

CE203

ساختمان داده ها و الگوریتم ها

سجاد شیرعلی شمرضا
پاییز 1400

جلسه ششم و هفتم: حل با روش جایگذاری و انتخاب k امین عضو

شنبه، 17 مهر 1400 و شنبه، 24 مهر 1400

اطلاع رسانی

- بخش مرتبط کتاب برای این جلسه: 4.3
- ارائه تمرین اول
 - مهلت ارسال تمرین اول: صبح شنبه، 24 مهر 1400 (ساعت 8 صبح)
- امتحانک اول:
 - دوشنبه همین هفته، 16 مهر 1400
 - به صورت آنلاین
 - از طریق سامانه کورسس
 - در طی ساعت کلاس

SUBSTITUTION METHOD

Algorithm, Proof of Correctness, Runtime

حل با روش جایگذاری

الگوریتم، اثبات درستی، زمان اجرا

SO FAR:

- Proving correctness:
 - Iterative algorithms: proof by induction on the iteration (e.g. Insertion Sort)
 - Recursive algorithms: proof by induction on the input size (e.g. MergeSort)
- Proving runtime:
 - Iterative algorithms: ~intuition~ (basically directly analyze the work being done)
 - Recursive algorithms: **Defining & Solving recurrence relations**

SO FAR:

- Proving correctness:
 - Iterative algorithms: proof by induction on the iteration (e.g. Insertion Sort)
 - Recursive algorithms: proof by induction on the input size (e.g. MergeSort)
- Proving runtime:
 - Iterative algorithms: ~intuition~ (basically directly analyze the work being done)
 - Recursive algorithms: **Defining & Solving recurrence relations**

We saw how to solve some recurrence relations with RECURSION TREES
& the MASTER THEOREM. Here's another way:

THE SUBSTITUTION METHOD

SUBSTITUTION METHOD

1. Guess what the answer is (expand for a few iterations)
2. Prove your guess is correct (using induction)

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This is a good technique to turn to if you find that the Master Theorem doesn't work. It's also especially helpful with recurrences that have differently sized subproblems (i.e. when the recursion tree & table aren't helpful either).

Let's try it on some example recurrences...

SUBSTITUTION METHOD: EXAMPLE

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

$$T(1) = 1$$

STEP 1: guess what the answer is!

You can “unravel” the recursion a few steps & follow the pattern to get a closed form expression!

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$$\begin{aligned} T(n) &= 2T(n/2) + n \\ &= 2(2T(n/4) + n/2) + n \\ &= 4T(n/4) + 2n \\ &= 4(2(T(n/8) + n/4)) + 2n \\ &= 8(T(n/8)) + 3n \\ &= \dots \end{aligned}$$

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$$\begin{aligned} T(n) &= \dots \\ &= nT(n/n) + (\log n)n \\ &= nT(1) + n \log n \\ &= n \log n + n \end{aligned}$$

SUBSTITUTION METHOD: EXAMPLE

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

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$$\begin{aligned} T(n) &= \dots \\ &= nT(n/n) + (\log n)n \\ &= nT(1) + n \log n \\ &= n \log n + n \end{aligned}$$

let's guess that
 $T(n) = n \log n + n$
and try to prove it!

SUBSTITUTION METHOD: EXAMPLE

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

$$T(1) = 1$$



Our guess from Step 1:

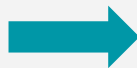
$$\mathbf{T(n) = n \log n + n}$$

STEP 2: Try to prove your guess!

SUBSTITUTION METHOD: EXAMPLE

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

$$T(1) = 1$$



Our guess from Step 1:

$$T(n) = n \log n + n$$

STEP 2: Try to prove your guess!

- **Inductive Hypothesis:** $T(n) = n \log n + n$

SUBSTITUTION METHOD: EXAMPLE

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

$$T(1) = 1$$



Our guess from Step 1:

$$T(n) = n \log n + n$$

STEP 2: Try to prove your guess!

- **Inductive Hypothesis:** $T(n) = n \log n + n$
- **Base case:** Prove IH holds for $n = 1$. $T(1) = 1 = 1 \log 1 + 1$.

SUBSTITUTION METHOD: EXAMPLE

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

$$T(1) = 1$$



Our guess from Step 1:

$$T(n) = n \log n + n$$

STEP 2: Try to prove your guess!

- **Inductive Hypothesis:** $T(n) = n \log n + n$
- **Base case:** Prove IH holds for $n = 1$. $T(1) = 1 = 1 \log 1 + 1$.
- **Inductive step:**
 - Let $k > 1$. Assume that the IH holds for all n such that $1 \leq n < k$.
 - $$\begin{aligned} T(k) &= 2 \cdot T(k/2) + k \\ &= 2 \cdot ((k/2)(\log(k/2)) + (k/2)) + k \\ &= 2 \cdot ((k/2)(\log k - 1 + 1)) + k \\ &= 2 \cdot (k/2)(\log k) + k \\ &= k \log k + k \end{aligned}$$

SUBSTITUTION METHOD: EXAMPLE

$$T(n) = 2 \cdot T(n/2) + n$$
$$T(1) = 1$$



Our guess from Step 1:

$$T(n) = n \log n + n$$

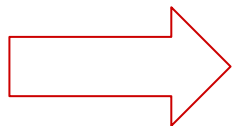
STEP 2: Try to prove your guess!

- **Inductive Hypothesis:** $T(n) = n \log n + n$
- **Base case:** Prove IH holds for $n = 1$. $T(1) = 1 = 1 \log 1 + 1$.
- **Inductive step:**
 - Let $k > 1$. Assume that the IH holds for all n such that $1 \leq n < k$.
 - $$\begin{aligned} T(k) &= 2 \cdot T(k/2) + k \\ &= 2 \cdot ((k/2)(\log(k/2)) + (k/2)) + k \\ &= 2 \cdot ((k/2)(\log k - 1 + 1)) + k \\ &= 2 \cdot (k/2)(\log k) + k \\ &= k \log k + k \end{aligned}$$
- **Conclusion:** By induction, $T(n) \leq n \log n + n$ for all $n > 0$.

This satisfies the
Big-O definition for
 $O(n \log n)$
(imagine choosing
 $c = 2, n_0 = 1$)

SUBSTITUTION METHOD: EXAMPLE 1

$$\begin{aligned} T(n) &\leq n \log n + n \\ \text{for all } n > 0 \end{aligned}$$

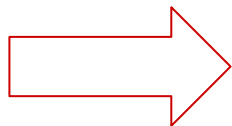


$$T(n) = O(n \log n)$$

1. Guess what the answer is (expand for a few iterations)
2. Prove your guess is correct (using induction)

SUBSTITUTION METHOD: EXAMPLE 1

$$T(n) \leq n \log n + n \\ \text{for all } n > 0$$



$$T(n) = O(n \log n)$$

1. Guess what the answer is (expand for a few iterations)
2. Prove your guess is correct (using induction)

But sometimes expanding gets complicated...



سوال؟

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$

Note:

While Example 1 could have also been solved with the Master Theorem, this one has differently sized subproblems, so the Master Theorem won't apply.

So... Time to use the Substitution Method!

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$

STEP 1: guess what the answer is!

Unraveling this expression gets ugly... (feel free to try it!).

You can also make a semi-educated guess and just hope for the best.

