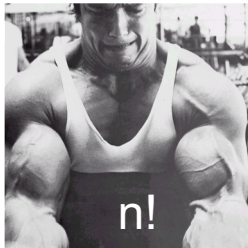
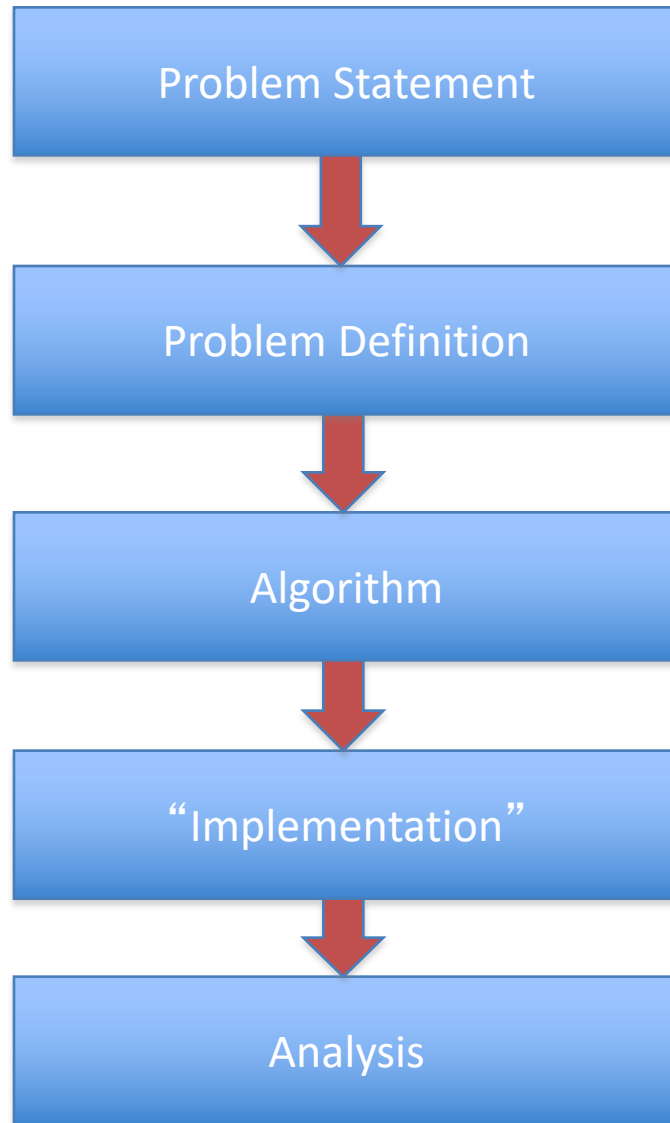


Lecture 9

CSE 331

Feb 19, 2021

Main Steps in Algorithm Design



Correctness Analysis

Definition of Efficiency

An algorithm is efficient if, when implemented, it runs quickly on real instances

Implemented where?



What are real instances?

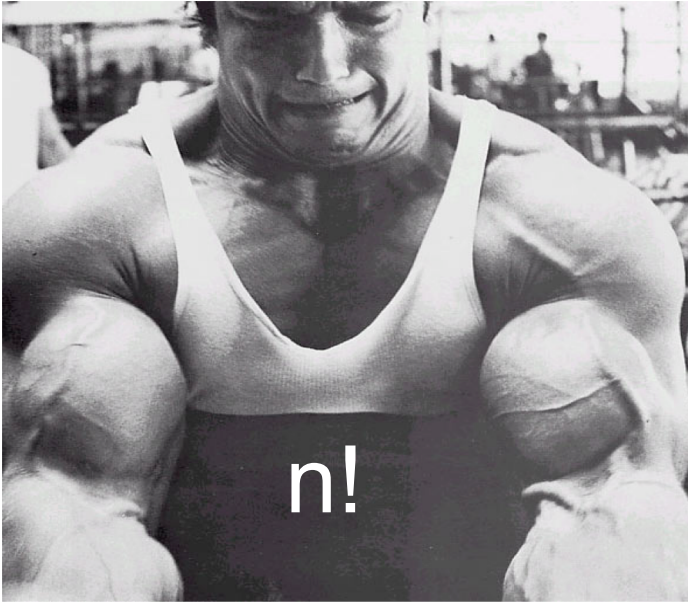
Worst-case Inputs

Efficient in terms of what?

$$N = 2n^2 \text{ for SMP}$$

Input size N

Definition-II



Analytically better than brute force

How much better? By a factor of 2?

Definition-III

Should scale with input size

If N increases by a constant factor,
so should the measure



Polynomial running time

At most $c \cdot N^d$ steps ($c > 0$, $d > 0$ absolute constants)

