CS2040S Data Structures and Algorithms

Welcome!



How to Search!

Algorithm Analysis

- Big-O Notation
- Model of computation

Searching

Peak Finding

- 1-dimension
- 2-dimensions

Admin

Tutorial and Recitation Registration

- Sorry for all the announcements!
- Still in progress.
- 95% settled.
- Please be patient!
- Please do not use ModReg to swap/petition/change your tutorial/recitation registration.

Admin

Midterm Exam: Week 7

- Trying to arrange in-person midterm.
 - Fewer problems in person!
 - No Zoom issues!
- Requirement: < 50 students / room.
- If we cannot find sufficient rooms, may reschedule:
 - E.g., an evening exam at 7pm
 - E.g., Saturday

How to Search!

Algorithm Analysis

- Big-O Notation
- Model of computation

Searching

Peak Finding

- 1-dimension
- 2-dimensions

Algorithm Analysis

Warm up: which takes longer?

```
void pushAdd(int k) {
    for (int i=0; i<= k; i++)
    {
        for (int j=0; j<= k; j++) {
            stack.push(i+j);
        }
    }
}</pre>
```

O(100k)



0(k2)

Algorithm Analysis

Warm up: which takes longer?

```
void pushAdd(int k) {
   for (int i=0; i<= k; i++)
   {
      for (int j=0; j<= k; j++) {
          stack.push(i+j);
      }
   }
}</pre>
```

100k push operations

 k^2 push operations

Which grows faster?

$$T(k) = 100k$$

$$T(k) = k^2$$

$$T\left(0\right) =0$$

$$T(1) = 100$$

$$T(100) = 10,000$$

$$T(1000) = 100,000$$

$$T(0) = 0$$

$$T(1) = 1$$

$$T(100) = 10,000$$

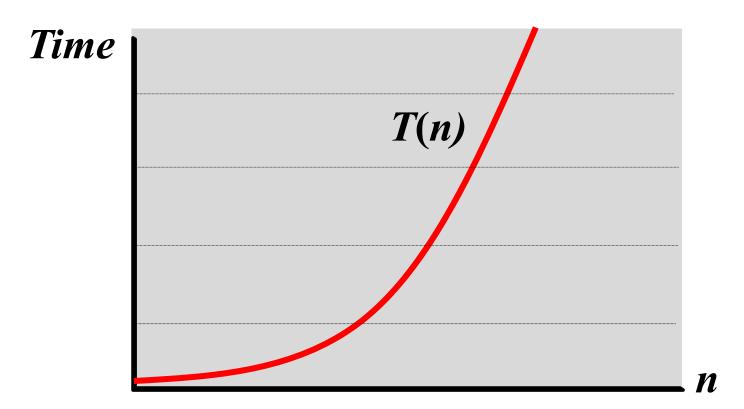
$$T(1000) = 1,000,000$$

Always think of big input



How does an algorithm scale?

- For large inputs, what is the running time?
- T(n) = running time on inputs of size n



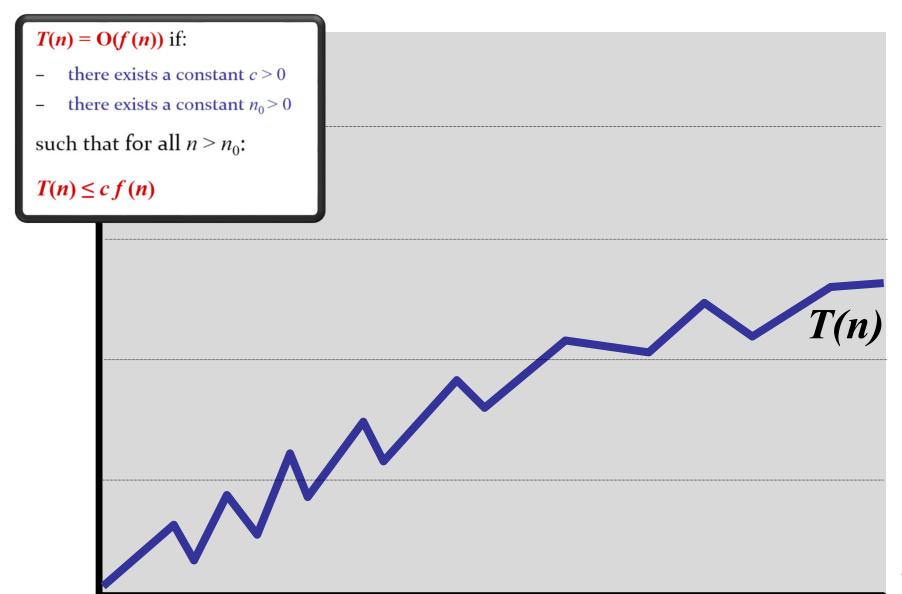
Definition: T(n) = O(f(n)) if T grows no faster than f

$$T(n) = O(f(n))$$
 if:

- there exists a constant c > 0
- there exists a constant $n_0 > 0$

such that for all $n > n_0$:

$$T(n) \le c f(n)$$



```
T(n) = O(f(n)) if:
                                    T(n) = O(f(n))
   there exists a constant c > 0
   there exists a constant n_0 > 0
such that for all n > n_0:
T(n) \le c f(n)
                                                                                 c f(n)
                                                                               T(n)
```

n

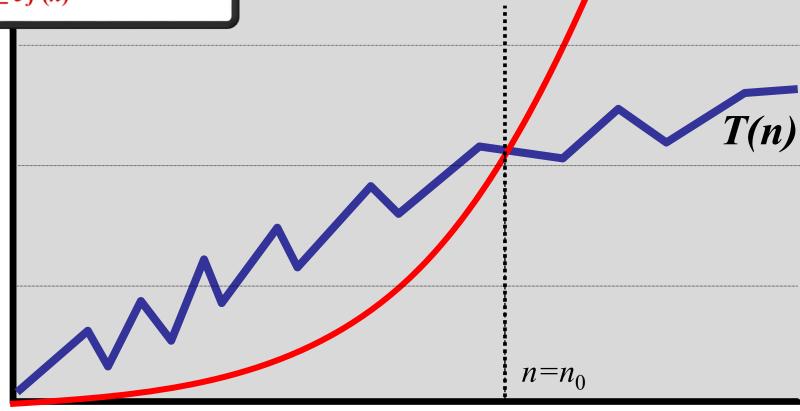
c f(n)

```
T(n) = O(f(n)) if:
```

- there exists a constant c > 0
- there exists a constant $n_0 > 0$

such that for all $n > n_0$:

$$T(n) \le c f(n)$$



T(n) = O(f(n))

n

Example proof: $T(n) = O(n^2)$

$$T(n) = 4n^2 + 24n + 16$$

Example

7	7	
1		l)

