

2. Insertion Sort, Merge Sort

Link: https://www.youtube.com/watch?v=Kg4bqzAqRBM&t=21s&index=5&list=PLSX2U_ZE4Huk19DPn34oZlygPbsig380X

Finding a median array in $A[0:n]$ (unsorted) $\rightarrow B[0:n]$ (sorted) ($B[n/2]$)

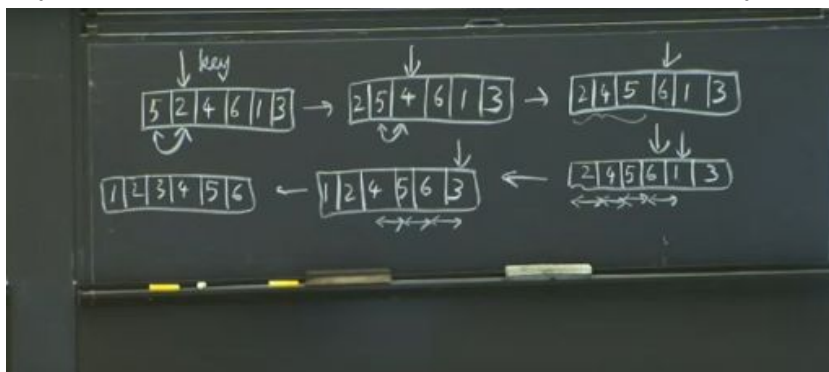
- It's constant time to find the median once you have a sorted list

Uses of Sorting Algorithms

- Binary Search
 - $A[0:n]$ looking for specific item can take linear time if searching one by one
 - $B[0:n]$ takes logarithmic time to complete if array is sorted by dividing the list in half
- Compression Algorithms
 - If data exists multiple times in a file, it can be represented as one item and duplicated upon decompression
 - Document distance
 - Computer graphics

Insertion Sort

- For $i = 1, 2, \dots, n$
 - Insert $A[i]$ into sorted array $A[0:i-1]$ by pair wise swaps down to the correct position
 - Key starts at the second position and moves incrementally

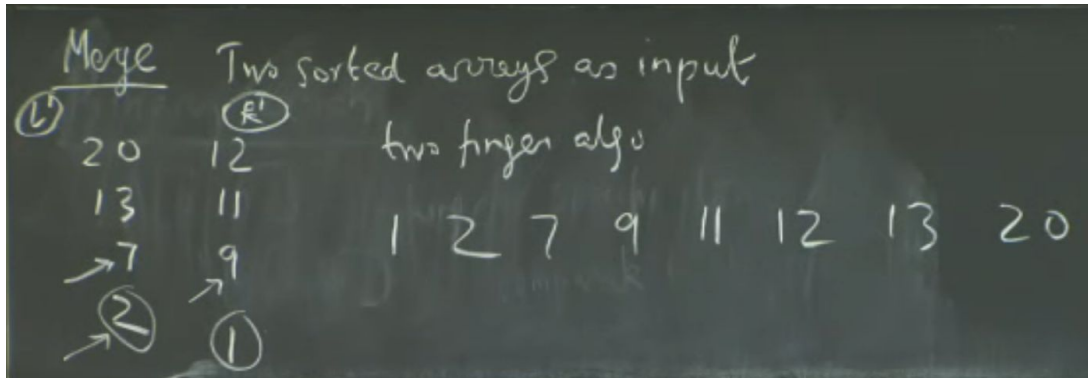


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- $O(n)$ steps (key positions)
- Each step is $O(n)$ comparisons and swaps
- If compares are more computationally expensive than swaps and you're more concerned about $O(n^2)$ comparison cost than $O(n)$ swap cost:
 - You can insert $A[1]$ into sorted array $A[0:i-1]$ by pairwise swaps (binary search) down to the correct position

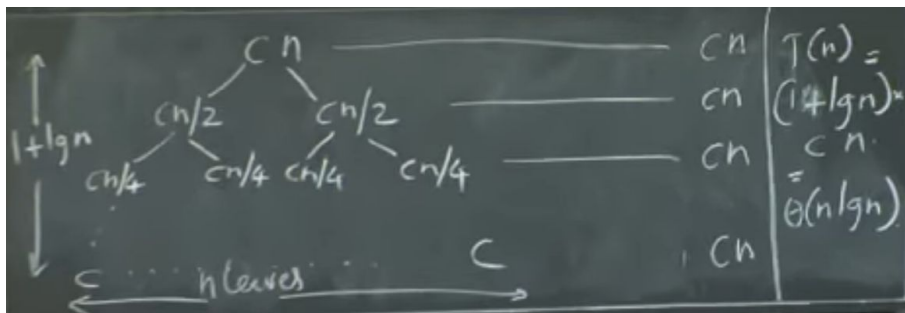
Merge Sort

- Divide and conquer recursive algorithm

- Merge - two sorted arrays as input



- The overall complexity is $O(n \lg n)$
- Complexity $T(n) = C1$ (Divide) + $2T(n/2)$ (Recursion) + $c * n$ (Merge component)
- Proof



- $2T(n/2) + O(n) + O(1)$

Comparison between Insert and Merge Sort

- The advantage insertion sort has over merge sort has to do auxiliary space
 - Merge sort requires $O(n)$ auxiliary space in order to recursively handle the split arrays
 - Insertion sort has $O(1)$ auxiliary space since it's done in place
 - Great for high volume arrays
 - In place merge sort
 - Impractical
 - Merge sort in python = $2.2n \lg(n)$ ms
 - Insertion sort in python = $.2n^2$ ms
 - Insertion sort in c = $0.01n^2$ ms