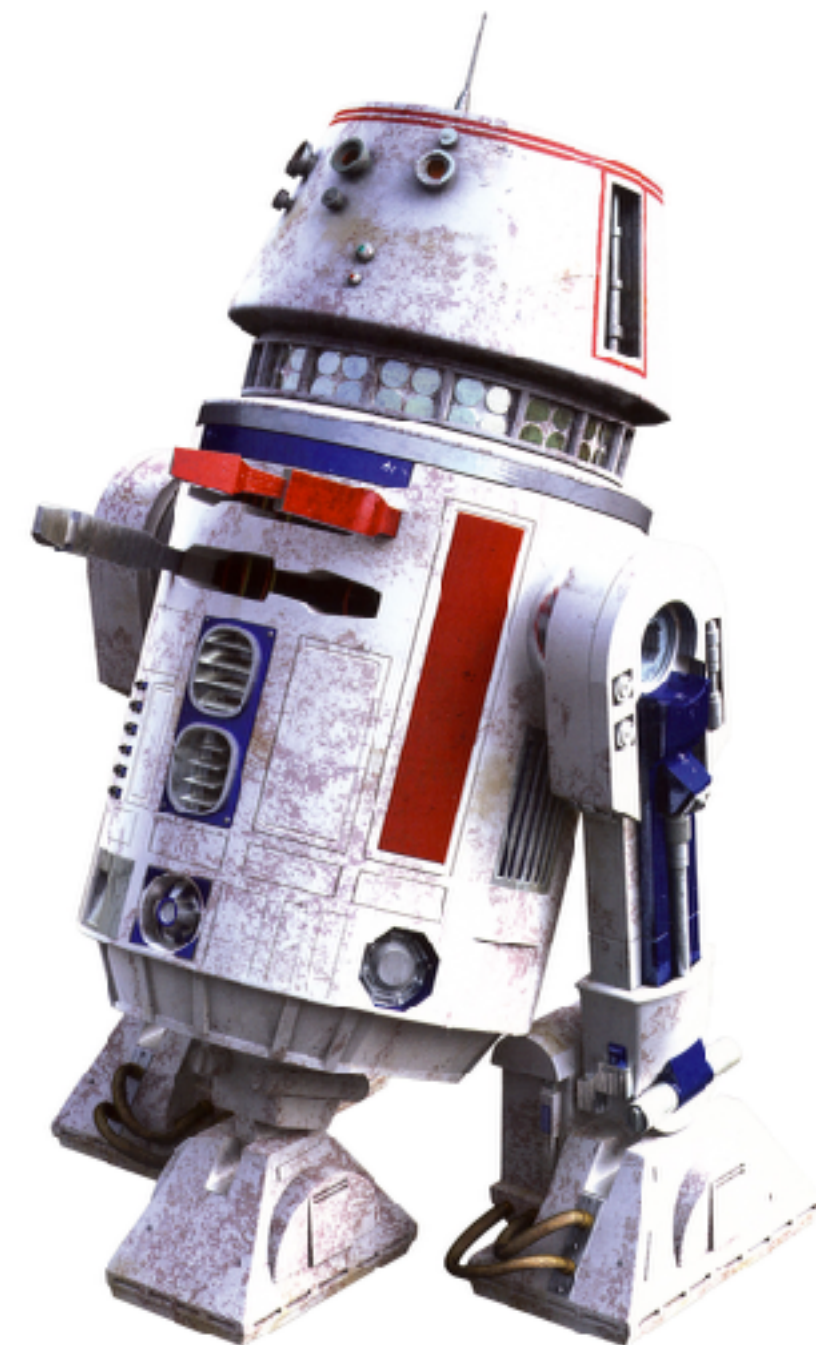


# BIG O

---

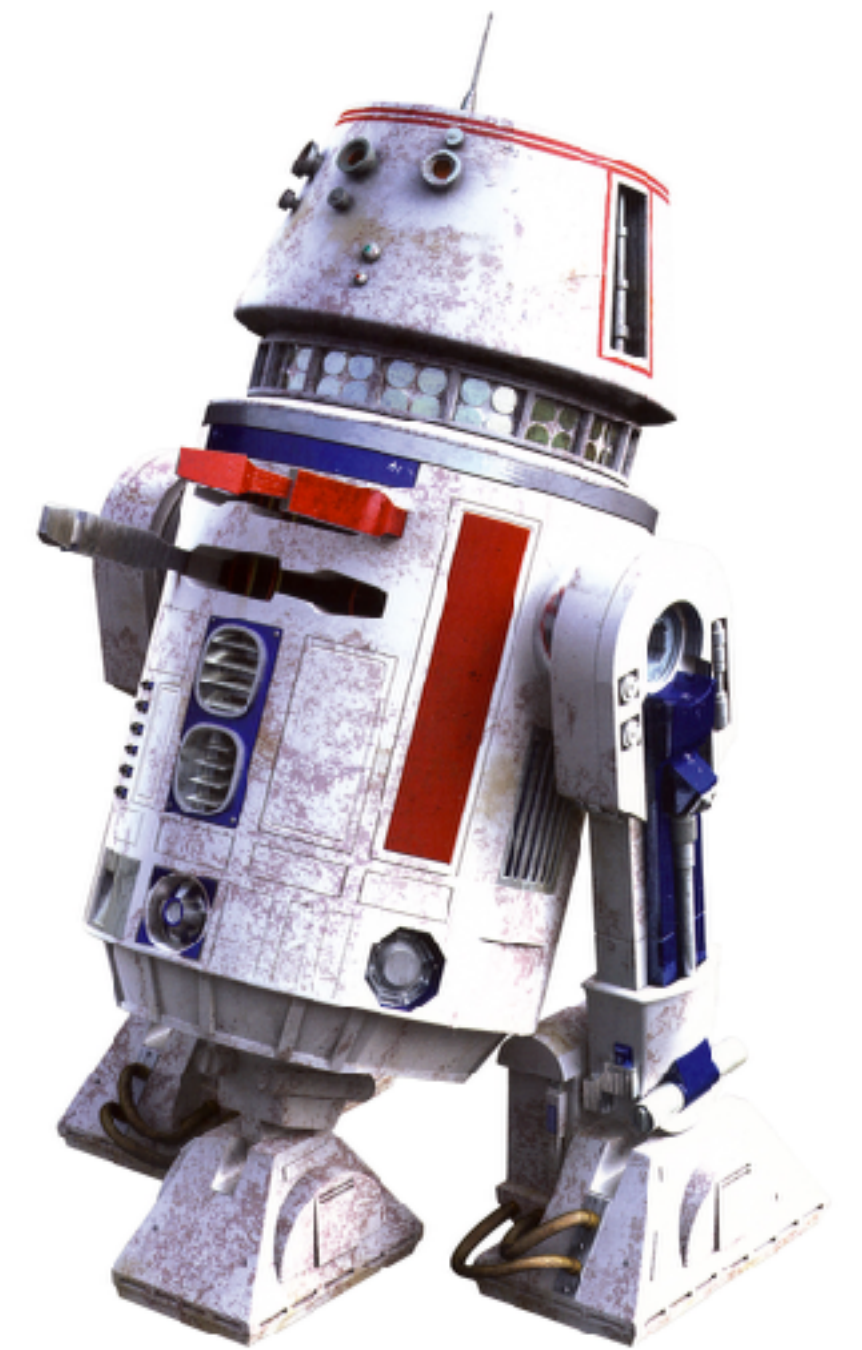
*Time and space complexity*

**BUT FIRST...LET'S GO BUY  
SOME DROIDS**





1 day = 5s



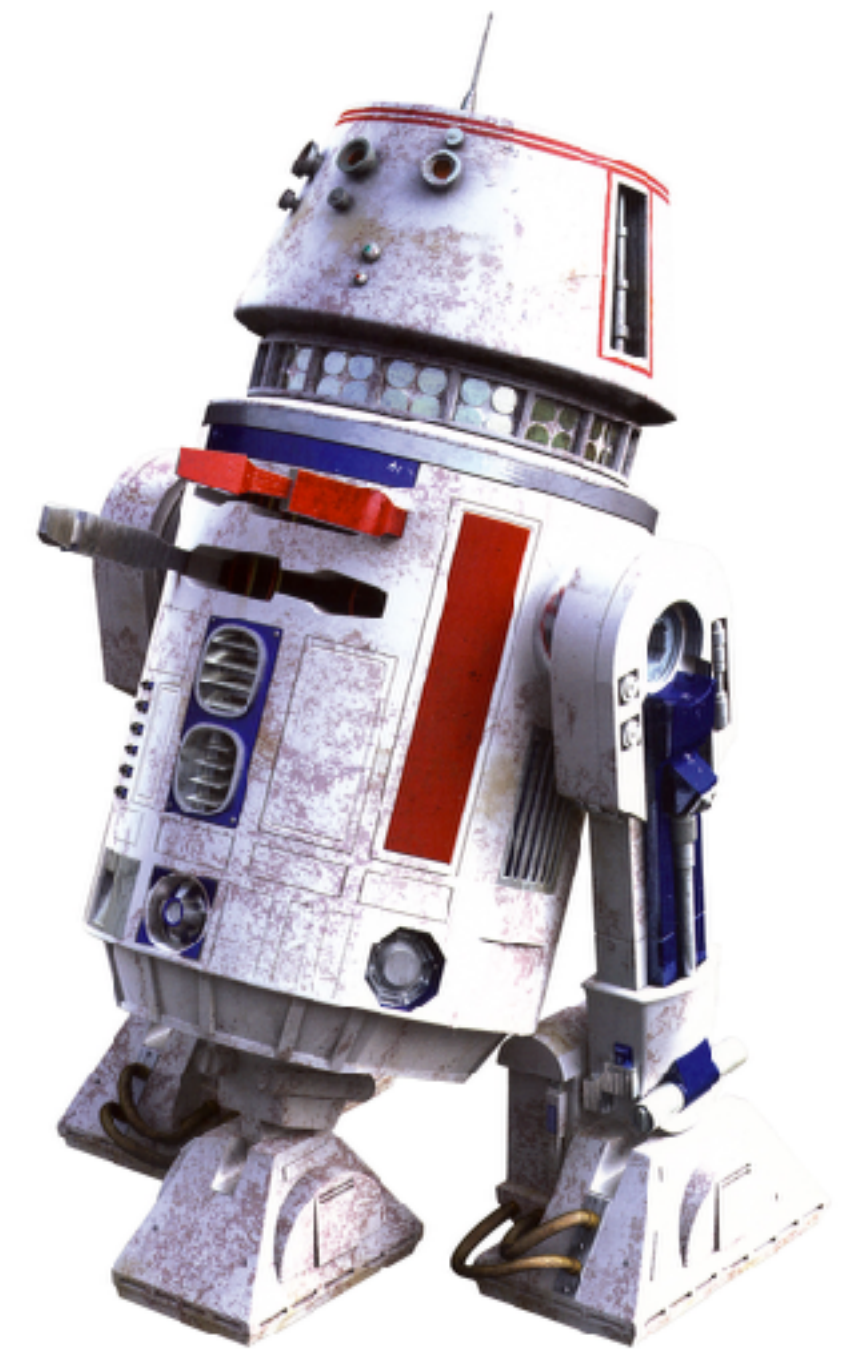
1 day = 1s





1 day = 5s

1 day = 1s

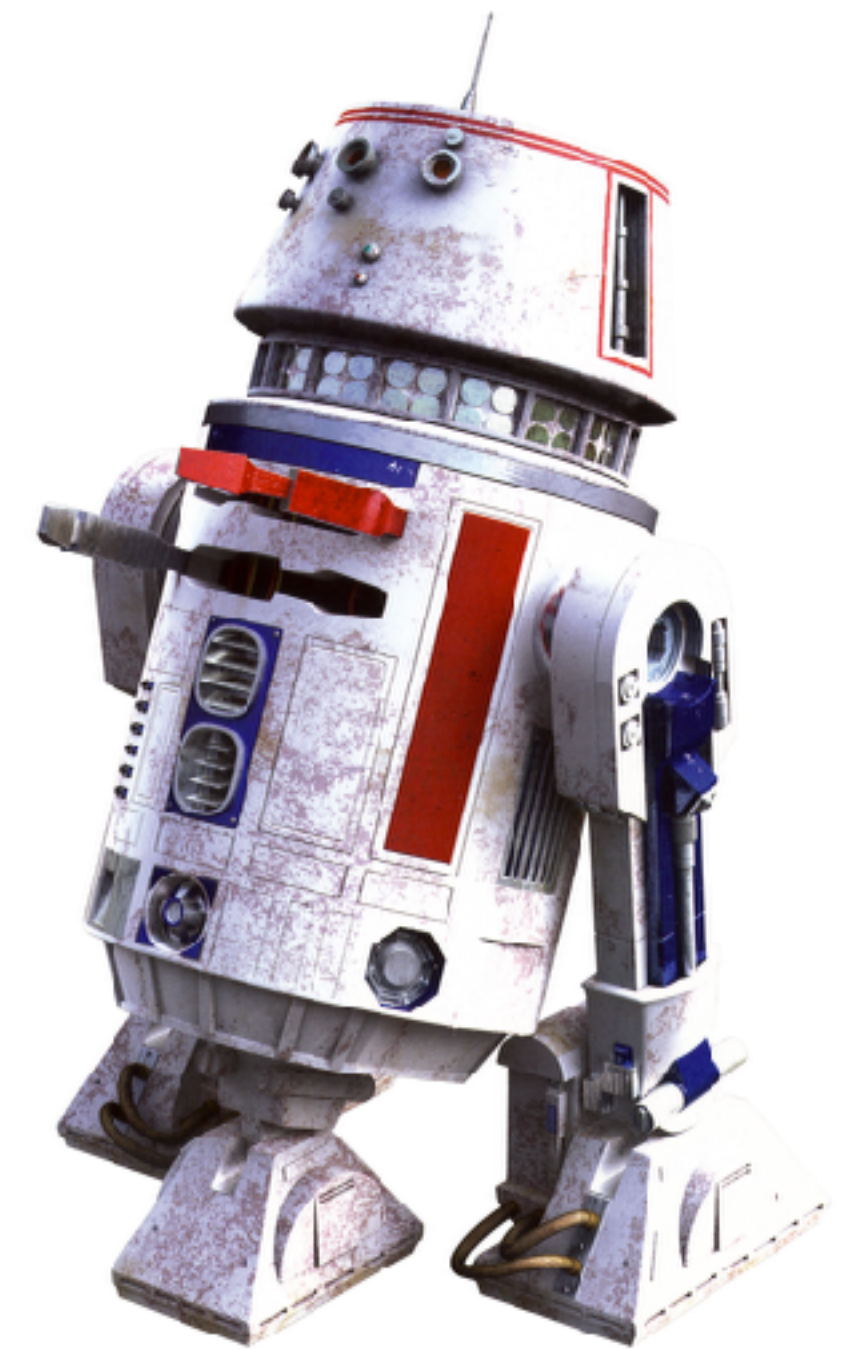




1 day = 5s

2 days = 10s

1 day = 1s



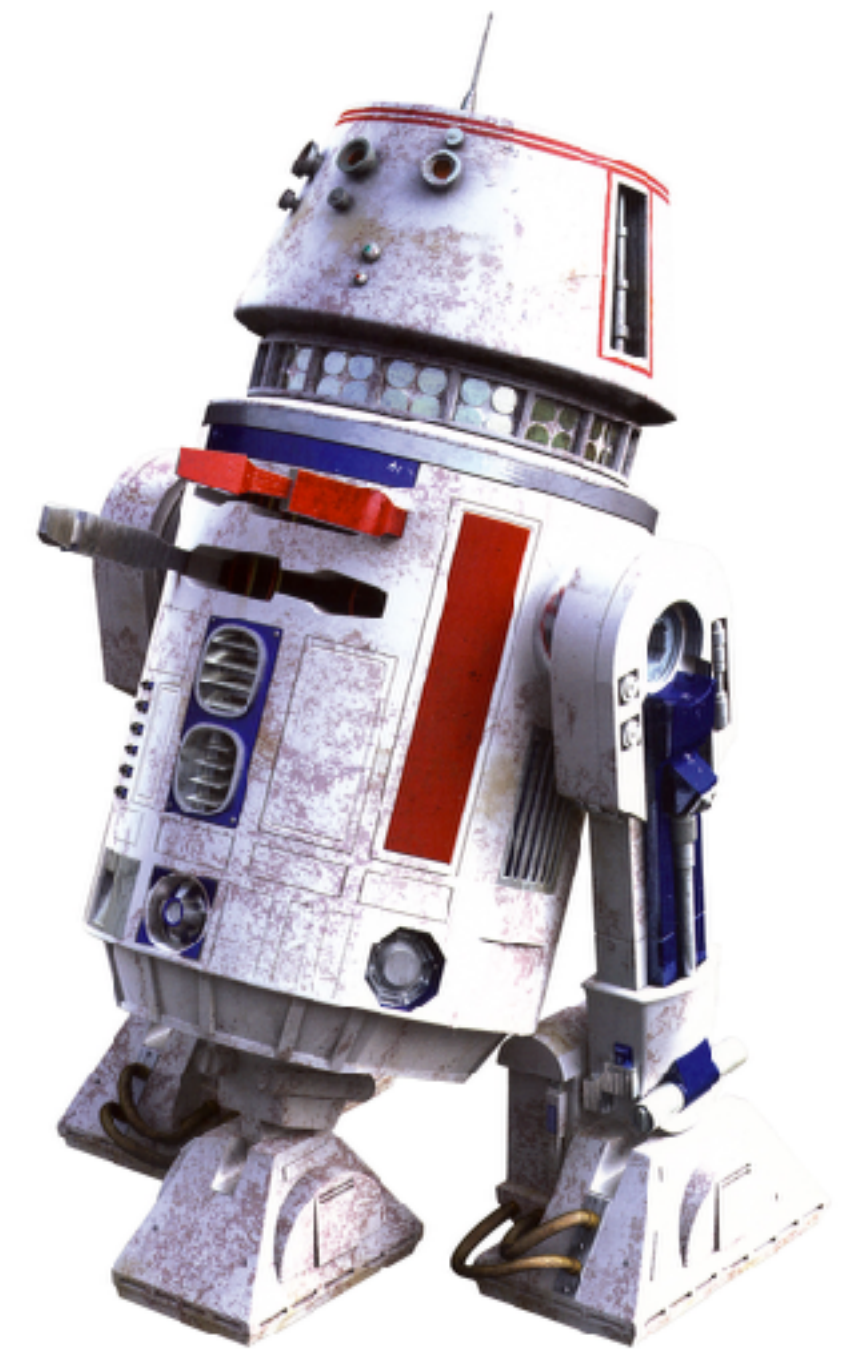
2 days = 4s





1 day = 5s  
2 days = 10s

1 day = 1s  
