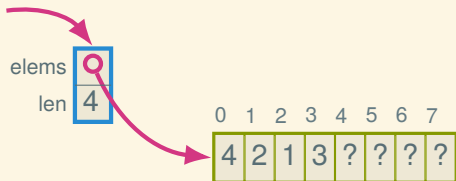
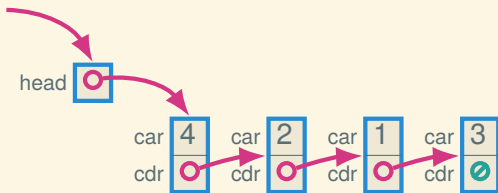


Asymptotic Complexity

CS 214, Fall 2019

A comparison



How long would it take to...

- Get or set the n th element?
- Add an element to the front?
- Add an element to the back?
- Determine whether x is an element?

Getting the n th element

```
def list_nth(lst, n):  
    def loop(i, link):  
        if not link: error('list_nth: out of bounds')  
        elif i == 0: return link.car  
        else:       return loop(i - 1, link.cdr)  
    loop(n, lst.head)
```

Getting the *n*th element

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def list_nth(lst, n):  
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        elif i == 0: return link.car  
        else:       return loop(i - 1, link.cdr)  
    loop(n, lst.head)  
  
def array_nth(array, n):  
    if n < array.len:  
        return array.elems[n]  
    else:  
        error('array_nth: out of bounds')
```

Getting the *n*th element

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def list_nth(lst, n):  
    def loop(i, link):  
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    loop(n, lst.head)  
  
def array_nth(array, n):  
    if n < array.len:  
        return array.elems[n]  
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        error('array_nth: out of bounds')
```

The loop in `list_nth` repeats *n* times. `array_nth` has no loop.

Adding an element to the front

```
def list_push_front(lst, val):  
    lst.head = cons(val, lst.head)
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Adding an element to the front

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def list_push_front(lst, val):  
    lst.head = cons(val, lst.head)  
  
def array_push_front(array, val):  
    if array.len == array.elems.len():  
        error('array_push_front: out of space')  
    let i = array.len  
    while i > 0:  
        array.elems[i] = array.elems[i - 1]  
        i = i - 1  
    array.elems[0] = val  
    array.len = array.len + 1
```


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    let i = array.len  
    while i > 0:  
        array.elems[i] = array.elems[i - 1]  
        i = i - 1  
    array.elems[0] = val  
    array.len = array.len + 1
```

`list_push_front` is loop-free, whereas `array_push_front` loops `array.len` times.

Breaking down list-nth

```
def list_nth(lst, n):  
    let link = lst.head  
    for i in n:  
        link = link.cdr  
    return link.car
```

(T stands for “time”)

$$T_{\text{list_nth}}(n) =$$

Breaking down list-nth

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Let $c_1 = T_{\text{get head}} + T_{\text{loop setup}} + T_{\text{get car}}$

Let $c_2 = T_{\text{get cdr}} + T_{\text{assign link}} + T_{\text{loop inc}}$

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$$T_{\text{list_nth}}(n) = c_1 + c_2n$$

Let $c_1 = T_{\text{get head}} + T_{\text{loop setup}} + T_{\text{get car}}$

Let $c_2 = T_{\text{get cdr}} + T_{\text{assign link}} + T_{\text{loop inc}}$

Operation time comparison

	list	array
nth	$c_1 + c_2n$	d_1
push_front	e_1	$f_1 + f_2n$

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No matter what the values of c_1 , c_2 , and e_1 are, if n gets large enough then $c_1 + c_2n$ will be larger than e_1 .

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nth	$c_1 + c_2n$	d_1
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No matter what the values of c_1 , c_2 , and e_1 are, if n gets large enough then $c_1 + c_2n$ will be larger than e_1 .

The same cannot be said when comparing $c_1 + c_2n$ to $f_1 + f_2n$.

Complexity classes

There's a sense in which $c_1 + c_2n$ and $f_1 + f_2n$ are similar.

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There's a sense in which $c_1 + c_2n$ and $f_1 + f_2n$ are similar.

We call this sense $\mathcal{O}(n)$.

Another example: insertion sort

```
# : Link[num?] -> Link[num?]  
def insertion_sort(lst):  
  def insert(elt, link):  
    if not link or elt < link.car:  
      cons(elt, link)  
    else:  
      cons(link.car, insert(elt, link.cdr))  
  
  let result = None  
  let curr = lst.head  
  while curr:  
    result = insert(curr.car, result)  
    curr = curr.cdr  
  result
```


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  let result = None  
  let curr = lst.head  
  while curr:  
    result = insert(curr.car, result)  
    curr = curr.cdr  
  result
```

