

Q1)

Rule The Master Theorem applies to recurrences of the following form:

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

where  $a \geq 1$  and  $b > 1$  are constants and  $f(n)$  is an asymptotically positive function.

There are 3 cases:

- 1) If  $f(n) = O(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$
- 2) If  $f(n) = \Theta(n^{\log_b a} \log^k n)$  with  $k \geq 0$ , then  $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$
- 3) If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  with  $\epsilon > 0$ , and  $f(n)$  satisfies the regularity condition, then  $T(n) = \Theta(f(n))$ .

Regularity condition:  $a f(n/b) \leq c f(n)$  for some constant  $c < 1$  and all sufficiently large  $n$ .

a)  $T(n) = 16T(n/4) + n!$

$a = 16$   $b = 4$   $f(n) = n!$

Case 3

$\Omega(n^{\log_4 16 + \epsilon}) \rightarrow n^{\log_4 16} = n^2$

$n! > n^2$

$T(n) = \Theta(f(n)) = \Theta(n!)$

b)  $T(n) = \sqrt{2} T(n/2) + \log n$

$a = 2^{1/2}$

$b = 2$

$f(n) = \log n$

$n^{\log_2 a} = n^{\log_2 \sqrt{2}} = n^{0.25}$

$\log n > n^{0.25}$

Case 1

$T(n) = \Theta(n^{\log_2 a}) = \Theta(n^{\log_2 \sqrt{2}}) = \Theta(n^{0.25})$

c)  $T(n) = 8T(n/2) + 4n^3$

$a = 8$   $b = 2$   $f(n) = 4n^3$

Case 2

$n^{\log_2 8} = n^{\log_2 8} = n^3$

$T(n) = \Theta(n^{\log_2 a} \log n)$

$T(n) = \Theta(n^3 \cdot \log n)$



$$d) T(n) = 64T\left(\frac{n}{8}\right) - n^2 \log n$$

Theorem doesn't apply because  $f(n)$  that is negative

$$e) T(n) = 3T\left(\frac{n}{3}\right) + \sqrt{n}$$

$$a = 3$$

$$b = 3$$

$$c = \frac{1}{2}$$

$$f(n) = \sqrt{n}$$

$$\log_a b = \log_3 3 = 1$$

Case 1

$$T(n) = \Theta(n^{\log_a b}) = \Theta(n)$$

$$f) T(n) = 2^n T\left(\frac{n}{2}\right) - n^n$$

$$a = 2^n \rightarrow \text{not constant}$$

$$b = 2$$

$$f(n) = -n^n \rightarrow f(n) \text{ is not asymptotically positive function}$$

Theorem does not apply

$$g) T(n) = 3T\left(\frac{n}{3}\right) + \frac{n}{\log n}$$

$$a = 3$$

$$b = 3$$

$$f(n) = n^0 (\log n)^{-1}$$

$$d = 1$$

$$a = b^d \Rightarrow 3 = 3^1$$

So, Case 2

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



Q2) What are the running times of each of these algorithms (in big-O notation), and which would you choose?

Rule)

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n) \quad a \rightarrow \text{number of subproblems}$$

$$\frac{n}{b} \rightarrow \text{size of the subproblems}$$

a) Algorithm x

$$T(n) = 9T\left(\frac{n}{3}\right) + n^2 \rightarrow \text{quadratic time}$$

Let's using master theorem:

$$\begin{matrix} a=9 \\ b=3 \end{matrix} \quad f(n)=n^2 \quad \Rightarrow \quad n^{\log_3 9} = n^{\log_3 3^2} = n^2$$

case 2

$$f(n) = \Theta(n^{\log_3 9}) \quad T_n = \Theta(n^2 \log n)$$

b) Algorithm y

$$T(n) = 8T\left(\frac{n}{2}\right) + n^3$$

Let's using master theorem:

$$\begin{matrix} a=8 \\ b=2 \end{matrix} \quad n^{\log_2 8} = n^3$$

$$f(n) = n^3$$

case 2

$$T(n) = \Theta(n^{\log_2 8} \cdot \log n)$$

$$= \Theta(n^3 \cdot \log n)$$

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C) Algorithm  $\mathbb{Z}$

$$T(n) = 2 \cdot T\left(\frac{n}{4}\right) + \sqrt{n}$$

$$a=2$$

$$b=4$$

$$f(n) = \sqrt{n}$$

$$n \log_b a = n \log_4 2 = \sqrt{n}$$

Case 2

$$T(n) = \Theta(n^{\log_b a} \cdot \log n)$$

$$= \Theta(\sqrt{n} \cdot \log n)$$

Running times of algorithm

compare algorithm  $x$  and algorithm  $y$

$$\lim_{n \rightarrow \infty} \frac{n^2 \log n}{n^3 \log n} = \frac{1}{n} = 0 \rightarrow \text{algorithm } x \text{ faster than } y$$

compare algorithm  $x$  and algorithm  $\mathbb{Z}$

$$\lim_{n \rightarrow \infty} \frac{n^2 \log n}{n \cdot \log n} = \frac{n^2}{n} = n = \infty \rightarrow \text{algorithm } \mathbb{Z} \text{ faster than } x$$

compare algorithm  $y$  and  $\mathbb{Z}$  =

$$\lim_{n \rightarrow \infty} \frac{n^3 \log n}{n \cdot \log n} = \frac{n^3}{n} = n^2 = \infty \rightarrow \text{algorithm } \mathbb{Z} \text{ faster than algo } y$$

Result,

According to fast = algo  $\mathbb{Z}$  > algo  $x$  > algo  $y$

I would prefer Algorithm  $\mathbb{Z}$  because it is the fastest algorithm.



Q3)

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a)

Rule)

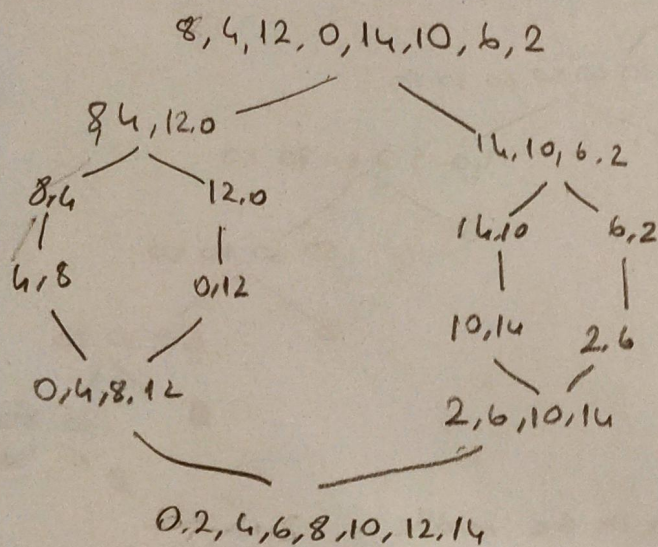
Merge sort is a divide and conquer algorithm. It divides the input array into two halves, calls itself for two halves and then merges the two sorted halves.

Time complexity = Best case =  $\Omega(n \log n)$

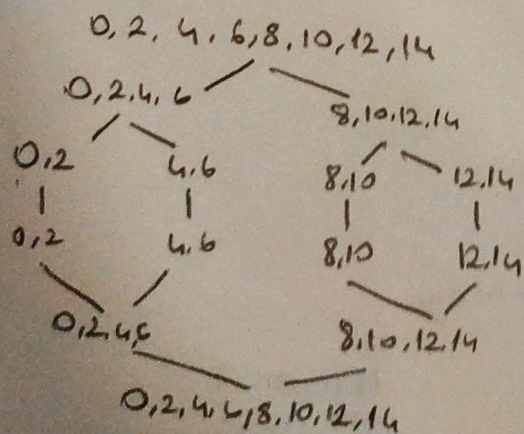
Average case =  $\Theta(n \log n)$

Worst case =  $O(n \log n)$

i) Merge Sort maximum comparison = All pair comparisons must be in the wrong order,  
worst case



ii) Merge Sort minimum comparison = All pair comparisons therefore pairs are always sorted.  
best case



Note = Best case is the same as worst case because comparison must occur at all costs.