2 days = 4s

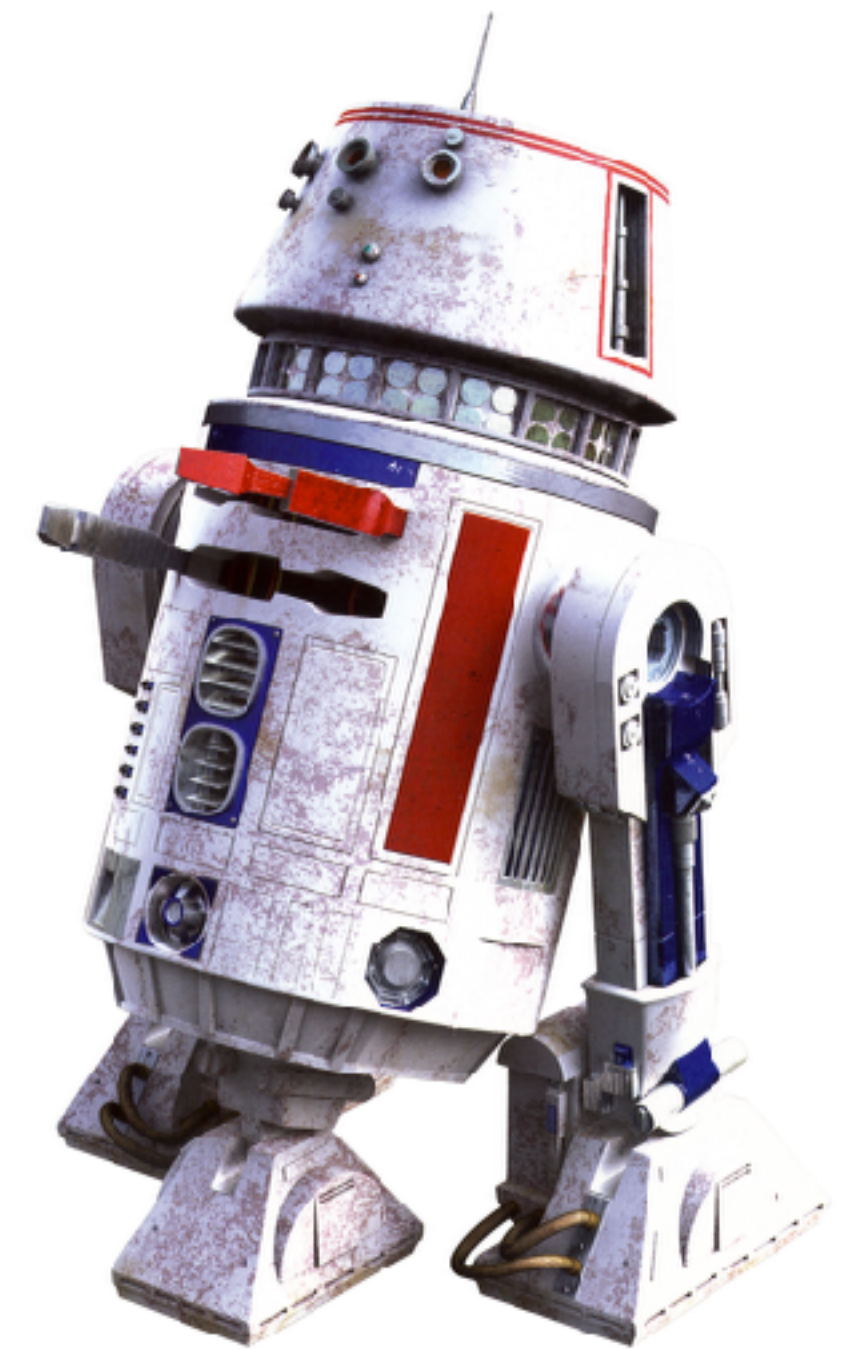




1 day = 5s  
2 days = 10s

3 days = 15s

1 day = 1s  
2 days = 4s



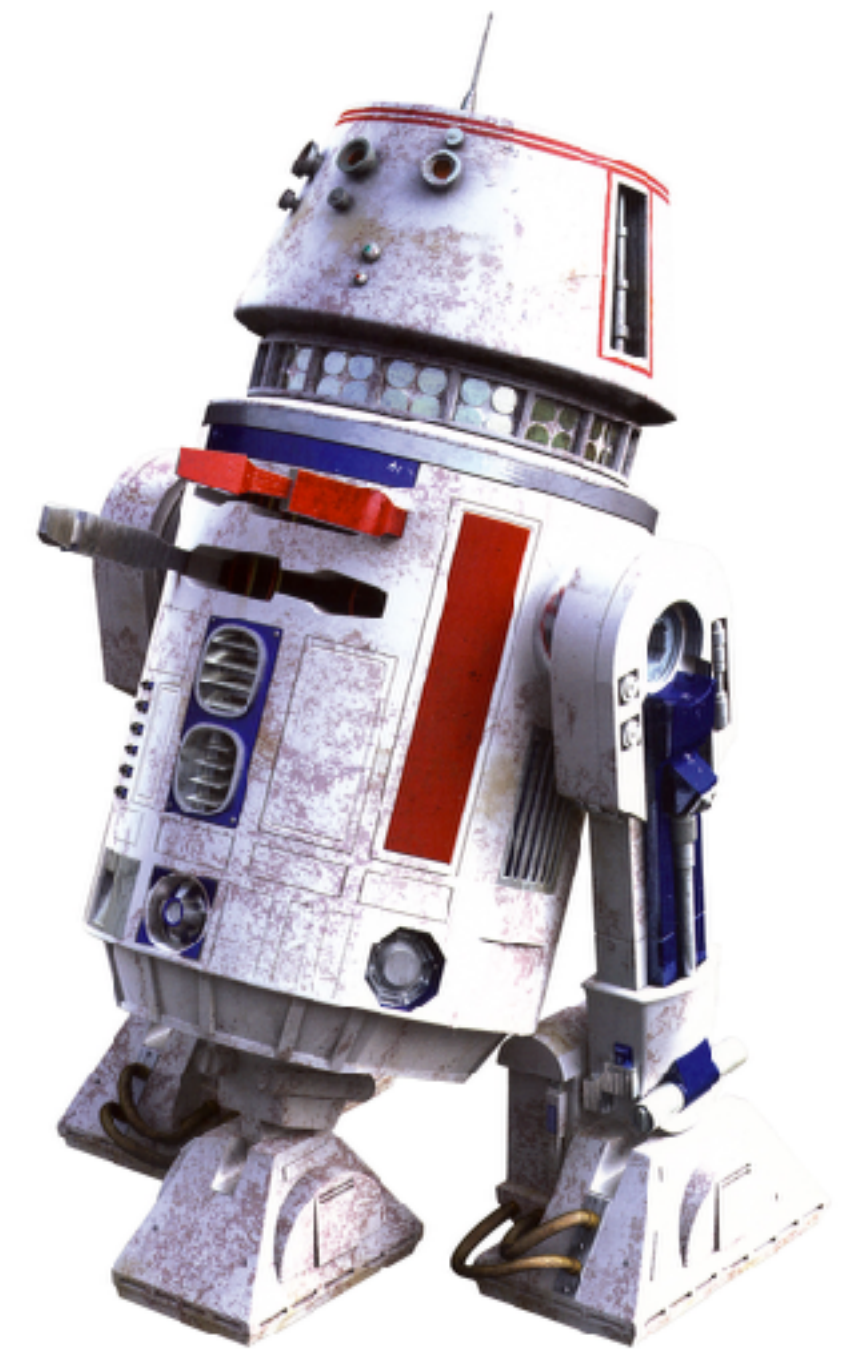
3 days = 9s





1 day = 5s  
2 days = 10s  
3 days = 15s

1 day = 1s  
2 days = 4s  
3 days = 9s





1 day = 5s

2 days = 10s

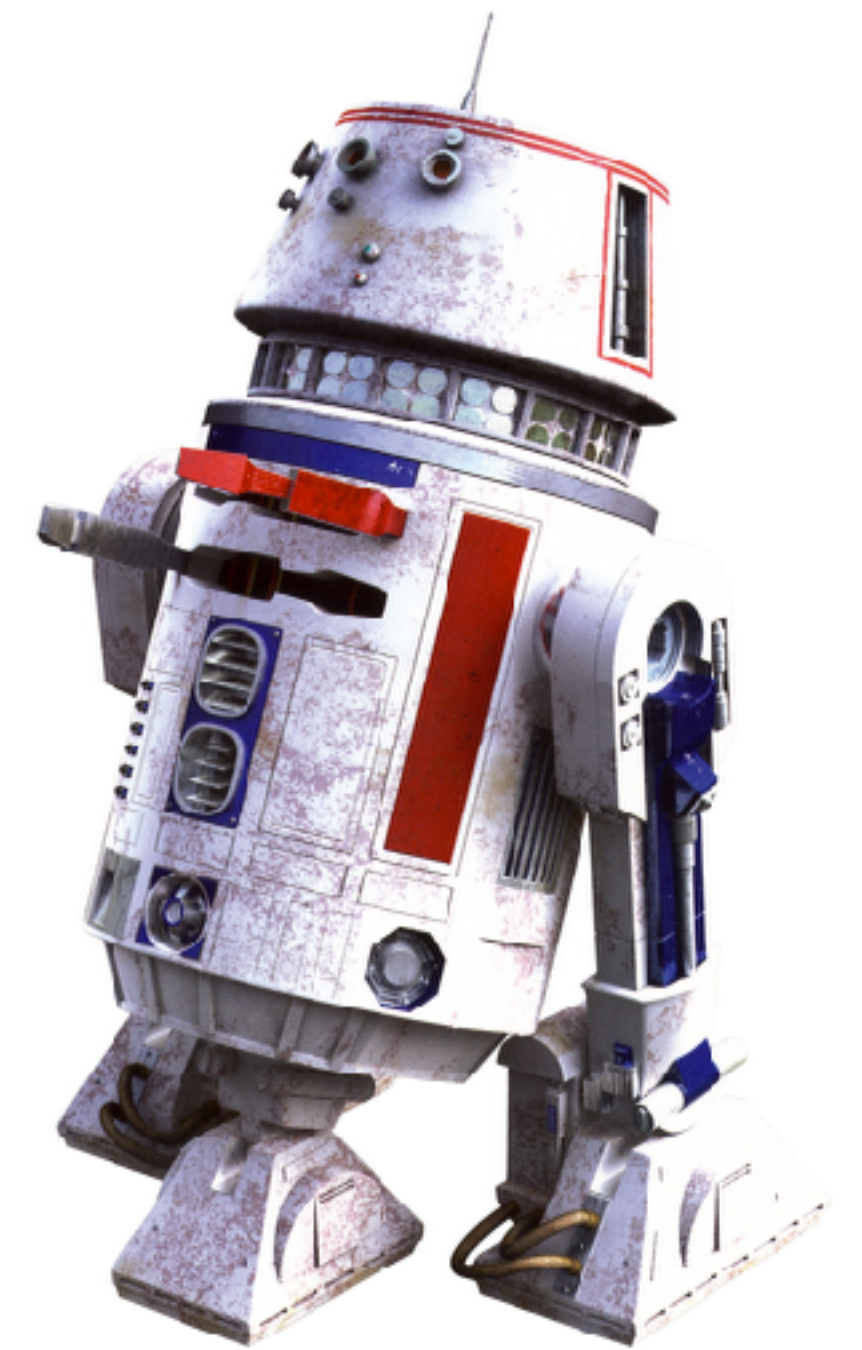
3 days = 15s

4 days = 20s

1 day = 1s

2 days = 4s

3 days = 9s



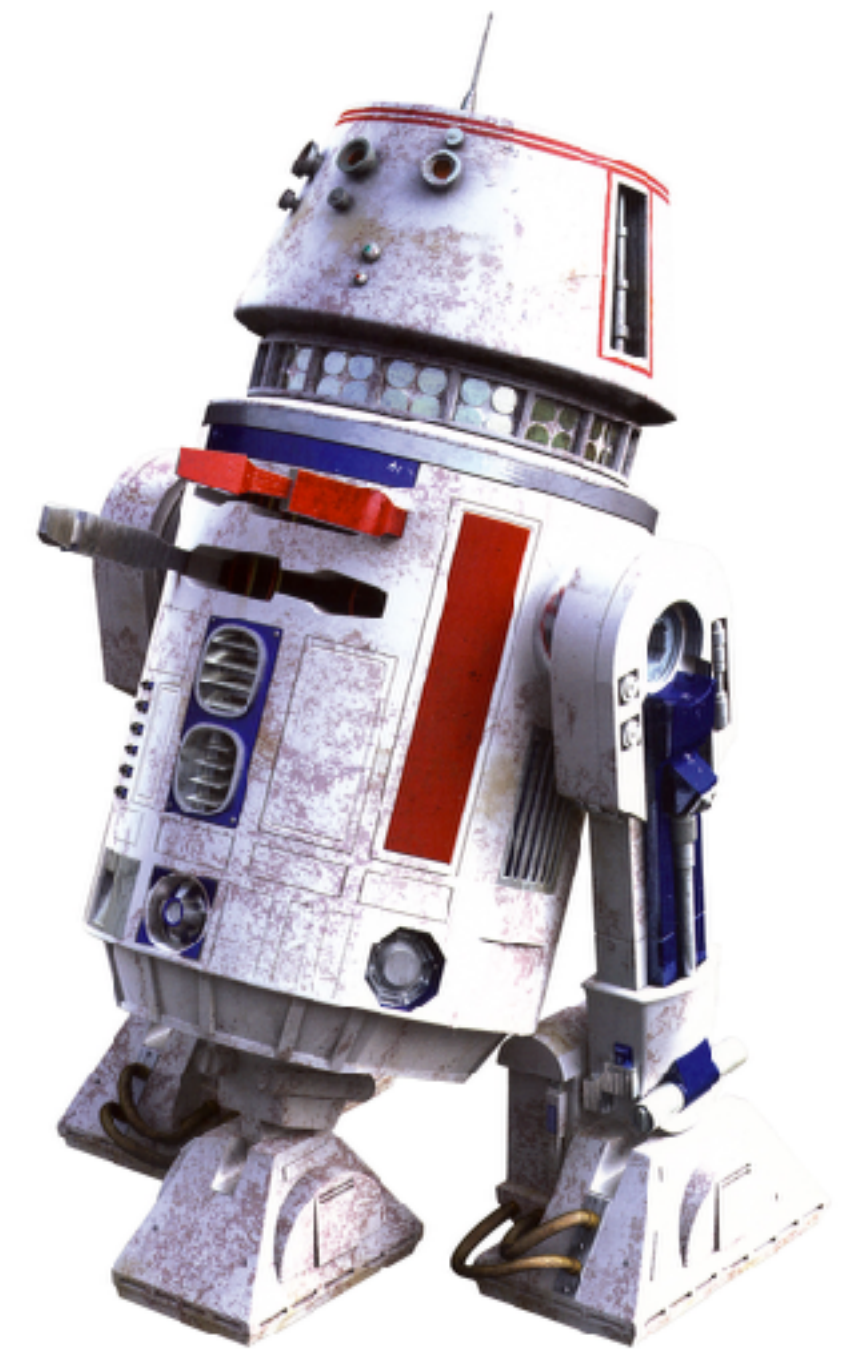
4 days = 16s





