

Week 03 - Lecture 1 Slides

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Lecture 1: Analysis of Algorithms (cont.)

Learning objectives: By the end of this lecture you should be able to

1. Use “Big-O” to describe program execution time

$O(T(n))$ and $T(n)$: assignment statement

(SETF *var form*) or (*:= var form*)

Example:

(setf x (+ y b)) ; Or using RUTIL's := macro (*:= x (+ y b)*)

As long as the operations in *form* do not involve loops or recursive functions:

- Running time: $T(n) = c$, where c is a constant, e.g., 1. Therefore, for our purposes $T(n) = 1$
- Order of magnitude: $O(1)$

$O(T(n))$ and $T(n)$: LET and LET* form

(LET *varInits form**) or (LET* *varInits form**)

As long as the operations in *varInits* do not involve loops or recursive functions: Example:

- we will assume the running time of each variable initialization is the running time of one assignment statement
- the running time of *varInits*, denoted $T(\text{varInits})$ is the summation of the running times of each of its variable initializations
- running time for the LET statement: $T(n) = T(\text{varInits}) + T(\text{form}^*)$

Example:

```
(let ((a 5) (b 6) (c 10) (d 0) (e 40)
      (k 1) (x 1) (y 1) (v 1) (w 1) (z 1))
  (:= a (* (/ b c e) (+ d k x y v w z))))
```

- Running time: $T(n) = 11 + 1 = 12$
- Order of magnitude: $O(1)$

$O(T(n))$ and $T(n)$: loops

(DOTIMES *VarTestRes bodyForm**) or (DO *VarTestRes bodyForm**)

As long as the variable initializations, variable update operations, test, and result form computation in *VarTestRes* do not

involve loops or recursive functions:

- We will ignore the computational time taken by DO and DOTIMES to carry out those operations.
 - We will only consider the running time of *bodyForm**

Example: Simple loop

```
(dotimes (i n)           ; Repeats n times.
  (:= x (+ i b))
  (:= y (* x 2)))
```

- Running time: $\sum_{i=0}^{n-1} (1 + 1) = 2n$
- Order of magnitude is therefore $O(n)$

$O(T(n))$ and $T(n)$: loops (cont.)

```
(dotimes (i n)           ; Repeats n times.
  (:= x (+ i b))         ; takes constant time, say, 1 time unit
  (dotimes (j n)         ; Repeats n times
    (:= y (+ a b))))     ; takes constant time, say, 1 time unit
```

- Running time: $\sum_{i=0}^{n-1} (1 + \underbrace{\sum_{j=0}^{n-1} 1}_{\text{inner loop}}) = \sum_{i=0}^{n-1} (1 + n) = n + n^2$
- Order of magnitude is therefore $O(n^2)$

Obtaining $O(T(n))$ and $T(n)$: another example

```
(let ((a 5) (b 6) (c 10) (d 0)
      (k 1) (x 1) (y 1) (v 1) (w 1) (z 1)) ; each initialization takes constant time
  (dotimes (i n)
    (dotimes (j n)
      (:= z (* i i)) ; each assignment statement takes some constant time
      (:= y (/ j j))
      (:= z (+ i j))))
  (dotimes (k n)
```

```
(:= w (+ (* a k) 45)) ; each assignment statement takes some constant time
(:= v (* b b)))
(:= d 33))
```

- Running time:

$$T(n) = 10 + \left(\sum_{i=0}^{n-1} \sum_{j=0}^{n-1} 3\right) + \left(\sum_{k=0}^{n-1} 2\right) + 1 = 3n^2 + 2n + 11$$

- Order of magnitude: $O(n^2)$

Question

- Complete the blanks in the expressions below that provide the execution time $T(n)$ and Big-O performance (aka "order of magnitude/growth") for the following code fragment:

```
(dotimes (i n) ;
  (dotimes (j n)
    (:= k (+ 2 2))))
(dotimes (i n)
  (:= k (+ 2 2)))
```

$$T(n) = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} \dots + \sum_{i=0}^{n-1} \dots = \dots + n$$

$$O(\dots)$$

Solution

```
(dotimes (i n)
  (dotimes (j n)
    (:= k (+ 2 2))))
(dotimes (i n)
  (:= k (+ 2 2)))
```

$$T(n) = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} 1 + \sum_{i=0}^{n-1} 1 = n^2 + n$$

