# Studio 3 Orders of Growth, Higher-Order Functions and Scope of Names

CS1101S AY20/21 SEM 1
Studio 03A

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# Studio X Agenda

- Admin and announcements
- Recap:
  - Order of growth
  - Higher order functions
  - Scope of names
- Studio sheets

#### Admin

# Admin Stuff Reading Assessment 1

- RA1 is this friday during lecture slot!!!
- Past year papers on LumiNUS
- All MCQs
- Don't get rekt



### Admin Stuff Asking Questions

- Ask in the studio telegram group first, or message your friends
- Refer to piazza, many questions have been answered already
- Message me as <u>LAST</u> resort
  - Please don't message me past 2300hrs
- Asking the right questions is an important skill!

#### Admin Stuff Mission - Curve Introduction

- Use of `connect\_rigidly`, `transform`:
  - Challenge yourself to not use these functions!
  - Don't need to tell me, just make sure you understand

#### Recap: Orders of Growth

#### Recap Orders of Growth - Terminology

- Time complexity, space complexity (resources needed)
- Described with: Big Omega, Big Theta, Big Oh
- Constant, Logarithmic, Linear, "Linearithmic", Quadratic, Exponential
- Efficiency

i don't think this is an actual word

#### Future mods:

• CS2040S, CS3230, CS3233 (not for the faint hearted)

#### Recap Orders of Growth - Definition

- "Big Oh" O(): Upper bound
  - Most common, used in analysing worse case scenarios
- "Big Theta" Θ(): Tight bound
  - Most useful, difficult to compute (or near impossible)
- "Big Omega"  $\Omega$ (): Lower bound
  - Not useful in most scenarios.

#### Recap Orders of Growth - Definition

• Let n denote the size of the problem, and let *r*(*n*) denote the resource needed solving the problem of size n.

#### Big Theta:

- the function r has order of growth  $\Theta(g(n))$ , if there are positive constants k1 and k2,
- such that  $k1 \cdot g(n) \le r(n) \le k2 \cdot g(n)$  for any sufficiently large value of n.

#### • Big Oh:

- the function r has order of growth O(g(n)) if there is a positive constant k,
- such that  $r(n) \le k \cdot g(n)$  for any sufficiently large value of n.

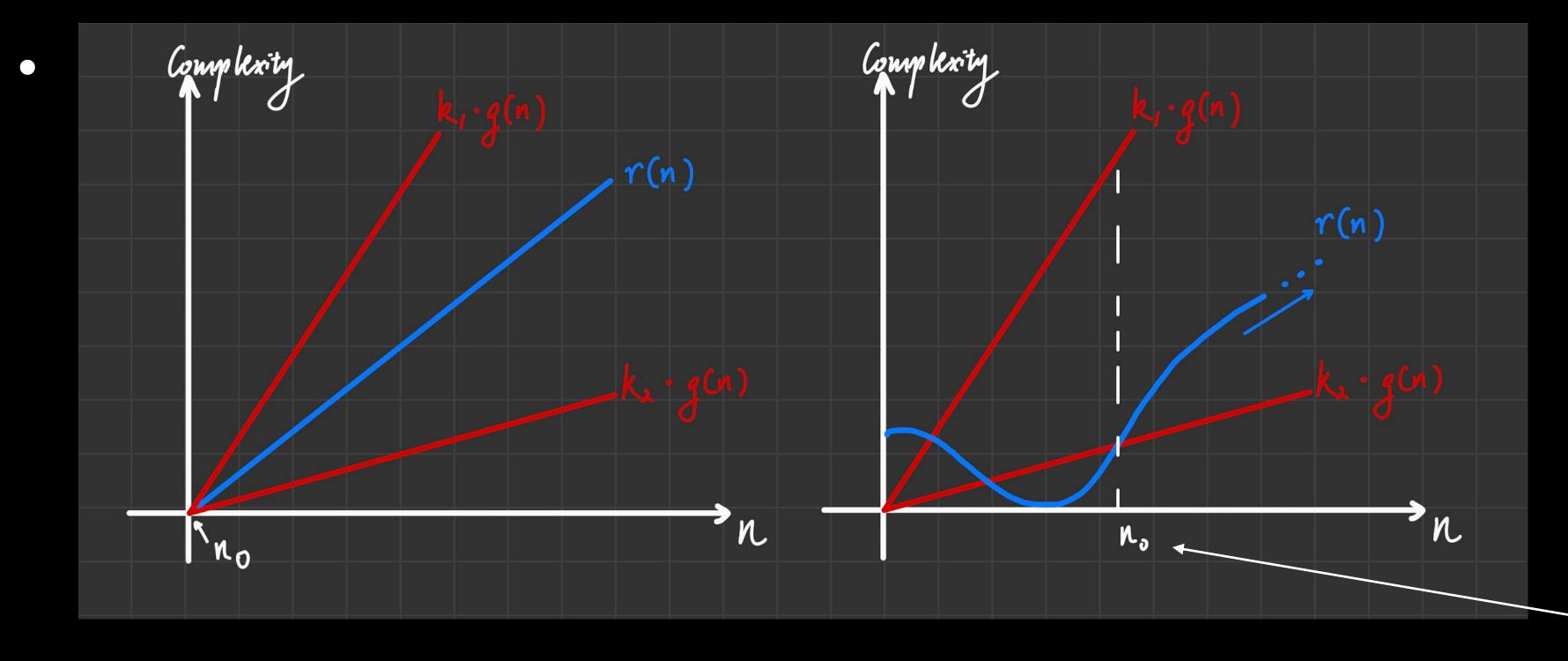
#### Big Omega:

- the function r has order of growth  $\Omega(g(n))$  if there is a positive constant k,
- such that  $k \cdot g(n) \le r(n)$  for any sufficiently large value of n.

