

Analysis Report

Question 1:

The time complexity in the iterative method is $O(n)$ as it has a loop that runs n time so the asymptotic big theta notation is $\theta(n)$

However, in the divide and conquer method the time complexity is $O(\log(n))$ so the asymptotic big theta notation is $\theta(\log n)$ as it constantly divides the problem into halves(subproblems)

Recurrence of divide and conquer:

$$T(n) = aT(n/b) + f(n)$$

Since we only call (recur) on the function on time inside it then $a = 1$ and when we call it we divide the result into half therefore the $b = 2$. Then the remaining function is the best case or when the 'y' is either odd or even

$$T(n) = T(n/2) + O(1)$$

To solve the recurrence: using Master theorem

$$a = 1, b = 2,$$

$$f(n) = O(1)$$

$$n^{(\log b \ a)} = n^{(0)} = 1$$

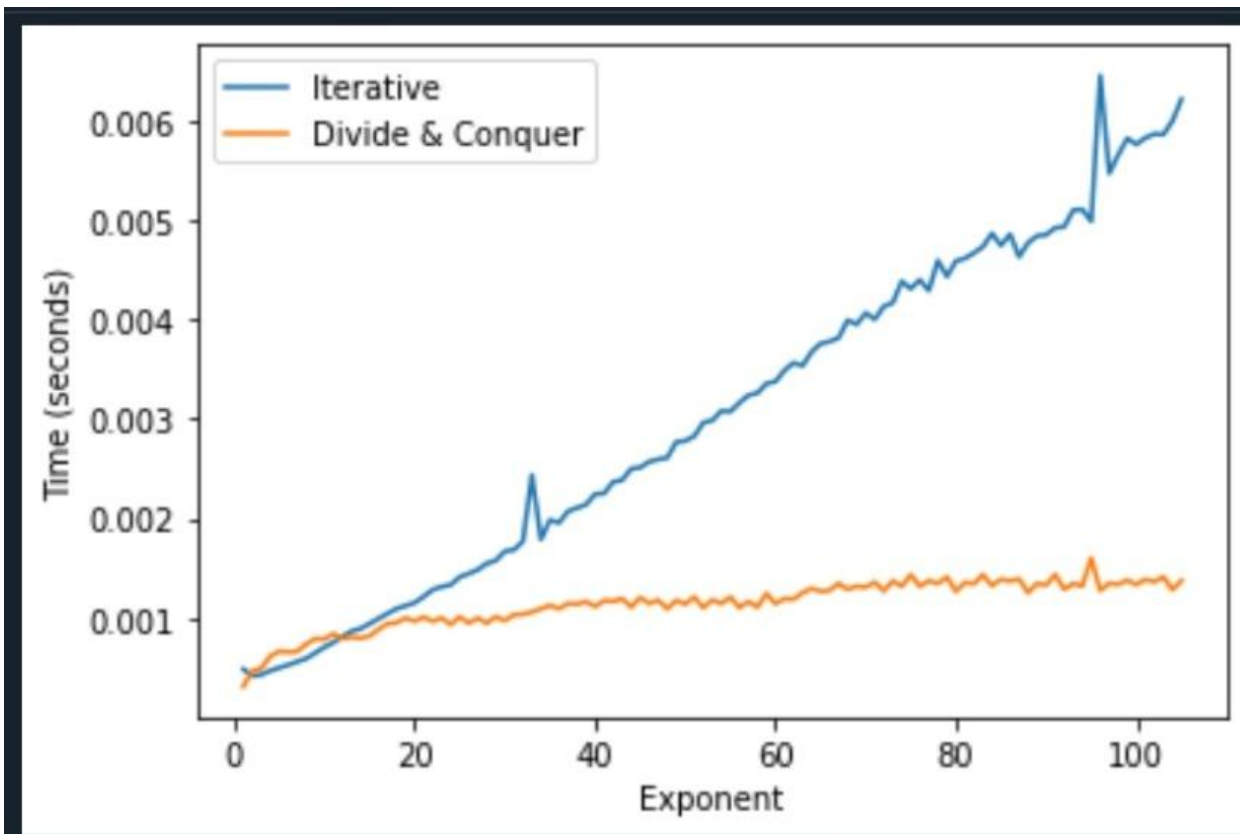
$$\text{therefore } f(n) = n^{(\log b \ a)}$$

then we will use case 2

$$\text{then } T(n) = \theta(n^{(\log b \ a \log n)})$$

$$T(n) = \theta(n^{(0) \log n}) = \theta(\log n)$$

-The empirically observed running time matches the running time predicted by the analysis as shown in the graph below



Question 2:

The two major methods here to determine the time complexity is the merge sort and binary search where binary search time complexity takes $O(\log(n))$ in the worst case however the merge sort takes $O(n\log(n))$ so the time complexity of the whole function is $O(n\log(n))$ so the asymptotic big theta notation is $\theta(n\log(n))$

Recurrence of divide and conquer:

$$T(n) = aT(n/b) + f(n)$$

Since the slower function is merge sort we will have its values;

We call it 2 times (recur) in the function so $a = 2$

And we call only its half so $b = 2$

While the remaining of the function calls loops on a value therefore $O(n)$

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

To solve the recurrence: using Master theorem

$$a = 2, b = 2,$$

$$f(n) = O(n)$$

$$n^{(\log_b a)} = n^{(\log_2 2)} = n^{(1)} = n$$

$$\text{therefore } f(n) = n^{(\log_b a)}$$

then we will use case 2

$$\text{then } T(n) = \theta(n^{(\log_b a)} \log n)$$

$$T(n) = \theta(n^{(1)} \log) = \theta(n \log n)$$

-The empirically observed running time matches the running time predicted by the analysis as shown in the graph below

