

Heap

- The *heap property*:
each node is more extreme (greater or less) than each of its children
- + Shape property
 - Binary heap (...) (by default for us)
 - Binomial heap
a forest of binomial trees satisfying the heap property
 - Fibonacci heap
a collection of trees satisfying the heap property

Binary heap

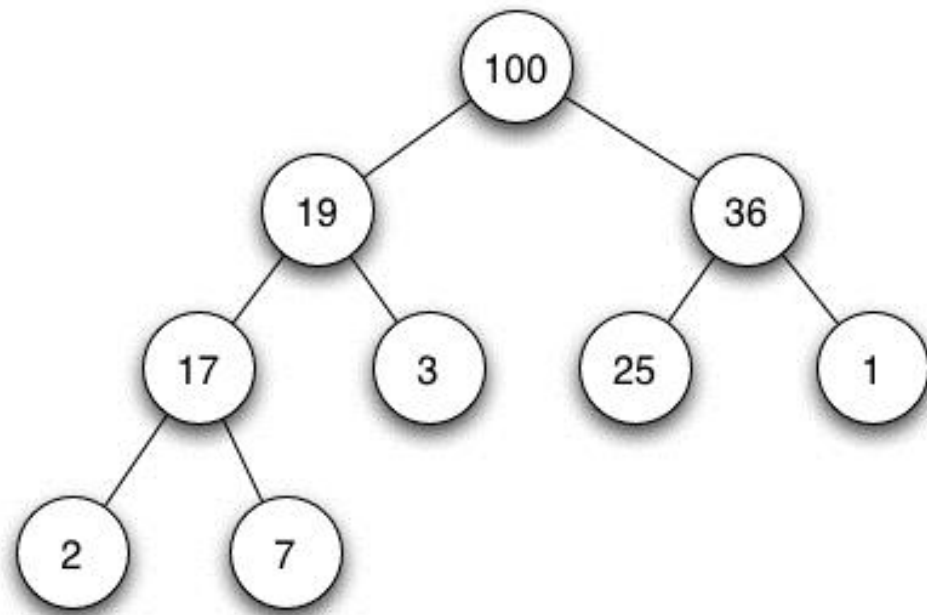
A binary tree with two additional constraints:

- The *shape property*
(almost) complete
- The *heap property*:
each node is more extreme (greater or less) than
each of its children or equal

NO ordering of siblings

Convention

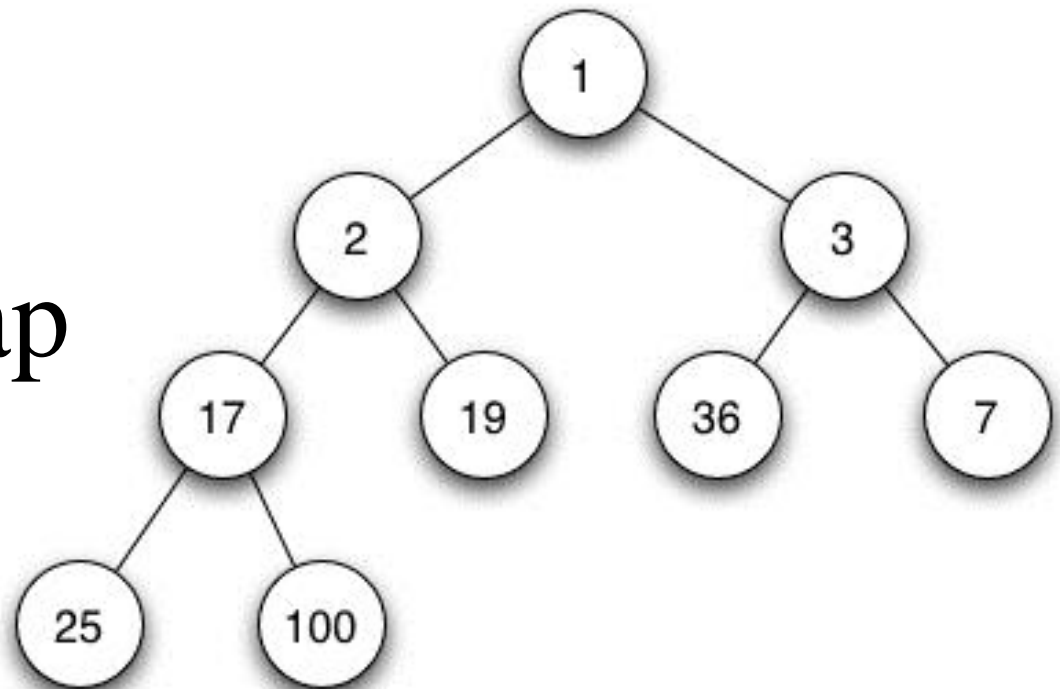
During these classes, the term heap will refer to max binary heap, when not explicitly specified otherwise.



