# طراحی الگوریتم ها (CE221)

جلسه ششم: مرتب سازی سریع

> سجاد شیرعلی شهرضا بهار 1401 *دوشنبه، 2 اسفند 1400*

# اطلاع رساني

بخش مرتبط کتاب برای این جلسه: 5

مرتب سازی سریع

یک نمونه واقعی و کاربردی از الگوریتم های تصادفی

#### QUICKSORT OVERVIEW

**EXPECTED RUNNING TIME** 

O (n log n)

**WORST-CASE RUNNING TIME** 

 $O(n^2)$ 

#### QUICKSORT OVERVIEW

#### **EXPECTED RUNNING TIME**

O (n log n)

#### **WORST-CASE RUNNING TIME**

 $O(n^2)$ 

In practice, it works great! It's competitive with MergeSort (& often better in some contexts!), and it runs *in place* (no need for lots of additional memory)

#### Let's use DIVIDE-and-CONQUER again!

Select a pivot at random

Partition around it

Recursively sort L and R!

Select a pivot



Select a pivot

3 2 7 6 1 5 4 8

Pick this pivot uniformly at random!

Partition around it

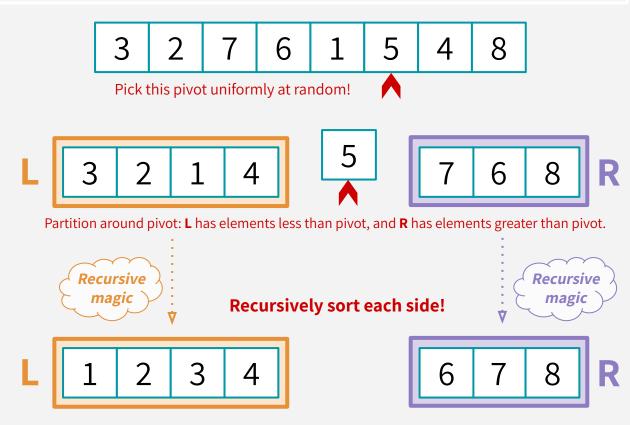


Partition around pivot: L has elements less than pivot, and R has elements greater than pivot.

Select a pivot

Partition around it

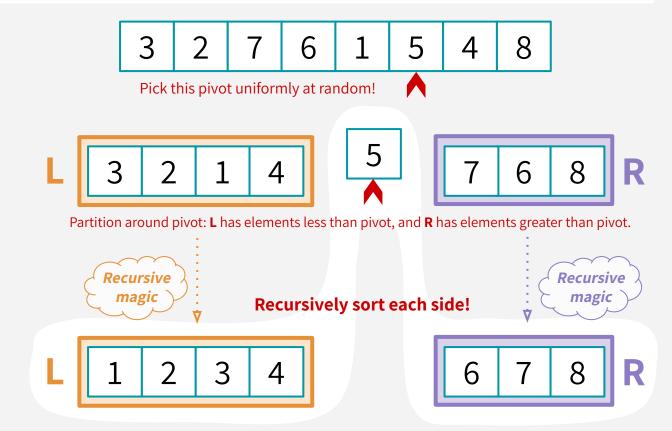
Recurse!



Select a pivot

Partition around it

Recurse!



#### QUICKSORT: PSEUDO-PSEUDOCODE

```
QUICKSORT(A):
    if len(A) <= 1:</pre>
        return
    pivot = random.choice(A)
    PARTITION A into:
        L (less than pivot) and
        R (greater than pivot)
    Replace A with [L, pivot, R]
    QUICKSORT(L)
    QUICKSORT(R)
```

#### RECURRENCE RELATION

```
QUICKSORT(A):
    if len(A) <= 1:
        return
    pivot = random.choice(A)
    PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
    Replace A with [L, pivot, R]
    QUICKSORT(L)
    QUICKSORT(R)
```

# Recurrence Relation for QUICKSORT

$$T(n) = T(|L|) + T(|R|) + O(n)$$
  
 $T(0) = T(1) = O(1)$ 

#### IDEAL RUNTIME?

```
QUICKSORT(A):
    if len(A) <= 1:
        return
   pivot = random.choice(A)
   PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
   Replace A with [L, pivot, R]
   QUICKSORT(L)
   QUICKSORT(R)
```

