

## Introduction

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- Algorithms and Data Structures
- Peaks
  - Algorithm 1
  - Algorithm 2
  - Algorithm 3

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## Algorithms and Data Structures

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- **Algorithmic problem.** Precisely defined relation between input and output.
- **Algorithm.** Method to solve an algorithmic problem.
  - **Discrete** and **unambiguous** steps.
  - Mathematical abstraction of a program.
- **Data structure.** Method for organizing data to enable queries and updates.

## Example: Find Max

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- **Find max.** Given an array  $A[0..n-1]$ , find an index  $i$ , such that  $A[i]$  is maximal.
  - **Input.** Array  $A[0..n-1]$ .
  - **Output.** An index  $i$  such that  $A[i] \geq A[j]$  for all indices  $j \neq i$ .
- **Algorithm.**
  - Process  $A$  from left to right and maintain value and index of maximal value seen so far.
  - Return index.

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	3	7	15	17	11	2	3	6	8	7	5	9	5	23

## Description of Algorithms

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- **Natural language.**
  - Process A from left-to-right and maintain value and index of maximal value seen so far.
  - Return index.
- **Program.**
- **Pseudocode.**

```
def findMax(A):  
    m = 0  
    for i in range(len(A)):  
        if A[i] > A[m]:  
            m = i  
    return m
```

```
FINDMAX(A, n)  
max = 0  
for i = 0 to n-1  
    if (A[i] > A[max]) max = i  
return max
```

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## Peaks

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- **Peak.**  $A[i]$  is a **peak** if  $A[i]$  is as least as large as its **neighbors**:
- $A[i]$  is a peak if  $A[i-1] \leq A[i] \geq A[i+1]$  for  $i \in \{1, \dots, n-2\}$
- $A[0]$  is a peak if  $A[0] \geq A[1]$ .
- $A[n-1]$  is a peak if  $A[n-2] \leq A[n-1]$ .

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	3	7	15	17	11	2	3	6	8	7	5	9	5	23

- **Peak finding.** Given an array  $A[0..n-1]$ , find **an** index  $i$  such that  $A[i]$  is a peak.
  - **Input.** A array  $A[0..n-1]$ .
  - **Output.** An index  $i$  such that  $A[i]$  is a peak.

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## Algorithm 1

- **Algorithm 1.** For each entry check if it is a peak. Return the index of the first peak.

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	3	7	15	17	11	2	3	6	8	7	5	9	5	23

- **Pseudocode.**

```
PEAK1(A, n)
  if A[0] ≥ A[1] return 0
  for i = 1 to n-2
    if A[i-1] ≤ A[i] ≥ A[i+1] return i
  if A[n-2] ≤ A[n-1] return n-1
```

- **Challenge.** How do we analyze the algorithm?

## Theoretical Analysis

- **Running time/time complexity.**
  - $T(n)$  = number of **steps** that the algorithm performs on input of size  $n$ .
- **Steps.**
  - Read/write to memory ( $x := y, A[i], i = i + 1, \dots$ )
  - Arithmetic/boolean operations ( $+, -, *, /, \%, \&\&, \&, |, \wedge, \sim$ )
  - Comparisons ( $<, >, =, \leq, \geq, =, \neq$ )
  - Program flow (if-then-else, while, for, goto, function call, ..)
- **Worst-case time complexity.** Maximal running time over all inputs of size  $n$ .

