ADT Dictionary

A Dictionary ADT is one that supports the following operations:

Init(), IsEmpty()

Insert(x, K) — x is element with key K.

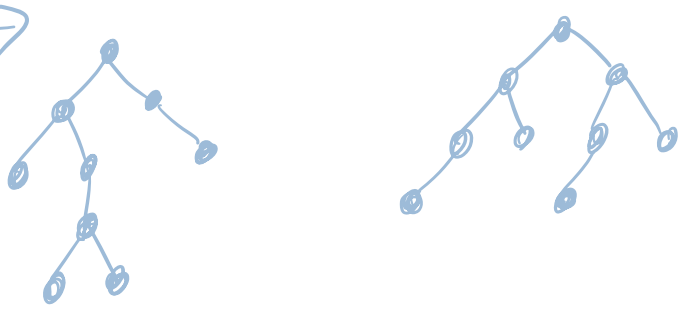
Search(K) — returns the element with key K, perhaps if missing, returns closest element with key < K

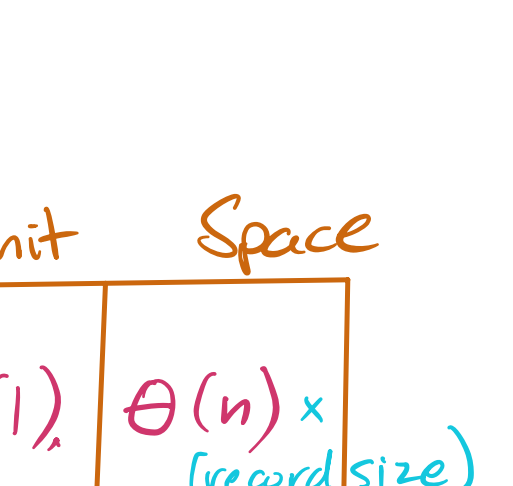
Delete(K) ← key from a totally ordered set

Differs from a PQ, wherein we always want to only extract the min.

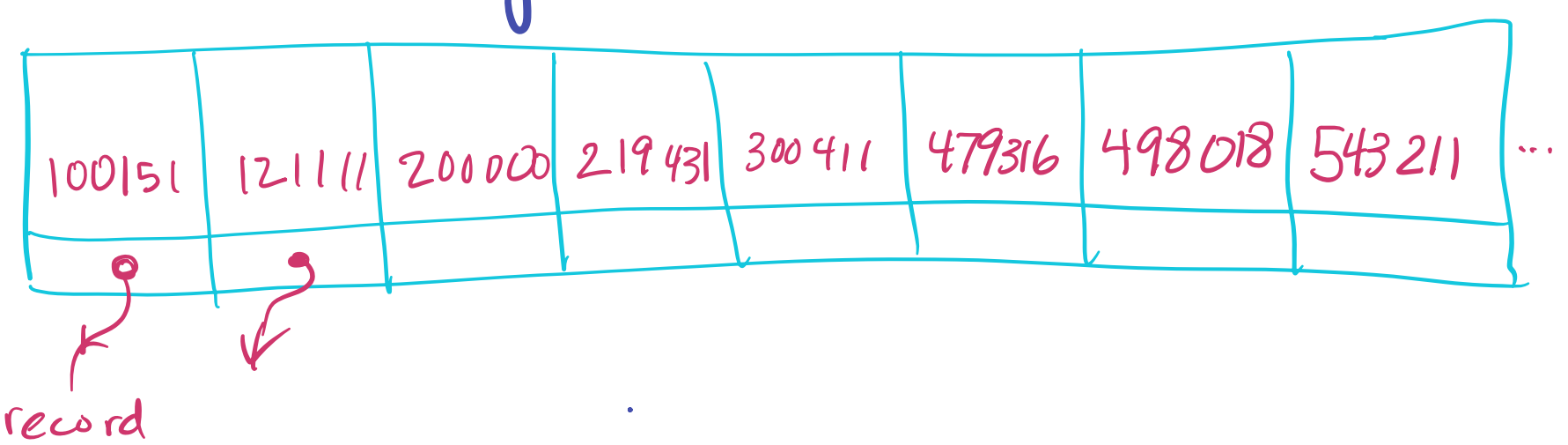
	Insert	Search	Delete	Init	Space
unsorted list	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$	$\Theta(1)$	$\Theta(n)$ x (record size)
sorted list	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$	$\Theta(1)$	$\Theta(n)$ x recordsize *
Sorted Array (packed)	$\Theta(n)$	$\Theta(\lg n)$	$\Theta(n)$	$\Theta(1)$	$\Theta(n)$ x recordsize. U = universe of key
Look-up Table (not packed	$\Theta(\ )$	$\Theta(\ )$	$\Theta(\ )$?	$\Theta(\ )$?	$\Theta(\ )$

