

Hash Tables

Data Structures for *Map ADT*

	Find	Insert	Delete
Unsorted Array	$\theta(n)$	$\theta(n)$	$\theta(n)$
Unsorted Linked List	$\theta(n)$	$\theta(n)$	$\theta(n)$
Sorted Array	$\theta(\log(n))$	$\theta(n)$	$\theta(n)$
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Binary Search Tree	$\theta(h), \theta(n)$	$\theta(h), \theta(n)$	$\theta(h), \theta(n)$
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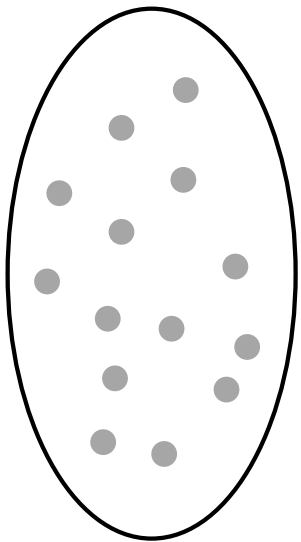
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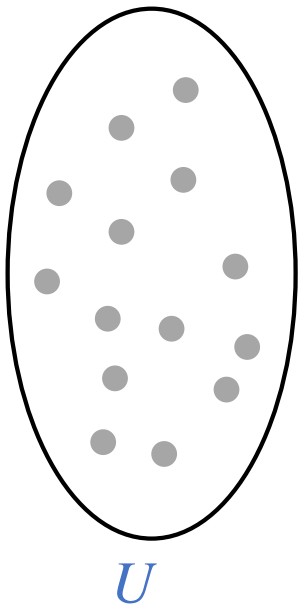
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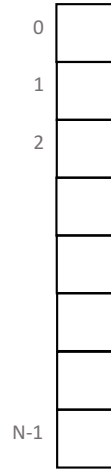
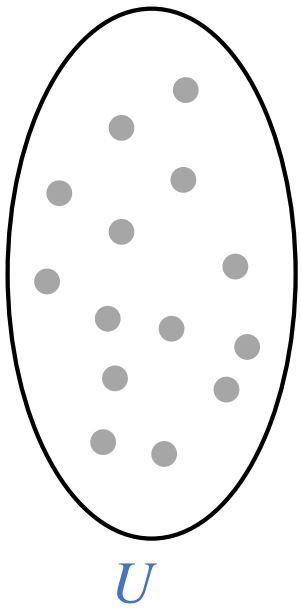
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Hash Tables



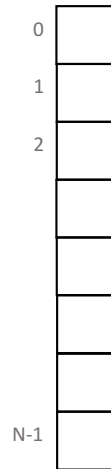
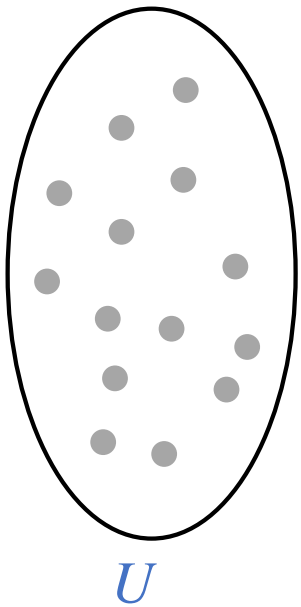
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Universe from which the keys will be taken

Hash Tables



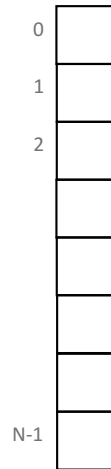
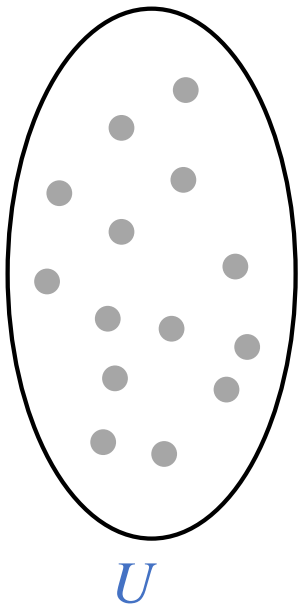
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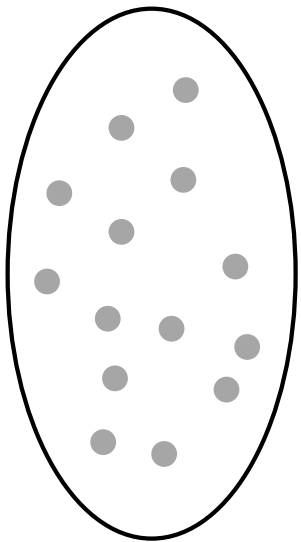
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Universe from which the keys will be taken
- $T = [None]*N$