SUBSTITUTION METHOD: EXAMPLE 2

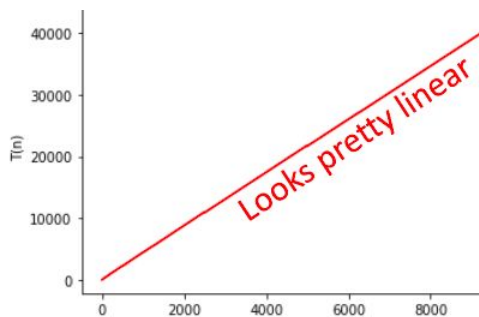
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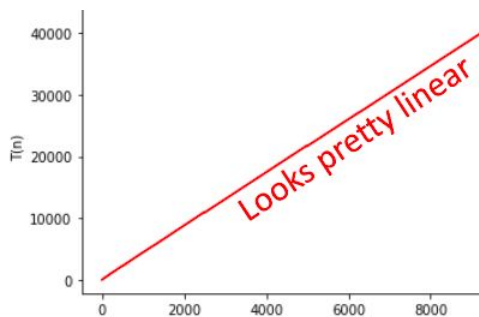
$$T(n) = T(n/5) + T(7n/10) + n$$

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STEP 1: guess what the answer is!

Unraveling this expression gets ugly... (feel free to try it!).

You can also make a semi-educated guess and just hope for the best.



It also feels like it could be better than $2T(n/2) + n$, which we know to be $O(n \log n)$...

SUBSTITUTION METHOD: EXAMPLE 2

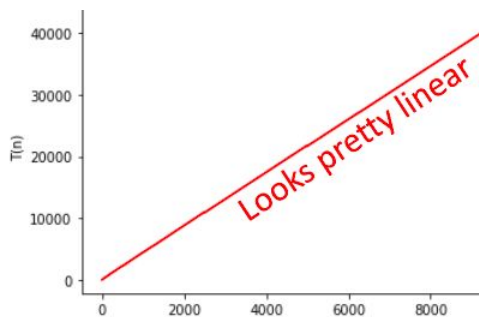
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STEP 1: guess what the answer is!

Unraveling this expression gets ugly... (feel free to try it!).

You can also make a semi-educated guess and just hope for the best.



It also feels like it could be better than $2T(n/2) + n$, which we know to be $O(n \log n)$...

Let's guess $O(n)$

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$



Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$



Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

WARNING:

You might be tempted to prove this with the inductive hypothesis
“ $T(n) = O(n)$ ”

But that doesn't make sense! Formally, this is what your IH would be saying:
“There is some $n_0 > 0$ and some $C > 0$ such that *for all* $n \geq n_0$, $T(n) \leq C \cdot n$ ”

Your IH is supposed to hold for a *specific* n , not an unbounded *range* of n !

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$



Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

You

Instead, we need to pick a C first,
and have our inductive hypothesis be

$$T(n) \leq C \cdot n$$

But

“There is some n_0 and some C such that for all $n \geq n_0$, $T(n) \leq C \cdot n$ ”

Your IH is supposed to hold for a *specific* n , not an unbounded *range* of n !

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

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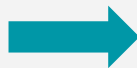
STEP 2: Prove it!

Use a placeholder **C** constant in the big-O proof. We don't know what C should be yet, but let's go through the proof leaving it as C and then figure out what works.

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

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Our guess from Step 1:

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STEP 2: Prove it!

Use a placeholder **C** constant in the big-O proof. We don't know what C should be yet, but let's go through the proof leaving it as C and then figure out what works.

- **Inductive Hypothesis:** $T(n) \leq Cn$
- **Base case:** Prove IH holds for $1 \leq n \leq 10$. $T(n) = 1 \leq Cn$

Whatever we choose
C to be, we know C
needs to be at least
1

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

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- **Inductive step:**
 - Let $k > 10$. Assume that the IH holds for all n such that $1 \leq n < k$.

Whatever we choose
C to be, we know C
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1

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$



Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

Use a placeholder **C** constant in the big-O proof. We don't know what C should be yet, but let's go through the proof leaving it as C and then figure out what works.

- **Inductive Hypothesis:** $T(n) \leq Cn$
- **Base case:** Prove IH holds for $1 \leq n \leq 10$. $T(n) = 1 \leq Cn$
- **Inductive step:**
 - Let $k > 10$. Assume that the IH holds for all n such that $1 \leq n < k$.
 - $$\begin{aligned} T(k) &= k + T(k/5) + T(7k/10) \\ &\leq k + C \cdot (k/5) + C \cdot (7k/10) \\ &= k \cdot (1 + C/5 + 7C/10) \\ &\leq Ck \end{aligned}$$
 ???
 - (If we find the right C, then we've shown IH holds for $n = k$)

Whatever we choose
C to be, we know C
needs to be at least
1

We can just solve for C:

$$1 + C/5 + 7C/10 \leq C$$

$$1 + 9C/10 \leq C$$

$$1 \leq C/10$$

So let's choose **C = 10!**

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$



Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

We can choose $C = 10$!

- **Inductive Hypothesis:** $T(n) \leq 10n$
- **Base case:** Prove IH holds for $1 \leq n \leq 10$. $T(n) = 1 \leq 10n$
- **Inductive step:**
 - Let $k > 10$. Assume that the IH holds for all n such that $1 \leq n < k$.
 - $$\begin{aligned} T(k) &= k + T(k/5) + T(7k/10) \\ &\leq k + 10 \cdot (k/5) + 10 \cdot (7k/10) \\ &= k + 2k + 7k \\ &= 10k \end{aligned}$$
 - Thus, the IH holds for $n = k$!
- **Conclusion:** With $C = 10$ and $n_0 = 1$, $T(n) \leq Cn$ for all $n \geq n_0$. By the Big-O definition, $T(n) = O(n)$.

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

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Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

We can choose $C = 10$!

- Induct
- Base
- Induc

-
-

Yay! Our guess worked!
But what if you make a bad guess?

= **10k**

- Thus, the IH holds for $n = k$!

- **Conclusion:** With $C = 10$ and $n_0 = 1$, $T(n) \leq Cn$ for all $n \geq n_0$. By the Big-O definition, $T(n) = O(n)$.



سوال؟

WHAT IF YOU MAKE A BAD GUESS?

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

$$T(1) = 1$$



Bad guess:

$$\mathbf{T(n) = O(n)}$$

WHAT IF YOU MAKE A BAD GUESS?

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Bad guess:

$$T(n) = O(n)$$

STEP 2: Prove it!

Use a placeholder C constant in the big- O proof. We don't know what C should be yet, but let's go through the proof leaving it as C and see if we run into any trouble.

- Inductive Hypothesis: $T(n) \leq Cn$
- Base case: $1 = T(1) \leq Cn$ for $n = 1$.
- Inductive step:
 - Let $k > 1$. Assume that the IH holds for all n such that $1 \leq n < k$.
 - $$\begin{aligned} T(k) &= 2 \cdot T(k/2) + k \\ &\leq 2 \cdot C(k/2) + k \\ &= Ck + k \\ &\leq Ck??? \end{aligned}$$

WHAT IF YOU MAKE A BAD GUESS?

$$T(n) = 2 \cdot T(n/2) + n$$
$$T(1) = 1$$



Bad guess:

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 - $$\begin{aligned} T(k) &= 2 \cdot T(k/2) + k \\ &\leq 2 \cdot C(k/2) + k \\ &= Ck + k \\ &\leq Ck??? \end{aligned}$$

We need this inequality to hold for the Inductive Step to be complete. However, no choice of C could ever make $Ck + k \leq Ck$!

