

# CS 1114:

## Sorting and selection (part one)

**Prof. Graeme Bailey**

<http://cs1114.cs.cornell.edu>

*(notes modified from Noah Snaveley, Spring 2009)*



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Computer Science

## Where we are so far

- Finding the lightstick
  - Attempt 1: Bounding box (not so great)
  - Attempt 2: Centroid isn't much better
  - Attempt 3: Trimmed mean
    - Seems promising
    - But how do we compute it quickly?
    - The obvious way doesn't seem so great...
    - But do we really know this?



# How do we define fast?

- We want to think about this issue in a way that doesn't depend on either:
  - A. Getting really lucky input
  - B. Happening to have really fast hardware



## Recap from last time

- We looked at the “trimmed mean” problem for locating the lightstick
  - Remove 5% of points on all sides, find centroid
- This is a version of a more general problem:
  - Finding the  $k^{\text{th}}$  largest element in an array
  - Also called the “selection” problem
- We considered an algorithm that repeatedly removes the largest element
  - How fast is this algorithm?



# A more general version of trimmed mean

- Given an array of  $n$  numbers, find the  $k^{\text{th}}$  largest number in the array
- Strategy:
  - Remove the biggest number
  - Do this  $k$  times
  - The answer is the last number you found



## How fast is this algorithm?

- An elegant answer exists
- You will learn it in later CS courses
  - But I'm going to steal their thunder and explain the basic idea to you here
  - It's called "big-O notation"
- Two basic principles:
  - Think about the average / worst case
    - Don't depend on luck
  - Think in a hardware-independent way
    - Don't depend on the chip!



# Performance of our algorithm

- What value of  $k$  is the worst case?
  - ~~$k = n$~~  we can easily fix this
  - $k = n/2$
- How much work will we do in the worst case?
  1. Examine  $n$  elements to find the biggest
  2. Examine  $n-1$  elements to find the biggest
  - ... keep going ...
  - $n/2$ . Examine  $n/2$  elements to find the biggest



## How much work is this?

- How many elements will we examine in total?

$$\underbrace{n + (n - 1) + (n - 2) + \dots + n/2}_{n / 2 \text{ terms}}$$

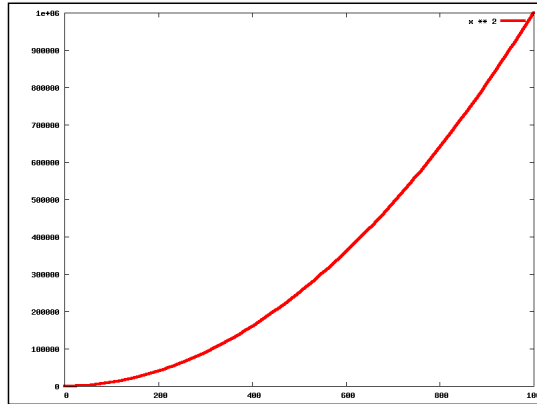
$$= ?$$

- We don't really care about the exact answer
  - It's bigger than  $(n / 2)^2$



# How much work is this?

- The amount of work grows *in proportion* to  $n^2$

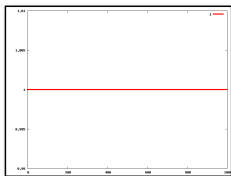


- We say that this algorithm is  $O(n^2)$
- [ Blackboard discussion of  $O(g(n))$  and  $o(g(n))$  ]

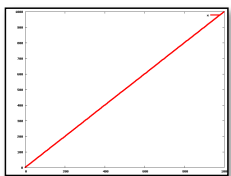


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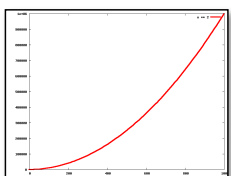
## Classes of algorithm speed



- Constant time algorithms,  $O(1)$ 
  - Do not depend on the input size
  - Example: find the first element



- Linear time algorithms,  $O(n)$ 
  - Constant amount of work for every input item
  - Example: find the largest element



- Quadratic time algorithms,  $O(n^2)$ 
  - Linear amount of work for every input item
  - Example: repeatedly removing max element

*Different hardware only affects the parameters (i.e., line slope)*



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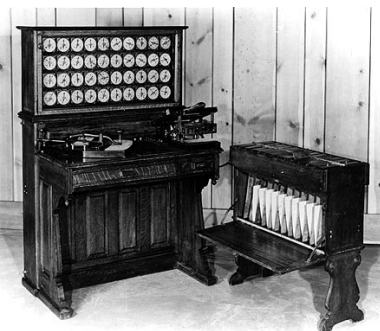
# How to do selection better?

- If our input were sorted, we can do better
  - Given 100 numbers in increasing order, we can easily figure out the 5<sup>th</sup> biggest or smallest
- Very important principle! (encapsulation)
  - Divide your problem into pieces
    - One person (or group) can provide **sort**
    - The other person can use **sort**
  - As long as both agree on what **sort** does, they can work independently
  - Can even “upgrade” to a faster **sort**



## How to sort?

- Sorting is an ancient problem, by the standards of CS
  - First important “computer” sort used for 1890 census, by Hollerith (the 1880 census took 8 years, 1890 took just one)
- There are many sorting algorithms



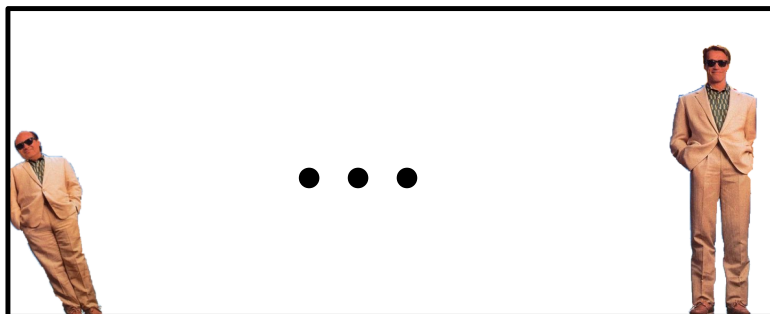
# How to sort?