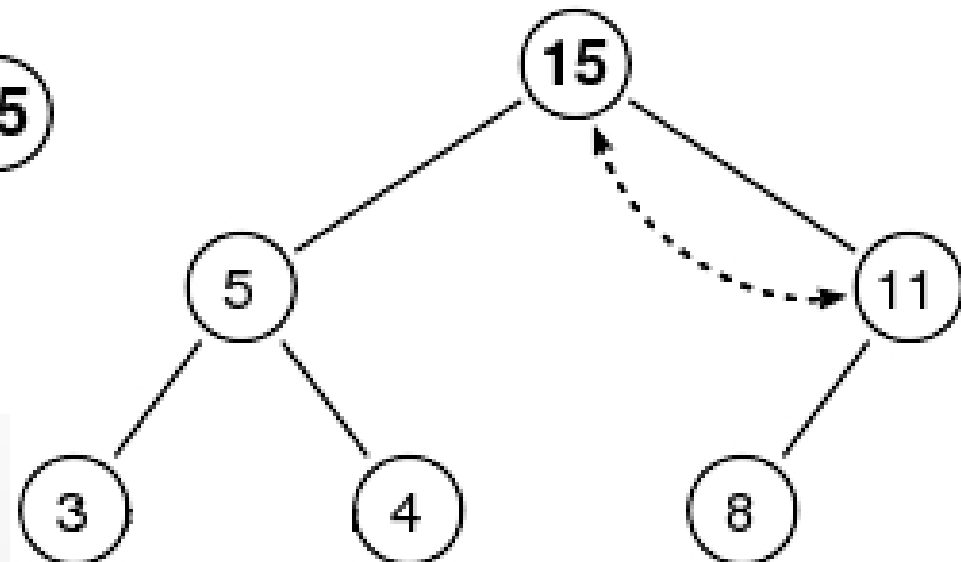
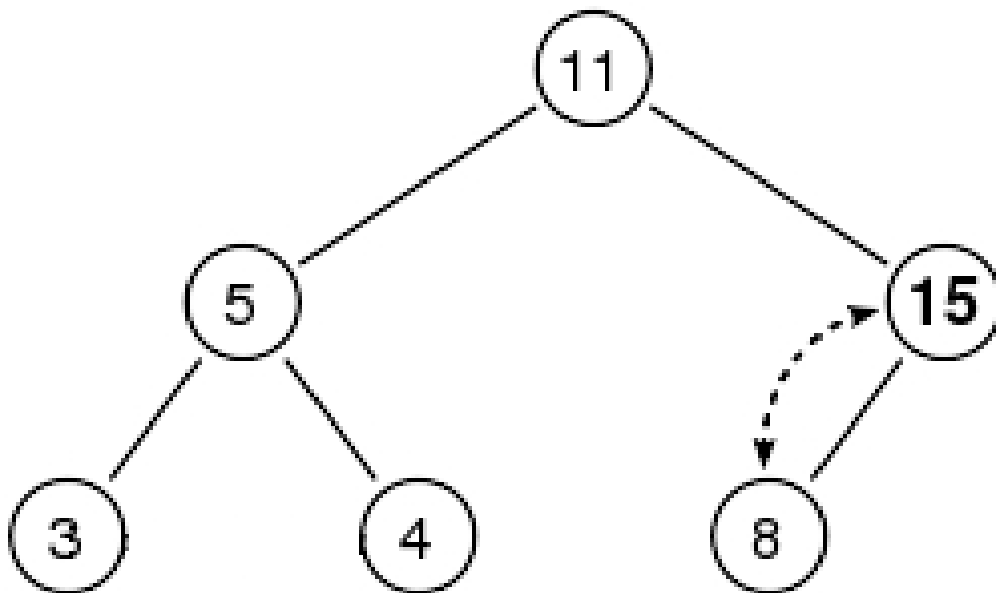
