

# Patterns in Functional Programming

Please take a handout &  
Sit in row H or forward

Recap:

Constructing and  
Deconstructing Lists

# Lists

`(cons 6 '(7 8 9))`

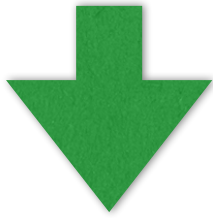


`'(6 7 8 9)`

`length`

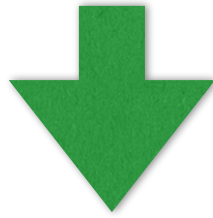


`first`



`6`

`rest`



`'(7 8 9)`

`4`

# Lists

```
(cons '(1 2 3) '(7 8 9))
```



```
'((1 2 3) 7 8 9)
```

length

first



```
'(1 2 3)
```



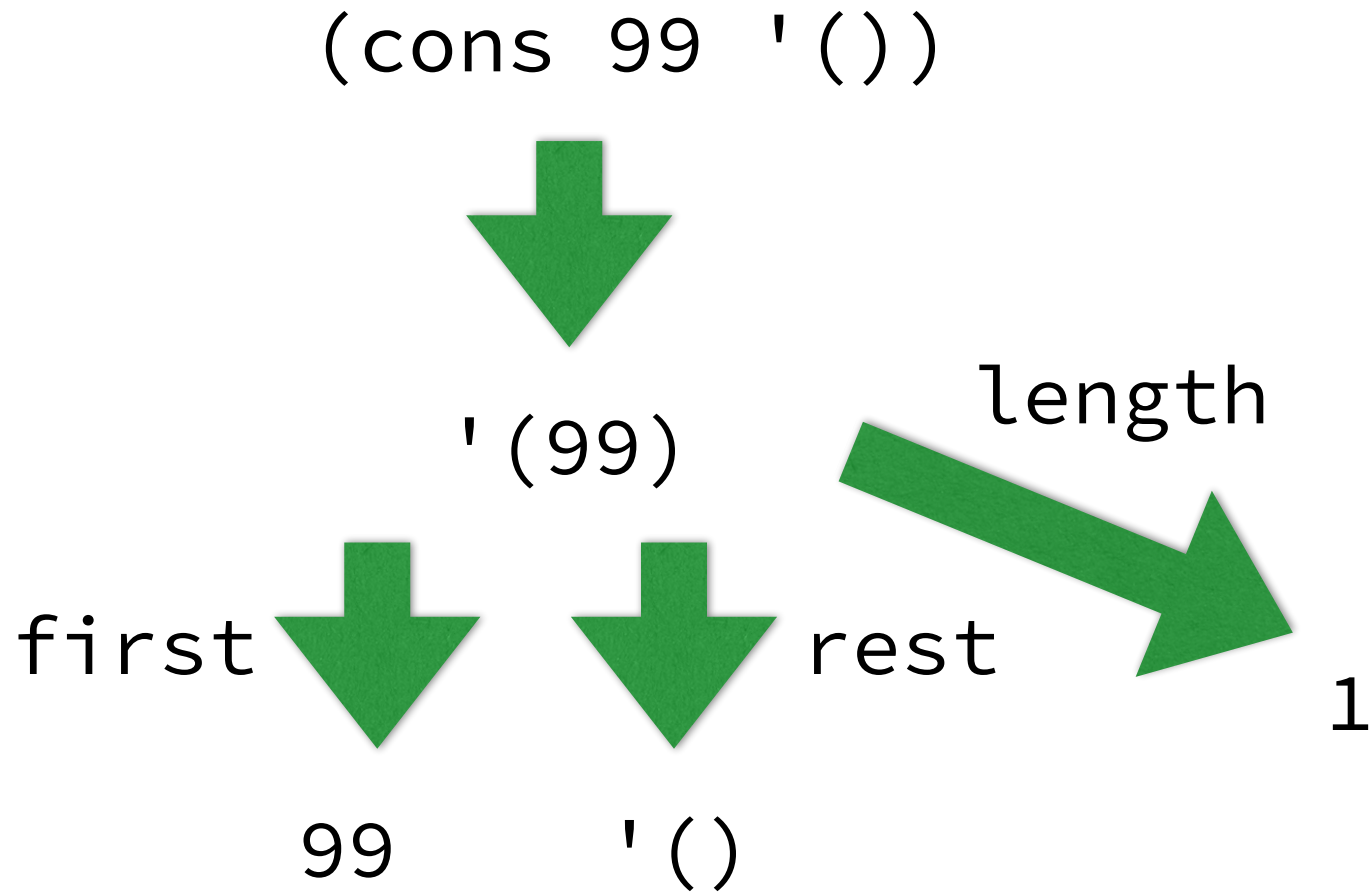
```
'(7 8 9)
```

rest



4

# Lists



# Complexity and Recurrence Relations

# In the World of Big $O$

## *Asymptotic Analysis*

We're always answering the same question:

How does the algorithm ***scale***  
(when we try larger and larger inputs)

NOT

- Exactly how many steps will it execute?
- How many seconds will it take?
- How many megabytes of memory will it need?

