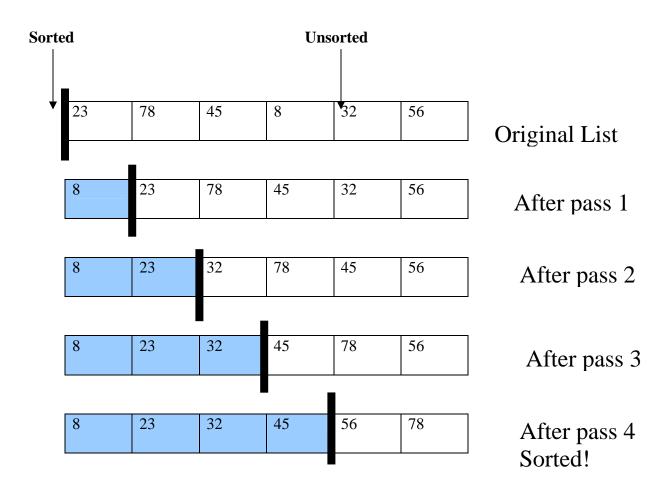
Then next pair of adjacent elements are considered and so on.

- \Rightarrow Thus the smallest element is bubbled from the unsorted list to the top of the array.
- After that, the wall moves one element ahead, increasing the number of sorted elements and decreasing the number of unsorted ones.
- ⇒ Each time an element moves from the unsorted part to the sorted part one sort pass is completed.
- ⇒ Given a list of n elements, bubble sort requires up to n-1 passes to sort the data.
- ⇒ Bubble sort was originally written to "bubble up" the highest element in the list. From an efficiency point of view it makes no difference whether the high element is bubbled or the low element is bubbled.

Bubble Sort Example

Trace of 1st pass of Bubble Sort Example:

78	45	8	32	56
78	45	8	32	56
	(
78	45	8	32	56
78	18	45	32	56
8	78	45	32	56
	<u> </u>			
23	78	45	32	56
	78	78 45 78 45 78 8 78 78	78 45 8 78 45 8 78 8 45 8 78 45	78 45 8 32 78 45 8 32 78 8 45 32 8 78 45 32



Bubble Sort Algorithm

```
/* Sorts list using bubble sort. Adjacent elements
    are compared and exchanged until list is
    completely ordered.

*/
void bubbleSort(int list[], int n)
{
    int cur, j, temp;
    for (cur = 0; cur <= n; cur++){
        for ( j = n; j > cur; j--)
            if(list[j] < list[ j - 1]){
            temp = list[ j ];
            list[ j ] = list[ j - 1];
            list[ j -1] = temp;
        }
    }
    return;
}</pre>
```

What is the time complexity of the Bubble sort algorithm?

The number of comparisons for the worst case (when the array is in the reverse order sorted) equals the number of iterations for the inner loop ,i.e. $O(n^2)$. How many comparisons are to be made for the best case? Well, it turns out that the number of comparisons remain the same.

What about the number of swaps? In the worst case, the number of movements is 3 n(n-1)/2. In the best case, when all elements are ordered, no swaps are necessary. If it is a randomly ordered array, then number of movements are around half of the worst case figure.

Compare it with insertion sort. Here each element has to be bubbled one step at a time. It does not go directly to its proper place as in insertion sort. It could be said that insertion sort is twice as fast as the bubble sort, for the average case, because bubble sort makes approximately twice as many comparisons.

In summary, we can say that bubble sort, insertion sort and selection sort are not very efficient methods, as the complexity is $O(n^2)$.

Mergesort

The *mergesort-sorting* algorithm uses the divide and conquer strategy of solving problems in which the original problem is split into two problems, with size about half the size of the original problem.

The basic idea is as follows. Suppose you have got a large number of integers to sort.

Write each integer on a separate slip of paper.

Make two piles of the slips.

So the original problem has been reduced to sorting individually two piles of smaller size.

Now reduce each pile to half of the existing size.

There would be now 4 piles with a smaller set of integers to sort.

Keep on increasing the number of piles by reducing their lengths by half every time.

Continue with the process till you have got piles with maximum two slips in each pile.

Sort the slips with smaller number on the top.

Now take adjacent piles and merge them such that resulting pile is in sorted form.

The piles would keep growing in size but now this time these are in sorted order.

Stop when all piles have been taken care of and there remains one single pile.