$$T(n) = 1000n$$

$$T(n) = 1000n$$

$$T(n) = n^2$$

$$T(\mathbf{n}) = 13n^2 + n$$

big-O

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

$$T(n) = O(n^2)$$

$$T(n) \neq O(n)$$

$$T(n) = O(n^2)$$

(n) Not tight

Definition: T(n) = O(f(n)) if T grows no faster than f

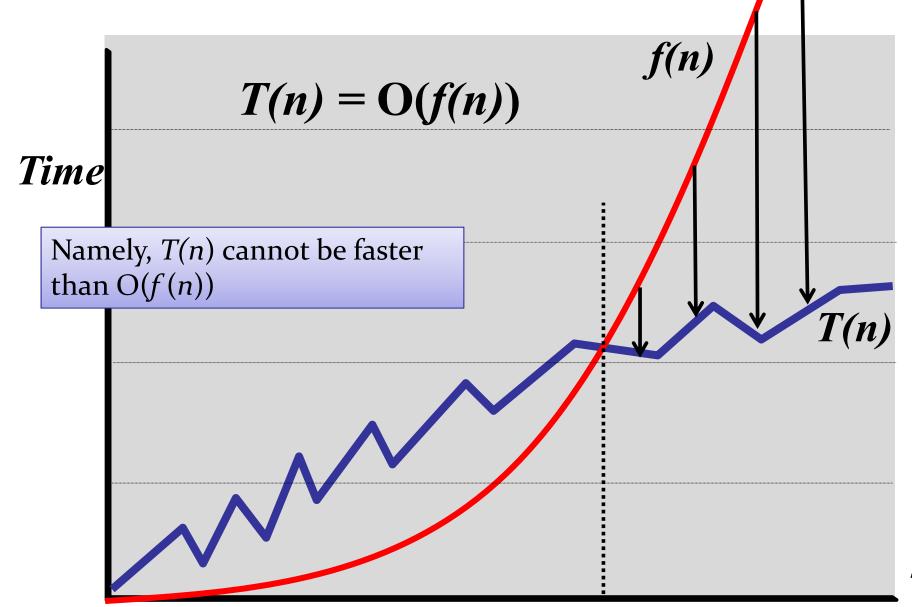
$$T(n) = O(f(n))$$
 if:

- there exists a constant c > 0
- there exists a constant $n_0 > 0$

such that for all $n > n_0$:

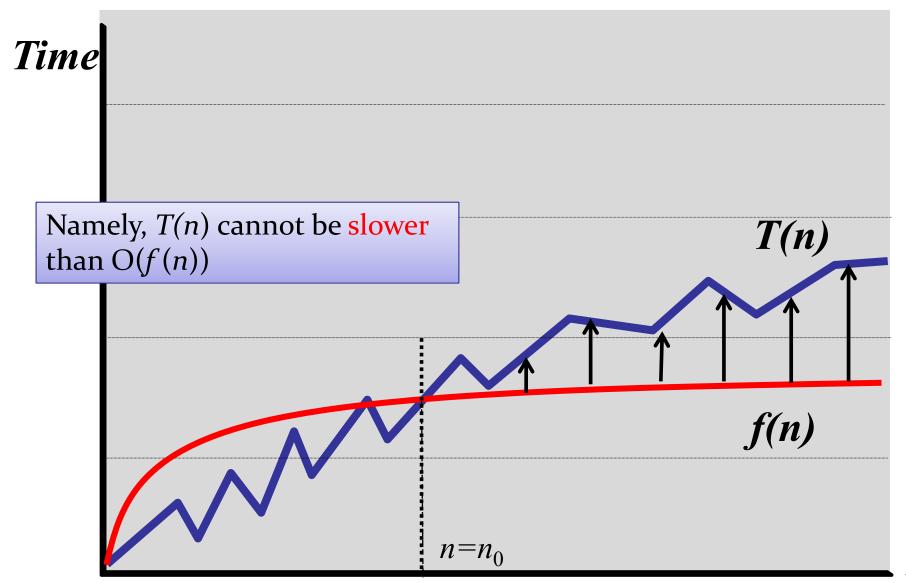
$$T(n) \le c f(n)$$

Big-O Notation as Upper Bound



How about Lower bound?





Definition: $T(n) = \Omega(f(n))$ if T grows no slower than f

$$T(n) = \Omega(f(n))$$
 if:

- there exists a constant c > 0
- there exists a constant $n_0 > 0$

such that for all $n > n_0$:

$$T(n) \ge c f(n)$$

Example

 $T(n) = 13n^2 + n$

T(n)	Asymptotic
T(n) = 1000n	$T(n) = \Omega(1)$
T(n) = n	$T(n) = \Omega(n)$
$T(n)=n^2$	$T(n) = \Omega(n)$

 $T(n) = \Omega(n^2)$

Exercise:

True or false:

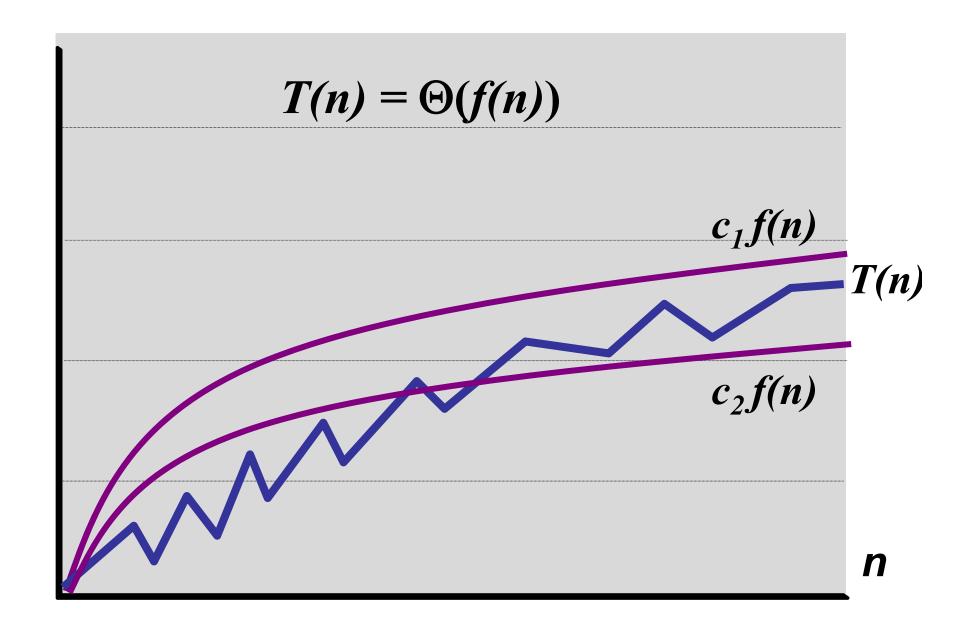
```
"f(n) = O(g(n)) if and only if g(n) = \Omega(f(n))"
```

Prove that your claim is correct using the definitions of O and Ω or by giving an example.

