

COMP 3804 Assignment 1:

①

$$\log(\log(n)) < \frac{\log(n)}{\log(\log n)} < \log n < (\log n)^3 < n^{2/3} < 2^{\log n} < n^{3/2} < 2^n < n! < n^n$$

Smallest

→ largest

$$\textcircled{2} \quad \log(n!) = \log(1 \cdot 2 \cdot 3 \cdots n) = \log(1) + \log(2) + \log(3) + \dots + \log(n-3) + \log(n-2) + \log(n-1) + \log(n)$$

$$\leq \log(n) + \log(n) \dots \log(n) = n \log(n)$$

To understand the above:

- $\log(n!)$ = the result of the factorial of all n's logged
- This results in: $\log(1) + \log(2) \dots \log(n)$

- The upperbound of n minus an Integer is always n, thus the upperbound of the factorial is the $\log(n)$ added to itself n times

- Multiplications can be thought of as repeated additions, we can represent $\log(n)$ added to itself n times as $n \log(n)$

$$\therefore \log(n!) \leq \Theta(n \log(n)) \text{ for all } n > 0$$

$$\begin{aligned} \log(n!) &= \log(1 \cdot 2 \cdot 3 \cdots n) = \log(1) + \log(2) + \log(3) + \dots + \log(n-2) + \log(n-1) + \log(n) \\ &= \log(1) + \log(2) + \log(3) + \dots + \log(\frac{n}{2}) + \dots + \log(n-2) + \log(n-1) + \log(n) \\ &\geq \log(\frac{n}{2}) + \log(\frac{n-1}{2}) + \dots + \log(\frac{1}{2}) + \log(n) = \frac{n}{2} \log(\frac{n}{2}) \end{aligned}$$

thus $\log(n!) \geq \frac{n}{2} \log(\frac{n}{2})$ by replacing all terms w/ their lower bounds

$$\log(n!) \geq \frac{n}{2} \log(\frac{n}{2})$$

$$= \frac{n}{2}(\log(n) - \log(2))$$

$$= \frac{n}{2}(\log(n) - 1)$$

$$= \frac{n}{2} \log(n) - \frac{n}{2} \rightarrow \text{Prove: } \frac{n \log n - n}{2} \geq c n \log n$$

$$[\log \text{ Rule: } \log_2 \frac{M}{N} = \log_2 M - \log_2 N]$$

For $n \geq 4$:

$$2 \leq \log n$$

$$[\log_2(4) = 2]$$

$$\frac{1}{2} \leq \frac{\log n}{4}$$

$$[\text{divide by 4}]$$

$$\frac{n}{4} \leq \frac{n \log n}{4}$$

$$[\text{multiply by } n]$$

$$0 \leq \frac{n \log n - n}{4}$$

$$[\text{Add } \frac{n}{4}]$$

$$\frac{n \log n}{4} \leq \frac{n \log n - n}{2}$$

$$[\text{Add } \frac{n \log n}{4}]$$

$$(C = 1/4)$$

thus we can see: $n \log n \geq \log(n!) \geq \frac{n \log n}{4}$

Base Case: $n=1$ (factorial can't have $n=0$)

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

$$\log(1!) = \Theta((1) \log(1))$$

$$\log(1) = \Theta(\log(1))$$

$$\Theta = \Theta(0)$$

∴ the base case is true

$$\text{Proves: } \frac{n \log n - n}{2} \geq \frac{n \log n}{4}$$

I'm showing $\frac{n \log n - n}{2}$ is greater than a multiple of $n \log n$:

$$\text{thus: } cn \log n \leq \frac{n \log n - n}{2}$$

$$\text{Since: } \frac{n \log n - n}{2} < \frac{n \log n}{2}, \text{ ?}$$

multiple of $\frac{1}{4}$ is chosen to

$$\text{allow: } \frac{n \log n - n}{2} \geq \frac{n \log n}{4}$$

$$\begin{aligned} \log(n!) &\geq \frac{n}{2} \log\left(\frac{n}{2}\right) & [\text{from earlier}] \\ &= \frac{n}{2} (\log n - \log 2) \\ &= \frac{n}{2} (\log n - 1) \\ &= \frac{n \log n - n}{2} & [\text{Algebra}] \end{aligned}$$

$$\frac{n \log n - n}{2} \geq \frac{n \log n}{4} & [\text{from limit comparison}]$$

$$\frac{n \log n}{4} = \Omega(n \log n) & [\text{Drop Constants}]$$

$$\text{thus } \log(n!) = \Omega(n \log n)$$

Since $\log(n!) = O(n \log n)$ & $\Omega(n \log n)$

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

(3) Prove: $\sum_{i=1}^n \frac{1}{i(i+1)} = \frac{n}{n+1}$

Base Step:

let $n=1$:

$$\frac{1}{1(1+1)} = \frac{1}{2} \rightarrow \frac{1}{1+1} = \frac{1}{1+1} \Rightarrow \frac{1}{2} = \frac{1}{2}$$