# Recurrence Relation for QUICKSORT

$$T(n) = T(|L|) + T(|R|) + O(n)$$
  
 $T(0) = T(1) = O(1)$ 

In an ideal world, the pivot would split the array exactly in half, and we'd get:

$$T(n) = T(n/2) + T(n/2) + O(n)$$

#### IDEAL RUNTIME?

```
Recurrence Relation for
QUICKSORT(A):
                                                  QUICKSORT
    if len(A) <= 1:
        return
                         In an ideal world:
                                                      + T(|R|) + O(n)
    pivot = random
                                                      T(1) = O(1)
    PARTITION A ir
                       T(n) = 2 \cdot T(n/2) + O(n)
        L (less th
                          T(n) = O(n \log n)
        R (greater
                                                      the pivot would split the
    Replace A with LL, pivot, KJ
                                            array exactly in half, and we'd get:
    QUICKSORT(L)
                                         T(n) = T(n/2) + T(n/2) + O(n)
    QUICKSORT(R)
```

#### **WORST-CASE RUNTIME**

```
QUICKSORT(A):
    if len(A) <= 1:
        return
    pivot = random.choice(A)
    PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
    Replace A with [L, pivot, R]
    QUICKSORT(L)
    QUICKSORT(R)
```

# Recurrence Relation for QUICKSORT

$$T(n) = T(|L|) + T(|R|) + O(n)$$
  
 $T(0) = T(1) = O(1)$ 

#### WORST-CASE RUNTIME

```
QUICKSORT(A):
    if len(A) <= 1:
        return
   pivot = random.choice(A)
   PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
   Replace A with [L, pivot, R]
   QUICKSORT(L)
   QUICKSORT(R)
```

# Recurrence Relation for QUICKSORT

$$T(n) = T(|L|) + T(|R|) + O(n)$$
  
 $T(0) = T(1) = O(1)$ 

With the unluckiest randomness, the pivot would be either min(A) or max(A):

$$T(n) = T(0) + T(n-1) + O(n)$$

#### WORST-CASE RUNTIME

```
Recurrence Relation for
QUICKSORT(A):
                                                    QUICKSORT
    if len(A) \ll 1:
        return
                 With the worst "randomness"
                                                           T(|R|) + O(n)
    pivot = ra
                                                            = O(1)
    PARTITION
                          T(n) = T(n-1) + O(n)
        L (less
                              T(n) = O(n^2)
        R (grea
                                                           domness, the pivot
                          (recursion tree/table or substitution method!)
    Replace A w.
                                                          nin(A) or max(A):
    QUICKSORT(L)
                                            T(n) = T(0) + T(n-1) + O(n)
    QUICKSORT(R)
```



#### AN **INCORRECT** PROOF:

#### AN **INCORRECT** PROOF:

• E[|L|] = E[|R|] = (n-1)/2

## AN ASIDE: why is E[|L|] = (n-1)/2?

$$E[|L|] = E[|R|]$$
 (by symmetry)

$$E[|L| + |R|] = n - 1$$

(because L and R make up everything except the pivot)

$$E[|L|] + E[|R|] = n - 1$$

(by linearity of expectation)

$$2 \cdot E[|L|] = n - 1$$

(plugging the first line)

$$E[|L|] = (n - 1)/2$$
(Solving for E[|L|])

#### AN **INCORRECT** PROOF:

- E[|L|] = E[|R|] = (n-1)/2
- If this occurs, then T(n) = T(|L|) + T(|R|) + O(n) could be written as T(n) = 2T(n/2) + O(n).

#### AN **INCORRECT** PROOF:

- E[|L|] = E[|R|] = (n-1)/2
- If this occurs, then T(n) = T(|L|) + T(|R|) + O(n) could be written as T(n) = 2T(n/2) + O(n).
- Therefore, the expected running time is O(n log n)!

#### AN **INCORRECT** PROOF:

- E[|L|] = E[|R|] = (n-1)/2
- If this occurs, then T(n) = T(|L|) + T(|R|) + O(n) could be written as T(n) = 2T(n/2) + O(n).
- Therefore, the expected running time is O(n log n)!

#### Why is this wrong?

Well, for starters, we can use the exact same argument to prove something false...

```
SLOW SORT(A):
   if len(A) <= 1:</pre>
       return randomly choose either!
   pivot = either max(A) OR min(A)
   PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
   Replace A with [L, pivot, R]
    SLOW SORT(L)
    SLOW SORT(R)
```

```
SLOW SORT(A):
   if len(A) <= 1:
       return
                      randomly choose either!
   pivot = either max(A) or min(A)
   PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
   Replace A with [L, pivot, R]
   SLOW SORT(L)
   SLOW SORT(R)
```

## Recurrence Relation for SLOW SORT

$$T(n) = T(|L|) + T(|R|) + O(n)$$
  
 $T(0) = T(1) = O(1)$ 

```
SLOW SORT(A):
   if len(A) <= 1:
       return
                      randomly choose either!
   pivot = either max(A) or min(A)
   PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
   Replace A with [L, pivot, R]
    SLOW SORT(L)
    SLOW SORT(R)
```

# Recurrence Relation for SLOW SORT

$$T(n) = T(|L|) + T(|R|) + O(n)$$
  
 $T(0) = T(1) = O(1)$ 

Same recurrence relation!