Nested loops of length n : $\mathcal{O}(n^2)$

Another example: merge sort helpers (1/2)

```
# : Link[num?] Link[num?] -> Link[num?]  
def merge(lnk1, lnk2):  
  if lnk1 and lnk2:  
    if lnk1.car < lnk2.car:  
      cons(lnk1.car, merge(lnk1.cdr, lnk2))  
    else:  
      cons(lnk2.car, merge(lnk1, lnk2.cdr))  
  else: lnk1 or lnk2
```

Another example: merge sort helpers (1/2)

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# : Link[num?] Link[num?] -> Link[num?]  
def merge(lnk1, lnk2):  
  if lnk1 and lnk2:  
    if lnk1.car < lnk2.car:  
      cons(lnk1.car, merge(lnk1.cdr, lnk2))  
    else:  
      cons(lnk2.car, merge(lnk1, lnk2.cdr))  
  else: lnk1 or lnk2
```

merge is $\mathcal{O}(|\text{lnk1}| + |\text{lnk2}|) = \mathcal{O}(n)$.

Another example: merge sort helpers (2/2)

```
def odds(link):  
    if link:  
        cons(link.car, evens(link.cdr))  
    else:  
        None  
  
def evens(link):  
    if link:  
        odds(link.cdr)  
    else:  
        None
```

Another example: merge sort helpers (2/2)

```
def odds(link):  
    if link:  
        cons(link.car, evens(link.cdr))  
    else:  
        None
```

```
def evens(link):  
    if link:  
        odds(link.cdr)  
    else:  
        None
```

odds and evens are both $\mathcal{O}(n)$.

Another example: merge sort

```
# : Link[num?] -> Link[num?]  
def merge_sort(link):  
  if not link or not link.cdr:  
    link  
  else:  
    merge(merge_sort(odds(link)),  
          merge_sort(evens(link)))
```

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In each recursion we take $\mathcal{O}(n)$. How many times do we recur?

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```

In each recursion we take $\mathcal{O}(n)$. How many times do we recur?
 $\mathcal{O}(\log n)$ times.

Merge sort versus insertion sort

Merge sort takes $\mathcal{O}(n \log n)$. Insertion sort takes $\mathcal{O}(n^2)$. What does this mean concretely?

n	n^2	$n \log n$
10	100	10
100	10,000	200
1,000	1,000,000	3,000
10,000	100,000,000	40,000
100,000	10,000,000,000	500,000

Merge sort versus insertion sort

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n	n^2	$n \log n$
1e1	1e2	1e1
1e2	1e4	2e2
1e3	1e6	3e3
1e4	1e8	4e4
1e5	1e10	5e5

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n	n^2	$10^{12}n \log n$
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1e3	1e6	3e3
1e4	1e8	4e4
1e5	1e10	5e5

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n	n^2	$10^{12}n \log n$
1e1	1e2	1e13
1e2	1e4	2e15
1e3	1e6	3e16
1e4	1e8	4e17
1e5	1e10	5e18

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1e4	1e8	4e17
1e5	1e10	5e18
1e13	1e26	1.3e26

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1e5	1e10	5e18
1e13	1e26	1.3e26
1e14	1e28	1.4e27

Formally

If f is a function, then $\mathcal{O}(f)$ is the set of functions that “grow no faster than” f

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“ g grows no faster than f ” means there exist some c and m such that for all $n > m$, $g(n) \leq cf(n)$

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If f is a function, then $\mathcal{O}(f)$ is the set of functions that “grow no faster than” f

“ g grows no faster than f ” means there exist some c and m such that for all $n > m$, $g(n) \leq cf(n)$

Intuitively: on large enough input (m), g grows no faster than f , up to getting a faster computer (c)

Another definition

$f \lll g$ means $f \in \mathcal{O}(g)$ but $g \notin \mathcal{O}(f)$

Big-O equalities

There are a bunch of rules we can apply to simplify complexity expressions:

- $\mathcal{O}(f(n) + c) = \mathcal{O}(f(n))$

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- $\mathcal{O}(cf(n)) = \mathcal{O}(f(n))$
- $\mathcal{O}(\log_k f(n)) = \mathcal{O}(\log_j f(n))$
- $\mathcal{O}(f(n) + g(n)) = \mathcal{O}(f(n))$ if $g \lll f$

Big-O inequalities

if $j < k$, then

$$1 \lll \log n$$

constants are less than logs...

Big-O inequalities

if $j < k$, then

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$$\lll n^j$$

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are less than polynomials...

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are less than poly-log...

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are less than exponentials...

$$\lll k^n$$

are less than higher-base exponentials.

Next time: Trees and Tree Walks