# CSE 12 – Basic Data Structures: Running Time Analysis

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[Slides borrowed from
Marina Langlois and Christine Alvarado]

### Big-O

We say a function f(n) is "big-O" of another function g(n), and write f(n) is  $O(g(n))^*$ , if there are positive constants c and  $n_0$  such that:

•  $f(n) \le c g(n)$  for all  $n \ge n_0$ .

In other words, for large n, can you multiply g(n) by a constant and have it always be bigger than or equal to f(n)

n is the "size of your problem" Some notations write f(n) = O(g(n)) rather than f(n) is O(g(n))

# Steps for calculating the Big O bound on code or algorithms

- 1. Count the number of instructions in your code (or algorithm) as precisely as possible as a function of n, which represents the size of your input (e.g. the length of the array). This is your f(n).
  - Make sure you know if you are counting best case, worst case or average case - could be any of these!
- Simplify your f(n) to find a simple g(n) such that f(n) = O(g(n))

Almost always, what we care about is the WORST CASE or the AVERAGE CASE. Best case is usually not that interesting, unless we can prove it's slow!

In CSE 12 when we do analysis, we are doing **WORST CASE** analysis unless otherwise specified.

#### Analyzing the worst case

```
boolean find( String[] theList, String toFind ) {
  for ( int i = 0; i < theList.length; i++ ) {
    if ( theList[i].equals(toFind) )
      return true;
  }
  return false;
}
boolean fastFind( String[] theList, String toFind ) {
  return false;
}</pre>
```

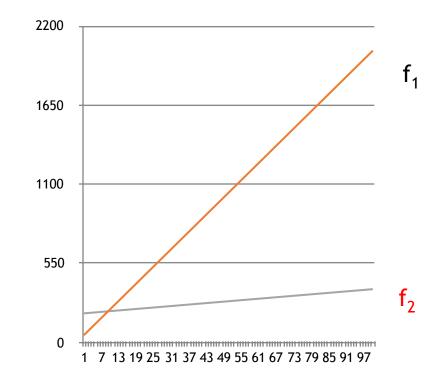
Which method is faster?

- A. find
- B. fastFind
- C. They are about the same

A. TRUE

B. FALSE

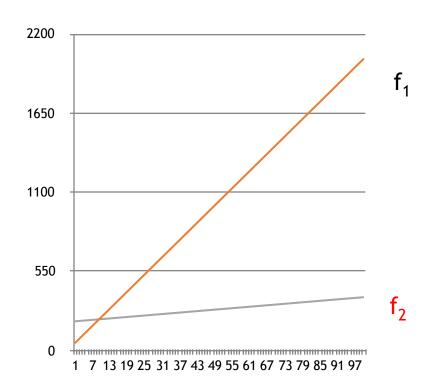
Why or why not?



<sup>\*</sup> You can't actually tell if you don't know the function, because it could do something crazy just off the graph, but we'll assume it doesn't.

- Obviously f<sub>2</sub> = O(f<sub>1</sub>)
   because f<sub>1</sub> > f<sub>2</sub> (after about n=10, so we set n<sub>0</sub>
   = 10)
  - f<sub>1</sub> is clearly an *upper*bound on f<sub>2</sub> and that's

    what big-O is all about



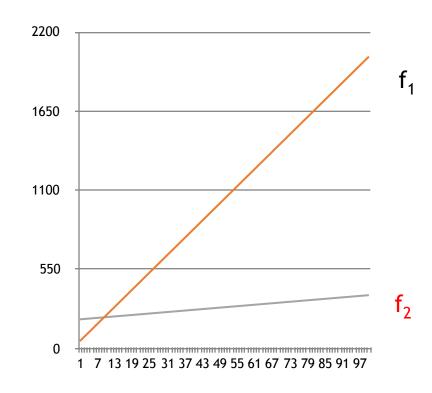
## $f_1$ is $O(f_2)$

f(n) = O(g(n)), if there are positive constants c and  $n_0$  such that  $f(n) \le c * g(n)$  for all  $n \ge n_0$ .

