

Eloise



Abelard

by Edmund Leighton, 1852—1922

Last time: the InsertSort algorithm, on an array of length n , has running time $\leq \frac{1}{2} k_1 n(n-1) + k_2(n-1)$.

Let's make life easier by only worrying about asymptotic costs.

Definition. Given two functions f and g , both $\mathbb{N} \rightarrow \mathbb{R}$, we say $f(n)$ is $O(g(n))$ if
 $\exists \kappa > 0$ and $n_0 \in \mathbb{N}$ such that $\forall n \geq n_0, |f(n)| \leq \kappa |g(n)|$

and we say $f(n)$ is $\Omega(g(n))$ if

$\exists \delta > 0$ and $n_0 \in \mathbb{N}$ such that $\forall n \geq n_0, |f(n)| \geq \delta |g(n)|$.

If $f(n)$ is $O(g(n))$ and also $\Omega(g(n))$ we say that $f(n)$ is $\Theta(g(n))$.

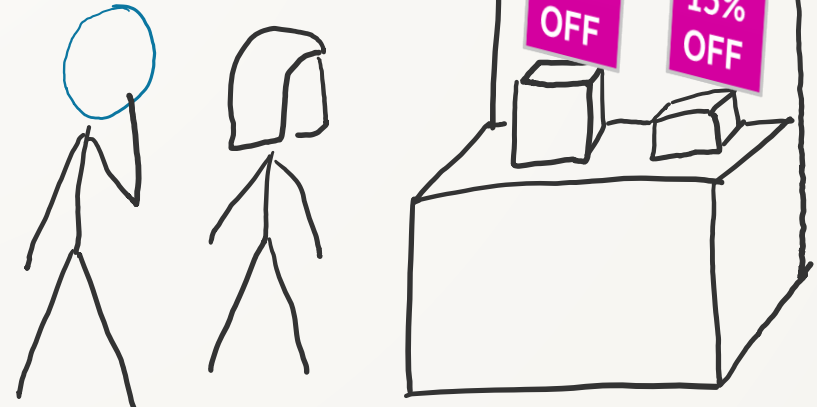
Let $f(n) = \frac{1}{2} k_1 n(n-1) + k_2(n-1)$ k_1, k_2 constants.

Then $f(n)$ is $O(n^3)$ since $f(n) \leq \frac{1}{2} k_1 n^2 + k_2 n = n^3 \left(\frac{\frac{1}{2} k_1}{n} + \frac{k_2}{n^2} \right) \leq 2n^3$ for $n \geq \max(\frac{1}{2} k_1, k_2)$.

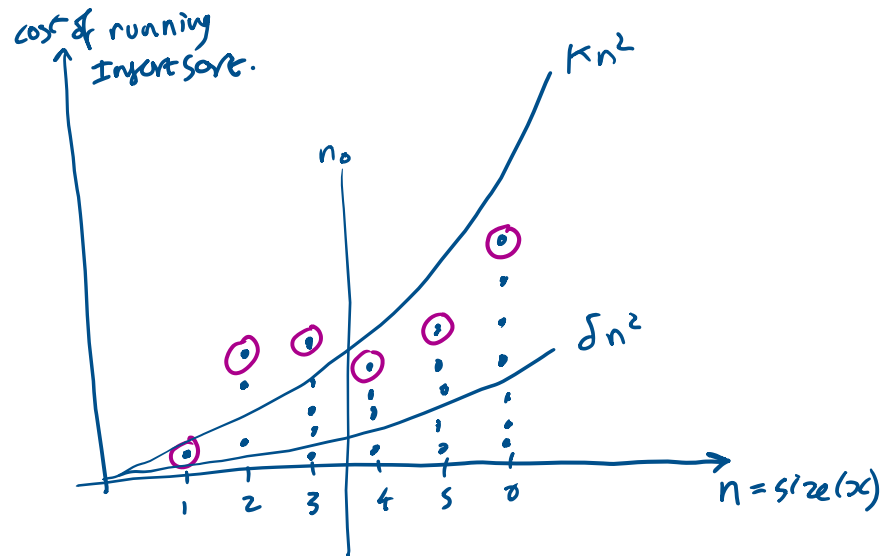
But also $f(n)$ is $O(n^2)$ by similar reasoning. And $O(e^n)$. And....

