#### Lecture 9

**CSE 331** 

#### Please have a face mask on

#### Masking requirement



<u>UB\_requires</u> all students, employees and visitors – regardless of their vaccination status – to wear face coverings while inside campus buildings.

https://www.buffalo.edu/coronavirus/health-and-safety/health-safety-guidelines.html

#### Proof Details of Lemma 1

### Gale Shapley algorithm terminates

This page collects material from Fall 17 incarnation of CSE 331, where we proof details for the claim that the Gale-Shapley algorithm terminates in  $O(n^2)$  iterations.

#### Where does the textbook talk about this?

Section 1.1 in the textbook has the argument (though not in as much detail as below).

#### Fall 2017 material

Here is the lecture video (it starts from the part where we d the proof details):



# Proof by contradiction

Assume the negation of what you want to prove

After some reasoning



Source: 4simpsons.wordpress.com

#### Two obervations

Obs 1: Once m is engaged he keeps getting engaged to "better" women

Obs 2: If w proposes to m' first and then to m (or never proposes to m) then she prefers m' to m

#### Proof of Lemma 3

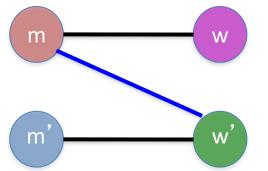
By contradiction

Assume there is an instability (m,w')

w' last proposed to m'

m prefers w' to w

w' prefers m to m'



# Contradiction by Case Analysis

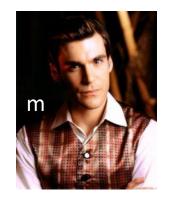
Depending on whether w' had proposed to m or not

Case 1: w' never proposed to m

w' prefers m' to m

By Obs 2

Assumed w' prefers m to m'











Source: 4simpsons.wordpress.com

## Case 2: w' had proposed to m

#### Case 2.1: m had accepted w' proposal

m is finally engaged to w

Thus, m prefers w to w'











#### Case 2.2: m had rejected w' proposal

m was engaged to w" (prefers w" to w')

By Algo def

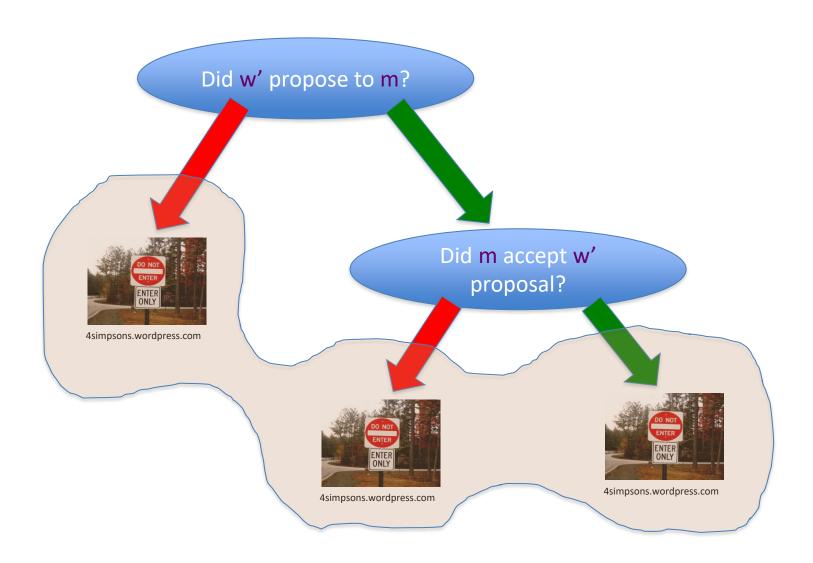
m is finally engaged to w (prefers w to w")

By Obs 1

m prefers w to w'



# Overall structure of case analysis



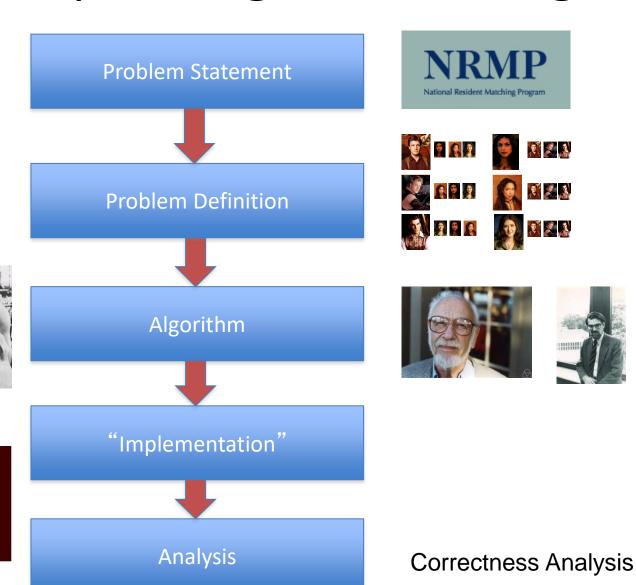
# Questions?