Thus the *mergesort* can be thought of as a recursive process. Let us assume that the elements are stored in an array.

mergesort

- 1. if the number of items to sort is 0 or 1, return.
- 2. Divide the array into two halves and copy the items in two subarrays.
- 3. Recursively mergesort the first and second halves separately.
- 4. Merge the two-sorted halves into a single sorted array.

What would be the complexity of the process?

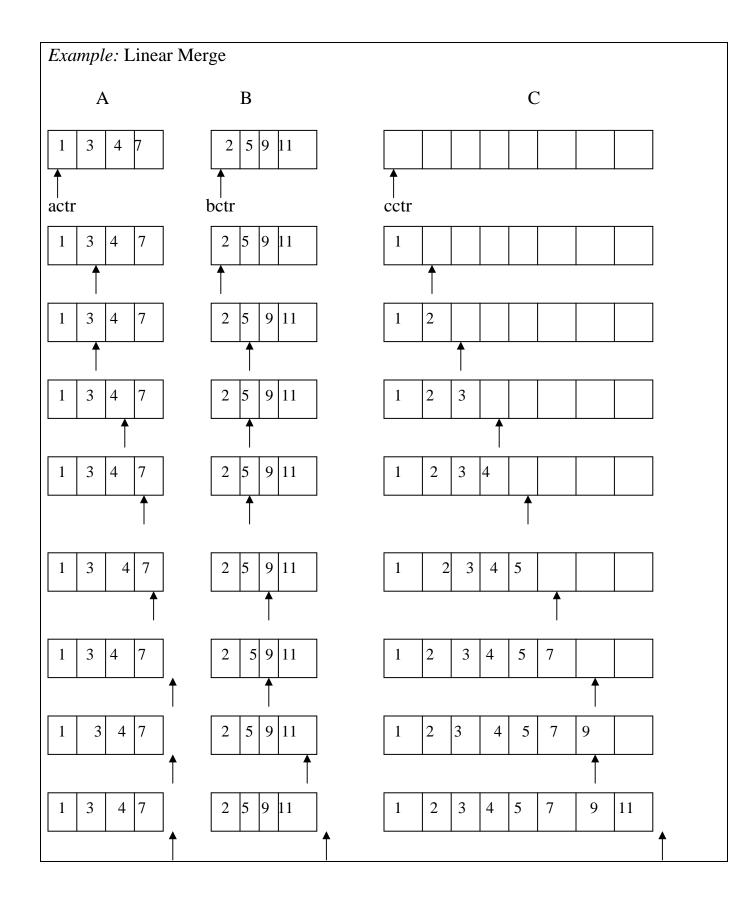
Since this algorithm uses the divide and conquer strategy and employs the halving principle, we can guess that the sorting process would have $O(\log_2 n)$ complexity. However, the merging operation would involve movement of all the n elements (linear time), and we shall show later that the overall complexity turns out to be $O(N \log_2 N)$.

We can merge two input arrays A and B to result in a third array C. Let the index counter for the respective arrays be actr, bctr, and cctr. The index counters are initially set to the position of the first element. The smaller of the two elements A[actr] and B[bctr] is stored in C[cctr] as shown below:

```
if \ A[actr] < B[bctr] \\ C[cctr] = A[actr]; \\ cctr++; \\ actr++; \\ \} \ else \ \{ \\ C[cctr] = B[bctr]; \\ cctr++; \\ bctr++; \\ \}
```

Let us take an example. Say at some point in the sorting process we have to merge two lists 1, 3, 4, 7 and 2, 5, 9, 11

We store the first list in Array A and the second list in Array B. The merging goes in following fashion:



The array is recursively split into two halves and mergesort function is applied on the two arrays separately. The arrays get sorted. Then these two arrays are merged. The mergesort and merge functions are shown below:

```
void mergesort(int array[], int n)
        int j,n1,n2,arr1[n],arr2[n];
        if (n<=1)return;</pre>
        n1=n/2;
        n2 = n - n1;
        for (j = 0; j < n1; j++)
           arr1[j]= array[ j];
        for ( j = 0; j < n2; j++ )
           arr2[j]= array[ j+n1];
        mergesort(arr1, n1);
        mergesort(arr2, n2);
        merge(array, arr1, n1, arr2, n2);
}
void merge ( int array[], int arr1[], int n1,int arr2[],
int n2)
{
        int j, p=0, p1=0, p2=0;
        while ( p1 < n1 \&\& p2 < n2 )
              if( arr1[p1] < arr2[p2 ])</pre>
                   array [p++] = arr1[p1++];
              else
                  array[p++] = arr2[p2++];
        while (p1 < n1)
                array [p++] = arr1[p1++];
        while (p2 < n2)
                array[p++] = arr2[p2++];
}
```

