

Lecture 2: Data Structures

Data Structure Interfaces

- A **data structure** is a way to store data, with algorithms that support **operations** on the data
- Collection of supported operations is called an **interface** (also API or ADT)
- Interface is a **specification**: what operations are supported (the problem!)
- Data structure is a **representation**: how operations are supported (the solution!)
- In this class, two main interfaces: **Sequence** and **Set**

Sequence Interface (L02, L07)

- Maintain a sequence of items (order is **extrinsic**)
- Ex: $(x_0, x_1, x_2, \dots, x_{n-1})$ (zero indexing)
- (use n to denote the number of items stored in the data structure)
- Supports sequence operations:

Container	<code>build(X)</code> <code>len()</code>	given an iterable X , build sequence from items in X return the number of stored items
Static	<code>iter_seq()</code> <code>get_at(i)</code> <code>set_at(i, x)</code>	return the stored items one-by-one in sequence order return the i^{th} item replace the i^{th} item with x
Dynamic	<code>insert_at(i, x)</code> <code>delete_at(i)</code> <code>insert_first(x)</code> <code>delete_first()</code> <code>insert_last(x)</code> <code>delete_last()</code>	add x as the i^{th} item remove and return the i^{th} item add x as the first item remove and return the first item add x as the last item remove and return the last item

- Special case interfaces:

stack	<code>insert_last(x)</code> and <code>delete_last()</code>
queue	<code>insert_last(x)</code> and <code>delete_first()</code>

Set Interface (L03-L08)

- Sequence about **extrinsic** order, set is about **intrinsic** order
- Maintain a set of items having **unique keys** (e.g., item x has key $x.key$)
- (Set or multi-set? We restrict to unique keys for now.)
- Often we let key of an item be the item itself, but may want to store more info than just key
- Supports set operations:

Container	<code>build(X)</code> <code>len()</code>	given an iterable x , build sequence from items in x return the number of stored items
Static	<code>find(k)</code>	return the stored item with key k
Dynamic	<code>insert(x)</code> <code>delete(k)</code>	add x to set (replace item with key $x.key$ if one already exists) remove and return the stored item with key k
Order	<code>iter_ord()</code> <code>find_min()</code> <code>find_max()</code> <code>find_next(k)</code> <code>find_prev(k)</code>	return the stored items one-by-one in key order return the stored item with smallest key return the stored item with largest key return the stored item with smallest key larger than k return the stored item with largest key smaller than k

- Special case interfaces:
 - dictionary** | set without the Order operations
- In recitation, you will be asked to implement a Set, given a Sequence data structure.

Array Sequence

- Array is great for static operations! `get_at(i)` and `set_at(i, x)` in $\Theta(1)$ time!
- But not so great at dynamic operations...
- (For consistency, we maintain the invariant that array is full)
- Then inserting and removing items requires:
 - reallocating the array
 - shifting all items after the modified item

Data Structure	Operation, Worst Case $O(\cdot)$				
	Container	Static	Dynamic		
	<code>build(X)</code>	<code>get_at(i)</code> <code>set_at(i, x)</code>	<code>insert_first(x)</code> <code>delete_first()</code>	<code>insert_last(x)</code> <code>delete_last()</code>	<code>insert_at(i, x)</code> <code>delete_at(i)</code>
Array	n	1	n	n	n

Linked List Sequence

- Pointer data structure (this is **not** related to a Python “list”)
- Each item stored in a **node** which contains a pointer to the next node in sequence
- Each `node` has two fields: `node.item` and `node.next`
- Can manipulate nodes simply by relinking pointers!
- Maintain pointers to the first node in sequence (called the head)
- Can now insert and delete from the front in $\Theta(1)$ time! Yay!
- (Inserting/deleting efficiently from back is also possible; you will do this in PS1)
- But now `get_at(i)` and `set_at(i, x)` each take $O(n)$ time... :(
- Can we get the best of both worlds? Yes! (Kind of...)

Data Structure	Operation, Worst Case $O(\cdot)$				
	Container	Static	Dynamic		
	<code>build(X)</code>	<code>get_at(i)</code> <code>set_at(i, x)</code>	<code>insert_first(x)</code> <code>delete_first()</code>	<code>insert_last(x)</code> <code>delete_last()</code>	<code>insert_at(i, x)</code> <code>delete_at(i)</code>
Linked List	n	n	1	n	n

Dynamic Array Sequence

- Make an array efficient for **last** dynamic operations
- Python “list” is a dynamic array
- **Idea!** Allocate extra space so reallocation does not occur with every dynamic operation
- **Fill ratio:** $0 \leq r \leq 1$ the ratio of items to space
- Whenever array is full ($r = 1$), allocate $\Theta(n)$ extra space at end to fill ratio r_i (e.g., $1/2$)
- Will have to insert $\Theta(n)$ items before the next reallocation
- A single operation can take $\Theta(n)$ time for reallocation
- However, any sequence of $\Theta(n)$ operations takes $\Theta(n)$ time
- So each operation takes $\Theta(1)$ time “on average”

Amortized Analysis

- Data structure analysis technique to distribute cost over many operations
- Operation has **amortized cost** $T(n)$ if k operations cost at most $\leq kT(n)$
- “ $T(n)$ amortized” roughly means $T(n)$ “on average” over many operations
- Inserting into a dynamic array takes $\Theta(1)$ **amortized** time
- More amortization analysis techniques in 6.046!

Dynamic Array Deletion

- Delete from back? $\Theta(1)$ time without effort, yay!
- However, can be very wasteful in space. Want size of data structure to stay $\Theta(n)$
- **Attempt:** if very empty, resize to $r = 1$. Alternating insertion and deletion could be bad...
- **Idea!** When $r < r_d$, resize array to ratio r_i where $r_d < r_i$ (e.g., $r_d = 1/4$, $r_i = 1/2$)
- Then $\Theta(n)$ cheap operations must be made before next expensive resize
- Can limit extra space usage to $(1 + \varepsilon)n$ for any $\varepsilon > 0$ (set $r_d = \frac{1}{1+\varepsilon}$, $r_i = \frac{r_d+1}{2}$)
- Dynamic arrays only support dynamic **last** operations in $\Theta(1)$ time
- Python List `append` and `pop` are amortized $O(1)$ time, other operations can be $O(n)$!
- (Inserting/deleting efficiently from front is also possible; you will do this in PS1)

Data Structure	Operation, Worst Case $O(\cdot)$				
	Container	Static	Dynamic		
	build(X)	get_at(i) set_at(i, x)	insert_first(x) delete_first()	insert_last(x) delete_last()	insert_at(i, x) delete_at(i)
Array	n	1	n	n	n
Linked List	n	n	1	n	n
Dynamic Array	n	1	n	$1_{(a)}$	n

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