WHAT IF YOU MAKE A BAD GUESS?

$$T(n) = 2 \cdot T(n/2) + n$$
$$T(1) = 1$$



Bad guess:
 $T(n) = O(n)$

STEP 2: Prove it!

Use a placeholder C constant in the big- O proof. We don't know what C should be yet, but let's go

-
-
-

A few tips:

If you stumble across impossible inequalities, then your guess was too small! If you end up with an inequality that seems too loose (e.g. $k \leq k^2$, $\log k \leq k$), maybe try a smaller guess.

$$= Ck + k$$
$$\leq Ck???$$

hold
for the inductive step to be
**complete. However, no choice of C
could ever make $Ck + k \leq Ck$!**

SO WHAT HAVE WE LEARNED?

- The substitution method can work when the master theorem doesn't
 - E.g. with different-sized sub-problems

- 1. Guess what the answer is (expand for a few iterations)**
- 2. Prove your guess is correct (using induction)**

In your final proof, pretend like you didn't do Steps 1 & 2 - no need to say how you unraveled the expression or why you made your guess. Just make sure your proof checks out!



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انتخاب k امین عضو

الگوریتم، اثبات درستی، زمان اجرا

THE SELECT PROBLEM

INPUT:

an unsorted array **A** of n elements (assume all elements are distinct),
& an integer **k** in $\{1, \dots, n\}$

7	2	6	9	1	5	4	11
---	---	---	---	---	---	---	----

OUTPUT of SELECT(A, k): the k^{th} smallest element of A

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7	2	6	9	1	5	4	11
---	---	---	---	---	---	---	----

OUTPUT of SELECT(A, k): the k^{th} smallest element of A

SELECT(A, 1) = 1

SELECT(A, 2) = 2

SELECT(A, 3) = 4

SELECT(A, 8) = 11

SELECT(A, 1) = MIN(A)

SELECT(A, $n/2$) = MEDIAN(A)

SELECT(A, n) = MAX(A)

**Note: k is a
1-indexed number!**

THE SELECT PROBLEM

INPUT:

an unsorted array **A** of n elements (assume all elements are distinct),
& an integer **k** in $\{1, \dots, n\}$

7	2	6	9	1	5	4	11
---	---	---	---	---	---	---	----


OUTPUT of SELECT(A, k): the k^{th} smallest element of A

Can you come up with an $O(n \log n)$ algorithm for SELECT?

AN $O(n \log n)$ ALGORITHM

```
SELECT(A,k):  
    A = MERGESORT(A)  
    return A[k-1]
```

It's k-1 (rather than k)
since my pseudocode
is 0-indexed and k is a
1-indexed number



Okay, great! We're done!



سوال؟

AN $O(n \log n)$ ALGORITHM

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THE QUESTION IS...
**CAN WE DO
BETTER?**

It's $k-1$ (rather than k)
since my pseudocode
is 0-indexed and k is a
1-indexed number

~~Okay, great! We're done!~~

GOAL: AN $O(n)$ ALGORITHM

If $k = 1$, then we want the minimum of A . There's an easy $O(n)$ algorithm for that:

Pretty much the same if $k = n$ (we're just finding $\text{MAX}(A)$ instead)


GOAL: AN $O(n)$ ALGORITHM

If $k = 1$, then we want the minimum of A . There's an easy $O(n)$ algorithm for that:


Pretty much the same if $k = n$ (we're just finding $\text{MAX}(A)$ instead)

SELECT-1(A):

$\text{result} = \text{infinity}$

for i **in** $[0, \dots, n-1]$: 

if $A[i] < \text{result}$:

The body of each iteration
is $O(1)$ work. 

$\text{result} = A[i]$

return result

Runtime of SELECT-1: $O(n)$

GOAL: AN $O(n)$ ALGORITHM

If $k = 2$, then we want the second-smallest element in A .

There's an easy-ish $O(n)$ algorithm for that:

(Not a very important algorithm, because this will end up being a bad idea...)

GOAL: AN $O(n)$ ALGORITHM

If $k = 2$, then we want the second-smallest element in A .

There's an easy-ish $O(n)$ algorithm for that:

(Not a very important algorithm, because this will end up being a bad idea...)

SELECT-2(A):

 result = infinity

 minSoFar = infinity

 for i in $[0, \dots, n-1]$:

 if $A[i] < \text{result} \ \& \ A[i] < \text{minSoFar}$:

 result = minSoFar

 minSoFar = $A[i]$

 else if $A[i] < \text{result} \ \& \ A[i] \geq \text{minSoFar}$

 result = $A[i]$

 return result

The body of each iteration
is still $O(1)$ work.

This loop runs $O(n)$ times

Runtime of SELECT-2: $O(n)$

GOAL: AN $O(n)$ ALGORITHM

If $k = n/2$, then we want the median element in A.

SELECT- $n/2$ (A):

```
result = infinity
minSoFar = infinity
secondMinSoFar = infinity
thirdMinSoFar = infinity
fourthMinSoFar = infinity
fifthMinSoFar = infinity
...
```

GOAL: AN $O(n)$ ALGORITHM

If $k = n/2$, then we want the median element in A .

SELECT- $n/2(A)$:

```
result = infinity
minSoFar = infinity
secondMinSoFar = infinity
thirdMinSoFar = infinity
fourthMinSoFar = infinity
fifthMinSoFar = infinity
...
```

Runtime of SELECT- $n/2$: $O(n^2)$

Clearly, this algorithm style isn't a good idea for large k (e.g. $n/2$).
This basically ends up looking like InsertionSort.

LINEAR SELECTION: THE IDEA

Let's use DIVIDE-and-CONQUER!

LINEAR SELECTION: THE IDEA

Let's use DIVIDE-and-CONQUER!

Select a pivot

Partition around it

Recurse!

LINEAR SELECTION: THE IDEA

Let's use DIVIDE-and-CONQUER!

Select a pivot

Partition around it

Recurse!

kind of like a “binary search” for the k^{th} smallest element (except that the array isn't sorted!)

LINEAR SELECTION: THE IDEA

3	2	9	8	1	6	4	11
---	---	---	---	---	---	---	----

LINEAR SELECTION: THE IDEA

Select a pivot

3	2	9	8	1	6	4	11
---	---	---	---	---	---	---	----

How do we pick a pivot?? We'll see this later.
For now, imagine we pick it randomly.



LINEAR SELECTION: THE IDEA

Select a pivot

3	2	9	8	1	6	4	11
---	---	---	---	---	---	---	----

How do we pick a pivot?? We'll see this later.
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Partition around it



Partition around pivot: **L** has elements less than pivot, and **R** has elements greater than pivot.
(Note that **L** and **R** remain unsorted).

LINEAR SELECTION: THE IDEA

Select a pivot



How do we pick a pivot?? We'll see this later.
For now, imagine we pick it randomly.

Partition around it



Partition around pivot: **L** has elements less than pivot, and **R** has elements greater than pivot.
(Note that **L** and **R** remain unsorted).

Recurse!

The pivot is in position 5. We have three cases:

1. **if $k = 5$: return pivot** the k^{th} smallest element is the pivot!
2. **if $k < 5$: return `SELECT(L, k)`** the k^{th} smallest element lives in L
3. **if $k > 5$: return `SELECT(R, k-5)`** the k^{th} smallest element is the $(k-5)^{\text{th}}$ smallest element in R

LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

1	12	4	20	31	6	18	9
---	----	---	----	----	---	----	---

LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

PICK A PIVOT

How do we pick a pivot???

