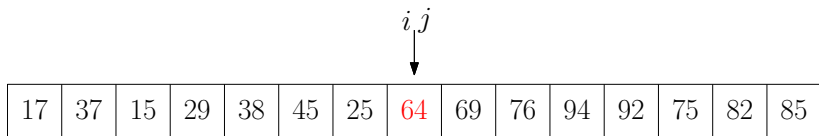


Quicksort Can Be Implemented as an “In-Place” Sorting Algorithm

- In-Place Sorting Algorithm: an algorithm that only uses “small” **extra** space.



- To partition the array into two parts, we only need $O(1)$ extra space.

partition(A, ℓ, r)

- 1: $p \leftarrow$ random integer between ℓ and r , swap $A[p]$ and $A[\ell]$
- 2: $i \leftarrow \ell, j \leftarrow r$
- 3: **while true do**
- 4: **while** $i < j$ and $A[i] < A[j]$ **do** $j \leftarrow j - 1$
- 5: **if** $i = j$ **then break**
- 6: swap $A[i]$ and $A[j]$; $i \leftarrow i + 1$
- 7: **while** $i < j$ and $A[i] < A[j]$ **do** $i \leftarrow i + 1$
- 8: **if** $i = j$ **then break**
- 9: swap $A[i]$ and $A[j]$; $j \leftarrow j - 1$
- 10: **return** i

In-Place Implementation of Quick-Sort

quicksort(A, ℓ, r)

- 1: **if** $\ell \geq r$ **then return**
- 2: $m \leftarrow \text{partition}(A, \ell, r)$
- 3: **quicksort**($A, \ell, m - 1$)
- 4: **quicksort**($A, m + 1, r$)

- To sort an array A of size n , call **quicksort**($A, 1, n$).

Note: We pass the array A by reference, instead of by copying.

Merge-Sort is Not In-Place

- To merge two arrays, we need a third array with size equaling the total size of two arrays

Merge-Sort is Not In-Place

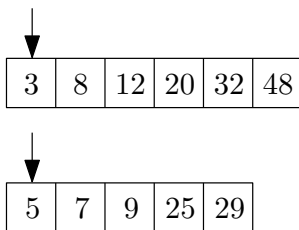
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3	8	12	20	32	48
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5	7	9	25	29
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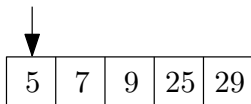
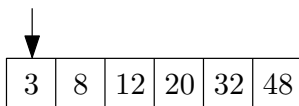
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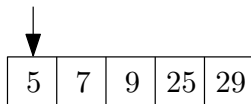
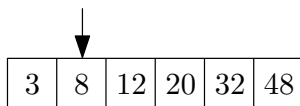
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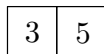
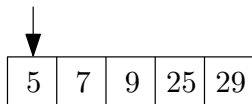
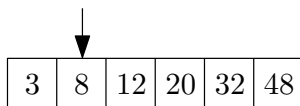
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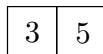
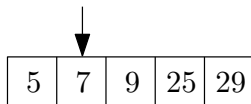
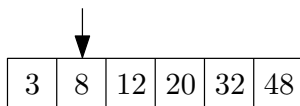
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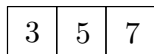
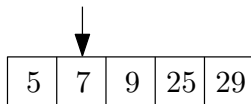
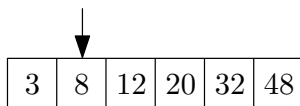
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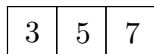
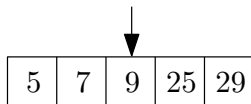
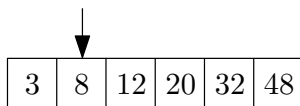
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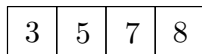
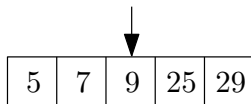
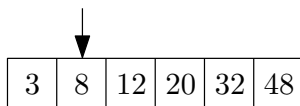
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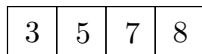
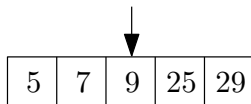
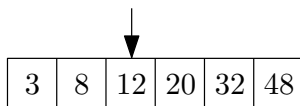
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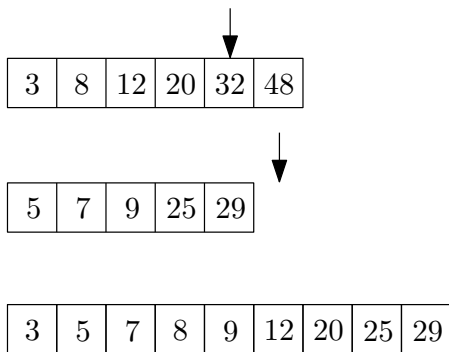
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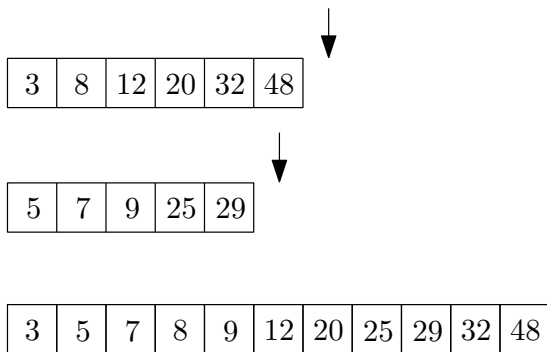
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Outline