Definition: $T(n) = \Theta(f(n))$ if T grows at the same rate as f

$$T(n) = \Theta(f(n))$$
 if and only if:

- T(n) = O(f(n)) and
- $T(n) = \Omega(f(n))$



Example

T(n)	big-O
T(n) = 1000n	$T(\mathbf{n}) = \Theta(n)$
T(n) = n	$T(n) \neq \Theta(1)$
$T(n) = 13n^2 + n$	$T(n) = \Theta(n^2)$
$T(n)=n^3$	$T(n) \neq \Theta(n^2)$

Some simple rules for most cases...

Order or size:

Function	Name
5	Constant
loglog(n)	double log
log(n)	logarithmic
$\log^2(n)$	Polylogarithmic
n	linear
nlog(n)	log-linear
n^3	polynomial
n³log(n)	
n ⁴	polynomial
2 ⁿ	exponential
2 ²ⁿ	
n!	factorial

Rules:

If T(n) is a polynomial of degree k then:

$$T(n) = O(n^k)$$

Example:

$$10n^5 + 50n^3 + 10n + 17 = O(n^5)$$

Rules:

If
$$T(n) = O(f(n))$$
 and $S(n) = O(g(n))$ then
$$T(n) + S(n) = O(f(n) + g(n))$$

Example:

$$10n^{2} = O(n^{2})$$

$$5nlog(n) = O(nlog(n))$$

$$10n^{2} + 5nlog(n) = O(n^{2} + nlog(n)) = O(n^{2})$$

Rules:

If
$$T(n) = O(f(n))$$
 and $S(n) = O(g(n))$ then:

$$T(n)*S(n) = O(f(n)*g(n))$$

Example:

$$10n^{2} = O(n^{2})$$

$$5n = O(n)$$

$$(10n^{2})(5n) = 50n^{3} = O(n*n^{2}) = O(n^{3})$$

$$n^4 + 3n^2 + n^2 + 17 = ?$$

- A. O(1)
- B. O(n)
- C. $O(n^2)$
- D. $O(n^3)$
- E. $O(n^4)$

Why don't you try a few?



$$4n^2\log(n) + 8n + 16 = ?$$

- 1. $O(\log n)$
- O(n)
- 3. O(nlog n)
- 4. $O(n^2 \log n)$
- 5. $O(2^n)$

$$2^{2n} + 2^n + 2 =$$

- 1. O(n)
- 2. $O(n^6)$
- 3. $O(2^n)$
- 4. $O(2^{2n})$
- 5. $O(n^n)$

~1×1...×1)

log(n!) = log(n!)

- 1. $O(\log n)$
- 2. O(n)
- 3. $O(n \log n)$
- 4. $O(n^2)$
- 5. $O(2^n)$

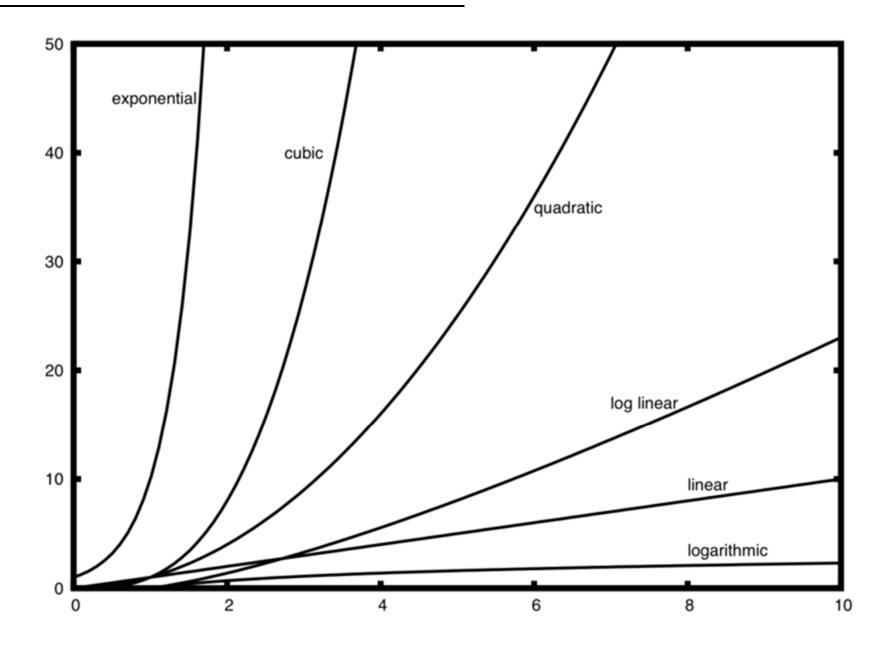
$$log(n!) =$$

- 1. $O(\log n)$
- O(n)
- 3. O(n log n)
- 4. $O(n^2)$
- 5. $O(2^n)$

Hint: Sterling's Approximation

$$n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$

In General



Model of Computation?

Different ways to "compute":

- Sequential (RAM) model of computation
- Parallel (PRAM, BSP, Map-Reduce)
- Circuits
- Turing Machine
- Counter machine
- Word RAM model
- Quantum computation
- Etc.

Model of Computation

Sequential Computer

One thing at a time

All operations take constant time
 Addition subtraction multiplication compariso

Addition, subtraction, multiplication, comparison

Algorithm Analysis

Example:

```
void sum(int k, int[] intArray) {
   int total=0;
   for (int i=0; i<= k; i++) {
        total = total + intArray[i];
        k array access
        k addition
        k assignment

1 assignment
        k+1 comparisons
        k increments
        k array access
        k addition
        k assignment

1 return</pre>
```

Total:
$$1 + 1 + (k+1) + 3k + 1 = 4k+4 = O(k)$$

Algorithm Analysis

Example:

What is the cost of this operation?

```
void sum(int k, int[] intArray) {
  int total=0;
  String name="Stephanie";
  for (int i=0; i <= k; i++) {
       total = total + intArray[i];
       name = name + "?"
  return total;
```

Not 1! Not constant! Not k!

Moral: all costs are not 1.

Loops

cost = (# iterations)x(max cost of one iteration)

```
int sum(int k, int[] intArray) {
   int total=0;
   for (int i=0; i<= k; i++) {
      total = total + intArray[i];
   }
   return total;
}</pre>
```

Nested Loops

cost = (# iterations)(max cost of one iteration)

```
int sum(int k, int[] intArray) {
   int total=0;
   for (int i=0; i <= k; i++) {
     for (int j=0; j <= k; j++) {
          total = total + intArray[i];
  return total;
```

Sequential statements

cost = (cost of first) + (cost of second)

```
int sum(int k, int[] intArray) {
   for (int i=0; i<= k; i++)
        intArray[i] = k;
   for (int j =0; j<= k; j++)
        total = total + intArray[i];
   return total;
}</pre>
```

```
if / else statements
cost = max(cost of first, cost of second)
  <= (cost of first) + (cost of second)</pre>
```

```
void sum(int k, int[] intArray) {
   if (k > 100)
        doExpensiveOperation();
   else
        doCheapOperation();
   return;
}
```

For recursive function calls.....