We also still have:

$$E[|L|] = E[|R|] = (n-1)/2$$

But now, one of |L| or |R| is always n-1 & the runtime is  $\Theta(n^2)$ , with probability 1

#### **SLOW** SORT(A): if len(A) return pivot = e**PARTITION** L (les R (gre Replace A **SLOW SORT** SLOW SORT (R)

# Recurrence Relation for SORT

We could use the exact same (incorrect) proof to prove that **SLOWSort** has expected runtime **O(n log n)**, when it actually has expected runtime of  $\Theta(n^2)$ ...

ll have: ] = (n-1)/2

& the runtime is ⊖(n²), with probability 1

#### AN **INCORRECT** PROOF:

- E[|L|] = E[|R|] = (n-1)/2
- If this occurs, then T(n) = T(|L|) + T(|R|) + O(n) could be written as T(n) = 2T(n/2) + O(n).
- Therefore, the expected running time is O(n log n)!

Why is this wrong?

AM

#### Basically:

E[f(x)] is *not necessarily* the same as f(E[x])

e.g.  $E[X^2]$  is not the same as  $(E[X])^2$ 

We were reasoning about T(E[x]) instead of E[T(x)]

wny is this wrong:

Instead, to prove that the expected runtime of QuickSort is O(n log n), we're going to count the **number of comparisons** that this algorithm performs, and take the expectation of that!

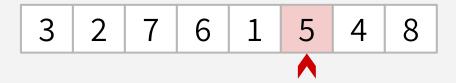
How many times are any two items compared?



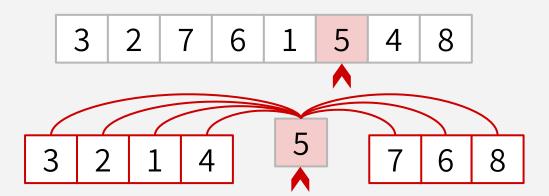
# امید ریاضی زمان اجرای مرتب سازی سریع

راه حل درست: تعداد مورد انتظار مقایسه دو عنصر با همدیگر چند بار است؟

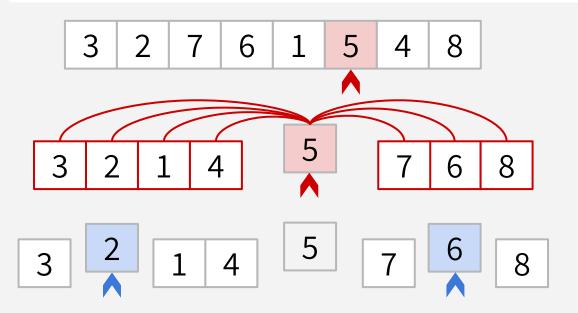
#### HOW MANY COMPARISONS?



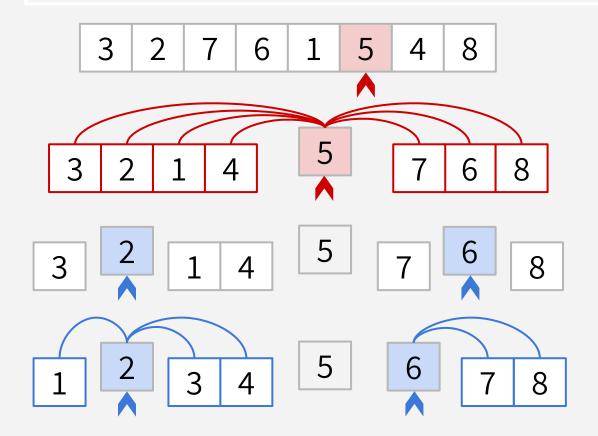
#### HOW MANY COMPARISONS?