Hash Tables

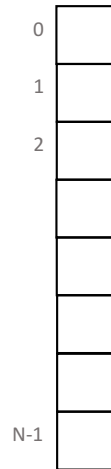


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Hash table of N slots, where the entries will be stored

Hash Tables



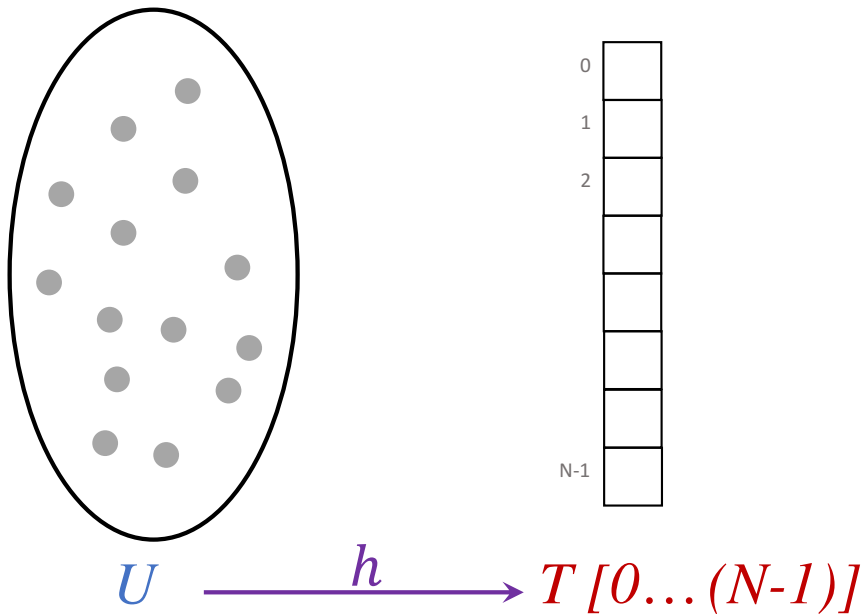
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$T[0 \dots (N-1)]$