#### huh?????

### **Recap**Orders of Growth - Big Theta (Tight Bound)

• Assume g(n) = n



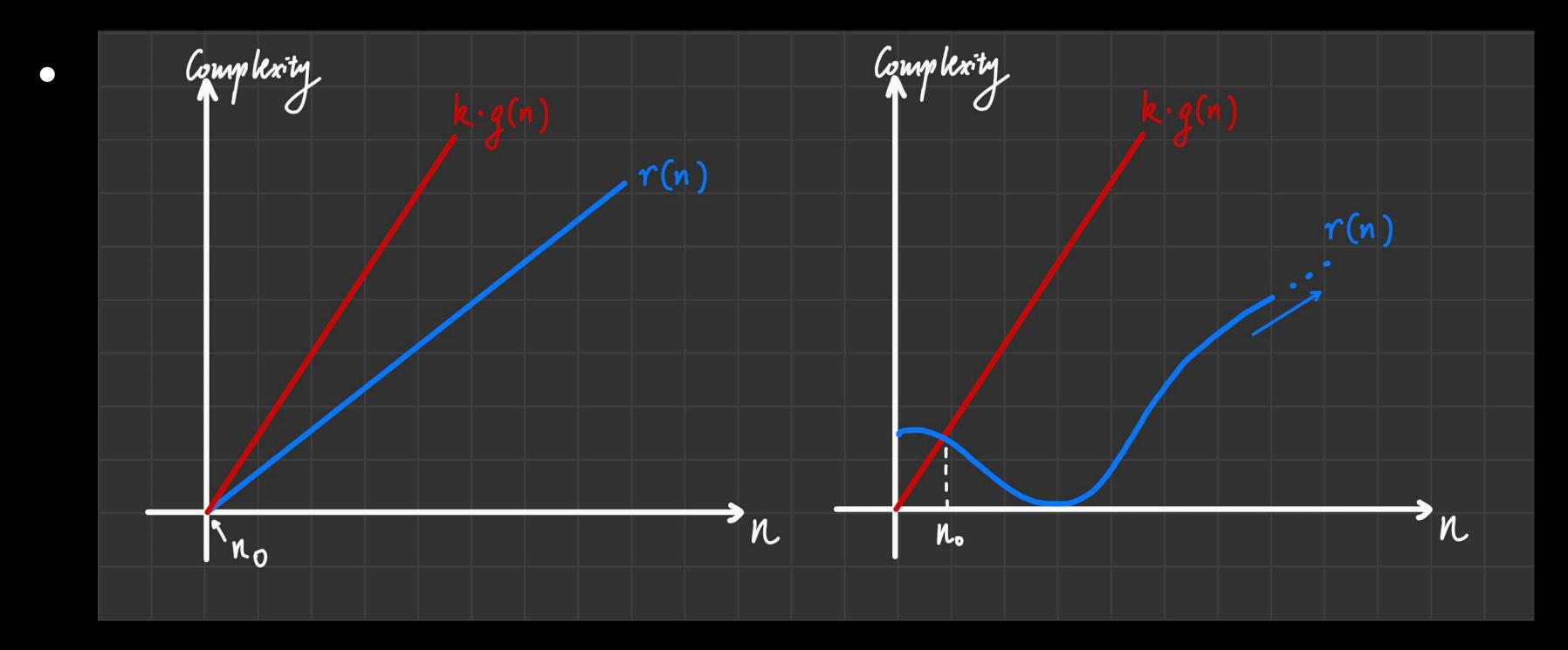
Hence, the functions have orders of growth  $\Theta(g(n))$  for  $n \ge n0$  More specifically, they have orders of growth  $\Theta(n)$ 

recall:

"for any sufficiently large values of n", this only applies for n >= n0

# Recap Orders of Growth - Big Oh (Upper Bound)

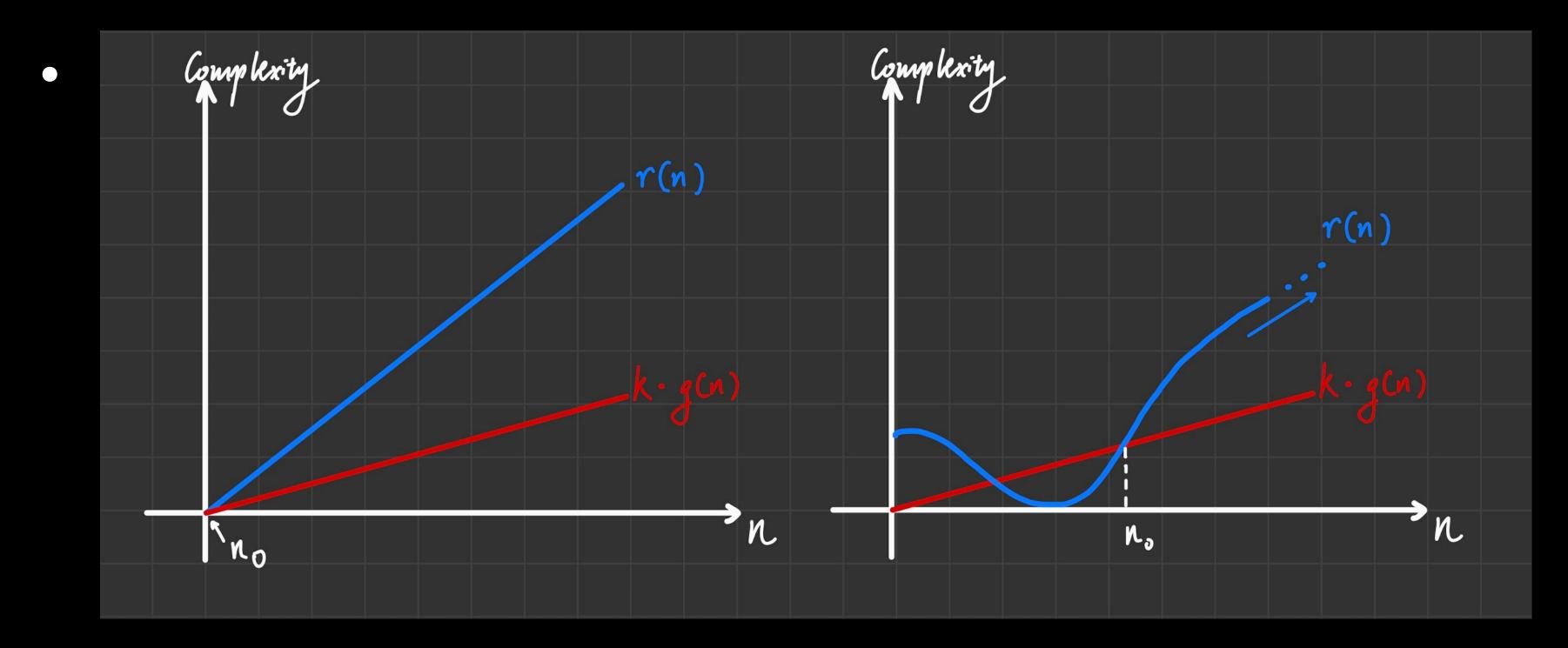
• Assume g(n) = n



Hence, the functions have orders of growth O(g(n)) for  $n \ge n0$ More specifically, they have orders of growth O(n)

## Recap Orders of Growth - Big Omega (Lower Bound)

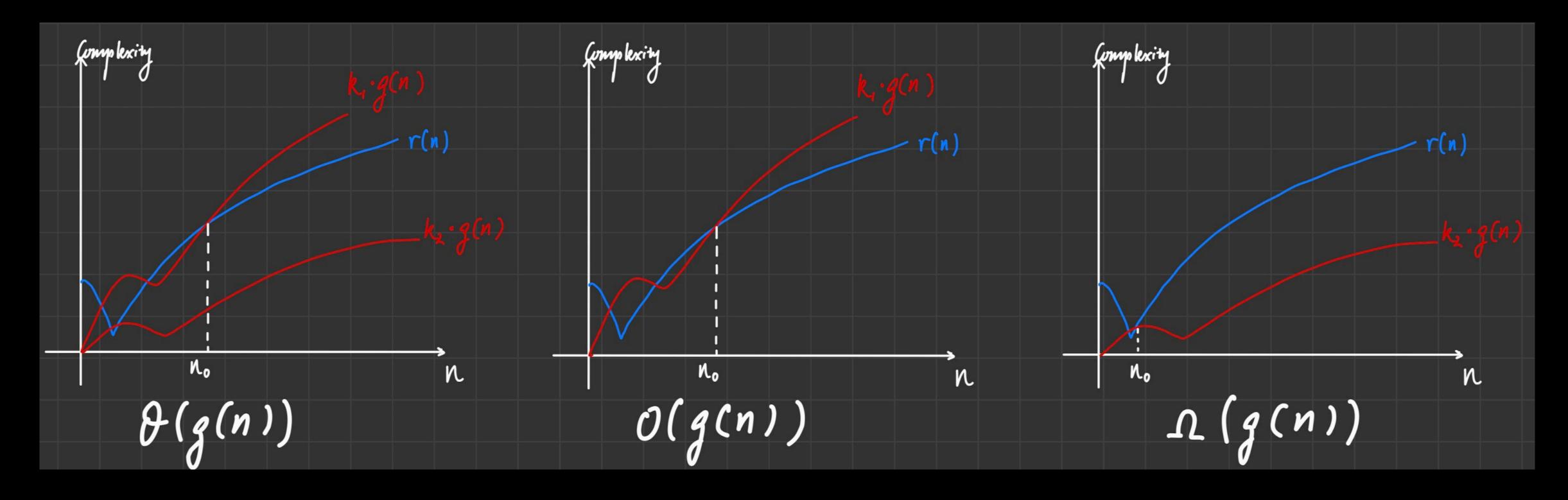
• Assume g(n) = n



Hence, the functions have orders of growth  $\Omega(g(n))$  for  $n \ge n0$ More specifically, they have orders of growth  $\Omega(n)$ 

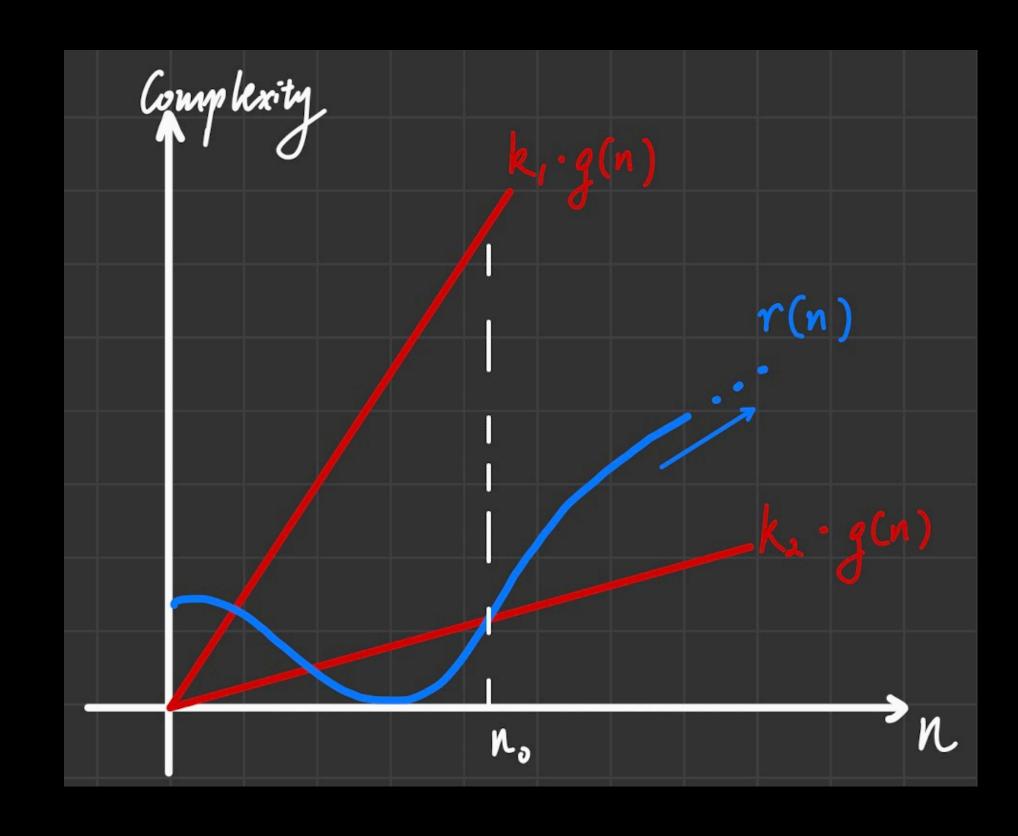
#### **Recap**Orders of Growth - in General

