

Date: 16/06/21

Black Board

Design and Analysis of Algorithms

Topics:

- **Randomized Quicksort Analysis**
 - **The Expected Running time and its Variance**

(i) Randomized Quick-sort Revisited

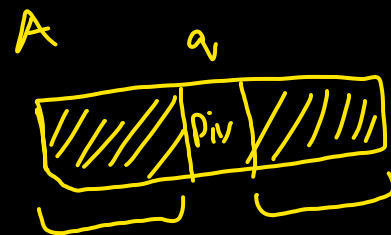
Quicksort(A, p, r)

IF $p < r$

$q \leftarrow \text{partition}(A, p, r)$

Quicksort(A, p, q-1)

Quicksort(A, q+1, r)



Worst Case:

$$\begin{aligned} T(n) &= T(n-1) + O(n) \\ &= O(n^2) \end{aligned}$$



Claim:

$$E[T(n)] = O(n \lg n)$$

$\xrightarrow{T(n)}$ R.V

$$\left\{ \begin{array}{l} O(n \lg n) \\ \vdots \\ cn^2 \end{array} \right.$$

$O(n \lg n)$

(ii) Proof That $E[T(n)] = O(n \lg n)$ for Rand QS

For any comp. based sort,

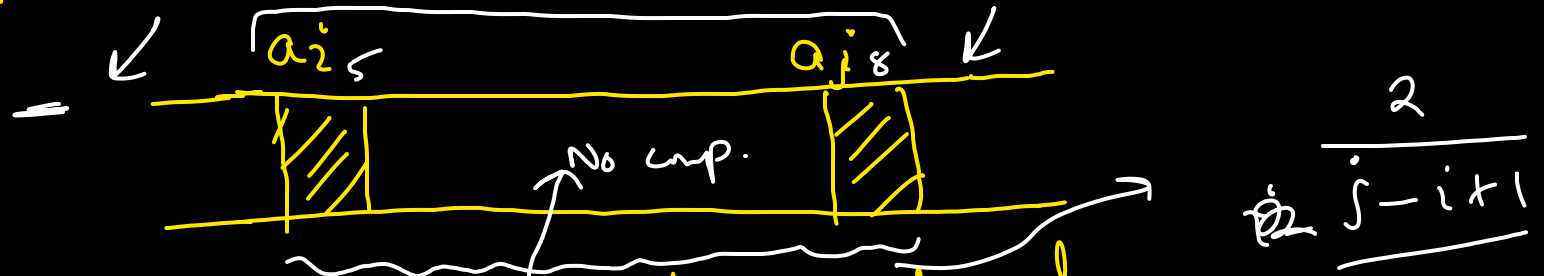
$$[T(n) = O(\# \text{ of comparisons between data})]$$

We wish to estimate the expected # of comparisons made by RQS.

Q = What is the probability that the i^{th} smallest
a_i of the input & the j^{th} smallest element a_j
of input are compared by RQS. ($i < j$)

$a_1 \ a_2 \ a_3 \ \dots \ a_i \ \dots \ a_j \ \dots \ a_n$

Random partition (consider #s in sorted order)



When will a_i & a_j be compared.

→ a_i & a_j will never be compared
if p is between them

$$a_i = 5$$

→ a_i & a_j will only be compared
if either a_i or a_j are picked

$$a_j = 20$$

as pivot.

$$P_r\{a_i \text{ \& } a_j \text{ are comp}\} = \frac{2}{j-i+1}$$

$$X_{ij} = \begin{cases} 0, & a_i \& a_j \text{ not comp.} \\ 1, & a_i \& a_j \text{ are comp.} \end{cases}$$

$$E[X_{ij}] = \frac{2}{j-i+1}$$

$X :=$ total number of comp in ROs

$$X = \sum_{i=1}^n \sum_{j=i+1}^n X_{ij}$$

X_{34} same X_{43}

$$E[X] = \sum_{i=1}^n \sum_{j=i+1}^n E[X_{ij}]$$

Linearity of
Expectation

$$E[X] = \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1}$$

$$= 2 \sum_{i=1}^n \left[\sum_{j=i+1}^n \frac{1}{j-i+1} \right]$$

$$= 2 \sum_{i=1}^n \left[\frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \frac{1}{5} + \dots + \frac{1}{n-i+1} \right]$$

$$\leq 2 \sum_{i=1}^n \left[1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \right]$$

$$\leq 2n \lg n$$

$$\Rightarrow T(n) = O(n \lg n)$$

Expected Running time of ROB

worst
 $T(n) = O(n^2)$ not realistic

(iii) How far does $E[T(n)]$ go from its mean $O(n \lg n)$

Are we likely to be close to $O(n \lg n)$ on most occasions?

Consider a loaded die

$$\left\{ \begin{array}{l} P\{X=1\} = \frac{1}{2} \\ P\{X=6\} = \frac{1}{2} \end{array} \right\}$$

$$P\{X=2\}, P\{X=3\}, P\{X=4\}$$

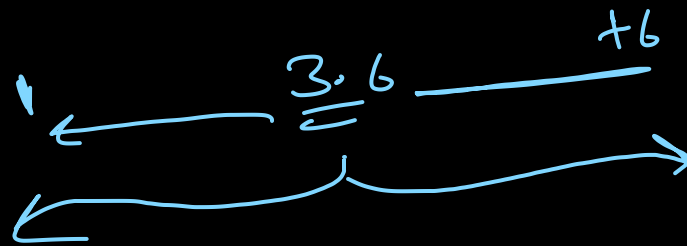
$$P\{X=5\} = 0$$

$$E[X] = 1 \cdot \frac{1}{2} + 6 \cdot \frac{1}{2} = \underline{3.5} = 7/2$$

Variance, $\sqrt{\text{variance}} = \underline{\underline{\text{s.d.}}}$

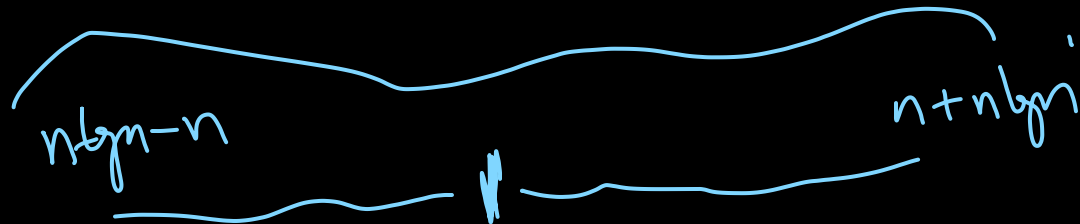
$$\begin{aligned} \cancel{E[X]} \quad \text{Var}(X) &= E\{(X - E[X])^2\} \\ &= E\left\{(X - \frac{7}{2})^2\right\} \\ &= E\left[X^2 - 7X + \frac{49}{4}\right] \\ &= E[X^2] - 7E[X] + \frac{49}{4} \end{aligned}$$

$$\text{s.d.} = \underline{\underline{2.5}} \approx 6$$



Variance of RQS X

$$s.d \approx n$$



$$nlg n \approx T(n) \leq n + nlg n$$

with high prob.

$$T(n) \leq 2nlg n$$