1 day = 5s  
2 days = 10s  
3 days = 15s  
4 days = 20s

1 day = 1s  
2 days = 4s  
3 days = 9s  
4 days = 16s



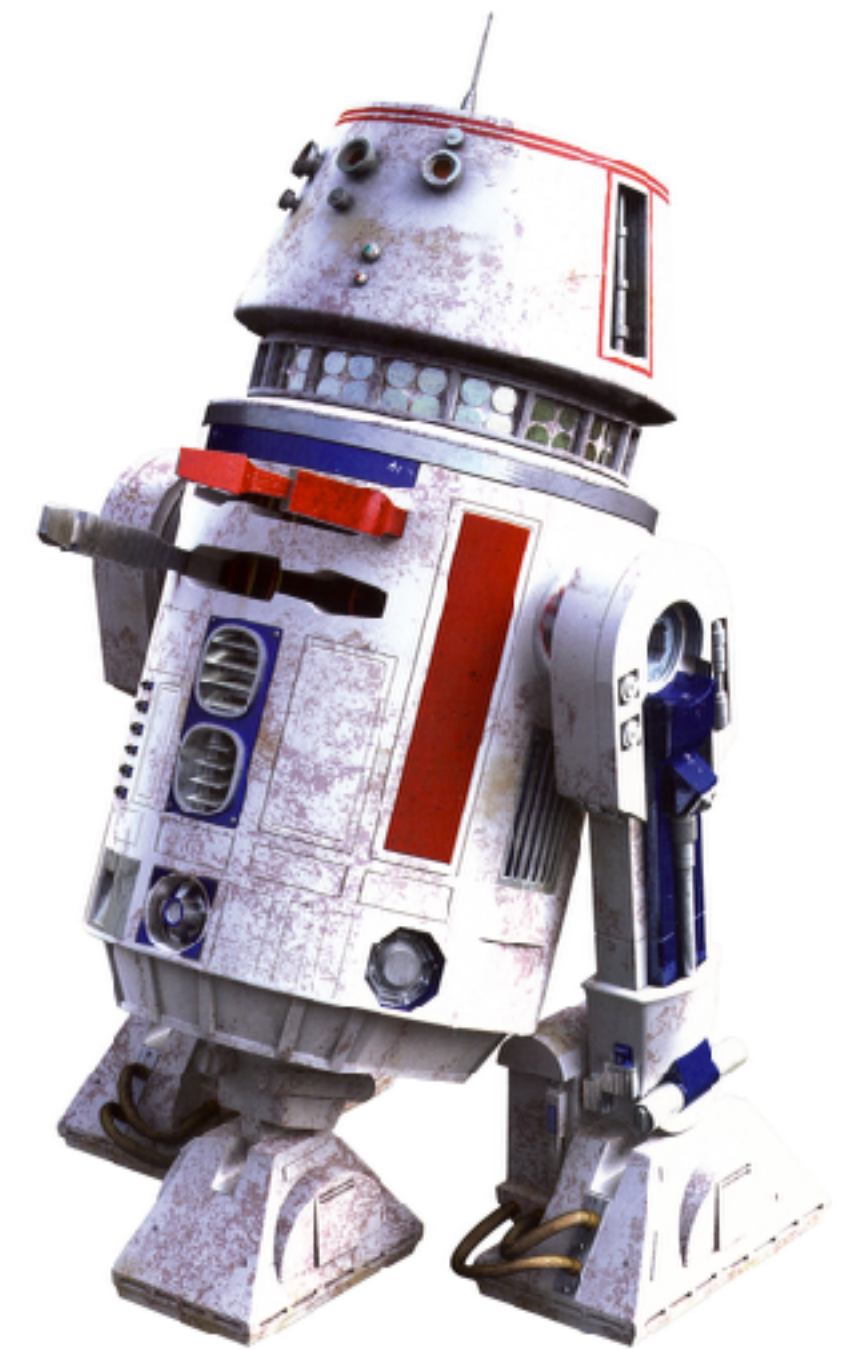




1 day = 5s  
2 days = 10s  
3 days = 15s  
4 days = 20s

5 days = 25s

1 day = 1s  
2 days = 4s  
3 days = 9s  
4 days = 16s

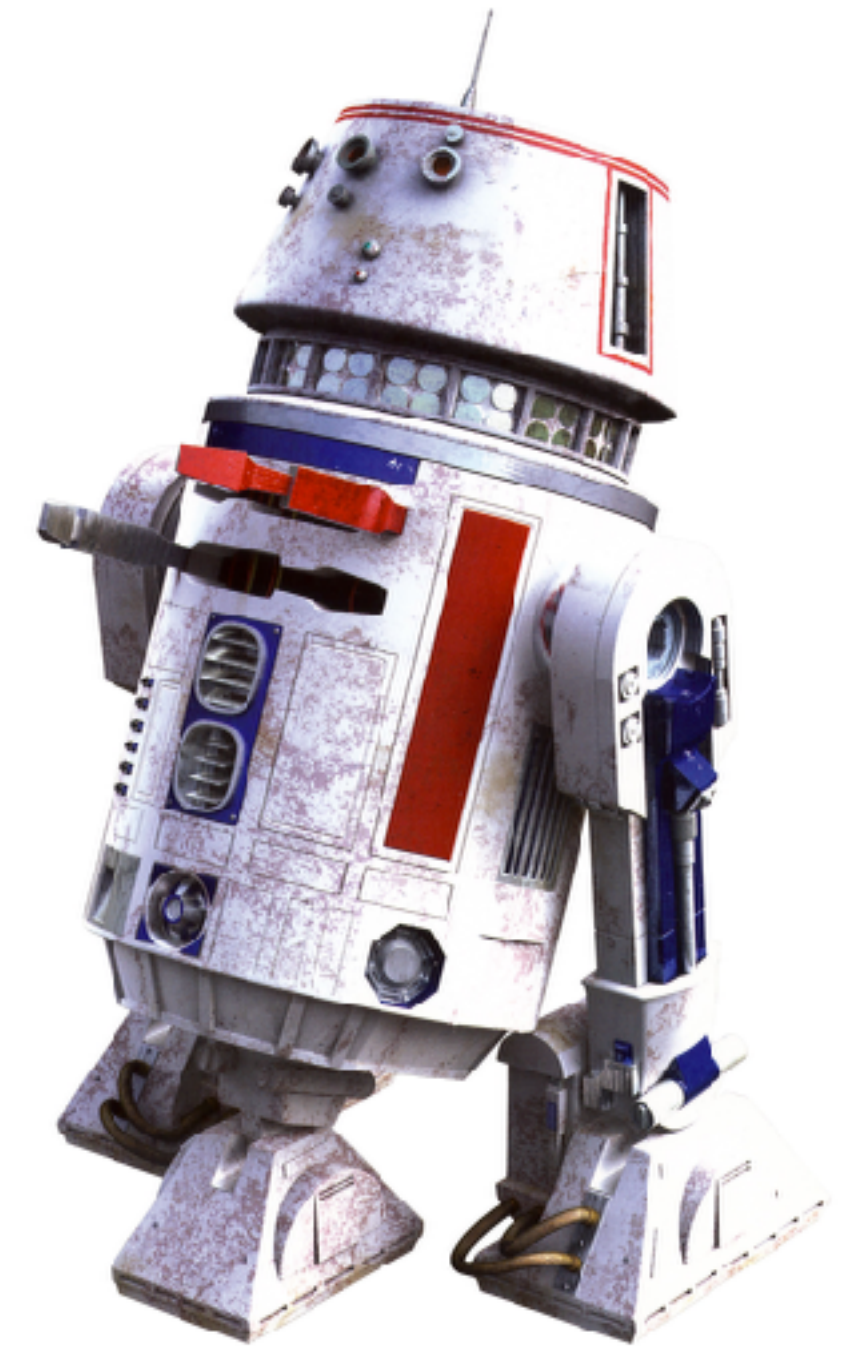


5 days = 25s



1 day = 5s  
2 days = 10s  
3 days = 15s  
4 days = 20s  
5 days = 25s

1 day = 1s  
2 days = 4s  
3 days = 9s  
