



## 2. Dictionaries and Sets

Programming and Algorithms II

Degree in Bioinformatics

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# Keeping counts. Example

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Input contains 9 6 30 -5 9 5 -5 30 4 1 5 9 5 2 4 -5

Output should be ideally the list

[ (-5,3) (1,1) (2,1) (4,2) (5,3) (6,1) (9,3) (30,2) ]

Or perhaps in another order (less good)

# A problem: Keeping counts

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Problem:

Given:

Input that contains a sequence of integers

Compute:

A list of pairs  $(x,c)$  where  $c$  is the number of times that  $x$  appears in the list (only for  $c > 0$ )

Optionally: the list should be sorted by  $x$

# First solution

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create an empty list L

for each element x in input

if x is not in L:

L.append((x,1))

else:

replace (x,c) with (x,c+1) in L

# First solution: Cost

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$N$  = number of integers in input

$D$  = number of **different** integers in input

Note  $\text{len}(L) \leq D$

“if  $x$  not in  $L$ ” cost up to  $O(D)$

even if implemented as  $L.\text{count}(x)$

Total cost  $O(ND)$

If  $D = N$  (all items different), cost is  $O(N^2)$

# Second solution: Keeping the list sorted

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Finding  $x$  in a list  $L$  is  $O(\text{len}(L))$

`L.count(x)` does not do magic

But if the list is sorted, we can do better

Binary or dichotomic search:

$O(\log D)$  time, where  $D = \text{len}(L)$

# Binary or dichotomic search

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$O(\log D)$  time, where  $D = \text{len}(L)$

Only on ordered structures

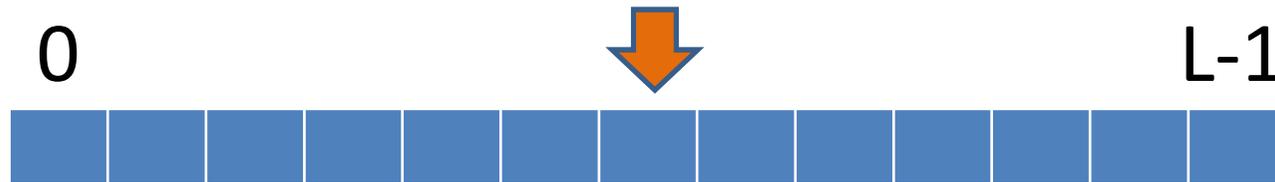
Only if  $O(1)$  Random Access... which lists have

For  $D=1,000$ ,  $\log D = 10$

For  $D = 1,000,000$   $\log D = 20$

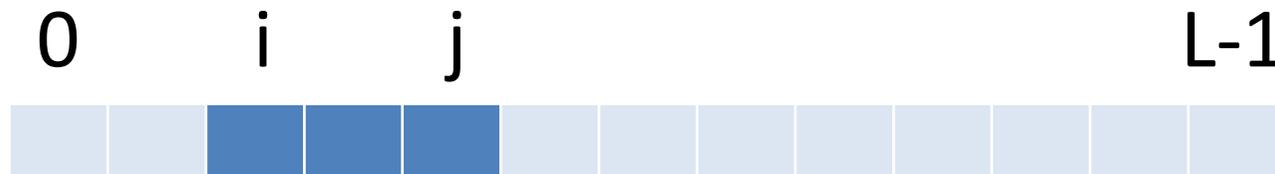
# Binary or dichotomic search

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In general, we keep two positions  $i, j$  such that

$x$  is in  $L$  if and only if it is in  $L[i:j]$



# Binary or dichotomic search

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```
def binary_search(L, x):
    i = 0
    j = len(L) - 1
    while i <= j:
        mid = (i+j)//2 # // is integer div
        if L[mid] > x:
            j = mid-1
        elif L[mid] < x:
            i = mid+1
        else:
            return mid
    return -1
```

If x is in L, returns **some** position of i that contains x

If x is not in L, returns -1

2 comparisons per iteration

# Binary or dichotomic search

---