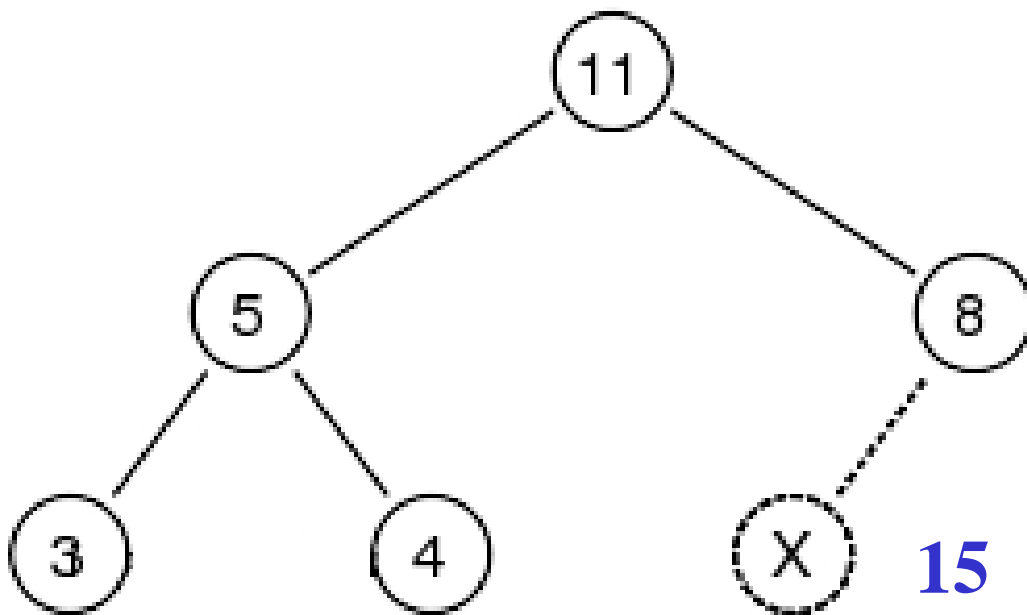
Max binary heap

