

## SelectionSort

To sort an array of size $n$ :
-Examine $\mathbf{a}[0]$ to $\mathbf{a}[\mathbf{n - 1}$ ]; find the smallest one and swap it with $\mathbf{a}$ [0]
$\square$ Examine $\mathbf{a}[1]$ to $\mathbf{a}[\mathbf{n - 1}$ ]; find the smallest one and swap it with $\mathbf{a}$ [1]
$\square$ In general, in step $i$, examine $\mathbf{a}[\mathbf{i}]$ to $\mathbf{a}[\mathbf{n}-1]$; find the smallest one and swap it with a[i]

This is the other common way for people to sort cards

Runtime

- Worst-case $O\left(n^{2}\right)$
- Best-case O( $\mathrm{n}^{2}$ )
- Expected-case $\mathrm{O}\left(\mathrm{n}^{2}\right)$


## Divide \& Conquer?

$\square$ lt often pays to
$\square$ Break the problem into smaller subproblems,
-Solve the subproblems separately, and then
$\square$ Assemble a final solution
$\square$ This technique is called divide-and-conquer -Caveat: It won't help unless the partitioning and assembly processes are inexpensive
$\square$ Can we apply this approach to sorting?

## MergeSort

```
\square \text { Quintessential divide-and-conquer algorithm}
Divide array into equal parts, sort each part,
then merge
Questions:
    Q1: How do we divide array into two equal parts?
    \square Al: Find middle index: a.length/2
    \squareQ: How do we sort the parts?
    \square A2: call MergeSort recursively!
    \squareQ3: How do we merge the sorted subarrays?
    \square A3: We have to write some (easy) code
```


## Merging Sorted Arrays A and B

- Create an array $\mathbf{C}$ of size $=$ size of $\mathbf{A}+$ size of $\mathbf{B}$
$\square$ Keep three indices:
- $\mathbf{i}$ into A
- $\mathbf{j}$ into $\mathbf{B}$
- k into C
- Initialize all three indices to $\mathbf{0}$ (start of each array)

Compare element $\mathbf{A}$ [ $\mathbf{i}]$ with $\mathbf{B}[\mathbf{j}]$, and move the smaller element into C [k]
Increment $\mathbf{i}$ or $\mathbf{j}$, whichever one we took, and $\mathbf{k}$
When either A or B becomes empty, copy remaining
elements from the other array ( $\mathbf{B}$ or $\mathbf{A}$, respectively) into $\mathbf{C}$


## MergeSort Notes

$\square$ Asymptotic complexity: $O(n \log n)$

- Much faster than $\mathrm{O}\left(\mathrm{n}^{2}\right)$
$\square$ Disadvantage
- Need extra storage for temporary arrays
- In practice, this can be a disadvantage, even though MergeSort is asymptotically optimal for sorting
- Can do MergeSort in place, but this is very tricky (and it slows down the algorithm significantly)
$\square$ Are there good sorting algorithms that do not use so much extra storage? - Yes: QuickSort


## MergeSort Analysis

| $\square$ Outline (detailed code on | Runtime recurrence <br> - Let $T(n)$ be the time to sort an |
| :--- | :--- |
| the website) | array of size $n$ |
| $\square$ Split array into two halves | $T(n)=2 T(n / 2)+O(n)$ |
| $\square$ Recursively sort each half | $T(1)=1$ |


| $\square$ Merge $=$ combine two | Can show by induction that |
| :--- | :--- |
| sorted arrays to make a | T(n) is O(n log $n$ ) |
| single sorted array | Alternately, can see that |
| $\square$ Rule: always choose the smallest item | T(n) is O(n log n) by looking at |
| Time: $O(n)$ where $n$ is the combined size of <br> the two arrays | tree of recursive calls |

Merge $=$ combine two

Rule: always choose the smallest item the two arrays

Runtime recurrence
array of size $n$
$+O(n)$

Can show by induction that $T(n)$ is $O(n \log n)$
$T(n)$ is $O(n \log n)$ by looking at tree of recursive calls