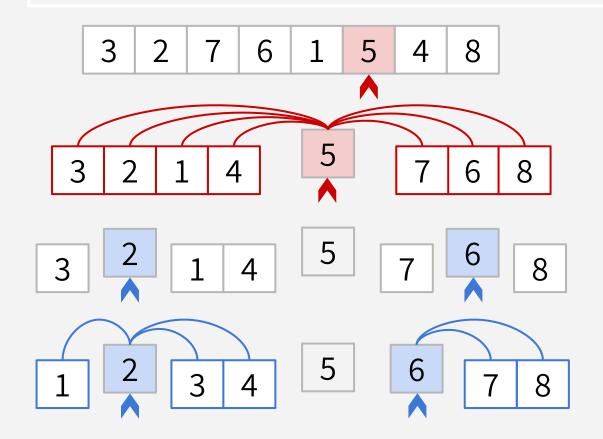
Everything is compared to 5 once in this first step... and then never again with **5**.



Everything is compared to 5 once in this first step... and then never again with **5**.



Everything is compared to 5 once in this first step... and then never again with **5**.



Everything is compared to 5 once in this first step... and then never again with **5**.

Only 1, 3, & 4 are compared to **2**.

And only 7 & 8 are compared with **6**.

No comparisons ever happen between two numbers on opposite sides of 5.



Seems like whether or not two elements are compared has something to do with pivots...



Everything is compared to 5 once in this first step... and then never again with **5**.

Only 1, 3, & 4 are compared to **2**.

And only 7 & 8 are compared with **6**.

No comparisons ever happen between two numbers on opposite sides of 5.

Each pair of elements is compared either **0** or **1** times.

Let  $\mathbf{X}_{\mathbf{a},\mathbf{b}}$  be a Bernoulli/indicator random variable such that:

$$X_{a,b} = 1$$

if **a** and **b** are compared

$$X_{a,b} = 0$$

otherwise

Each pair of elements is compared either **0** or **1** times.

Let  $\mathbf{X}_{\mathbf{a},\mathbf{b}}$  be a Bernoulli/indicator random variable such that:

$$X_{a,b} = 1$$
 if **a** and **b** are compared

$$X_{a,b} = 0$$
 otherwise

In our example,  $\mathbf{X}_{2,5}$  took on the value **1** since **2** and **5** were compared. On the other hand,  $\mathbf{X}_{3,7}$  took on the value **0** since **3** and **7** are *not* compared.

Each pair of elements is compared either **0** or **1** times.

Let  $\mathbf{X}_{\mathbf{a},\mathbf{b}}$  be a Bernoulli/indicator random variable such that:

$$X_{a,b} = 1$$

 $X_{a,b} = 1$  if **a** and **b** are compared  $X_{a,b} = 0$  otherwise

$$X_{a,b} = 0$$

In our example,  $X_{2.5}$  took on the value **1** since **2** and **5** were compared. On the other hand,  $X_{3,7}$  took on the value **0** since **3** and **7** are *not* compared.

#### **Total number of comparisons =**

$$\mathbb{E}\left[\sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}X_{a,b}\right] = \sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}\mathbb{E}\left[X_{a,b}\right]$$

Each pair of elements is compared either **0** or **1** times.

Let  $\mathbf{X}_{\mathbf{a},\mathbf{b}}$  be a Bernoulli/indicator random variable such that:

$$X_{a,b} = 1$$
 if **a** and **b** are compared

otherwise

In our example,  $X_{2.5}$  took on the value **1** since **2** and **5** were compared. On the other hand,  $X_{3,7}$  took on the value **0** since **3** and **7** are *not* compared.

#### **Total number of comparisons =**

$$\mathbb{E}\left[\sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}X_{a,b}\right] = \sum_{\substack{\text{by linearity of}\\ \text{expectation!}}}^{n-2}\sum_{a=0}^{n-1}\sum_{b=a+1}^{n-1}\mathbb{E}\left[X_{a,b}\right]$$

$$\sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}\mathbb{E}$$

We need to figure out this value!

So, what's  $E[X_{a,b}]$ ?

$$E[X_{a,b}] = 1 \cdot P(X_{a,b} = 1) + 0 \cdot P(X_{a,b} = 0) = P(X_{a,b} = 1)$$