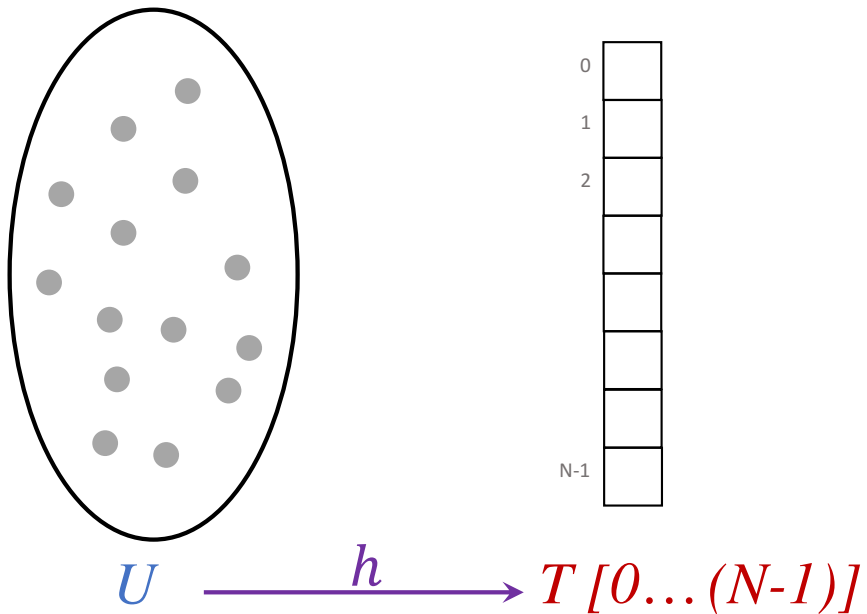
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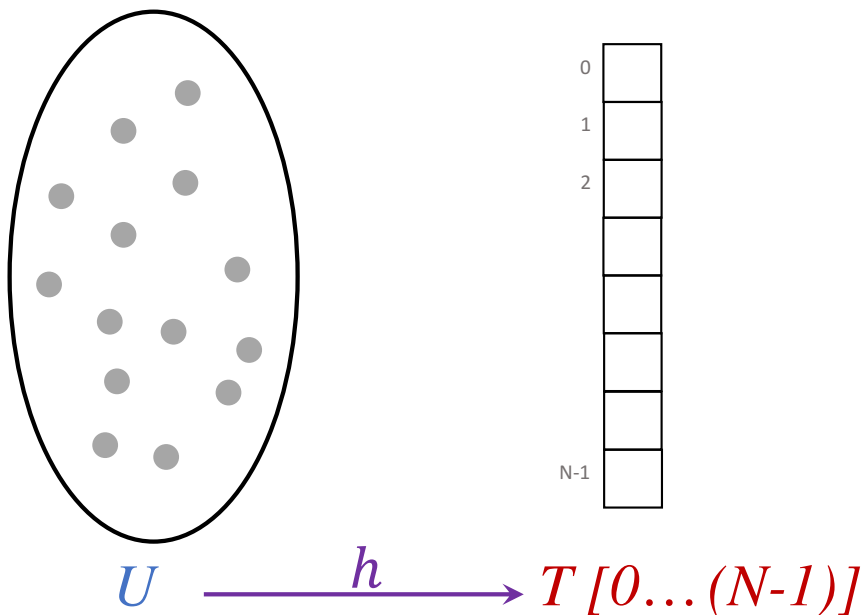
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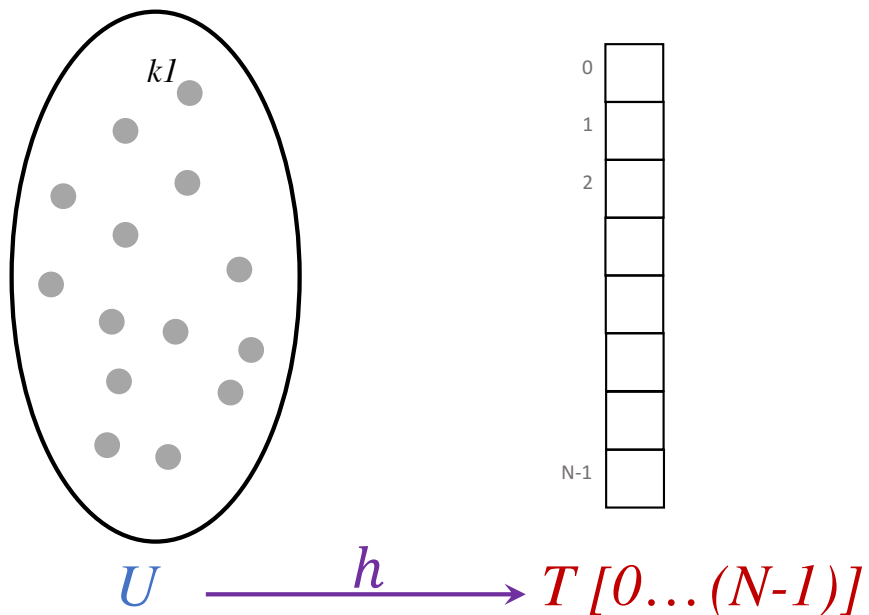
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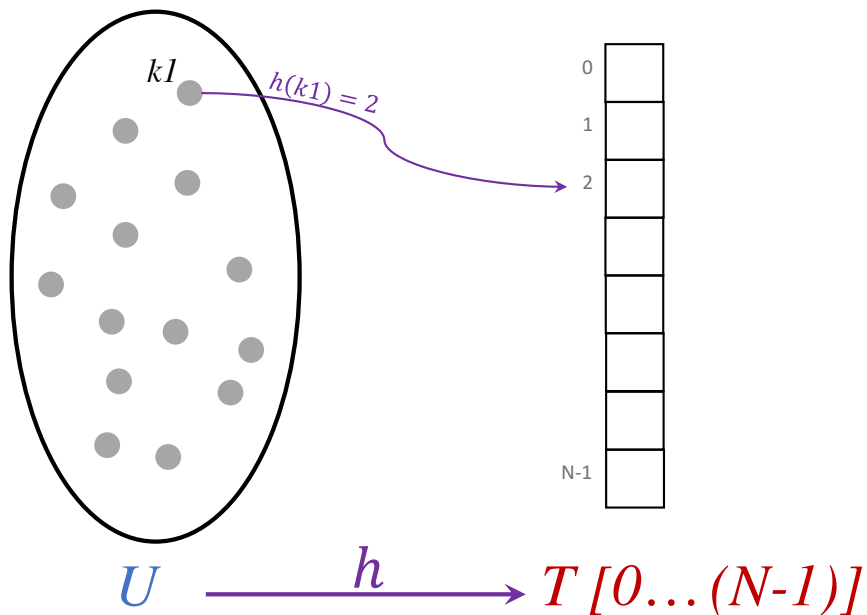