Also, $f(n)$ is $\Omega(n^2)$, and $\Omega(\log n)$, and $\Omega(1)$...
by similar reasoning.

Since $f(n)$ is $O(n^2)$, and $\Omega(n^2)$, it is $\Theta(n^2)$.



In this course, we're typically interested in an algorithm's worst-case running time as a function of input size.



Plot a dot • for every possible input x .

For each n , circle ⊙ the input that's the worst case.

We've shown that for every input x of size n , the cost is $\leq \kappa n^2$ (for some $\kappa > 0$, and sufficiently large n). In other words, all the blue dots are $\leq \kappa n^2$.

In other words, the purple circles are $\leq \kappa n^2$.

In other words, if we define the worst-case cost to be $h(n) = \max_{x: \text{size}(x)=n} \text{cost}(x)$, then $h(n)$ is $O(n^2)$.

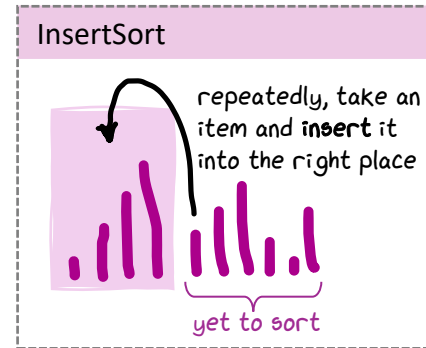
Can we find a matching Ω bound, i.e. show that $h(n)$ is $\Omega(n^2)$?

In other words, can we show that the purple circles are $\geq \delta n^2$ (for some $\delta > 0$, and sufficiently large n)?

In other words, can we find for each n a specific input x whose cost is $\geq \delta n^2$?

In this course, we're typically interested in an algorithm's worst-case running time as a function of input size.

```
1 def insert_sort(x):
2   for i in 1..(len(x)-1):
4     j = i - 1
5     while j >= 0 and x[j] > x[j+1]:
6       swap x[j] with x[j+1]
7     j = j - 1
```

