

➤ Quicksort

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Reading: Chapter 7

➤ Aims of this lecture

- To introduce the **QuickSort** algorithm: a popular algorithm which is fast in practice, despite a $\Theta(n^2)$ worst case time.
- To show an **average-case analysis**, revealing why QuickSort is fast in practice.
- To see another example of **divide-and-conquer**.

➤ Idea behind QuickSort

- **Divide:**
 - Pick some element called **pivot**.
 - Move it to its final location in the sorted sequence such that **all smaller elements are to its left, larger ones are to its right**.
- **Conquer:**
 - Recursively sort subarrays for smaller and larger elements
- **Combine:**
 - No work needed here – after the recursion the array is sorted.

➤ QuickSort: The Algorithm

```
QUICKSORT( $A, p, r$ )
```

```
1: if  $p < r$  then
2:    $q = \text{PARTITION}(A, p, r)$ 
3:   QUICKSORT( $A, p, q - 1$ )
4:   QUICKSORT( $A, q + 1, r$ )
```

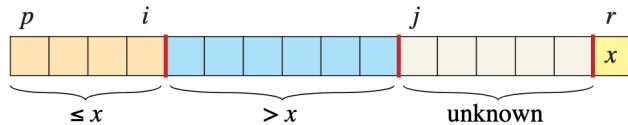
Initial call: $\text{QUICKSORT}(A, 1, A.\text{length})$

Differences to MergeSort:

- Split the array at q , the position of the pivot in sorted array
 - We don't know q in advance, it is revealed by Partition
- No combine step at the end
- Partition plays a similar role to Merge

➤ Partition(A, p, r)

- Rearranges the subarray $A[p..r]$ in place, using swaps
- Takes the last element $A[r]$ as pivot element.
- Idea:
 - Scan the subarray from left to right
 - Build up a subarray $A[p..i]$ of elements smaller or equal to the pivot
 - Build up a subarray $A[i+1..j-1]$ of elements larger than the pivot
 - When reaching the end of the array, put the pivot in the right place



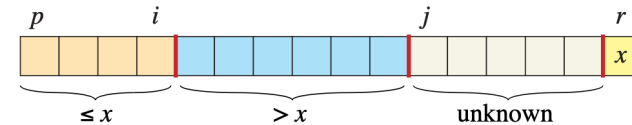
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➤ Partition: Pseudocode

```

PARTITION( $A, p, r$ )
1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i = i + 1$ 
6:     exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 
  
```



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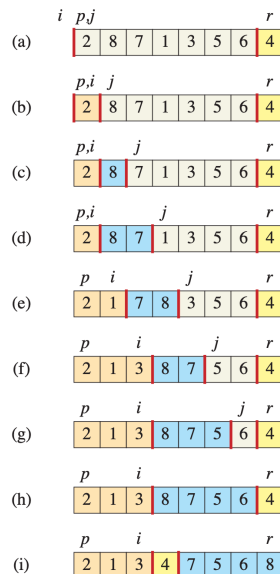
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➤ Partition: Example

```

PARTITION( $A, p, r$ )
1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i = i + 1$ 
6:     exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 
  
```

i : 已经做完的小于等于x的
队列的末尾index;
 j : 已经做完的大于x的队列
的末尾index;

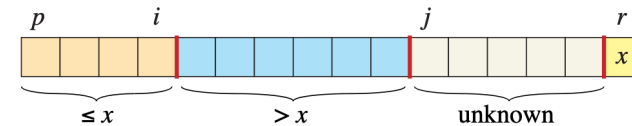


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(step (i) swaps pivot into place, line 7)

➤ Partition: Correctness (1)



```

PARTITION( $A, p, r$ )
1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
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5:      $i = i + 1$ 
6:     exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 
  
```

Loop invariant:
At the beginning of the j -th
iteration:

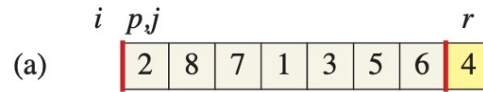
$A[p]..A[i] \leq x$
and
 $A[i+1]..A[j-1] > x$.