* Binary Search Trees

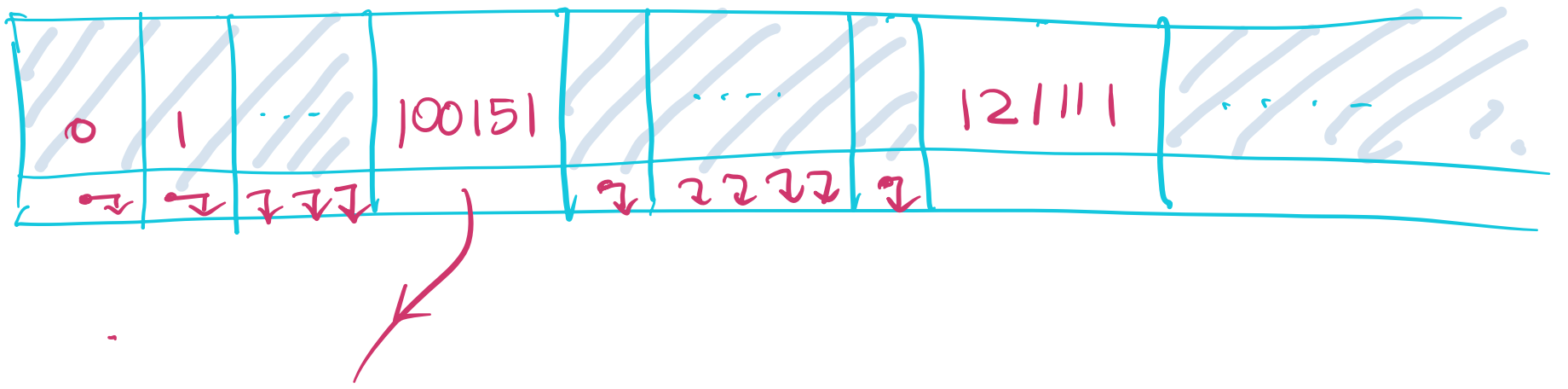




Packed Array



Look-up Table = "Direct Address Table"



The Look-up Table is indexed by the Keys.