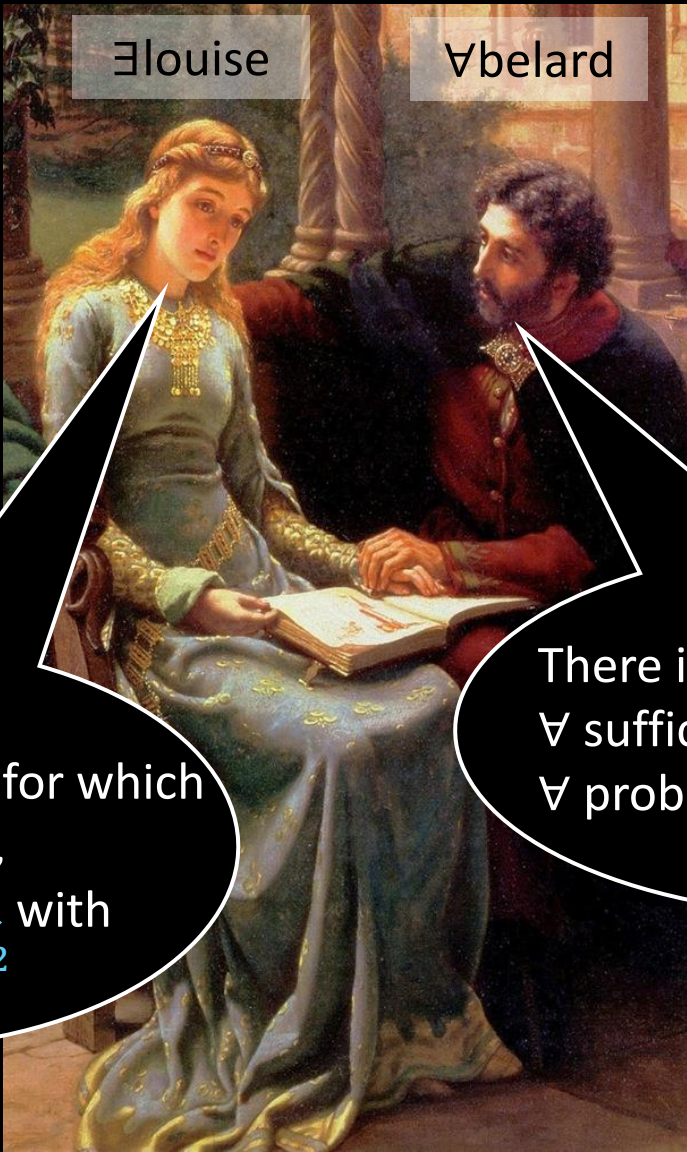


Q. Given an arbitrary n , what is an input of size n that gives the worst possible running time?

For input $[n, n-1, \dots, 1]$
cost is $\Omega(n^2)$

ChatGPT struggles with \exists problems.

For example, see the "vulnerability report" at <https://hackerone.com/reports/2298307>



Elouise

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There is some $\delta > 0$ for which
 \forall sufficiently large n ,
 \exists a problem of size n with
 $\text{cost} \geq \delta n^2$

There is some $\kappa > 0$ for which
 \forall sufficiently large n and
 \forall problems of size n
 $\text{cost} \leq \kappa n^2$

After we show that our algorithm is $O(n^2)$,
it's good manners to also demonstrate that
the worst case is $\Omega(n^2)$.

MON Simple sorting algorithms compared

WED Two optimal algorithms

FRI Better than optimal!?

2.5 Minimum cost of sorting

Can we do better than InsertSort's $\Theta(n^2)$ worst-case running time?

Complexity of Comparison Sort?

- typically count the number of comparisons $C(n)$
- there are $n!$ permutations of n elements
- each comparison eliminates *half* of the permutations
 $2^{C(n)} \geq n!$
- therefore $C(n) \geq \log(n!) \approx n \log n - 1.44n$
- The lower bound of comparison is $O(n \log n)$

ALERT! We don't expect to see "lower bound" and " O " in the same sentence!