#### **Extensions**

Fairness of the GS algorithm

Different executions of the GS algorithm

## Main Steps in Algorithm Design



# Definition of Efficiency

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

Implemented where?



What are real instances?

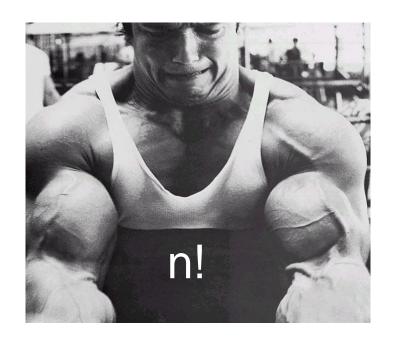
Worst-case Inputs

 $N = 2n^2$  for SMP

Efficient in terms of what?

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·N<sup>d</sup> steps (c>0, d>0 absolute constants)

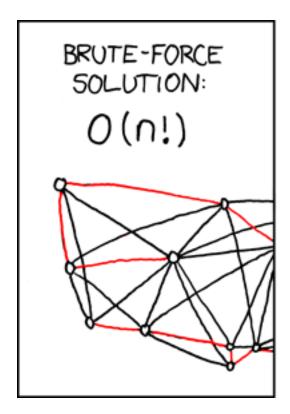
Step: "primitive computational step"

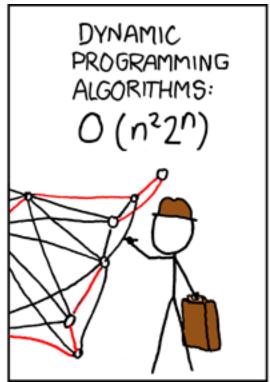
## More on polynomial time

#### Problem centric tractability

Can talk about problems that are not efficient!