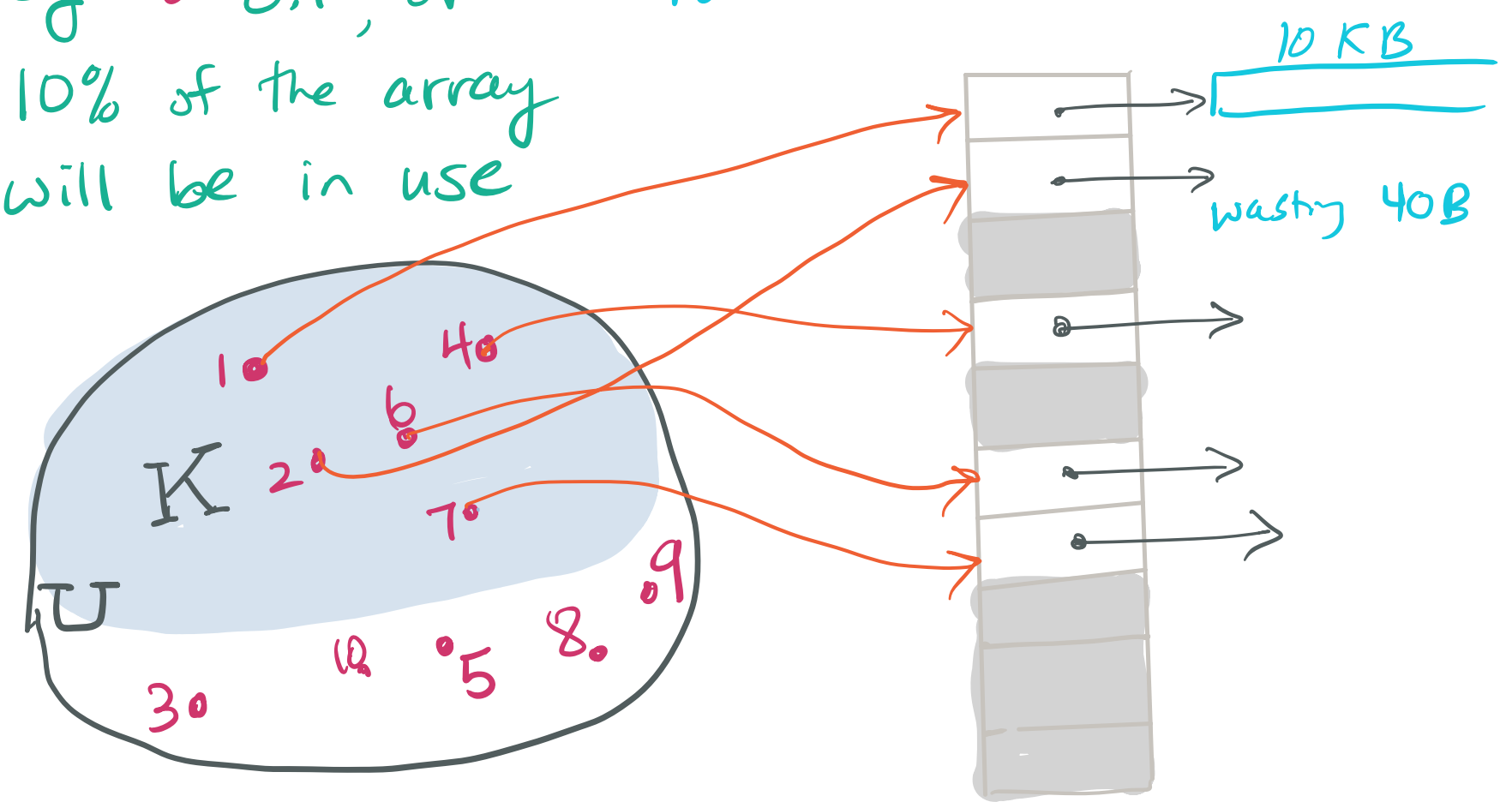
It is a fantastic implementation for running time (time complexity)

When "keys in use" as a proportion of "all possible Keys" is small, which is usual, the space complexity is too much. γ is that ratio. $\gamma \cdot |U|$ is Keys-in-use

$$\gamma = \frac{|K|}{|U|}$$

$|U|$ = Universe of Key values ≈ space needed for Look-up Table.

eg $\gamma = 0.1$, or 10% of the array will be in use



Direct-Address implementation

DirectAddressSearch(k)

return T[k]



DirectAddressInsert(x, k)

T[k] = a pointer to x

DirectAddressDelete(k)

T[k] = NULL

The main problem with DirectAddress is that many applications do NOT have high γ (high key density)

Can we achieve DirectAddress-like

behaviour when γ is low?