## Theoretical Analysis

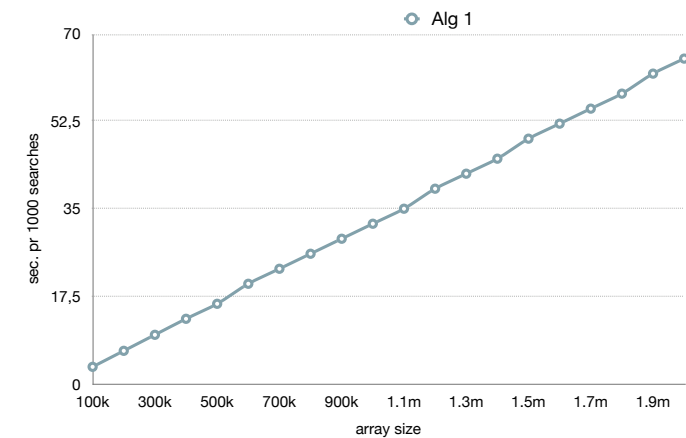
- **Running time.** What is the running time  $T(n)$  for algorithm 1?

```
PEAK1(A, n)
  if A[0] ≥ A[1] return 0
  for i = 1 to n-2
    if A[i-1] ≤ A[i] ≥ A[i+1] return i
  if A[n-2] ≤ A[n-1] return n-1
```

$c_1$   
 $(n-2) \cdot c_2$   
 $c_3$

$$T(n) = c_1 + (n-2) \cdot c_2 + c_3$$

- $T(n)$  is a linear function of  $n$ :  $T(n) = an + b$
- **Asymptotic notation.**  $T(n) = O(n)$
- **Experimental analysis.**
  - What is the experimental running time of algorithm 1?
  - How does the experimental analysis compare to the theoretical analysis?



## Peaks

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- Algorithm 1 finds a peak in  $O(n)$  time.
- Theoretical and experimental analysis agrees.
- **Challenge.** Can we do better?

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## Algorithm 2

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- **Observation.** A maximal entry  $A[i]$  is a peak.
- **Algorithm 2.** Find a maximal entry in  $A$  with  $\text{FINDMAX}(A, n)$ .

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	3	7	15	17	11	2	3	6	8	7	5	9	5	23

```
FINDMAX(A, n)
max = 0
for i = 0 to n-1
    if (A[i] > A[max]) max = i
return max
```

## Theoretical Analysis

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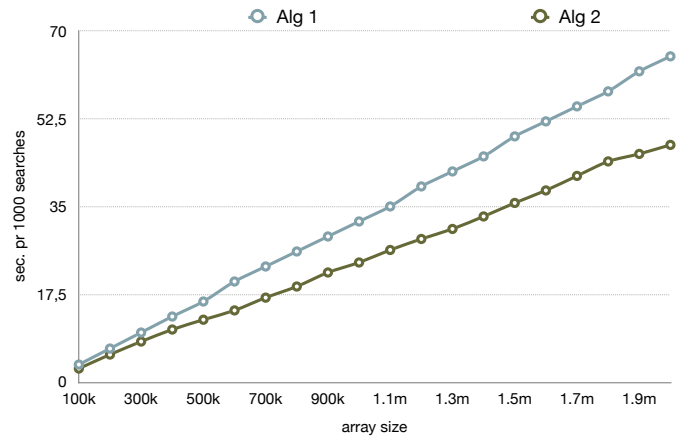
- **Running time.** What is the running time  $T(n)$  for algorithm 2?

```
FINDMAX(A, n)
max = 0
for i = 0 to n-1
    if (A[i] > A[max]) max = i
return max
```

$C_4$   
 $n \cdot C_5$   
 $C_6$

$$T(n) = C_4 + n \cdot C_5 + C_6 = O(n)$$

- **Experimental analysis.** Better constants?



## Peaks

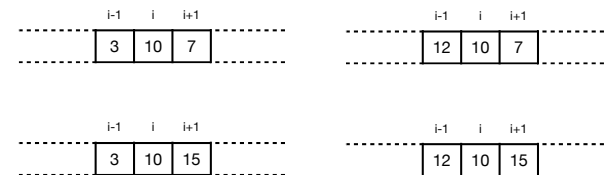
- **Theoretical analysis.**
  - Algorithm 1 and 2 find a peak in  $O(n)$  time.
- **Experimental analysis.**
  - Algorithm 1 and 2 run in  $O(n)$  time in practice.
  - Algorithm 2 is a constant factor faster than algorithm 1.
- **Challenge.** Can we do **significantly** better?

