Quick-Sort

- To understand quick-sort, let's look at a high-level description of the algorithm
- 1) **Divide** : If the sequence *S* has 2 or more elements, select an element *x* from *S* to you **pivot**. Any arbitrary element, like the last, will do. Remove all the elements of *S* and divide them into 3 sequences:
 - *L*, holds *S*'s elements less than *x*
 - *E*, holds *S*'s elements equal to *x*
 - *G*, holds *S*'s elements greater than *x*
- 2) **Recurse**: Recursively sort *L* and *G*
- 3) Conquer: Finally, to put elements back into *S* in order, first inserts the elements of *L*, then those of *E*, and those of *G*.
- Here are some pretty diagrams....







sorting













Analysis of Running Time

- Consider a quick-sort tree *T*:
 - Let $s_i(n)$ denote the sum of the input sizes of the nodes at depth *i* in *T*.
- We know that $s_0(n) = n$ since the root of *T* is associated with the entire input set.
- Also, $s_1(n) = n 1$ since the pivot is not propagated.
- Thus: either s₂(*n*) = *n* 3, or *n* 2 (if one of the nodes has a zero input size).
- The worst case running time of a quick-sort is then:

$$O\binom{n-1}{\sum_{i=0}^{n-1} s_i(n)}$$

Which reduces to:

$$O\binom{n-1}{\sum_{i=0}^{n-1}(n-i)} = O\binom{n}{\sum_{i=1}^{n-1}i} = O(n^2)$$

• Thus quick-sort runs in time $O(n^2)$ in the worst case.

Analysis of Running Time (contd.)

- Now to look at the best case running time:
- We can see that quicksort behaves optimally if, whenever a sequence S is divided into subsequences L and G, they are of equal size.
- More precisely:
 - $s_0(n) = n$

-
$$s_1(n) = n - 1$$

-
$$s_2(n) = n - (1 + 2) = n - 3$$

-
$$s_3(n) = n - (1 + 2 + 2^2) = n - 7$$

-
$$s_i(n) = n - (1 + 2 + 2^2 + ... + 2^i - 1) = n - 2^i + 1$$

...

- This implies that T has height $O(\log n)$
- Best Case Time Complexity: *O*(*n*log *n*)

Randomized Quick-Sort

- The main drawback to quick-sort is that it achieves its worst-case time complexity on data sets that are common in practice: sequences that are already sorted (or mostly sorted)
- To avoid this, we modify quick-sort so that it selects the pivot as a *random* element of the sequence
- The expected time of a randomized quick-sort on a sequence of size *n* is *O*(*n*log *n*).
- Justification: we say that an invocation of quicksort, on an input sequence of size *m* is "good" if neither L nor G is less than *m*/4.
 - there are m/2 "good" pivots and m/2 "bad" ones
 - The probablility that an invocation is "good" is 1/2
 - Suppose we choose a good pivot at node *v*: the algorithm recurs on sequences with size at most (3/4)*m* each
 - On average, the height of the tree representing a randomized quick-sort is at most $2\log_{4/3} n$
- Total time complexity: $O(n \log n)$

In-Place Quick-Sort

• **Divide step**: *l* scans the sequence from the left, and *r* from the right.



• A swap is performed when *l* is at an element larger than the pivot and *r* is at one smaller than the pivot.





In Place Quick Sort (contd.)

• pseude-code fragment 8.7

How Fast Can We Sort?

- Proposition: The running time of any comparisonbased algorithm for sorting an *n*-element sequence S is Ω(*n*log *n*).
- Justification:
- The running time of a comparison-based sorting algorithm must be equal to or greater than the depth of the decision tree *T* associated with this algorithm.
- Each internal node of *T* is associated with a comparison that establishes the ordering of two elements of S.
- Thus every external node of *T* represents a distinct permutation of the elements of S.
- Hence *T* must have at least *n*! external nodes which again implies T has a height of at least log(*n*!)
- Since n! has at least n/2 terms that are greater than or equal to n/2, we can see:
- $\log(n!) \quad \log(n/2)n/2 = (n/2)\log(n/2)$
- Total Time Complexity: $\Omega(n \log n)$.

How Fast Can We Sort? (contd.)

• A graphical representation of a comparison-based algorithm's decision tree.