Assume True for $f(n-1)$: when $n>1$ $f(n) = \sum_{i=1}^n \frac{1}{i(i+1)}$

$$\sum_{i=1}^n \frac{1}{i(i+1)} = \frac{1}{2} + \frac{1}{6} + \frac{1}{12} + \dots + \frac{1}{(n-2)(n-1)} + \frac{1}{(n-1)n} + \frac{1}{n(n+1)}$$

$$\left[\text{Value of } n-1 \right] \left[\sum_{i=1}^{n-1} \frac{1}{i(i+1)} \right] + \frac{1}{n(n+1)} \left[\text{Value of } n \text{ in the sum} \right]$$

Inductive Hypothesis: for any value less than n , assume true for $n>1$:

$$\frac{n-1}{n} = \sum_{i=1}^{n-1} \frac{1}{i(i+1)}$$

$$\text{thus: } \frac{n-1}{n} + \frac{1}{n(n+1)} = \frac{n-1(n+1)+1}{n(n+1)} = \frac{n^2+n-n-1+1}{n(n+1)} = \boxed{\frac{n}{n+1}}$$

\therefore Its true that $\sum_{i=1}^n \frac{1}{i(i+1)} = \frac{n}{n+1}$ for all $n>1$

$$\textcircled{4} \quad a) 2n^2 - 5n - 12 = O(n^3)$$

Show: $2n^2 - 5n - 12 \leq cn^3, \forall n > d$ where $c > 0$ & d

$$2n^2 - 5n - 12 \leq 2n^3 \rightarrow \forall n > 1$$

$\therefore c=2$ & $d=1$ its true

\therefore its **true** that $O(n^3)$ is big Oh for $2n^2 - 5n - 12$

$$b) 3n^2 \log n - 2n \log n = \Omega(n^2)$$

$$3n^2 \log n - 2n \log n = cn^2$$

$$3\log n - \frac{2\log n}{n} \geq c \quad [\text{definition of } \Omega]$$

$$\lim_{n \rightarrow \infty} \frac{3\log n}{1} - \lim_{n \rightarrow \infty} \frac{2\log n}{n} = \infty - \lim_{n \rightarrow \infty} \frac{2\log n}{n} \stackrel{H}{=} \infty - \lim_{n \rightarrow \infty} \frac{1/n \ln(2)}{1}$$

$$= \infty - \lim_{n \rightarrow \infty} \frac{1}{n \ln(2)} = \infty - \frac{1}{\infty} = \infty > c = 1 \quad \because \lim_{n \rightarrow \infty} \frac{f(x)}{g(x)} > 0$$

\therefore Its **true** that $\Omega(n^2)$ is the lower bound for $f(n)$

$$c) 2^n = \Omega(2^{n+1})$$

Must Show that:

$$2^n \geq c2^{n+1} \quad \forall n \geq d$$

$$\frac{2^n}{2^n} \geq c \frac{2^{n+1}}{2^n} \quad \forall n \rightarrow 1 \quad [2^{n+1} = 2^n 2^1 = 2(2^n)]$$

$$1 \geq 2c \rightarrow \frac{1}{2} \geq c$$

$\therefore c = \frac{1}{2}$ & $d = 1$ its true

\therefore Its **true** that $\Omega(2^{n+1})$ is big-omega for 2^n

$$d) n \log n = O(n^2) \quad \lim_{n \rightarrow \infty} \frac{f(x)}{g(x)} < \infty \quad \frac{n \log n}{n^2} \leq c$$

Must Show that:

$$\lim_{n \rightarrow \infty} \frac{n \log n}{n^2} \leq \lim_{n \rightarrow \infty} \frac{(\log(n)+1)/\log(2)}{2n} = \lim_{n \rightarrow \infty} \frac{\log n + 1}{\log(2) 2n} \stackrel{H}{=} \lim_{n \rightarrow \infty} \frac{1/n}{2 \log(2)} = \lim_{n \rightarrow \infty} \frac{1}{n 2 \log(2)}$$

$$= 0 < c = 1$$

$0 < \infty \quad \therefore$ True

Its **true** that $O(n^2)$ dominates $n \log n$.