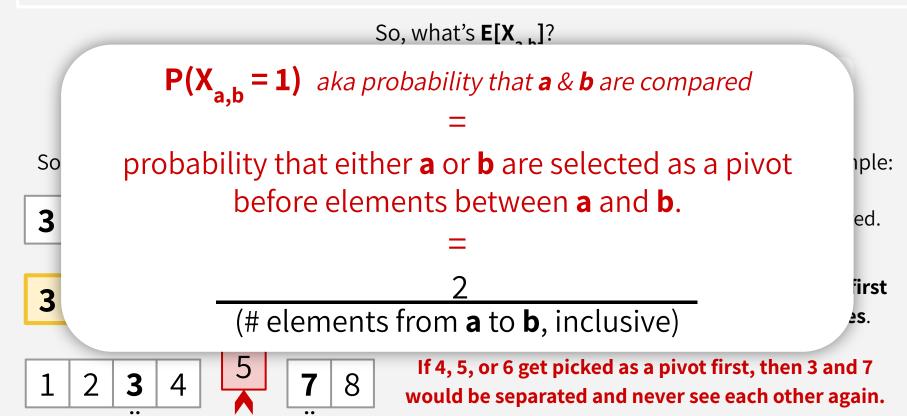
So, what's  $P(X_{a,b} = 1)$ ? It's the probability that **a** and **b** are compared. Consider this example:

 $P(X_{3,7} = 1)$  is the probability that 3 and 7 are compared.



This is exactly the probability that either 3 or 7 is first picked to be a pivot out of the highlighted entries.

If 4, 5, or 6 get picked as a pivot first, then 3 and 7 would be separated and never see each other again.



So, what's **E[X, L]**?

$$P(X_{a,b} = 1)$$
 aka probability that  $a \& b$  are compared

probability that either **a** or **b** are selected as a pivot before elements between **a** and **b**.

$$\frac{2}{\mathbf{b} - \mathbf{a} + 1}$$

So

3

1 2 3 4 5 7 8 If 4, 5, or 6 get picked as a pivot first, then 3 and 7 would be separated and never see each other again.

ıple:

ed.

irst

2S.

Total number of comparisons =

$$\sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}\mathbb{E}ig[X_{a,b}ig]$$

Total number of comparisons =

$$\sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}\mathbb{E}ig[X_{a,b}ig] = \sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}rac{2}{b-a+1}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Total number of comparisons =

$$egin{aligned} \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} \mathbb{E}ig[X_{a,b}ig] &= \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} rac{2}{b-a+1} \ &= \sum_{a=0}^{n-2} \sum_{c=1}^{n-a-1} rac{2}{c+1} \end{aligned}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Introduce c = b – a to make notation nicer

Total number of comparisons =

$$egin{align} \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} \mathbb{E}ig[X_{a,b}ig] &= \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} rac{2}{b-a+1} \ &= \sum_{a=0}^{n-2} \sum_{c=1}^{n-a-1} rac{2}{c+1} \ &\leq \sum_{a=0}^{n-1} \sum_{c=1}^{n-1} rac{2}{c+1} \end{aligned}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Introduce c = b - a to make notation nicer

Increase summation limits to make them nicer (hence the ≤)

Total number of comparisons =

$$egin{align} \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} \mathbb{E}ig[X_{a,b}ig] &= \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} rac{2}{b-a+1} \ &= \sum_{a=0}^{n-2} \sum_{c=1}^{n-a-1} rac{2}{c+1} \ &\leq \sum_{a=0}^{n-1} \sum_{c=1}^{n-1} rac{2}{c+1} \ &= 2n \sum_{1}^{n-1} rac{1}{c+1} \end{aligned}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Introduce c = b – a to make notation nicer

Increase summation limits to make them nicer (hence the ≤)

Nothing in the summation depends on a, so pull 2 out

Total number of comparisons =

$$egin{align} \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} \mathbb{E}ig[X_{a,b}ig] &= \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} rac{2}{b-a+1} \ &= \sum_{a=0}^{n-2} \sum_{c=1}^{n-a-1} rac{2}{c+1} \ &\leq \sum_{a=0}^{n-1} \sum_{c=1}^{n-1} rac{2}{c+1} \ &= 2n \sum_{c=1}^{n-1} rac{1}{c+1} \ &\leq 2n \sum_{c=1}^{n-1} rac{1}{c} \end{aligned}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Introduce c = b – a to make notation nicer

Increase summation limits to make them nicer (hence the ≤)

Nothing in the summation depends on a, so pull 2 out

decrease each denominator → we get the harmonic series!

Total number of comparisons =

$$egin{aligned} \sum_{=a+1}^{n-1} \mathbb{E}ig[X_{a,b}ig] &= \sum_{a=0}^{n-2} \sum_{b=a+1}^{n-1} rac{2}{b-a+1} \ &= \sum_{a=0}^{n-2} \sum_{c=1}^{n-a-1} rac{2}{c+1} \ &\leq \sum_{a=0}^{n-1} \sum_{c=1}^{n-1} rac{2}{c+1} \ &= 2n \sum_{c=1}^{n-1} rac{1}{c+1} \ &\leq 2n \sum_{c=1}^{n-1} rac{1}{c} \ &= O(n \log n) \end{aligned}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Introduce c = b – a to make notation nicer

Increase summation limits to make them nicer (hence the ≤)

Nothing in the summation depends on a, so pull 2 out

decrease each denominator → we get the harmonic series!