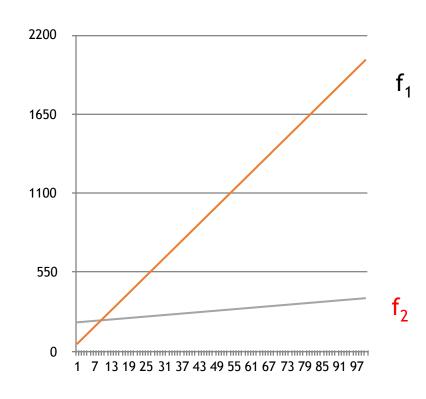
- A. TRUE
- B. FALSE

#### Why or why not?

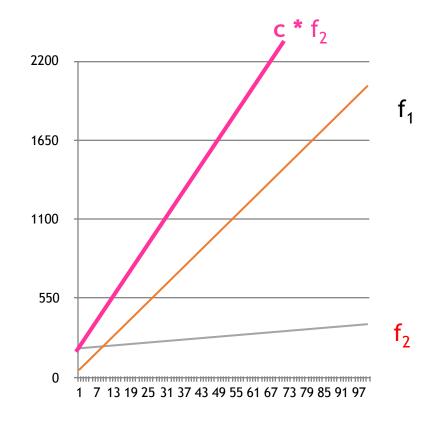
In other words, for large n, can you multiply  $f_2$  by a constant and have it always be bigger than  $f_1$  for large enough n?



- Obviously  $f_2 = O(f_1)$ because  $f_1 > f_2$  (after about n=10, so we set  $n_0$ = 10)
  - f<sub>1</sub> is clearly an *upper* bound on f<sub>2</sub> and that's what big-O is all about
- But  $f_1 = O(f_2)$  as well!
  - We just have to use the "c" to adjust so f<sub>2</sub> that it moves above f<sub>1</sub>



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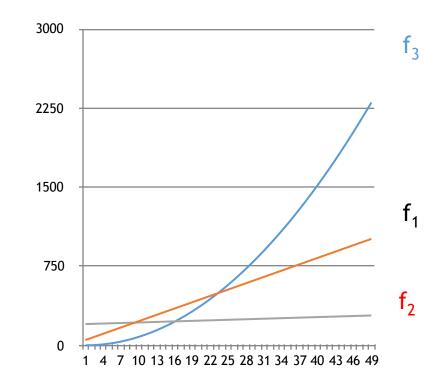


$$f_1$$
 is  $O(f_3)$ 

A. TRUE

B. FALSE

Why or why not?

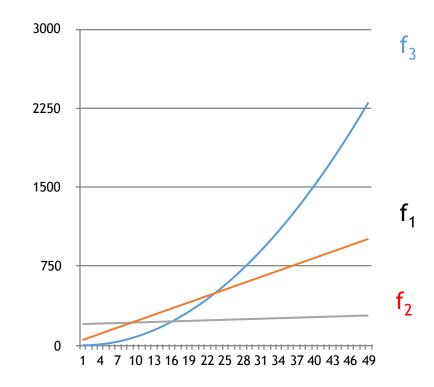


$$f_3$$
 is  $O(f_1)$ 

A. TRUE

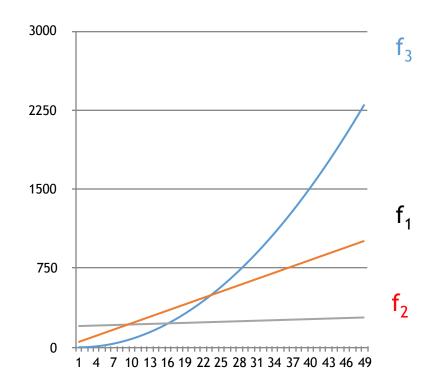
B. FALSE

Why or why not?



### $f_1$ is $O(f_3)$ but $f_3$ is not $O(f_1)$

There is no way to pick a c that would make an O(n) function (f<sub>1</sub>) stay above an O(n<sup>2</sup>) function (f<sub>3</sub>).



#### Common Big-O confusions when trying to argue that $f_2$ is $O(f_1)$ :

- What if we multiply  $f_2$  by a large constant, so that  $c^*f_2$  is larger than  $f_1$ ? Doesn't that mean that  $f_2$  is not  $O(f_1)$ ? No, because we get to control the constants to our advantage, and only on  $f_1$ .
- What about when n is less than 10?
   Isn't f2 larger than f1?
   Remember, we get to pick our n0, and only consider n larger than n0.