# Scaling in Time and Space

“Given an array of length  $n$ ,  
the MergeSort algorithm  
requires time  $O(n \log n)$   
and uses  $O(n)$  space.”



# In the limit (for VERY LARGE inputs)

The running time is bounded  
regardless of the input size.  $O(1)$

An input twice as big takes  
no more than twice as long.  $O(n)$

An input twice as big takes  
no more than four times as long.  $O(n^2)$

An input one bigger takes  
no more than twice as long.  $O(2^n)$

# If We Only Care About Scalability...

*What are the consequences?*

Constant factors can be ignored.

**n** and **6n** and **200n** scale identically (“linearly”)

Small summands can be ignored.

**n<sup>2</sup>** and **n<sup>2</sup> + n + 999999** are indistinguishable when n is huge.

# Grouping Algorithms by Scalability

$O(1)$  takes 6 steps  
takes 1 (big) step  
no more than 4000 steps  
somewhere between 2 and 47 steps, depending on the input

$O(n)$  takes  $100n + 3$  steps  
takes  $n/20 + 10,000,000$  steps  
anywhere between 3 and 68 steps per item, for  $n$  items.

$O(n^2)$  takes  $2n^2 + 100n + 3$  steps  
takes  $n^2/17$  steps  
somewhere between 1 and 40 steps per item, for  $n^2$  items  
anywhere between 1 and  $7n$  steps per item, for  $n$  items.

# Big O — Step 2

Measuring inputs

How does the algorithm ***scale***  
(when we try larger and larger inputs)

The first step is defining a “measure” of input size.  
Conventionally we call this  $n$  or  $N$ , but not always.

Suppose we are analyzing a function  
whose input is a single string?

The function is  $O(\dots)$ ,  
where  $n$  is the length of the input string.

# Big O — Step 2

## Measuring inputs

Suppose we are analyzing functions whose input is an array of strings?

The function takes  $O(\dots)$  time,  
where  $n$  is the length of the array.

The function takes  $O(\dots)$  time,  
where  $n$  is the length of the longest string.

The function takes  $O(\dots)$  time,  
where  $n$  is the total number of characters.

The function takes  $O(\dots)$  time,  
where  $n$  is the length of the array  
and  $m$  is the length of the longest string.

# The scalability of sum revisited

Step 1: Choose an appropriate measure of input “size”

```
;; input: a list of numbers  
;; output: the sum of the given numbers  
(define (sum L)  
  (if (empty? L)  
      0  
      (+ (first L) (sum (rest L)))))
```

We care about what happens when the input is large.  
For **this** function, how should we measure L?

- The sum of the numbers in L?
- The biggest number in L?
- The total number of digits in the numbers in L?
- Something else?

# The scalability of sum revisited

Step 2: Choosing what to count

```
;; input: a list of numbers  
;; output: the sum of the given numbers  
(define (sum L)  
  (if (empty? L)  
      0  
      (+ (first L) (sum (rest L))))))
```

Suppose we're interested in how **running-time** scales.

What should we count?

- Additions ?
- `empty?` calls ?
- `first` and `rest` operations ?
- Maybe all of these ?

# Counting Steps

For our asymptotic analysis

```
;; input: a list of numbers
;; output: the sum of the given numbers
(define (sum L)
  (if (empty? L)
      0
      (+ (first L) (sum (rest L)))))
```

Let  $n$  be the length of the given list.

We want  $T(n)$ , the number of “steps” required.

When  $n = 0$ , we do  $O(1)$  steps.

When  $n > 0$ , we do  $O(1)$  steps  
in addition to making a recursive call  
on an input of size  $n-1$ .



