

Introduction

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

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Algorithms and Data Structures

- **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

- **Find max.** Given a 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

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

```
public static int findMax(int[] A) {  
    int max = 0;  
    for(i = 0; i < A.length; i++)  
        if (A[i] > A[max]) max = i;  
    return max;  
}
```

```
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

- Peak. $A[i]$ is a **peak** if $A[i]$ is at 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
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- Peak finding. Given an array $A[0..n-1]$, find **an** index i such that $A[i]$ is a peak.
 - Input. An 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 ($x := y$, $A[i]$, $i = i + 1$, ...)
 - Arithmetic/boolean operations ($+$, $-$, $*$, $/$, $\%$, $\&\&$, $\|$, $\&$, $|$, \wedge , \sim)
 - Comparisons ($<$, $>$, $=<$, $=>$, $=$, \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
```

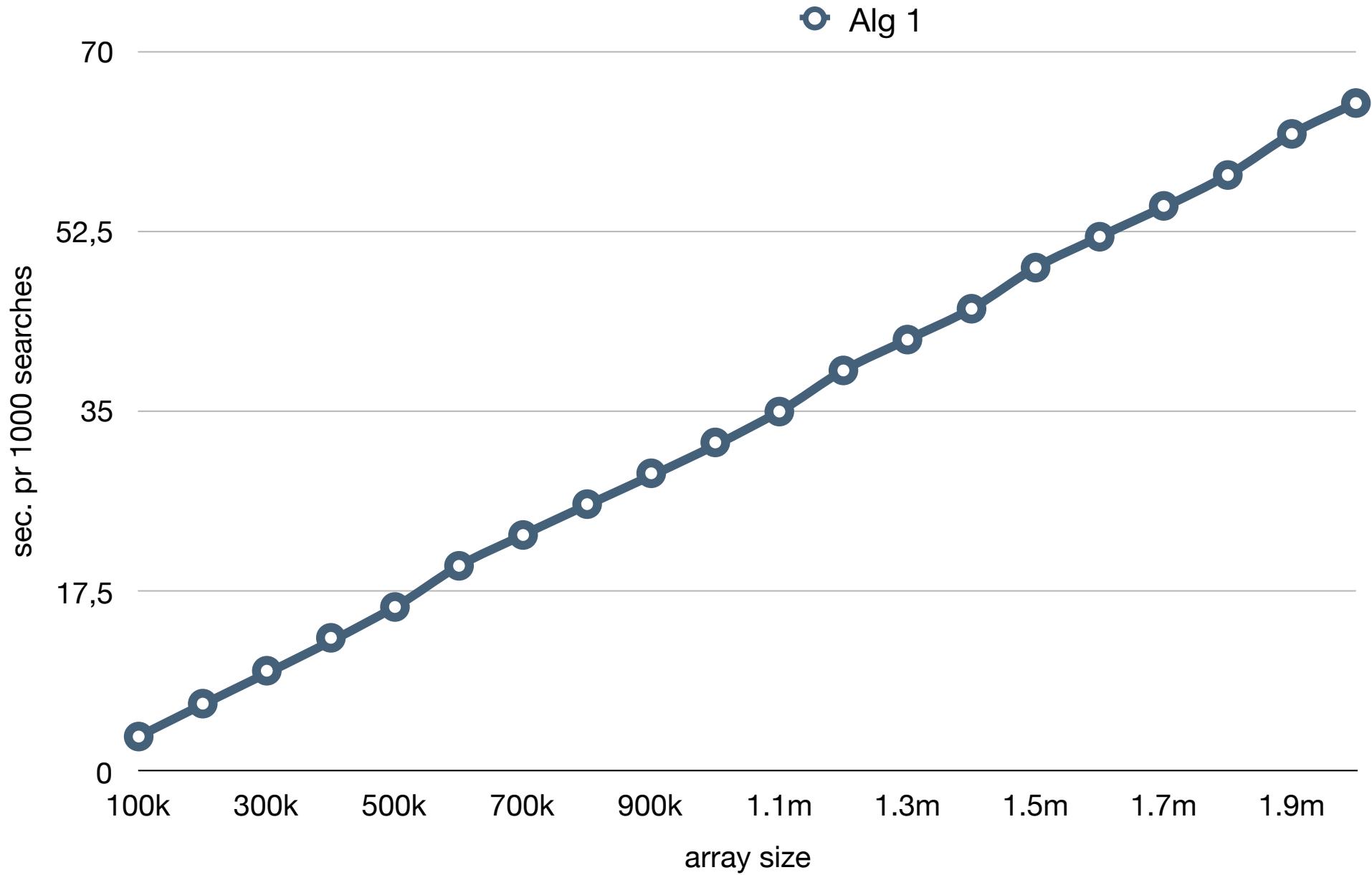
c_1

$(n-2) \cdot c_2$

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.** $T(n) = \Theta(n)$
- **Experimental analysis.**
 - What is the experimental running time of algorithm 1?
 - How does the experimental analysis compare to the theoretical analysis?



Peaks

- Algorithm 1 finds a peak in $\Theta(n)$ time.
- Theoretical and experimental analysis agrees.
- [Challenge](#). Can we do better?

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Algorithm 2

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

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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

