

Algorithm and Data Structures HW 2

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1 Merge Sort

1.1 Merge Sort Variant

```
void mergeSort(int arr[], int sub, int l, int r)
{
    //sub is the size k of the sub array
    //in witch we will applied insertion sort
    int size = r - l + 1 ;

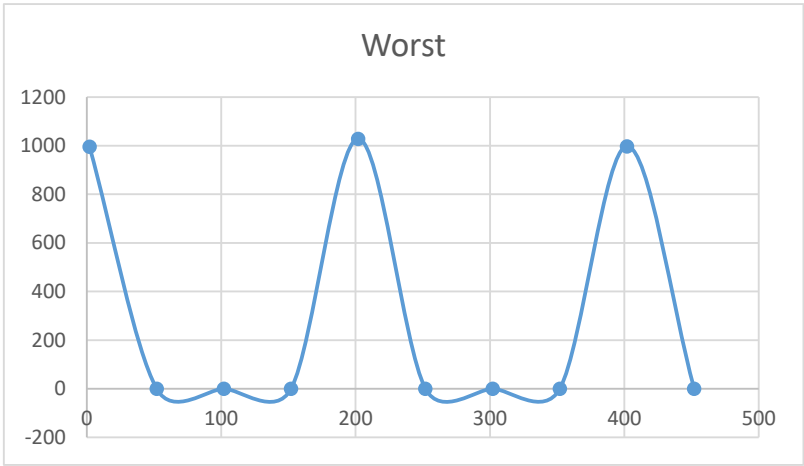
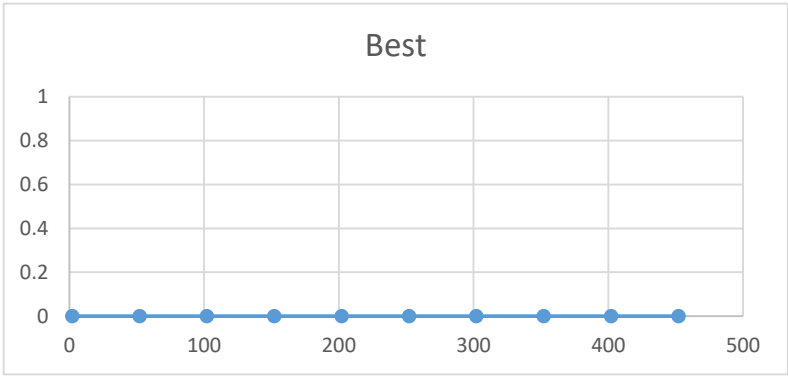
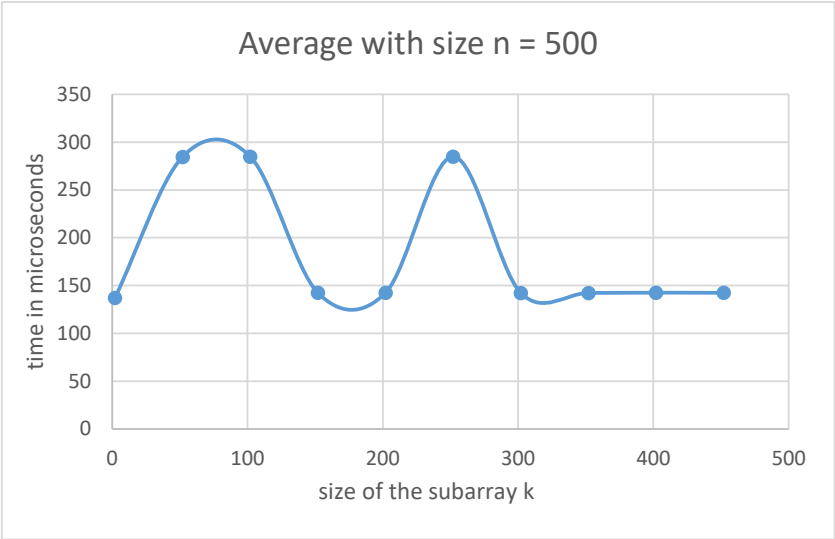
    if(sub <= size){
        insertion_sort(arr, l, r);
    }
    else
    {
        int m = l+(r-l)/2;

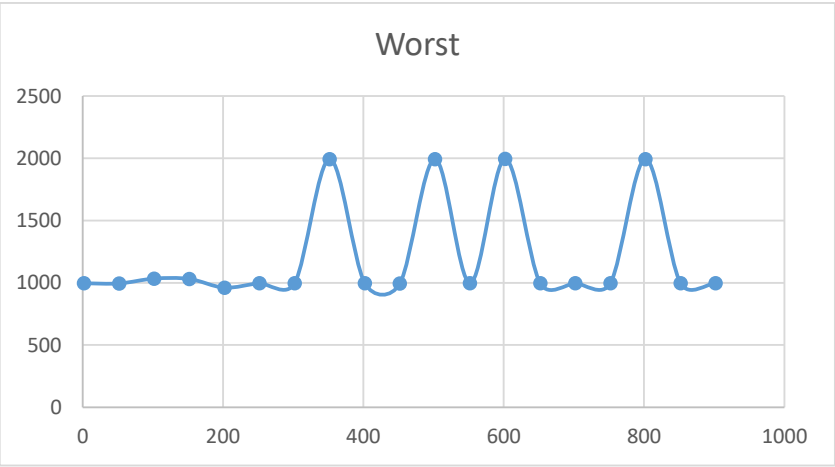
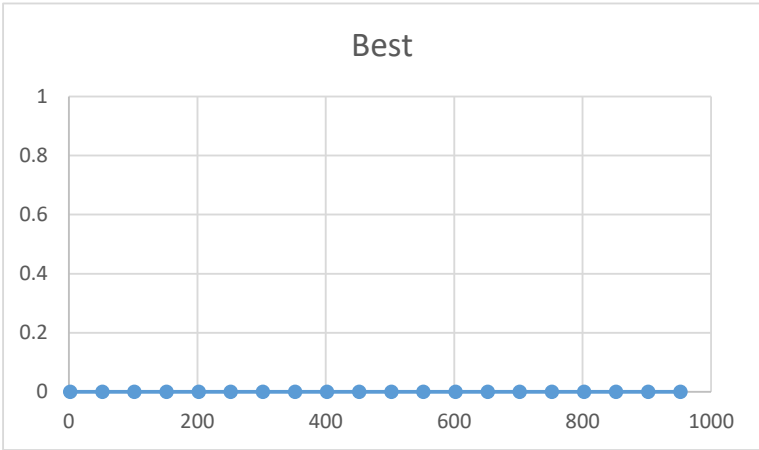