We'll see later...

1	12	4	20	31	6	18	9
---	----	---	----	----	---	----	---



LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

1	12	4	20	31	6	18	9
---	----	---	----	----	---	----	---

PARTITION

L

1	12	4	6	9
---	----	---	---	---

18

20	31
----	----

R

LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

1	12	4	20	31	6	18	9
---	----	---	----	----	---	----	---



Recurse here (since 18 occupies index 6 and $k = 7 > 6$)

RECURSE

SELECT(R, 1):

20	31
----	----

$1 = 7 - 6$
(aka k minus pivot position)

LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

1	12	4	20	31	6	18	9
---	----	---	----	----	---	----	---



Recurse here (since 18 occupies index 6 and $k = 7 > 6$)

SELECT(R, 1):

20	31
----	----

PICK A PIVOT

How do we pick a pivot??
We'll see later...

LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

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---	----	---	----	----	---	----	---



Recurse here (since 18 occupies index 6 and $k = 7 > 6$)

SELECT(R, 1):

20	31
----	----



PARTITION

LINEAR SELECTION: EXAMPLE

SELECT(A, 7):

1	12	4	20	31	6	18	9
---	----	---	----	----	---	----	---



Recurse here (since 18 occupies index 6 and $k = 7 > 6$)

SELECT(R, 1):

20	31
----	----



20 is in the 1th position, and $k = 1$!
No need to recurse further!

20 IS OUR ANSWER!
(20 is the 1th smallest in R,
and 7th smallest overall)

LINEAR SELECTION: PSEUDOCODE

Base Case:
if $\text{len}(A) = 1$, then just
go ahead and return
the element itself

```
SELECT(A,k):  
  { if len(A) == 1:  
    return A[0]  
    p = GET_PIVOT(A)  
    L, R = PARTITION(A,p)  
    if len(L) == k-1:  
      return p  
    else if len(L) > k-1:  
      return SELECT(L, k)  
    else:  
      return SELECT(R, k-len(L)-1)
```

Case 1:

We got lucky and found
exactly the k^{th} smallest!

Case 2:

The k^{th} smallest is in the
first part of the array (L)

Case 3:

The k^{th} smallest is in the
second part of the array (R)

LINEAR SELECTION: PSEUDOCODE

```
SELECT(A,k):  
    if len(A) == 1:  
        return A[0]  
    p = GET_PIVOT(A)  
    L, R = PARTITION(A,p)  
    if len(L) == k-1:  
        return p  
    else if len(L) > k-1:  
        return SELECT(L, k)  
    else:  
        return SELECT(R, k-len(L)-1)
```

```
PARTITION(A, pivot):  
    L, R = [], []  
    for i in [1,...,len(A)]:  
        if A[i] == pivot:  
            continue  
        else if A[i] < pivot:  
            add A[i] to L  
        else:  
            add A[i] to R
```



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LINEAR SELECTION: SO FAR

- Intuition:
 - Partition the array around a pivot (how do we select?? still TBD)
 - Either return the pivot itself or recurse on the left or right subarrays (but not both!)

```
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LINEAR SELECTION: SO FAR

- Intuition:
 - Partition the array around a pivot (how do we select?? still TBD)
 - Either return the pivot itself or recurse on the left or right subarrays (but not both!)
- Our two favorite questions:
 - Does this work?
 - What's the runtime?

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    if len(A) == 1:  
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```

LINEAR SELECTION: DOES IT WORK?

RECURSIVE ALGORITHMS

1. **Inductive hypothesis:** your algorithm is correct for sizes *up to* i
2. **Base case:** IH holds for $i < \text{small constant}$
3. **Inductive step:**
 - assume IH holds for $k \Rightarrow$ prove $k+1$, OR
 - assume IH holds for $\{1, 2, \dots, k-1\} \Rightarrow$ prove k .
4. **Conclusion:** IH holds for $i = n \Rightarrow$ yay!

FROM LAST WEEK!

INDUCTION PROOF

INDUCTIVE HYPOTHESIS (IH)

When run on an array A of size i and an integer $1 \leq k \leq i$, $\text{SELECT}(A, k)$ correctly returns the k^{th} smallest element of A .

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BASE CASE

The IH holds for $i = 1$: We know k must be 1, so SELECT does indeed return the smallest (and only) element of A .

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(OUTLINE OF) INDUCTIVE STEP (*strong/complete induction*)

Let j be an integer, where $j > 1$. Assume that the IH holds for all i where $1 \leq i < j$. We want to show that the IH holds for $i = j$, i.e. that for an array A of size j and an integer $k \leq j$, SELECT returns the k^{th} smallest element of A .

We consider three cases, depending on the pivot chosen by GET_PIVOT . PARTITION gives us L , and R .

- **CASE 1:** $|L| = k-1$.
 - **CASE 2:** $|L| > k-1$.
 - **CASE 3:** $|L| < k-1$.
- We use **STRONG** induction because cases 2 and 3 rely on the correctness of the smaller recursive calls.

Thus, in each of the three cases, $\text{SELECT}(A, k)$ returns the k^{th} smallest element of A . This establishes the IH for $i = j$.

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Thus, in each of the three cases, $\text{SELECT}(A, k)$ returns the k^{th} smallest element of A . This establishes the IH for $i = j$.

CONCLUSION

By induction, we conclude that the IH holds for all $1 \leq i \leq n$. Thus, we conclude that $\text{SELECT}(A, k)$ returns the k^{th} smallest element of A on any array A , provided that $1 \leq k \leq |A|$. That is, SELECT is correct!



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RUNTIME

```
SELECT(A,k):  
    if len(A) == 1:  
        return A[0]  
    p = GET_PIVOT(A)  
    L, R = PARTITION(A,p)  
    if len(L) == k-1:  
        return p  
    else if len(L) > k-1:  
        return SELECT(L, k)  
    else if len(L) < k-1:  
        return SELECT(R, k-len(L)-1)
```

Recurrence Relation for SELECT

For now, assume we'll pick the pivot in time $O(n)$

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Recurrence Relation for SELECT

For now, assume we'll pick the pivot in time $O(n)$

$$T(n) = \begin{cases} O(n) & \text{len(L) == k-1} \\ T(\text{len(L)}) + O(n) & \text{len(L) > k-1} \\ T(\text{len(R)}) + O(n) & \text{len(L) < k-1} \end{cases}$$

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But what are len(L) and len(R) ?
That depends on how we pick the pivot...

RUNTIME

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    if len(A) == 1:  
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    p = GET_PIVOT(A)  
    L, R = PARTITION(A,p)  
    if len(L) == k-1:  
        return p  
    else if len(L) > k-1:  
        return SELECT(L, k)  
    else if len(L) < k-1:  
        return SELECT(R, k-len(L)-1)
```

What's a “good” pivot?
What's a “bad” pivot?

Relation for SELECT

we'll pick the pivot in time $O(n)$

$$T(n) = \begin{cases} O(n) & \text{len(L) == k-1} \\ T(\text{len(L)}) + O(n) & \text{len(L) > k-1} \\ T(\text{len(R)}) + O(n) & \text{len(L) < k-1} \end{cases}$$

But what are **len(L)** and **len(R)**?
That depends on how we pick the pivot...

THE WORST PIVOT

The WORST pivot: picking the max or the min each time!