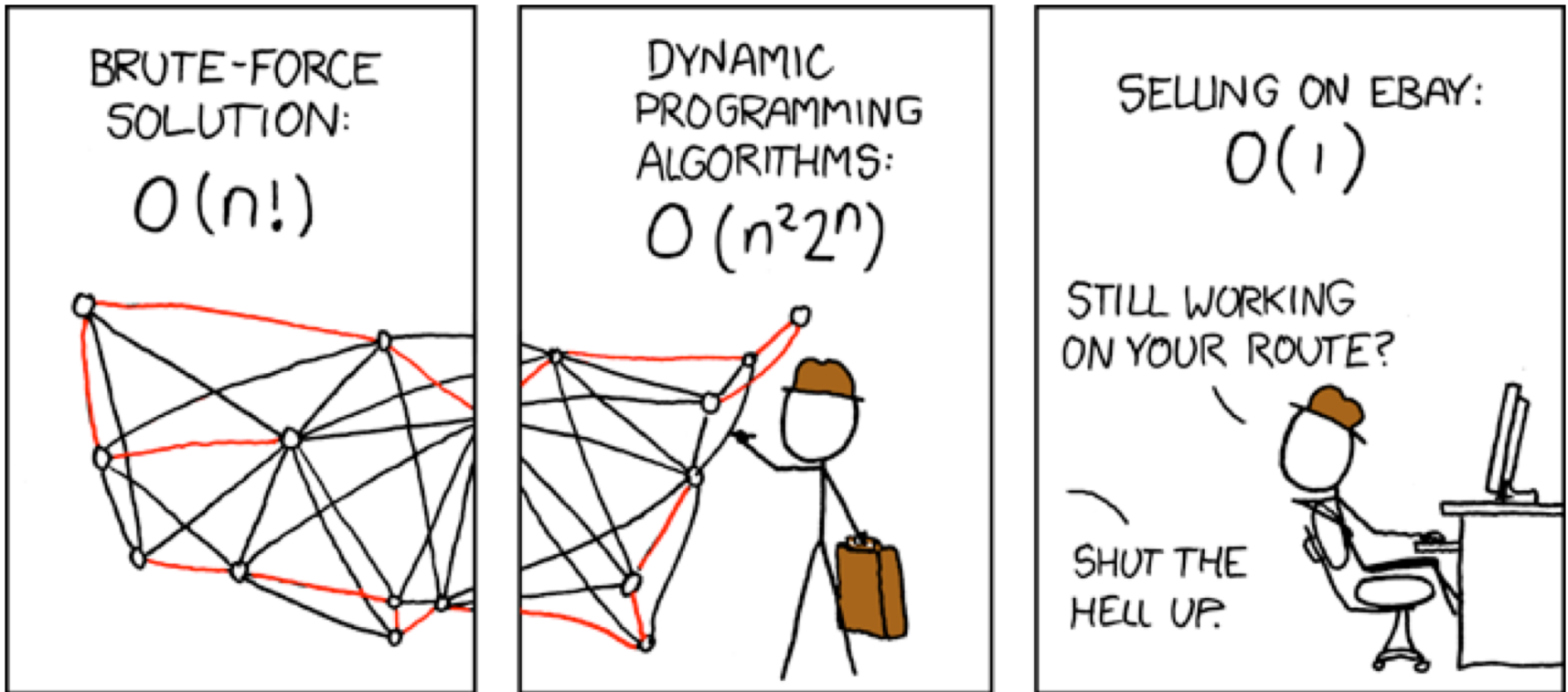
Step: “primitive computational step”

More on polynomial time

Problem centric tractability

Can talk about problems that are not efficient!