# Recurrence Relations in Big-O

We want to describe a function  $T(n)$  such that sum takes  $O(T(n))$  steps on an input of size  $n$

$$T(0) = 1$$

$$T(n) = 1 + T(n-1) \quad \text{when } n > 0$$

$$T(n) = \begin{cases} 1 & \text{if } n = 0 \\ 1 + T(n-1) & \text{otherwise} \end{cases}$$

```
;; Input: n, the length of the input list
;; Output: T(n), the asymptotic number of steps
;;         taken by sum on an input of size n.
(define (T n)
  (if (= n 0)
      1
      (+ 1 (T (- n 1)))))
```

# “Solving” the Recurrence

sum requires  $O(T(n))$  steps on an input of size  $n$

```
Welcome to DrRacket, version 6.3 [3m].  
Language: racket; memory limit: 128 MB.  
> (T 0)  
1  
> (T 1)  
2  
> (T 2)  
3  
> (T 3)  
4  
> (T 1000)  
1001
```

aha!  $T(n) = n+1$  for all inputs  
and I (or Mathematica) could prove it!

$\therefore$  sum requires  $O(n)$  steps for a list of length  $n$

# An Error That TDD Might Not Detect

What Changed?

```
;; input: a list of numbers
;; output: the sum of the given numbers
(define (sum L)
  (if (equal? (length L) 0)
      0
      (+ (first L) (sum (rest L)))))
```

Let  $n$  be the length of the given list.

We want  $T(n)$ , the number of “steps” required.

When  $n = 0$ , we do  $O(1)$  steps.

When  $n > 0$ , we do  $O(n)$  steps  
in addition to making a recursive call  
on an input of size  $n-1$ .

# Recurrence Relation for sum

We want to describe a function  $T(n)$  such that sum takes  $O(T(n))$  steps on an input of size  $n$

$$T(0) = 1$$

$$T(n) = n + T(n-1) \quad \text{when } n > 0$$

$$T(n) = \begin{cases} 1 & \text{if } n = 0 \\ n + T(n-1) & \text{otherwise} \end{cases}$$

```
;; Input: n, the length of the input list
;; Output: T(n), the asymptotic number of steps
;;         taken by summ on an input of size n.
(define (T n)
  (if (= n 0)
      1
      (+ n (T (- n 1)))))
```

# Solving the New Recurrence for sum

```
Welcome to DrRacket, version 6.3 [3m].  
Language: racket; memory limit: 128 MB.  
> (T 0)  
1  
> (T 1)  
2  
> (T 2)  
4  
> (T 3)  
7  
> (T 1000)  
500501
```

aha!  $T(n) = \frac{1}{2}n^2 + \frac{1}{2}n + 1$  for all inputs  
and I (or Mathematica) could prove it!

$\therefore$  This sum requires  $O(n^2)$  steps on an input of size  $n$

# Appending Two Lists: Intuition

(append '(1 2 3) '(4 5))



(cons 1 (append '(2 3) '(4 5)))



(cons 2 (append '(3) '(4 5)))



(cons 3 (append '() '(4 5)))



'(4 5)

```
(define (append L M)
  (if (empty? L)
      M
      (cons (first L)
            (append (rest L) M)))))
```

# Appending Two Lists: Intuition

`(append '(1 2 3) '(4 5)) ===`

`(cons 1 (append '(2 3) '(4 5))) ===`

`(cons 1 (cons 2 (append '(3) '(4 5)))) ===`

`(cons 1 (cons 2 (cons 3 (append '() '(4 5))))) ===`

`(cons 1 (cons 2 (cons 3 '(4 5)))) ===`

`'(1 2 3 4 5)`

# Counting Steps

For our asymptotic analysis

```
(define (append L M)
  (if (empty? L)
      M
      (cons (first L)
            (append (rest L) M)))))
```

Let  $n$  be the length of list  $L$ .

We want  $T(n)$ , the number of “steps” required.

When  $n = 0$ , we do  $O(1)$  steps.

When  $n > 0$ , we do  $O(1)$  steps  
in addition to making a recursive call  
on an input of size  $n-1$ .



# The Recurrence Relation

We want to describe a function  $T(n)$  such that append takes  $O(T(n))$  steps on an input of size  $n$

$$T(0) = 1$$

$$T(n) = 1 + T(n-1) \quad \text{when } n > 0$$

$$T(n) = \begin{cases} 1 & \text{if } n = 0 \\ 1 + T(n-1) & \text{otherwise} \end{cases}$$

Wait, we've solved this recurrence already!