4 days = 16s  
5 days = 25s







1 day = 5s

2 days = 10s

3 days = 15s

4 days = 20s

5 days = 25s

6 days = 30s

1 day = 1s

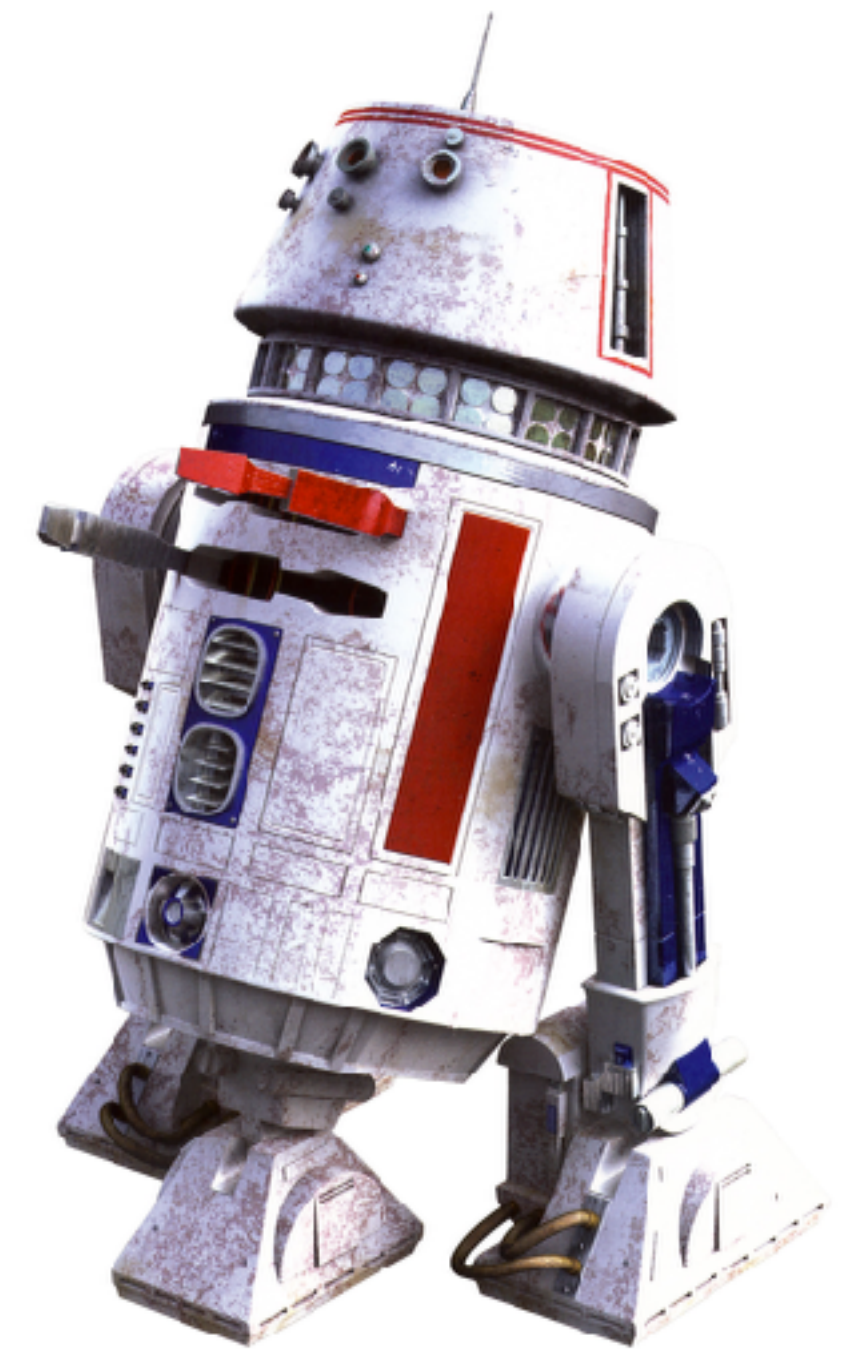
2 days = 4s

3 days = 9s

4 days = 16s

5 days = 25s

6 days = 36s

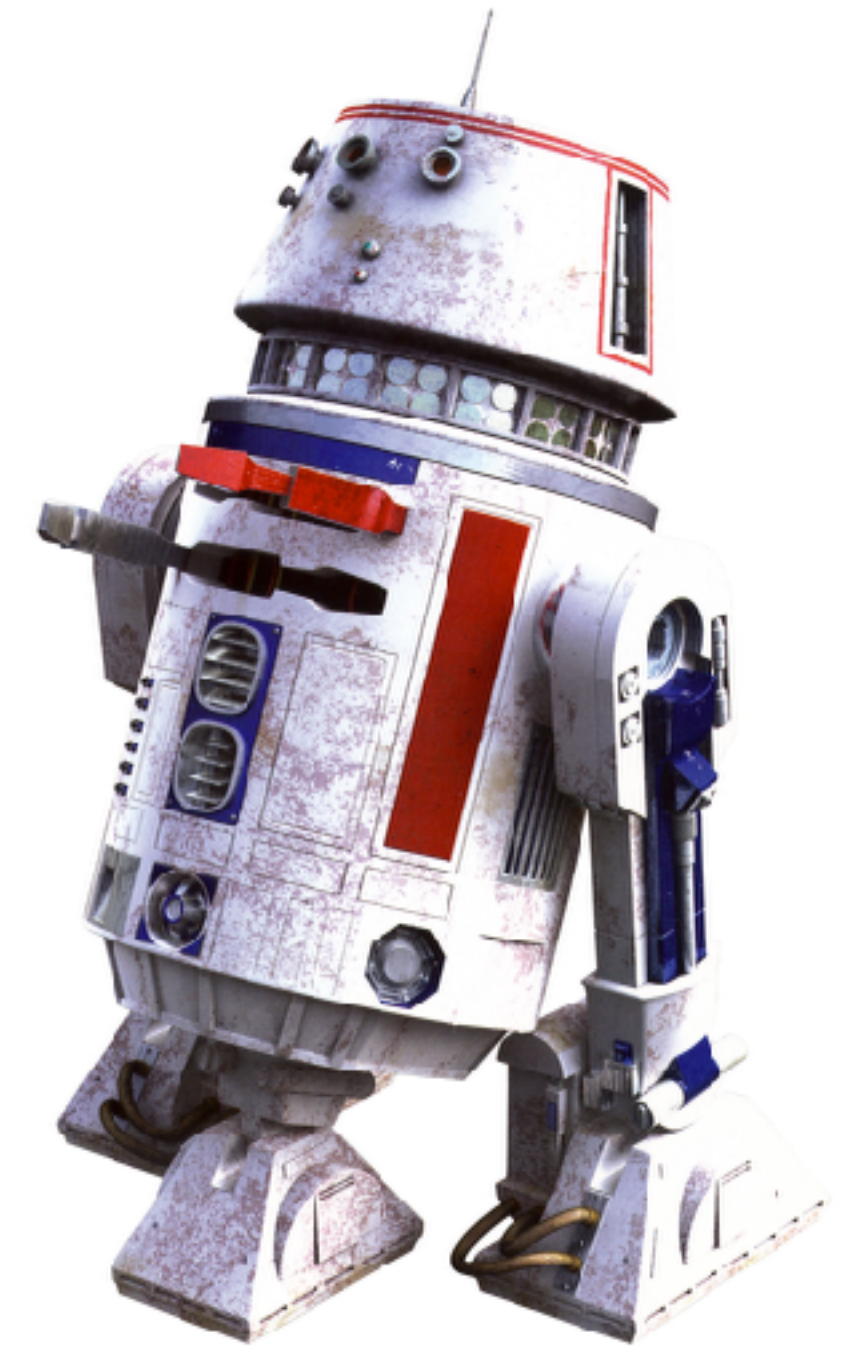






1 day = 5s  
2 days = 10s  
3 days = 15s  
4 days = 20s  
5 days = 25s  
6 days = 30s

1 day = 1s  
2 days = 4s  
3 days = 9s  
4 days = 16s  
5 days = 25s  
6 days = 36s