Then, in the worst case, the recurrence relation looks like $T(n) = T(n-1) + O(n)$.

$$T(n) = \begin{cases} O(n) & \text{len}(L) == k-1 \\ T(\text{len}(L)) + O(n) & \text{len}(L) > k-1 \\ T(\text{len}(R)) + O(n) & \text{len}(L) < k-1 \end{cases} \quad \Rightarrow \quad T(n) \leq T(n-1) + O(n)$$

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This ends up being $\Omega(n^2)$!

A call to `SELECT(A, n/2)` would already consist of $\sim n/2$ recursive calls
(each with a subarray of length at least $n/2$)!

THE IDEAL PIVOT

The IDEAL pivot: splits the input array exactly in half!

$$\text{len(L)} = \text{len(R)} = (n-1)/2$$

$$T(n) = \begin{cases} O(n) & \text{len(L)} == k-1 \\ T(\text{len(L)}) + O(n) & \text{len(L)} > k-1 \\ T(\text{len(R)}) + O(n) & \text{len(L)} < k-1 \end{cases} \quad \Rightarrow \quad T(n) \leq T(n/2) + O(n)$$

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$a = 1$
 $b = 2$
 $d = 1$
 $a < b^d$

Suppose $T(n) = a \cdot T(n/b) + O(n^d)$. The Master Theorem states:

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b(a)}) & \text{if } a > b^d \end{cases}$$

THE IDEAL PIVOT

The IDEAL pivot: splits the input array exactly in half!

$$T(n) = \begin{cases} O(n) \\ T(\text{len}(L)) + O(n) \\ T(\text{len}(R)) + O(n) \end{cases}$$

*With the ideal
pivot, the runtime
would be:*

$O(n)$

$$T(n) \leq T(n/2) + O(n)$$

$$\begin{aligned} a &= 1 \\ b &= 2 \\ d &= 1 \end{aligned}$$

$$a < b^d$$

Suppose $T(n) = a \cdot T(n/b)$, Master Theorem states:

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THE IDEAL PIVOT

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Sadly, the pivot to divide the input in half is the

MEDIAN

*aka **SELECT(A, n/2)***

aka exactly the problem we're trying to solve...

$$T(n) = \begin{cases} O(n) \\ T(n/2) \\ T(n/2) \end{cases}$$

+ O(n)

b^d

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b(a)}) & \text{if } a > b^d \end{cases}$$



سوال؟

THE GOOD-ENOUGH PIVOT

The GOOD-ENOUGH pivot: splits the input array kind of in half!

$$3n/10 < \text{len}(L) < 7n/10$$

$$3n/10 < \text{len}(R) < 7n/10$$

THE GOOD-ENOUGH PIVOT

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$$3n/10 < \text{len(L)} < 7n/10$$

$$3n/10 < \text{len(R)} < 7n/10$$

If we could fetch this good-enough pivot in time $O(n)$, let's say, the recurrence looks like:

$$T(n) = \begin{cases} O(n) & \text{len(L)} == k-1 \\ T(\text{len(L)}) + O(n) & \text{len(L)} > k-1 \\ T(\text{len(R)}) + O(n) & \text{len(L)} < k-1 \end{cases} \quad \Rightarrow \quad T(n) \leq T(7n/10) + O(n)$$

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$a = 1$
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$$T(n) = \begin{cases} O(n) \\ T(\text{len}(L)) + O(n) \\ T(\text{len}(R)) + O(n) \end{cases}$$

*This good-enough pivot
would still give us:*

$$O(n)$$

$$T(n) \leq T(7n/10) + O(n)$$

$$a = 1$$

$$b = 10/7$$

$$d = 1$$

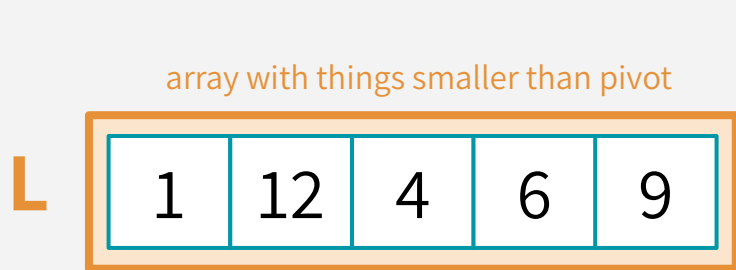
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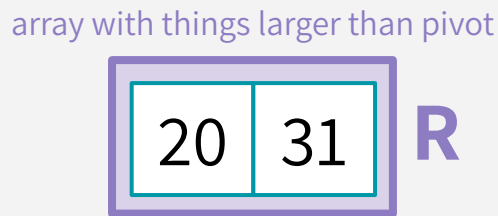
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OUR GOAL

Efficiently pick the pivot in time $O(n)$ so that



$$3n/10 < \text{len}(L) < 7n/10$$



$$3n/10 < \text{len}(R) < 7n/10$$

Then, our recurrence $T(n) \leq T(7n/10) + O(n)$ comes out to **$O(n)$** !



سوال؟

میانه ی میانه ها!

ایده اصلی الگوریتم خطی برای انتخاب k امین عضو

MEDIAN-OF-MEDIANS

The ideal world wasn't feasible because we can't just compute $\text{SELECT}(A, n/2) \Rightarrow$ that would throw us into infinite recursion since problem sizes aren't shrinking between recursive calls...

But we can instead generate a ***smaller*** list and call SELECT on that smaller list!

MEDIAN-OF-MEDIANS

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But we can instead generate a ***smaller*** list and call SELECT on that smaller list!

OUR GAME PLAN:

We'll make a smaller list out of SUB-MEDIANS.

Then, we'll use SELECT to find the median of the sub-medians.

This “median of medians” will be our proxy for the true median!

MEDIAN-OF-MEDIANS

GOAL: get a proxy for the true median by finding the exact median of all the sub-medians!

1	14	4	18	25	6	17	9	3	5	10	16	12	23	19	13	20	8	15	24	7	21	22	2	11
---	----	---	----	----	---	----	---	---	---	----	----	----	----	----	----	----	---	----	----	---	----	----	---	----

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Divide the original list into $\lceil n/5 \rceil$ groups (each group has ≤ 5 elements)

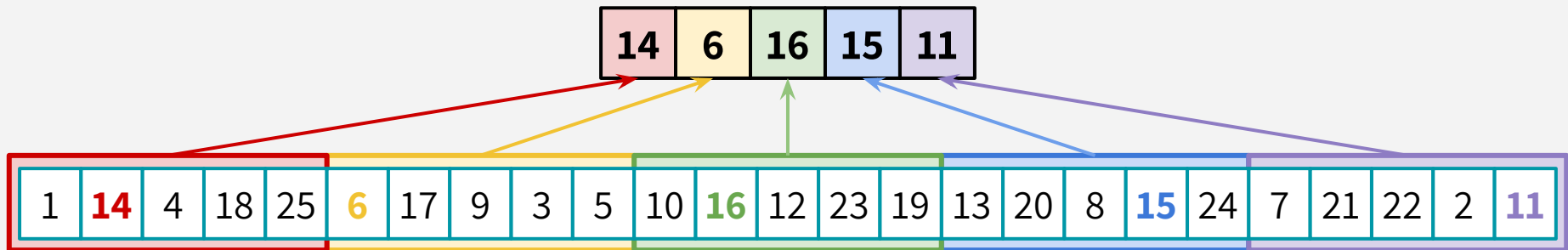
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Divide the original list into $\lceil n/5 \rceil$ groups (each group has ≤ 5 elements)