To sort an array A of size n, the call from the main program would be **mergesort**(A, n). Here is a typical run of the algorithm, for an array of size 8. Unsorted array A = $\begin{bmatrix} 31 & 45 & 24 & 15 & 23 & 92 & 30 & 77 \end{bmatrix}$

The original array is kept on splitting in two halves till it reduces to array of one element. Then these are merged.

```
After merging [31 ] &[45 ] merged array is [31 45 ]

After merging [24 ] & [15 ] merged array is [15 24 ]

After merging [31 45 ] & [15 24 ] merged array is [15 24 31 45 ]

After merging [23 ] & [92 ] merged array is [23 92 ]

After merging [30 ] & [77 ] merged array is [30 77 ]

After merging [23 92 ] & [30 77 ] merged array is [23 30 77 92 ]

After merging [15 24 31 45 ] & [23 30 77 92 ]

merged array is [15 23 24 30 31 45 77 92 ]
```

Computational Complexity:

Intuitively we can see that as the **mergesort** routine reduces the problem to half its size every time, (done twice), it can be viewed as creating a tree of calls, where each level of recursion is a level in the tree. Effectively, all n elements are processed by the **merge** routine the same number of times as there are levels in the recursion tree. Since the number of elements is divided in half each time, the tree is a balanced binary tee. The height of such a tree tends to be log n.

The **merge** routine steps along the elements in both halves, comparing the elements. For n elements, this operation performs n assignments, using at most n-1 comparisons, and hence it is O(n). So we may conclude that $[\log n \cdot O(n)]$ time merges are performed by the algorithm.

The same conclusion can be drawn more formally using the method of recurrence relations. Let us assume that n is a power of 2, so that we always split into even halves. The time to **mergesort** n numbers is equal to the time to do two recursive mergesorts of size n/2, plus the time to **merge**, which is linear.

For n = 1, the time to mergesort is constant.

We can express the number of operations involved using the following recurrence relations:

$$T(1) = 1$$

 $T(n) = 2 T(n/2) + n$

Using same logic and going further down

$$T(n/2) = 2 T(n/4) + n/2$$

Substituting for T(n/2) in the equation for T(n), we get

$$T(n) = 2[2 T(n/4) + n/2] + n$$

= $4 T(n/4) + 2n$

Again by rewriting T(n/4) in terms of T(n/8), we have

$$T(n) = 4 [2 T(n/8) + n/4] + 2n$$

$$= 8 T(n/8) + 3 n$$

$$= 2^{3} T(n/2^{3}) + 3 n$$

The next substitution would lead us to

$$T(n) = 2^4 T(n/2^4) + 4 n$$

Continuing in this manner, we can write for any k,

$$T(n) = \ 2^k \ T(n/\ 2^k\) + k \ n$$

This should be valid for any value of k. Suppose we choose k = log n, i.e. $2^k = n$. Then we get a very neat solution:

$$T(n) = n T(1) + n \log n$$
$$= n \log n + n$$

Thus
$$T(n) = O(n \log n)$$

This analysis can be refined to handle cases when n is not a power of 2. The answer turns out to be almost identical.

Although mergesort's running time is very attractive, it is not preferred for sorting data in main memory. The main problem is that merging two sorted lists uses linear extra memory (as you need to copy the original array into two arrays of half the size), and the additional work spent copying to the temporary array and back, throughout the algorithm, has the effect of slowing down the sort considerably.

The copying can be avoided by judiciously switching the roles of list and temp arrays at alternate levels of the recursion. For serious internal sorting applications, the algorithm of choice is the Quicksort, which we shall be studying next.

Quicksort

As the name implies the quicksort is the fastest known sorting algorithm in practice. It has the best average time performance. Like merge sort, Quicksort is also based on the *divide-and-conquer* paradigm.

- But it uses this technique in a somewhat opposite manner, as all the hard work is done *before* the recursive calls.
- It works by partitioning an array into two parts, then sorting the parts independently, and finally combining the sorted subsequences by a simple concatenation.

In particular, the quick-sort algorithm consists of the following three steps:

1. Choosing a pivot::.

To partition the list, we first choose some element from the list which is expected to divide the list evenly in two sublists. This element is called a *pivot*.

2. Partitioning:

Then we partition the elements so that all those with values less than the pivot are placed in one sublist and all those with greater values are placed in the other sublist.