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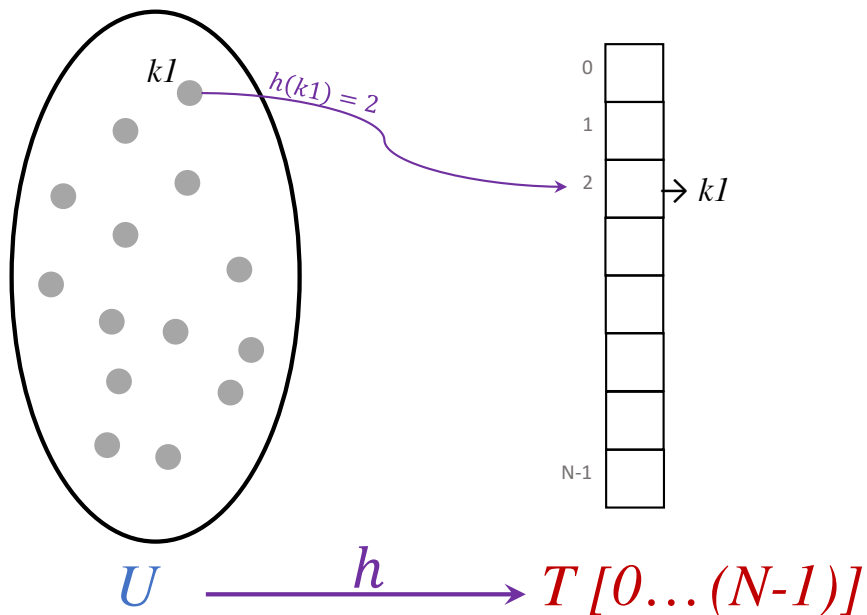
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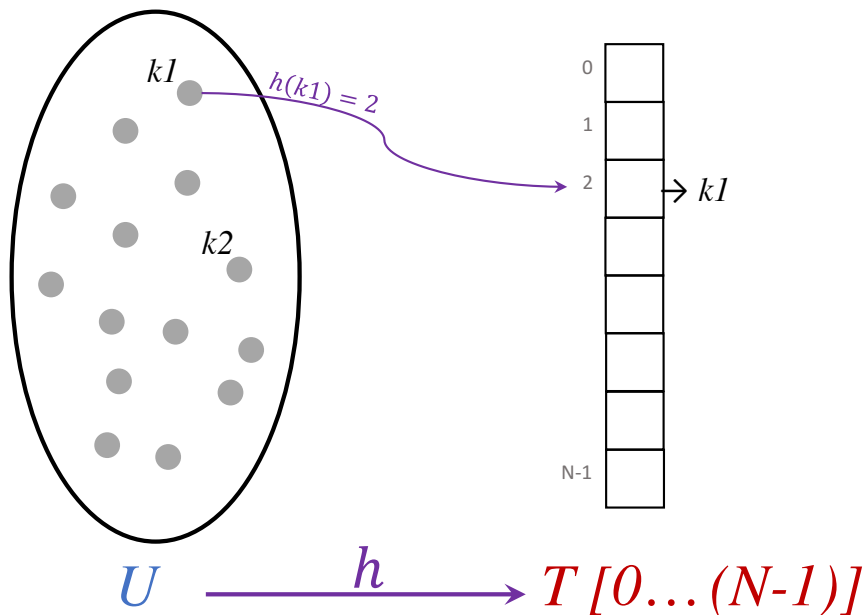
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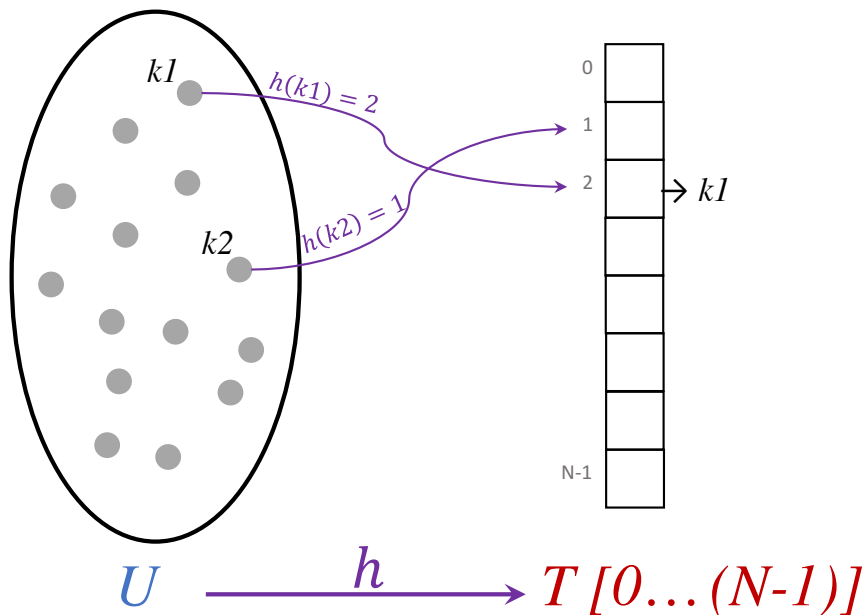
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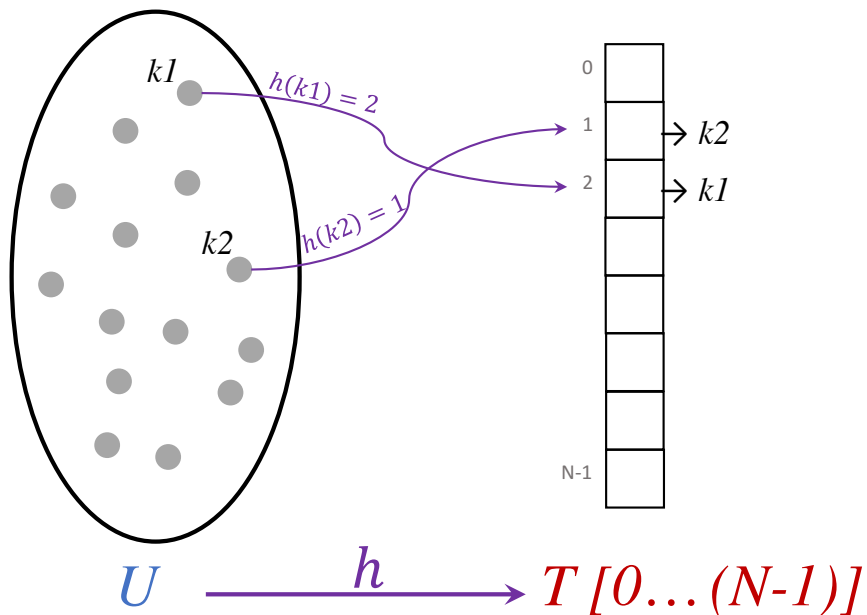
