# Scientific Programming: Part B

Lecture 5

Luca Bianco - Academic Year 2020-21 luca.bianco@fmach.it [credits: thanks to Prof. Alberto Montresor]

# **Dictionary: ADT**

### DICTIONARY

% Returns the value associated to key k, if present; returns **none** otherwise

```
OBJECT lookup(OBJECT k)
```

% Associates value v to key kinsert(OBJECT k, OBJECT v)

% Removes the association of key k

remove $(OBJECT \ k)$ 

Note: insert replaces the object associated to the key if already present

# Possible implementations of a dictionary

	Unordered	Ordered	Linked	RB Tree	Ideal
	array	array	$\mathbf{List}$		impl.
insert()	O(1), O(n)	O(n)	O(1), O(n)	$O(\log n)$	O(1)
lookup()	O(n)	$O(\log n)$	O(n)	$O(\log n)$	O(1)
remove()	O(n)	O(n)	O(n)	$O(\log n)$	O(1)

### Ideal implementation: hash tables

- Choose a hash function h that maps each key  $k \in \mathcal{U}$  to an integer h(k)
- The key-value  $\langle k, v \rangle$  is stored in a list at position h(k)
- This vector is called hash table

### Hash table: definitions

- All the possible keys are contained in the universe set  $\mathcal{U}$  of size u
- The table is stored in list  $T[0 \dots m-1]$  with size m
- An hash function is defined as:  $h: \mathcal{U} \to \{0, 1, \dots, m-1\}$



### Hash table: collisions

- When two or more keys in the dictionary have the same hash values, we say that a collision has happened
- Ideally, we want to have hash functions with no collisions



# Direct access tables

- All the possible keys are contained in the universe set  $\mathcal U$  of size u
- The table is stored in list  $T[0 \dots m-1]$  with size m
- An hash function is defined as:  $h: \mathcal{U} \to \{0, 1, \dots, m-1\}$

In some cases: the set  $\mathcal{U}$  is already a (small) subset of  $\mathbb{Z}^+$ 

#### Example: days of the year

### Direct access tables

- We use the identity function h(k) = k as hash function
- We select m = u

#### Problems

- If u is very large, this approach may be infeasible
- If u is large but the number of keys that are actually recorded is much smaller than u = m, memory is wasted

### Perfect hash function

- All the possible keys are contained in the universe set  $\mathcal{U}$  of size u
- The table is stored in list  $T[0 \dots m-1]$  with size m
- An hash function is defined as:  $h: \mathcal{U} \to \{0, 1, \dots, m-1\}$

### Definition

A hash function h is called **perfect** if h is **injective**, i.e.  $\forall k_1, k_2 \in \mathcal{U} : k_1 \neq k_2 \Rightarrow h(k_1) \neq h(k_2)$ 

#### Examples

- Students ASD 2005-2016
   N. matricola in [100.090, 183.864]
   h(k) = k 100.090, m = 83.774
- Studentes enrolled in 2014
   N. matricola in [173.185, 183.864]
   h(k) = k 173.185, m = 10.679

#### Problems

- Universe space is often large, sparse, unknown
- To obtain a perfect hash function is difficult

# Hash functions

#### If collisions cannot be avoided

- Let's try to minimize their number
- We want hash functions that uniform distribute the keys into hash indexes  $[0 \dots m 1]$



we will have to deal with collisions anyway. More on this later...

### Simple uniformity

- Let P(k) be the probability that key k is inserted in the table
- Let Q(i) be the probability that a key ends up in the *i*-th entry of the table

$$Q(i) = \sum_{k \in \mathcal{U}: h(k) = i} P(k)$$

• An hash function h has simple uniformity if:  $\forall i \in [0, \dots, m-1] : Q(i) = 1/m$ 

# Hash functions

To obtain a hash function with simple uniformity, the probability distribution P should be known

#### Example

if  $\mathcal{U}$  is given by real number in [0,1[ and each key has the same probability of being selected, then  $H(k) = \lfloor km \rfloor$  has simple uniformity