Properly-stated theorem

Given any sorting alg. A

let $g_A(x) = \# \text{ comparisons when we run } A \text{ on input } x$

let $f_A(n) = \max_{x: \text{size}(x)=n} g_A(x)$

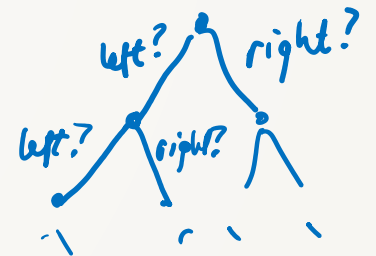
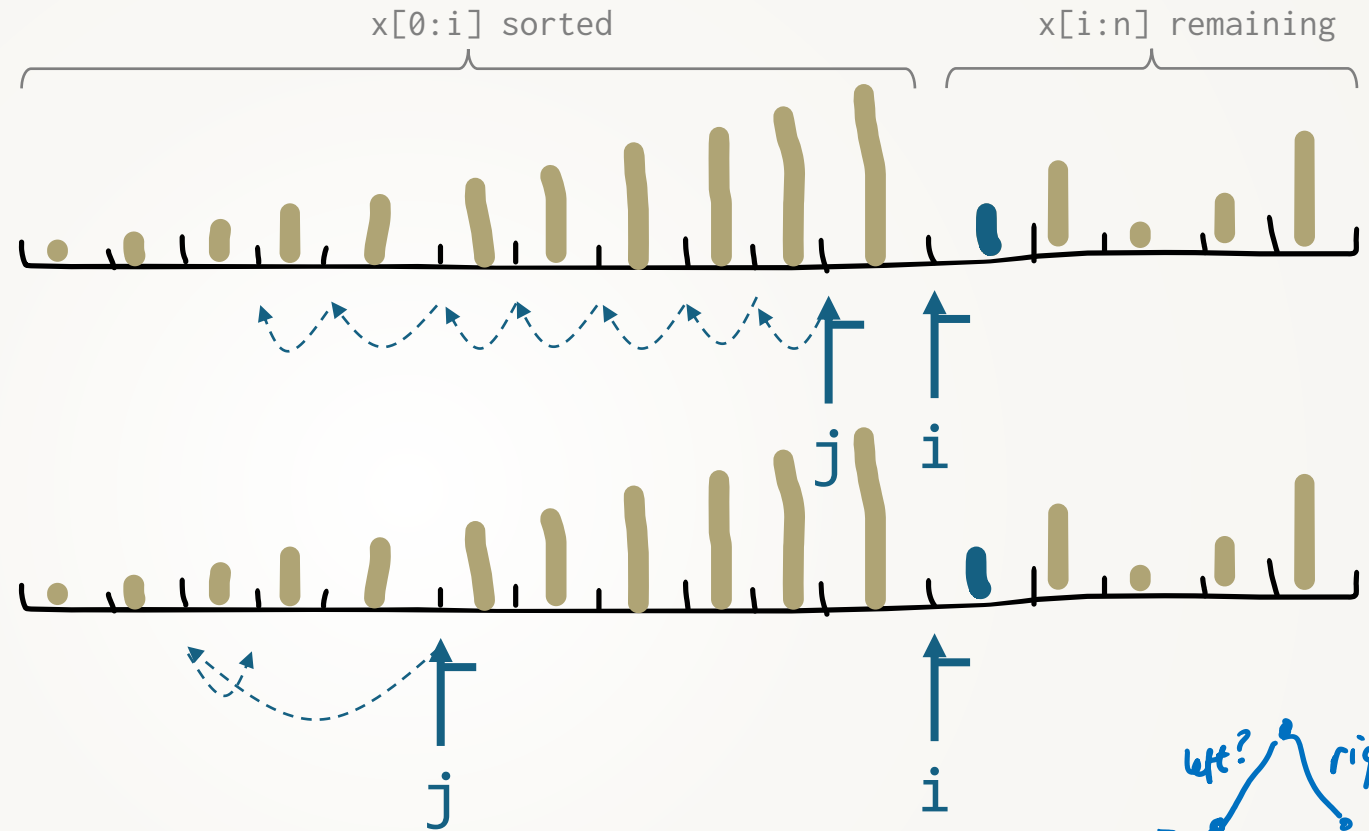
Then $f_A(n)$ is $\Omega(n \log n)$.

§2.7 Binary InsertSort

Can we sort using only $O(n \log n)$ comparisons?

```
def insert_sort(x):  
    for i in 1..(len(x)-1):  
        do a linear search for  
        where x[i] should go, and  
        insert it there
```

```
def binary_insert_sort(x):  
    for i in 1..(len(x)-1):  
        do a binary search for  
        where x[i] should go, and  
        insert it there
```



QUESTION

What's a big-O bound on the number of comparisons for BinaryInsertSort?

$$x \leq \lceil x \rceil < 1 + x$$

$\log n!$ is $\Theta(n \log n)$.

