# Randomized Algorithm II Randomized QuickSort

### Randomized Quicksort

- Recall *deterministic* Quicksort
- Depending on the choice pivot, could be  $O(n^2)$
- What if we pick the pivot uniformly at random?
  - We saw in randomized selection that this leads to good pivots half of the time

Quicksort(
$$A$$
):  
If  $|A| < 3 : SetElse: choose a
 $A_{< p}, A_{> p} \in$   
Quicksort( $A$ )$ 

Sort(A) directly a pivot element  $p \leftarrow A$   $\leftarrow$  Partition around p  $(A_{< p})$  $(A_{> p})$ 

### Randomized Quicksort

- Intuitively half the pivots will be good, half bad
- We will analyze quick sort using another accounting trick (see the textbook for example similar to selection's approach of analyzing "phases")
- Total work done can be split into to types:
  - Work done making recursive calls (this is a lower order term, it turns out)
  - Work partitioning the elements
- How many recursive calls in the worst case?
  - Imagine worst pivot being chosen each time
  - *O*(*n*)

### Randomized Quicksort

- We thus need to bound the work partitioning elements
- Partitioning an array of size n around a pivot p takes exactly n-1 comparisons
- We won't look at partitions made in each recursive call, which depend on the choice of random pivot
- Idea: Instead, account for the total work done by the partition step by summing up the total number of comparisons made
- Two ways to count total comparisons:
  - Look at the size of arrays across recursive calls and sum •
  - Look at all pairs of elements and count total # of times they are compared (this is easier to do in this case)

### Aside: Randomized Analysis

- Often multiple ways to determine a randomized algorithm's cost
- We can split into phases, or count the cost directly. We can calculate each probability, or use linearity of expectation
- Intrinsically some "cleverness" involved in choosing the way that gets you a clean answer
- We'll focus on problems where there's a clear path to finding the solution (either it follows directly from the question, or we'll revisit problems you've seen before). More complex problems abound if you look!
- That said, here's a very clever way to calculate Quicksort's running time

# Counting Total Comparisons

- Just for analysis, let B denote the sorted version of input array A, that is, B[i] is the  $i^{th}$  smallest element in A
- Define random variable  $X_{ij}$  as the number of times Quicksort compares B[i] and B[j]

• Observation:  $X_{ii} = 0$  or  $X_{ii} = 1$ , why?

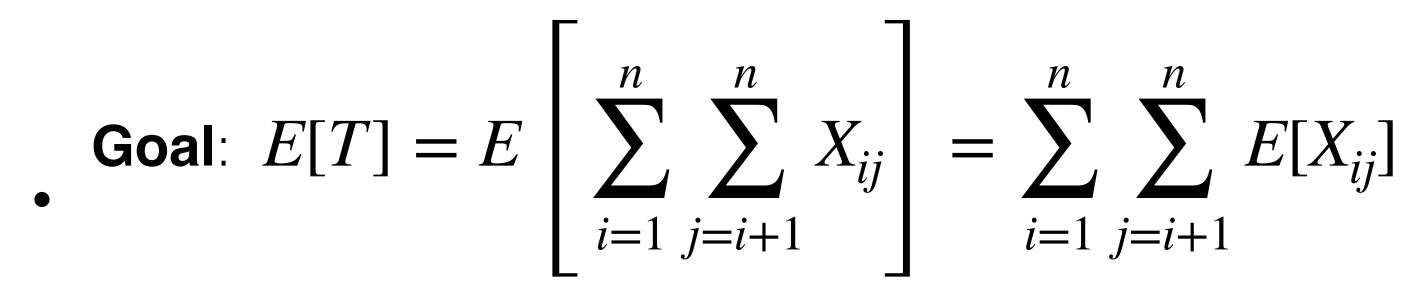
• B[i], B[j] only compared when one of them is the current pivot; pivots are excluded from future recursive calls

Let 
$$T = \sum_{i=1}^{n} \sum_{j=i+1}^{n} X_{ij}$$
 be the total num

by randomized Quicksort

nber of comparisons made





- $E[X_{ii}] = \Pr[X_{ii} = 1]$
- When is  $X_{ii} = 1$ ? That is, when are B[i] and B[j] compared?
- Consider a particular recursive call. Let rank of pivot p be r.
  - Let's think about where B[i], B[j] lie with respect to p