Hash Tables

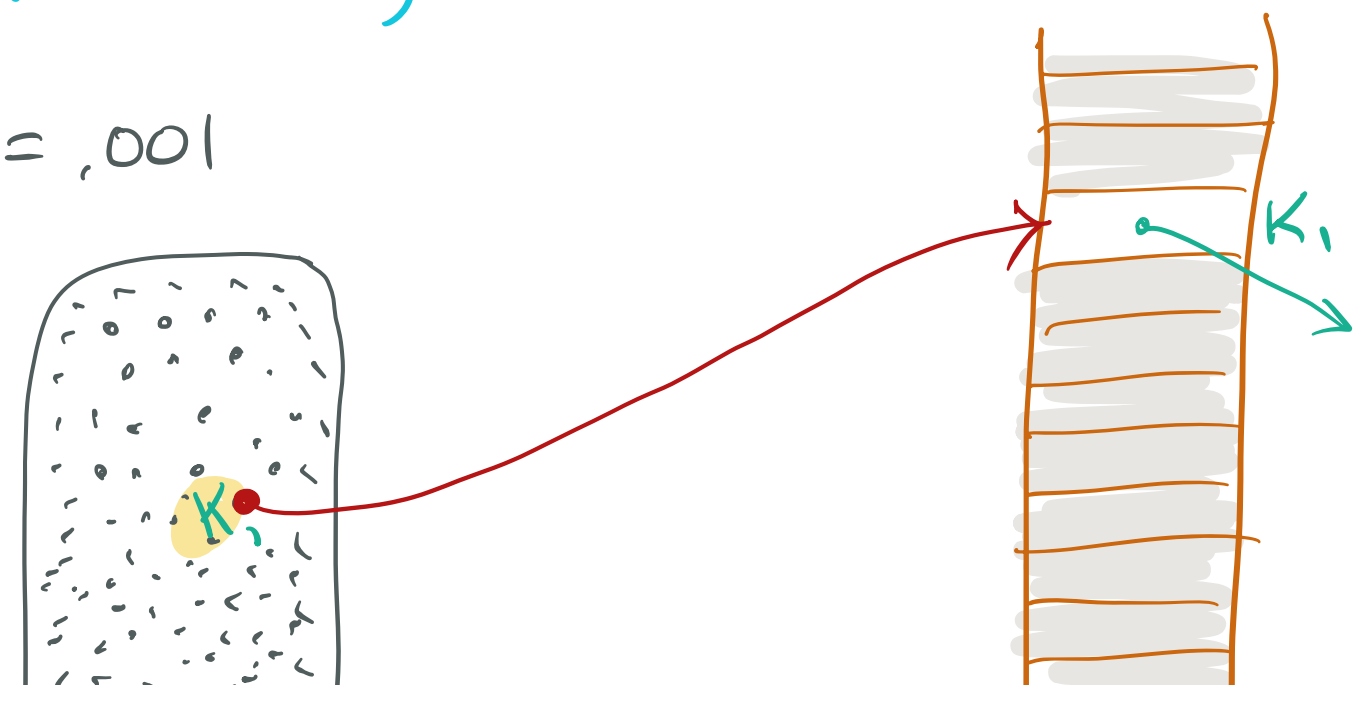
Suppose you have a function h

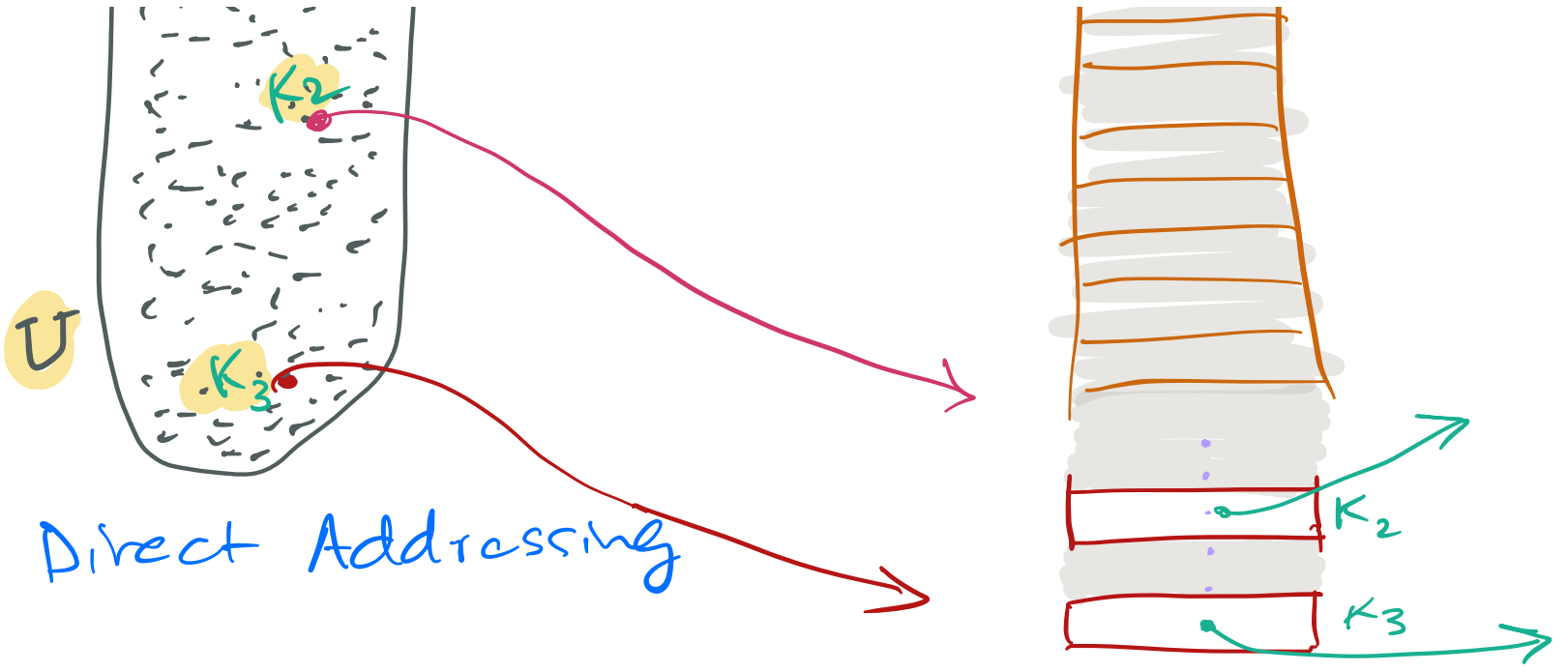
$$h : U \to \{0, \ldots, m-1\}$$

i.e. f maps the universe of keys to a much smaller set of values...