#### **Asymptotic Analysis**



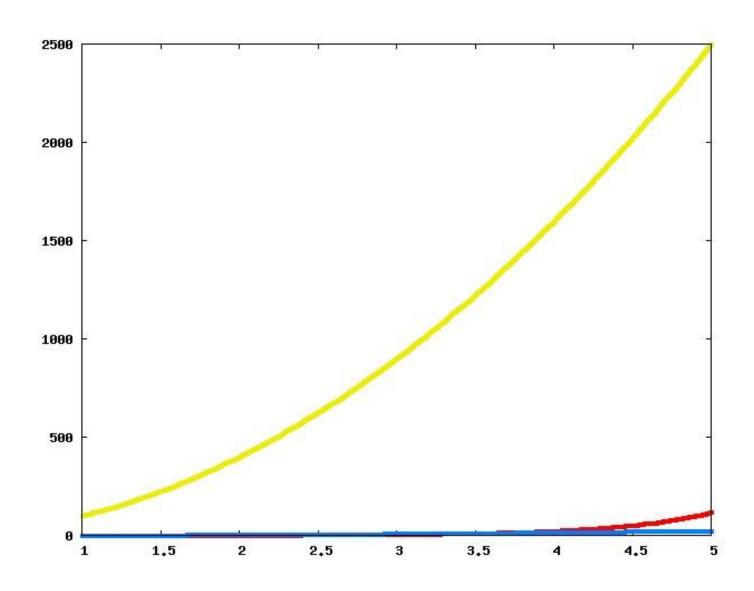




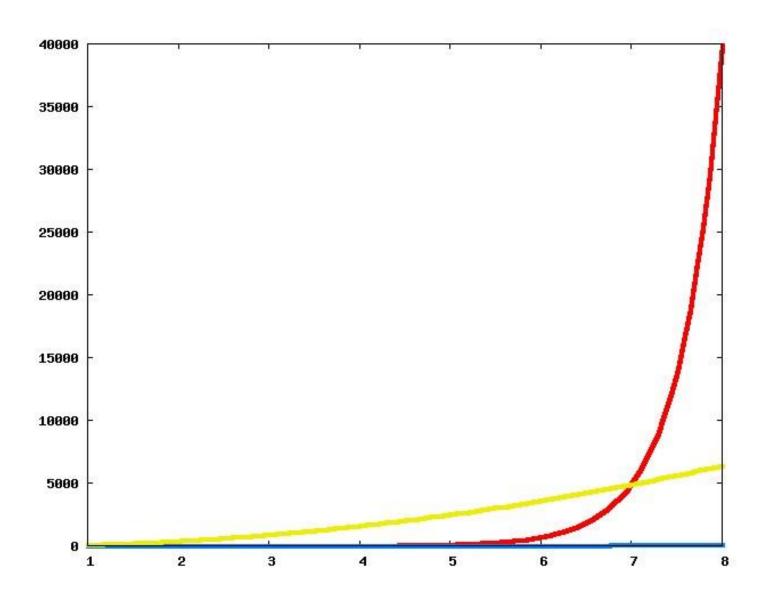
Travelling Salesman Problem

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

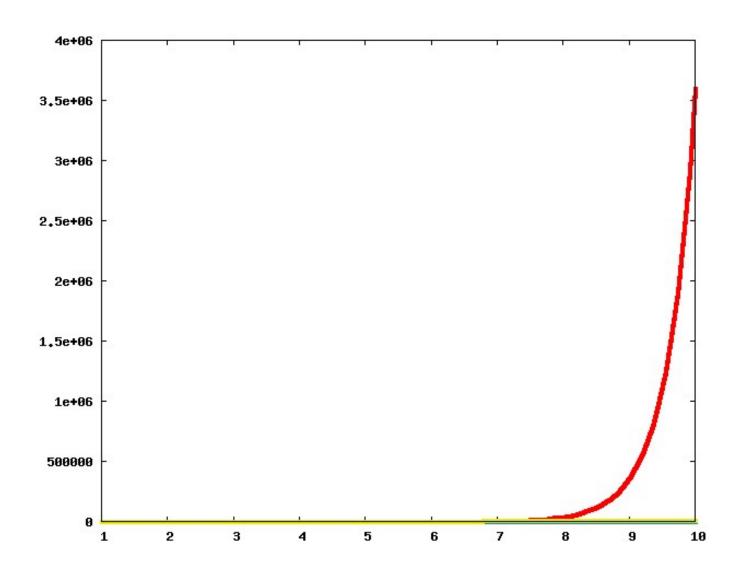
#### Which one is better?



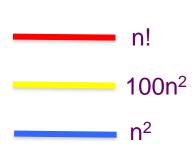
### Now?