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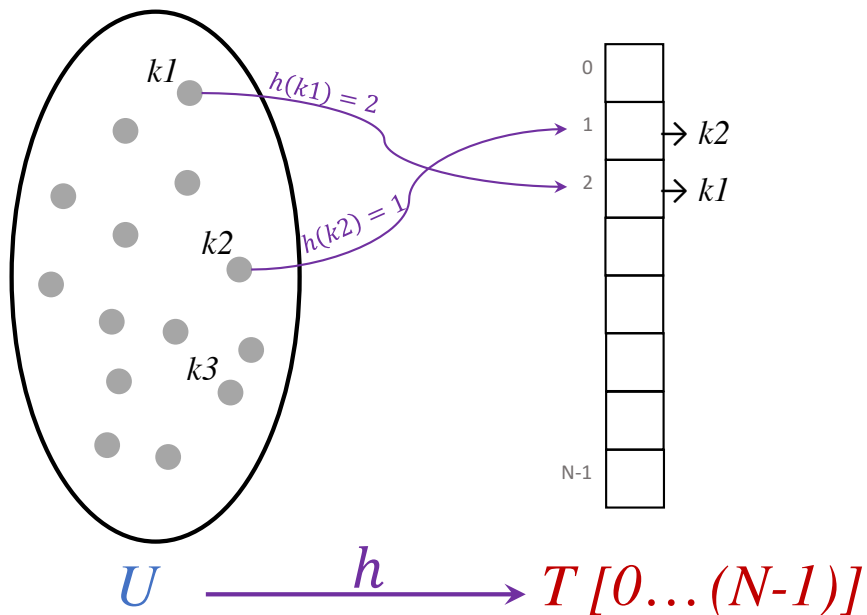
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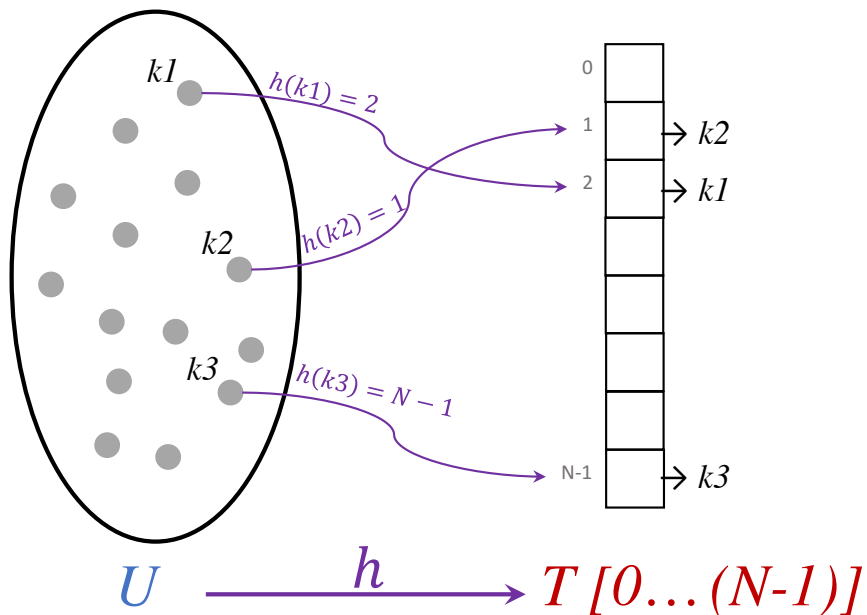
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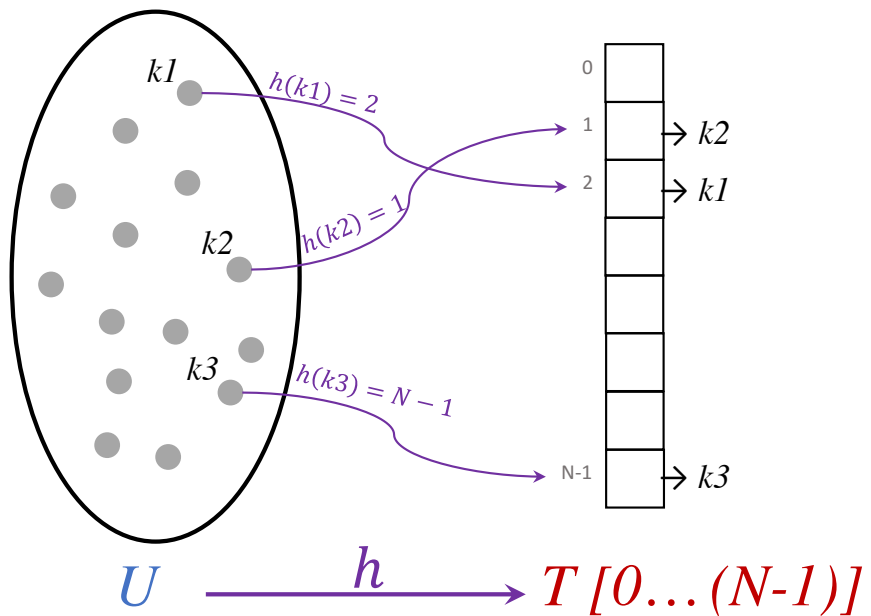
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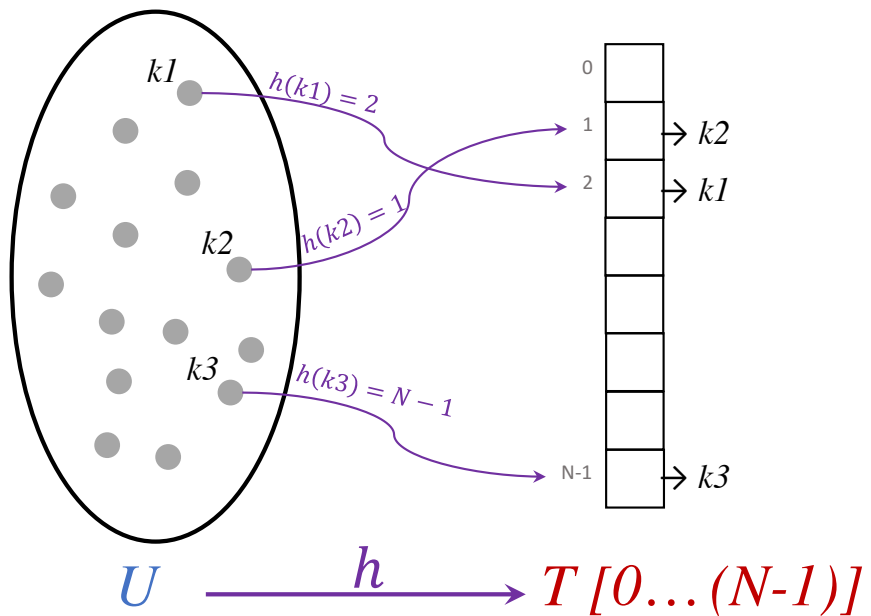