Asymptotic Analysis

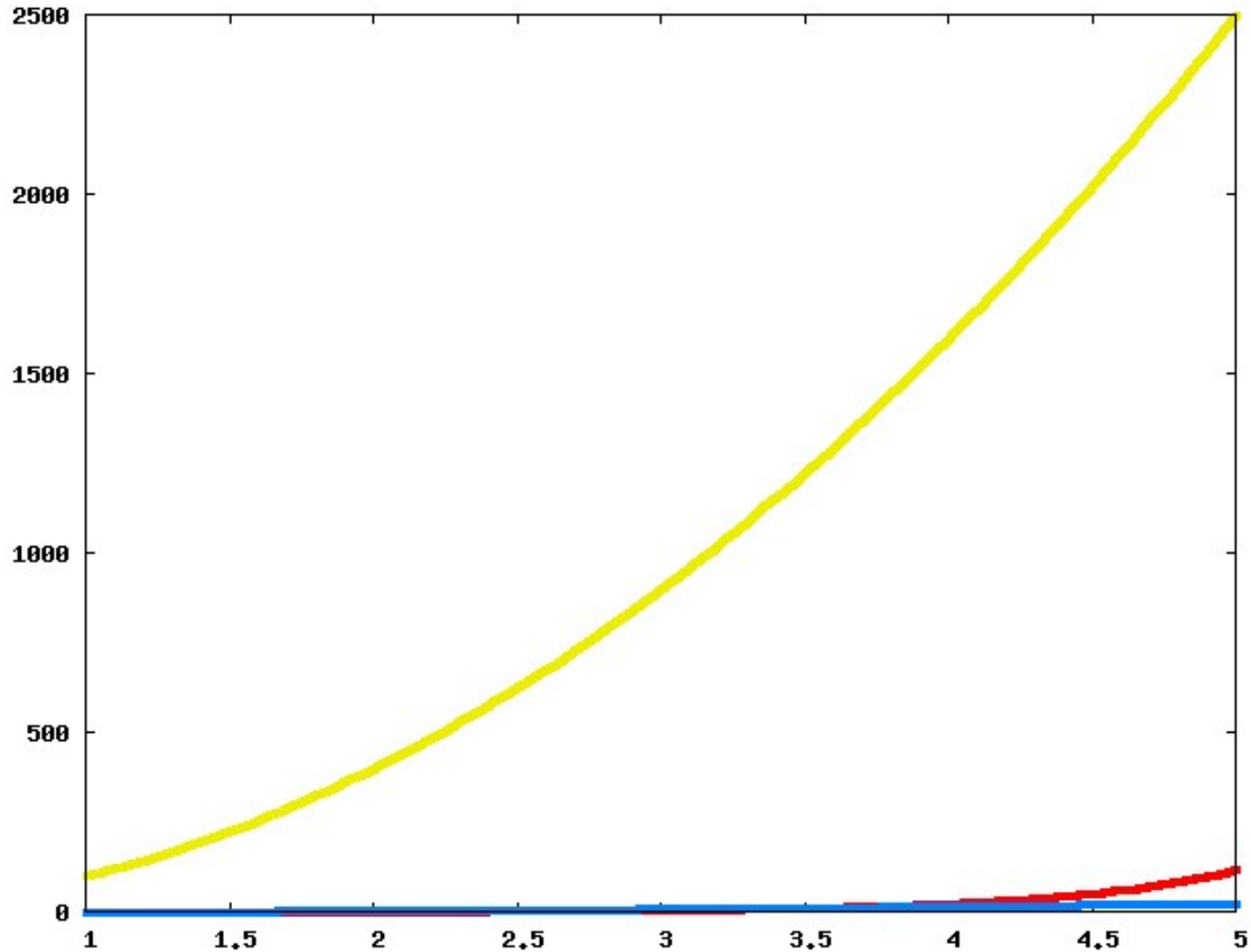


Travelling Salesman Problem

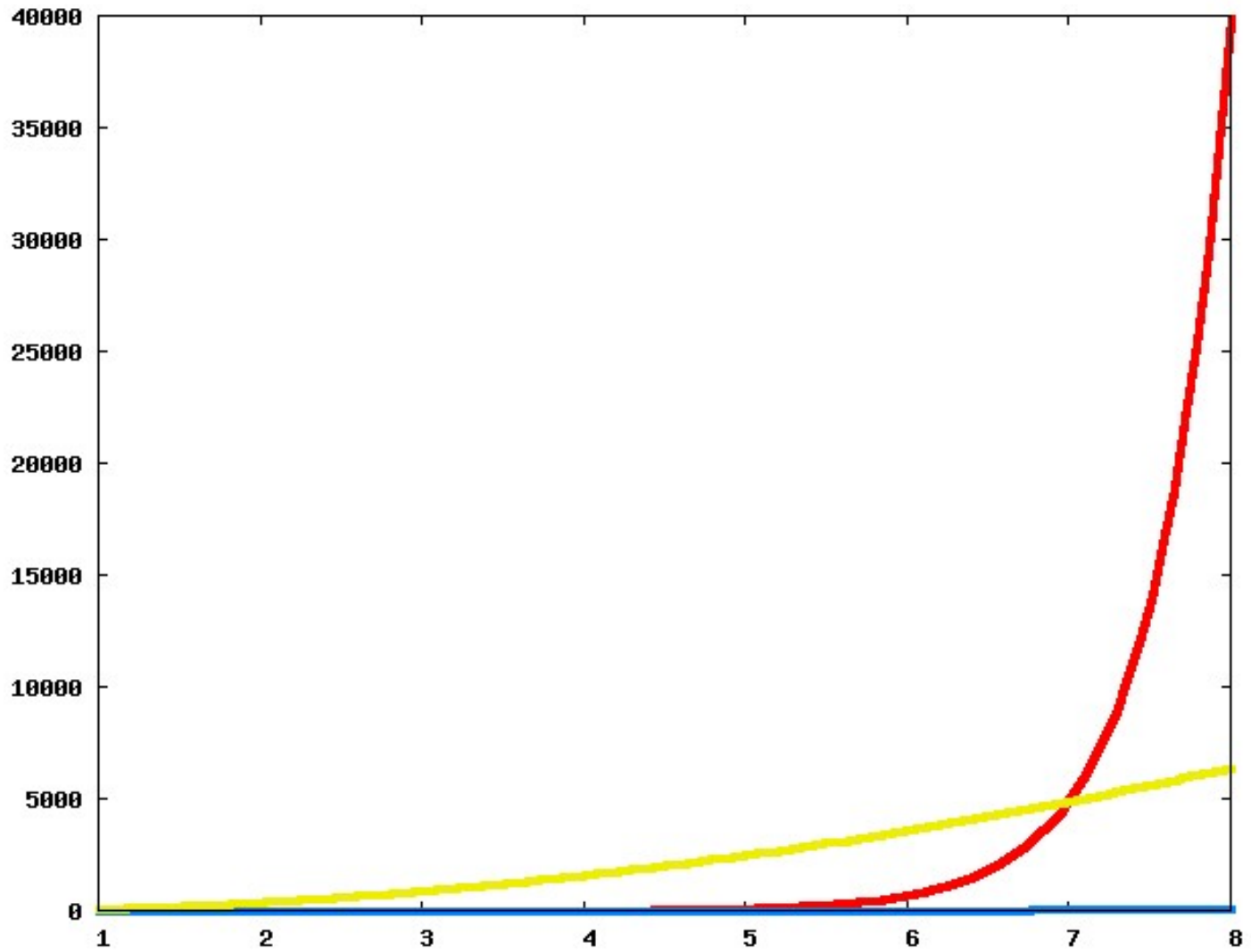
(<http://xkcd.com/399/>)

See the reading assignment!