1 day = 5s

2 days = 10s

3 days = 15s

4 days = 20s

5 days = 25s

6 days = 30s

**10 days = 50s**

1 day = 1s

2 days = 4s

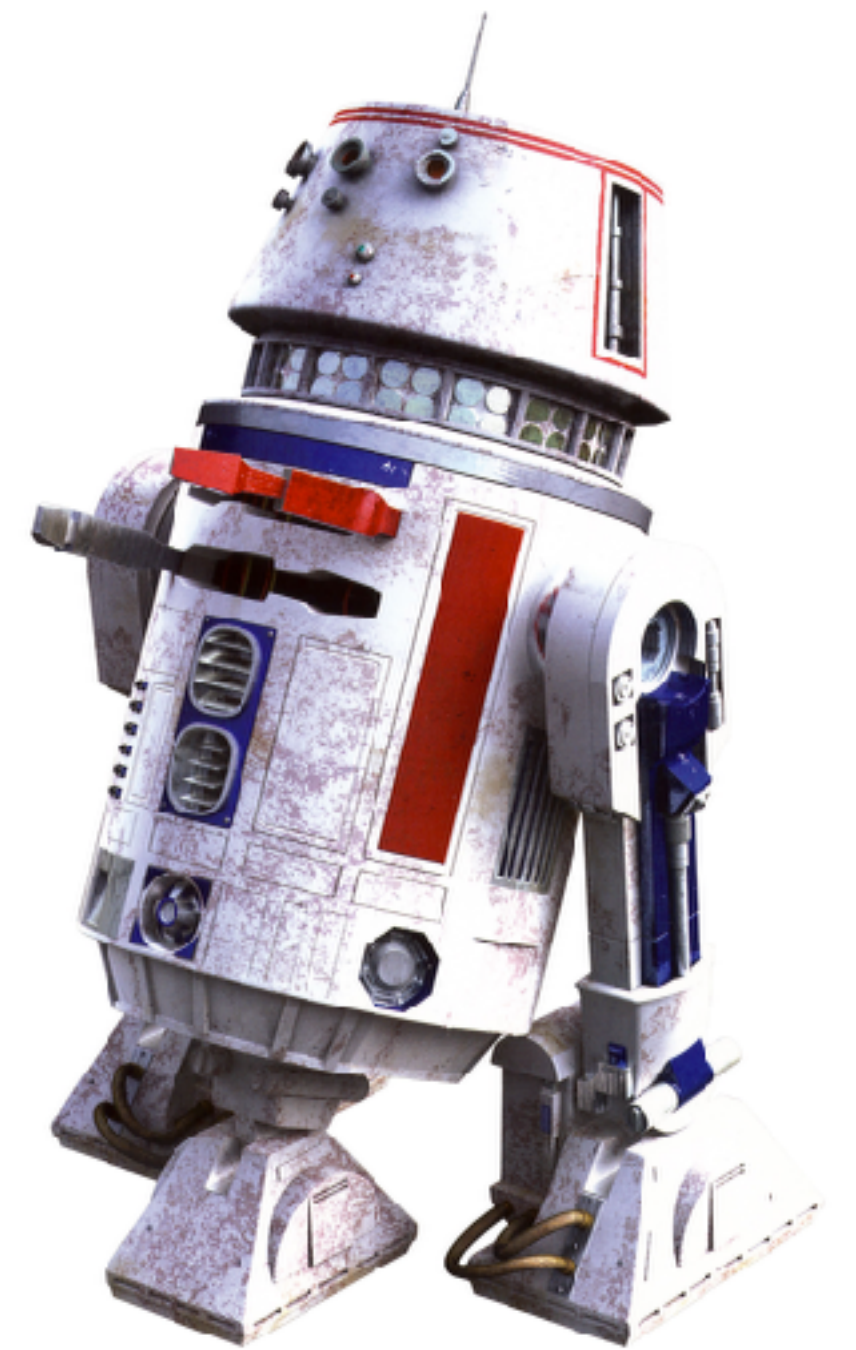
3 days = 9s

4 days = 16s

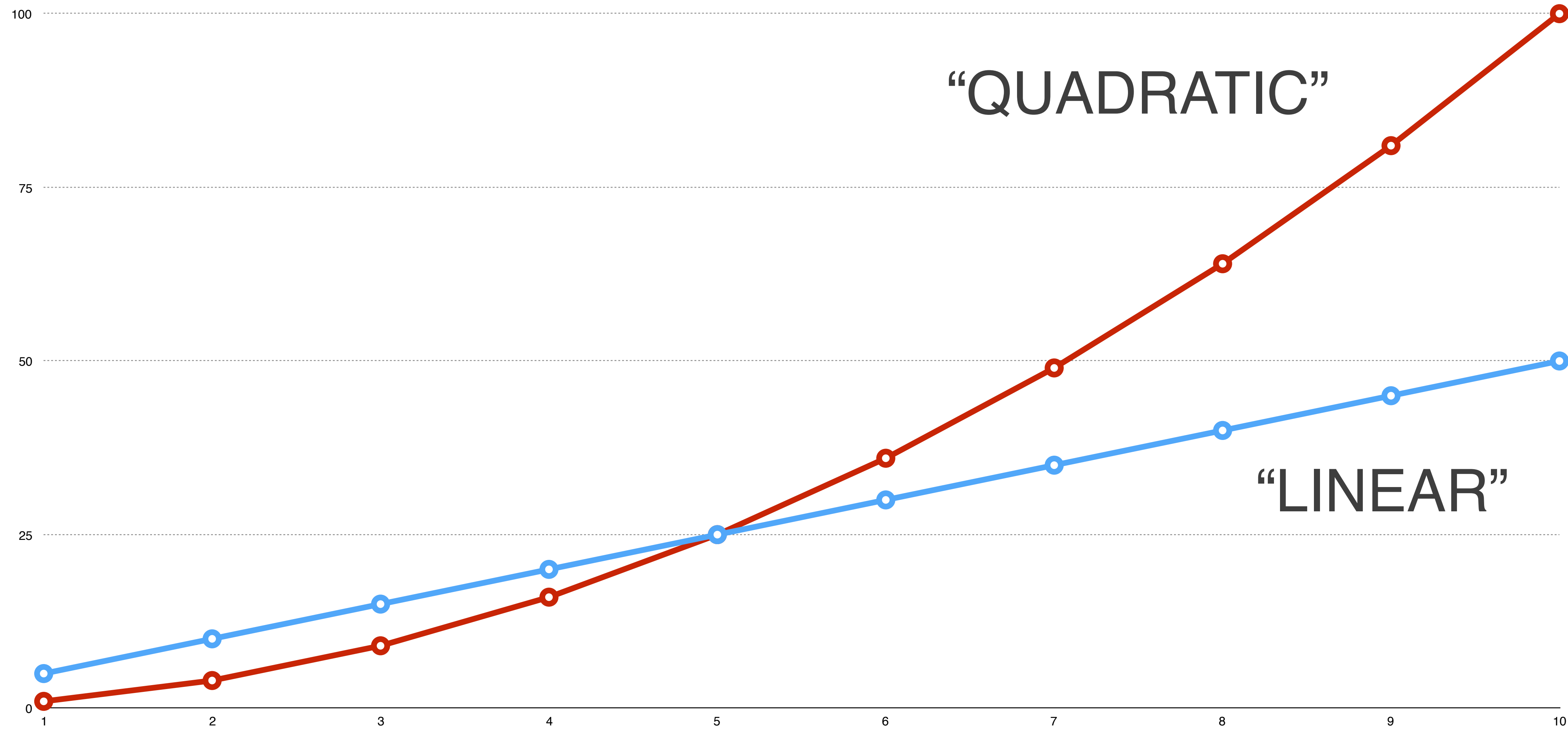
5 days = 25s

6 days = 36s

**10 days = 100s**









# BIG O

**THE ANALYSIS OF HOW MUCH TIME  
(OR SPACE) AN OPERATION TAKES UP,  
RELATIVE TO ITS INPUT, AS THAT  
INPUT GETS BIGGER AND BIGGER**

# BIG O

- Concerns *growth*, not *actual space or time*
- Focus on the “limiting behavior” (the behavior as input gets bigger and bigger)

# Example 1



```
function example (array) {  
  console.log(array.length)  
  let someNumber = 4  
  someNumber += array.length  
  return someNumber  
}
```

```
function example (array) {  
  console.log(array.length) // 1  
  let someNumber = 4  
  someNumber += array.length  
  return someNumber  
}
```

```
function example (array) {  
  console.log(array.length) // 1  
  let someNumber = 4 // 1  
  someNumber += array.length  
  return someNumber  
}
```

```
function example (array) {  
  console.log(array.length) // 1  
  let someNumber = 4 // 1  
  someNumber += array.length // 1  
  return someNumber  
}
```



```
function example (array) {  
  console.log(array.length) // 1  
  let someNumber = 4 // 1  
  someNumber += array.length // 1  
  return someNumber // 1  
}
```

```
function example (array) {  
  console.log(array.length) // 1  
  let someNumber = 4 // 1  
  someNumber += array.length // 1  
  return someNumber // 1  
}  
// 0(1 + 1 + 1 + 1) = 0(4) = 0(1)
```

# Example 2



```
// re-naming the array 'n'
function example (n) {
  const len = n.length
  let sum = 0

  for (let i = 0; i < len; i++) {
    sum += n[i]
  }

  return sum
}
```

```
// re-naming the array 'n'
function example (n) {
  const len = n.length           // 1
  let sum = 0                     // 1

  for (let i = 0; i < len; i++) {
    sum += n[i]
  }

  return sum                      // 1
}
```

```
// re-naming the array 'n'
function example (n) {
  const len = n.length           // 1
  let sum = 0                     // 1

  for (let i = 0; i < len; i++) { // n
    sum += n[i]                   // 1
  }

  return sum                      // 1
}
```



```

// re-naming the array 'n'
function example (n) {
  const len = n.length           // 1
  let sum = 0                     // 1

  for (let i = 0; i < len; i++) { // n
    sum += n[i]                  // 1
  }

  return sum                      // 1
}
// 0(1 + 1 + (n * 1) + 1) = 0(3 + n) = 0(n)

```

# Example 3

```
function example (n) {  
  const len = n.length  
  
  for (let i = 0; i < len; i++) {  
    console.log(n[i])  
  }  
  
  for (let j = 0; j < len; j++) {  
    if (n[i] > 5) {  
      console.log(n[i])  
    }  
  }  
  
  return len  
}
```



```
function example (n) {  
  const len = n.length // 1  
  
  for (let i = 0; i < len; i++) {  
    console.log(n[i])  
  }  
  
  for (let j = 0; j < len; j++) {  
    if (n[i] > 5) {  
      console.log(n[i])  
    }  
  }  
  
  return len // 1  
}
```

```
function example (n) {  
  const len = n.length // 1  
  
  for (let i = 0; i < len; i++) { // n  
    console.log(n[i]) // 1  
  }  
  
  for (let j = 0; j < len; j++) { // n  
    if (n[i] > 5) { // assume this always runs  
      console.log(n[i]) // 1  
    }  
  }  
  
  return len // 1  
}
```

```

function example (n) {
  const len = n.length // 1

  for (let i = 0; i < len; i++) { // n
    console.log(n[i]) // 1
  }

  for (let j = 0; j < len; j++) { // n
    if (n[i] > 5) { // assume this always runs
      console.log(n[i]) // 1
    }
  }

  return len // 1
}
// 0(1 + (n * 1) + (n * 1) + 1) = 0(2 + 2n) = 0(2n) = 0(n)

```

# Example 4



```
function example (n) {  
  for (let i = 0; i < n.length; i++) {  
    for (let j = 0; j < n.length; j++) {  
      console.log(n[i] + n[j])  
    }  
  }  
}
```

```
function example (n) {  
  for (let i = 0; i < n.length; i++) { // n  
    for (let j = 0; j < n.length; j++) {  
      console.log(n[i] + n[j])  
    }  
  }  
}
```

```
function example (n) {  
  for (let i = 0; i < n.length; i++) { // n  
    for (let j = 0; j < n.length; j++) { // n  
      console.log(n[i] + n[j])  
    }  
  }  
}
```

```
function example (n) {  
  for (let i = 0; i < n.length; i++) { // n  
    for (let j = 0; j < n.length; j++) { // n  
      console.log(n[i] + n[j])          // 1  
    }  
  }  
}
```



```
function example (n) {  
  for (let i = 0; i < n.length; i++) { // n  
    for (let j = 0; j < n.length; j++) { // n  
      console.log(n[i] + n[j]) // 1  
    }  
  }  
}  
// O((n * (n * 1)) = O(n^2)
```

# Example 5

```
// now, n is a number
function example (n) {
  let counter = 0

  while (n > 1) {
    n = n / 2
    counter++
  }

  return counter
}
```

```
// now, n is a number
function example (n) {
  let counter = 0 // 1

  while (n > 1) {
    n = n / 2
    counter++
  }

  return counter // 1
}
```



```
// now, n is a number
function example (n) {
  let counter = 0 // 1

  while (n > 1) { // ?
    n = n / 2
    counter++
  }

  return counter // 1
}
```

```
// now, n is a number
function example (n) {
  let counter = 0 // 1

  while (n > 1) { // log(n)
    n = n / 2
    counter++
  }

  return counter // 1
}
```

```
// now, n is a number
function example (n) {
    let counter = 0 // 1

    while (n > 1) { // log(n)
        n = n / 2
        counter++
    }

    return counter // 1
}
//  $O(2 + \log(n)) = O(\log(n))$ 
```

# Quick review of logarithms

Logarithms are just the opposite of exponents

$$\log_2(n)$$

Read as: *what power do we need to raise 2 to in order to get n?*



$$\log_2(2) = 1$$

$$\log_2(4) = 2$$

$$\log_2(8) = 3$$

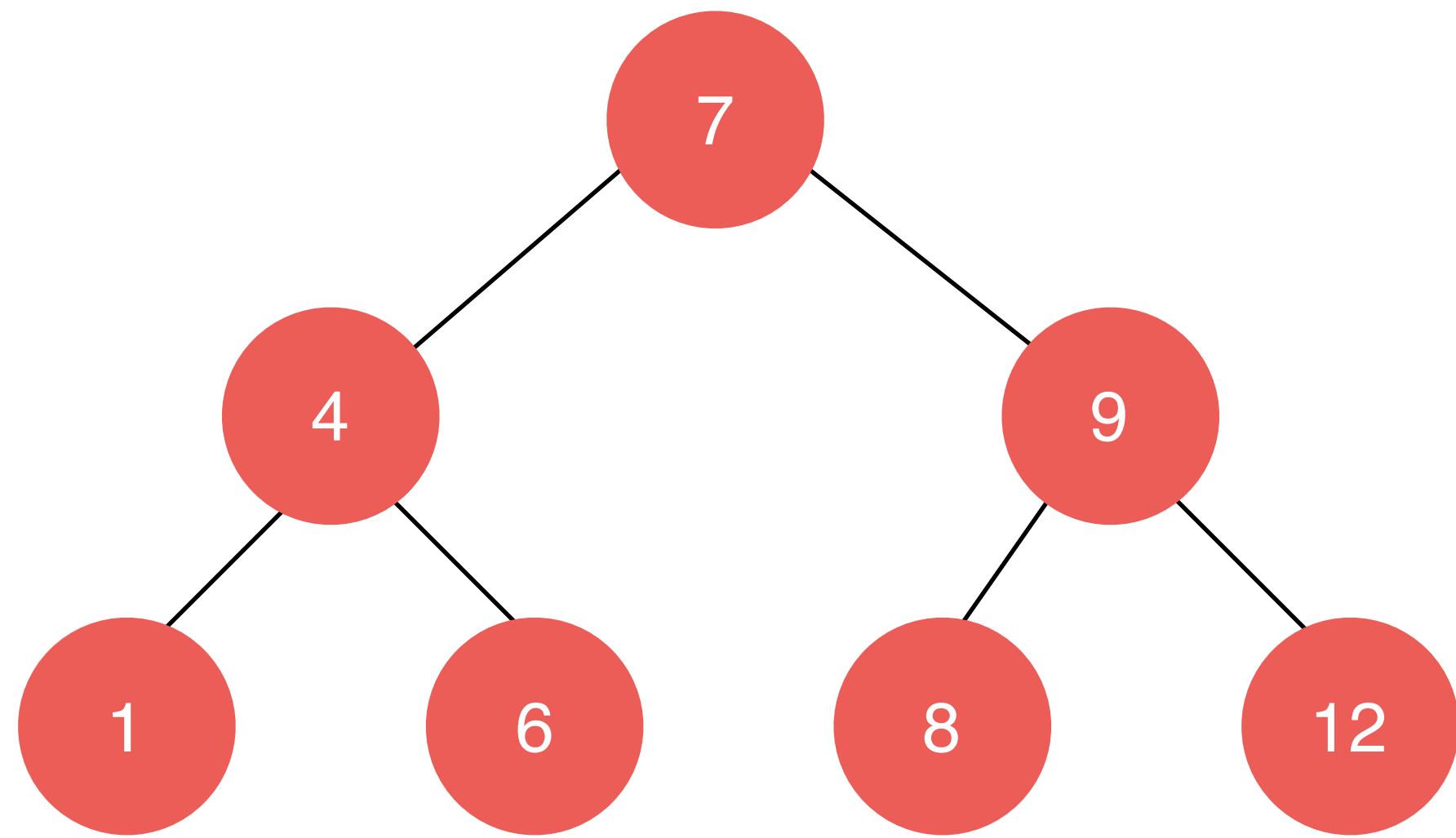
$$\log_2(5) = 2.32192809489$$

# Beyond the Basics

- Recursion
- Space Complexity
- Multivariate algorithms

# Recursion

- It's helpful to think of recursion as a tree
- When there is only one “branch” of recursion, this is usually like a standard “for” loop, where the number of times we recursive corresponds to the input size
- When there are multiple recursive branches, the runtime will often be similar to  $O(\text{branches}^{\text{depth}})$ 
  - Each level of “depth” has “branch” number more calls than the level before - an exponential relationship!



$$n = 7$$

$$\text{branches} = 2$$

$$\text{depth} = 3$$

$$\log_2 7 \approx 3$$

$$\text{depth} \approx \log_2 n$$

$$\text{branches}^{\text{depth}} = 2^{\log_2 n}$$

$$\text{branches}^{\text{depth}} = n$$

```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```