- g(n) and r(n) can be all sorts of funny functions!
- O,  $\Omega$ ,  $\Theta$  stand as long as there's some function g(n) that binds r(n) for  $n \ge n0$



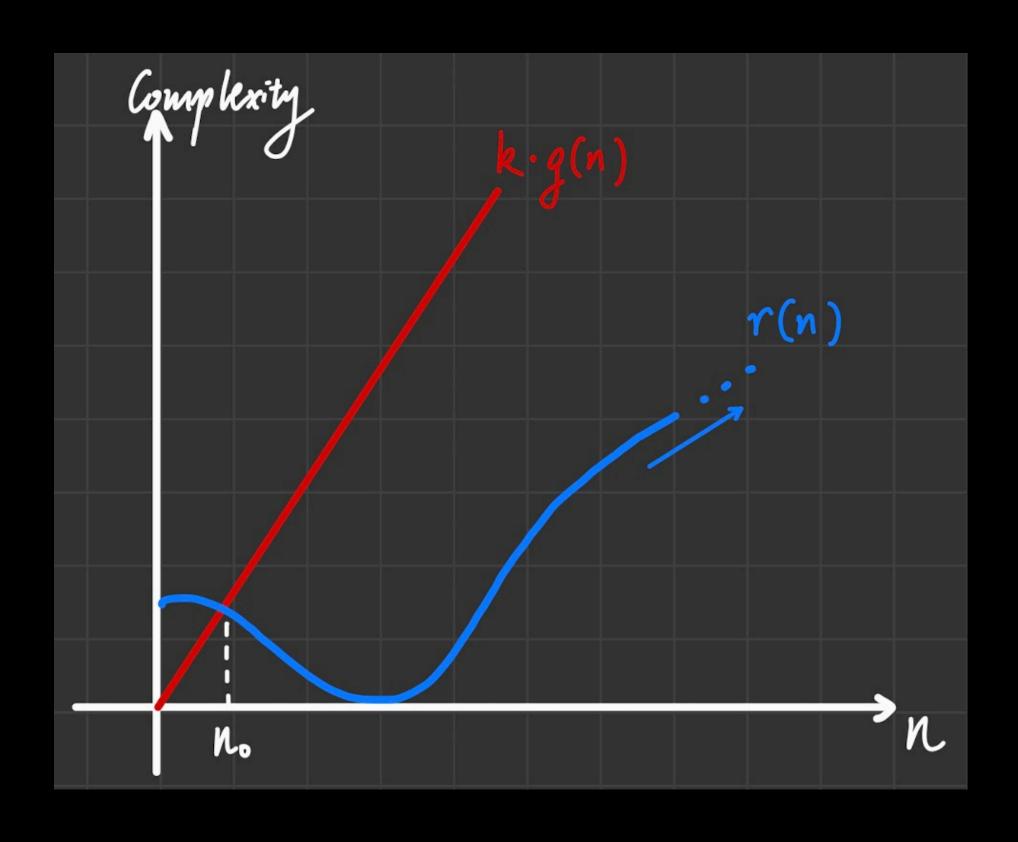
#### Recap Orders of Growth - Why Do We Use Them?

- Why <u>Big Theta?</u>
  - Tightest bound we can find for the algorithm's complexity
  - Represents the complexity most accurately



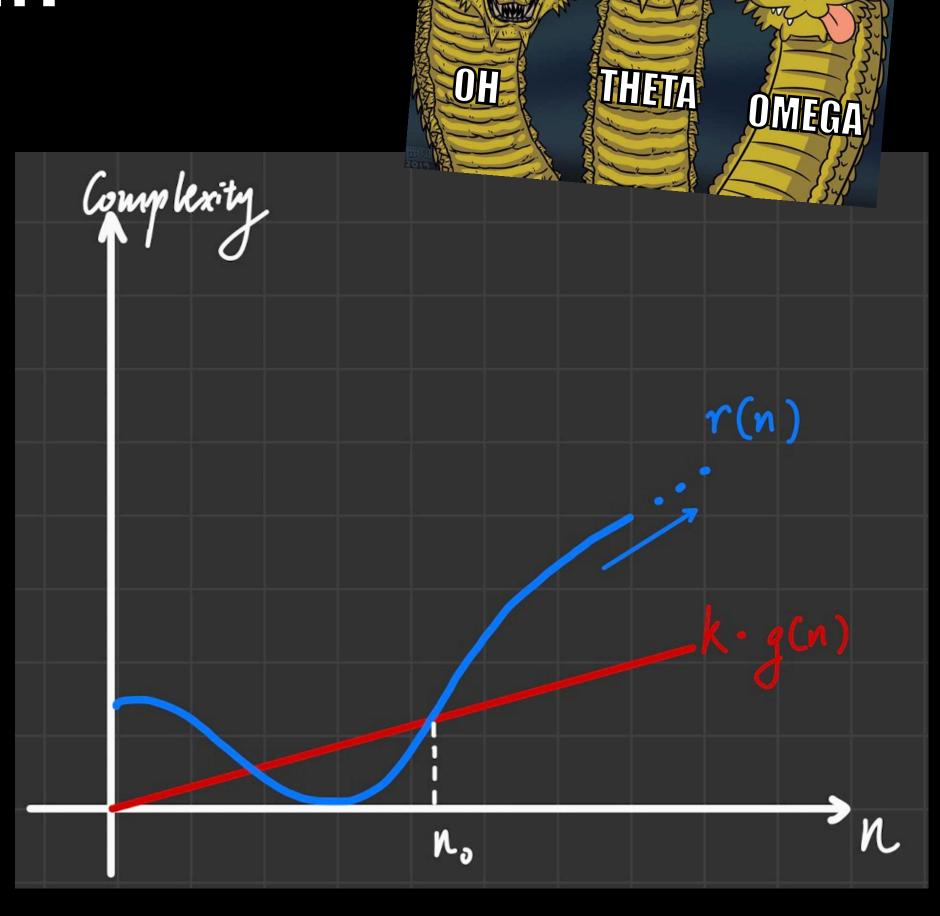
#### Recap Orders of Growth - Why Do We Use Them?