- See picture above -

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➤ Partition: Initialisation



PARTITION(A, p, r)

```

1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i = i + 1$ 
6:   exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 

```

Loop invariant:
See picture above –

$$A[p]..A[i] \leq x$$

and

$$A[i + 1]..A[j - 1] > x.$$

Trivially true at initialisation.
(both sets are empty)

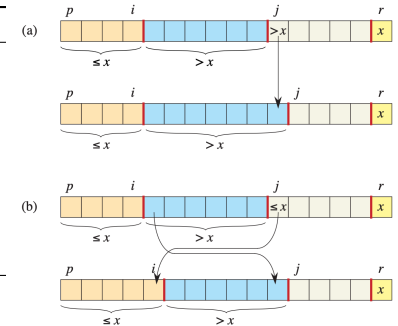
➤ Partition: Maintaining the loop invariant

PARTITION(A, p, r)

```

1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i = i + 1$ 
6:   exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 

```



Maintenance:

- If line 4 is false : picture (a)
- If line 4 true: picture (b)
- In both cases after one iteration of j the loop invariant is maintained.

Loop invariant:

$$A[p]..A[i] \leq x$$

and

$$A[i + 1]..A[j - 1] > x.$$

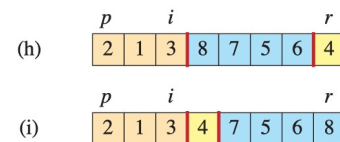
➤ Partition: termination

PARTITION(A, p, r)

```

1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i = i + 1$ 
6:   exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 

```



Loop invariant:

$$A[p]..A[i] \leq x$$

and

$$A[i + 1]..A[j - 1] > x.$$

Termination:

After the last swap in line 7,
 $A[p]..A[i] \leq x < A[i + 2]..A[r]$
 and Partition returns the position of x .

➤ Exercise: Analyse the Runtime of Partition

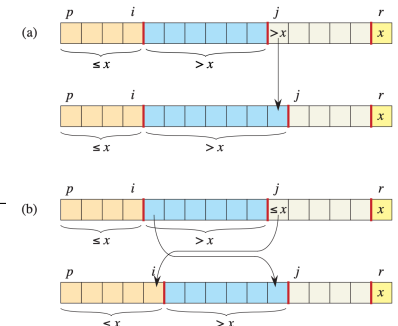
Q: What is the runtime of Partition on a subarray of size n ?

PARTITION(A, p, r)

```

1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i = i + 1$ 
6:   exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 

```



➤ QuickSort: The Algorithm

QUICKSORT(A, p, r)

```

1: if  $p < r$  then
2:    $q = \text{PARTITION}(A, p, r)$ 
3:   QUICKSORT( $A, p, q - 1$ )
4:   QUICKSORT( $A, q + 1, r$ )

```

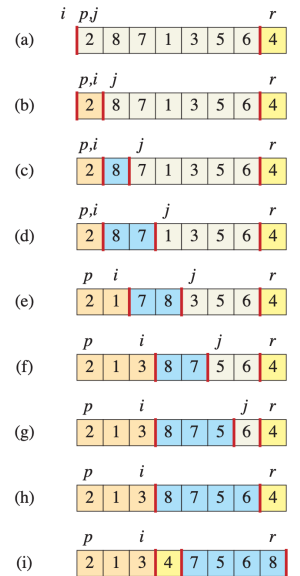
PARTITION(A, p, r)

```

1:  $x = A[r]$ 
2:  $i = p - 1$ 
3: for  $j = p$  to  $r - 1$  do
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5:      $i = i + 1$ 
6:     exchange  $A[i]$  with  $A[j]$ 
7: exchange  $A[i + 1]$  with  $A[r]$ 
8: return  $i + 1$ 

```

Runtime?