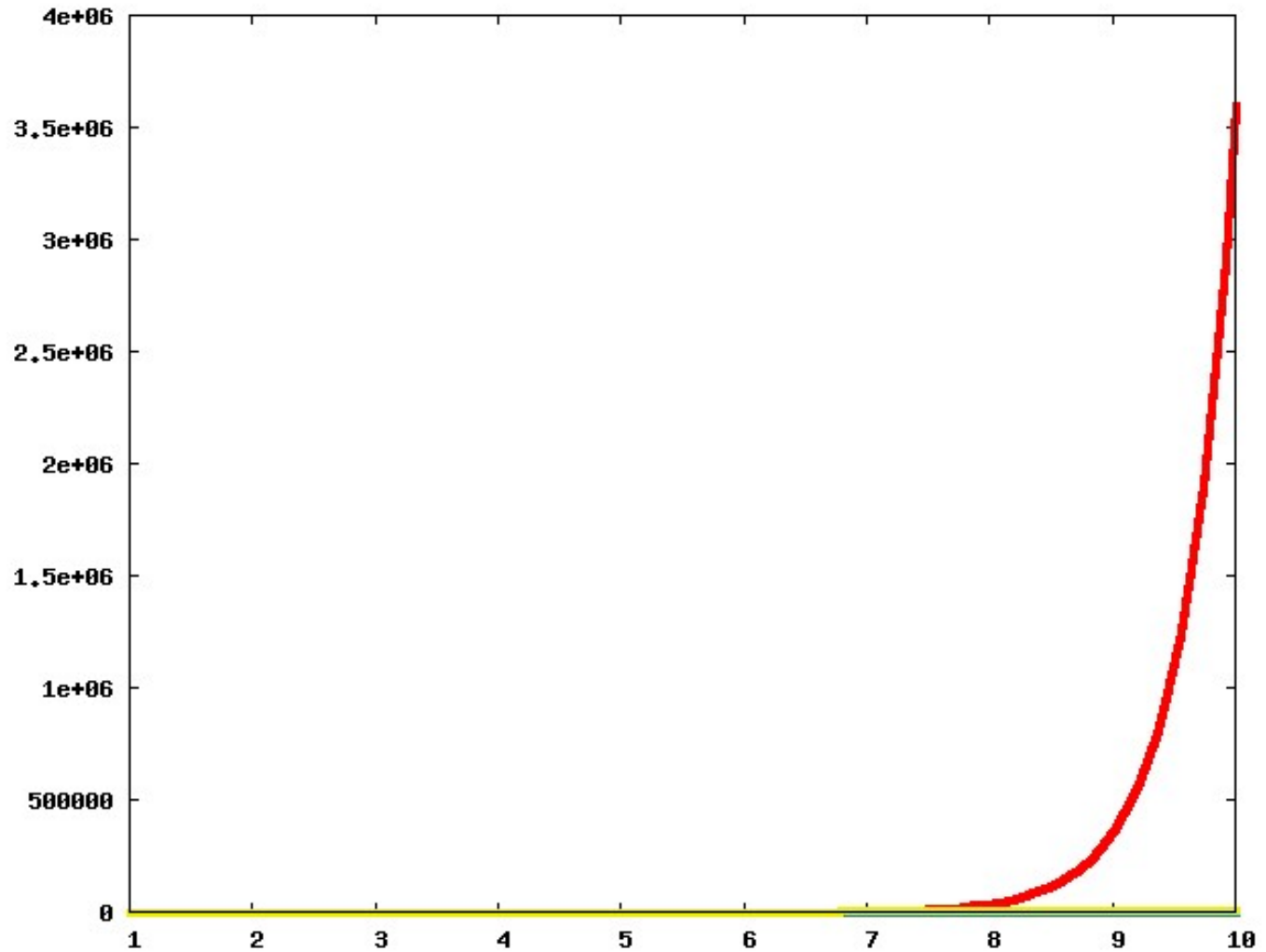
Which one is better?