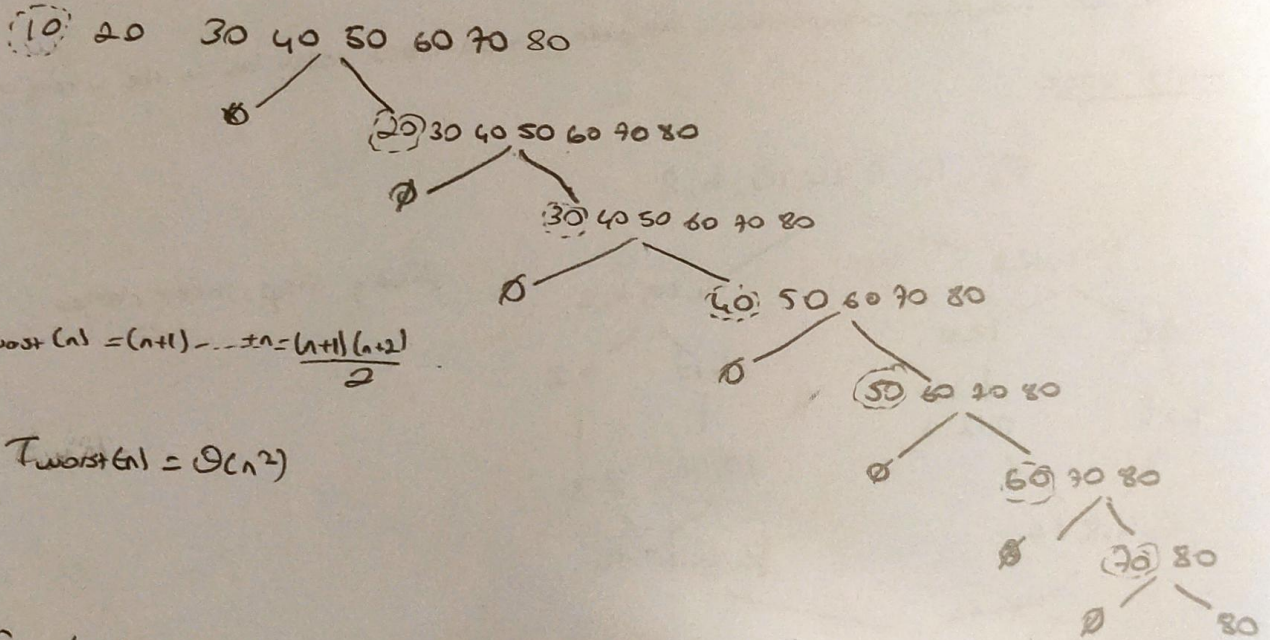
b) Rule QuickSort is a Divide and Conquer algorithm. It picks an element as pivot and partitions the given array around the picked pivot.

Time complexity = Best case =  $\Omega(n \log n)$

Average case =  $\Theta(\log n)$

Worst case =  $O(n^2)$  and space

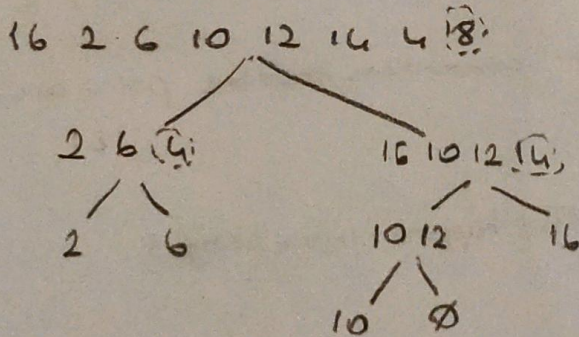
1) For worst case the list is already sorted in either ascending or descending order. So there is nothing to compare to in one side