by default, for us

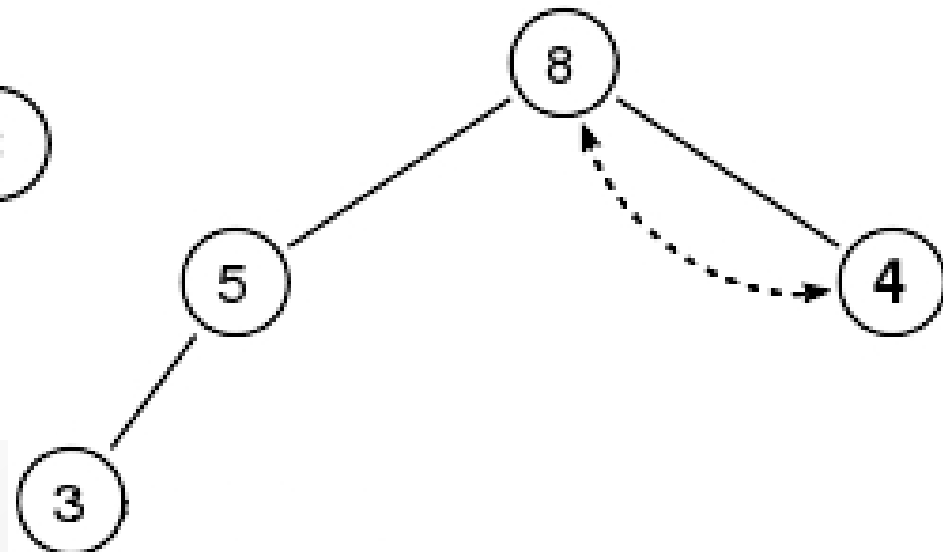
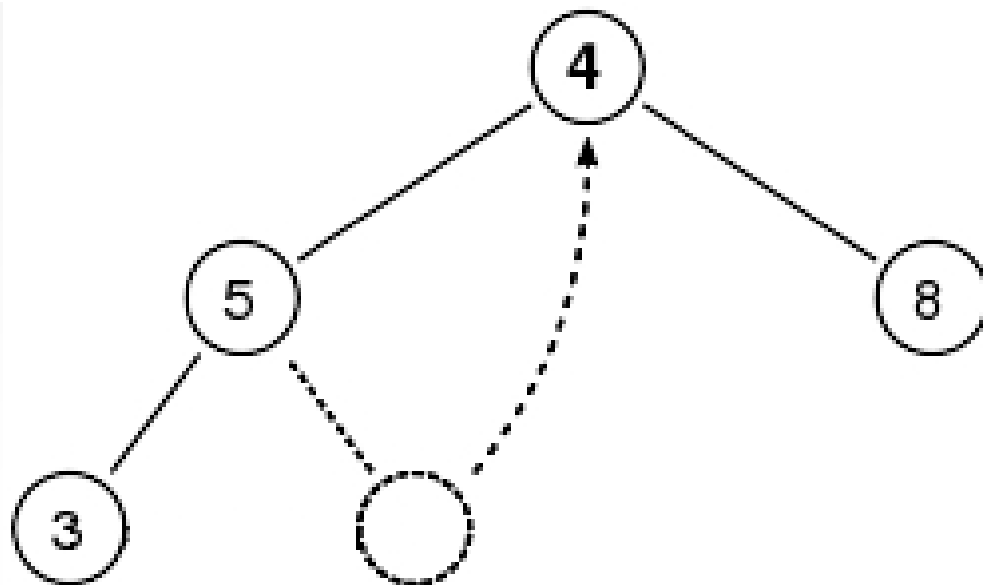
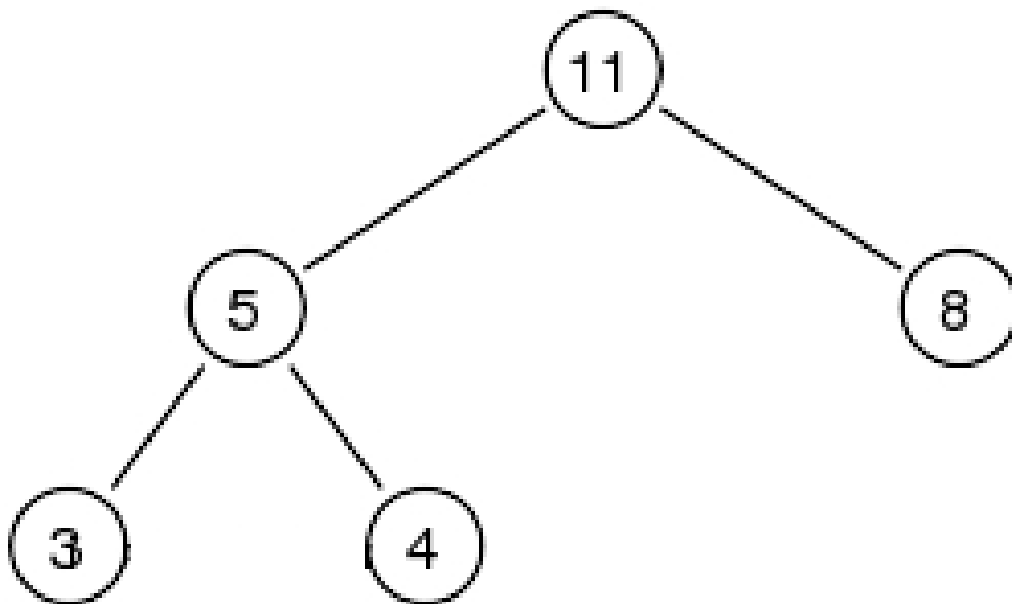
Min binary heap