Recurrences

 $= O(2^n)$

if (n <= 1)

else

return n;

```
T(n) = 1 + T(n-1) + T(n-2)
                                        = 1 + \left(1 + T(n-2) + T(n-3)\right)
                                                 4 (1+T(n-3)+T(n-4))
T(n) = 1 + T(n - 1) + T(n - 2)
                                        = 3 + T(n-2)+3T(9-3)+T
                                               T(n-2)
                                               T(n/X)
                                    T(n-1)
              int fib(int n) {
                     return fib(n-1) + fib(n-2);
```

What is the running time?

```
for (int i = 0; i<n; i++)

for (int j = 0; j<i; j++)

store[i] = i + j;</pre>
```

- 1. O(1)
- O(n)
- 3. $O(n \log n)$
- $\widehat{4.} O(n^2)$
 - 5. $O(n^2 \log n)$
 - 6. $O(2^n)$

Today: Divide and Conquer!

Algorithm Analysis

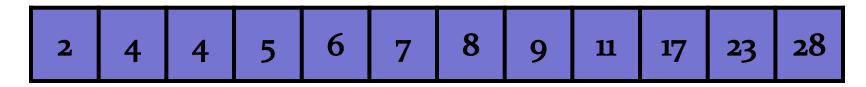
- Big-O Notation
- Model of computation

Searching

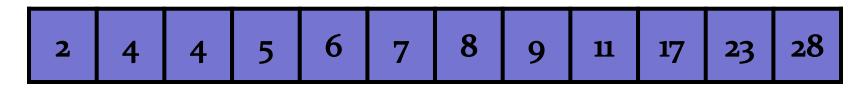
Peak Finding

- 1-dimension
- 2-dimensions

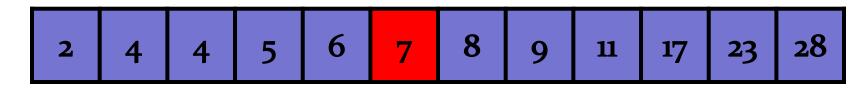
Sorted array: A [0 . . n-1]



Sorted array: A [0..n-1]



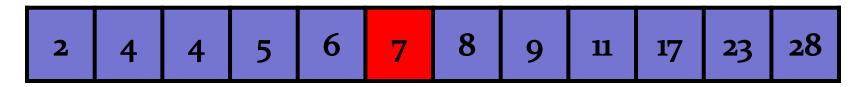
Sorted array: A [0..n-1]



Search for 17 in array A.

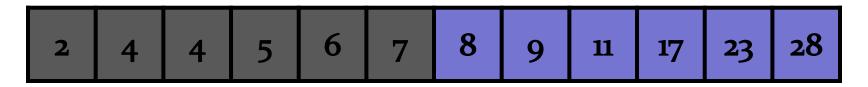
- Find middle element: 7

Sorted array: A [0..n-1]



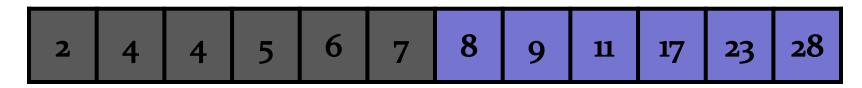
- Find middle element: 7
- Compare 17 to middle element: 17 > 7

Sorted array: A [0..n-1]



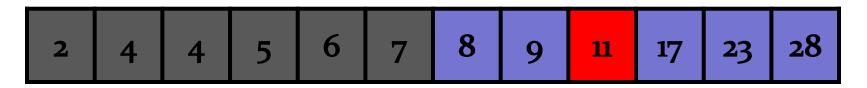
- Find middle element: 7
- Compare 17 to middle element: 17 > 7

Sorted array: A [0..n-1]



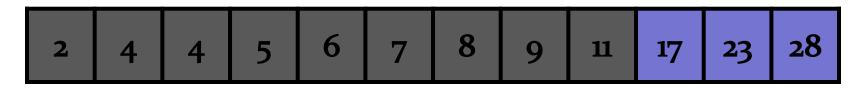
- Find middle element: 7
- Compare 17 to middle element: 17 > 7
- Recurse on right half

Sorted array: A [0..n-1]



- Find middle element
- Compare 17 to middle element
- Recurse

Sorted array: A [0..n-1]



- Find middle element
- Compare 17 to middle element
- Recurse

Sorted array: A [0..n-1]



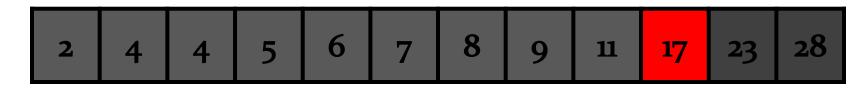
- Find middle element
- Compare 17 to middle element
- Recurse

Sorted array: A [0..n-1]



- Find middle element
- Compare 17 to middle element
- Recurse

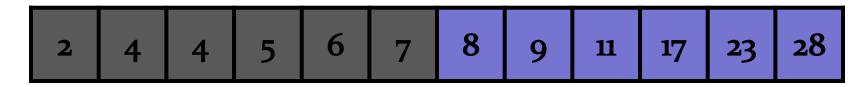
Sorted array: A [0..n-1]



- Find middle element
- Compare 17 to middle element
- Recurse

Problem Solving: Reduce the Problem

Sorted array: A [0..n-1]



Reduce-and-Conquer:

- Start with *n* elements to search.
- Eliminate half of them.
- End with n/2 elements to search.
- Repeat.

programming pearls

By Jon Bentley

WRITING CORRECT PROGRAMS

The Challenge of Binary Search

Even with the best of designs, every now and then a programmer has to write subtle code. This column is about one problem that requires particularly careful code: binary search. After defining the problem and sketching an algorithm to solve it, we'll use principles of program verification in several stages as we develop the program.



Jon Bentley

Most programmers think that with the above description in hand, writing the code is easy; they're wrong. The only way you'll believe this is by putting down this column right now, and writing the code yourself. Try it.

programming pearls

By Jon Bentley

I've given this problem as an in-class assignment in courses at Bell Labs and IBM. The professional programmers had one hour (sometimes more) to convert the above description into a program in the language of their choice; a high-level pseudocode was fine. At the end of the specified time, almost all the programmers reported that they had correct code for the task. We would then take 30 minutes to examine their code, which the programmers did with test cases. In many different classes and with over a hundred programmers, the results varied little: 90 percent of the programmers found bugs in their code (and I wasn't always convinced of the correctness of the code in which no bugs were found).

I found this amazing: only about 10 percent of professional programmers were able to get this small program right. But they aren't the only ones to find this task difficult. In the history in Section 6.2.1 of his Sorting and Searching, Knuth points out that while the first binary search was published in 1946, the first published binary search without bugs did not appear until 1962.