$$e) n^2 \log n = \omega(n^3) \therefore \frac{n^2 \log n}{n^3} > c$$

$$\lim_{n \rightarrow \infty} \frac{n^2 \log n}{n^3} = \lim_{n \rightarrow \infty} \frac{\log(n)}{n} \stackrel{H}{=} \lim_{n \rightarrow \infty} \frac{1/n \log(2)}{1} = \lim_{n \rightarrow \infty} \frac{1}{n \log(2)}$$

$$= \frac{1}{\infty} = \boxed{0} > c = 1$$

$O < \infty$: False as $O < c \leq \infty$

\therefore Its **False** that $\omega(n^3)$ dominates $n^2 \log n$

(5) a)

$$T(n) \leq \begin{cases} 1 & \text{if } n=1 \\ 1+2T\left(\frac{n}{3}\right) & \text{if } n \geq 3 \end{cases} \quad a=2, b=3, d=0 \quad \alpha \geq 1, b > 1, d \geq 0$$

\therefore **use Masters Theorem**

b) $\log_3(2) > 0 \quad T(n) = O(n^{\log_3(2)})$

$$T(n) \leq \begin{cases} 1 & \text{if } n=2 \\ 1+T(\sqrt{n}) & \text{if } n \geq 4 \end{cases} \quad a=1, b=1, d=0 \quad \alpha \geq 1, b > 1, d \geq 0$$

\therefore **do NOT use Masters**

$$T(n) = 1 + T(\sqrt{n})$$

Assume n is a power of a power of 2: $n = 2^{2^K}$ & $K = \log \log n$

$$\sqrt{n} = n^{1/2} = (2^{2^K})^{1/2} = 2^{\frac{2^K}{2}} = 2^{2^{K-1}}$$

thus:

$$T(n) = 1 + T(\sqrt{2^{2^K}}) = T(2^{2^{K-1}}) + 1$$

now: let $V = 2^K$; $V \in \mathbb{R}$

$$n^{1/4} = (2^{2^K})^{1/4} = 2^{\frac{2^K}{4}} = 2^{(2^K \cdot 2^{-2})} = \boxed{2^{2^{K-2}}}$$

we can say:

$$n^{1/V} = (2^{2^K})^{1/V} = 2^{\frac{2^K}{V}} = 2^{(2^K \cdot 2^{-V})} = \boxed{2^{2^{K-V}}}$$

this describes the recurrence. V is defined in here; it equal to $V = 2^K$; $V \in \mathbb{R}$

thus:

$$\begin{aligned} T(n) &= 1 + T(\sqrt{2^{2^K}}) = T(2^{2^{K-1}}) + 1 = T(2^{2^{K-2}}) + 2 \\ &= (1 + T(2^{2^{K-3}}) + 3) = \dots = (K + T(2^{2^0})) \\ &= T(2) + K \end{aligned}$$

$\therefore O(\log \log n)$

then:

$$T(2) = T(\sqrt{2}) + 1 = T(2) + 1 = 2 \quad \therefore T(2^K) = K + 2$$

thus if $T(2^K) = K + 2$ & n is a power of a power of two then:

$$2^{2^K} = n \quad \& \quad K = \log \log n \text{ thus:}$$

$$T(2^K) = T(n) \quad \therefore T(n) = (\log \log n) + 2 \quad \text{when } n > 1$$

$$T(n) = \boxed{O(\log \log n)} + 2$$

$$c) T(1) = 1$$

$$T(n) = 5T\left(\frac{n}{7}\right) + n \text{ for } n \geq 2$$

Assume $n = 7^k$ for some $k > 0$ [Assignment Specifications]

$$T(n) \leq \begin{cases} 1 & \text{if } n=1 \\ 5T\left(\frac{n}{7}\right) + n & \text{if } n \geq 2 \end{cases}$$

$a=5, b=7, d=1$

$a \geq 1, b > 1, d \geq 0$

\therefore Use Masters Theorem

$$\therefore d \leq \log_b a$$

$$\therefore 1 > \log_7(5)$$

$$\therefore 1 > \frac{\log_{10} 5}{\log_{10} 7} \rightarrow 1 > \sim 0.8271$$

$$\text{thus: } d > \log_b a \therefore T(n) = O(n^d) = O(n)$$

the big O of the given recurrence is $O(n)$.