Find the sub-median of each small group (3rd smallest out of the 5)



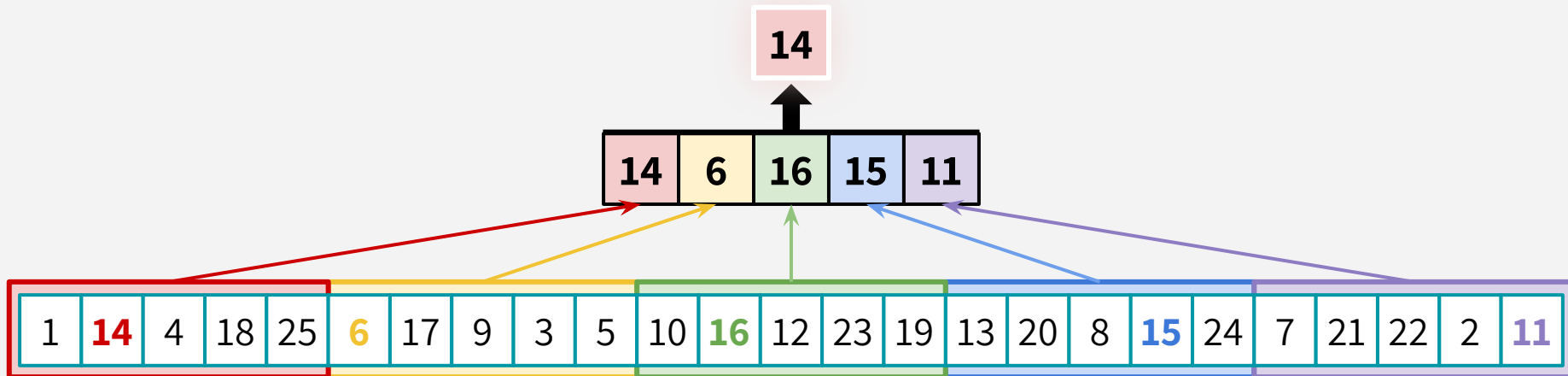
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Find the median of all the sub-medians (call SELECT)



MEDIAN-OF-MEDIANS

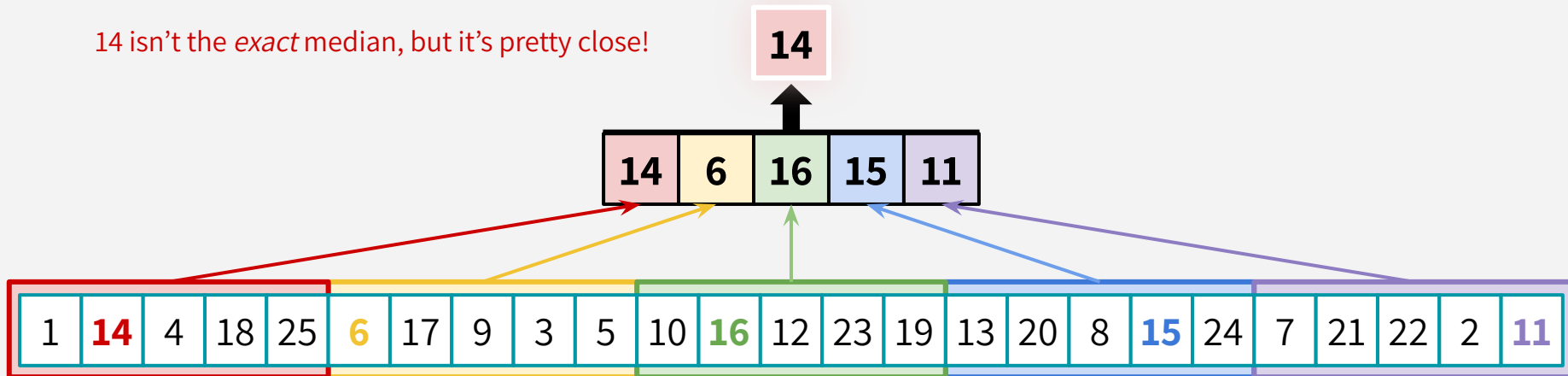
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Find the median of all the sub-medians (call SELECT)

14 isn't the *exact* median, but it's pretty close!



MEDIAN-OF-MEDIANS

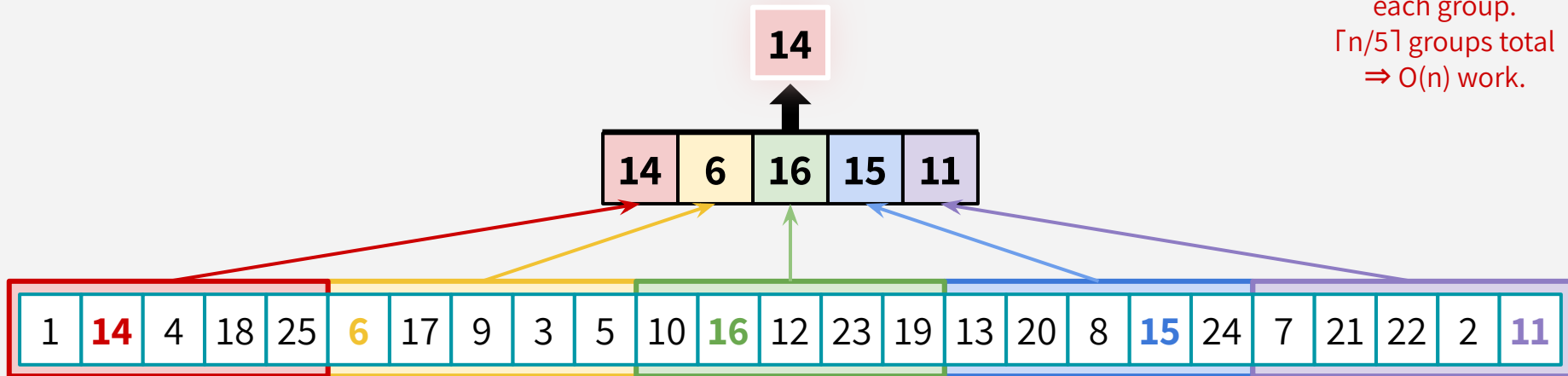
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Find the median of all the sub-medians (call SELECT)

constant work for
each group.
 $\lceil n/5 \rceil$ groups total
 $\Rightarrow O(n)$ work.



MEDIAN-OF-MEDIANS

GOAL: get a proxy for the true median by finding the exact median of all the sub-medians!

Divide the original list into $\lceil n/5 \rceil$ groups (each group has ≤ 5 elements)

Find the sub-median of each small group (3rd smallest out of the 5)

Find the median of all the sub-medians (call SELECT)

constant work for
each group.
 $\lceil n/5 \rceil$ groups total
work.

14

To compute our pivot:

Do $O(n)$ work to set up (divide into groups & get a list of submedians),
then make a call to **SELECT**(Submedians, $\lfloor \text{Submedians} \rfloor / 2$)





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ANALYZING RUNTIME

```
SELECT(A,k):  
    if len(A) == 1:  
        return A[0]  
    p = MEDIAN_OF_MEDIANS(A)  
    L, R = PARTITION(A,p)  
    if len(L) == k-1:  
        return p  
    else if len(L) > k-1:  
        return SELECT(L, k)  
    else:  
        return SELECT(R, k-len(L)-1)
```

What does the recurrence relation for $T(n)$ look like?

