

$$T(n) = T(n-2q-1) + T(3q/2) + T(q/2) + \Theta(1) \quad T(n) > 2$$

$$T(n) = 1 \quad \text{when } T(n) \leq 2$$

Depends on $q \in [0, n/4]$

$$T(n-2q-1) = T(n-2q-1-2q-1) + T(3q/2) + T(q/2) + \Theta(1) + \dots$$

$$= T(n-4q-2) + 2T(3q/2) + 2T(q/2) + \Theta(1)$$

$$T(n-4q-2) = T(n-8q-4) + 3T(3q/2) + 3T(q/3) + \Theta(1)$$

\Downarrow

$$\sum_{i=1}^{2n} i \cdot x$$

$O(n)$ runtime

$$(b) \quad T(n) = T(n-q-1) + T(n/2 - q) + \Theta(n)$$

$$T(n-q-1) = T(n-2q-2) + T(n/2 - q/2 - 1/2 - q) + \Theta(n)$$

$$= T(n-2q-2) + T(n/2 + q/2 - 1/2) + \Theta(n) \\ + T(n/2 - q)$$

$$T(n-2q-2) = T(n-4q-4) + T(n/2 - q/2 - 1/2) + \Theta(n)$$

$$+ T(n/2 + q/2 - 1/2) + T(n/2 - q)$$

$$+ \Theta(n)$$

\dots

$$O(n \log n)$$

$$\begin{aligned}
(c) \quad T(n) &= T(n-q-1) + T(3q) + \Theta(n) \\
&\Rightarrow T(n-1) + T(0) + \Theta(n) \\
&\Rightarrow T(n-1) + \Theta(n) \\
&\Rightarrow O(n) \text{ worst case}
\end{aligned}$$

Given a sorted array A , the expected runtime of the silly randomized variant of quicksort is $O(\log n)$.

We can state this as we know that on the first occurrence of the algorithm, we have a $1/n$ of finding the v .

On the second $2/n$

Third $4/n$

and so on and so forth when dealing with randomization.

We can derive $\frac{2^{x-1}}{n}$, thus the average number

of recursive calls is:

$$\frac{1}{n} \sum_{i=1}^{\log(n)} i 2^{i-1} \Rightarrow i 2^{i-1} < \log(n) \cdot 2^{i-1}$$

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

$$\frac{\log(n)}{n} \cdot (n-1) \Rightarrow \log(n)$$

average case run time is $\log(n)$
 worst case is $O(n)$ if it has v as the
 $A[0]$ position and $A[\text{length}-1]$ as the
 random choice for comparison on every iteration,
 but that's highly unlikely

$$5. T(n) = 4T(n/2) + n^2 \lg n$$

$$a=4 \quad b=2 \Rightarrow n^{\log_2(4)} = n^2$$

$$f(n)/n^2 \lg n = \lg n \text{ which shows that } f(n)$$

is asymptotically larger, not polynomially. Thus, the master theorem

can't be applied. Using the recursion tree method:

$$T(n) = 4T(n/2) + n^2 \lg n \dots$$

$$T(n/2) = 4T(n/4) + (n/4) \lg(n/4) + n^2 \lg n \dots$$

$$T(n/4) = 4T(n/8) + (n/8) \lg(n/8) + (n/4) \lg(n/4) + n^2 \lg n \dots$$

$$T(n) \leq n^2 \lg n (\lg n - 1) + n^2 \Rightarrow n^2 (\lg n)^2$$

$$O(n^2 (\lg n)^2)$$

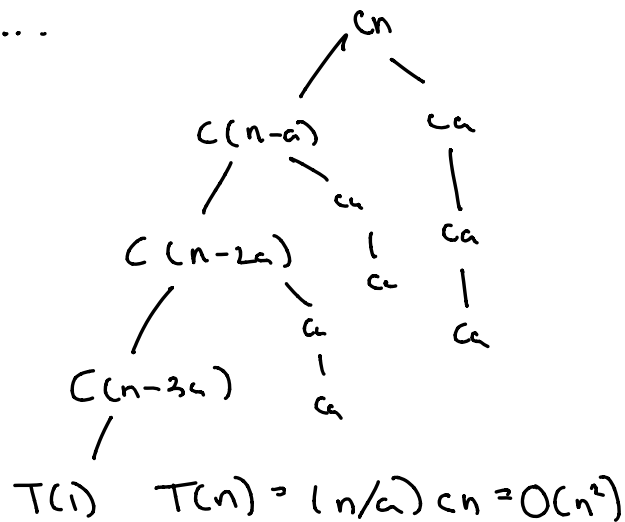
(a) $T(n) = T(n-a) + T(a) + cn$, where $a \geq 1$ and $c > 0$ are constants

$$T(n-a) = T(n-2a) + T(a) + c(n-a) + cn$$

$$T(n-2a) = T(n-4a) + T(a) + c(n-3a) + c(n-a) + cn$$

$$T(n-4a) = T(n-8a) + T(a) + c(n-7a) + c(n-6a) + \dots + c(n)$$

....



b. $T(n) = T(an) + T((1-a)n) + cn$ where $a \in (0,1)$, and $c > 0$ are constants.

$$\begin{array}{ccc} an & (1-a)n & cn \\ | & | & \downarrow \\ a(n+1) & (1-a)(n+1) & c(n+1) \end{array}$$

$O(n)$ constant run time

$$a. \quad T(n) = 4T(n/3) + n$$

$$a=4 \quad b=3$$

$$n^{\log_3(4)}$$

$$T(n) = \Theta(n^{\log_3 4}) \text{ by case 1}$$

$$\begin{aligned} T(n) &\leq 4(cn^{\log_3 4}/3) + n \\ &= \frac{4}{3} cn^{\log_3 4} + n \end{aligned}$$

this holds as long as $c > 1$ ✓

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

$$a = 4 \quad b = 2$$

$$n^{\log_2(4)} = n^2 \quad f(n) = O(n^{2-1}) \text{ by}$$

case 1 ✓

$$\Theta(n^2)$$

$$\begin{aligned} T(n) &= 4(c \cdot n^2/2) + n \\ &= \frac{4}{2}(c \cdot n^2) + n \end{aligned}$$

$$= 2cn^2 + n$$

where $c > 1$ ✓

$$T(n) = 3T(\sqrt{n}) + \log n$$

$$x = \log(n)$$

$$T(n) = 3T(\sqrt{n}) + x$$

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

$$A(x) \leq c x^{\log^3} + dx$$

$$A(x) \leq 3(c(x/2)^{\log^3} + d(x/2)) + x \text{ when } d \geq -2$$

$$\leq c x^{\log^3} + (\frac{3}{2}d + 1)x$$

$$\leq c x^{\log^3} + dx$$

$$A(x) \geq c x^{\log^3} + dx \text{ when } d \leq -2$$

....

$$\leq c x^{\log^3} + dx$$

$$\Rightarrow A(x) = \Theta(x^{\log^3}) \Rightarrow T(n) = \Theta(\log^{\log^3} n)$$

Knowing that for any $n > 0$, the number of leaves of an almost complete binary tree is $n/2$

Base case: We know that $h=0$ is true from the previous statement.

Proof by induction: Let's assume the statement is

true for $h-1$. Let N be the # of nodes at height h . Looking at a tree T , it will have
 $n' = n - \lceil n/2 \rceil$ nodes thus

$$\begin{aligned} N &= n' / 2^h \\ &= n/2 / 2^h \\ &= (n/2) / 2^h \\ &= \frac{n}{2^{h+1}} \end{aligned}$$