Total number of comparisons =

$$\sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}\mathbb{E}ig[X_{a,b}ig] = \sum_{a=0}^{n-2}\sum_{b=a+1}^{n-1}rac{2}{b-a+1}$$

If E[ # comparisons ] = O(n log n), does this mean E[ running time ] is also O(n log n)?

YES! Intuitively, the runtime is dominated by comparisons.

$$egin{aligned} &= \sum_{a=0}^{n-2} \sum_{c=1}^{n-a-1} rac{2}{c+1} \ &\leq \sum_{a=0}^{n-1} \sum_{c=1}^{n-1} rac{2}{c+1} \ &= 2n \sum_{c=1}^{n-1} rac{1}{c+1} \ &\leq 2n \sum_{c=1}^{n-1} rac{1}{c} \ &= O(n \log n) \end{aligned}$$

We just computed  $E[X_{a,b}] = P(X_{a,b} = 1)$ 

Introduce c = b – a to make notation nicer

Increase summation limits to make them nicer (hence the ≤)

Nothing in the summation depends on a, so pull 2 out

decrease each denominator → we get the harmonic series!

## QUICKSORT

```
QUICKSORT(A):
    if len(A) <= 1:
        return
    pivot = random.choice(A)
    PARTITION A into:
       L (less than pivot) and
       R (greater than pivot)
    Replace A with [L, pivot, R]
    OUICKSORT(L)
    OUICKSORT(R)
```

Worst case runtime: **O(n<sup>2</sup>)** 

Expected runtime: O(n log n)

## BETTER WORST CASE RUNTIME

- Select a better pivot
  - Ideally, split the array into two equal parts
  - Select the median as pivot
- If the pivot is median, then we will have:
  - $\circ$  T(n) = 2T(n/2) + O(n) = O(n log n)
- How to select the median in O(n)?
  - Will see how to do it next week



مرتب سازی سریع در عمل

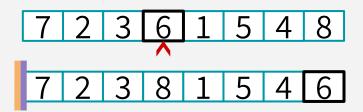
چگونگی پیاده سازی (و آیا واقعا کسی از آن استفاده می کند؟)

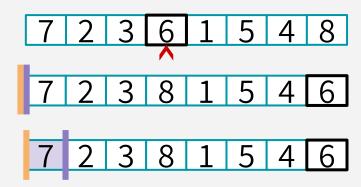
## IMPLEMENTING QUICKSORT

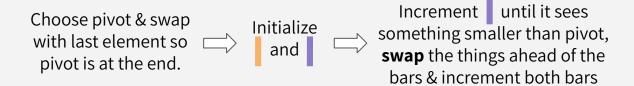
In practice, a more clever approach is used to implement PARTITION, so that the entire QuickSort algorithm can be implemented "in-place" (i.e. via swaps, rather than constructing separate L or R subarrays)

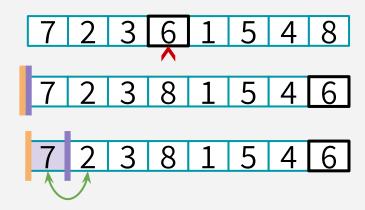
7 2 3 6 1 5 4 8

Choose pivot & swap with last element so pivot is at the end.



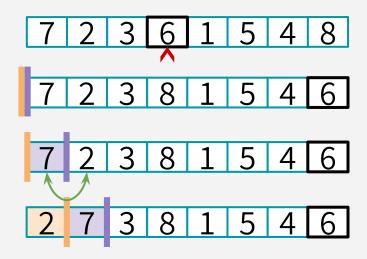






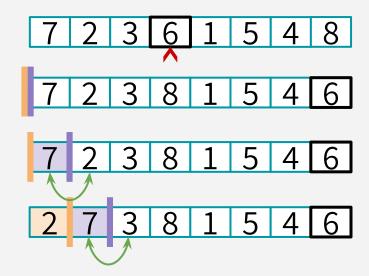
Choose pivot & swap with last element so pivot is at the end.





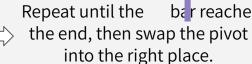
Choose pivot & swap with last element so pivot is at the end.