- Why Big Oh?
  - What does the red line represent?
  - The worst case scenario of the algorithm!
  - i.e. it says: "when n is large enough, r(n) will be bad, but it can't do worse than me!"



### Recap Orders of Growth - Why Do We Use Them?

- Why <u>Big Theta</u>?
  - Red line: the *best case* scenario of the algorithm
  - We usually don't concern ourselves with that
    - or at least not in CS1101S



#### **Recap**Orders of Growth - Constants

- We do <u>not</u> consider constants in complexity
  - $O(1) \equiv O(2), \Theta(n) \equiv \Theta(99999 * n), etc.$
  - Changing base:

$$\log_a N = \frac{1}{\log_b a} \log_b N$$

$$\log_a N = C \log_b N$$

$$O(\log_a N) = O(C \log_b N)$$

$$O(\log_a N) = O(\log_b N) \longrightarrow \text{in general, O(log N)}$$

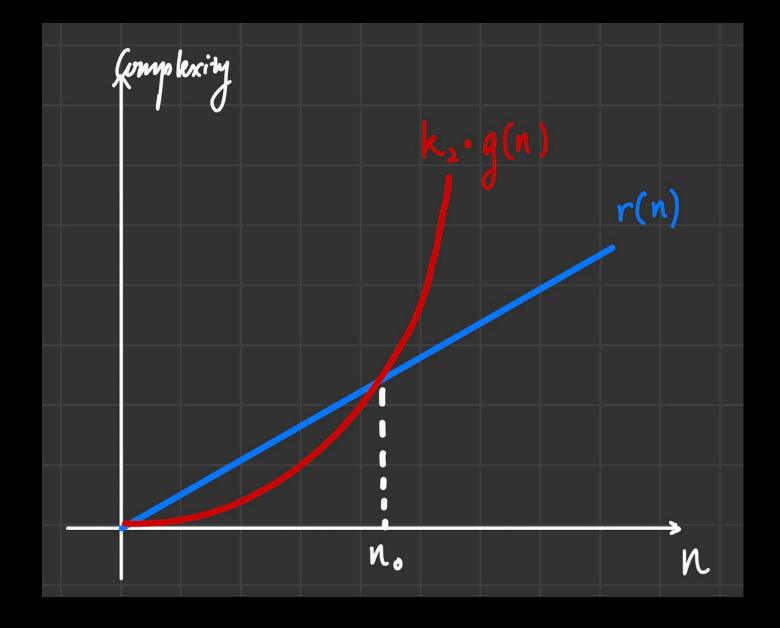
#### Recap Orders of Growth - Quiz

- Claim: "if my algorithm runs in Θ(1) time then it runs in O(n) time"
  - true or false?
  - Answer: true!

#### some tips

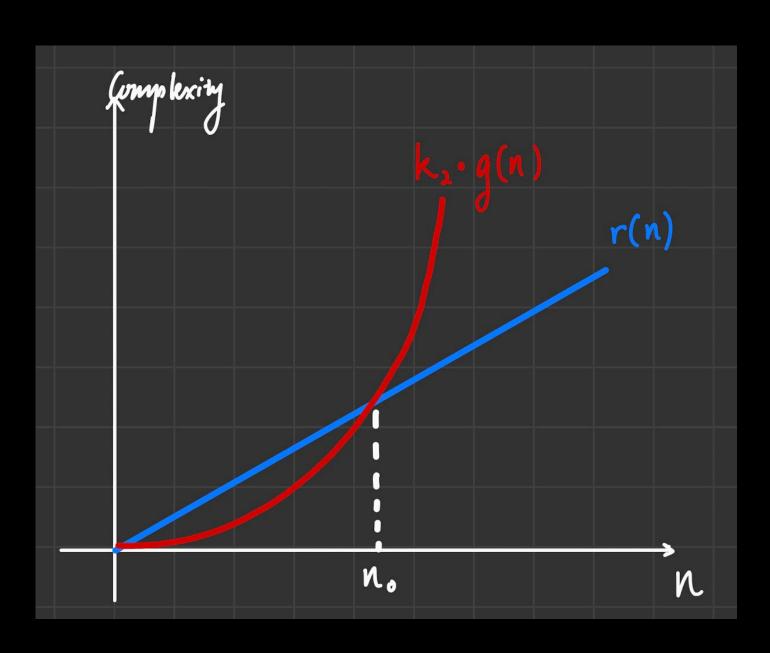
#### Recap Orders of Growth - Equivalence Relationships

- Assume a function with complexity  $\Theta(n)$  and a resource function r(n) = n
  - Thus,  $k1 \cdot g(n) \le r(n) \le k2 \cdot g(n)$  for some k1, k2, g and n0
- Can we find a k2 and g such that  $r(n) \ge k2 \cdot g(n)$  for  $n \ge n0$ ?