Jon Bentley

Binary Search (buggy)

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
Search (A, key, n)
    begin = 0
    end = n
    while begin != end do:
         if key < A[(begin+end)/2] then</pre>
                end = (begin+end)/2 - 1
         else begin = (begin+end)/2
    return A[begin]
```

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
Search (A, key, n)
    begin = 0
                                  array out of bounds!
     end = n \leftarrow
    while begin != end do
           if key < A[ //begin+end) /2] then
           end = (begin+end)/2 - 1
else begin = (begin+end)/2
     return A[end]
```

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
Search (A, key, n)
                            array out of bounds!
    begin = 0
                            (Can't happen because of other bugs...)
    while begin != end/do:
          if key < A[(begin+end)/2] then</pre>
                  end \neq (begin+end)/2 - 1
                       = (begin+end)/2
    return A[end]
```

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
Search (A, key, n)
    begin = 0
    end = n-1
    while begin != end do:
         if key < A[(begin+end)/2] then</pre>
                end = (begin+end)/2 - 1
         else begin = (begin+end)/2
    return A[end]
```

Sorted arra

2 4

Search (A,

Example: search(7)

- begin = 0, end = 1
- mid = (0+1)/2 = 0
- key >= A[mid] → begin = 0

5 10

May not terminate!

end = n-1

begin = 0

while begin != end do:

key < A[(begin+end)/2] then

round down

and - (bogin and) /2 1

end = (begin+end)/2 - 1

else begin = (begin+end)/2

return A[end]

Sorted arra

2 4

Search (A,

```
Example: search(2)
```

- begin = 0, end = 1
- mid = (0+1)/2 = 0
- key $< A[mid] \rightarrow end = 0 1 = -1$

5 10

```
begin = 0
end < begin
end = n-1
while begin != end do:
    if key < A[(begin+end)/2] then
        end = (begin+end)/2 - 1
else begin = (begin+end)/2
return A[end]</pre>
```

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
Search (A, key, n)
   begin = 0
    end = n-1
    while begin != end do:
         if key < A[(begin+end)/2] then</pre>
               end = (begin+end)/2 - 1
         else begin = (begin+end)/2
    return A[end] ← Useful return value?
```

Specification:

- Returns element if it is in the array.
- Returns "null" if it is not in the array.

Alternate Specification:

- Returns index if it is in the array.
- Returns -1 if it is not in the array.

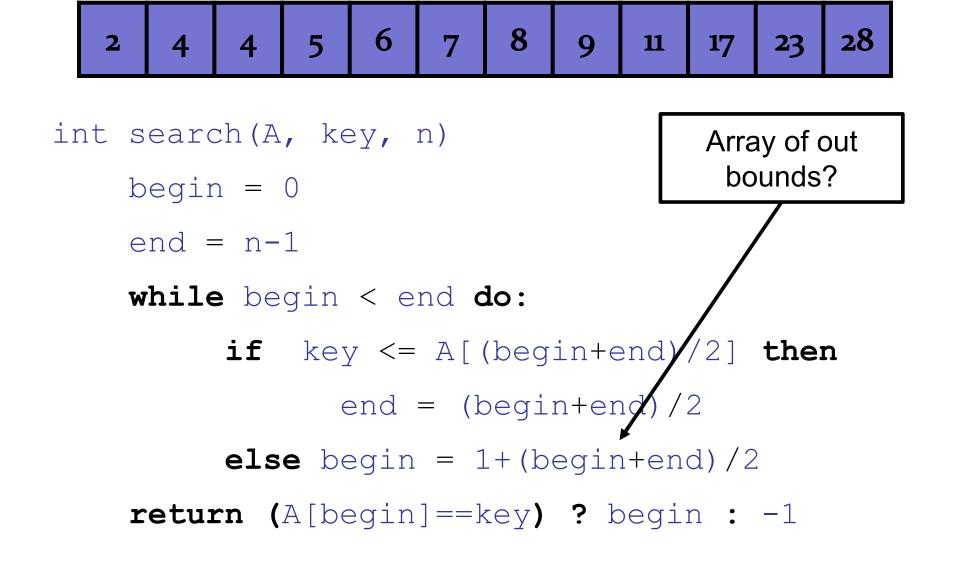
```
Sorted array: A [0.p-1]
                               7 8
                   6
                           8
                                               28
               5
                                   11
                                       17
                                           23
                                         50900 beg 66

ord=5 end=2

mid=1
                                   beg=0
int search (A, key, n)
                                   Wid = 5
    begin = 0
    end = n-1
    while begin < end do:
              key <= A[(begin+end)/2] then
                 end = (begin+end)/2
          else begin = 1+(begin+end)/2
    return (A[begin] == key) ? begin : -1
```

```
6
                                  11
                                     17
                                         23
int search (A, key, n)
                                less-than-or-equal
    begin = 0
    end = n-1
    while begin < end do
         if key < = A[(begin+end)/2] then
                end = (begin+end)/2
         else begin = 1+(begin+end)/2
    return (A[begin] == key) ? begin : -1
```

```
6
                                     11
                                         17
                                             23
int search (A, key, n)
                                   strictly greater than
    begin = 0
    end = n-1
    while begin < end do:</pre>
          if key <= A[(begin+end)/2] then</pre>
                 end = (begin+end)/2
          else begin = 1 + (begin + end) / 2
    return (A[begin] == key) ? begin : -1
```



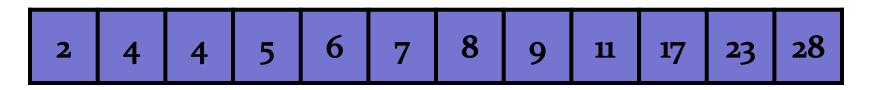


```
int search (A, key, n)
                                        Array of out
                                         bounds?
    begin = 0
                                        No: division
    end = n-1
                                       rounds down.
    while begin < end do:</pre>
              key \le A[(begin+end)/Z]
                 end = (begin+end)/2
          else begin = 1 + (begin + end) / 2
    return (A[begin] == key) ? begin : -1
```

Bug 4

```
6
                                    11
                                       17
                                           23
int search (A, key,
                         What if begin > MAX INT/2?
    begin = 0
    end = n-1
    while begin < end do:</pre>
          if key <= A[(begin+end)/2] then</pre>
                 end = (begin+end)/2
          else begin = 1+(begin+end)/2
    return (A[begin] == key) ? begin : -1
```

Sorted array: A [0..n-1]



```
int search(A, key,

begin = 0

end = n-1
```

What if begin > MAX_INT/2?

Overflow error: begin+end > MAX_INT

while begin < end do:</pre>

```
if key <= A[(begin+end)/2] then
        end = (begin+end)/2

else begin = 1+(begin+end)/2

return (A[begin]==key) ? begin : -1</pre>
```

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
int search (A, key, n)
    begin = 0
    end = n-1
    while begin < end do:
         mid = begin + (end-begin)/2;
         if key <= A[mid] then</pre>
                end = mid
         else begin = mid+1
    return (A[begin] == key) ? begin : -1
```

Moral of the Story

Easy algorithms are *hard* to write correctly.

Binary search is 8 lines of code.

If you can't write 8 correct lines of code, how do you expect to write thousands of lines of bug-free code??