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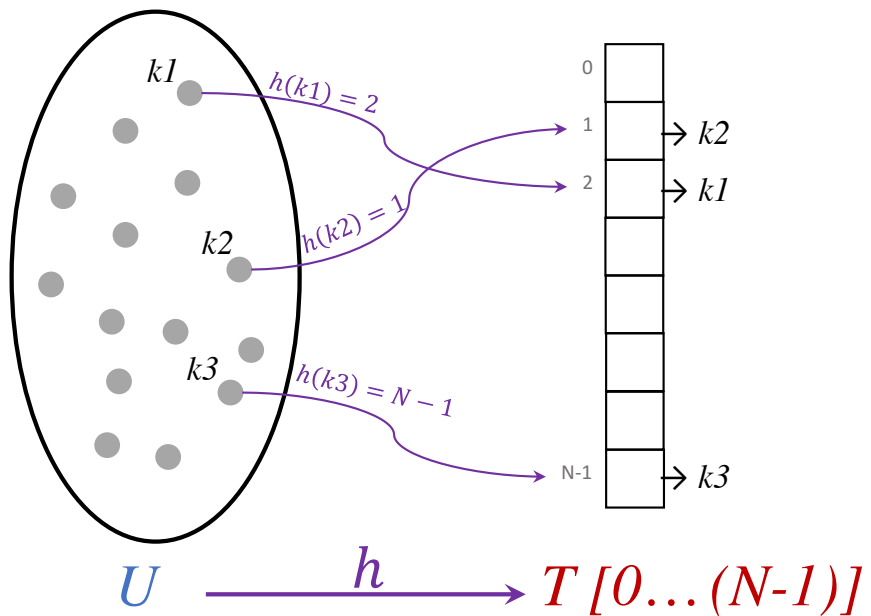


