Common data structures

Implementation and complexity

What's a data structure?

(Wikipedia) "a collection of **data values**, the **relationships** among them, and the functions or **operations** that can be applied to the data"

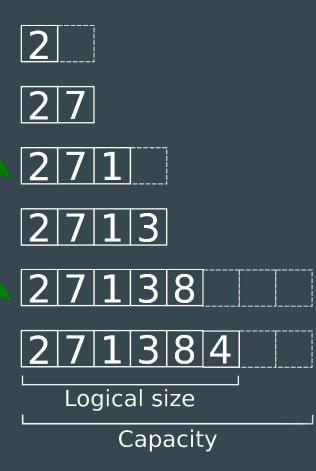
- an **abstraction** of data that is easy for a programmer to work with
- contains more than data: the data is organized in a specific way
- well-defined operations can be applied to the data
 - → it is important to know what data structures exist and which operations can be applied on them

Dynamic arrays (Python lists)

Goal: quickly access indexed items in a container and append new ones (or remove the last one)

Implementation:

- the language uses more room than needed
- while there is room, appending costs nothing
- when there is no more room, create a new array with more room and copy everything



Dynamic arrays (Python lists)

What's the complexity of adding a new item?

if you reach the capacity c_1 and extend the size to $c_2 = 2 * c_1$, the cost is c_1 only once but then the cost will be const for the next c_1 append operations

 \rightarrow on average, the cost is O(1)

What's the complexity of accessing an item?

O(1), position in RAM deduced from its index

2

2 7

271

2 7 1 3

27138

271384

Logical size

Capacity

Dynamic arrays: in Python and C++

Common operations:

эншнон орегацонs: 	Python	C++	Complexity
Access the i-th item	arr[i]	arr[i]	0(1)
Add v at the end	arr.append(v)	arr.push_back(v)	0(1) avg 0(n) max
Insert v at position i	arr.insert(i, v)	arr.insert(i, v)	0(n)
Find the position of v	arr.index(v)	std::find()	0(n)

When to use:

There are all kinds of use cases. If you often need to perform operations that are not O(1), check if another data structure matches your needs

Quick implementation cheat sheet - Python array

	Instruction	Complexity
Create a new one	L = list(), L = []	0(1)
Access the i-th item	arr[i]	0(1)
Add v at the end	arr.append(v)	0(1) avg 0(n) max
Insert v at position i	arr.insert(i, v)	0(n)
Find the position of v	arr.index(v) O(n)	

Stacks & Queues



Quick implementation cheat sheet - Python deque

Python: deque in collections : an implementation of double-ended queue

	Python	Complexity
Create a deque	dq = deque()	O(1)
Access the i-th element	dq[i]	O(n)
Add v at the head	dq.appendleft(v)	O(1)
Remove the head and put it in v	<pre>v = dq.popleft()</pre>	O(1)
Add v at the tail	dq.append(v)	O(1)
Remove the tail and put it in v	v = dq.pop()	O(1)

Double-ended queues - usage

When to use:

whenever you need a **queue** (e.g in *breadth-first search* algorithm)

When not to use:

- if you want to access very often elements in the middle of the sequence
- you don't need it when inserting and removing elements only on one side (e.g a a stack). Just don't do that on the left side

Dictionaries / hash maps

Goals:

- associate **values** to **keys**
- retrieve the value associated to a key in O(1) time
- → unlike real-life dictionaries, these ones are **not ordered**

(in real life, finding a word in a dictionary is O(log n) unless you have forgotten the alphabetical order)

Dictionaries / hash maps

The underlying structure is called a **hash table**

- a hash function turns the keys into an index
- the keys and values are stored in the data structure based on this index
- multiple strategies exist to manage collisions (cf next slides)

Example of hash function: sum of the ASCII codes for a string, modulo 42

"hello" \rightarrow 28

"world" \rightarrow 6

"INSAlgo" \rightarrow 33

Dictionaries / hash maps - separate chaining

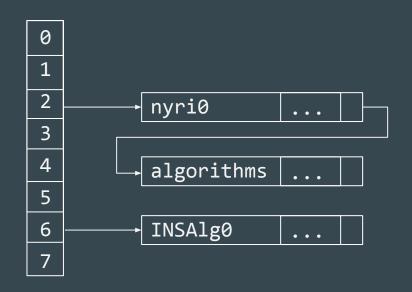
(key, value) couples where keys have the same hash are stored in a common data structure

These structures are kept small enough to have O(1) time access. When too filled, the table is re-created bigger

