

## Loop Invariant for Partition

Initialization: Before the loop starts, all of the loop invariant is satisfied because  $A[\text{end}] = \text{pivot}$  + the subarray  $A[\text{start}..i]$  +  $A[i+1..j-1]$  are empty.

Maintenance: While the loop is running, if  $A[j] \leq \text{pivot}$  then  $A[i+1]$  +  $A[j]$  are swapped +  $i + j$  are incremented. If  $A[j] > \text{pivot}$  then only  $j$  is incremented.

Termination: when the loop terminates  $j = \text{end}$  so all of the elements are partitioned into one of the three cases:  $A[\text{start}..i] \leq \text{pivot}$ ,  $A[i+1..end-1] \geq \text{pivot}$  +  $A[\text{end}] = \text{pivot}$

Time for partitioning  $\Theta(n)$

## Performance of Quicksort

- Can run as well as merge sort if subarrays are balanced + as bad insertion sort otherwise.

### Worst Case

- Completely unbalanced (sorted in ascending or descending / all same #)

- Recurrence Relation

$$T(0) = c$$

$$T(n) = T(n-1) + T(0) + \theta(n)$$

$$T(n-1) = T(n-2) + T(0) + \theta(n-1)$$

$$T(n) = T(n-2) + T(0) + \theta(n-1) + T(0) + \theta(n)$$

$$T(n-2) = T(n-3) + T(0) + \theta(n-2)$$

$$T(n) = T(n-3) + 3T(0) + \theta(n-2) + \theta(n-1) + \theta(n)$$

$$T(n) = T(0) + cn + \sum_{i=1}^n \theta(i)$$

$$T(n) = c(n+1) + \theta\left(\frac{n(n+1)}{2}\right)$$

$$T(n) = \theta(n^2)$$

### Best Case

- Completely balanced ( $\leq \frac{n-1}{2}$ ) both array sizes

$$T(n) = 2T\left(\frac{n}{2}\right) + T(0) + \theta(n)$$

$$= \theta(n \log n)$$

## Average Case

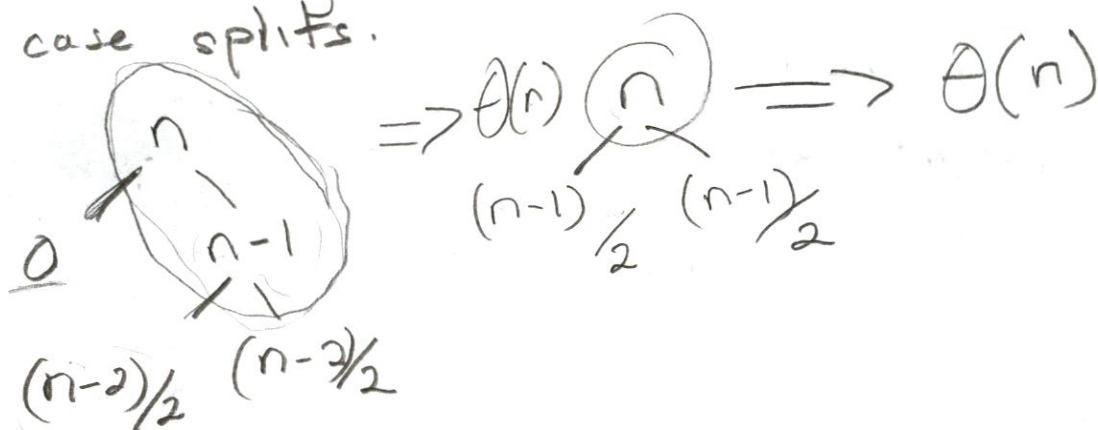
- Imagine partition always produces a 9-1 split

$$T(n) = T\left(\frac{9n}{10}\right) + T\left(\frac{n}{10}\right) + \Theta(n)$$

- As long as the base of a log is constant, it's still  $\log n$  in asymptotic notation.

## Intuition

- Alternating between best + worst case splits.



Both of these recursion trees will be  $\Theta(n \log n)$  but the constant hidden from the notation is higher in the left tree.

## Pseudo code

```
def random-partition(A, start, end):  
    idx = random.randint(start, end)  
    swap(A[idx], A[end])  
    return partition(A, start, end)
```

## Analysis on Randomized Quick sort

- The dominant cost of this algorithm is the partition function.
  - Partition removes the pivot element from future consideration each time.
  - Therefore, the partition function is called at most  $n$  times.
  - Partition work = constant + comparisons made in the loop.
  - Let  $X$  = total # of comparisons performed in all calls to partition.
  - Total work for entire execution  
 $= O(n + X)$
  - Now we'll compute a bound on  $(X)$ 
    - Rename the elements of  $A$  to  $\{z_0, z_1, \dots, z_n\}$   
 where  $z_i$  is the  $i^{\text{th}}$  smallest element.
    - Define a set  $Z_{ij} = \{z_i, z_{i+1}, \dots, z_j\}$
- $A = [5, 3, 1, 4, 2]$     
  $\begin{matrix} z_0 & z_1 & z_2 & z_3 & z_4 \\ 1 & 2 & 3 & 4 & 5 \end{matrix}$     
 $z_{13} = \{2, 3, 4\}$



- Each element is at most compared once.
- Let  $X_{ij} = \mathbb{I}\{z_i \text{ is compared to } z_j\}$   
 $X_{23} = 1 \text{ or } 0$

Total # of comparisons made by this algorithm.

$$X = \sum_{i=0}^{n-1} \sum_{j=i+1}^n X_{ij}$$

$$E[X] = E\left[\sum_{i=0}^{n-1} \sum_{j=i+1}^n X_{ij}\right]$$

$$= \sum_{i=0}^{n-1} \sum_{j=i+1}^n E[X_{ij}]$$

$$= \sum_{i=0}^{n-1} \sum_{j=i+1}^n \Pr\{z_i \text{ is compared to } z_j\}$$

- All we have to do now is define  $\Pr\{z_i \text{ is compared to } z_j\}$  in order to find the expected amount of comparisons.

-  $\Pr(z_i \text{ is compared to } z_j) = \Pr(\text{either } z_i \text{ or } z_j \text{ are the first pivots chosen in } z_{ij})$