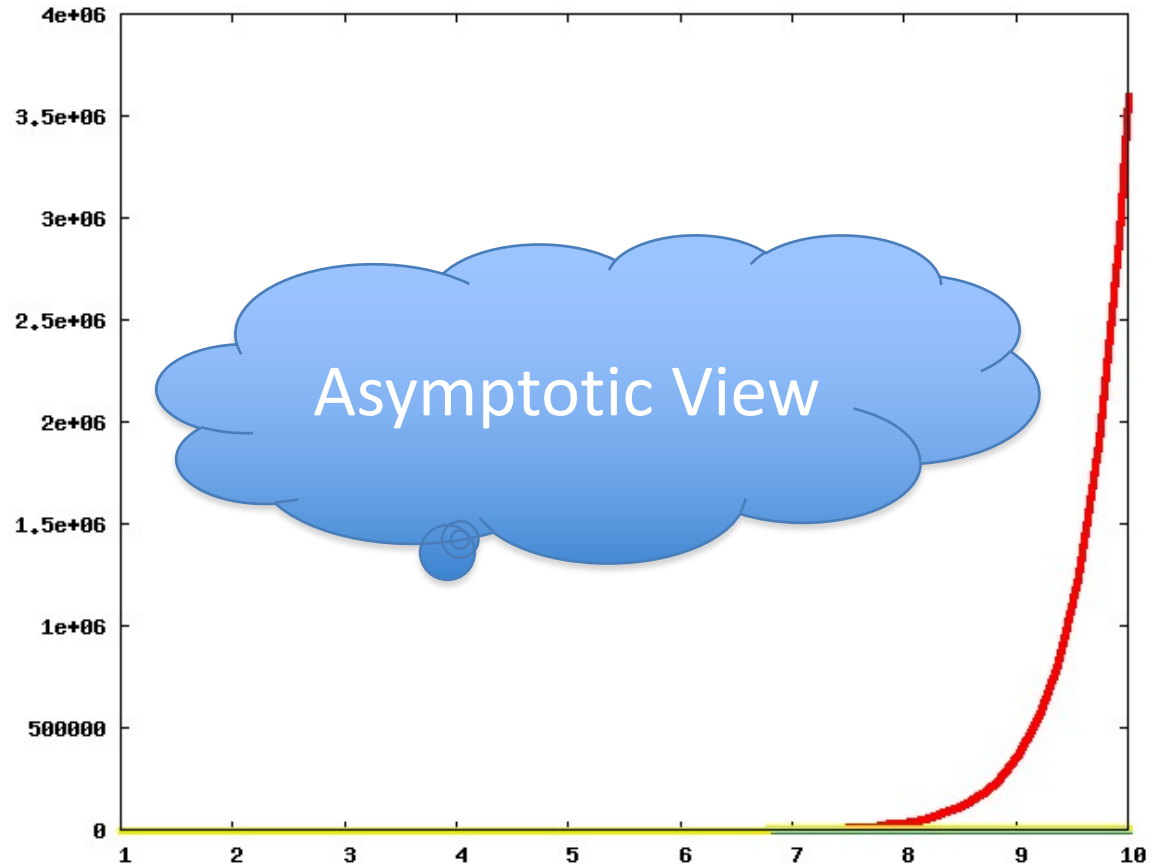
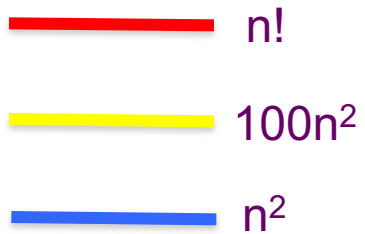
Now?



And now?



The actual run times



Asymptotic Notation



\leq is O with glasses

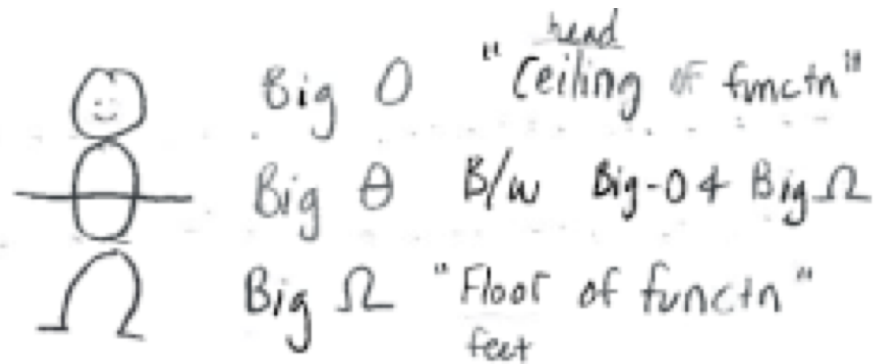
\geq is Ω with glasses

$=$ is Θ with glasses

Another view

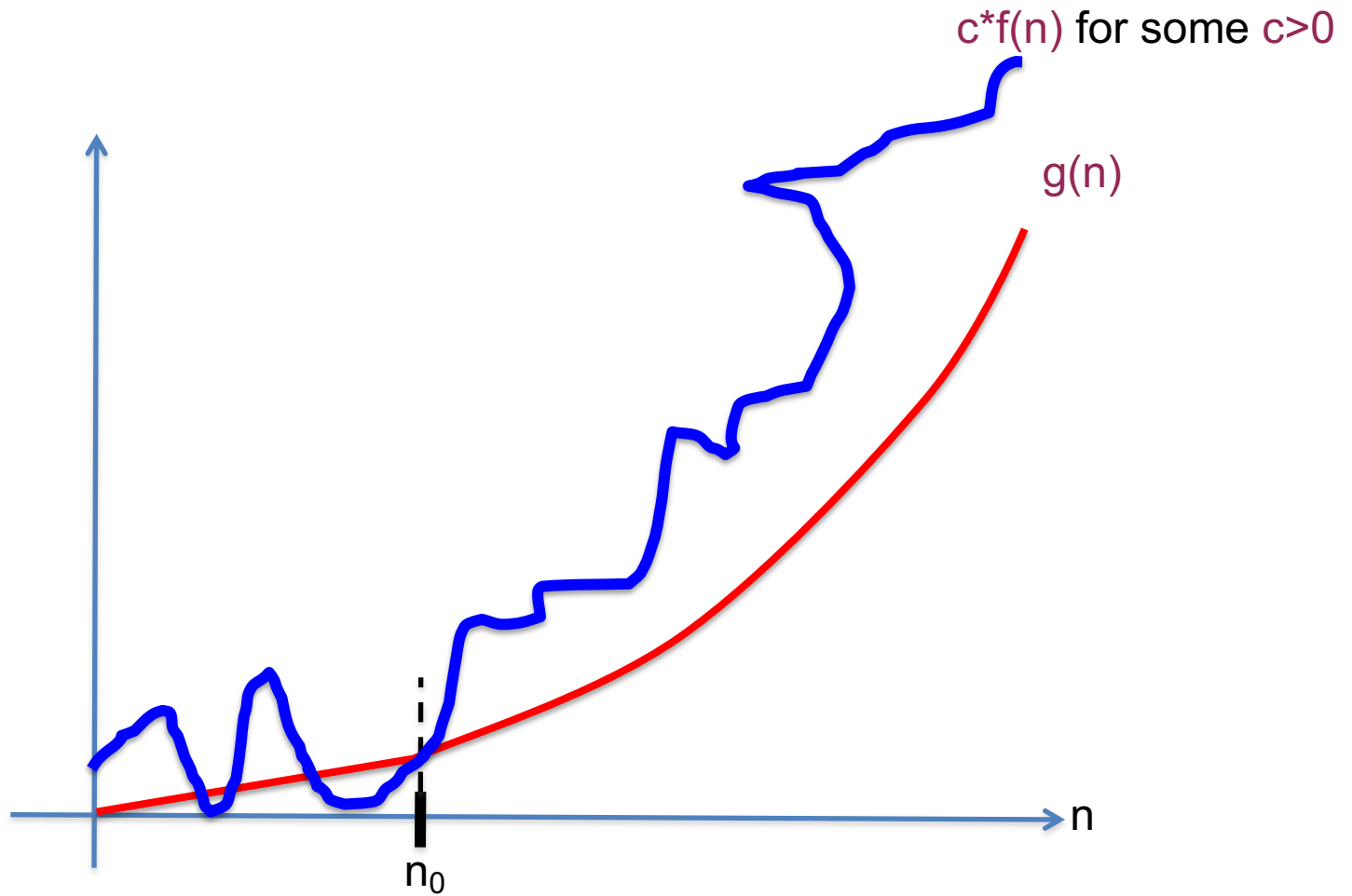
remain anonymous on the web, let me know).