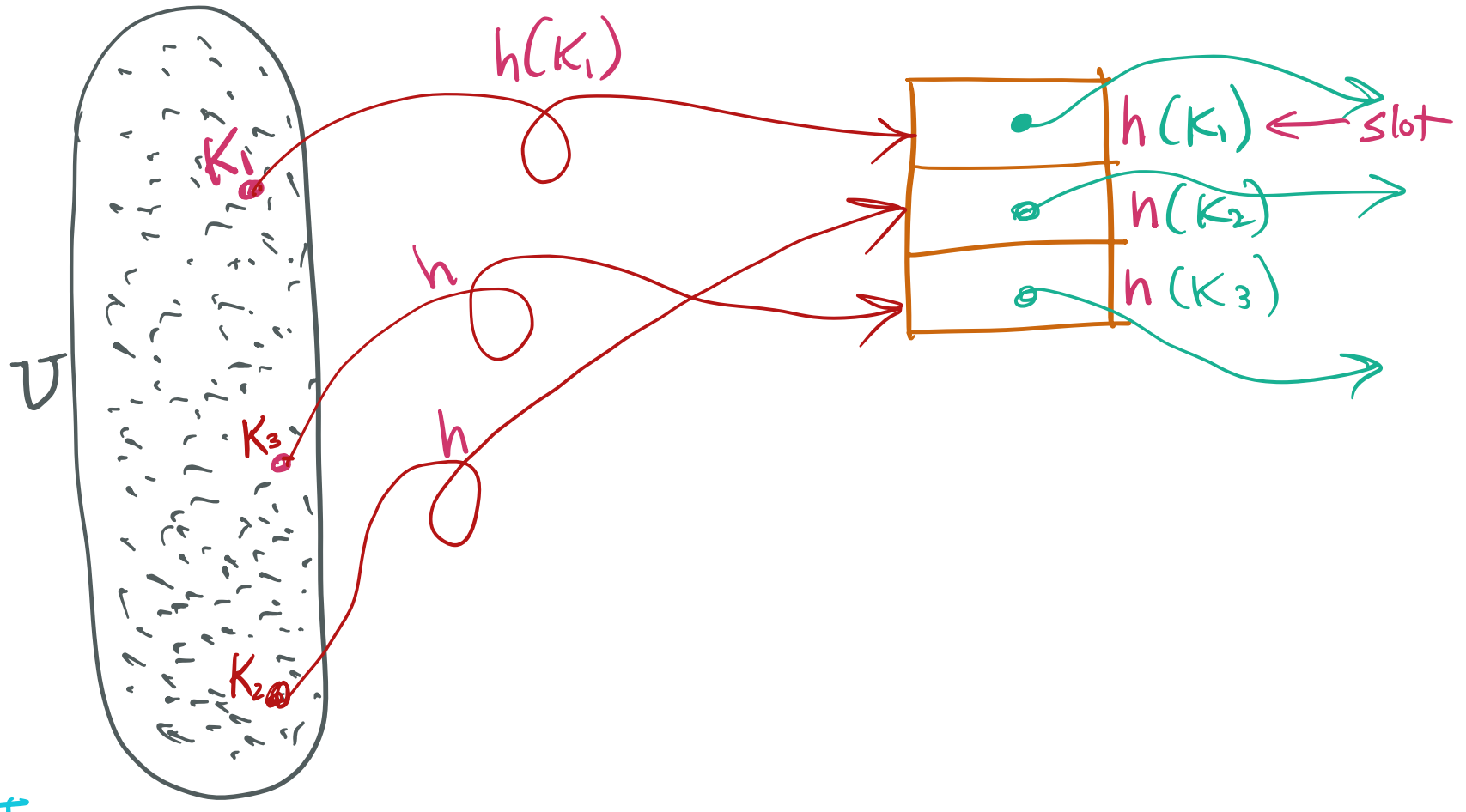
ideally, values that index into an appropriately-sized array

$\gamma = .001$





Direct Addressing



BUT can we find a function h that maps:

- exactly to $\{0, \ldots, m-1\}$ with $m = |K|$?
  - don't need this
- so as to have no two distinct keys used, K_1 and K_2, where