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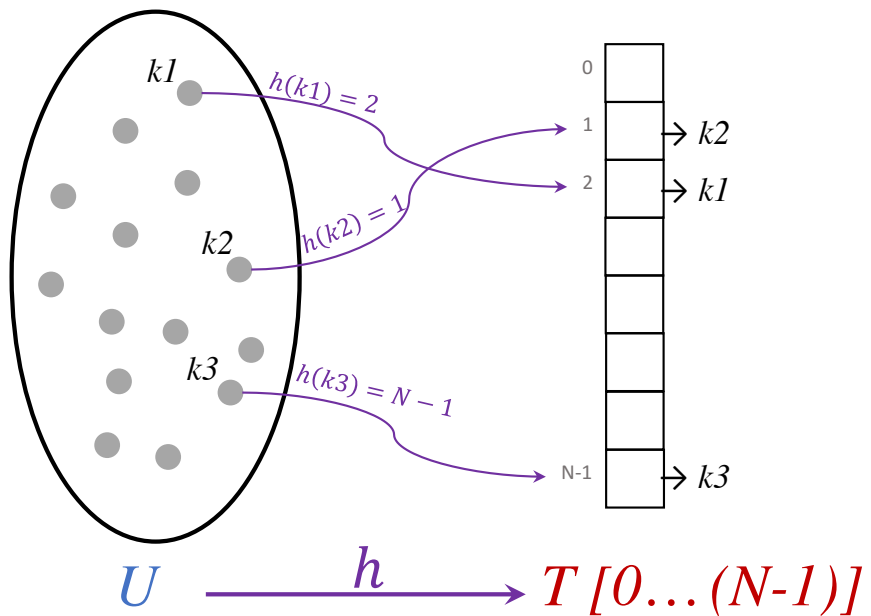
insert(key, value):