comparisons to place $x[i]$ $\leq \lceil \log_2(i+1) \rceil$

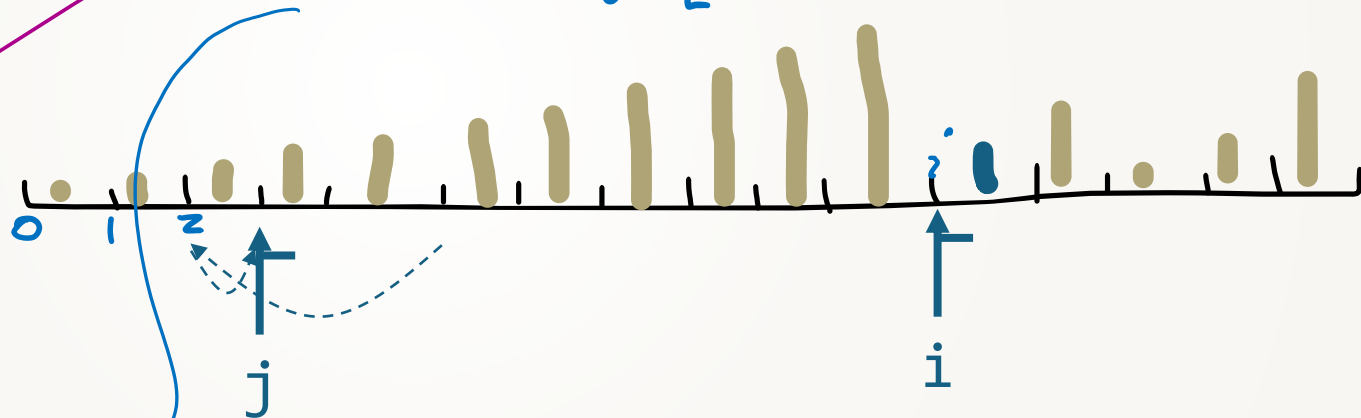
So total # comparisons

$$\leq \sum_{i=1}^{n-1} \lceil \log_2(i+1) \rceil \leq \sum_{i=1}^{n-1} (1 + \log_2(i+1))$$

$$= n-1 + \sum_{i=1}^{n-1} \log_2(i+1)$$

$$= n-1 + \log_2 [n \times (n-1) \times \dots \times 2] = n-1 + \log_2 n!$$

```
def binary_insert_sort(x):
    for i in 1..(len(x)-1):
        do a binary search for
        where x[i] should go, and
        insert it there
```



We used \leq right at the beginning. This "contaminates" all the rest of the working, and means we can only end up with a $O(\cdot)$ conclusion.

$$= n-1 + \frac{\log n!}{\log 2} \leq n-1 + \frac{k}{\log 2} n \log n \quad \text{for some } k \text{ for } n \text{ suff. large}$$

So total # comparisons is $O(n \log n)$.

QUESTION

What's the asymptotic worst-case number of swaps?

Recall: sum of arithmetic series.

$$1 + 2 + \dots + n = \frac{1}{2}n(n+1)$$

- To place $x[i]$ we might need i swaps.

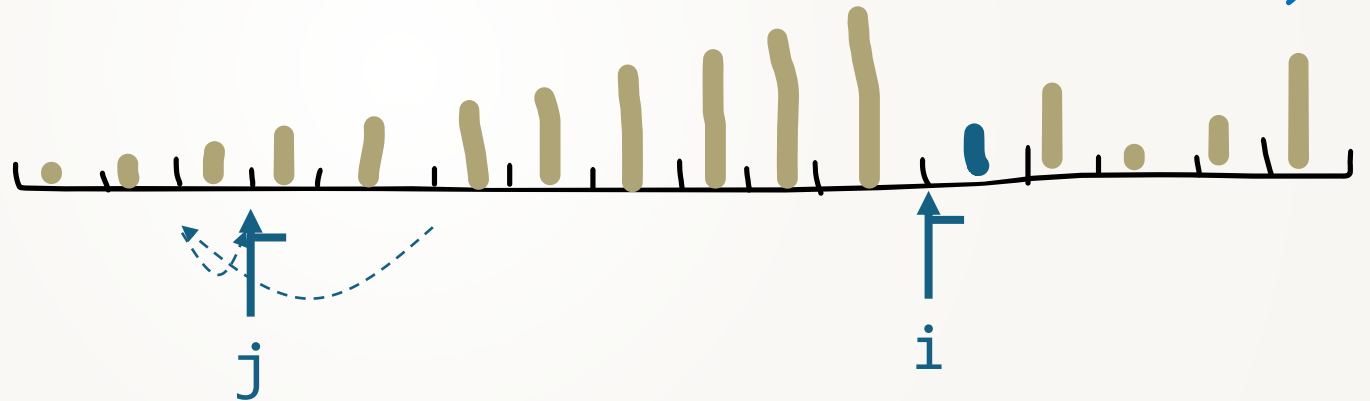
$$\text{Total \#swaps} \leq \sum_{i=1}^{n-1} i$$