In the real world

- The key distribution may unknown or partially known
- Heuristic techniques are used to obtain an approximation of simple uniformity

#### Simple uniformity

- Let P(k) be the probability that key k is inserted in the table
- Let Q(i) be the probability that a key ends up in the *i*-th entry of the table

$$Q(i) = \sum_{k \in \mathcal{U}: h(k) = i} P(k)$$

• An hash function h has simple uniformity if:  $\forall i \in [0, \dots, m-1]: Q(i) = 1/m$ 

# Hash functions: possible implementations

#### Assumption

Each key can be translated in a numerical, non-negative values, by reading their internal representation as a number.

#### Example: string transformation

- ord(c): ordinal binary value of character c in ASCII
- bin(k): binary representation of key k, by concatenating the binary values of its characters
- int(b): numerical value associated to the binary number b
- int(k) = int(bin(k))

### Hash functions: possible implementations (the code)

ord → ascii

a character

Replace the b

that stands for

binary!

representation of

```
def H(in string):
    d = "".join([str(bin(ord(x))) for x in in string]).replace("b","")
    int d = int(d,2)
    return int d
                                L: ord(L) = 76 bin(76) = 0b1001100
L = "Luca"
                                u: ord(u) = 117 bin(117) = 0b1110101
D = "David"
                                c: ord(c) = 99 bin(99) = 0b1100011
C = "Massimiliano"
                                a: ord(a) = 97 bin(97) = 0b1100001
E = "Andrea"
                                0100110001110101010001101100001 -> 1,282,761,569
A = "Alberto"
A1 = "Alan Turing"
people = [L, D, C, E, A, A1]
for p in people:
    print("H('{}')\t=\t{:,}".format(p, H(p)))
H('Luca')
                         1,282,761,569
H('David')
                         293,692,926,308
                 =
H('Massimiliano')
                                  23,948,156,761,864,131,868,341,923,439
H('Andrea')
                         71,942,387,426,657
                 =
H('Alberto')
                         18,415,043,350,787,183
                 =
H('Alan Turing')
                                  39.545.995.566.905.718.680.940.135
                         =
```

### Hash function implementation

So far, we translated strings into big numbers.

**Question for you:** how do we convert these big numbers into values in [0, ..., m-1] where m is the size of the hash table?

H('Luca') 1,282,761,569 = H('David') 293,692,926,308 = 23,948,156,761,864,131,868,341,923,439 H('Massimiliano') = 71,942,387,426,657 H('Andrea') = H('Alberto') 18,415,043,350,787,183 = H('Alan Turing') 39,545,995,566,905,718,680,940,135 =

### Hash function implementation

#### **Division** method

- Let *m* be a odd number (prime)
- $H(k) = int(k) \mod m$

Be careful that: m = 2<sup>i</sup> means to consider the i least significant bits people = [L, D, C, E, A, A1]
prime = 383
for p in people:
 print("{} \t {:,} mod {}\t\t Index: {}".format(p, H(p),prime,my\_hash\_fun(p,prime)))

def H(in string):

int d = int(d,2)

def my hash fun(key str, m = 383):

return int d

h = H(key\_str) hash key = h % m

return hash key

C = "Massimiliano" E = "Andrea" A = "Alberto" A1 = "Alan Turing"

L = "Luca"
D = "David"

Luca	1,282,761,569 mod 383	Index: 351
David	293,692,926,308 mod 383	Index: 345
Massimiliano	23,948,156,761,864,131,868,341,923,439 mod 383	Index: 208
Andrea	71,942,387,426,657 mod 383	Index: 111
Alberto	18,415,043,350,787,183 mod 383	Index: 221
Alan Turing	39,545,995,566,905,718,680,940,135 mod 383	Index: 314

d = "".join([str(bin(ord(x))) for x in in string]).replace("b","")

# Conflicts: separate chaining

### Idea

- The keys with the same value *h* are stored in a monodirectional list / dynamic vector
- The H(k)-th slot in the hash table contains the list/vector associated to k



Another possible method is to look for another place in the hash table where we can put the value (open addressing).