#### **Recap**Orders of Growth - Equivalence Relationships

- So what can we deduce from this?
  - No upper limit to Big Oh, O(infinity) perhaps
  - Lower limit to Big Omega: Ω(1)

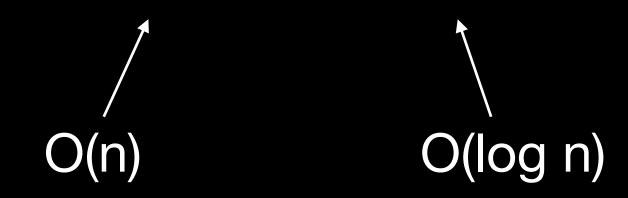


#### **Recap**Orders of Growth - Equivalence Relationships

- If a algorithm runs in  $\Theta(n)$ , then the algorithm is also:
  - $\Omega(1)$ ,  $\Omega(n)$  and O(n),  $O(n^2)$
- Every algorithm's complexity is technically  $\Omega(1)$  and O(infinity)
  - trivial (if you don't understand, go back to the definitions!)
  - but this doesn't help us with analysis:(

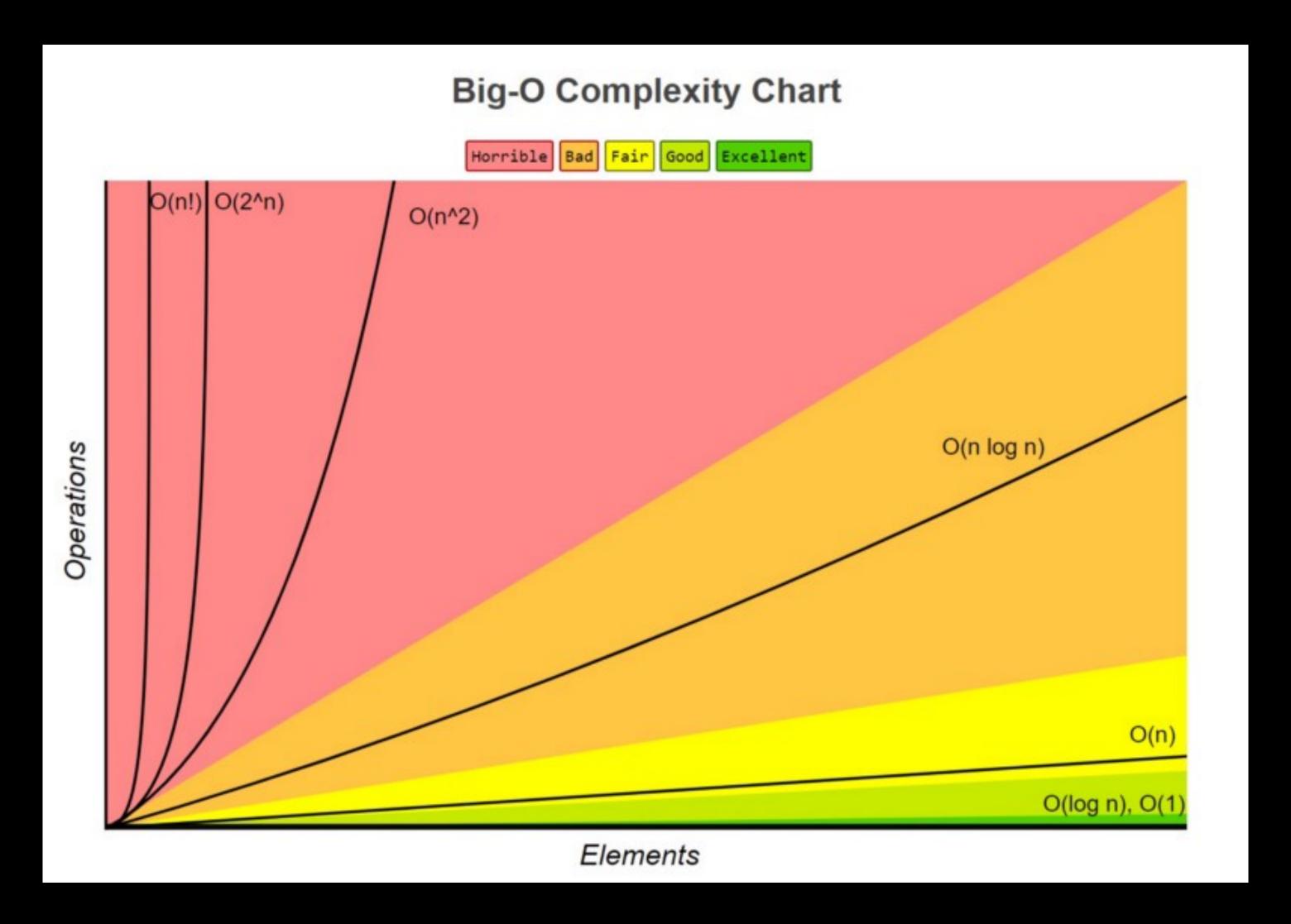
#### Recap Orders of Growth - Ranked

- From least resource complex to most:
  - $O(1) < O(\log n) < O(n \cdot \log n) < O(n^2) < O(n^3) < O(2^n) < O(n!)$
- Aim to find a more efficient algorithm!
  - Recall from S3: `expt` vs `fast\_expt`



for the interested: trying to brute force the Travelling Salesman Problem, or finding all possible permutations

# Recap Orders of Growth - Ranked



#### Recap Orders of Growth - Summation

- When adding: drop the less significant terms
  - $O(n^2 + 2n + 1) = O(n^2)$
  - $O(log(n) + n) \equiv O(n)$

#### Recap Orders of Growth - Multiplication

- When multiplying: take the product
  - $O(n^2 * n) \equiv O(n^3)$
  - $O(\log(n) * n) = O(n \cdot \log n)$