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## Algorithm 3

- **Clever idea.**
  - Consider any entry  $A[i]$  and its neighbors  $A[i-1]$  and  $A[i+1]$ .
  - Where can a peak be relative to  $A[i]$ ?
    - Neighbor are  $\leq A[i] \Rightarrow A[i]$  is a peak.
    - Otherwise  $A$  is increasing in at **least** one direction  $\Rightarrow$  peak must exist in that direction.



- **Challenge.** How can we turn this into a fast algorithm?

## Algorithm 3

### Algorithm 3.

- Consider the **middle** entry  $A[m]$  and neighbors  $A[m-1]$  and  $A[m+1]$ .
- If  $A[m]$  is a peak, return  $m$ .
- Otherwise, continue search **recursively** in half with the increasing neighbor.

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	3	7	15	17	11	2	3	6	8	7	5	9	5	23

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
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## Algorithm 3

### Algorithm 3.

- Consider the **middle** entry  $A[m]$  and neighbors  $A[m-1]$  and  $A[m+1]$ .
- If  $A[m]$  is a peak, return  $m$ .
- Otherwise, continue search **recursively** in half with the increasing neighbor.

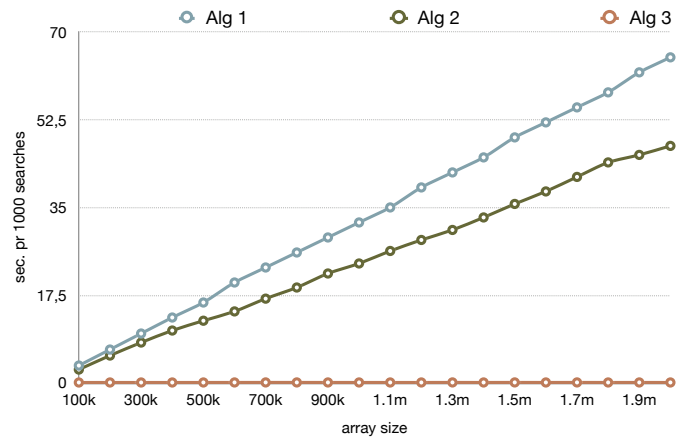
```
PEAK3(A, i, j)
  m = ⌊(i+j)/2⌋
  if A[m] ≥ neighbors return m
  elseif A[m-1] > A[m]
    return PEAK3(A, i, m-1)
  elseif A[m] < A[m+1]
    return PEAK3(A, m+1, j)
```

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
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## Algorithm 3

```
PEAK3(A, i, j)
  m = ⌊(i+j)/2⌋
  if A[m] ≥ neighbors return m
  elseif A[m-1] > A[m]
    return PEAK3(A, i, m-1)
  elseif A[m] < A[m+1]
    return PEAK3(A, m+1, j)
```

- Running time.**
- A recursive call takes constant time.
- How many recursive calls?
  - A recursive call **halves** the size of the interval. We stop when the array has size 1.
    - 1<sup>st</sup> recursive call:  $n/2$
    - 2<sup>nd</sup> recursive call:  $n/4$
    - ....
    - $k^{\text{th}}$  recursive call:  $n/2^k$
    - ....
  - ⇒ After  $\sim \log_2 n$  recursive call array has size  $\leq 1$ .
  - ⇒ Running time is  $O(\log n)$
- Experimental analysis.** Significantly better?



## Peaks

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- **Theoretical analysis.**
  - Algorithm 1 and 2 finds a peak in  $O(n)$  time.
  - Algorithm 3 finds a peak in  $O(\log n)$  time.
- **Experimental analysis.**
  - Algorithm 1 and 2 run in  $O(n)$  time in practice.
  - Algorithm 2 is a constant factor faster than algorithm 1.
  - Algorithm 3 is much, much faster than algorithm 1 and 3.

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