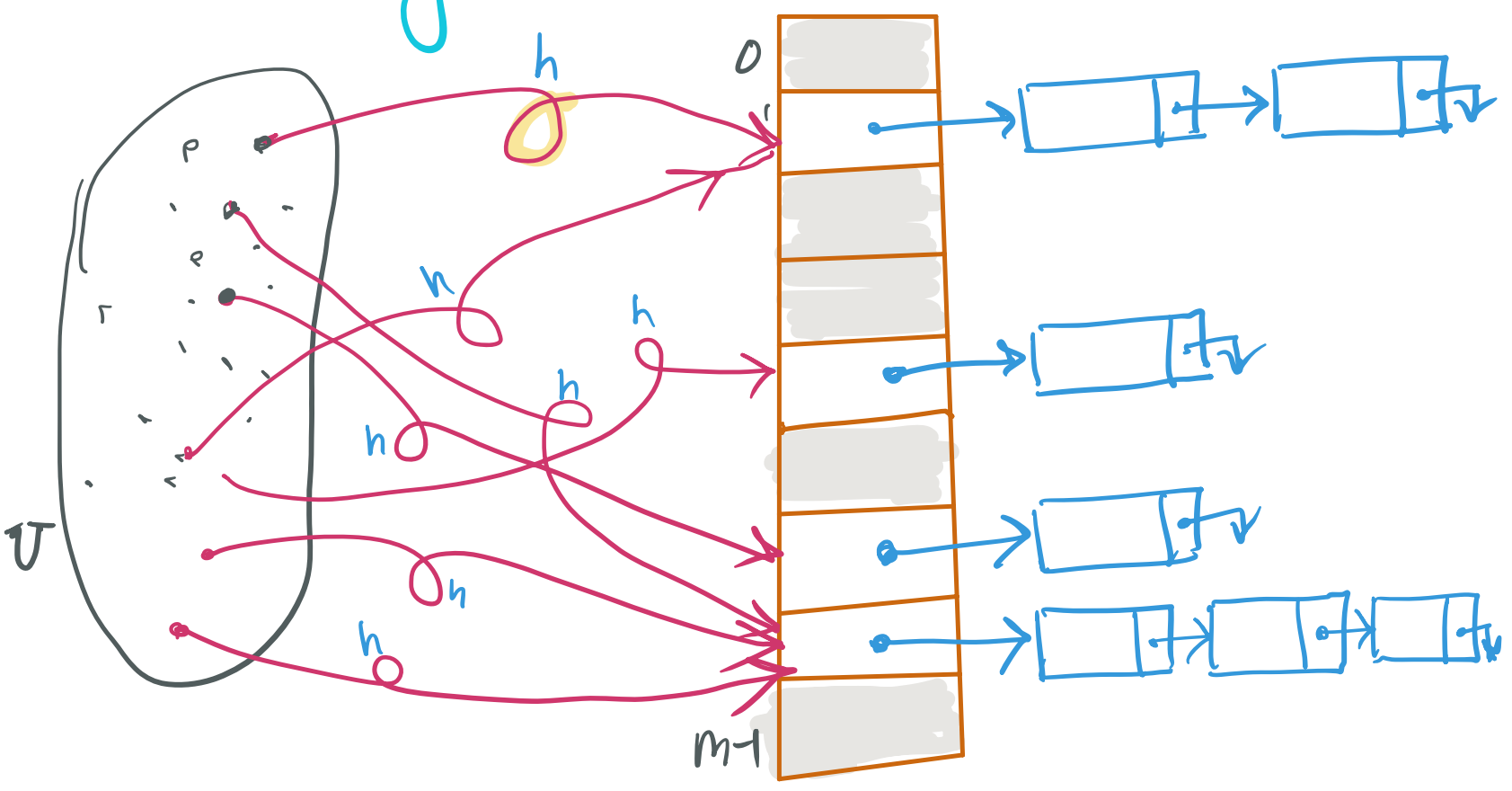
$h(K_1) = h(K_2)$?

↑ this is called a collision

Generally, we cannot avoid collisions, because we don't always know in advance what subset of U will be in use.

Two strategies for dealing with collisions.

I. Chaining



Analysis of hashing with chaining

In this context, we are actually interested in expected behaviour more than worst-case behaviour.

Why?

- – Worst case behaviour is pretty bad
  – acts just like unsorted linked list
- – The behaviour is not just a function of the inputs, but also of the hash function we chose ....
  worst case is decoupled from dependence on inputs alone.
  There will always be a hash function that has good worst case behaviour on the same inputs.

Given hash table T with m slots and n elements

$\alpha = \frac{n}{m}$ is called the load factor

= average number of elements stored per chain.

Suppose we pick a hash function so that a randomly selected element of U is equally likely to hash to each of the m slots = "uniform distribution"

(this assumption is simple uniform hashing)

Also assume computing $h(k)$ is $\in O(1)$.