# Separate chaining: complexity

n	Number of keys stored in the hash table
m	Size of the hash table
$\alpha = n/m$	Load factor
I(lpha)	Average number of memory accesses to search a key that is not in the table (insuccess)
S(lpha)	Average number of memory accesses to search a key that is not in the table (success)

Worst case analysis

- All the keys are inserted in a unique list
- insert():  $\Theta(1)$
- lookup(), remove():  $\Theta(n)$

# Separate chaining: complexity

Average case analysis

• Let's assume the hash function has simple uniformity



• Hash function computation:  $\Theta(1)$ , to be added to all searches

all places have the same probability of contain one element

How long the lists are?

 The expected length of a list is equal to α = n/m





alpha is the average length of each list

# Separate chaining: complexity

#### Insuccess

- When searching for a missing key, all the keys in the list must be read
- Expected cost:  $\Theta(1) + \alpha$

#### Success

When searching for a key included in the table, on average half of the keys in the list must be read.
Expected cost: Θ(1) + α/2

What is the meaning of the load factor?

- The cost factor of every operation is influenced by the load factor
- If  $m = O(n), \ \alpha = O(1)$
- In such case, all operations are O(1) in expectation
- If  $\alpha$  becomes too large, the size of the hash table can be doubled through dynamic vectors

### Hash table: rules for hashing objects

Rule: If two objects are equal, then their hashes should be equal

• If you implement \_\_eq\_\_(), then you should implement function \_\_hash\_\_() as well

Rule: If two objects have the same hash, then they are likely to be equal

• You should avoid to return values that generate collisions in your hash function.

Rule: In order for an object to be hashable, it must be immutable

• The hash value of an object should not change over time

[https://www.asmeurer.com/blog/posts/what-happens-when-you-mess-with-hashing-in-python/]

### Hash table: sample code (m = 11)

#### class HashTable:

```
# the table is a list of m empty lists
def __init__(self, m):
    self.table = [[] for i in range(m)]
```

```
#converts a string into an integer (our keys will be strings only)
def H(self, key):
    d = "".join([str(bin(ord(x))) for x in key]).replace("b","")
    int_d = int(d,2)
    return int d
```

```
#gets a string and converts it into a hash-key
def hash_function(self,str_obj):
    #m is inferred from the length of the table
    m = len(self.table)
    h = self.H(str_obj)
    hash_key = h % m
    return hash_key
```

```
#adds a pair (key, value) to the hash table
def insert(self, key, value):
   index = self.hash function(key)
                                                  pair to deal
   self.table[index].append((key, value))
#removes the value associated to key if it exists with collisions
def remove(self, kev):
    index = self.hash function(key)
   for el in self.table[index]:
        if el[0] == key:
           self.table[index].remove(el)
            break
#returns the value associated to key or None
def search(self, key):
   index = self.hash function(key)
   for el in self.table[index]:
       if el[0] == key:
            return el[1]
#converts the table to a string
def str (self):
    return str(self.table)
```

```
if __name__ == "__main_":
    myHash = HashTable(11)
    myHash.insert("Luca",27)
    myHash.insert("David",5)
    myHash.insert("Andrea",15)
    myHash.insert("Alberto",12)
    myHash.insert("Alberto",12)
    myHash.insert("Alan",1)
    print(myHash)
    key = "Luca"
    print("{} -> {}".format(key, myHash.search(key)))
    myHash.remove("Luca")
    key = "Thomas"
    print("{} -> {}".format(key, myHash.search(key)))
    print("{} -> {}".format(key, myHash.search(key)))
    print("{} -> {}".format(key, myHash.search(key)))
    print(myHash)
```

[[('Andrea', 15)], [**('Luca', 27), ('David', 5), ('Alberto', 12)**], [], [], [[, [('Alan', 1)], [], [('Massimiliano', 12)], [], [], [], []]