Heap - insertion



Heap – delete root



Binary heap

Max binary heap

getMax	$O(1)$
insert	$O(\log n)$
deleteMax	$O(\log n)$

The height of binary heap: $O(\log n)$

Heap – stored in array

tree root item has index 1

n tree elements: $a[1] .. a[n]$

element $a[i]$

children: $a[2i]$ and $a[2i+1]$

parent $a[\text{floor}(i/2)]$

tree root item has index 0

n tree elements: $a[0] .. a[n-1]$

element $a[i]$

children: $a[2i+1]$ and $a[2i+2]$

parent $a[\text{floor}((i-1)/2)]$

Heap: record

n: Integer

els: array [1..MAX] of TComparable

end

Extract maximum (root)

```
Funct. extractMax (H)           //if size(H)>=1
    extractMax :=H.els [1]
    H. els[1]:=H. els[H.n]
    H.n := H.n -1
    downHeap(H,1)
end_extractMax
```


subalg. downHeap(H,poz)

el:=H.Element[poz];

p:=poz; ch:=2*poz

while ch<=H.n do

 if ch<H.n then

 if H. els[ch]<H. els[ch+1] then

 ch:=ch+1

 endif endif

 if H. els[ch]< el then *break*;

 else H. els[p]:=H. els[ch]

 p:=ch; ch:=2*ch

 endif

endwhile

H. els[p]:=el

end_downHeap

•

add

```
subalg. add (H,el)
H.n := H.n + 1
H. els[H.n] := el
upHeap (H, H.n)
end_add
```

```
subalg. upHeap (H, i)
el := H. els[i]
ch:=i
p:=ch div 2
while (p>=1) and (H. els[p]<el) do
    H. els[ch] := H. els[p]
    ch:=p
    p:=p div 2
endwhile
H. els[ch]:=el
end_upHeap
```

build heap - complexity

- A heap could be built by successive insertions.
 $O(n \log_2 n)$
- optimal method:
 - starts by randomly putting the elements
 - then: build the *heap property*

// build the heap property

Subalg. buildHeapProp(H)

for i:=[H.n / 2] , 1 , step = -1 do

 downHeap (H,i)

endfor

endbuildHeapProp

$$nrNodes_h \leq \left\lceil \frac{n}{2^{h+1}} \right\rceil$$

build heap - complexity

- obvious: complexity $\in O(n \cdot \log_2(n))$
- not obvious, but proved: complexity $\in O(n)$

proof ideas

- nr. nodes of height h

$$nrNodes_h \leq \left\lceil \frac{n}{2^{h+1}} \right\rceil$$

- complexity
(nr. of oper.)

$$\sum_{h=0}^{\lfloor \lg n \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil O(h) = O \left(n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h} \right)$$

$$\begin{aligned} \sum_{k=0}^{\infty} kx^k &= \frac{x}{(1-x)^2} \\ &\leq O \left(n \sum_{h=0}^{\infty} \frac{h}{2^h} \right) \\ &= O(n) \end{aligned}$$

HeapSort

- build a heap $\Rightarrow O(n)$
- repeatedly extract maximum $\Rightarrow n * O(\log(n))$

$\Rightarrow O(n * \log(n))$ (even in the worse case)

Heap - usage

- used in the sorting algorithm

heapsort

one of the best sorting methods

with no quadratic worst case scenarios

- used to implement priority queues

Java util: Priority Queue

based on a priority heap

*head of this queue is the **least** element*

C++ STL

.

Standard Template Library: Algorithms

Heap:

push_heap

pop_heap

make_heap ...

(uses RandomAccessIterator)

sort_heap ...