Precondition and Postcondition

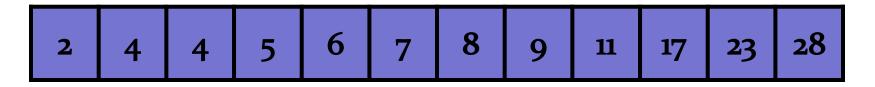
Precondition:

- Fact that is true when the function begins.
- Usually important for it to work correctly.

Postcondition:

- Fact that is true when the function ends.
- Usually useful to show that the computation was done correctly.





```
int search(A, key, n)
                                      What are useful
    begin = 0
                                     preconditions and
                                      postconditions?
    end = n-1
    while begin < end do:
          mid = begin + (end-begin)/2;
          if key <= A[mid] then</pre>
                 end = mid
          else begin = mid+1
    return (A[begin] == key) ? begin : -1
```

Functionality:

- If element is in the array, return index of element.
- If element is not in array, return -1.

Functionality:

- If element is in the array, return index of element.
- If element is not in array, return -1.

Preconditions:

- Array is of size n
- Array is sorted

Functionality:

- If element is in the array, return index of element.
- If element is not in array, return -1.

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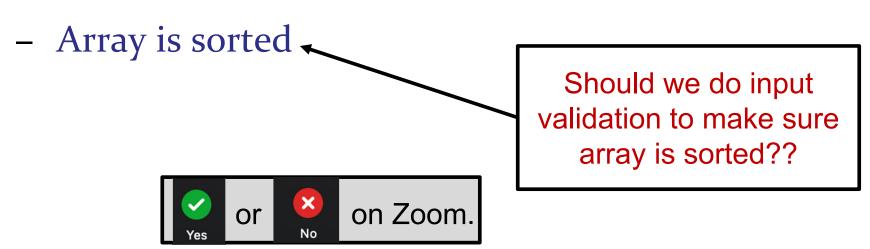
You can usually check this directly.

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- If element is not in array, return -1.

Preconditions:

Array is of size n



Functionality:

- If element is in the array, return index of element.
- If element is not in array, return -1.

Preconditions:

- Array is of size n
- Array is sorted •

Should we do input validation to make sure array is sorted??

NO! Too slow!

Functionality:

- If element is in the array, return index of element.
- If element is not in array, return -1.

Preconditions:

- Array is of size n
- Array is sorted

Postcondition:

- If element is in the array: A [begin] = key

Invariants

Invariant:

relationship between variables that is always true.

Invariants

Invariant:

relationship between variables that is always true.

Loop Invariant:

 relationship between variables that is true at the beginning (or end) of each iteration of a loop.



Sorted array: A [0..n-1]

int search (A, key, n)



```
begin = 0
end = n-1

while begin < end do:
    mid = begin + (end-begin) / 2;
    if key <= A[mid] then
        end = mid

else begin = mid+1

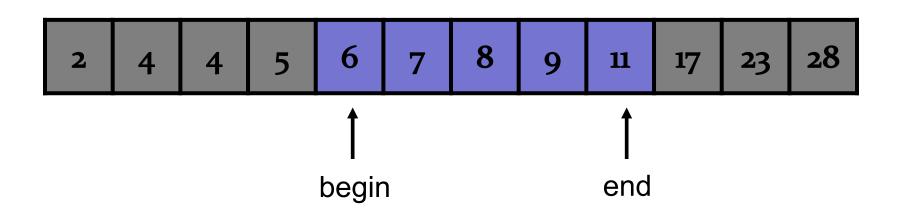
return (A[begin] == key) ? begin : -1</pre>
```

Loop invariant:

 $- A[begin] \le key \le A[end]$

Interpretation:

- The key is in the range of the array



Loop invariant:

 $- A[begin] \le key \le A[end]$

Interpretation:

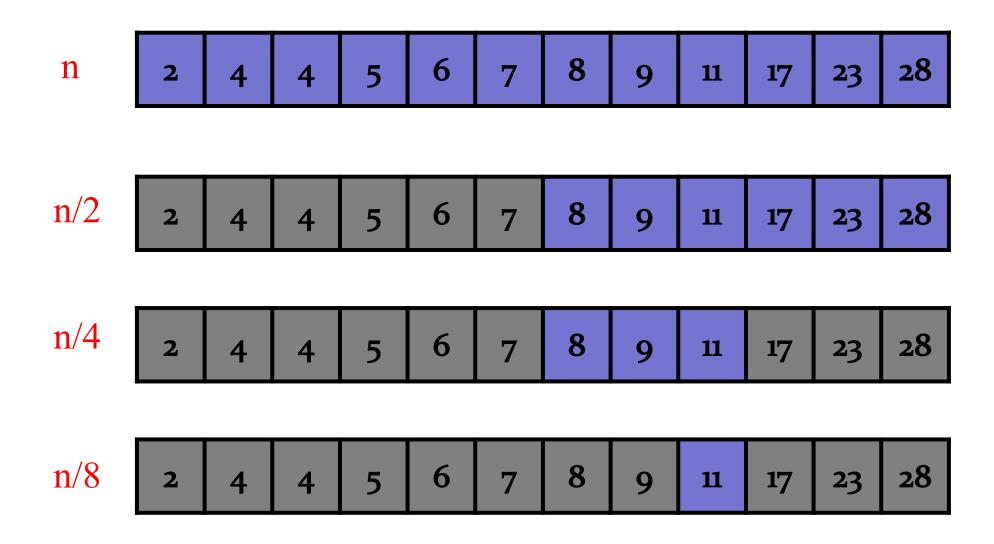
- The key is in the range of the array

Error checking:

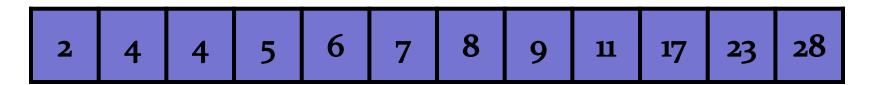
```
if ((A[begin] > key) or (A[end] < key))
     System.out.println("error");</pre>
```

```
2 4 4 5 6 7 8 9 11 17 23 28
```

```
int search (A, key, n)
    begin = 0
    end = n-1
    while begin < end do:
         mid = begin + (end-begin)/2;
         if key <= A[mid] then</pre>
                end = mid
         else begin = mid+1
    return (A[begin] == key) ? begin : -1
```



Sorted array: A [0..n-1]



Iteration 1: (end - begin) = n

Iteration 2: (end - begin) = n/2

Iteration 3: (end - begin) = n/4

. . .

Iteration k: (end – begin) = $n/2^k$

Another invariant!

$$n/2^k = 1$$
 \rightarrow $k = \log(n)$

Key Invariants:

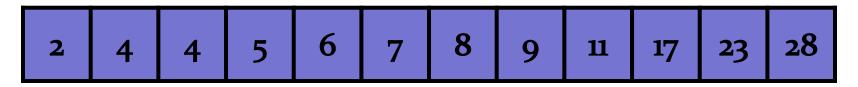
Correctness:

- A[begin] \leq key \leq A[end]

Performance:

- (end-begin) \leq n/2^k in iteration k.