Luca -> 27 Thomas -> None

[[('Andrea', 15)], [**('David', 5), ('Alberto', 12)**], [], [], [], [('Alan', 1)], [], [('Massimiliano', 12)], [], [], [], []]

#### **SOME CONFLICTS!**

### Hash table: sample code (m = 17)

#### class HashTable:

```
# the table is a list of m empty lists
def __init__(self, m):
    self.table = [[] for i in range(m)]
```

```
#converts a string into an integer (our keys will be strings only)
def H(self, key):
    d = "".join([str(bin(ord(x))) for x in key]).replace("b","")
    int_d = int(d,2)
    return int_d
```

```
#gets a string and converts it into a hash-key
def hash_function(self,str_obj):
    #m is inferred from the length of the table
    m = len(self.table)
    h = self.H(str_obj)
    hash_key = h % m
    return hash_key
```

```
#adds a pair (key, value) to the hash table
def insert(self, kev, value):
   index = self.hash function(key)
   self.table[index].append((key, value))
#removes the value associated to key if it exists
def remove(self, key):
    index = self.hash function(key)
   for el in self.table[index]:
        if el[0] == key:
            self.table[index].remove(el)
            break
#returns the value associated to key or None
def search(self, key):
    index = self.hash function(key)
   for el in self.table[index]:
       if el[0] == key:
            return el[1]
#converts the table to a string
```

def str (self):

return str(self.table)

```
if __name__ == "__main__":
    myHash = HashTable(17)
    myHash.insert("Luca",27)
    myHash.insert("David",5)
    myHash.insert("Andrea",15)
    myHash.insert("Alberto",12)
    myHash.insert("Alberto",12)
    myHash.insert("Alan",1)
    print(myHash)
    key = "Luca"
    print("{} -> {}".format(key, myHash.search(key)))
    myHash.remove("Luca")
    key = "Thomas"
    print("{} -> {}".format(key, myHash.search(key)))
    print("{} -> {}".format(key, myHash.search(key)))
    print("{} -> {}".format(key, myHash.search(key)))
    print(myHash)
```

[[], [], [], [], [], [], [('Alan', 1)], [], [], [('Andrea', 15)], [], [], [('David', 5)], [('Massimiliano', 12)], [], [('Luca', 27)], [('Alberto', 12)]] Luca -> 27 Thomas -> None [[], [], [], [], [], [], [('Alan', 1)], [], [], [('Andrea', 15)], [], [], [('David', 5)], [('Massimiliano', 12)], [], [], [('Alberto', 12)]]

#### **NO CONFLICTS!**

# In python...

Python sets and dict

- Are implemented through hash tables
- Sets are degenerate forms of dictionaries, where there are no values, only keys

Unordered data structures

• Order between keys is not preserved by the hash function; this is why you get unordered results when you print them

# Python built-in: set

Operation		Average case	Worst case
x in S	Contains	O(1)	O(n)
S.add(x)	Insert	O(1)	O(n)
S.remove(x)	Remove	O(1)	O(n)
SIT	Union	O(n+m)	$O(n \cdot m)$
S&T	Intersection	$O(\min(n,m))$	$O(n \cdot m)$
S-T	Difference	O(n)	$O(n \cdot m)$
for x in S	Iterator	O(n)	O(n)
len(S)	Get length	O(1)	O(1)
<pre>min(S), max(S)</pre>	Min, Max	O(n)	O(n)

$$n = \mathtt{len}(\mathtt{S}), m = \mathtt{len}(\mathtt{T})$$

https://docs.python.org/2/library/stdtypes.html#set

# Python built-in: dictionary

Operation		Average case	Worst case	
x in D	Contains	O(1)	O(n)	
D[] =	Insert	O(1)	O(n)	
= D[]	$\operatorname{Lookup}$	O(1)	O(n)	
del D[]	Remove	O(1)	O(n)	
for x in S	Iterator	O(n)	O(n)	
len(S)	Get length	O(1)	O(1)	

$$n = \texttt{len}(\texttt{S}), m = \texttt{len}(\texttt{T})$$

#### Stack

A linear, dynamic data structure, in which the operation "remove" returns (and removes) a predefined element: the one that has remained in the data structure for the least time