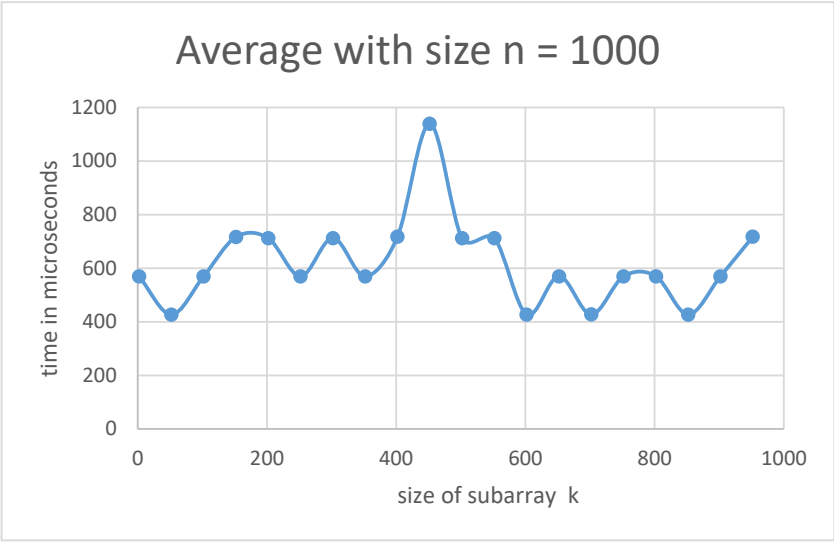
        mergeSort(arr, sub, l, m);
        mergeSort(arr, sub, m+1, r);

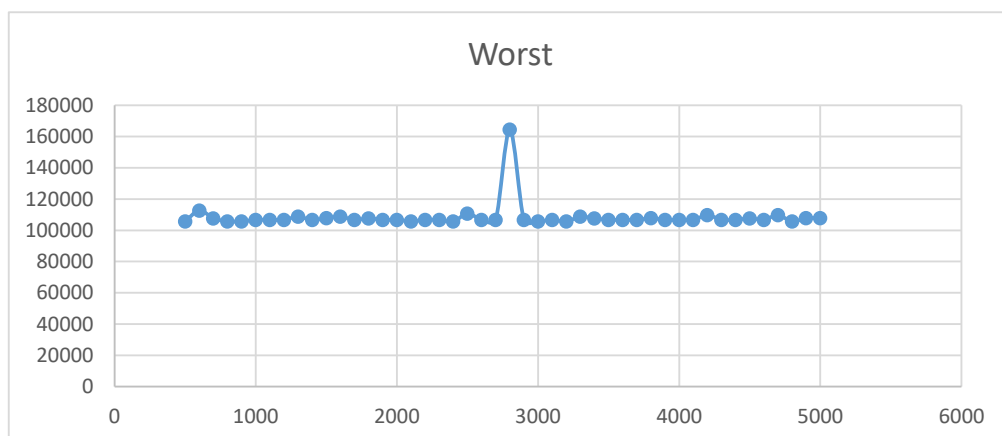
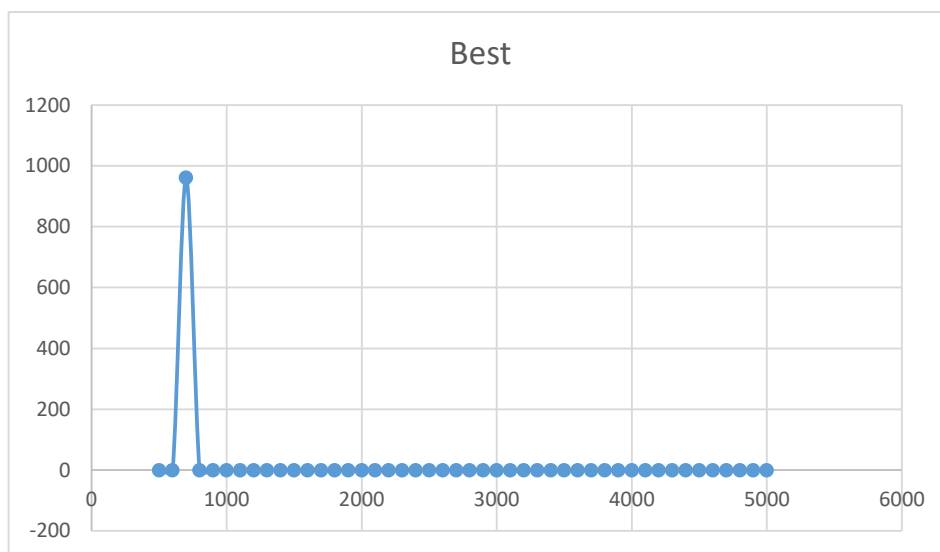
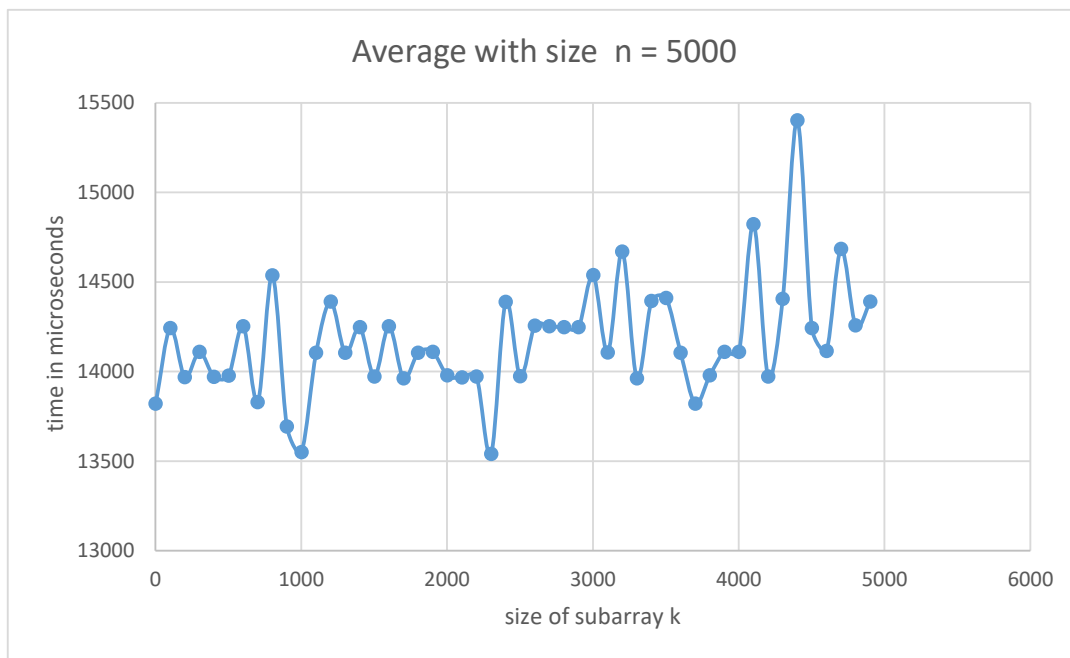
        merge(arr, l, m, r);
    }
}
```