```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```

```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```

fib(4)

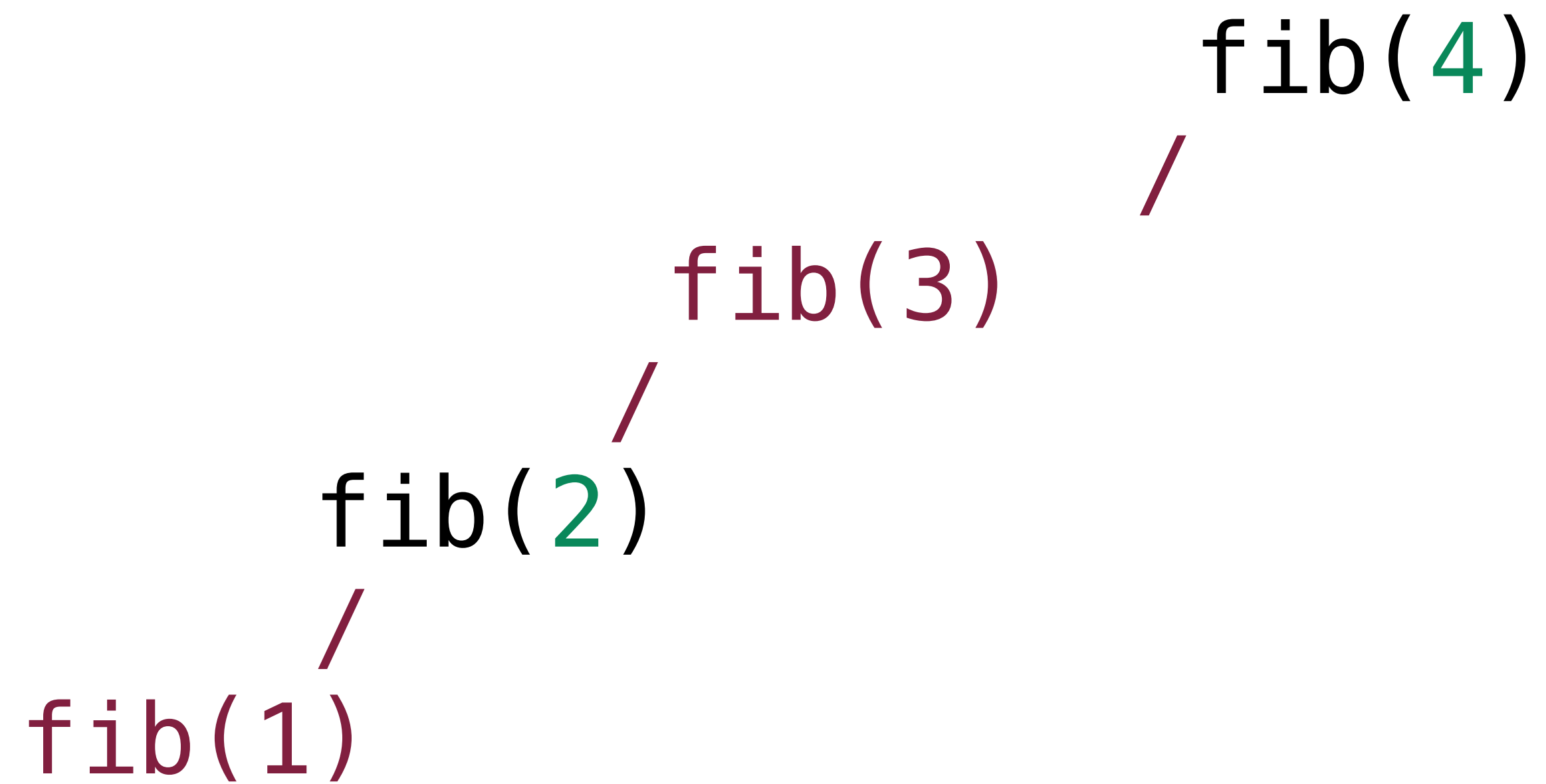
```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```

fib(4)  
/  
fib(3)

```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```

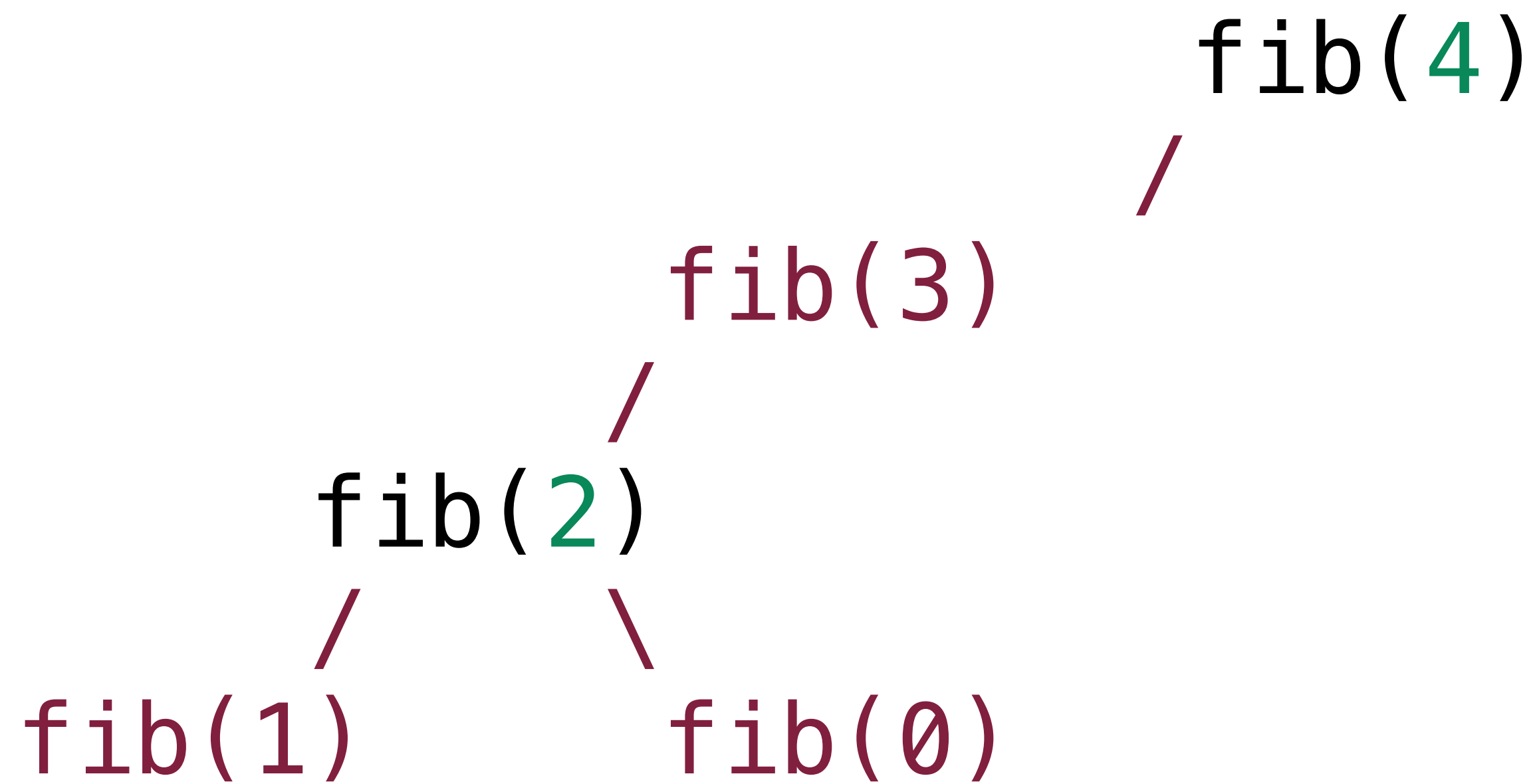
fib(2) / fib(3) / fib(4)