#### Recap Orders of Growth - Coefficients

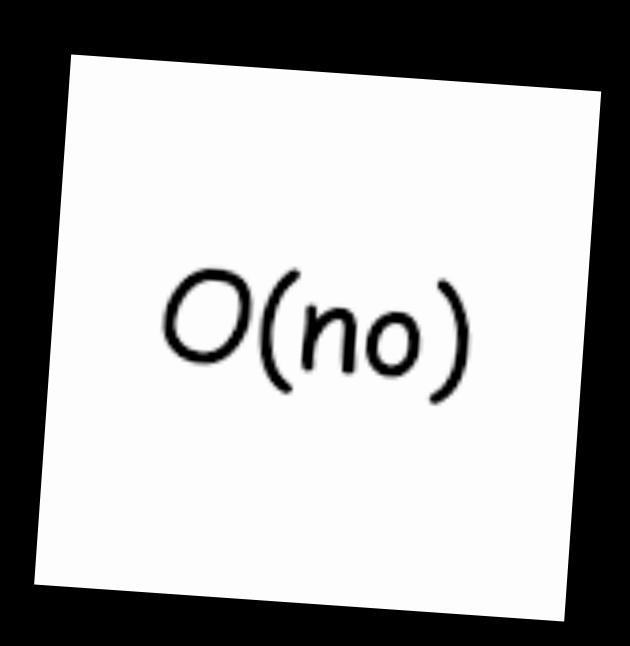
- Drop coefficients
  - $O(n^2 \cdot 0.5) = O(n^2)$
  - O(3) = O(3 \* 1) = O(1)

#### Recap Orders of Growth - Summary

- Complexities are used to indicate the resources used for an algorithm
- Most commonly used: Θ and O
- Addition and multiplication rules
- Coefficients are discarded

#### Recap Orders of Growth - Summart

- You WILL be asked to do complexity analysis during exams.
  - Analyse a given programme
  - Write a programme in XXX time complexity
  - Write a programme, THEN analyse it!



#### Any questions?

#### Recap: Higher Order Functions

#### **Recap**Higher Order Functions - Function Definitions

• In S2, we've learnt how to define functions:

```
function add_one(x) {
   return x + 1;
}
```

- Now we can define "anonymous" functions (arrow functions)
  - Makes your programmes more concise

#### Recap

#### Higher Order Functions - Function Definitions

Defining anonymous functions:

```
x \Rightarrow x + 1; // no name

(x, y) \Rightarrow x + y; // no name

parameters expression
```

To give names:

To call this function:

```
add_one(2); // same as normal functions
```

```
equivalent to:
function add_one(x) {
   return x + 1;
}
```

#### **Recap**Higher Order Functions - Function Signatures

- Every function has a "signature"
- It takes in something(s) and return something(s)
- Functions are sensitive to their signatures
  - Can't run if signature is incorrect!

### Recap

### Higher Order Functions - Function Signatures

- Example:
  - const sum =  $(x, y) \Rightarrow x + y;$
  - // Signature: (number, number) -> number
  - i.e. function `sum` takes in 2 numbers, does some stuff, and returns 1 number
- Recall: abstraction!

# **Recap**Higher Order Functions - Function Signatures

- Spot the mistake:
  - show(stackn(heart, sail));
- Error:
  - Signature: stackn: (number, rune, rune) -> rune
  - But we only gave two runes!

# **Recap**Higher Order Functions - Function Signatures

- Another thing to take note:
  - Type is important too!

```
function add_one(x) {
    return x + 1;
}
add_one(rcross);
```

How can we possibly add a number to a rune???

### Recap Higher Order Functions - Quiz

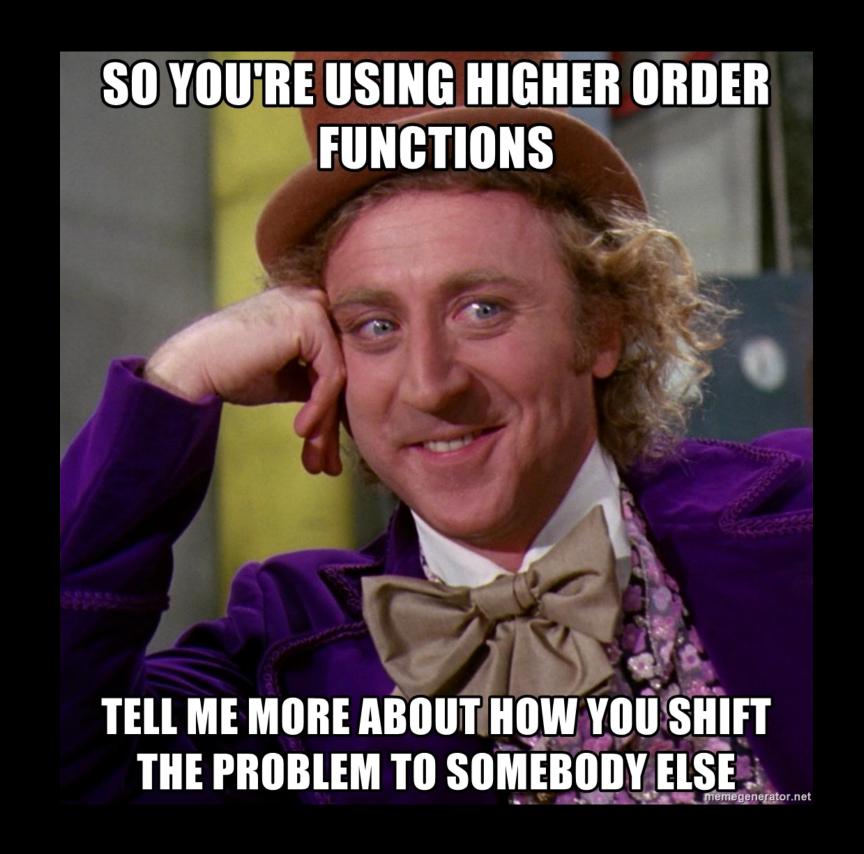
Identify the signatures:

```
    const f1 = () => 1;
    const f2 = some_val => some_val ? false : true;
    const f3 = func => x => func(x);
    const f4 = (x, y, z) => 0;
```

Discuss in your pairs!