```
def binary_search(L, x) :
    i = 0
    j = len(L) - 1
    while i < j:
        mid = (i+j)//2
        if L[mid] >= x:
            j = mid
        else:
            i = mid+1
    if i < len(L) and L[i] == x:
        return i
    else:
        return -1
```

If x is in L, returns **the first** position of L that contains x

If x is not in L, returns -1

1 comparison per iteration

# Why $\log_2(n)$

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At every iteration,  $j-i$  is divided by 2:

- Either  $i$  is the same and  $j \leq (i+j)/2$
- Or  $j$  is the same and  $i \geq (i+j)/2$

When  $j < i$ , we stop

How many times can we divide  $\text{len}(L)$  by 2 before we get to 0?

$$\text{len}(L) \approx 2^k \iff k \approx \log_2(\text{len}(L))$$

$$\text{example } 256 = 2^8 \rightarrow 128 \ 64 \ 32 \ 16 \ 8 \ 4 \ 2 \ 1 \ 0$$

# Second solution: Keeping L sorted

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Create an empty list L

For each element x in input

$i = \text{binarySearch}(L, x)$

if  $i == -1$ :

    insert (x,1) in L, in place that keeps L sorted

else

$x, c = L[i][0], L[i][1]$  # get pair (x,c) in pos. i

$L[i] = (x, c+1)$

# Second solution: Cost

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Remember that  $\text{len}(L) \leq D$

- Looking if  $x$  is in  $L$  (and where):  $O(\log D)$
- Adding it in right place if not in  $L$ :  $O(D)$
- Incrementing count if already in  $L$ :  $O(1)$
- $O(D)$  cost for every one of  $D$  distinct elements

Solution 1:  $O(ND)$

Solution 2:  $O(N \log D) + D * O(D) + (N-D) * O(1) =$   
 $= O(N \log D + D^2)$

Is still  $O(N^2)$  if  $D=N$ , but better if  $D \ll N$

## Solution 3

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1. read the whole input into a list A
2. sort A
3. go over A sequentially  
for each element A[i]  
if  $(i == \text{len}(L) - 1)$  or  $(A[i] < A[i+1])$  :  
append A[i] to L

1. A=[3,2,1,3,5,3,1,2,1,5,1]
2. A=[1,1,1,1,2,2,3,3,3,5,5]
3. L=[1,2,3,5]

## Solution 3

---

1. read the whole input into a list A
2. sort A
3. go over A sequentially  
for each element A[i]  
if  $(i == \text{len}(L) - 1)$  or  $(A[i] < A[i+1])$  :  
append A[i] to L

1.  $O(N)$
2.  $O(N \log N)$
3.  $O(N)$

Total time  $O(N \log N)$

# Recap: Costs

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Solution 1:  $O(ND)$

Solution 2:  $O(N \log D + D^2)$

Solution 3:  $O(N \log N)$

Solution 3 faster than 2 for  $D$  large w.r.t.  $N$

But uses memory  $O(N)$ , not  $O(D)$ .

No good for Big Data

(There are  $O(N \log N)$  algorithms for sorting files not in memory)

# The problem

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	Adding x	Searching x	Getting all elements in order
Unsorted list	.append $O(1)$	.count or linear search $O(D)$	.sort $O(D \log D)$
Sorted list	.insert $O(D)$	binary search $O(\log D)$	Trivial $O(D)$

# The problem

	Adding x	Searching x	Getting all elements in order
Unsorted list	.append $O(1)$	.count or linear search $O(D)$	sort $O(D \log D)$
Sorted list	.insert $O(D)$	binary search $O(\log D)$	Trivial $O(D)$
Hash table	$O(1)$	$O(1)$	$O(D \log D)$
Balanced tree	$O(\log D)$	$O(\log D)$	$O(D)$

- You will understand how they are built next term
- Detail: Hashing is  $O(1)$  “on average”
- **Python dictionaries use hash tables**

# Python dictionary operations

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Operation	Time
<code>d1 = d.copy()</code>	$O(n)$
Get, <code>x = d[key]</code> or <code>d.get(key, def)</code>	$O(1)$
Set, <code>d[key] = v</code>	$O(1)$
<code>key (not) in d</code>	$O(1)$
<code>del d[key]</code>	$O(1)$
<code>len(d)</code>	$O(1)$

Technically, \*average\* time. If you have really, really bad luck, all is  $O(n)$

# Python dictionary iterators

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Operation	Returns
<code>d.items()</code>	<code>[(k1,v1),(k2,v2),...,(kn,vn)]</code> <b>IN NO PARTICULAR ORDER</b>
<code>d.iteritems()</code>	Iterator over the list above
<code>d.keys()</code>	Iterator over the set of keys of <code>d</code>
<code>d.itervalues()</code>	Iterator over the set of values of <code>d</code>

Note: In Python2 this really returned a list.

In Python 3, it returns an “observer object”, a list that will change if the dictionary changes. Just put the result inside `list(...)` to get a proper list

# Keeping counts, Solution 4

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```
d = {}
for each x in input:
    if x in d:
        c = d[x]
    else:
        c = 0
    d[x] = c + 1
```

To sort at the end:

```
L = []
for x in sorted(d.keys()):
    L.append((x, d[x]))
```

## Solution 4, cost

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- Checking if  $x$  in dictionary  $O(1)$
- Getting count from dictionary  $O(1)$
- Adding / updating dictionary  $O(1)$
- $N$  iterations, each cost  $O(1)$
- Cost  $O(N)$
- Independent of  $D$ !!
  
- Sorting at the end,  $O(D \log D)$
- Cost  $O(N + D \log D)$

# Dictionaries vs. tables

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- Dictionaries preferable if only operations are “find” and “insert/update”
- Lists have (small) ranges of ints as keys
- Dictionaries take any hashable type as key
- If a list suffices, don't use a dictionary:  
Same  $O()$  but bigger constant in time and memory

# Sets

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- Special case of Dictionaries, if you want
- No value associated to key
- Just “is in” / “is not in”

# Sets

## Operation

## Time

`s1 = s.copy()`

$O(n)$

`s.isdisjoint(s1)`, `s.issubset(s1)`,  
`s.issuperset(s1)`

$O(n_1+n_2)$

`s.union(s1)`, `s.intersection(s1)`,  
`s.difference(s1)`

$O(n_1+n_2)$

`x (not) in s`

$O(1)$

`s.add(x)`

$O(1)$

`s.remove(x)` (exception if not in)  
`s.discard(x)` (no exception)

$O(1)$

`len(s)`

$O(1)$

`for x in s ...`

$O(n)$

# To remember

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- Dictionaries: Functions keys → values
- Sets: add, remove, is / is not
- **Huge speedups** over lists in many problems
- **They are your friends**
- Central to Python