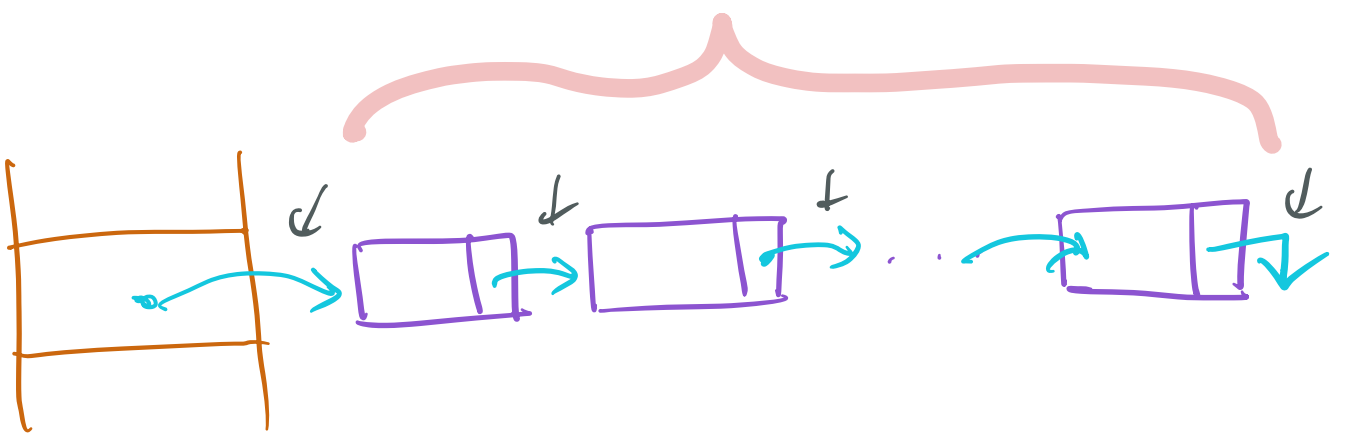
Theorem 12.1 (CLRS)

In a hash table with chaining, under simple uniform hashing, an unsuccessful search takes time

$\Theta(1+\alpha)$ on average. to search the list for hashing the Key.

α elements on average

Proof



$1 + \alpha$

∎

Theorem 12.2 (CLRS)

In a hashtable with chaining, under assumption of simple uniform hashing, a successful search takes time

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

Proof.

Suppose we change Insert so that it traverses the list to the end and places the new item there.

This Insert has same running time as successful search.

Consider all the inserts, and all the element comparisons done during these (inefficient) inserts:

$$\sum_{i=1}^{n}\left(1+\frac{i-1}{m}\right) = n+\frac{1}{m}\sum_{i=1}^{n}(i-1)$$

$\frac{i-1}{m}$ = #elements in table

When i was inserted

Divide by n to get average per insertion:

$$1 + \frac{1}{nm}\sum_{i=0}^{nm} i = 1 + \frac{1}{nm}\left(\frac{(n-1)\,n}{2}\right)$$

$$= 1 + \frac{n^2 - n}{2nm}$$

$$= 1 + \alpha - \frac{1}{2m}$$

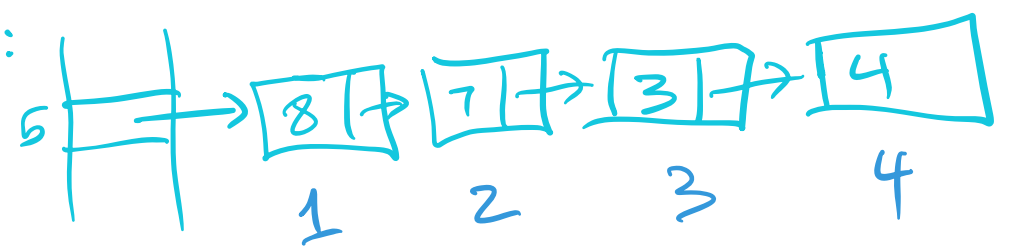
Number of comparisons in successful search $= 1 +$ comparisons for insert

So is $\Theta(2 + \alpha - \frac{1}{2m}) = \Theta(1+\alpha)$ ▨

all hash to slot 5

ins(4)... ins(3) ... ins(7) ... ins(8)

good:



Bad ins(8) .. ins(7) ins(3) ins(4),