Silly way to remember
asymptotic notation....
Stick figure:

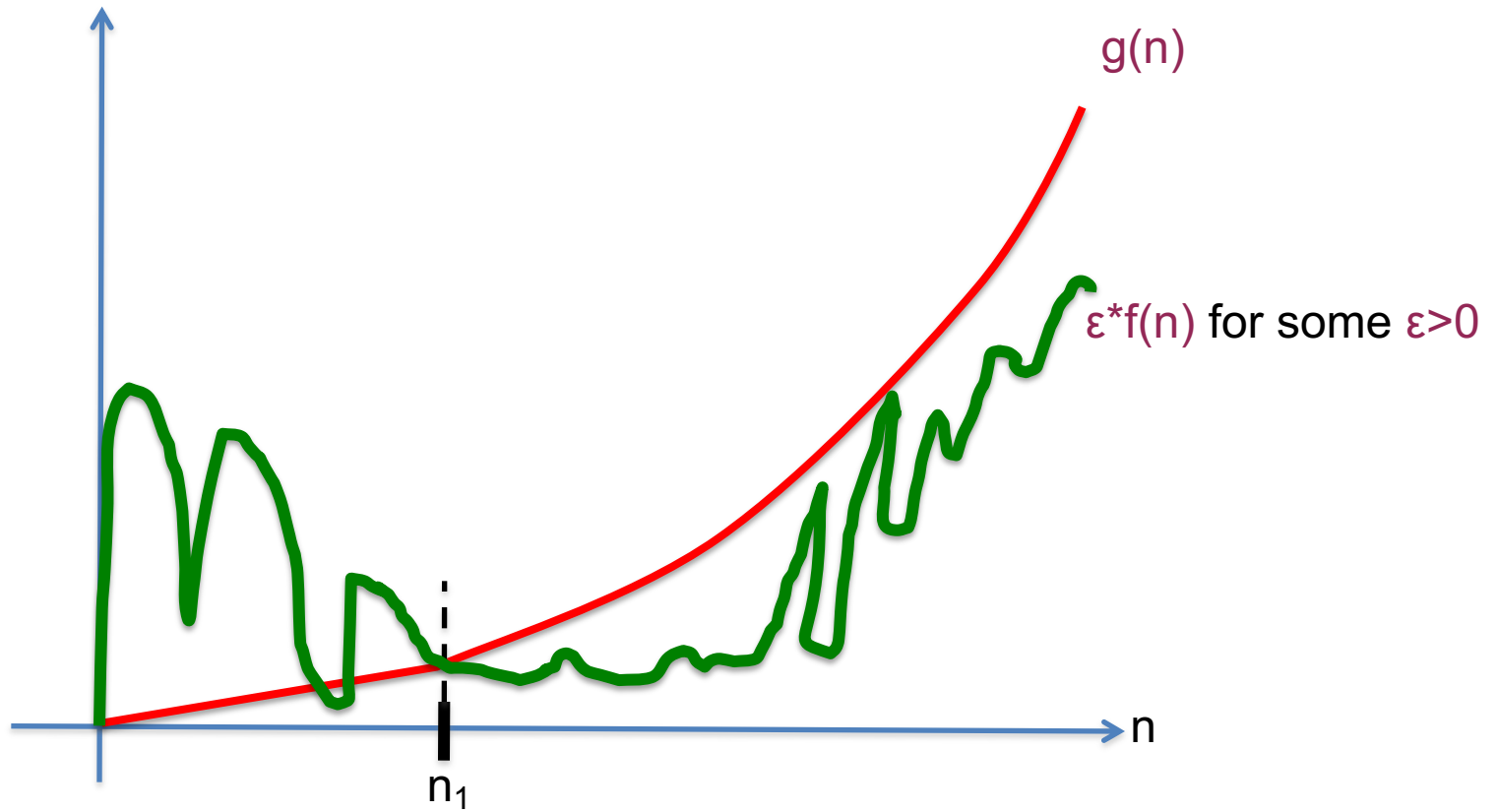


© Aleksandra Patrzalek, 2012

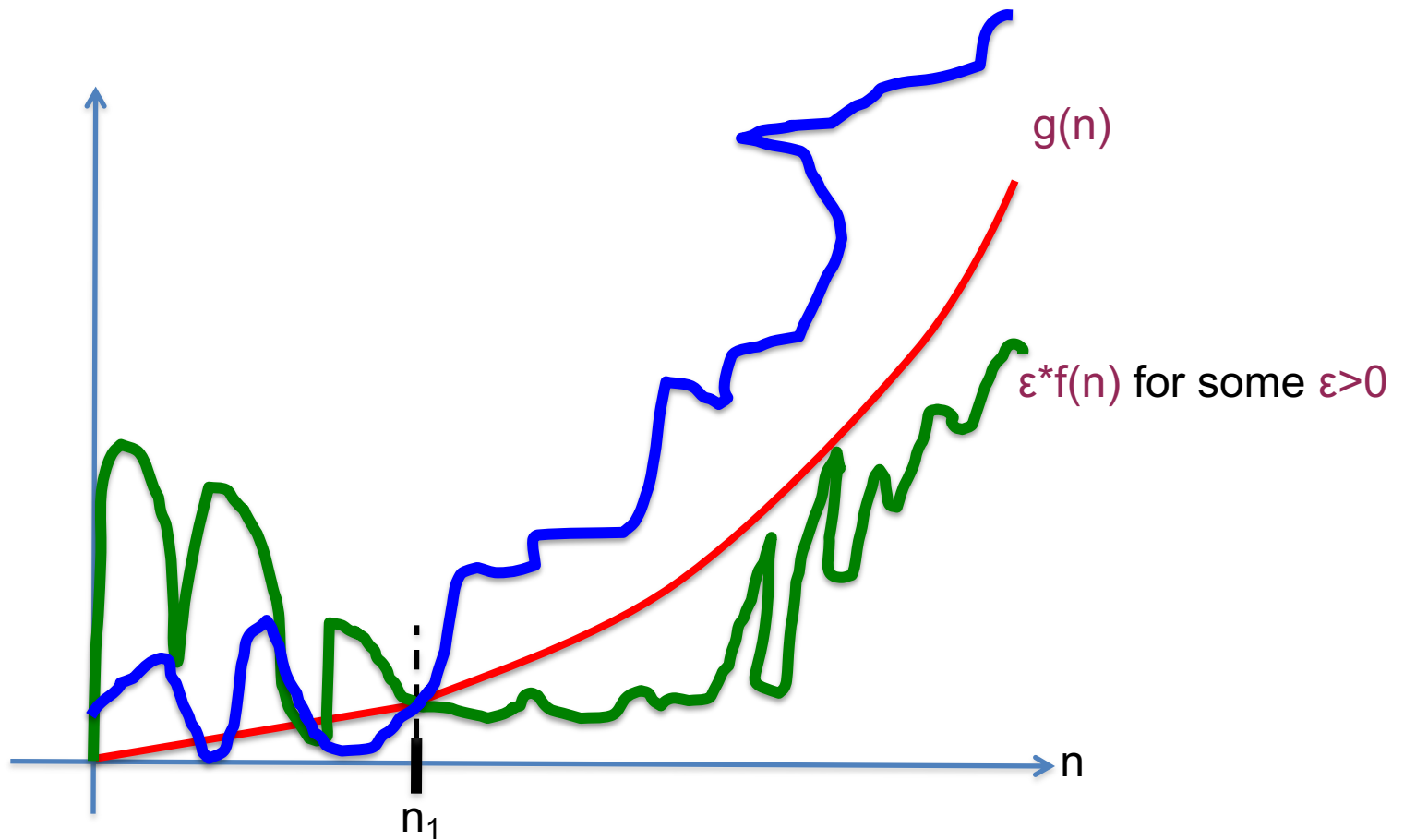
$g(n)$ is $O(f(n))$



$g(n)$ is $\Omega(f(n))$



$g(n)$ is $\Theta(f(n))$



Properties of O (and Ω)

Transitive

g is $O(f)$ and f is $O(h)$ then
 g is $O(h)$

```
Step 1 // O(n) time  
Step 2 // O(n) time
```

Additive

g is $O(h)$ and f is $O(h)$ then
 $g+f$ is $O(h)$

Overall:
 $O(n)$ time

Multiplicative

g is $O(h_1)$ and f is $O(h_2)$ then
 $g*f$ is $O(h_1*h_2)$

Overall:
 $O(n^2)$ time

```
While (loop condition) //  $O(n^2)$  iterations  
  Stuff happens //  $O(1)$  time
```

Another Reading Assignment

Analyzing the worst-case runtime of an algorithm

Some notes on strategies to prove Big-Oh and Big-Omega bounds on runtime of an algorithm.

The setup

Let \mathcal{A} be the algorithm we are trying to analyze. Then we will define $T(N)$ to be the worst-case run-time of \mathcal{A} over all inputs of size N . Slightly more formally, let $t_{\mathcal{A}}(\mathbf{x})$ be the number of steps taken by the algorithm \mathcal{A} on input \mathbf{x} . Then

$$T(N) = \max_{\mathbf{x}: \mathbf{x} \text{ is of size } N} t_{\mathcal{A}}(\mathbf{x}).$$

In this note, we present two useful strategies to prove statements like $T(N)$ is $O(g(N))$ or $T(N)$ is $\Omega(h(N))$. Then we will analyze the run time of a very simple algorithm.