- Goal:  $E[T] = E \left| \sum_{i=1}^{n} \sum_{j=i+1}^{n} X_{ij} \right| = \sum_{i=1}^{n} \sum_{j=i+1}^{n} X_{ij}$
- $E[X_{ii}] = \Pr[X_{ii} = 1]$
- When is  $X_{ii} = 1$ ? That is, when are B[i] and B[j] compared?
- Consider a particular recursive call. Let rank of pivot p be r.
  - Case 1. One of them is the pivot: r = i or r = j
  - Case 2. Pivot is between them: r > i and r < j
  - Case 3. Both less than the pivot: r > i, j
  - Case 4. Both greater than the pivot: r < i, j

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

### Comparisons for Each Case

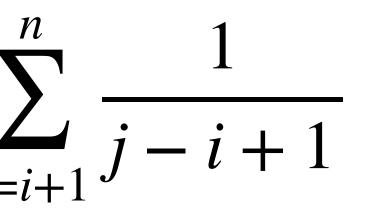
• **Case 1**. r = i or r = j

- B[i] and B[j] are compared once and one of them is excluded from all future calls
- **Case 2**. r > i and r < j
  - *B*[*i*] and *B*[*j*] are both compared to the pivot but not to each other, after which they are in different recursive calls: will never be compared again
- **Case 3**. r > i, j and **Case 4**. r < i, j
  - *B*[*i*] and *B*[*j*] are not compared to each other, they are both in the same subarray and may be compared in the future
- Takeaway: B[i], B[j] are compared for the 1st time when one of them is chosen as pivot from B[i], B[i + 1], ..., B[j] & never again

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

$$E[T] = \sum_{i=1}^{n} \sum_{j=i+1}^{n} E[X_{ij}] = 2 \sum_{i=1}^{n} \sum_{j=i-1}^{n} \sum_{j=i-1}^{n} \sum_{j=i-1}^{$$

•  $\Pr[X_{ij} = 1] = \Pr(\text{one of them is picked as pivot from } B[i], B[i + 1], ..., B[j])$ 



- range B[i], B[i + 1], ..., B[j]
- $\Pr[X_{ij} = 1] = \frac{2}{j i + 1}$
- $E[T] = \sum_{i=1}^{n} \sum_{j=i+1}^{n} E[X_{ij}] = 2 \sum_{i=1}^{n} \sum_{j=i+1}^{n} \sum_{i=1}^{n} \sum_{j=i+1}^{n} E[X_{ij}] = 2 \sum_{i=1}^{n} \sum_{j=i+1}^{n} \sum_{j=i+1}^{n} \sum_{i=1}^{n} \sum_{j=i+1}^{n} \sum$
- For fixed i, inner sum is  $\frac{1}{2} + \frac{1}{3} + \frac{1}{4}$
- Thus, expected number of comparisons is:  $\bullet$  $E[T] = O(n \log n)$

• B[i] and B[j] are compared iff one of them is the first pivot chosen from the

At each round, the probability that  $X_{ii} = 1$  conditioned on the event that we are in Case 1 or Case 2. (In Cases) 3 and 4, we "kick the can" until another round)

$$\sum_{i=1}^{n} \frac{1}{j-i+1} + \dots + \frac{1}{n-i+1} \le \sum_{\ell=2}^{n} \frac{1}{\ell} = O(\log n)$$

# Quick Sort Summary

- Las Vegas algorithms like Quicksort and Selection are always correct and their running time guarantees hold *in expectation*
- We can actually prove that the number of comparisons made by Quicksort is  $O(n \log n)$  with high probability
  - W.H.P. means that the the probability that the running time of quicksort is more than a constant c factor away from its expectation is very small (polynomially small: less than  $1/n^c$  for  $c \geq 1$ )
  - Whp bounds are called **concentration bounds**
  - Whp: ideal guarantees possible for a randomized algorithm

# Acknowledgments

- Some of the material in these slides are taken from lacksquare
  - Shikha Singh
  - Kleinberg Tardos Slides by Kevin Wayne (<u>https://</u>  $\bullet$ www.cs.princeton.edu/~wayne/kleinberg-tardos/pdf/ 04GreedyAlgorithmsI.pdf)
  - $\bullet$ algorithms/book/Algorithms-JeffE.pdf)

Jeff Erickson's Algorithms Book (<u>http://jeffe.cs.illinois.edu/teaching/</u>