$$d) T(n) = \begin{cases} 1 & \text{if } n=1 \\ T(n-1) + 2n & \text{if } n \geq 2 \end{cases}$$

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

$a \geq 1, b \neq 1, d \geq 2$

\therefore Do not use Masters Theorem

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

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

$$T(n-2) = T(n-3) + 2(n-2)$$

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

$$= T(n-2) + 2(n-1) + 2n = T(n-2) + 4n - 2$$

$$= T(n-3) + 2(n-2) + 2(n-1) + 2n =$$

$$= T(n-4) + 2(n-3) + 2(n-2) + 2(n-1) + 2n$$

if we substitute n for $n-1$ then index value:

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

$$T(n-2) = T(n-3) + 2(n-2)$$

$$T(n-3) = T(n-4) + 2(n-3)$$

$$T(n-4) = T(n-5) + 2(n-4)$$

increase by 1 element, \therefore there's a recurrence

the recurrence can be expressed as follows: for $K > 1$

$$T(n) = T(n-K) + 2(n-K+1) + 2(n-K+2) + \dots + 2n - K$$

finding Base case w/ $T(n-K)$; let $n=1$ such that $T(0)$, solve for K

$$T(n) = T(n-n) + 2(n-n+1) + \dots + 2n - n$$

$$= (0 + 2 + 4 + 6 + \dots + 2n) - n$$

$$= \sum_{i=0}^n 2i - n$$

[Convert $0 + 2 + 4 + \dots + 2n$ to a summation]

$$= \frac{2n(n+1)}{2} - n = n^2 + n - n = \boxed{O(n^2)}$$

⑥ a) this problem can be modeled as a recurrence relation where $a=5, b=2, \& d=1$
Such that $ST(\frac{n}{2}) + n$ represents the recurrence relation.

$$\begin{cases} ST\left(\frac{n}{2}\right) + n & \text{if } n \geq 0 \\ \end{cases}$$

$$a=5, b=2, d=1 \quad \& \quad a \geq 1, b \geq 1, d \geq 1$$

∴ use Masters Theorem

$$d \rightarrow \log_2 5$$

$$d < \log_2 5$$

$$1 < \log_2 5$$

$$1 < \frac{\log_{10} 5}{\log_{10} 2} \Rightarrow 1 < \approx 2.322 \therefore T(n) = O(n^{\log_2 5})$$

$$T(n) = O(n^{\log_2 5})$$

b) This can be represented as a recursive function where
 $2T(n-2) + 2$ represents the recurrence relation.

$b=1$ ∴ Masters Theorem Can't be used, as $b \neq 1$

To Solve via Algebra:

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

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

$$T(n-2) = 2T(n-3) + 2$$

$$T(n-3) = 2T(n-4) + 2$$

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

$$2(2T(n-2)) + 2 + 2$$

$$2(2(2T(n-3))) + 2 + 2 + 2$$

if we substitute for $n \rightarrow n-1$ then: index value:

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

2

increase by 1 element ∵

$$T(n-2) = 2T(n-3) + 2$$

3

recurrence

$$T(n-3) = 2T(n-4) + 2$$

4

Thus the recurrence can be expressed as follows: for $K \geq 1 \& r > 1$

$$T(n) = 2^r T(n-K) + 2 + 2 + 2 + \dots + 2$$

Add n many times

$$= 2^r T(n-K) + 2 \sum_{i=0}^{r-1} 2^i = 2^r T(n-K) + 2(2^r - 1)$$

Finding base case w/ $T(n-K)$: let $n=K$ such that $T(0)$, solve for K

$$T(n) = 2^r T(n-n) + 2(2^r - 1)$$

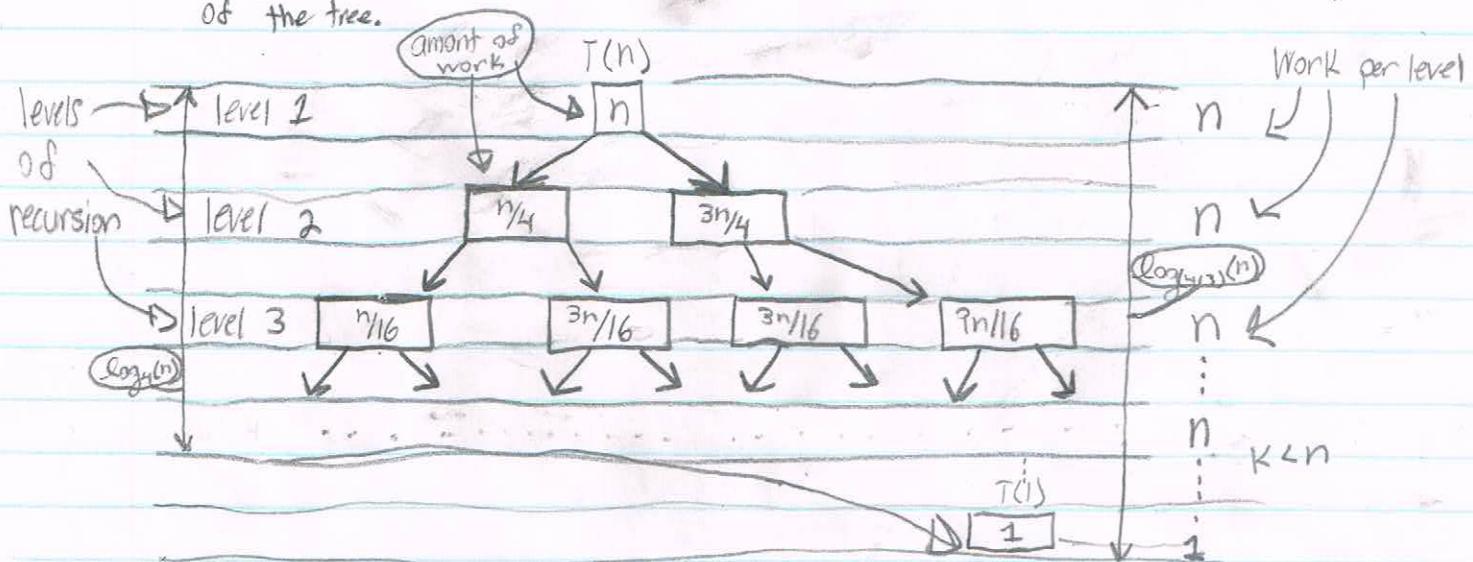
$$= 2^r(T(0) + 2) - 2$$

$$= \Theta(2^n) \quad \text{Simplified}$$

c) This problem can be modeled as a recursive relation where $b_1 = \frac{n}{4}$ & $b_2 = \frac{3n}{4}$ added and joined in $O(n)$ time.

$$T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + O(n) \text{ represents the recurrence relation}$$

Using the recursion tree method, we can determine the time complexity of the tree.



$$n \log_4(n) \leq T(n) \leq n \log_{4/3}(n)$$

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

- Explanation: For the left side of the tree

- We assume: $n = 4^K$
- We make it to a leaf in the tree when we're at $T(1)$, the terminating step
- To get to $T(1)$, we must divide n by 4^K such that it equals one
- $\frac{n}{4^K} = 1$ thus: $n = 4^K$ thus $K = \log_4(n)$

- For the Right side of the tree: we do the same as above; $n = \frac{n}{3}$

$$n^{(3/4)^K} = 0 \rightarrow \log_{4/3}(n)$$

- Thus we still get the estimate of: $\Theta(n \log n)$, but since the left half of the tree is shorter, we can't prove $\Omega(n \log n)$

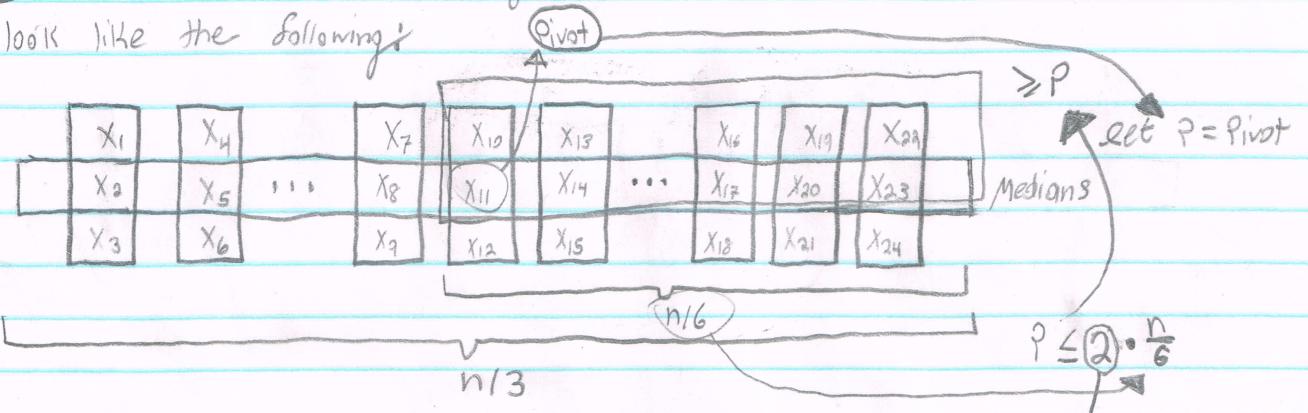
- Since the final solution is combined in $O(n)$ time (defined by question), there is no deterministic ω , $\therefore \Theta(n \log n)$ is the running time.