so worst-case total #swaps is $O(n^2)$

- Thinking of the input $[n, n-1, \dots, 1]$,

Worst-case total #swaps is $\Omega(n^2)$

```
def binary_insert_sort(x):  
    for i in 1..(len(x)-1):  
        do a binary search for  
        where  $x[i]$  should go, and  
        insert it there
```



§2.6 SelectSort

What's a lower bound for the worst-case number of swaps to sort an array of length n ?

Theorem. For any sorting algorithm, the worst-case number of swaps is $\Omega(n)$.

Proof. Given arbitrary n , consider the input $x = [2, 3, \dots, n, 1]$.

Every item starts in the wrong place, so every item needs to be "touched" by a swap.

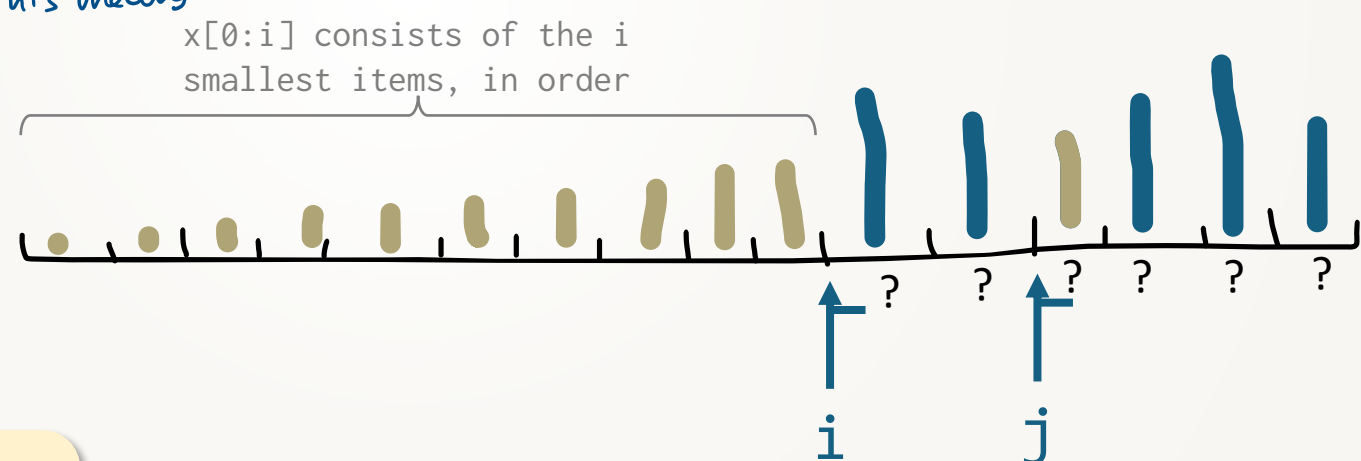
Each swap touches two items.

Thus $\#swaps \geq \lceil n/2 \rceil$, which is $\Omega(n)$.

Can we sort using only $O(n)$ swaps?