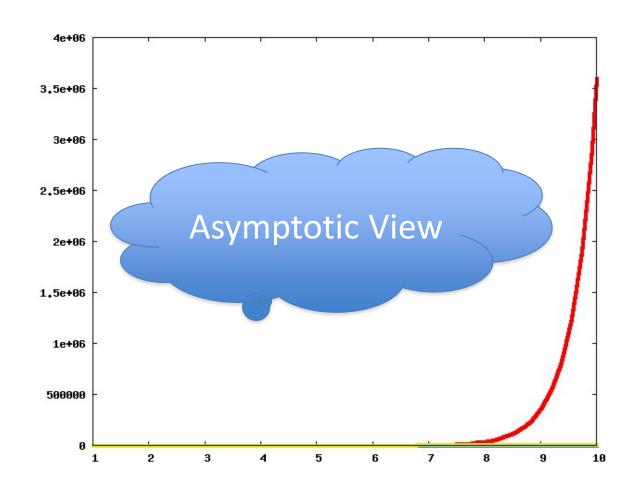


#### And now?

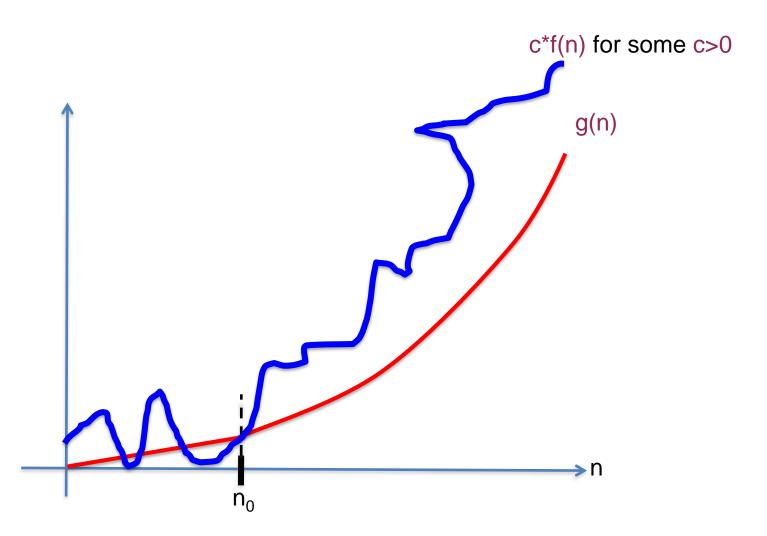


#### The actual run times

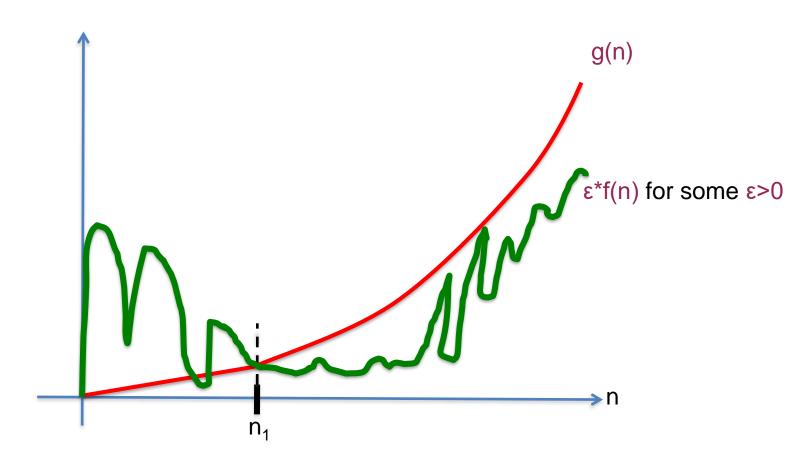




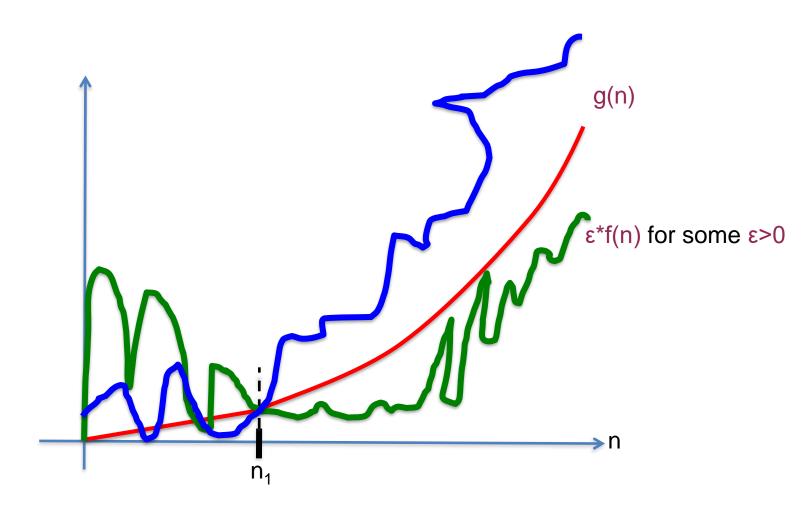
# g(n) is O(f(n))



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



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



## Properties of O (and $\Omega$ )

**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(n²) time

While (loop condition) // O(n²) iterations
Stuff happens // O(1) time

## **Another Reading Assignment**

**CSE 331** 

Support Pages -

# 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 O(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).

Sections 1.1, 1.2, 2.1, 2.2 and 2.4 in [KT]

## Gale-Shapley Algorithm

Intially all men and women are free

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
```

Output the engaged pairs as the final output

### Implementation Steps

How do we represent the input?

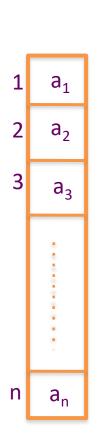
How do we find a free woman w?

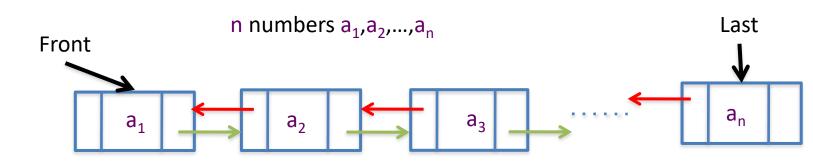
How would w pick her best unproposed man m?

How do we know who m is engaged to?

How do we decide if m prefers w' to w?

# **Arrays and Linked Lists**





	Array	Linked List
Access ith 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

#### Rest on the board...