• Thus its possible to get lower bound than $n \log n$, thus theres no Θ , only O .

- Out of the Algorithms from parts A,B,&C from question 6, I would choose to use Algorithm C.
- This is because the big Oh of Alg. C is the lowest, the runtimes for the Algs. is as follows:
 - A: $O(n^{\log_2 6})$
 - B: $O(2^n)$ Simplified
 - C: $O(n \log n)$

Thus Alg. C has the lowest time complexity w/ a big Oh of $O(n \log n)$, so I would use Algorithm C.

- ⑦ If we Partitioned the array into $n/3$ medians of Size 3, it would look like the following:



the value of '2' comes from the fact that there are 2 elements per row greater than or equal to the current pivot.

- a) As seen above, $P \leq 2 \cdot \frac{n}{6}$; thus the elements Smaller than the pivot would be the remainder of this value: Let P_s = elements smaller than the pivot

$$2 \cdot \frac{n}{6} = \frac{2n}{6} = \frac{n}{3} \rightarrow n - \frac{n}{3} = \frac{2n}{3}$$

$\therefore P < \frac{2n}{3}$ elements may be Smaller than the Pivot

The number of elements guaranteed to be larger than the pivot are:

$$P_l \geq 2 \cdot \left(\frac{1}{2}\left(\frac{n}{3}\right) - 1\right) + 2 \geq \frac{2n}{6} - 2$$

→ the # of elements included/excluded from equation

- 2 : the number of elements per median-group which is equal to or greater than the pivot
 1 : the number of elements not greater than the pivot, which includes only the main pivot itself

▷ UNSimplified for clarity

half

$\frac{n}{6}$: the number of elements in the list; each element is a median-group of size 3

$\frac{2n}{3}$: the number of elements greater than or equal to the pivot, unsimplified for clarity (or smaller)

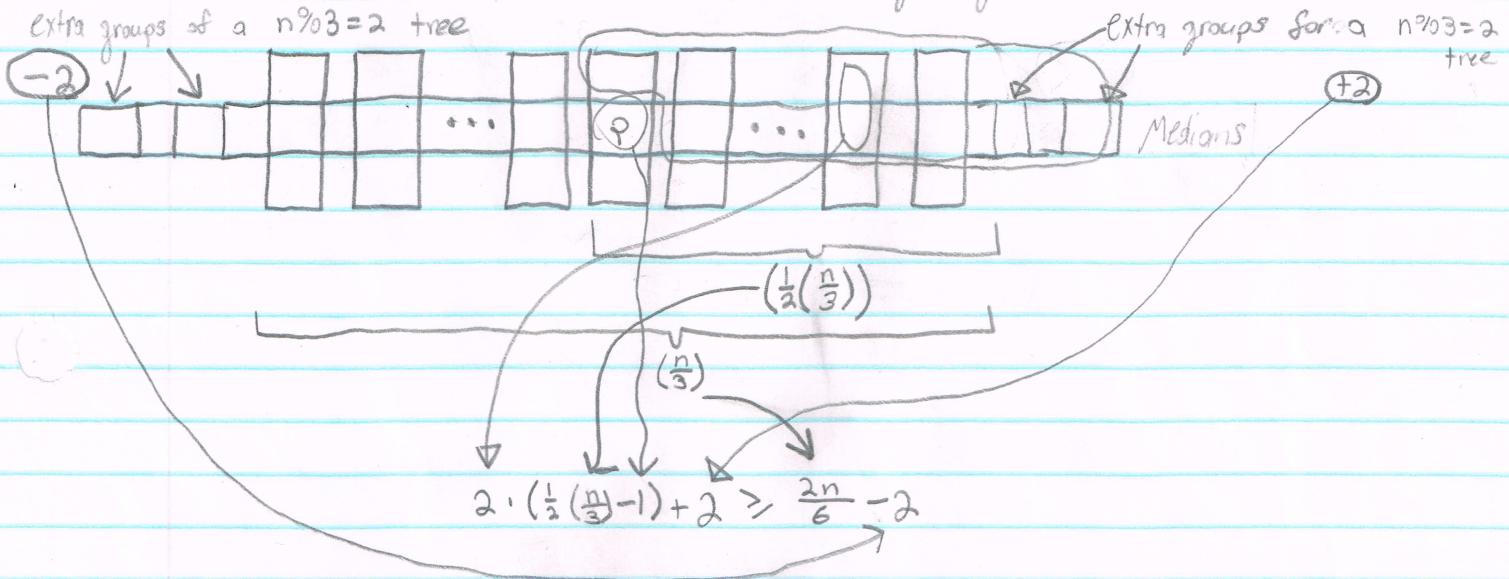
thus: the guaranteed elements to be larger than the pivot is within the range of! let P_L = elements larger than the pivot, P_S = elements smaller than the pivot