1.2 Plot of running time of the algorithm

I am using 3 different graphs for each case, each graph represent the sorting algorithm applied on an array that have a size constant $500 < n < 5000$ with $2 < k < n$







1.3 Graph analysis

As we can see on the graphs of the average cases in most of them when k increases is not a predictable tendency of the running time but when k is bigger the algorithm works more as an insertion sort algorithm instead of mergesort but when k is small is more like a mergesort.

For the Best case the algorithm is really lineal because as we know insertion sort in the best case it has a lineal order but in some cases when n is really big and k is between $n/2$ sometimes the time has variations so k does not affect anything at all.

Worst Case is the most interesting the peaks of the worst case graph is when k gets higher the algorithm works as a insertion sort being $O(n^2)$ compare to the mergesort $O(n \lg n)$ is slightly slower in those parts.

1.4 How to chose k

In practice from my point of view the algorithm k should have a value between 2 and 30 because is when the insertion sort is pretty fast and makes the merge way simpler so you do not have the necessity of doing to many swaps

2 Recurrences

2.1 A

$$T(n) = 36T(n/6) + 2n$$

$$a = 36 \quad b = 6$$

$$n^{\log_6 36} = n^2$$

$$F(n) \in O(n^{2-\epsilon}) \text{ for } \epsilon = 1$$

$$\text{thus } T(n) = \Theta(n^2)$$

2.2 B

$$T(n) = 5T(n/3) + 17n^{1.2}$$

$$a = 5 \quad b = 3$$

$$n^{\log_3 5}, \log_3 5 > 1.2$$

$$\text{and we have that } \log_3 5 - 1.2 > 0$$

$$F(n) \in O(n^{\log_3 5 - \epsilon}) \text{ for } \epsilon = (\log_3 5 - 1.2)$$

$$\text{thus } T(n) = \Theta(n^{\log_3 5})$$

2.3 C

$$T(n) = 12T(n/2) + n^2 \lg(n)$$

$$a = 12 \quad b = 2$$

$$n^{\log_2 12}, \quad n > \log_2 n$$

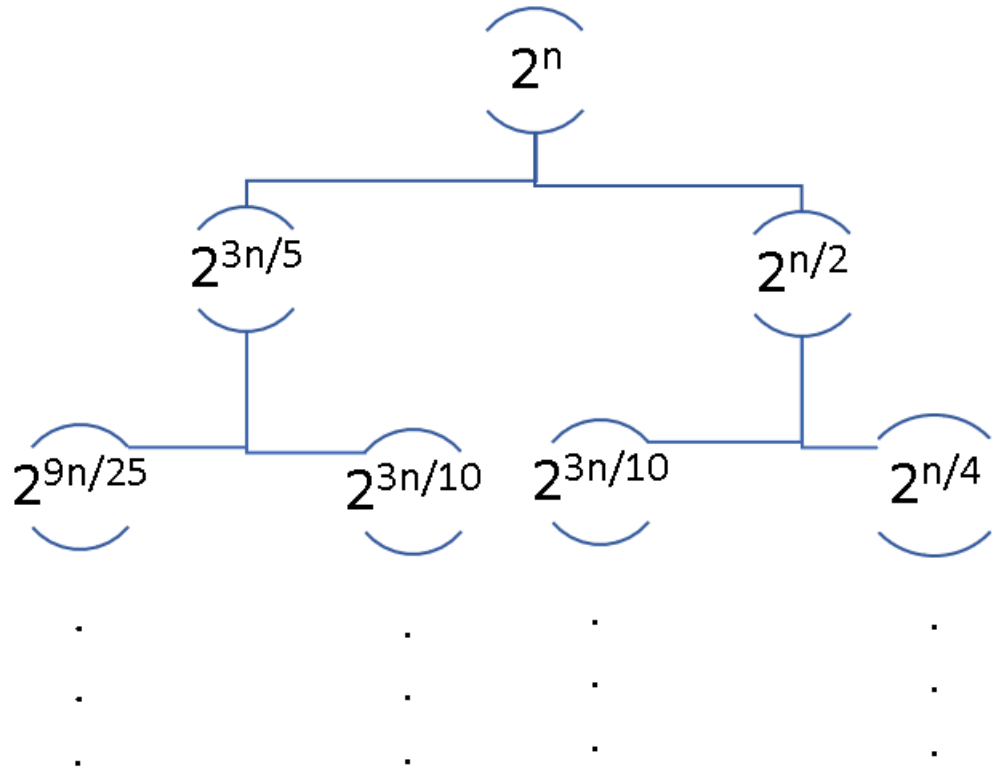
$$F(n) \in O(n^{\log_2 12 - \epsilon}) \text{ for } \epsilon = 1$$