```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```

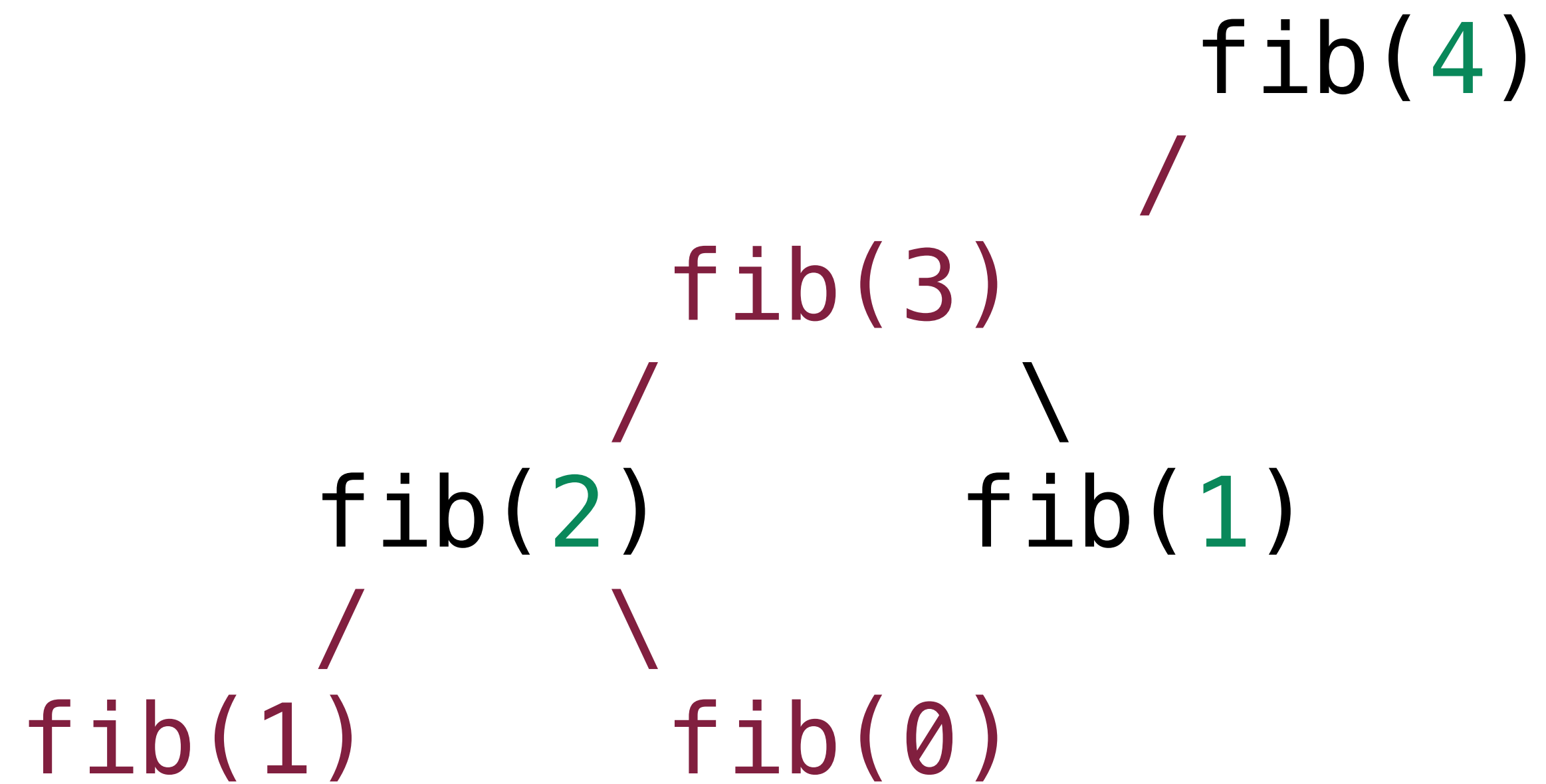




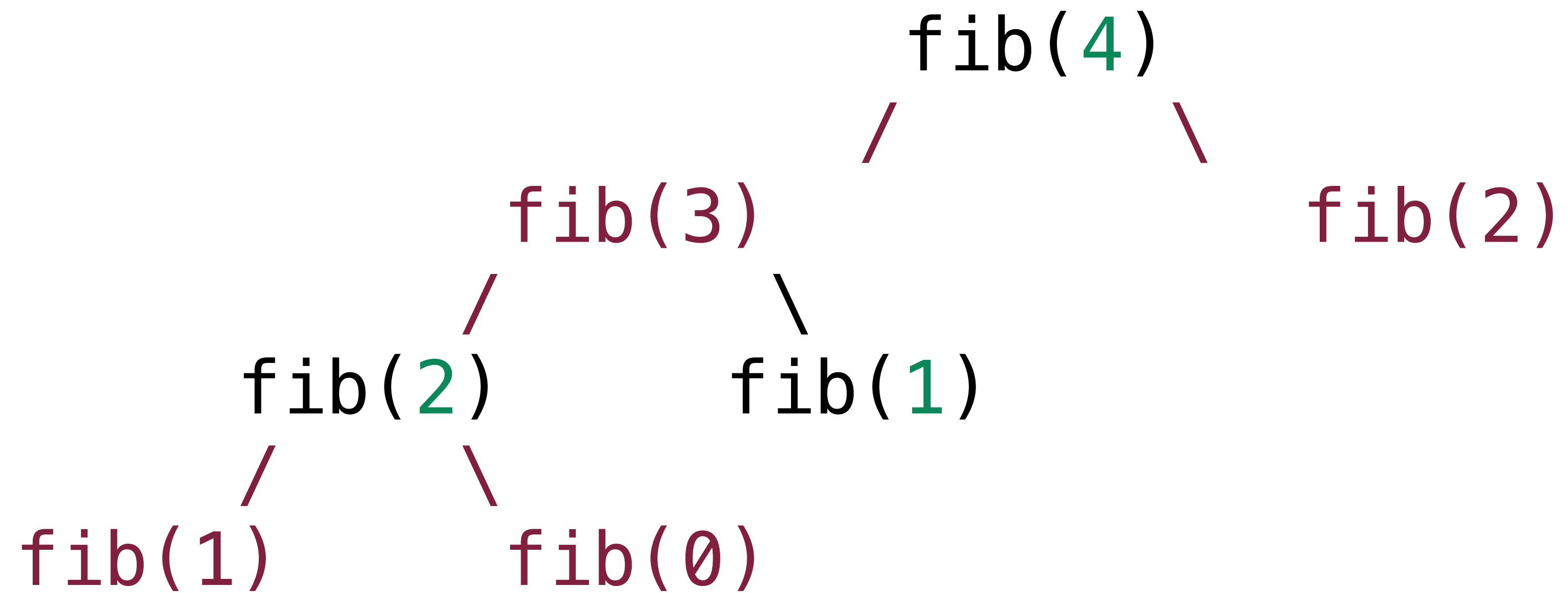
```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```



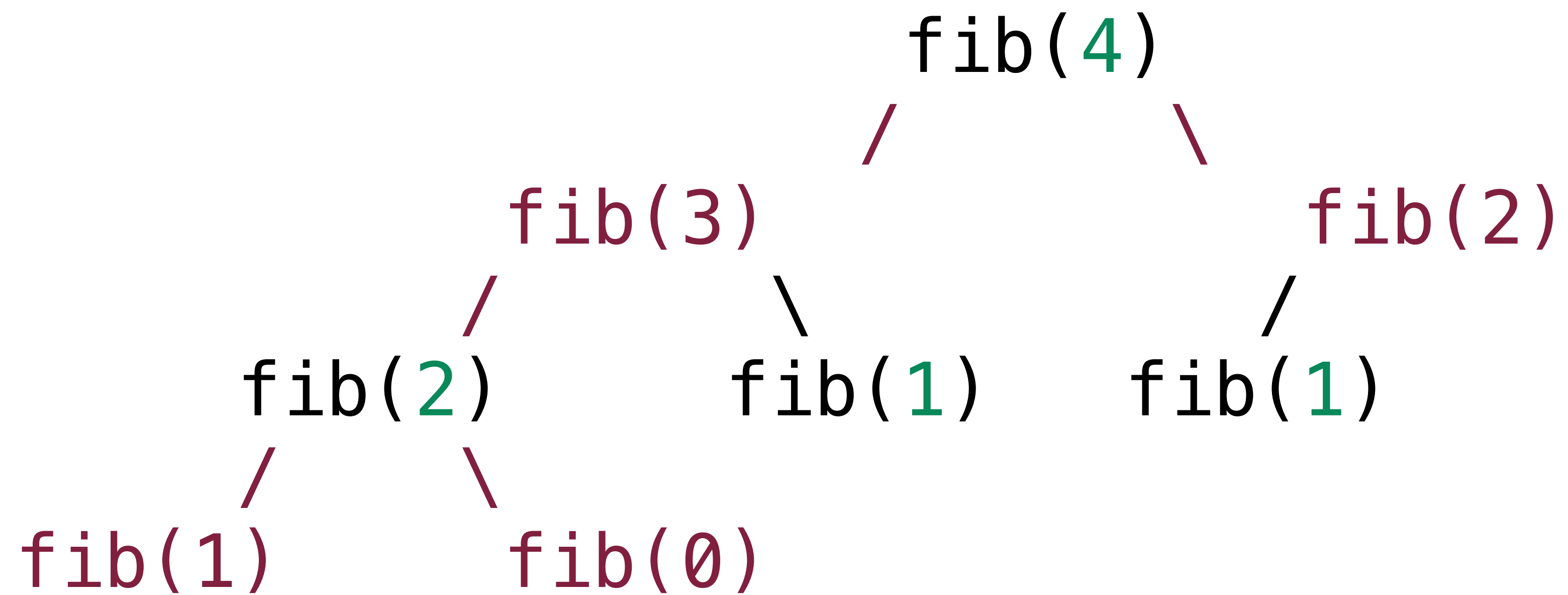
```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```



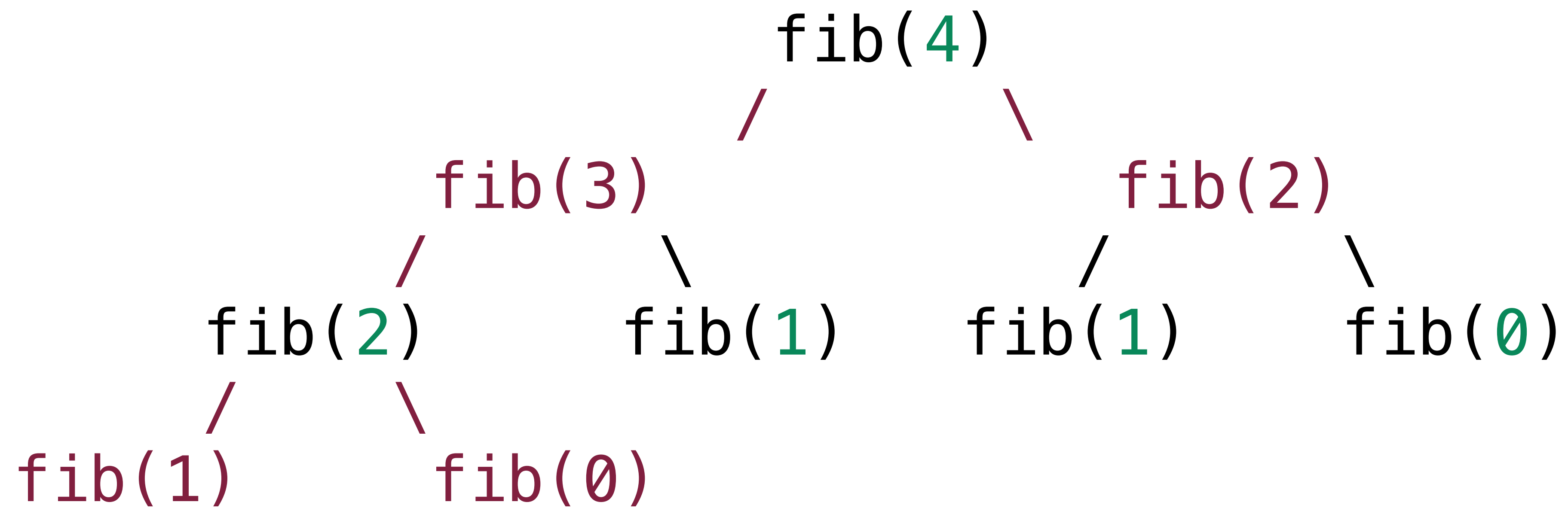
```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```



```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```



```
function fib (n) {  
  if (n === 1 || n === 0) return n;  
  else return fib(n - 1) + fib(n - 2);  
}
```





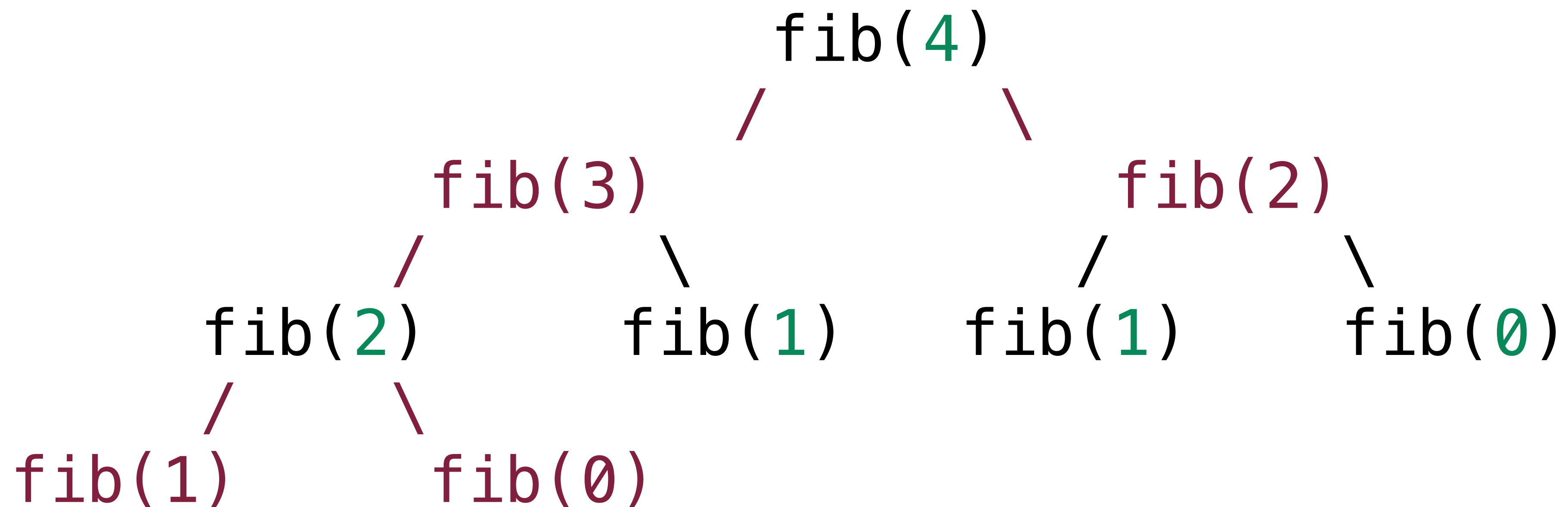
```
function fib (n) {
  if (n === 1 || n === 0) return n;
  else return fib(n - 1) + fib(n - 2);
}
```

our input is equal to 4:  $n = 4$

we go four levels deep, so  $\text{depth} = n$

we branch twice with each recursive call

therefore, runtime is  $O(2^n)$ !



```
function fib (n, memo = {}) {  
  if (n === 1 || n === 0) return n;  
  else if (memo[n]) return memo[n];  
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);  
  return memo[n];  
}
```

```
function fib (n, memo = {}) {  
  if (n === 1 || n === 0) return n;  
  else if (memo[n]) return memo[n];  
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);  
  return memo[n];  
}
```

```
function fib (n, memo = {}) {  
  if (n === 1 || n === 0) return n;  
  else if (memo[n]) return memo[n];  
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);  
  return memo[n];  
}
```

```
memo = {  
  
}
```

fib(4)

```
function fib (n, memo = {}) {  
  if (n === 1 || n === 0) return n;  
  else if (memo[n]) return memo[n];  
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);  
  return memo[n];  
}
```

```
memo = {  
  
}
```

fib(4)  
/  
fib(3)



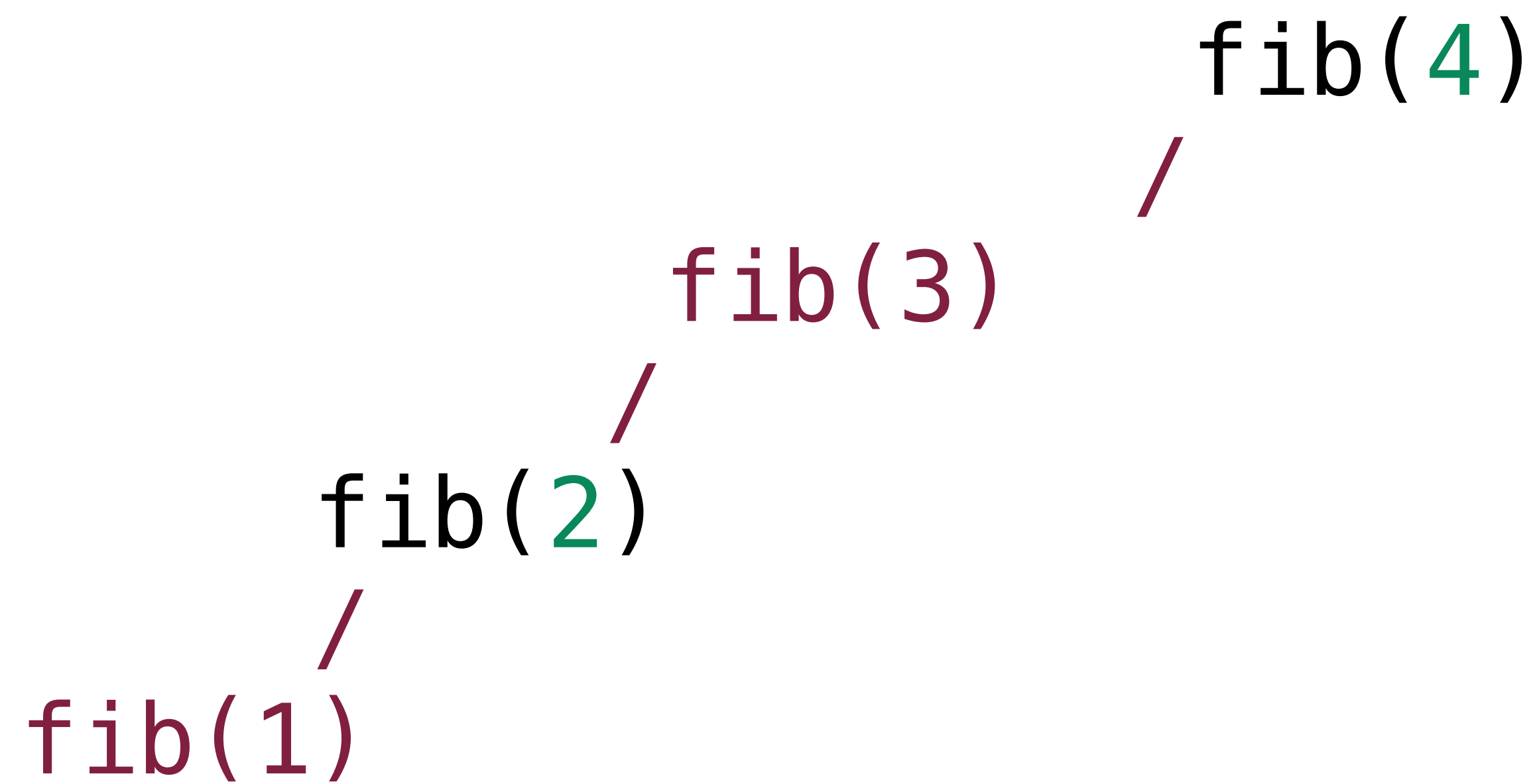
```
function fib (n, memo = {}) {  
  if (n === 1 || n === 0) return n;  
  else if (memo[n]) return memo[n];  
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);  
  return memo[n];  
}
```

```
memo = {  
  
}
```

fib(4)  
/  
fib(3)  
/  
fib(2)