C++ STL

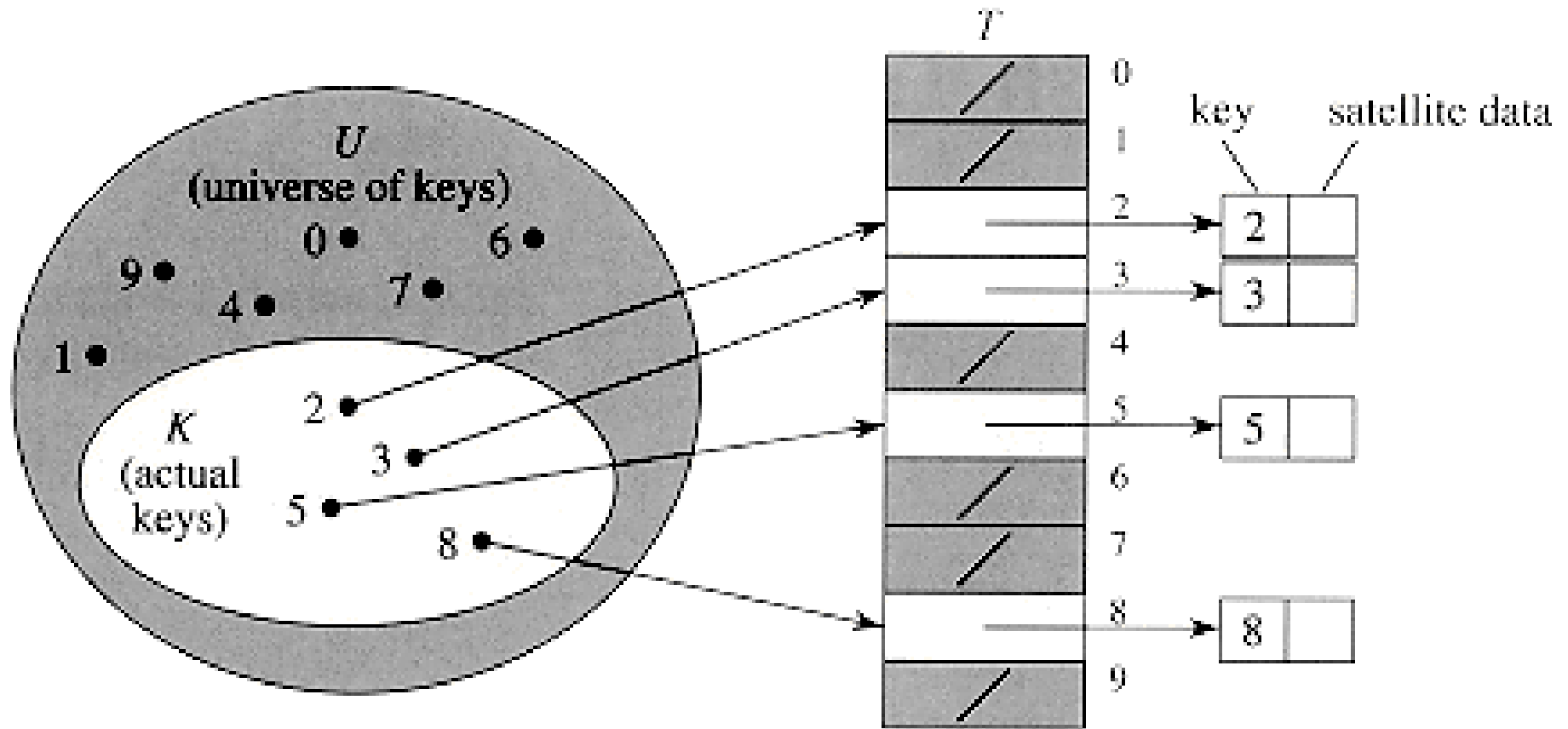
priority queue

Priority queues are implemented as container adaptors

The underlying container

- accessible through random access iterators
- operations:
 - front()
 - push_back()
 - pop_back()
- random access iterators is required to keep a heap structure internally
- container adaptor call make_heap, push_heap and pop_heap

Direct address table



Direct address table

idea:

- allocate an array that has one position for every possible key

applicable: *when we can afford to ...*

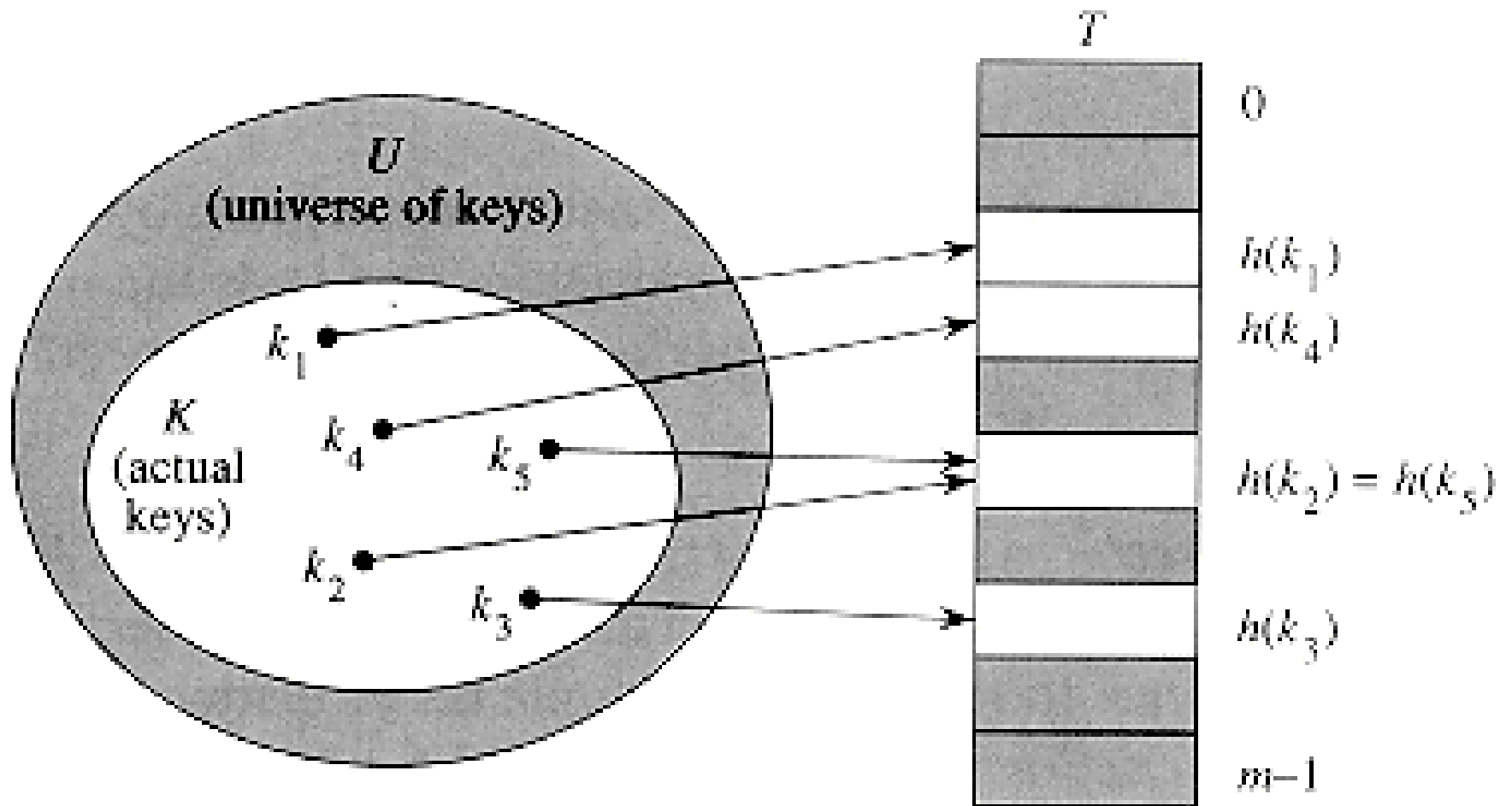
the universe U of keys is reasonably small

- each element has a key
- drawn from the universe $U = \{0, 1, \dots, m - 1\}$, where m is not too large.
- no two elements have the same key.

Possible ways to store elements:

1. satellite data - object external to the direct-address table
with a pointer from a slot in the table to the object
2. the elements can be stored in the direct-address table itself.

Hash table



Collision problem

- ideal solution - avoid collisions

a well-designed hash function

- minimize collisions
- deterministic:
a given input k should always produce the same output $h(k)$

If $|U| > m$

- there must be two keys that have the same hash value
- avoiding collisions altogether is therefore impossible
(sometimes?)

Collision resolution by chaining