$$P_L \geq 2 \cdot \left(\frac{1}{2} \cdot \frac{n}{3} - 1\right) + 2 \geq \frac{2n}{6} - 2$$

$$P_S \geq 2 \cdot \left(\frac{1}{2} \cdot \frac{n}{3} - 1\right) + 2 \geq \frac{2n}{6} - 2$$

this can be best illustrated w/ the following diagram:

extra groups of a $n \% 3 = 2$ tree



b) The recurrence for the running time for the Alg. can be broken into four steps:

Step 1: Divide S into $\frac{n}{3}$ groups, each of Length 3

↳ $O(n)$ time, we must iterate through the list 1 time in order to accomplish this

Step 2: For $i = 1, 2, 3, \dots, \frac{n}{3}$ groups; Compute the median of the i^{th} group, & let this median be m_i

↳ $O(n)$ time, we must iterate through $\frac{n}{3}$ groups of 3, which is exactly $\frac{n}{3} \cdot 3 = n$ iterations

Step 3: Compute the median p of $m_1, m_2, \dots, m_{n/3}$

↳ $T(\frac{n}{3})$ recurrences, we can simply recursively compute the $\frac{n}{6}^{th}$ smallest element of the sequence $m_1, m_2, \dots, m_{n/3}$ & call $\text{Select}(\{m_1, m_2, \dots, m_{n/3}\}, \frac{n}{6})$ which takes $T(\frac{n}{3})$ time.

Let Select be the algorithm described in this question (27)

Step 4: With p as the pivot, use the Select Algorithm: $\text{Select}(S, k)$

$\hookrightarrow \mathcal{O}(n)$ time to iterate through the elements in a list
of size $\frac{n}{3}$
+ $T(\frac{2n}{3})$ recurrences to search through the longest list, the list of elements guaranteed to be smaller than the pivot.

$$\hookrightarrow \text{thus: } \mathcal{O}(n) + T\left(\frac{2n}{3}\right)$$

thus: the recurrence for the running time is

$$T(n) = \mathcal{O}(n) + \mathcal{O}(n) + T\left(\frac{n}{3}\right) + \mathcal{O}(n) + T\left(\frac{2n}{3}\right)$$

Step 1 Step 2 Step 3 Step 4

$$= 2\mathcal{O}(n) + T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right)$$

$$= \mathcal{O}(n) + T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) \quad [\mathcal{O}(n) \text{ is just } \mathcal{O}(n)]$$

C) we can find the running time of the algorithm in terms of n via induction:

Claim: $T(n) \leq cn$ for some constant c

Proof:

Base Case: Choose large c or for small n

$$\text{let } c=10 \text{ & } n=1$$

$$T(1) = 0$$

$$0 \leq 10 \cdot 1$$

Inductive Hypothesis:

Let n be "large", assume $T(n') \leq cn'$ for all $1 \leq n' \leq n$

Inductive Step:

$$T(n) = \mathcal{O}(n) + T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right)$$

$$\stackrel{d>0}{\leq} dn + \frac{cn}{3} + \frac{2cn}{3}$$

$$= dn + cn$$

$$\leq cn \text{ if } dn \leq 0 \text{ or } \rightarrow dn \leq 0$$

Contradiction, $0 \mathcal{O}(n)$ is not the runtime

\hookrightarrow this is a contradiction,

the list must be greater than 0 elements!