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➤ Worst-case and Best-case Partitionings

- The overall runtime depends on **how the array is partitioned** as that determines the sizes $q - 1$ and $r - q$ of the subarray to be sorted recursively.
 - Recall that we don't know in advance where the pivot will end up.
- Questions:
 - What might be a **worst-case partitioning** for the runtime?
 - What might be a **best-case partitioning** for the runtime?

QUICKSORT(A, p, r)

```

1: if  $p < r$  then
2:    $q = \text{PARTITION}(A, p, r)$ 
3:   QUICKSORT( $A, p, q - 1$ )
4:   QUICKSORT( $A, q + 1, r$ )

```

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➤ Worst-case Partitioning

- The worst case is attained when Partition always produces one subproblem with $n - 1$ and one with 0 elements.
- This is the case, for example, when the array is already sorted.
- This leads to the following recurrence:

$$\begin{aligned}
 T(n) &= T(n - 1) + T(0) + \Theta(n) \\
 &= T(n - 1) + \Theta(n).
 \end{aligned}$$

- Solving this gives $T(n) = \Theta(n^2)$.

➤ Best-case Partitioning

- Best case: split into two subproblems of sizes $\left\lfloor \frac{n}{2} \right\rfloor$ and $\left\lceil \frac{n}{2} \right\rceil - 1$.
- Ignoring floors, ceilings, and -1 we get the recurrence:

$$T(n) = 2T(n/2) + \Theta(n)$$

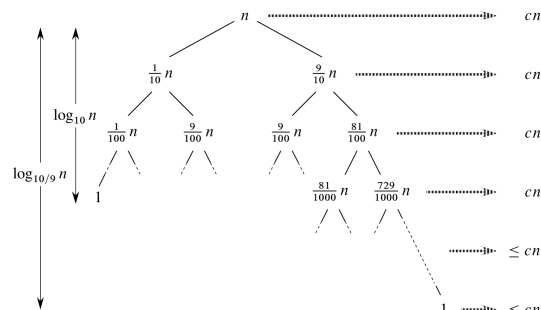
- Deja vu?
- This is $\Theta(n \log n)$ from the analysis of MergeSort.
- True to the spirit of divide-and-conquer.

➤ **Towards an average case**

- What if the split was always $\frac{9}{10} \cdot n$ and $\frac{1}{10} \cdot n$?

- Getting the recurrence

$$T(n) = T(9n/10) + T(n/10) + cn$$



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 $O(n \lg n)$

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➤ Average case analysis

- Assume all elements of the array are distinct.
- Assume each split $q = 1, 2, \dots, n$ was equally likely.
- This situation occurs when the input is chosen **uniformly at random** amongst all $n! = 1 \cdot 2 \cdot 3 \cdot \dots \cdot n$ possible orderings.
- Then

$$\begin{aligned} T(n) &= \frac{1}{n} \cdot \sum_{q=1}^n (T(q-1) + T(n-q) + \Theta(n)) \\ &= \frac{1}{n} \cdot \sum_{q=1}^n T(q-1) + \frac{1}{n} \cdot \sum_{q=1}^n T(n-q) + \frac{1}{n} \cdot \sum_{q=1}^n \Theta(n) \\ &= \frac{1}{n} \cdot \sum_{k=0}^{n-1} 2T(k) + \Theta(n) \end{aligned}$$

- Average over all problem sizes for 2 subproblems $+\Theta(n)$.
- Solving this recurrence gives a bound of $O(n \log n)$.
- We prove this next!