3. Recur:

Recursively sort the sublists separately. Repeat the partition process for both the sublists. Choose again two pivots for the two sublists and make 4 sublists now. Keep on partitioning till there are only *one* cell arrays that do not need to be sorted at all. By dividing the task of sorting a large array into two simpler tasks and then dividing those tasks into even simpler task, it turns out that in the process of getting prepared to sort, the data have already been sorted. This is the core part of the quicksort. The steps involved in the quicksort algorithm are explained through an example.

Example array:

- 1. Choose first element 56 as pivot.
- 2. Move rh index to left until it coincides with lh or points to value smaller than pivot. In this example it already points to value smaller than the pivot.
- 3. Move lh index to right until it coincides with rh or points to value equal to or greater than pivot.

[56	25	37	58	95	19	73	30]
			lh				rh

4. If lh and rh not pointing to same element, exchange the elements.

5. Repeat steps 2 to 4 until lh and rh coincide.

move lh to right till it finds an element larger than 56

Now exchange

move rh to left

No exchange now, as both lh and rh point to the same element

6. This will be the smallest value. Exchange this with the pivot.

7. This will result in two sub arrays, one to left of pivot and one to right of pivot.

8. Repeat steps 1 to 6 for the two sub arrays recursively.

Here is the code for the quick sort method:

```
void sorting(int values[] , int n)
        int mid;
        if(n < 2)
        return;
        mid = partitioning( values, n);
        sorting(values, mid);
        sorting(values+ mid+1 ,     n-mid-1);
  }
int partitioning( int values[],int n)
    int pivot, left, right, temp;
    pivot= values[0];
    left=1 ;
    right= n-1;
    while(1)
       while(left<right && values[right]>= pivot)
                   right--;
       while(left<right && values[left]< pivot)</pre>
                 left++;
       if( left == right)
                break;
       temp= values[left];
       values[left] = values[right];
       values[right] = temp;
    if(values[left] >= pivot)
          return 0;
    values[0] = values[left];
    values[left] = pivot;
    return left;
}
```

Picking the Pivot:

The algorithm would work, no matter which element is chosen as a pivot. However, some choices are going to be obviously better than the other ones. A popular choice would be to use the first element as a pivot. This may work if the input is random.

But, if the list is presorted or in reverse order, then what would be the result of choosing such a pivot? This may consistently happen throughout the recursive calls and turn the whole process into a quadratic time algorithm.

If the data is presorted, the first element is not chosen as the pivot element. A good strategy would be to take the first value and swap it with the center value, and then choose the first value as a pivot. A better strategy is to take the median of the first, last and the center elements. This works pretty well in many cases.

Analysis of Quicksort:

To do the analysis let us use the recurrence type relation used for analyzing mergesort. We can drop the steps involved in finding the pivot as it involves only constant time. We can take T(0) = T(1) = 1.

The running time of quicksort is equal to the running time of the two recursive calls, plus the linear time spent in the partition. This gives the basic quicksort relation

T(N) the time for Quicksort on array of N elements, can be given by

$$T(N) = T(i) + T(N-i-1) + N,$$

Where j is the number of elements in the first sublist.

Best case analysis:

The best case analysis assumes that the pivot is always in the middle. To simplify the math, we assume that the two sublists are each exactly half the size of the original. Then we can follow the same analysis as in merge sort and can show that

$$T(N) = T(N/2) + T(N/2) + 1$$

leads to

$$T(N) = O(N \log N)$$

Worst Case Analysis:

The partitions are very lopsided, meaning that either left partition |L| = 0 or N - 1 or right partition |R| = N - 1 or 0, at each recursive step. Suppose that Left contains no elements, Right contains all of the elements except the pivot element (this means that the pivot element is always chosen to be the smallest element in the partition),

1 time unit is required to sort 0 or 1 elements, and N time units are required to partition a set containing N elements.

Then if N > 1 we have:

$$T(N) = T(N-1) + N$$

This means that the time required to quicksort N elements is equal to the time required to recursively sort the N-1 elements in the Right subset plus the time required to partition the N elements. By telescoping the above equation we have:

$$T(N) = T(N-1) + N$$

$$T(N-1) = T(N-2) + (N-1)$$

$$T(N-2) = T(N-3) + (N-2)$$
.....
$$T(2) = T(1) + 2$$

$$T(N) = T(1) + 2 + 3 + 4 + ... + N$$

$$= N(N+1)/2$$

$$= O(N^2)$$

This implies that whenever a wrong pivot is selected it leads to unbalanced partitioning.