Next, let's try $n \log n$:

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

Claim: $T(n) \leq cn \log n$ for some constant n

Proof:

use the previous base case

Inductive hyp.:

Let n be large, assume $T(n') \leq cn' \log n'$ for all $1 \leq n' < n$

Inductive Step:

$$T(n) = O(n) + T(\frac{n}{3}) + T(\frac{2n}{3})$$

$$\leq dn + \frac{cn}{3} + \frac{2cn}{3}$$

$$= dn + cn$$

$$\leq cn \log n \text{ if } dn \leq c \log n \text{ iff } c \geq \frac{dn}{\log n}$$

i.e. the k^{th} smallest element in a sequence of n #'s can be computed in $O(n \log n)$ time.

We can use Masters Theorem to support this Claim:

$$T(n) = T(\frac{n}{3}) + T(\frac{2n}{3}) + O(n)$$

$$a \geq 1 \text{ & } b > 1, d \geq 0$$

is similar to $\Rightarrow G(n) = 3G(\frac{n}{3}) + O(n)$ where $a=3, b=3, d=1$

\therefore Masters applies

$$\bullet d = \log_b(a) \text{ then } T(n) = O(n^{\log_b(a)}) = O(n \log n)$$

$$\bullet l = \log_3(3)$$

$$\bullet l = 1$$

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

Therefore, the running time is $O(n \log n)$ through Masters Theorem & induction.

⑧ Given an unsorted array of n numbers, the Second Smallest number can be found in $n + (\log n) - 2$ Comparisons:

- To find the Second Smallest element, we must also find the Smallest.

- We can build a tree & find the Smallest element in $O(n)$ time! Use tournament comparison

□ We must pair all elements into groups of two & Compare them
we compare i_{an} w/ i_{an+1} for list $0 \leq n \leq \frac{n-1}{2}$, if $i_{an} > i_{an+1}$ interchange the elements.

Iteration: We can draw the following pattern: Let $0 \leq k \leq (n-2^{k-2})(2^{-k})$

1: if $i_{an} > i_{an+1}$ change
2: if $i_{4n} > i_{4n+1}$ change
|
K: if $i_{2^k n} > i_{2^k n + 2^{k-1}}$ change

Swap elements
if true

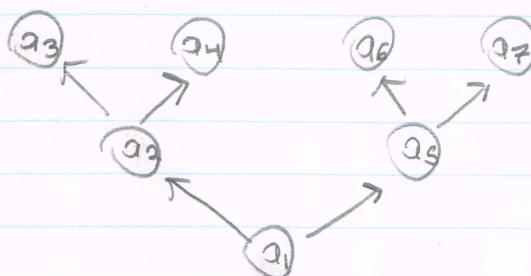


This runs for $n \leq 2^{k-1}$ trials, thus if the last pass is the k^{th} Pass then we must make fewer than the number of elements.

This is because we must compare every number in the set except the smallest:

∴ $n-1$ Comparisons

To illustrate:



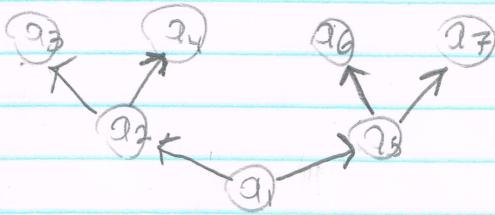
Can result in $n-1$ Comparisons if every elements Compared & Swapped.

The Smallest # in the Set will become the first element, & every other element will have been compared.
Hence: $n-1$

- the Set is now in a Sorted tree. We can simply iterate through the trees elements which have lost direct comparisons to the smallest element.

- let g be the number of comparisons made there were a total of $2^{g-1} < n \leq 2^g$ Comparisons to make the tree, thus the upperbound for n can be $n = 2^g$.
 $\therefore g = \log n$
- thus the tournament-tree can be searched through in g ($\log n$) time
- we already have the smallest element, thus there are only $0, 1, 2, \dots, g-1$ Comparisons that were smaller than the smallest element.
- Worst case: $g-1$ Comparisons to find the second smallest when g is the time to search the tournament-tree, when $g = \log n$
 $\therefore \log n - 1$ Comparisons

↳ Assuming we use the floor of $\log n$ as the heap may be this can be illustrated as: Unbalanced.



Such that the elements

a) are all sorted down
Step one.

Thus we have a tree, i.e.
we know the smallest element
 $\therefore 1$ less comparison

- In total, building the tree via comparison & finding the smallest element is: $(n-1)$

Searching the tree minus the smallest element is:

$\log n - 1$

\therefore the total searches are:

$$(n-1) + (\log n - 1)$$

$$= n + \log n - 2$$