ANALYZING RUNTIME

```
SELECT(A,k):  
    if len(A) == 1:  
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```

$O(n)$ work outside of recursive calls

(base case, set-up within
MEDIAN_OF_MEDIANS, partitioning)

$T(n/5)$ work hidden in this recursive call

(remember, MEDIAN_OF_MEDIANS calls
SELECT on $\lceil n/5 \rceil$ -size array)

$T(???)$ work hidden in this recursive call

What is the maximum size of
either L or R?

ANALYZING RUNTIME

```
SELECT(A,k):  
    if len(A) == 1:
```

What is the smallest
number of elements that
could be smaller than our
MEDIAN OF MEDIANS?

```
else:  
    return SELECT(R, k-len(L)-1)
```

**$O(n)$ work outside of
recursive calls**

(base case, set-up within
MEDIAN_OF_MEDIANS, partitioning)

**$T(n/5)$ work hidden in
this recursive call**

(remember, MEDIAN_OF_MEDIANS calls
SELECT on $\lceil n/5 \rceil$ -size array)

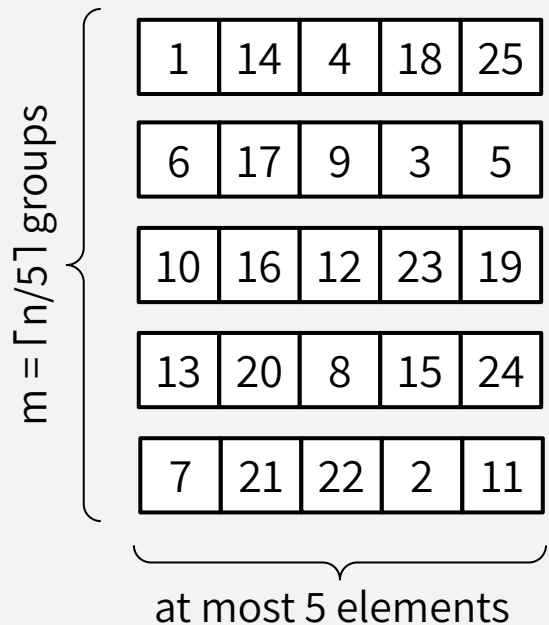
**$T(???)$ work hidden in
this recursive call**

What is the maximum size of
either L or R?

ANALYZING RUNTIME

MEDIAN_OF_MEDIANS will choose a pivot greater than at least $3n/10 - 6$ elements

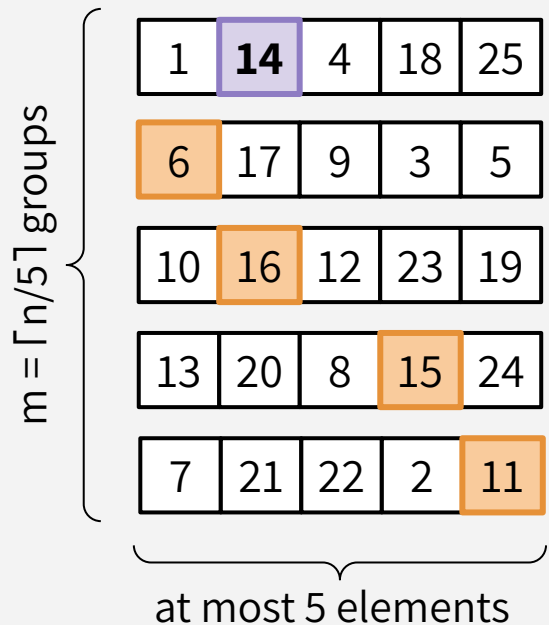
(The same reasoning we're about to do also shows that the pivot will be less than at least $3n/10 - 6$ elements)



ANALYZING RUNTIME

MEDIAN_OF_MEDIANS will choose a pivot greater than at least $3n/10 - 6$ elements

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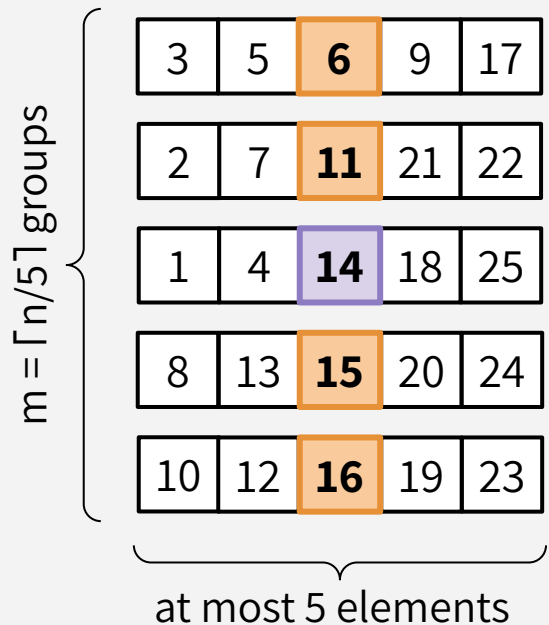


At least how many elements are guaranteed to be **smaller** than the median of medians?

ANALYZING RUNTIME

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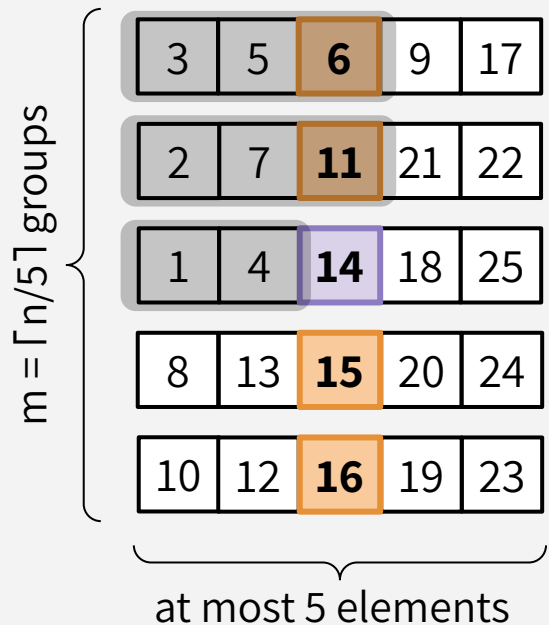


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(The same reasoning we're about to do also shows that the pivot will be less than at least $3n/10 - 6$ elements)



At least how many elements are guaranteed to be **smaller** than the median of medians?

3 elements from each group that has a **median** smaller than the **median of medians** + 2 elements from the group containing the **median of medians**

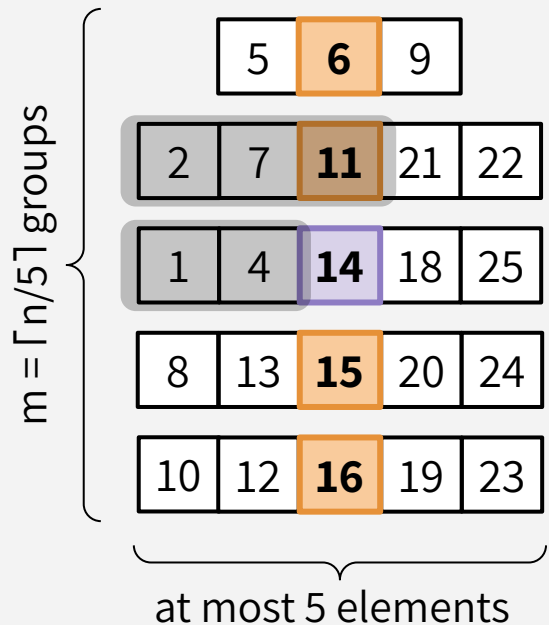
$$3 \cdot (\lceil m/2 \rceil - 1) + 2$$

To exclude the group with the **median of medians**

ANALYZING RUNTIME

MEDIAN_OF_MEDIANS will choose a pivot greater than at least $3n/10 - 6$ elements

(The same reasoning we're about to do also shows that the pivot will be less than at least $3n/10 - 6$ elements)



At least how many elements are guaranteed to be **smaller** than the median of medians?

3 elements from each (non-leftover) group that has a **median** smaller than the **median of medians** + 2 elements from the group containing the **median of medians**

$$3 \cdot (\lceil m/2 \rceil - 1 - 1) + 2$$

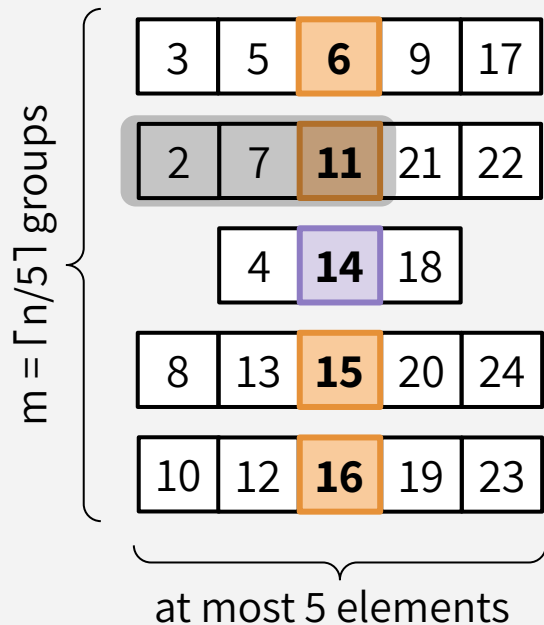
To exclude the group with the **median of medians**

To exclude any of those groups that might be a "leftover" group!

ANALYZING RUNTIME

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(The same reasoning we're about to do also shows that the pivot will be less than at least $3n/10 - 6$ elements)



At least how many elements are guaranteed to be **smaller** than the median of medians?