#### STACK

% Returns **True** if the stack is empty **boolean** isEmpty()

% Returns the size of the stack int size()

% Inserts v on top of the stack push(OBJECT v) % Removes the top element of the stack and returns it to the caller OBJECT pop()

% Read the top element of the stack, without modifying it OBJECT peek()





Stack Operation	Stack Contents	Return Value
s.isEmpty()	[]	True
s.push(4)	[4]	
<pre>s.push('dog')</pre>	[4,'dog']	
s.peek()	[4,'dog']	'dog'
s.push(True)	<pre>[4,'dog',True]</pre>	
s.size()	<pre>[4,'dog',True]</pre>	3
s.isEmpty()	<pre>[4,'dog',True]</pre>	False
s.push(8.4)	[4,'dog',True,8.4]	
s.pop()	<pre>[4,'dog',True]</pre>	8.4
s.pop()	[4,'dog']	True
s.size()	[4,'dog']	2



Possible uses

- In languages like Python:
  - Compiler: To balance parentheses
  - In the the interpreter: A new activation record is created for each function call
- In graph analysis:
  - To perform visits of the entire graph

### Possible implementations

- Through bidirectional lists
  - reference to the top element
- Through vectors
  - limited size, small overhead







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top

top

Daila	henen	1	- anna
OG	🔨 Last In -	First Out	
Push	I		P
	Data Element	Data Element	
	Data Element	Data Element	
	Data Element	Data Element	
	Data Element	Data Element	
	Data Element	Data Element	
	Stock	Stock	

def my\_func(x): if x <= 2: return x else: print("{} + my\_func({})".format(x,x//4)) return x + my\_func(x//4)





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def my func(x):

top

top







if x <= 2: return x print("{} + my func({})".format(x,x//4)) return x + my func(x//4)



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top



def my func(x): if x <= 2: return x else: print("{} + my func({})".format(x,x//4)) return x + my func(x//4)





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top





def my func(x): if x <= 2: return x print("{} + my func({})".format(x,x//4)) return x + my func(x//4)





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top





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Possible uses

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### Possible implementations

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def my func(x):

if x <= 2:

return x

top

top







Possible uses

- In languages like Python:
  - Compiler: To balance parentheses
  - In the the interpreter: A new activation record is created for each function call
- In graph analysis:
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### Possible implementations

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- Through vectors
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top

top



my func(20)

my func(80)



#### def my func(x): if x <= 2: return x print("{} + my func({})".format(x,x//4)) return x + my func(x//4)

26

Possible uses

- In languages like Python:
  - Compiler: To balance parentheses
  - In the the interpreter: A new activation record is created for each function call
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Possible uses

- In languages like Python:
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f my\_func(x):
 if x <= 2:
 return x
 else:
 print("{} + my\_func({})".format(x,x//4))
 return x + my\_func(x//4)</pre>



Possible uses

- In languages like Python:
  - Compiler: To balance parentheses
  - In the the interpreter: A new activation record is created for each function call
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  - To perform visits of the entire graph



- Through bidirectional lists
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- Through vectors
  - limited size, small overhead