- 1 Divide-and-Conquer
- 2 Counting Inversions
- 3 Quicksort and Selection**
 - Quicksort
 - **Lower Bound for Comparison-Based Sorting Algorithms**
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Comparison-Based Sorting Algorithms

Q: Can we do better than $O(n \log n)$ for sorting?

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Comparison-Based Sorting Algorithms

- To sort, we are only allowed to **compare** two elements
- We can not use “internal structures” of the elements

Lemma The (worst-case) running time of any comparison-based sorting algorithm is $\Omega(n \lg n)$.

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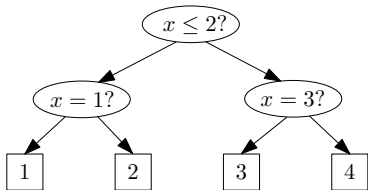
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- Our goal: $O(n)$ running time

Recall: Quicksort with Median Finder

quicksort(A, n)

- 1: **if** $n \leq 1$ **then return** A
- 2: $x \leftarrow$ lower median of A
- 3: $A_L \leftarrow$ elements in A that are less than x ▷ Divide
- 4: $A_R \leftarrow$ elements in A that are greater than x ▷ Divide
- 5: $B_L \leftarrow$ quicksort($A_L, A_L.size$) ▷ Conquer
- 6: $B_R \leftarrow$ quicksort($A_R, A_R.size$) ▷ Conquer
- 7: $t \leftarrow$ number of times x appear A
- 8: **return** the array obtained by concatenating B_L , the array containing t copies of x , and B_R

Selection Algorithm with Median Finder

selection(A, n, i)

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- 5: **if** $i \leq A_L.size$ **then**
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- 7: **else if** $i > n - A_R.size$ **then**
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- Solving recurrence: $T(n) = O(n)$

Randomized Selection Algorithm

selection(A, n, i)

- 1: **if** $n = 1$ **then return** A
- 2: $x \leftarrow$ **random element** of A (called **pivot**)
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- 5: **if** $i \leq A_L.size$ **then**
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- **expected** running time = $O(n)$

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Input: two polynomials of degree $n - 1$

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- **Input:** $(4, -5, 2, 3), (-5, 6, -3, 2)$
- **Output:** $(-20, 49, -52, 20, 2, -5, 6)$

polynomial-multiplication(A, B, n)

- 1: let $C[k] \leftarrow 0$ for every $k = 0, 1, 2, \dots, 2n - 2$
- 2: **for** $i \leftarrow 0$ to $n - 1$ **do**
- 3: **for** $j \leftarrow 0$ to $n - 1$ **do**
- 4: $C[i + j] \leftarrow C[i + j] + A[i] \times B[j]$
- 5: **return** C

Naïve Algorithm

polynomial-multiplication(A, B, n)

- 1: let $C[k] \leftarrow 0$ for every $k = 0, 1, 2, \dots, 2n - 2$
- 2: **for** $i \leftarrow 0$ to $n - 1$ **do**
- 3: **for** $j \leftarrow 0$ to $n - 1$ **do**
- 4: $C[i + j] \leftarrow C[i + j] + A[i] \times B[j]$
- 5: **return** C

Running time: $O(n^2)$

Divide-and-Conquer for Polynomial Multiplication

$$p(x) = 3x^3 + 2x^2 - 5x + 4 = (3x + 2)x^2 + (-5x + 4)$$

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- $p(x)$: degree of $n - 1$ (assume n is even)
- $p(x) = p_H(x)x^{n/2} + p_L(x)$,
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$$pq = (p_Hx^{n/2} + p_L)(q_Hx^{n/2} + q_L)$$

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$$\begin{aligned}pq &= (p_Hx^{n/2} + p_L)(q_Hx^{n/2} + q_L) \\ &= p_Hq_Hx^n + (p_Hq_L + p_Lq_H)x^{n/2} + p_Lq_L\end{aligned}$$