Sorted array: A[o..n-1]



Not just for searching arrays:

Assume a complicated function:

int complicatedFunction(int s)

Assume the function is always increasing:

complicatedFunction(i) < complicatedFunction(i+1)</pre>

Find the minimum value j such that:

complicatedFunction(j) > 100

Tutorial allocation

Tutorial allocation

T₁

T₂

T₃

Tutorials

(in order of tutor preference)

T4

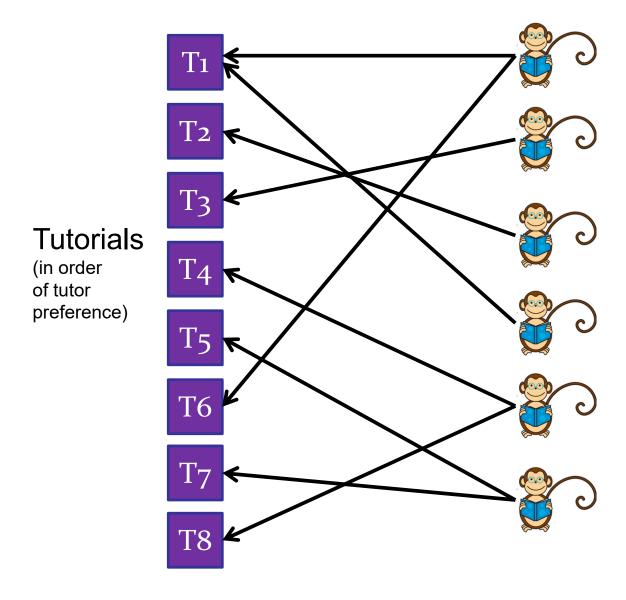
T₅

T6

T₇

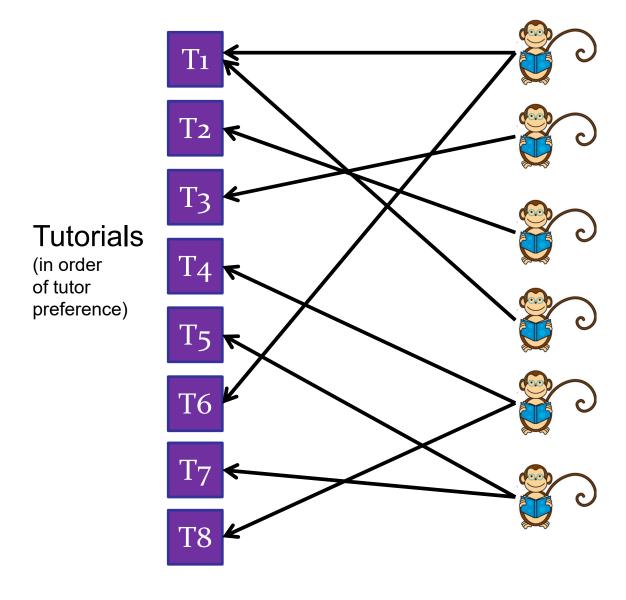
T8

Tutorial allocation



Students want certain tutorials.

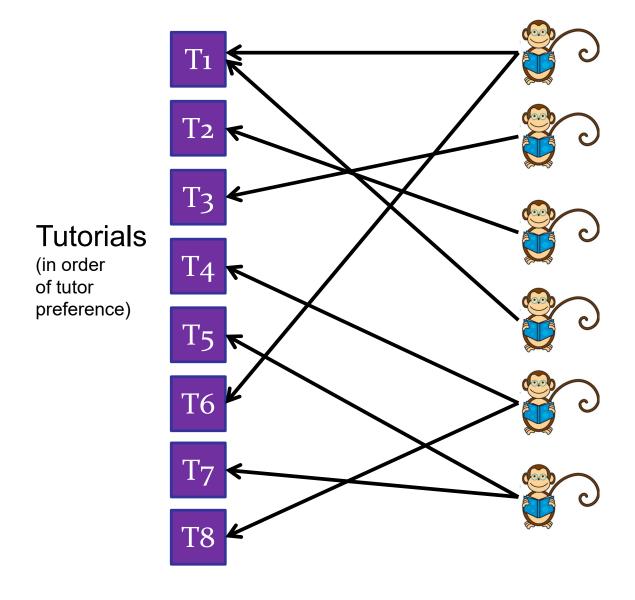
Tutorial allocation



Students want certain tutorials.

We want each tutorial to have < 15 students...

Tutorial allocation

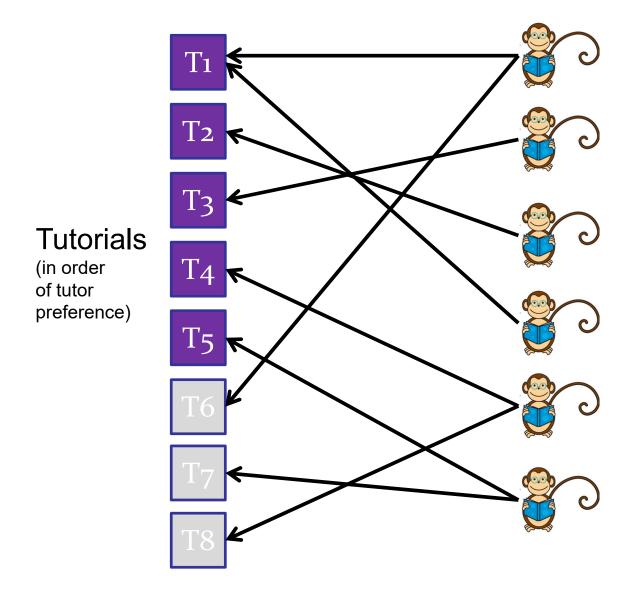


Students want certain tutorials.

We want each tutorial to have < 15 students...

How many tutorials do we need to run?

Tutorial allocation



Students want certain tutorials.

We want each tutorial to have < 15 students...

How many tutorials do we need to run?

Tutorial allocation

 T_1 T2 T₃ **Tutorials** (in order of tutor preference) T₅ T6 **T8**

Can we do greedy allocation?

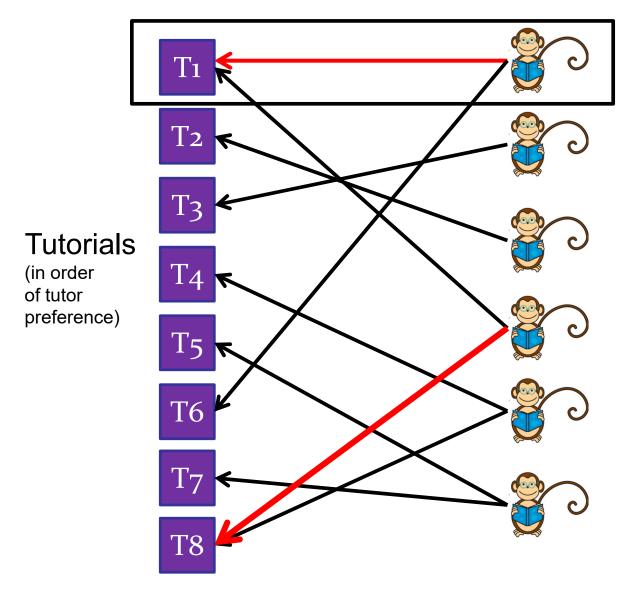
First, fill T1. Then fill T2. Then fill T3.