3 elements from each (non-leftover) group that has a **median** smaller than the **median of medians** + ~~2 elements from the group containing the **median of medians**~~

$$3 \cdot (\lceil m/2 \rceil - 1 - 1) + 2$$

To exclude the group with the **median of medians**

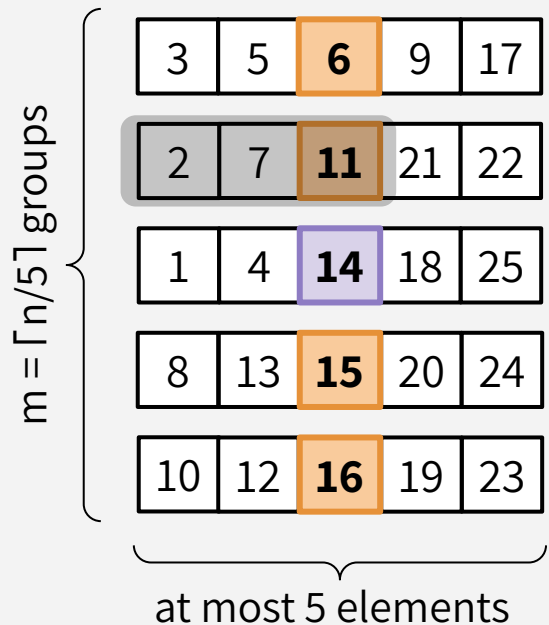
To exclude any of those groups that might be a "leftover" group!

The group with the **median of medians** might be a "leftover" group! Might as well just get rid of the +2 to be safe

ANALYZING RUNTIME

MEDIAN_OF_MEDIANS will choose a pivot greater than at least $3n/10 - 6$ elements

(The same reasoning we're about to do also shows that the pivot will be less than at least $3n/10 - 6$ elements)



At least how many elements are guaranteed to be **smaller** than the median of medians?

3 elements from each (non-leftover) group that has a **median** smaller than the **median of medians**

$$\begin{aligned} & 3 \cdot (\lceil m/2 \rceil - 2) \\ &= 3 \cdot (\lceil \lceil n/5 \rceil / 2 \rceil - 2) \\ &\geq 3 \cdot (n/10 - 2) \\ &= 3n/10 - 6 \end{aligned}$$

ANALYZING RUNTIME

We just showed:

$$3n/10 - 6 \leq \text{len}(L)$$

$$\text{len}(R) \leq 7n/10 + 5$$

ANALYZING RUNTIME

We can similarly show the inverse:

$$3n/10 - 6 \leq \text{len}(L) \leq 7n/10 + 5$$

$$3n/10 - 6 \leq \text{len}(R) \leq 7n/10 + 5$$

ANALYZING RUNTIME

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What is the maximum size of
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ANALYZING RUNTIME

We can similarly show the inverse:

$$3n/10 - 6 \leq \text{len}(L) \leq 7n/10 + 5$$

$$3n/10 - 6 \leq \text{len}(R) \leq 7n/10 + 5$$

What does the recurrence relation for $T(n)$ look like?

$$T(n) \leq T(n/5) + T(???) + O(n)$$

ANALYZING RUNTIME

We can similarly show the inverse:

$$3n/10 - 6 \leq \text{len}(L) \leq 7n/10 + 5$$

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What does the recurrence relation for $T(n)$ look like?

$$T(n) \leq T(n/5) + T(7n/10) + O(n)$$

ANALYZING RUNTIME

$$T(n) \leq T(n/5) + T(7n/10) + O(n)$$

We solved this recurrence using the Substitution Method at the start of class!

SUBSTITUTION METHOD: EXAMPLE 2

$$T(n) = T(n/5) + T(7n/10) + n$$

$$T(n) = 1 \text{ when } 1 \leq n \leq 10$$



Our guess from Step 1:

$T(n)$ is $O(n)$

STEP 2: Prove it!

We can choose $C = 10$!

- **Inductive Hypothesis:** $T(n) \leq 10n$
- **Base case:** Prove IH holds for $1 \leq n \leq 10$. $T(n) = 1 \leq 10n$
- **Inductive step:**
 - Let $k > 10$. Assume that the IH holds for all n such that $1 \leq n < k$.
 - $$\begin{aligned} T(k) &= k + T(k/5) + T(7k/10) \\ &\leq k + 10 \cdot (k/5) + 10 \cdot (7k/10) \\ &= k + 2k + 7k \\ &= 10k \end{aligned}$$
 - Thus, the IH holds for $n = k$!
- **Conclusion:** With $C = 10$ and $n_0 = 1$, $T(n) \leq Cn$ for all $n \geq n_0$. By the Big-O definition, $T(n) = O(n)$.

ANALYZING RUNTIME

$$T(n) \leq T(n/5) + T(7n/10) + O(n)$$

We solved this recurrence using the Substitution Method at the start of class!



$$O(n)$$

Worst-case Runtime!

LINEAR-TIME SELECTION

```
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```

$O(n)$

Worst-case Runtime!



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