# Elementary Data Structures and Hash Tables 

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■ Common concepts and notation
■ Stacks
■ Queues

■ Linked lists

■ Trees

- Direct-access tables

■ Hash tables

- A data structure is a way to organize and store information
- to facilitate access, or for other purposes


## Concepts

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■ A data structure stores data and possibly meta-data

- e.g., a heap needs an array $A$ to store the keys, plus a variable $A$. heap-size to remember how many elements are in the heap

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- Interface
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- Push $(S, x)$ pushes the value $x$ onto the stack $S$
- Pop(S) extracts and returns the value on the top of the stack $S$

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- Implementation
- using an array
- using a linked list
- ...


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StACK-Empty(S)
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2 return TRUE
3 else return FALSE
```

```
Push(S,x)
S.top = S.top + 1
2 S[S.top] = x
```

Pop(S)
1 if Stack-Empty (S) error "underflow"
else S.top $=$ S.top -1
return $S[S$. top + 1]

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■ Implementation

- $Q$ is an array of fixed length $Q$. length
- i.e., $Q$ holds at most $Q$.length elements
- enqueueing more than $Q$ elements causes an "overflow" error
- Q.head is the position of the "head" of the queue
- Q.tail is the first empty position at the tail of the queue

```
Enqueue(Q,x)
1 if Q.queue-full
2 error "overflow"
3 else Q[Q.tail] = x
if Q.tail < Q.length
5 Q.tail = Q.tail +1
else Q.tail = 1
7 if Q.tail == Q.head
8 Q.queue-full = TRUE
    Q.queue-empty = FALSE
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Q.head


- Interface
- List-Insert $(L, x)$ adds element $x$ at beginning of a list $L$
- List-Delete( $(L, x)$ removes element $x$ from a list $L$
- List-Search $(L, k)$ finds an element whose key is $k$ in a list $L$
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- Implementation
- a doubly-linked list
- each element $x$ has two "links" $x$.prev and $x$. next to the previous and next elements, respectively
- each element $x$ holds a key $x$.key
- it is convenient to have a dummy "sentinel" element L.nil


## Linked List With a "Sentinel"

```
LIst-Init(L)
1 L.nil.prev = L.nil
2 L.nil.next = L.nil
```


## List-Insert $(L, x)$

1 x.next $=$ L.nil.next
2 L.nil.next.prev $=x$
L.nil.next $=x$
x.prev $=$ L.nil

List-Search $(L, k)$

```
x = L.nil.next
```

while $x \neq$ L.nil $\wedge x$.key $\neq k$
$x=x . n e x t$
return $x$

Trees

■ Structure

- fixed branching
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- Implementation
- for each node $x \neq$ T.root, $x$.parent is $x$ 's parent node
- fixed branching:
e.g., x.left-child and x.right-child in a binary tree
- unbounded branching:
$x$. left-child is x's first (leftmost) child $x$.right-sibling is $x$ closest sibling to the right

Complexity

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$\underline{\underline{\text { Algorithm Complexity }}}$

Algorithm Complexity

## Stack-Емрту

Complexity

| Algorithm | Complexity |
| :--- | :---: |
| STACK-EMPTY | $O(1)$ |
| PUSH |  |


| Algorithm | Complexity |
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| STACK-Empty | $O(1)$ |
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| Pop | $O(1)$ |
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| Dequeue | $O(1)$ |
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| List-Delete | $O(1)$ |
| List-SeArch |  |


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| LISt-InSERT | $O(1)$ |
| LISt-DeLete | $O(1)$ |
| LISt-SEARCH | $\Theta(n)$ |

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- hash tables

Direct-Address Table

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- an array $T$ of size $M$
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Direct-Address-Insert $(T, k)$
1 T[k] = TRUE

Direct-Address-Delete $(T, k)$
$1 T[k]=$ FALSE

> Direct-Address-Search $(T, k)$
> 1 return $T[k]$

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- the represented set is typically much smaller than |U|
- i.e., a direct-address table usually wastes a lot of space
- Can we have the benefits of a direct-address table but with a table of reasonable size?


# Hash Table 

- Idea
- use a table $T$ with $|T| \ll|U|$
- map each key $k \in U$ to a position in $T$, using a hash function

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| $\operatorname{Hash}-\operatorname{Insert}(T, k)$ | $\operatorname{Hash}-\operatorname{Delete}(T, k)$ |
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| Hash-Insert $(T, k)$ | Hash-Delete $(T, k)$ |
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| 1 | $T[h(k)]=$ True |

> HASH-SEARCH $(T, k)$
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Are these algorithms correct? No!
What if two distinct keys $k_{1} \neq k_{2}$ collide? (I.e., $h\left(k_{1}\right)=h\left(k_{2}\right)$ )

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## Analysis

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■ So, given $n$ distinct keys, the expected length $n_{i}$ of the linked list at position $i$ is

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■ We further assume that $h(k)$ can be computed in $O(1)$ time
■ Therefore, the complexity of Chained-HAsh-SeArch is

$$
\Theta(1+\alpha)
$$

Open-Address Hash Table


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Hash-Insert $(T, k)$
$1 j=h(k)$
2 for $i=1$ to $T$.length
3 if $T[j]==$ NIL $T[j]=k$ return $j$
elseif $j<T$.length
$j=j+1$
else $j=1$
9 error "overflow"

# Open-Addressing (2) 

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■ When a collision occurs, we simply find another free cell in $T$
■ A sequential "probe" may not be optimal

- can you figure out why?

```
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■ Notice that $h(k, \cdot)$ must be a permutation

- i.e., $h(k, 1), h(k, 2), \ldots, h(k,|T|)$ must cover the entire table $T$