Notation: this means "return the k that achieves the minimum"

```
def select_sort(x):
    for i in 0..(len(x)-2):
        # Find what belongs in x[i]
        j = arg min x[k]
            i ≤ k < len(x)
        swap x[i] with x[j]
```



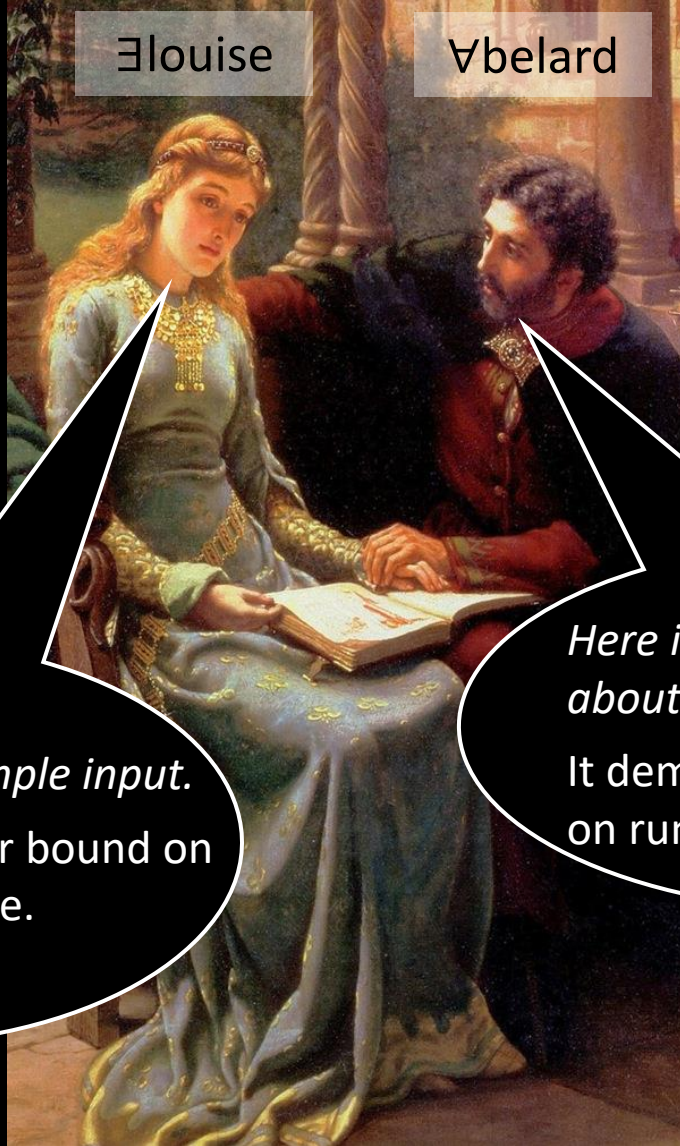
QUESTION
What's the asymptotic worst-case number of comparisons?

Total #comparisons = $\sum_{i=0}^{n-2} (n-i-1) = \Theta(n^2)$.
(for every input)

	comparisons	swaps
any algorithm	worst case is $\Omega(n \log n)$	worst case is $\Omega(n)$
InsertSort	worst case is $O(n^2)$ worst case is $\Omega(n^2)$	} worst case is $O(n^2)$ } worst case is $\Omega(n^2)$
BinaryInsertSort	worst case is $O(n \log n)$	
SelectSort	every case is $\Theta(n^2)$	worst case is $O(n)$

\exists louse

\forall belard



Here is a concrete example input.
It demonstrates a lower bound on worst-case running time.

Here is a universal argument about the worst that can happen.
It demonstrates an upper bound on running time.

If our bounds don't agree, we should think harder!

- Can we find a better example, one that hits our upper bound?
- Or maybe the algorithm isn't as bad as we thought: can we find a tighter upper bound?

As well as O and Ω and Θ , we also use o and ω [see notes]

O is pronounced “big-O”

o is pronounced “little-o”

Ω is pronounced “big-Omega”

ω is pronounced “little-omega”

literally means big o!