$$\text{thus } T(n) = \Theta(n^{\log_2 12})$$

2.4 D

$$T(n) = 3T(n/5) + T(n/2) + 2^n$$

for this problem I am using recursion tree



for the first 3 levels we got the following sums

$$2^n + 2^{3n/5} + 2^{9n/25} + 2^{3n/10} + 2^{3n/10} + 2^{n/4} + \dots$$

we take common factor n^2 we get the following

$$2^n (1 + 2^{(-2n/5)} + 2^{(-16n/25)} + 2^{(-7n/10)} + 2^{(-7n/10)} + 2^{(-3n/4)} + \dots)$$

we are interested when n goes to infinity so we get the following

$$2^n * (1 + 0 + 0 + 0 + 0 + 0 + 0 + 0 + \dots)$$

$$2^n$$

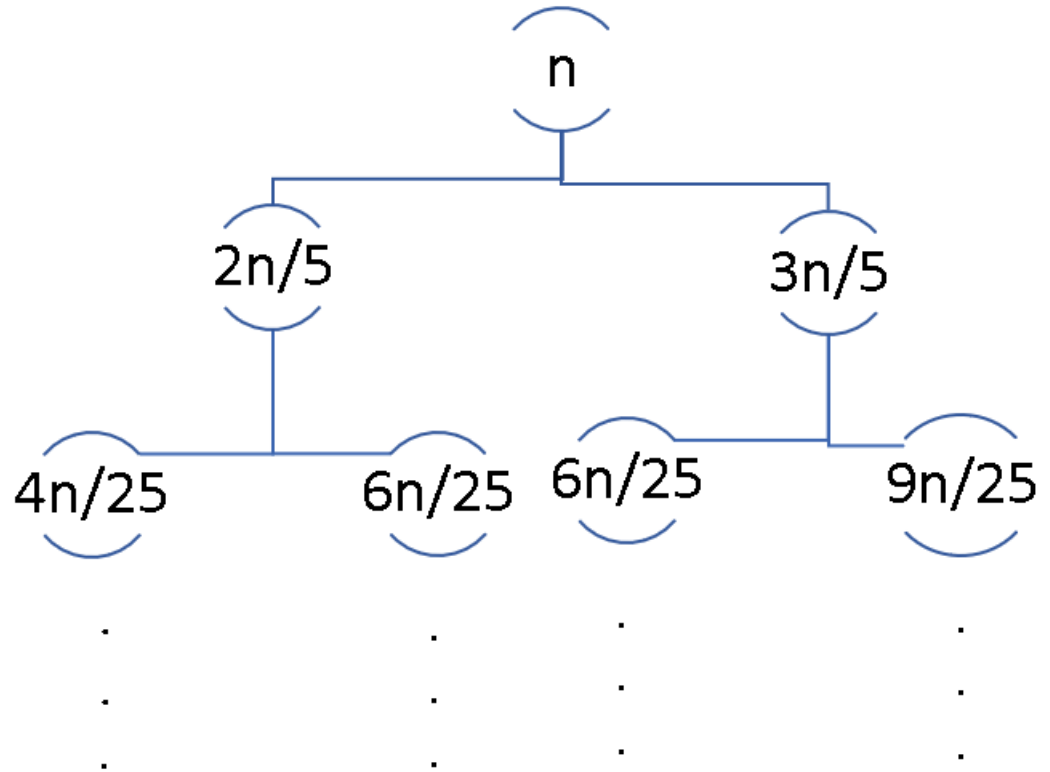
thus

$$T(n) \in \Theta(2^n)$$

2.5 E

$$T(n) = T(2n/5) + T(3n/5) + \Theta(n)$$

we can assume that $\Theta(n)$ is a representation of how the $f(n)$ behaves so I take $f(n) = cn$ being c a constant but at the end it does not affect the overall order so I will use $f(n) = n$ so I will start my recursion tree using n as my root



for the first 3 levels we got the following sums

$$n + n + n + \dots$$

as the tree continues growing we see that is a sum of n and as we can know the high of the tree because it divide in 5 branches every time the high is $\log_5 n$ thus

$$T(n) \in \Theta(n \log n)$$