$$T_{\text{worst}}(n) = (n+1) + \dots + n = \frac{(n+1)(n+2)}{2}$$

$T_{\text{worst}}(n) = \mathcal{O}(n^2)$

ii] For best case split must happen in the middle



$$T_{\text{best}}(n) = 2 T_{\text{best}}\left(\frac{n}{2}\right) + n, \quad T_{\text{best}}(n) = n \log n$$

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O<sub>2</sub> on GFC/lin



Q4) algorithm(left, right)  
 mid = (left + right) / 2  
 if A[mid] == 0  
   return mid  
 else  
   if A[mid] > 0  
   right = mid  
   algorithm(left, right)  
   else  
   left = mid  
   algorithm(left, right)

This algorithm is Binary Search algorithm. Search a sorted array by repeatedly dividing the search interval in half. Begin with an interval covering the whole array. If the value of the search key is less than the item in the middle of the interval, narrow the interval to the lower half. Otherwise, narrow it to the upper half. Repeatedly check until the value is found or the interval is empty.

This algorithm:

1. check A[mid] == 0
2. check A[mid] > 0, delete the upper half
3. check A[mid] < 0, delete the lower half

Thus the recurrence relation  $\rightarrow T(n) = T\left(\frac{n}{2}\right) + 1 \rightarrow$  firstly step

$$T\left(\frac{n}{2^{k-1}}\right) = T\left(\frac{n}{2^k}\right) + k \rightarrow k^{\text{th}} \text{ step } \log_2 \text{ step } n=1$$

$$n = 2^k \rightarrow \log n = k$$

$$T(n) = T\left(\frac{n}{2^{\log_2 n}}\right) + \log n$$

base case

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

$$\text{So, } T(n) = O(\log n)$$



a) Algorithm in pseudocode

matchs (gifts, boxes, low, high)

if low < high

    pivot = partition (gifts, boxes, low, high, boxes[high])

    partition (boxes, low, high, gifts[pivot])

    match (gifts, boxes, low, pivot-1)

    match (gifts, boxes, pivot+1, high)

partition (arr, low, high, pivot)

    i = low

    j = low

while (j < high)

    if (arr[j] < pivot)

        arr[i], arr[j] = arr[j], arr[i]

        i++

    elif (arr[j] == pivot)

        arr[j], arr[high] = arr[high], arr[j]

        j++

    j++

arr[i], arr[high] = arr[high], arr[i]

return [i]

b) This algorithm firstly a partition by selecting the last element of the boxes array as the pivot, rearranges the gift array, and rotates the partition index 'i' so that all gifts smaller than gifts[i] are on the left side and all gifts are larger than gifts[i] is on the right. We can then segment the boxes array using gifts[i]

This process also makes the string of gifts and boxes nicely segmented.

When we divide both gifts and boxes, the total time complexity will be

Recurrence Relation

$$T(0) = T(1)$$

$$T(N) = 2T\left(\frac{N}{2}\right) + N$$

$$\frac{T(N)}{N} = \frac{N}{N} + 2\frac{T(N/2)}{N}$$

$$\frac{T\left(\frac{N}{n-2}\right)}{\frac{N}{n-2}} = 1 + \frac{T\left(\frac{N}{N}\right)}{\frac{N}{N}} = 1 + T(1)$$

$$\frac{T(N)}{N} = 1 + 1 + 1 + \dots + 1 = \log N$$

$$\frac{T(N)}{N} = \log N$$

$$\underline{\underline{T(N) = N \log N}}$$

$$O(N \log N)$$