top

top

Push

Last In - First Out

Data Element Data Element

Data Element

Data Element Data Element

Stack

Data Element

Data Element Data Element

Data Element

Data Element

Pop

# Stack: implementation

#### class Stack:

*# initializer, the inner structure is a list* # data is added at the end of the list # for speed def init (self): could have used a deque, self. data = [] linked list .... # returns the length of the stack (size) def len (self): return len(self. data) # returns True if stack is empty def isEmpty(self): return len(self. data) == 0 name == " if main ": # returns the last inserted item of the stack S = Stack()# and shrinks the stack def pop(self): print(S) if len(self. data) > 0: print("Empty? {}".format(S.isEmpty())) return self. data.pop() S.push("Luca") S.push(1) S.push(27)# returns the last inserted element without print(S) # removing it (None if empty) S.push([1,2,3]) def peek(self): if len(self. data) > 0: print("The stack has {} elements".format(len(S))) return self. data[-1] print(S) else: print("Last inserted: {}".format(S.peek())) return None print("Removed: {}".format(S.pop())) print("Stack now:") # adds an element to the stack print(S) def push(self, item): self. data.append(item) Stack([]) # transforms the Stack into a string Empty? True def str (self): Stack([27 | 1 | Luca]) if len(self. data) == 0: The stack has 4 elements return "Stack([])" Stack([[1, 2, 3] | 27 | 1 | Luca]) else: Last inserted: [1, 2, 3] out = "Stack([" + str(self. data[-1]) Removed: [1, 2, 3] for i in range(len(self. data) -2,-1, -1): out += " | " + str(self. data[i]) Stack now: out += "])" Stack([27 | 1 | Luca]) return out

#### STACK

% Returns **True** if the stack is empty **boolean** isEmpty()

% Returns the size of the stack **boolean** size()

% Inserts v on top of the stack push(OBJECT v)

% Removes the top element of the stack and returns it to the caller OBJECT pop()

% Read the top element of the stack, without modifying it OBJECT peek()



### Stack: uses

• Check whether the following sets of parentheses are balanced

```
• { { ( [ ] [ ] ) } ( ) }
• [ [ { { ( ( ) ) } ] ]
• [ [ ] [ ] ( ) { }
• ( [ ) ]
• ( ( ( ) ] ) )
• [ { ( ( ( ) ] ) )
• [ { ( ( ) ] }
```

### Stack: exercise

Ideas on how to implement **par\_checker** using a Stack?

Simplifying assumption: only characters allowed in input are "{ [ ( ) ] }"

#### **Possible solution:**

Loop through the input string and...

- push opening parenthesis to stack
- when analyzing a closing parenthesis, pop one element from the stack and compare: if matching keep going, else return False

#### **Desired output**



### Stack: exercise

```
def par match(open p, close p):
    openers = "{[("
    closers = "}])"
    if openers.index(open p) == closers.index(close p):
        return True
    else:
        return False
def par checker(parString):
    s = Stack()
    for symbol in parString:
        if symbol in "([{":
            s.push(symbol)
        else:
            if s.isEmpty():
                return False
            else:
                top = s.pop()
                if not par match(top,symbol):
                    return False
    return s.isEmpty()
```

#### **Desired output**

{{([][])}()}	balanced:	True
[{()]	balanced:	False
{[(())][{[]}]}	balanced:	True
{[(())][{[]}]	balanced:	False

# Queue: First in, first out queue (FIFO)



### Queue

A linear, dynamic data structure, in which the operation "remove" returns (and removes) a predefined element: the one that has remained in the data structure for the longest time)

### QUEUE

% Returns **True** if queue is empty **boolean** isEmpty()

% Returns the size of the queue int size()

% Inserts v at the end of the queue enqueue(OBJECT v)

% Extracts q from the beginning of the queue OBJECT dequeue()

% Reads the element at the top of the queue OBJECT top()

# Queue: example



QUEUE	
% Returns <b>True</b> if queue is empty <b>boolean</b> isEmpty()	% Extracts $q$ from the beginning of the queue
% Returns the size of the queue int size()	OBJECT dequeue() % Reads the element at the top of
% Inserts $v$ at the end of the queue enqueue (OBJECT $v$ )	the queue OBJECT top()

Queue Operation	Queue Contents	Return Value
q.isEmpty()	[]	True
q.enqueue(4)	[4]	
q.enqueue('dog')	['dog',4]	
q.enqueue(True)	[True,'dog',4]	
q.size()	[True,'dog',4]	3
q.isEmpty()	[True,'dog',4]	False
q.enqueue(8.4)	[8.4,True,'dog',4]	
q.dequeue()	<pre>[8.4,True,'dog']</pre>	4
q.dequeue()	[8.4,True]	'dog'
q.size()	[8.4,True]	2