Preliminaries

We now collect two properties of asymptotic notation that we will need in this note (we saw these in class today).

Reading Assignments

Sections 1.2, 2.1, 2.2 and 2.4 in [KT]

Questions?

Today's agenda

$O(n^2)$ implementation of the Gale-Shapley algorithm

More practice with run time analysis



Gale-Shapley Algorithm

Initially all men and women are **free**

At most n^2 iterations

While there exists a free woman who can propose

Let w be such a woman and m be the best man she has not proposed to

w proposes to m

If m is free

(m,w) get **engaged**

Else (m,w') are engaged

If m prefers w' to w

w remains **free**

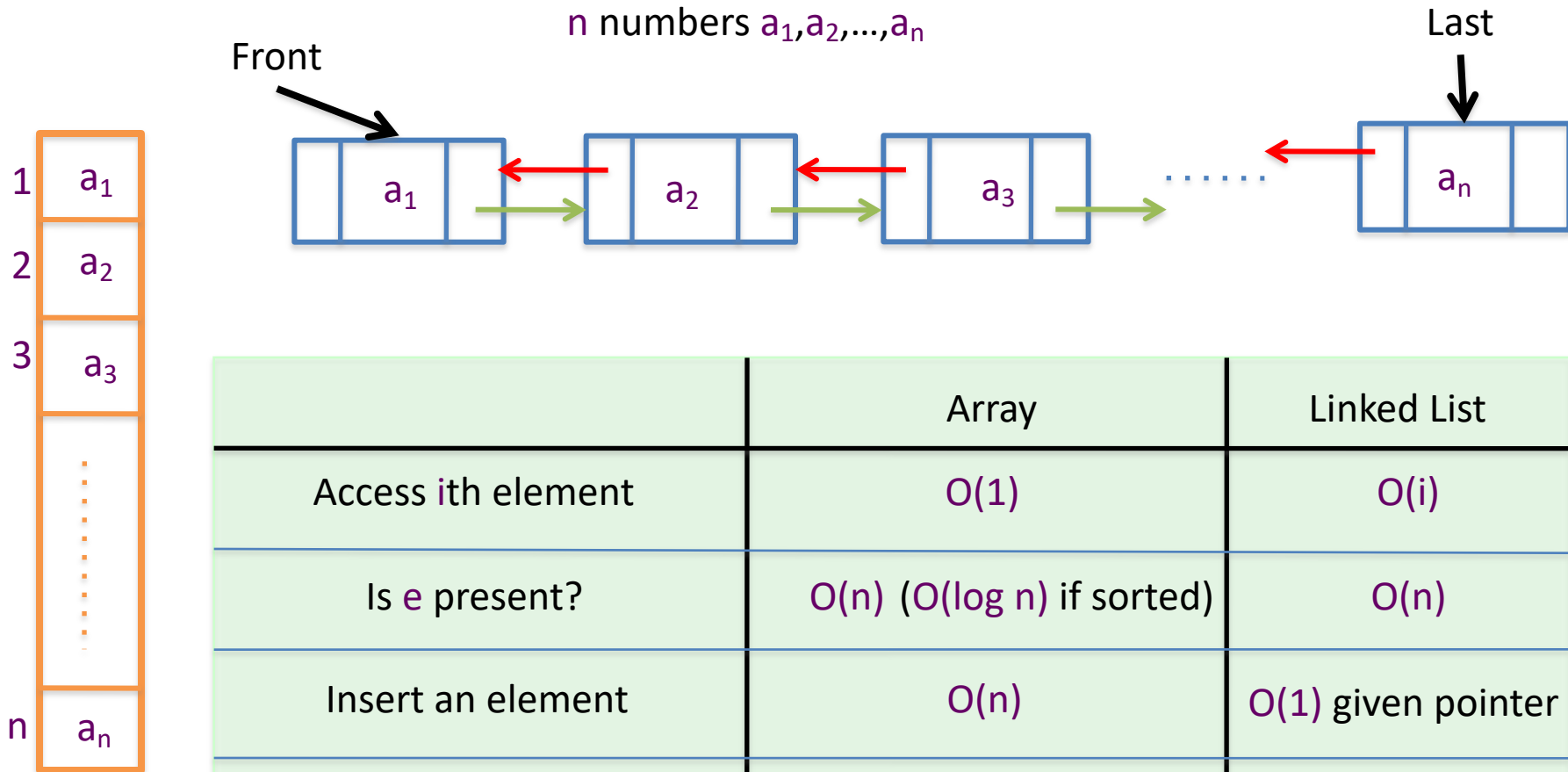
Else

(m,w) get **engaged** and w' is **free**

$O(1)$ time
implementation

Output the engaged pairs as the final output

Arrays and Linked Lists



	Array	Linked List
Access i th element	$O(1)$	$O(i)$
Is e present?	$O(n)$ ($O(\log n)$ if sorted)	$O(n)$
Insert an element	$O(n)$	$O(1)$ given pointer
Delete an element	$O(n)$	$O(1)$ given pointer
Static vs Dynamic	Static	Dynamic

Implementation Steps

(0) How to represent the input?

(1) How do we find a free woman w ?

(2) How would w pick her best unproposed man m ?

(3) How do we know who m is engaged to?

(4) How do we decide if m prefers w' to w ?

Overall running time

Init(1-4)



n^2 X (Query/Update(1-4))