$O(n^2)$

Another example

Let's obtain $T(n)$ and Big-O for the code fragment below:

```
(:= k 0)
(do ((i 1 (* i 2))) ; Note: increment of i is (setf i (* i 2))
    ((>= i n))
    (:= k (+ K (+ 2 2)))))
```

So we rush to get $T(n)$: $T(n) = 1 + \sum_{i=1}^n \text{????}$

- The sigma (Σ) notation assumes the increment is by ones
- i is not being incremented by ones!
- i does not grow linearly, i.e., 1, 2, 3, 4, 5, It grows exponentially: 1, 2, 4, 8, 16, ...

To calculate $T(n)$ using a summation (Σ) we need to transform the increment of i in increment by ones

Example (Cont.)

- Notice that i grows exponentially: 1, 2, 4, 8, ...

```
(:= k 0)
(do ((i 1 (* i 2))) ; Note: increment of i is (setf i (* i 2))
    ((>= i n))
    (:= k (+ K (+ 2 2)))))
```

- If we make $i = 2^p$, with p starting from 0, incrementing p by 1. When $i = n$, $p = \log n$ (log n on base 2), then the loop becomes:

```
(:= k 0)
(do ((p 0 (1+ p))) ; ignore computation time
    ((>= p (log n 2))) ; ignore computation time
    (:= k (+ k (+ 2 2))))) ; 1 unit of time
```

Now we are ok. The increment of the variable p that control the loop is by ones.

Therefore, $T(n)$ and Big-O for the revised code are

$$T(n) = 1 + \sum_{p=0}^{\log n - 1} 1 = 1 + \log n$$

$$O(\log n)$$

Question

What is $T(n)$ and Big-O for the code fragment below?

```
(dotimes (i n)
  (do ((j 1 (* j 2))) ; ignore computation time
      ((>= j n))      ; ignore computation time
      (:= a (+ b c))) ; 1 unit of time
```

$$T(n) = \sum_{i=0}^{\dots} (\sum_{p=0}^{\dots} 1) = (\sum_{i=0}^{\dots} 1 \dots n) = 1 \dots \log n$$

$$O(\dots \log n)$$

Solution

```
(dotimes (i n)
  (do ((j 1 (* j 2)))
      ((>= j n))
      (:= a (+ b c)))
```

$$T(n) = \sum_{i=0}^{n-1} (\sum_{p=0}^{\log n - 1} 1) = (\sum_{i=0}^{n-1} 1 \log n) = 1n \log n$$

$$O(n \log n)$$

Homework

What is $T(n)$ and Big-O for function G?

```
(defun f (a b)
  (let ((acc 0))
```

```

(dotimes (i a)
  (incf acc))))

(defun g(n)
  (dotimes (i n)
    (dotimes (j n (f i j))
      (setf x (+ i j))))))

```

Solution

$$T_g(n) = \sum_{i=0}^{n-1} \left(\sum_{j=0}^{n-1} (1) + T_f(n) \right)$$

$$T_f(n) = 1 + \sum_{i=0}^{a-1} 1 = 1 + a \quad \text{Substituting } T_f(n) \text{ in } T_g(n)$$

$$T_g(n) = \underbrace{\sum_{i=0}^{n-1} \sum_{j=0}^{n-1} 1}_{\sum_{i=0}^{n-1} n} + \underbrace{\sum_{i=0}^{n-1} 1}_n + \underbrace{\sum_{i=0}^{n-1} i}_{\frac{n(n-1)}{2}} \quad \text{Note: } a \text{ in function } f \text{ is } i \text{ in function } g$$

$$T_g(n) = \sum_{i=0}^{n-1} n + n + \frac{n(n-1)}{2} = n^2 + n + \frac{1}{2}(n^2 - n) \quad \therefore$$

$$\text{Running time of function G: } T_g(n) = n^2 + n + \frac{1}{2}(n^2 - n)$$

$$\text{Big-O: } O(n^2)$$