# Queue: uses and implementation





# Queue: as a list (with deque)

#### from collections import deque

class Queue:

```
def init (self):
    self. data = deque()
def len (self):
    return len(self. data)
def str (self):
    return str(self. data)
def isEmpty(self):
    return len(self. data) == 0
def top(self):
    if len(self. data) > 0:
       return self. data[-1]
def enqueue(self, item):
    self. data.appendleft(item)
def dequeue(self):
    if len(self. data) > 0:
       return self. data.pop()
```

```
name == " main ":
if
   0 = 0ueue()
    print(Q)
    print("TOP: {}".format(Q.top()))
    print(Q.isEmpty())
    Q.enqueue(4)
    Q.enqueue('dog')
    0.engueue(True)
    print(Q)
    print("Size: {}".format(len(Q)))
    print(Q.isEmpty())
    0.engueue(8.4)
    print("Removing: '{}'".format(Q.dequeue()))
    print("Removing: '{}'".format(Q.dequeue()))
    print(0)
    print("Size: {}".format(len(Q)))
deque([])
TOP now: None
True
deque([True, 'dog', 4])
Size: 3
False
Removing: '4'
```

QUEUE

% Returns  $\mathbf{True}$  if queue is empty  $\mathbf{boolean}$  is Empty()

% Returns the size of the queue  ${\tt int\ size}()$ 

% Inserts v at the end of the queue enqueue(OBJECT v)

% Extracts q from the beginning of the queue OBJECT dequeue()
% Beads the element at the top of

% Reads the element at the top of the queue OBJECT top()

Not very interesting implementation.

```
Just pay attention to the case when
the Queue is empty in top and
dequeue
```

Makes use of efficient deque object that provides ~ O(1) push/pop https://docs.python.org/3.7/library/collections.html#collections.deque

Removing: 'dog'

Size: 2

deque([8.4, True])

# Queue as a circular list

- Implementation based on the modulus operation
- Pay attention to overflow problems (full queue)

















skipping a few typing steps...

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skipping a few typing/reading steps...

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### Queue as a circular list: exercise

Implement the CircularQueue data structure

(without going to the next slide...)

QUEUE

% Returns **True** if queue is empty<br/>**boolean** isEmpty()% Extracts q from the beginning<br/>of the queue% Returns the size of the queue<br/>**int size()**OBJECT dequeue()% Inserts v at the end of the<br/>queue% Reads the element at the top of<br/>the queue<br/>OBJECT top()

enqueue(OBJECT v)

		QUEUE	
Queue as a circular lis	st: the code	% Returns True if queue is empty boolean isEmpty() % Returns the size of the queue	<pre>y % Extracts q from the beginning of the queue OBJECT dequeue()</pre>
<pre>class CircularQueue: definit(self, N): selfdata = [None for i in range(N)] selfhead = 0</pre>		<pre>int size() % Inserts v at the end of the queue enqueue(OBJECT v)</pre>	% Reads the element at the top of the queue OBJECT top()
<pre>selftail = 0 selfsize = 0 selfmax = N  def top(self):     if selfsize &gt; 0:</pre>			None [0] H < Head [1] E [2] L [3] L [4] 0 [5]
<pre>return selfdata[selfhead]  def dequeue(self):     if selfsize &gt; 0:         ret = selfdata[selfhead]         selfhead = (selfhead + 1) % selfmax         selfsize -= 1         return ret</pre>	<pre>ifname == "main"     CQ = CircularQueue(10     print(CQ.dequeue())     text = "HELLO W"     text2 = "IKIPEDIA"     for t in text:</pre>	: ))	[6] W [7] None < Tail [8] None [9] None [0] H [1] E [2] L [3] L
<pre>def enqueue(self, item): if selfmax &gt; selfsize: selfdata[selftail] = item selftail = (selftail + 1) % selfmax selfsize += 1 else: raise Exception("The queue is full. Cannot add to it")</pre>	CQ.enqueue(t) print(CQ) out_txt = "" for i in range(6): out_txt += str(CQ)	).dequeue())	[5] [6] W < Head [7] None < Tail [8] None [9] None HELLO [0] P [1] E
<pre>deflen(self):     return selfsize  def isEmpty(self):     return selfsize == 0  defstr(self):     out = ""</pre>	<pre>print(CQ) print(out_txt) for t in text2:         CQ.enqueue(t) print(CQ)</pre>		[2] D [3] I [4] A [5] < Tail [6] W < Head [7] I [8] K [9] I
<pre>if len(selfdata) == 0: return "" for i in range(len(selfdata)): out += "[{}] ".format(i) + str(selfdata[i]) if i == selfhead: out += " &lt; Head" if i == selftail: out += " &lt; Tail" out += "\n"</pre>	<pre>while not CQ.isEmpty( out_txt += str(CQ print(out_txt) print(CQ)</pre>	): dequeue())	HELLO WIKIPEDIA [0] P [1] E [2] D [3] I [4] A [5] < Head < Tail [6] W [7] I [8] K
return out			[9] I