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➤ Average case analysis (2): Substitution method

To prove:

$$\frac{1}{n} \cdot \sum_{k=0}^{n-1} 2T(k) + \Theta(n) \leq \mathbf{c} \, n \ln n$$

Base case: $n=2$ Prove: $T(2) \leq c2 \ln 2$

$$T(2) = T(0) + T(1) + \Theta(2) = 2c' + c^* \leq c 2 \ln 2 \quad (\text{for e.g., } c > 2c' + c^*)$$

Inductive case: Assume true for $\leq n$ ($T(k) \leq c k \ln k$ for $k \leq n$) and prove for $n+1$

$$T(n) = \frac{2}{n} \left[T(0) + T(1) + \sum_{k=2}^{n-1} c k \ln k \right] + \Theta(n)$$

$$\leq \frac{2}{n} [c' + c' + \sum_{k=2}^{n-1} c k \ln k] + \Theta(n) =$$

$$= \frac{4c'}{n} + \left(\frac{2c}{n} \sum_{k=2}^{n-1} k \ln k \right) + \Theta(n) \leq \frac{4c'}{n} + \frac{2c}{n} \left[\frac{n^2 \ln n}{2} - \frac{n^2}{4} \right] + \Theta(n)$$

$$= cn \ln n - \frac{cn}{2} + \frac{4c'}{n} + \Theta(n) < cn \ln n \Leftrightarrow \frac{cn}{2} > \frac{4c'}{n} + c^* n, \text{ (e.g. for } c > 3c^*)$$

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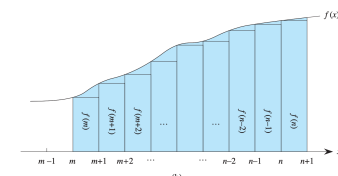
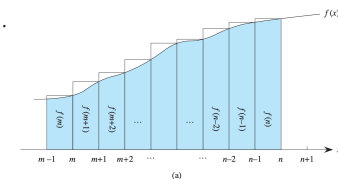
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➤ Average case analysis (3)

$$\sum_{k=2}^{n-1} k \ln k \leq \int_2^n k \ln k \, dk \leq \left[\frac{n^2 \ln n}{2} - \frac{n^2}{4} \right]$$

When a summation has the form $\sum_{k=m}^n f(k)$, where $f(k)$ is a monotonically increasing function, you can approximate it by integrals:

$$\int_{m-1}^n f(x) dx \leq \sum_{k=m}^n f(k) \leq \int_m^{n+1} f(x) dx.$$



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➤ Improvements to QuickSort

- QuickSort is fast in practice because of small constants in the asymptotic running time.
- **Improvements for handling equal values** (exercise)
 - Partition into smaller, equal and larger elements
 - Only need to sort smaller and larger subarrays
- **Choose the pivot as median of 3 elements**
 - Slightly faster in practice, but still quadratic worst case
- **Dual-Pivot QuickSort** by Vladimir Yaroslavskiy
 - Use two pivots instead of one and partition array in 3 areas
 - Used in Java 7

➤ Summary

- QuickSort is used in modern programming languages
 - QuickSort has a bad worst-case runtime of $\Theta(n^2)$
 - Average-case performance on **random inputs** is $O(n \log n)$.
- Why is it popular?
 - Constants hidden in the asymptotic terms are small.
- Next week we'll see how randomisation allows to avoid the worst case runtime

➤ A Randomised Version of QuickSort

- Choosing the right pivot element can be tricky – we have no idea *a priori* which pivot elements are good.
- **Solution: leave it to chance!**

```
RANDOMISED-PARTITION( $A, p, r$ )
1:  $i = \text{RANDOM}(p, r)$ 
2: exchange  $A[r]$  with  $A[i]$ 
3: return PARTITION( $A, p, r$ )

RANDOMISED-QUICKSORT( $A, p, r$ )
1: if  $p < r$  then
2:    $q = \text{RANDOMISED-PARTITION}(A, p, r)$ 
3:   RANDOMISED-QUICKSORT( $A, p, q - 1$ )
4:   RANDOMISED-QUICKSORT( $A, q + 1, r$ )
```

"Random" picks pivot uniformly at random among all elements.