Stop when all

students are allocated



Tutorial allocation



Can we do greedy allocation?

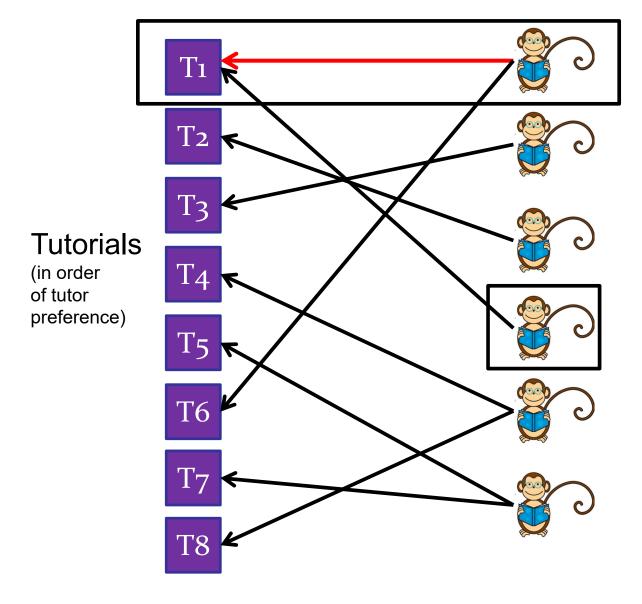
NO

Assume max tutorial size is 1.

Assign student 1 to tutorial 1.

Now we need all 8 tutorials.

Tutorial allocation



Can we do greedy allocation?

NO

Assume max tutorial size is 1.

Assign student 1 to tutorial 1.

Now one student has no feasible allocation!

Tutorial allocation

T2 **Tutorials** (in order of tutor preference) T5 **T6** T8

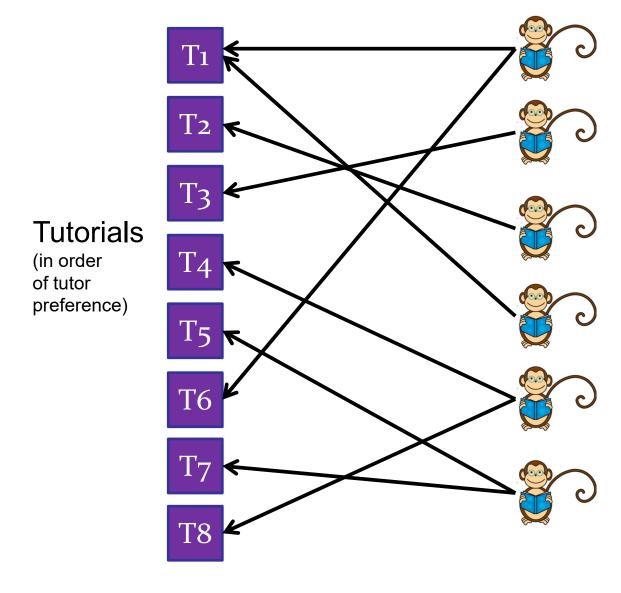
Assume we can solve allocation problem:

Given a fixed set of tutorials and a fixed set of students, find an allocation where every student has a slot.

Warning:

- may be > 15 students in a slot!
- minimizes max students in a slot.

Tutorial allocation



How to find minimum number of tutorials that we need to open to ensure: no tutorial has more than 15 students.

Tutorial allocation

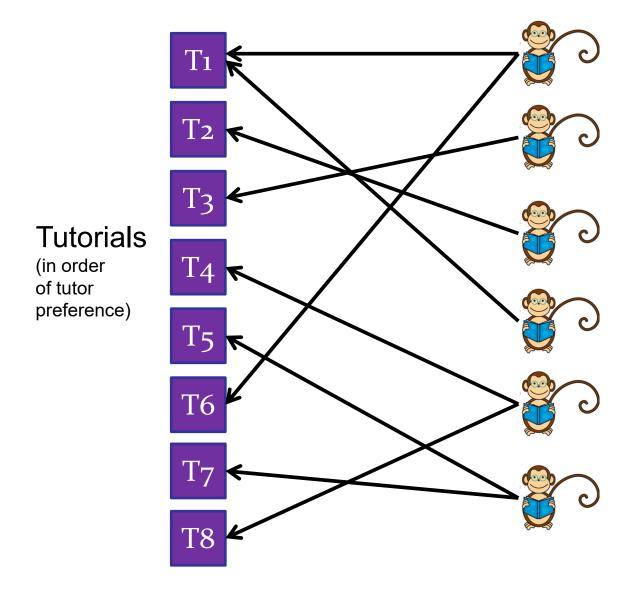
 T_1 T2 T3 **Tutorials** (in order of tutor preference) T₅ T6 T8

Observation:

Number of students in WORST a tutorial only decreases as number of tutorials increases.

Monotonic function of number of tutorials!

Tutorial allocation



Monotonic function of number of tutorials!



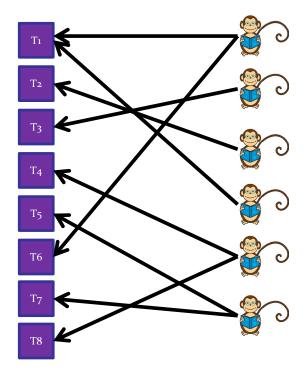
Binary Search

Tutorial allocation

Solution:

Binary Search

Define:

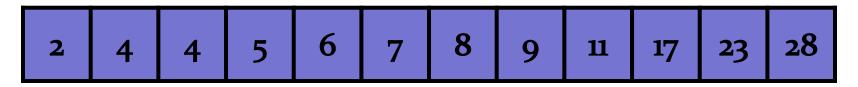


MaxStudents(x) = number of students in most crowded tutorial, if we offer x tutorials.

return begin

```
MaxStudents(x) = number of students in most
               crowded tutorial,
               if we offer x tutorials.
 Search (n)
      begin = 0
      end = n-1
      while begin < end do:</pre>
            mid = begin + (end-begin)/2;
            if MaxStudents(mid) <= 15 then</pre>
                    end = mid
            else begin = mid+1
```

Sorted array: A[o..n-1]



Not just for searching arrays:

Assume a complicated function:

int complicatedFunction(int s)

Assume the function is always increasing:

complicatedFunction(i) < complicatedFunction(i+1)</pre>

Find the minimum value j such that:

complicatedFunction(j) > 100

Today: Divide and Conquer!

Algorithm Analysis

- Big-O Notation
- Model of computation

Searching

Peak Finding

- 1-dimension
- 2-dimensions