- **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
```

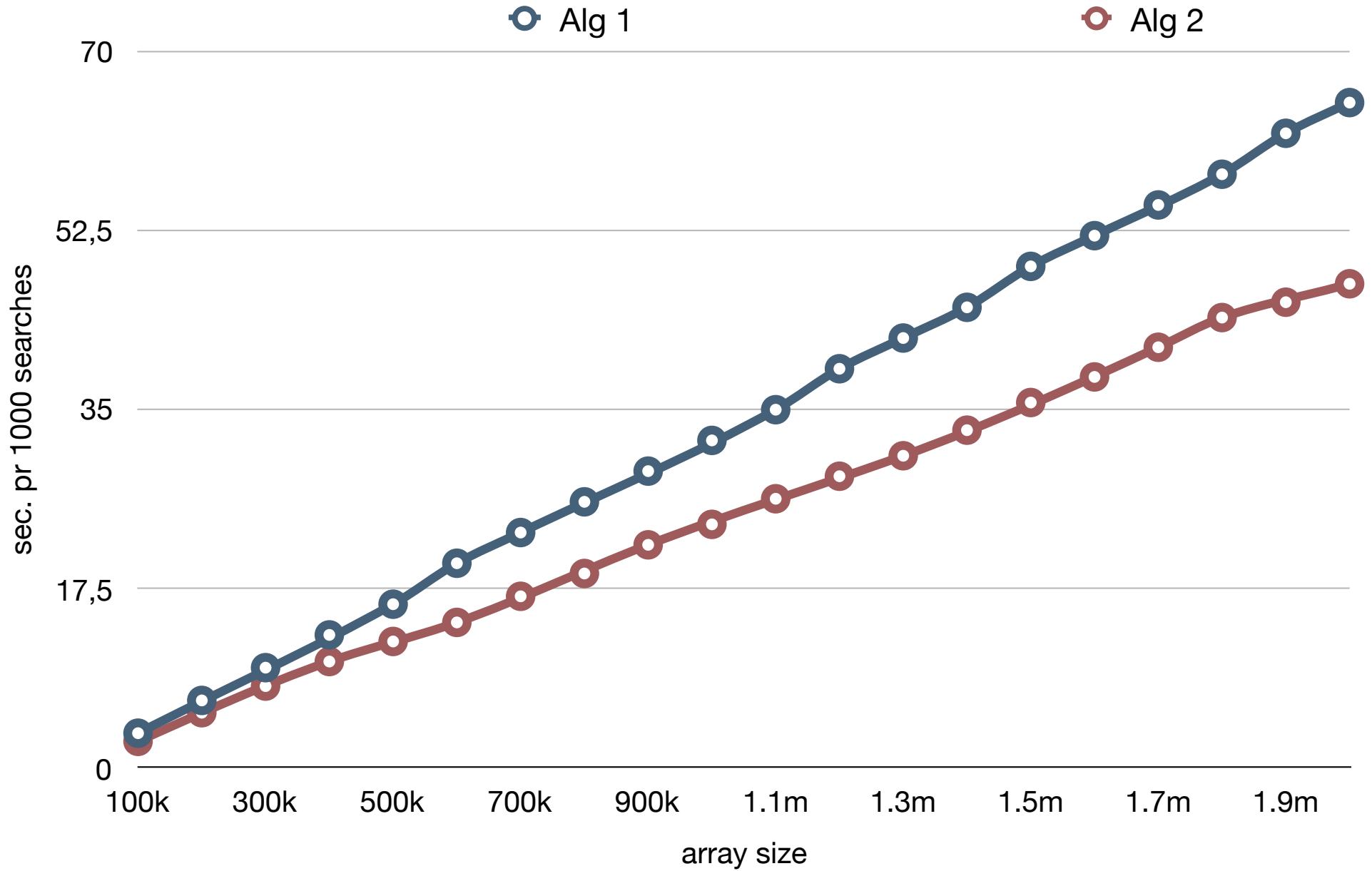
C₄

n·C₅

C₆

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

- **Experimental analysis.** Better constants?



Peaks

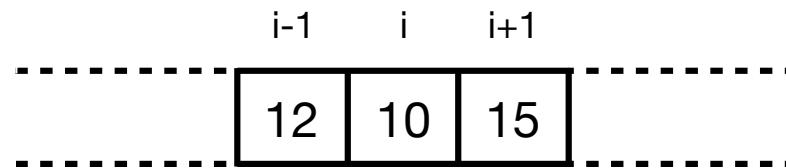
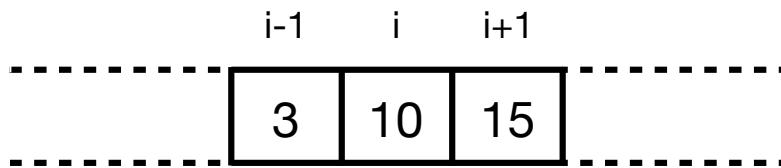
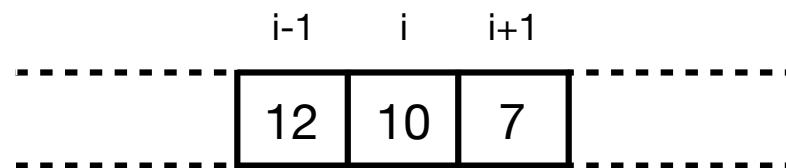
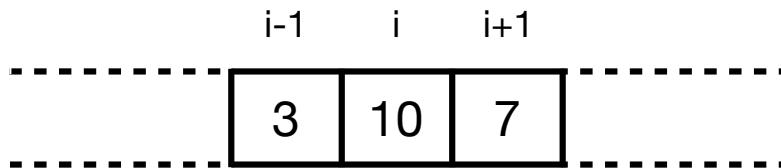
- Theoretical analysis.
 - Algorithm 1 and 2 find a peak in $\Theta(n)$ time.
- Experimental analysis.
 - Algorithm 1 and 2 run in $\Theta(n)$ time in practice.
 - Algorithm 2 is a constant factor faster than algorithm 1.
- Challenge. Can we do significantly better?

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Algorithm 3

- Clever idea.
 - Consider any entry $A[i]$ and it's 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

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)
```