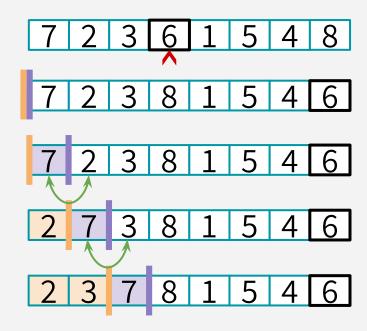




Choose pivot & swap with last element so pivot is at the end.

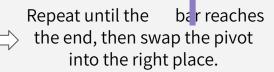


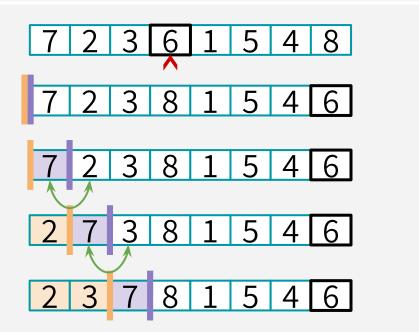




Choose pivot & swap with last element so pivot is at the end.

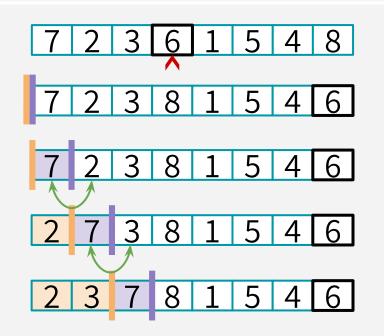






Choose pivot & swap with last element so pivot is at the end.

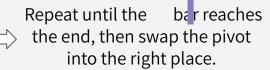


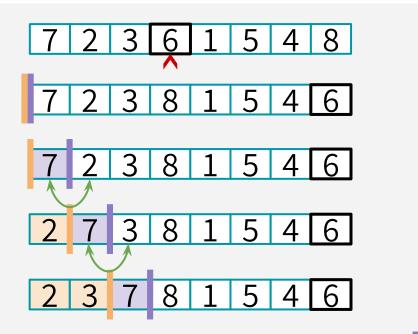


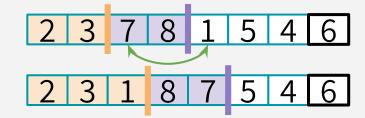
23781546

Choose pivot & swap with last element so pivot is at the end.









Choose pivot & swap with last element so pivot is at the end.

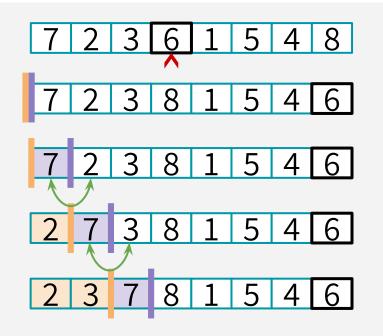


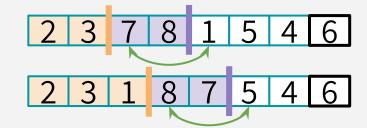
 $\qquad \qquad \Longrightarrow \qquad$ 

Increment until it sees something smaller than pivot, **swap** the things ahead of the bars & increment both bars



Repeat until the bar reaches the end, then swap the pivot into the right place.





Choose pivot & swap with last element so pivot is at the end.

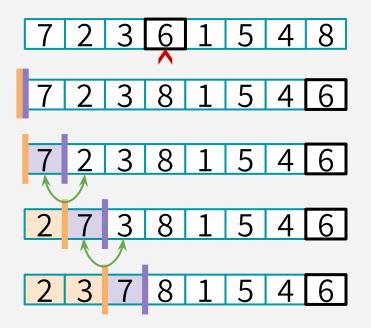


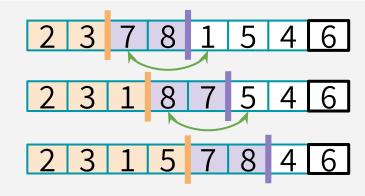
Initialize and

Increment until it sees something smaller than pivot, **swap** the things ahead of the bars & increment both bars



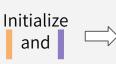
Repeat until the bar reaches the end, then swap the pivot into the right place.

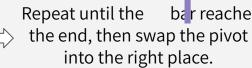


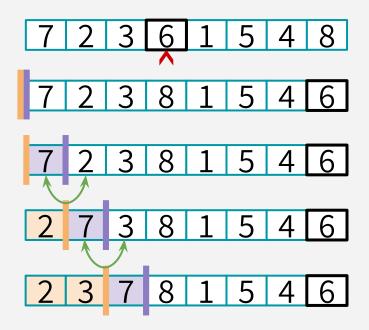


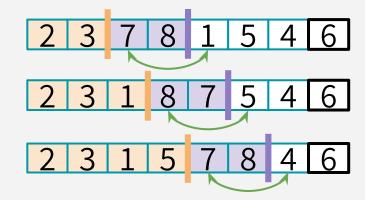
Choose pivot & swap with last element so pivot is at the end.