# **Recap**Higher Order Functions - Functions as Arguments

- Functions can take in functions
- Functions can return functions
- i.e. make some other dude do your dirty work



## **Recap**Higher Order Functions - Quiz

- Functions as arguments follow parameter name binding rules
  - Creates a key-value pair, where the value is a function, instead of numbers, booleans or runes

```
function foo(f, x) {
  return f(x);
}
foo(x => x + 1, 10); // binds `f` in `foo` to (x => x + 1)
```

# **Recap**Higher Order Functions - Interpreting HoF

- Strategy to interpret higher order functions:
  - Find the left most arrow
  - Consider the two sides:
  - Whatever's on the left of the arrow are the parameters
  - Whatever's on the right of the arrow is returned

## Recap

#### Higher Order Functions - Interpreting HoF

```
    const f3 = f => x => f(x + 1);
    Parameter: `f` (some function)
    Returns: `x => f(x + 1)` (another function)
```

• Signature: function -> function

# Recap Higher Order Functions - Summary

- Functions can be anonymous
- Functions can be declared as `constants`
- Functions can take in functions and return functions

## Recap

#### Higher Order Functions - Another Quiz

```
const bar = abc => def => abc * def;
```

const foo = a => a;

- Evaluate these expressions:
  - foo(117);
  - bar(5)(10);
  - bar(foo(1))(foo(7));
  - bar(bar(3)(4))(bar(5)(foo(6)));

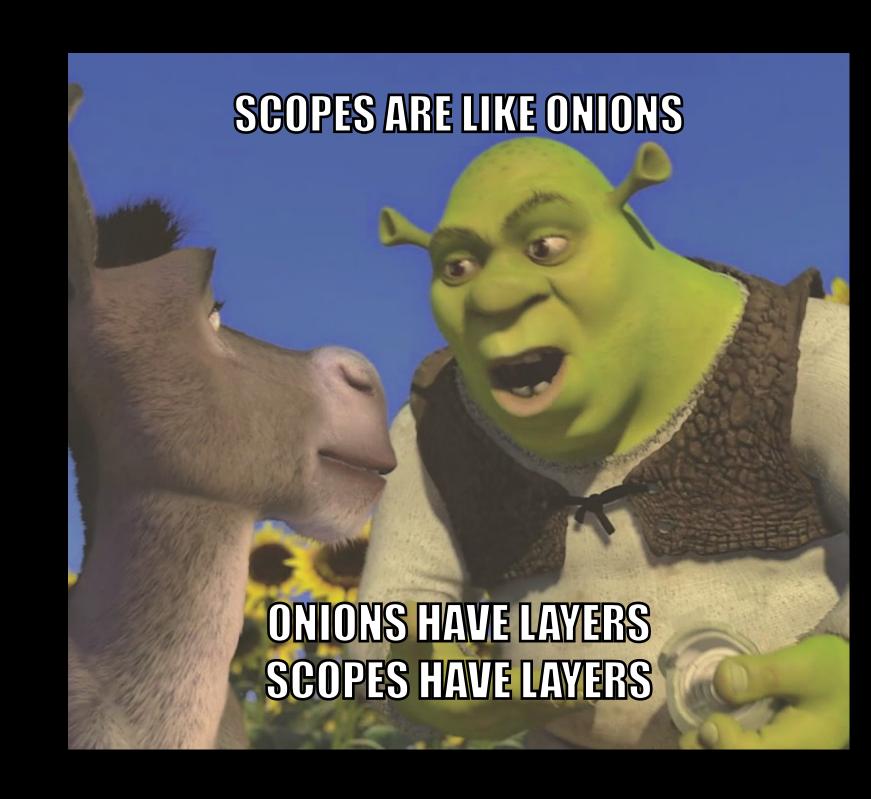
## Recap: Scope of Names

### Recap Scope of Names - Lexical Scoping

- Scoping rules:
  - A name occurrence refers to the closest surrounding declaration

### Recap Scope of Names - Overview

- Scopes are like onions: there are layers
  - When we refer to a name, the interpreter looks for it in the current scope
  - If it doesn't exist, it looks outwards in the layers until it finds it, or the global scope is reached
  - If it's still not found in the global layer, then the interpreter raises an error for "name XXX not declared"



### Recap Scope of Names - Overview

- What entails a scope?
  - Global context is a scope (the big universe)
    - Contains predefined functions and values (display, math\_PI, etc)
  - Braces provide block scope `{ ... }`
  - Functions create scope

### Recap Scope of Names - Shadowing

Consider this programme:

```
const x = 7;
function f(y) {
   const x = 0;
   return x * y;
}
f(10); // what's the result?
```

### Recap Scope of Names - Shadowing

Consider this NEW programme: (without using SourceAcademy)

```
const x = 7;
function f(y) {
    // const x = 0;
    return x * y;
}
f(10); // what's the new result?
```

### Recap Scope of Names - Shadowing

- What are the differences?
- Why did this happen?
- Discuss among your pairs

# Recap Scope of Names - Lexical Scoping

- Scoping rules:
  - A name occurrence refers to the <u>closest surrounding declaration</u>
  - The interpreter can only look upwards and outwards (to the left)

## End of Recap

## Challenge

## Challenge Start Your Brains

- Consider this programme:
- Attempt without running the programme in playground

```
const cs1101s = path => (quest, mission) => contest => contest(quest, mission);
const cs = ceg => ((infosec, bza, infosys) => bza);

cs(cs1101s(false))(NaN, 117, "hello");

// NaN is just a pre-declared value in the global scope
```