$\therefore$  append requires  $O(n)$  steps  
when the **first** input list has length  $n$



# Patterns of Control Flow

# double-list

```
(define (double n)
  (* 2 n))
```

```
(define (double-list L)
  (if (empty? L)
      '()
      (cons (double (first L))
            (double-list (rest L)))))
```

*; tests*

```
(check-equal? (double-list '()) '())
(check-equal? (double-list '(1)) '(2))
(check-equal? (double-list '(1 2)) '(2 4))
(check-equal? (double-list '(2 3 2)) '(4 6 4))
```

# Tracing double-list

```
> (double-list '(1 2 3 4 5))  
>(double-list '(1 2 3 4 5))  
> (double-list '(2 3 4 5))  
> >(double-list '(3 4 5))  
> > (double-list '(4 5))  
> > >(double-list '(5))  
> > > (double-list '())  
< < < '()  
< < <'(10)  
< < '(8 10)  
< <'(6 8 10)  
< '(4 6 8 10)  
<'(2 4 6 8 10)  
'(2 4 6 8 10)
```

# Tracing (double-list '(1 2 3 4 5))

```
> (double-list '(1 2 3 4 5))  
> (double-list '(1 2 3 4 5))  
> (double-list '(2 3 4 5))
```

```
< '(4 6 8 10)  
< '(2 4 6 8 10)  
'(2 4 6 8 10)
```

```
(cons (double (first L))  
      (double-list (rest L)))
```

# square-list

```
(define (square x)
  (* x x))
```

```
(define (square-list L)
  (if (empty? L)
      '()
      (cons (square (first L))
              (square-list (rest L)))))
```

*; tests*

```
(check-equal? (square-list '()) '())
(check-equal? (square-list '(1)) '(1))
(check-equal? (square-list '(1 2)) '(1 4))
(check-equal? (square-list '(2 3 2)) '(4 9 4))
```

# The map pattern

```
(define (double-list L)
  (if (empty? L)
      '()
      (cons (double (first L))
              (double-list (rest L)))))
```

```
(define (square-list L)
  (if (empty? L)
      '()
      (cons (square (first L))
              (square-list (rest L)))))
```



# The map pattern

```
(define (map f L)
  (if (empty? L)
      '()
      (cons (f (first L))
              (map f (rest L)))))
```

```
(define (double-list L)
  (map double L))
```

```
(define (square-list L)
  (map square L))
```

# Fun Fact

The “built-in” map function is even more general.

```
> (map double '(1 2 3))  
'(2 4 6)  
> (map list '(1 2 3) '(10 11 12))  
'((1 10) (2 11) (3 12))  
> (map + '(1 2 3) '(10 11 12))  
'(11 13 15)
```

If we provide two lists,  
we also need a two-argument function

# sum

```
(define (sum L)
  (if (empty? L)
      0
      (+ (first L) (sum (rest L)))))
```

and what if we wanted product?

*; tests*

```
(check-equal? (sum '()) 0)
```

```
(check-equal? (sum '(1)) 1)
```

```
(check-equal? (sum '(1 2)) 3)
```

```
(check-equal? (sum '(2 3 2)) 7)
```

# product

```
(define (product L)
  (if (empty? L)
      1
      (* (first L) (product (rest L)))))
```

*; tests*

```
(check-equal? (product '()) 1)
(check-equal? (product '(1)) 1)
(check-equal? (product '(1 2)) 2)
(check-equal? (product '(2 3 2)) 12)
```

# The foldr pattern

```
(define (foldr f b L)
  (if (empty? L)
      b
      (f (first L) (foldr f b (rest L)))))
```

```
(define (sum L)
  (foldr + 0 L))
```

```
(define (product L)
  (foldr * 1 L))
```

# Summing using foldr

`(sum '(1 2 3))`

`(foldr + 0 '(1 2 3))`

`(+ 1 (foldr + 0 '(2 3)))`

`(+ 2 (foldr + 0 '(3)))`

`(+ 3 (foldr + 0 '()))`

`0`

# Summing using foldr

`(sum ' (1 2 3)) ===`

`(foldr + 0 ' (1 2 3)) ===`

`(+ 1 (foldr + 0 ' (2 3))) ===`

`(+ 1 (+ 2 (foldr + 0 ' (3)))) ===`

`(+ 1 (+ 2 (+ 3 (foldr + 0 ' ()))) ===`

`(+ 1 (+ 2 (+ 3 0))) ===`

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# Define append using foldr



```
(define (append L M)
  (if (empty? L)
      M
      (cons (first L)
            (append (rest L) M)))))
```

```
(append '(1 2 3) '(4 5)) ==
(cons 1 (cons 2 (cons 3 '(4 5))))
```

```
(define (foldr f b L)
  (if (empty? L)
      b
      (f (first L) (foldr f b (rest L)))))
```

```
(define (append L M)
  (foldr cons M L))
```