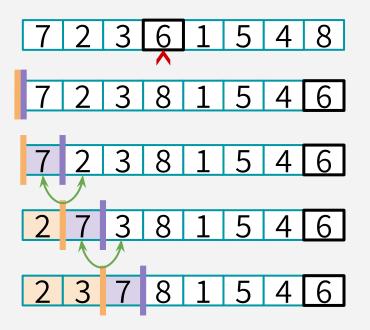
Choose pivot & swap with last element so pivot is at the end.

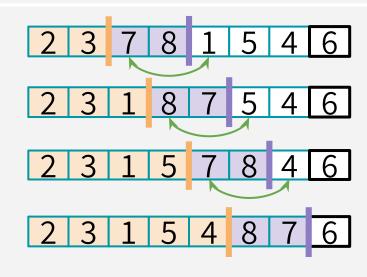


Increment until it sees something smaller than pivot, **swap** the things ahead of the bars & increment both bars



Repeat until the bar reaches the end, then swap the pivot into the right place.





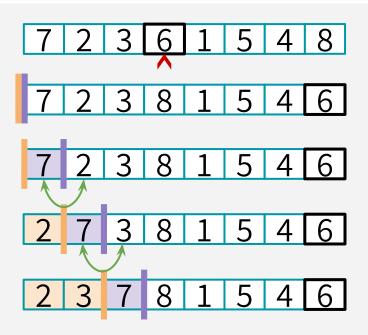
Choose pivot & swap with last element so pivot is at the end.

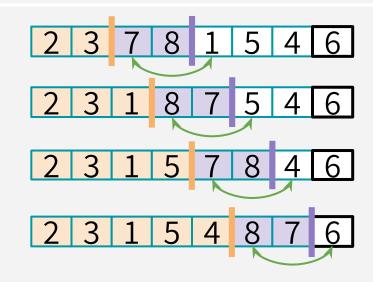


Initialize and

Increment until it sees something smaller than pivot, **swap** the things ahead of the bars & increment both bars

Repeat until the bar reaches the end, then swap the pivot into the right place.





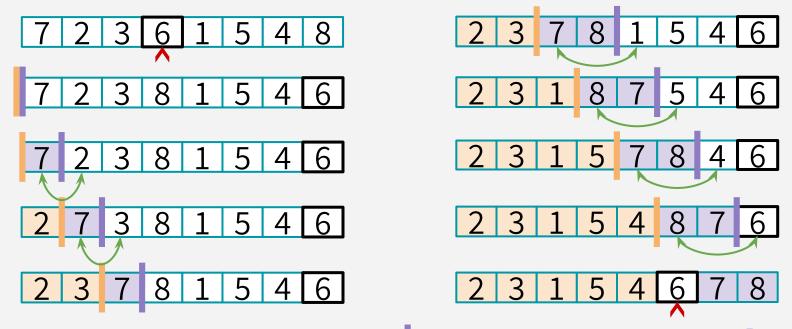
Choose pivot & swap with last element so pivot is at the end.



⇒ so

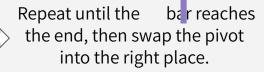
Increment until it sees something smaller than pivot, **swap** the things ahead of the bars & increment both bars

Repeat until the bar reaches the end, then swap the pivot into the right place.



Choose pivot & swap with last element so pivot is at the end.





## IMPLEMENTING QUICKSORT

There's another in-place partition algorithm called Hoare Partition that's even more efficient as it performs less swaps.

(you're not responsible for knowing it in this class)



## QUICKSORT vs. MERGESORT

	QuickSort (random pivot)	MergeSort (deterministic)
Runtime	Worst-case: O(n²) Expected: O(n log n)	Worst-case: O(n log n)
Used by	Java (primitive types), C (qsort), Unix, gcc	Java for objects, perl
In-place? (i.e. with O(log n) extra memory)	Yes, pretty easily!	Easy if you sacrifice runtime (O(nlogn) MERGE runtime).  Not so easy if you want to keep runtime & stability.
Stable?	No	Yes
Other Pros	Good cache locality if implemented for arrays	Merge step is really efficient with linked lists

You do not need to understand any of this stuff