Consider the following code (where s is a list of n elements). What is its complexity? Note: res is a string!

```
def reverse(s):
    n = len(s)-1
    res = ""
    while n >= 0:
        res = res + s[n]
        n -= 1
    return res
```

Consider the following code (where s is a list of n elements). What is its complexity? **Note: res is a string!** 

```
def reverse(s):
    n = len(s)-1
    res = ""
    while n >= 0:
        res = res + s[n]
        n -= 1
    return res
```



#### • *n* string sums

• Each sum copies all the characters in a new string





Consider the following code (where s is a list of n elements). What is its complexity?

```
def reverse(s):
    res = []
    for c in s:
        res.insert(0, c)
    return "".join(res)
```



Consider the following code (where s is a list of n elements). What is its complexity?

```
def reverse(s):
    res = []
    for c in s:
        res.insert(0, c)
    return "".join(res)
```



### Complexity: $\Theta(n^2)$

- n list inserts
- Each insert moves all characters one position up in the list

Consider the following code (where s is a list of n elements). What is its complexity?

```
def reverse(s):
    n = len(s)-1
    res = []
    while n >= 0:
        res.append(s[n])
        n -= 1
    return "".join(res)
```

Consider the following code (where s is a list of n elements). What is its complexity?

```
def reverse(s):
  n = len(s)-1
  res = []
  while n >= 0:
    res.append(s[n])
    n -= 1
  return "".join(res)
```

Better solution

```
def reverse(s):
    return s[::-1]
```



Complexity:  $\Theta(n)$ 

- n list append
- Each append has an amortized cost of O(1)

Note that: "".join(res) has complexity O(n)

Consider the following code (where L is a list of n elements). What is its complexity?

```
def deduplicate(L):
    res=[]
    for item in L:
        if item not in res:
            res.append(item)
    return res
```



Consider the following code (where L is a list of n elements). What is its complexity?

```
def deduplicate(L):
    res=[]
    for item in L:
        if item not in res:
            res.append(item)
    return res
```



#### Complexity: $\Theta(n^2)$

- n list append
- *n* checks whether an element is already present
- Each check costs O(n)

Consider the following code (where L is a list of n elements). What is its complexity?

```
def deduplicate(L):
    res=[]
    present=set()
    for item in L:
        if item not in present:
            res.append(item)
            present.add(item)
    return res
```

Consider the following code (where L is a list of n elements). What is its complexity?

```
def deduplicate(L):
    res=[]
    present=set()
    for item in L:
        if item not in present:
            res.append(item)
            present.add(item)
    return res
```

Other possibility – destroy original order

```
def deduplicate(L):
    return list(set(L))
```

Complexity:  $\Theta(n)$ 



- n list append
- *n* checks whether an element is already present
- Each check costs O(1)