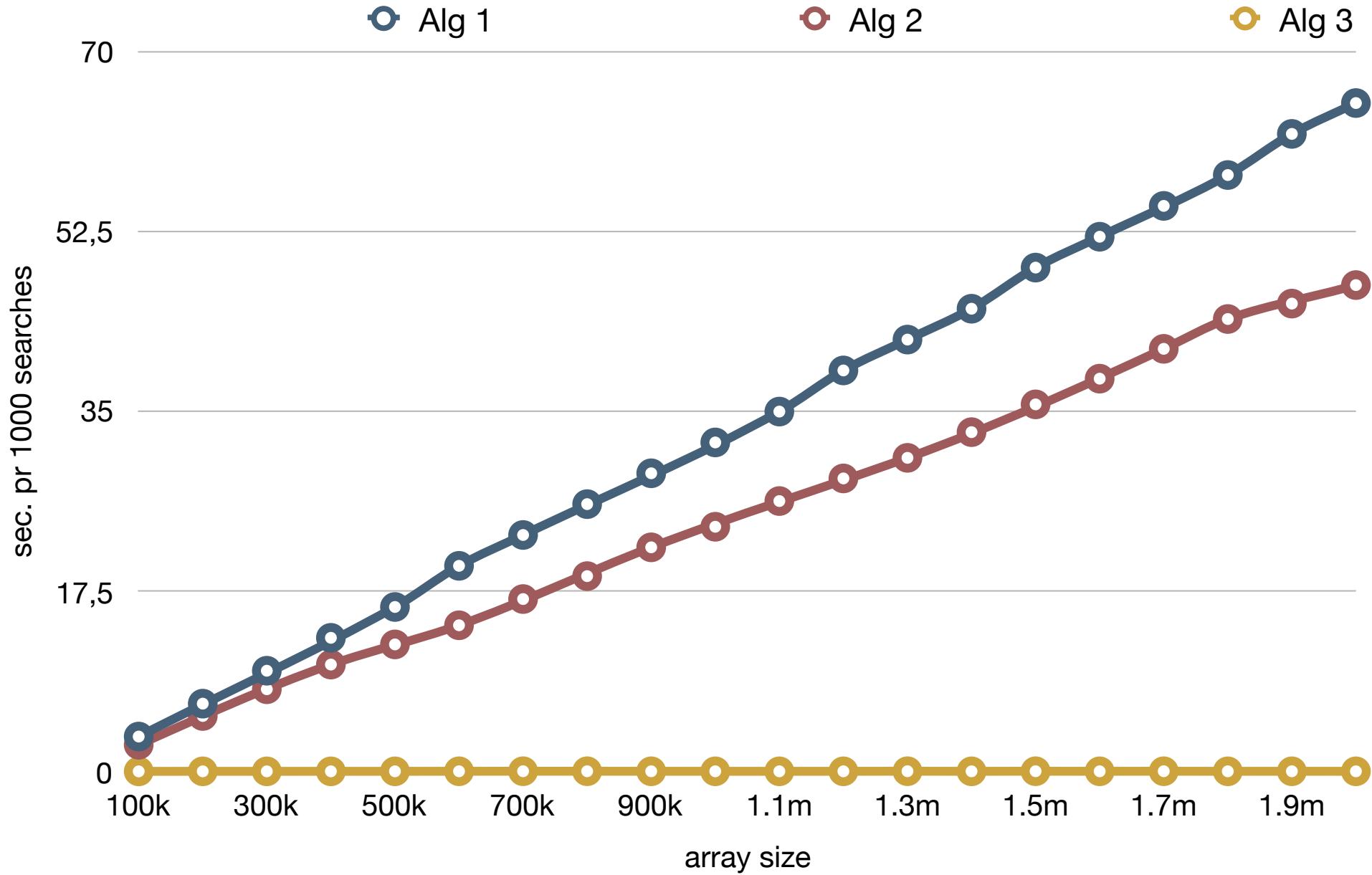
0 1 2 3 4 5 6 7 8 9 10 11 12 13 14

1	3	7	15	17	11	2	3	6	8	7	5	9	5	23
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Algorithm 3

- Running time.
- A recursive call takes constant time.
- How many recursive calls?
- A recursive call halves size of interval. We stop when array has size 1.
 - 1st recursive call: $n/2$
 - 2nd recursive call: $n/4$
 -
 - kth recursive call: $n/2^k$
 -
- \Rightarrow After $\sim \log_2 n$ recursive call array has size ≤ 1 .
- \Rightarrow Running time is $\Theta(\log n)$
- Experimental analysis. Significantly better?

```
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)
```



Peaks

- Theoretical analysis.
 - Algorithm 1 and 2 finds a peak in $\Theta(n)$ time.
 - Algorithm 3 finds a peak in $\Theta(\log n)$ time.
- Experimental analysis.
 - Algorithm 1 and 2 run in $\Theta(n)$ 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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