- Given an array of numbers:  
[10 2 5 30 4 8 19 102 53 3]
- How can we produce a sorted array?  
[2 3 4 5 8 10 19 30 53 102]



# How to sort?

- A concrete version of the problem
  - Suppose I want to sort all actors by height



- How do I do this?



# Sorting, 1<sup>st</sup> attempt

- Idea: Given  $n$  actors
  1. Find the shortest actor (D. Devito), put him first
  2. Find the shortest actor in the remaining group, put him/her second

... Repeat ...

  - n. Find the shortest actor in the remaining group (one left), put him/her last



# Sorting, 1<sup>st</sup> attempt

## Algorithm 1

1. Find the shortest actor put him first
  2. Find the shortest actor in the remaining group, put him/her second
- ... Repeat ...
- n. Find the shortest actor in the remaining group put him/her last

- What does this remind you of?
- This is called *selection sort*
- After round  $k$ , the first  $k$  entries are sorted





# Selection sort – pseudocode

```
function [ A ] = selection_sort(A)
% Returns a sorted version of array A
%   by applying selection sort
%   Uses in place sorting
n = length(A);
for i = 1:n
    % Find the smallest element in A(i:n)
    % Swap that element with something (what?)
end
```



## Filling in the gaps

- `% Find the smallest element in A(i:n)`
- We pretty much know how to do this

```
m = 10000; m_index = -1;
for j in i:n
    if A(j) < m
        m = A(j); m_index = j;
    end
end
```

[ 10 13 41 **6** 51 11 ]  
% After round 1,  
% m = 6, m\_index = 4

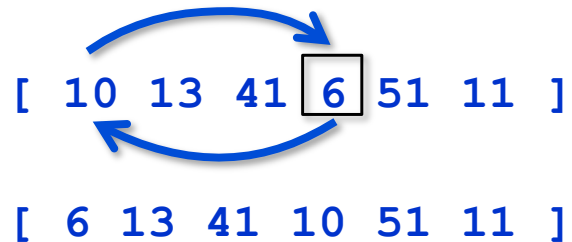


# Filling in the gaps

- % Swap the smallest element with something
- % Swap element A(m\_index) with A(i)

```
A(i) = A(m_index);  
A(m_index) = A(i);
```

```
tmp = A(i);  
A(i) = A(m_index);  
A(m_index) = tmp;
```

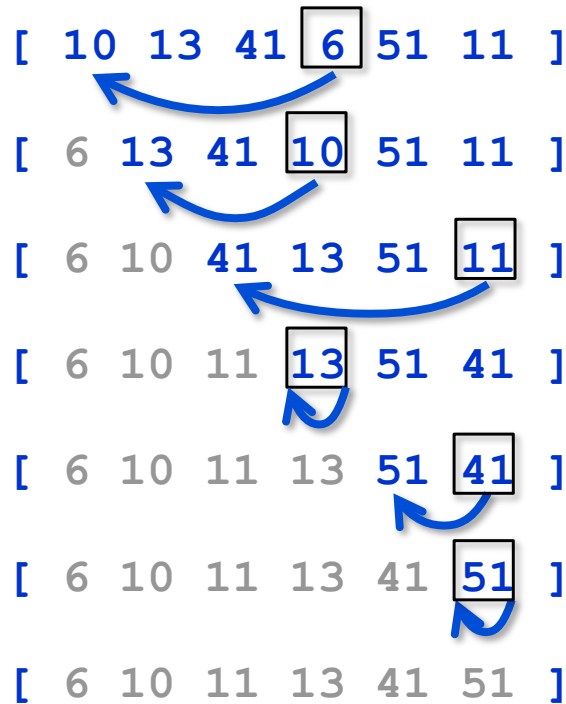


## Putting it all together

```
function [ A ] = selection_sort(A)  
% Returns a sorted version of array A  
len = length(A);  
for i = 1:len  
    % Find the smallest element in A(i:len)  
    m = 10000; m_index = -1;  
    for j in i:n  
        if A(j) < m  
            m = A(j); m_index = j;  
        end  
    end  
    % Swap element A(m_index) with A(i)  
    tmp = A(i);  
    A(i) = A(m_index);  
    A(m_index) = tmp;  
end
```



# Example of selection sort



## Speed of selection sort

- Let  $n$  be the size of the array
- How fast is selection sort?

$O(1)$     $O(n)$     $O(n^2)$    ?

- How many comparisons ( $<$ ) does it do?
- First iteration:  $n$  comparisons
- Second iteration:  $n - 1$  comparisons
- ...
- $n^{\text{th}}$  iteration:  $1$  comparison



# Speed of selection sort

- Total number of comparisons:

$$n + (n - 1) + (n - 2) + \dots + 1$$

$$\sum_{i=1}^n i = \frac{n(n+1)}{2}$$

- Work grows in proportion to  $n^2 \rightarrow$   
selection sort is  $O(n^2)$



## Is this the best we can do?

- Wait and see !!!!
- Don't forget to spend time in the lab getting help on hw 1 – we have a really fun exercise coming up for the next homework 😊