Often used:

- linked lists
- trees

With sum of ASCII codes modulo 8, and using linked lists:



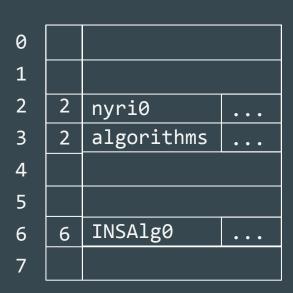
Dictionaries / hash maps - Open adressing

- No other data structure behind the array
- Collisions are solved with *probing*
- When the fill rate becomes too big, copy the data in a bigger hash table

Linear probing = put at next cell available

Randomized probing: follow a random sequence which seed is given by the hash

 \rightarrow used in Python



(sum of ASCII codes modulo 8, open adressing with linear probing)

Dictionaries / hash maps - in Python and C++

Python: open adressing with randomized probing (congruential RNG)

initial size 8, resized when ²/₃ full

C++: unordered map is a hash table, map a red-black tree

(we will talk about trees later in the year)

Common operations:

	Python	C++ unordered_map	Complexity
Find or set the value associated to key	dic[key]	dic[key]	0(1) avg 0(n) max
Check if key exists in the dictionary	key in dic	dic.find(key)	0(1) avg 0(n) max

Dictionaries / hash maps - in Python and C++

When to use:

Whenever you need to associate a value to a key.

In Python, the ease of use and high performance make the dictionaries a **very powerful tool**.

When not to use:

When your keys are 0, 1, ..., n You're better than that.

Quick implementation cheat sheet - Python dict

Python: dict, built in.

Interesting variation : defaultdict in collections

	Python	Complexity
Create a deque	<pre>d = dict(), d = {}</pre>	O(1)
Access an element	d[key]	Pseudo O(1)
Put an element	d[key] = value	Pseudo O(1)
Find if a key is in dict	key in d	Pseudo O(1)
Remove a key	del d[key]	Pseudo O(1)

Sets

Goals:

- store unique values
- quickly check if a value is in the set or not

Two common implementations:

- tree-based sets are ordered
- hash sets are faster but unordered (they're basically hash tables without values)

Sets - in Python and C++

Python: set is a hash set

C++: unordered_set is a hash table, set a tree

Common operations:	Python	C++ set and unordered_set	Complexity	Complexity - C++ set
Add the value v to the set	s.add(v)	s.emplace(v)	0(1) avg 0(n) max	O(log n)
Remove the value v from the set	s.remove(v)	s.erase(v)	0(1) avg 0(n) max	O(log n)
Check if v is in the set	v in s	s.find(v)	0(1) avg 0(n) max	O(log n)

Quick implementation cheat sheet - Python set

Python: set, built in : implemented as a hash set

	Python	Complexity
Create a set	S = set()	O(1)
Access a specific element	IMPOSSIBLE	
Put an element	s.add(element)	Pseudo O(1)
Find if an element is in a set	Element in s	Pseudo O(1)
Remove an element	s.remove(element)	Pseudo O(1)

Set arithmetics in Python

		Average complexity
s1 <= s2, s1 < s2	check if s1 is a [proper] subset of s2	O(n1)
s1 >= s2, s1 > s2	check if s1 is a [proper] superset of s2	O(n2)
s1 s2 sk	union of s1, s2,, sk	O(n1 + n2 + + sk)
s1 & s2 & & sk	intersection of s1, s2, …, sk	O(min(n1, n2,, sk))
s1 - s2	all elements of s1 that are not in s2	O(n1)
s1 ^ s2	all elements of s1 or s2 but not both	O(n1 + n2)

Sets too are cool:)

Sets - in Python and C++

When to use:

- to mark values already seen
- to keep a collection of unordered unique elements

When not to use:

• to make coffee (you just can't)

Python - python.collections classes

A default Python module

- collections.defaultdict(default_factory) : if an element doesn't exist, create it using the default_factory function
- collections.orderedDict: a dict which remembers the order of creation of its elements
- collections.deque: an implementation of a double-ended queue
- collections.Counter : counts occurrences in a list

To be continued...

You probably want to hear about red-black trees, heaps (priority queues), etc...

Well, see you in 2019! ... this was last year, see you in 2020 perhaps ;)...

2020 note : see you in 2021 if the world hasn't collapsed!

Slides: Louis Sugy, Arthur Tondereau for INSAlgo. Updated by Louis Gombert Schema of dynamic arrays: Wikipedia