DDG University

Informal Goals

- 1. Learn new things related to technology.
- 2. Learn from each other.
- 3. Foster inter-team building.
- 4. To become better engineers.

Search for DDG University in Asana.

Structure and Interpretation of Computer Programs (SICP)

by Harold Abelson and Gerald Jay Sussman

1.2 Procedures and the Processes They Generate

- 1. The shape of recursion and iteration
- 2. Orders of growth
- 3. Big O Notation
- 4. Greatest Common Divisors
- 5. Improving Prime Number Searching

Factorials

$$n! = n * (n-1) * (n-2) ... 3 * 2 * 1$$

 $6! = 6 * 5 * 4 * 3 * 2 * 1 = 120$

Recursive Process

Visualized with substitution model expansion:

```
(factorial 6)
(* 6 (factorial 5))
(* 6 (* 5 (factorial 4)))
(* 6 (* 5 (* 4 (factorial 2))))
(* 6 (* 5 (* 4 (* 3 (factorial 1)))))
(* 6 (* 5 (* 4 (* 3 (* 2 (factorial 1))))))
(* 6 (* 5 (* 4 (* 3 (* 2 1)))))
(* 6 (* 5 (* 4 (* 3 2))))
(* 6 (* 5 (* 4 6)))
(* 6 (* 5 24))
(* 6 120)
```

- 1. Defers execution
- 2. State is on the stack
- 3. Uses space equal to n

Iterative Process

Visualized with substitution model expansion:

```
(factorial 6)
(fact-iter 1 1 6)
(fact-iter 1 2 6)
(fact-iter 2 3 6)
(fact-iter 6 4 6)
(fact-iter 24 5 6)
(fact-iter 120 6 6)
(fact-iter 720 7 6)
```

- 1. Execution is not deferred
- 2. State is held variables
- 3. Uses constant space

Recursive Process vs. Procedure

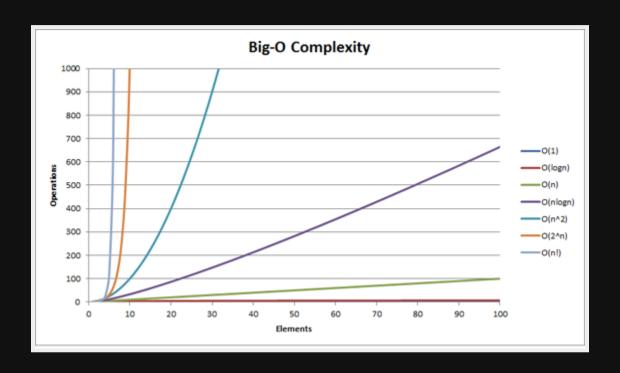
- A recursive *procedure* function that refers to itself. This is purely syntactic.
- A recursive *process* describes the way a function evolves in time and space.

Orders of Growth

We can describe the *shapes* of the above processes with the notion of *order of growth*. We are looking to measure the resources required as the input increases.

Big O Notation

Describes the growth behaviour of a function relative to its input.



Big Oh Theta Notation

Actually we're talking about Big Theta (Θ) notation. O() measures the upper bound of growth whereas Θ() measures the exact growth.

$$F(n) = O(n^3) - F(n)$$
 grows no faster than n^3
 $F(n) = Θ(n^3) - F(n)$ grows as fast as n^3
The more you know!

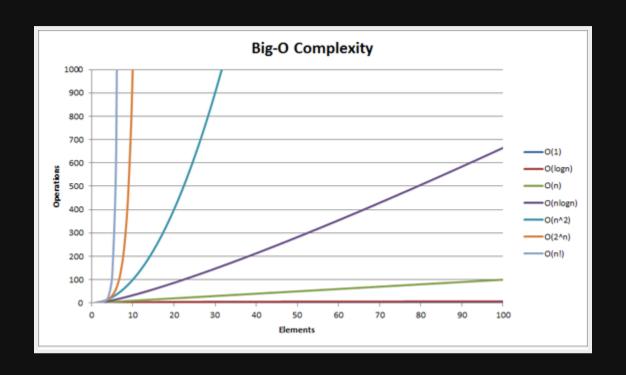
Big O Notation

Describes the growth behaviour of a function relative to its input.

If we were to measure the execution time of a function:

- O(1): If the input size changes, the execution stays the same.
- O(n): If the input size doubles, the execution time doubles.
- $O(n^2)$: If the input size doubles, the execution time quadruples.
- O(log n): If the input size doubles, the execution time increases by one.
- $O(2^n)$: If the input size increases by one, the execution time doubles.

Big O Notation (cont.)



http://stackoverflow.com/questions/487258/plain-english-explanation-of-big-o

Recursive Growth

time =
$$O(n)$$

space = $O(n)$

Iterative Growth

time =
$$O(n)$$

space = $O(1)$

Greatest Common Divisors

The GCD of two integers a and b is defined to be the largest integer that divides both a and b with no remainder.

Euclid's Algorithm:

```
GCD(a,b) = GCD(b,r)
```

For example:

```
GCD(206,40) = GCD(40,6)
= GCD(6,4)
= GCD(4,2)
= GCD(2,0) = 2
```

An iterative process to express this:

Improving Primality Tests

Our benchmarking code:

We are going to be measuring (prime?) for 10^{10} , 10^{11} , 10^{12} , and 10^{13} .

First iteration:

When testing for primality you only need to check divisors 2 thru √n. So for 81 you only need to check 2 thru 9.

This means that our first iteration is $O(\sqrt{n})$ time.

First iteration results:

```
> (timed-prime-test 10000000019)
10000000019 *** .09

> (timed-prime-test 100000000003)
100000000003 *** .26

> (timed-prime-test 1000000000039)
1000000000039 *** .8

> (timed-prime-test 1000000000037)
10000000000037 *** 2.51
```

We can do better.

Second iteration:

```
(define (next n)
  (if (= n 2) 3 (+ n 2)))
(define (smallest-divisor n)
  (find-divisor n 2))
(define (find-divisor n test-divisor)
  (cond ((> (square test-divisor) n)
        ((divides? test-divisor n)
        test-divisor)
        (else (find-divisor
           (next test-divisor)))))     <---- LESS SLOW!!!</pre>
(define (divides? a b)
  (= (remainder b a) 0))
(define (prime? n)
  (= n (smallest-divisor n)))
```

You would think that our second iteration is $O(\sqrt{n}/2)$ time. But it's not quite twice as fast.

Second iteration results:

We can still do better.

Third iteration:

```
(define (expmod base exp m)
  (cond ((= exp 0) 1)
        ((even? exp)
         (remainder
          (square (expmod base (/ exp 2) m))
          m))
        (else
         (remainder
          (* base (expmod base (- exp 1) m))
          m))))
(define (fermat-test n)
  (define (try-it a)
    (= (expmod a n n) a))
  (try-it (+ 1 (random (- n 1)))))
(define (prime? n times)
  (cond ((= times 0) true)
        ((fermat-test n)
         (prime? n (- times 1)))
        (else false)))
```

Our final iteration is O(log n) time, or fast as hell.

Third iteration results:

```
> (timed-prime-test 10000000019)
10000000019 *** 9.99999999999787e-3

> (timed-prime-test 100000000003)
100000000003 *** 9.9999999999999787e-3

> (timed-prime-test 10000000000039)
1000000000039 *** 1.000000000000675e-2

> (timed-prime-test 10000000000037)
10000000000037 *** 9.999999999999787e-3
```

That's all for section 1.2. Thanks!