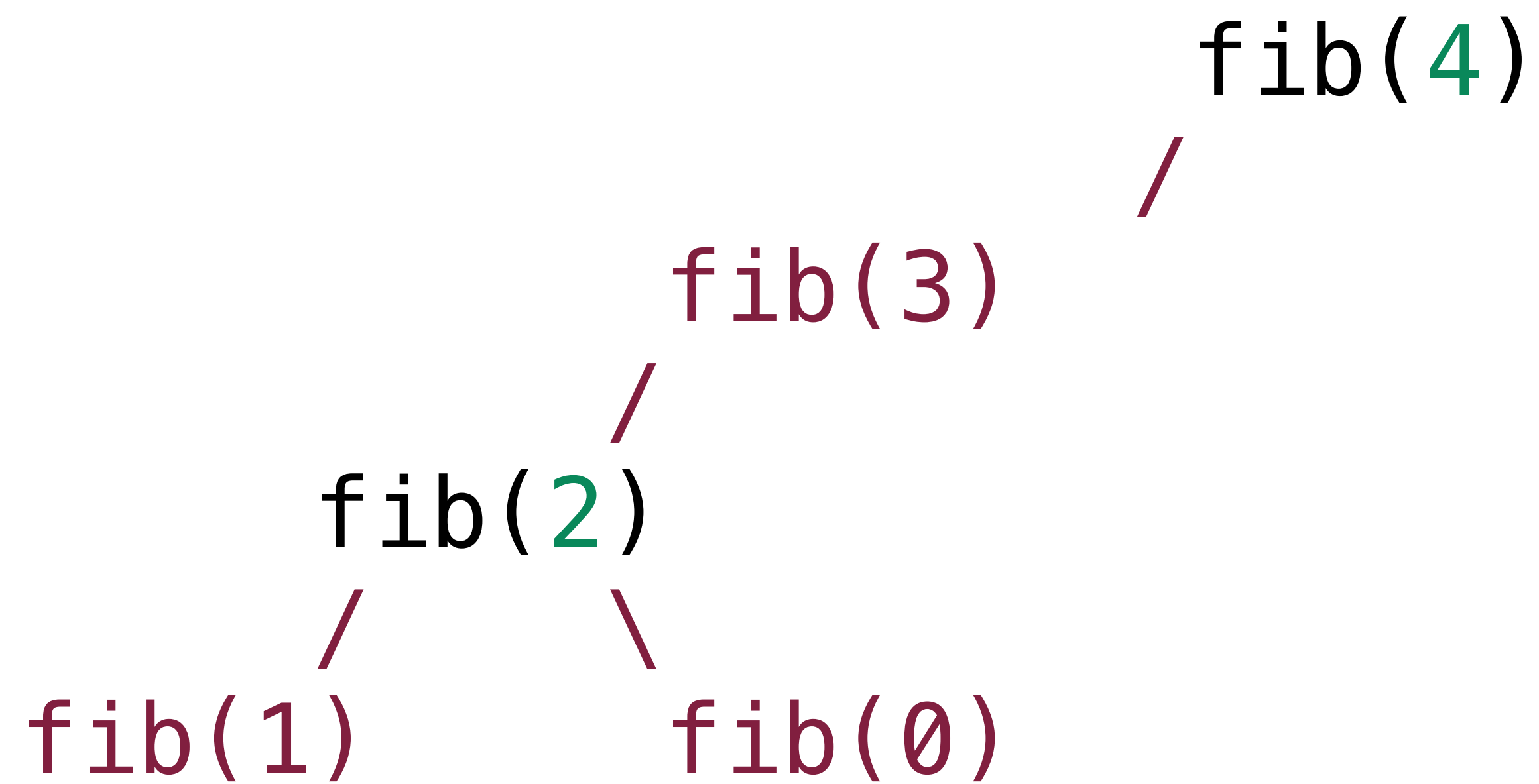
```
function fib (n, memo = {}) {
  if (n === 1 || n === 0) return n;
  else if (memo[n]) return memo[n];
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);
  return memo[n];
}
```

```
memo = {
  1: 1
}
```



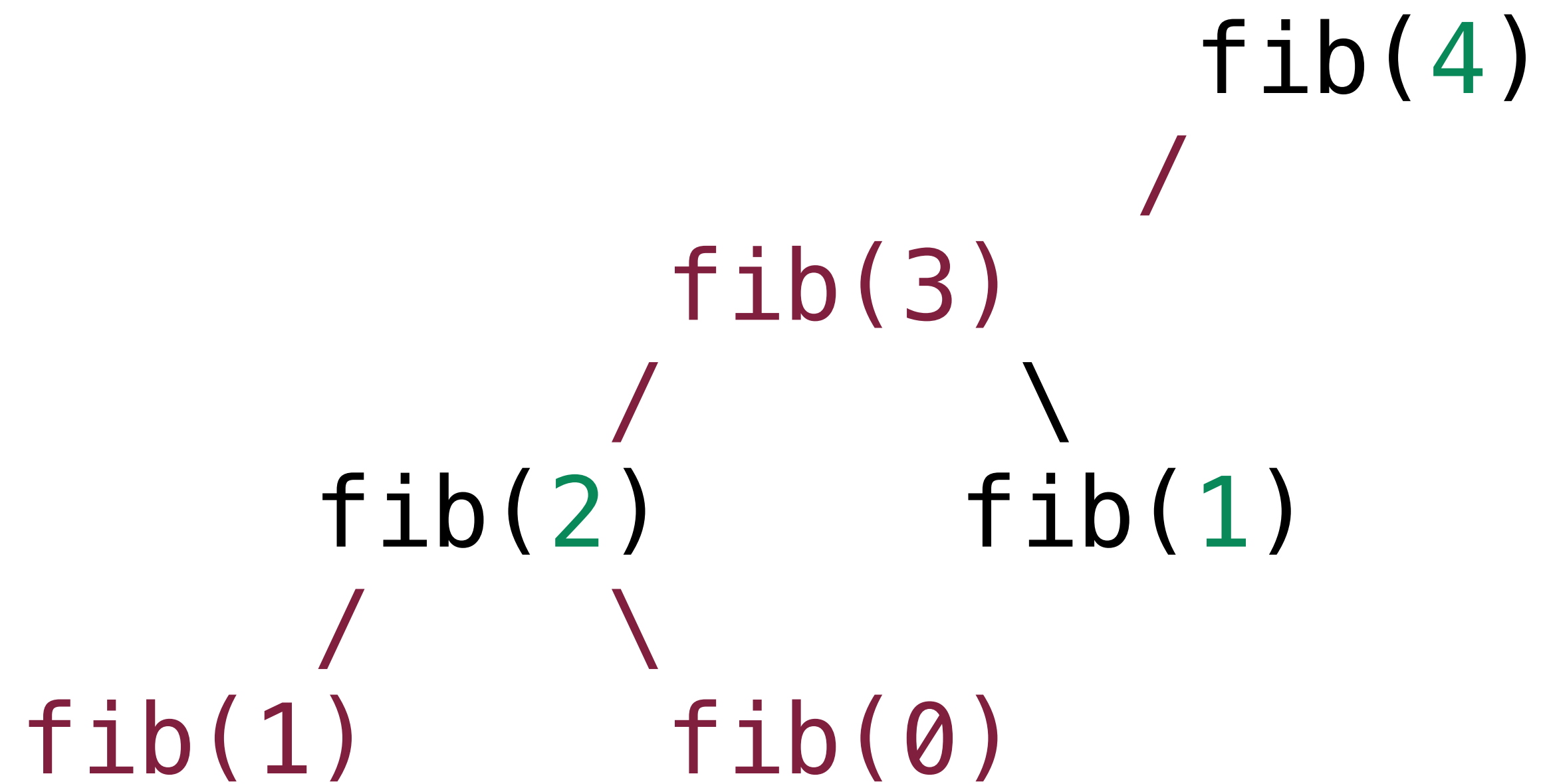
```
function fib (n, memo = {}) {
  if (n === 1 || n === 0) return n;
  else if (memo[n]) return memo[n];
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);
  return memo[n];
}
```

```
memo = {
  0: 0,
  1: 1,
  2: 1,
}
```



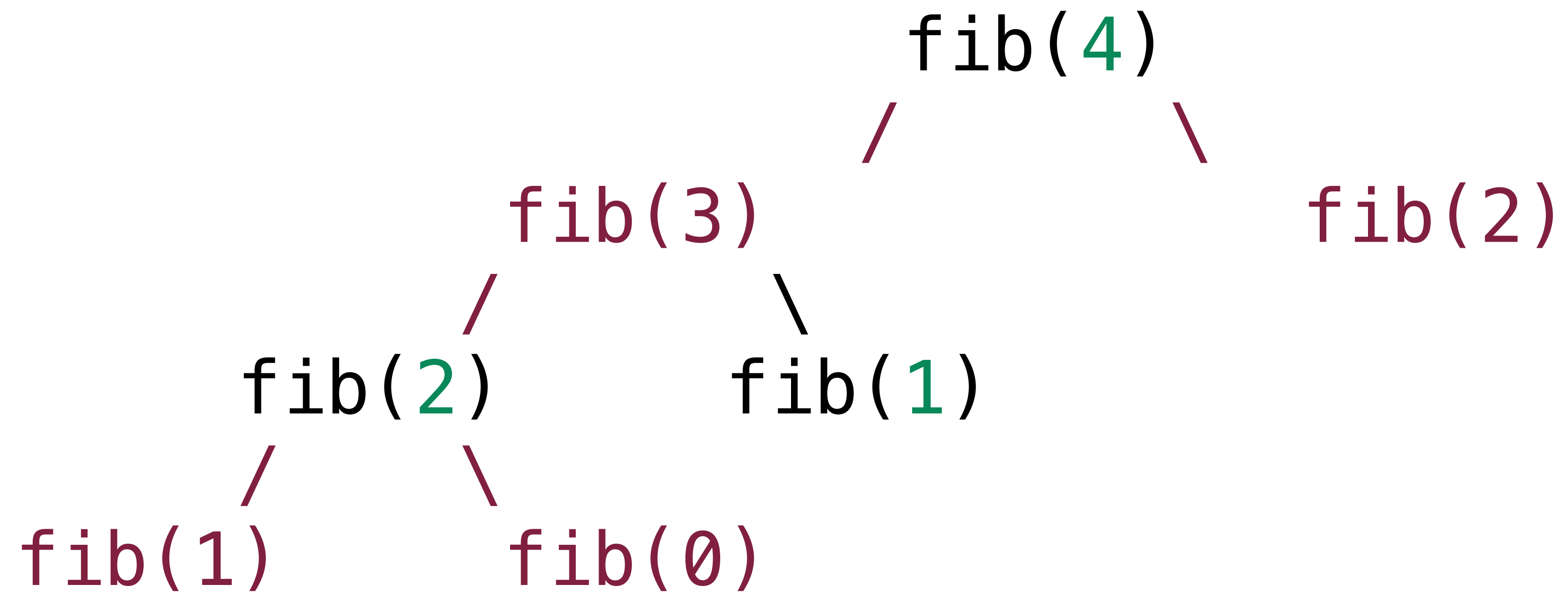
```
function fib (n, memo = {}) {
  if (n === 1 || n === 0) return n;
  else if (memo[n]) return memo[n];
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);
  return memo[n];
}
```

```
memo = {
  0: 0,
  1: 1,
  2: 1,
  3: 2
}
```



```
function fib (n, memo = {}) {
  if (n === 1 || n === 0) return n;
  else if (memo[n]) return memo[n];
  else memo[n] = fib(n - 1, memo) + fib(n - 2, memo);
  return memo[n];
}
```

```
memo = {
  0: 0,
  1: 1,
  2: 1,
  3: 2
}
```



# Space Complexity

- Big O can also express space complexity
- Measures how much space (i.e. memory) we use relative to the input (ex. by storing values in arrays and hash tables, and simultaneous calls on the call stack
  - Remember: what matters is the growth curve. not the actual number of bytes we store!
- Space can be taken a freed up again - the same can't be said of time!
- Usually, we have enough space...but not enough time!



```
// assume `callback` performs an  $O(1)$  operation
function map (arr, callback) {
  const newArr = []
  for (let i = 0; i < arr.length; i++) {
    newArr.push(callback(arr[i]))
  }
  return newArr
}
```

```
// assume `callback` performs an  $O(1)$  operation
function map (arr, callback) {
  const newArr = []
  for (let i = 0; i < arr.length; i++) {
    newArr.push(callback(arr[i]))
  }
  return newArr
}
```

# Multivariate Algorithms

- What if you have an algorithm that uses another algorithm?  
For example, what if you loop over an array of strings and sort each string?
- Be careful to not to confuse the input and runtime of the “outer” algorithm with the input and runtime of the “inner” algorithm

```
function sortedStrings (arr) {  
  for (let i = 0; i < arr.length; i++) {  
    arr[i].sort();  
  }  
}
```

// Let's say that .sort is  $O(n \log n)$

```
function sortedStrings (arr) {  
  for (let i = 0; i < arr.length; i++) { // 0(n)  
    arr[i].sort(); // 0(s log s)  
  }  
}
```

# Algorithm Analysis: Big O Notation

- A *comparative* way to classify different algorithms
- Based on *shape* of *growth curve* (time vs input size(s))
- For *big enough* inputs
  - Might not be true when  $n$  is small, but who cares when  $n$  is small?
- Establishing an *upper bound* on the time
  - Not worse than this. Might be better, but it ain't worse!
- Including just the *highest order* term
  - In  $f(n) = n^3 + 5n + 3$ , only  $n^3$  matters as  $n$  gets large
- *Ignores constants* (mostly irrelevant;  $0.1 \cdot n^2$  will overtake  $10 \cdot n$ )



$n!$   $2^n$   $n^2$

$n \cdot \log(n)$

$n$

Time for  
Function  
to Complete

Common Big O Notations

$\log(n)$