## QuickSort

$\square$ Intuitive idea
$\square$ Given an array $\mathbf{A}$ to sort, choose a pivot value $\mathbf{p}$
$\square$ Partition $\mathbf{A}$ into two subarrays, $\mathbf{A X}$ and $\mathbf{A Y}$

- $\mathbf{A X}$ contains only elements $\leq p$
- AY contains only elements $\geq \mathbf{p}$
$\square$ Sort subarrays $A X$ and $A Y$ separately
$\square$ Concatenate (not merge!) sorted AX and AY to get sorted A
- Concatenation is easier than merging - $\mathrm{O}(1)$



## QuickSort Questions

| $\square$ Key problems | Choosing a pivot |
| :---: | :---: |
| - How should we choose a pirot? | - Ideal pivot is the median, since this splits array in half |
| -How do we partition an array in place? | - Computing the median of an unsorted array is $\mathrm{O}(\mathrm{n})$, but algorithm is quite complicated |
|  | - Popular heuristics: |
| Partitioning in place - Can be done in $\mathrm{O}(\mathrm{n})$ time (next slide) | - Use first value in array (usually not a good choice) |
|  | - Use middle value in array |
|  | - Use median of first, last, and middle values in array |
|  | - Choose a random element |

## In-Place Partitioning



How can we move all the blues to the left of all the reds?

- Keep two indices, LEFT and RIGHT
- Initialize LEFT at start of array and RIGHT at end of array Invariant: all elements to left of LEFT are blue all elements to right of RIGHT are red
- Keep advancing indices until they pass, maintaining invariant

$\square$ Once indices cross, partitioning is done
- If you replace blue with $\leq \mathbf{p}$ and red with $\geq \mathbf{p}$, this is exactly what we need for QuickSort partitioning
$\square$ Notice that after partitioning, array is partially sorted
- Recursive calls on partitioned subarrays will sort subarrays
- No need to copy/move arrays, since we partitioned in place



## QuickSort Analysis

$\square$ Runtime analysis (worst-case)
$\qquad$

- Partition can work badly, producing this:
- Runtime recurrence
- $\mathrm{T}(\mathrm{n})=\mathrm{T}(\mathrm{n}-1)+\mathrm{n}$
- This can be solved to show worst-case $\mathrm{T}(\mathrm{n})$ is $\mathrm{O}\left(\mathrm{n}^{2}\right)$
$\square$ Runtime analysis (expected-case)
G More complex recurrence
- Can solve to show expected $T(n)$ is $O(n \log n)$
$\square$ Improve constant factor by avoiding
QuickSort on small sets
- Switch to InsertionSort (for example) for sets of size, say, $\delta 9$
- Definition of small depends on language, machine, etc

java.lang. Comparable<T> Interface
Lower Bound for Comparison Sorting
- How many input permutations are possible? $n!\sim 2^{\text {n } \log n}$
$\square$ For a comparison-based sorting algorithm to be correct, it must have at least that many leaves in its comparison tree
to have at least $n!\sim 2^{n} \log n$ leaves, it must have height at least $n \log n$ (since it is only binary branching, the number of nodes at most
doubles at every depth)
-therefore its longest path must be of length at least $n \log n$, and that it its worst-case running time


- Suppose we want to sort the elements in an array B[]
- Assume the elements of $\mathbf{B}[$ ] are distinct
- Any permutation of the elements is initially possible
$\square$ When done, $\mathbf{B}[$ ] is sorted
- But the algorithm could not have taken the same path in the comparison tree on different input permutations

public int compareTo(T x);
- Returns a negative, zero, or positive value
- negative if this is before $x$
- 0 if this.equals $(x)$
- positive if this is after $\mathbf{x}$

Many classes implement Comparable

- String, Double, Integer, Character, Date,
- If a class implements Comparable, then its compareTo method is considered to define that class's natural ordering

Comparison-based sorting methods should work with Comparable for maximum generality