Hash Tables



insert(key, value):
 $i = h(\text{key})$

Hash Tables

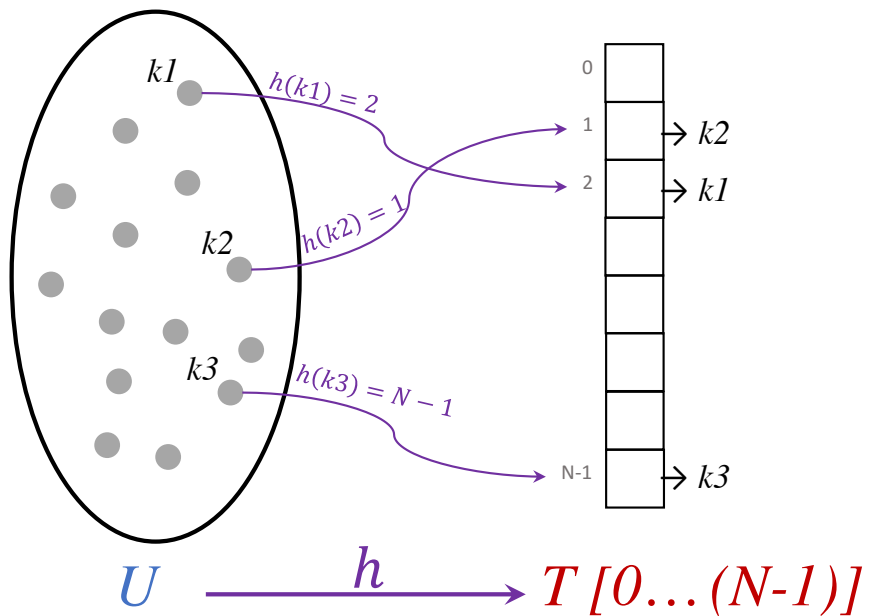


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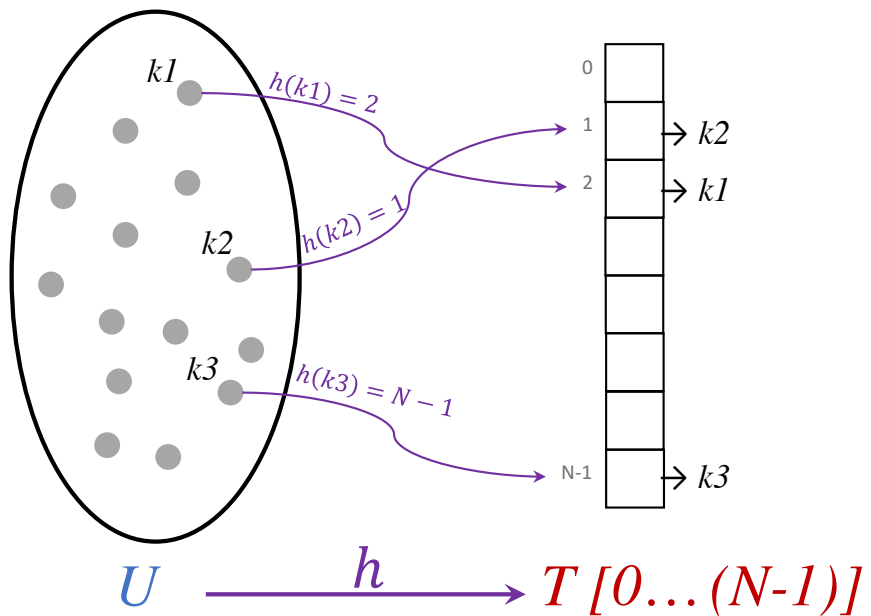
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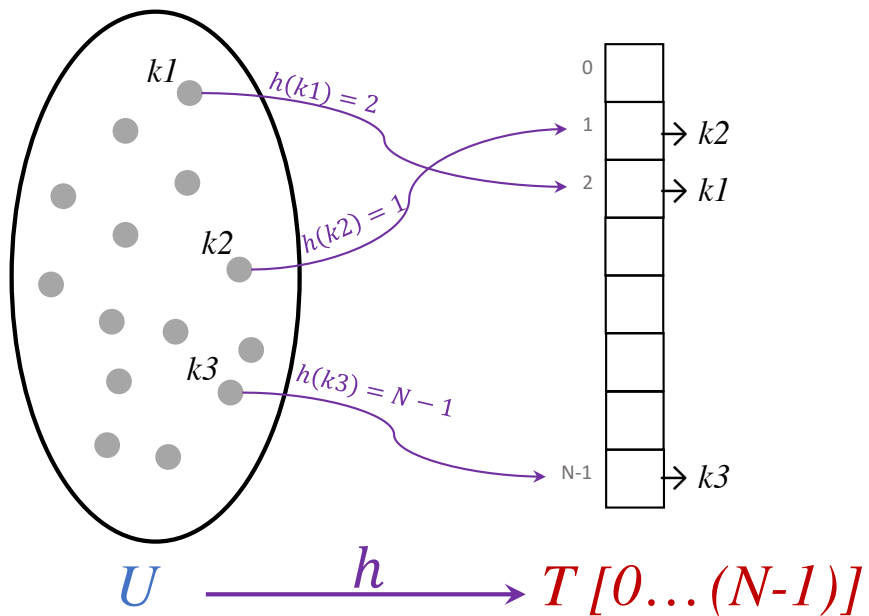
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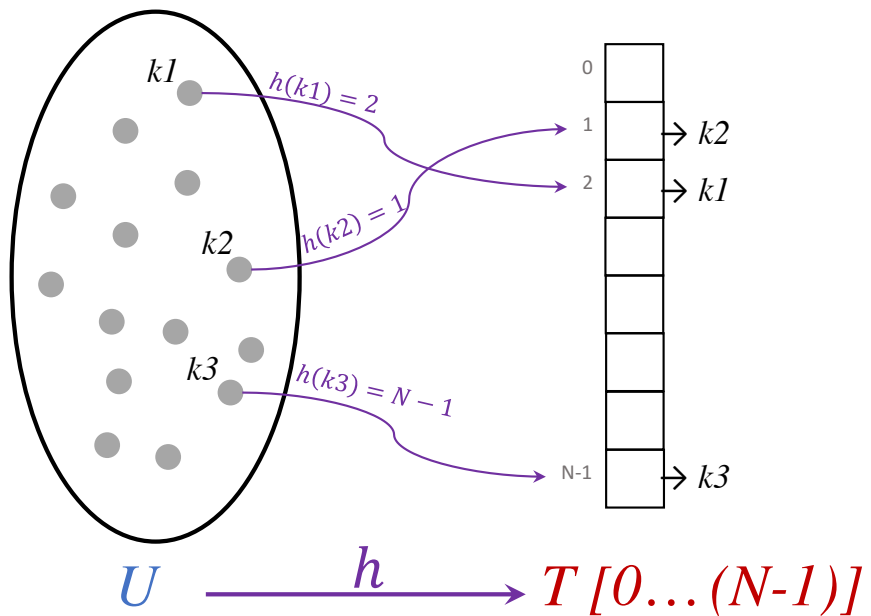
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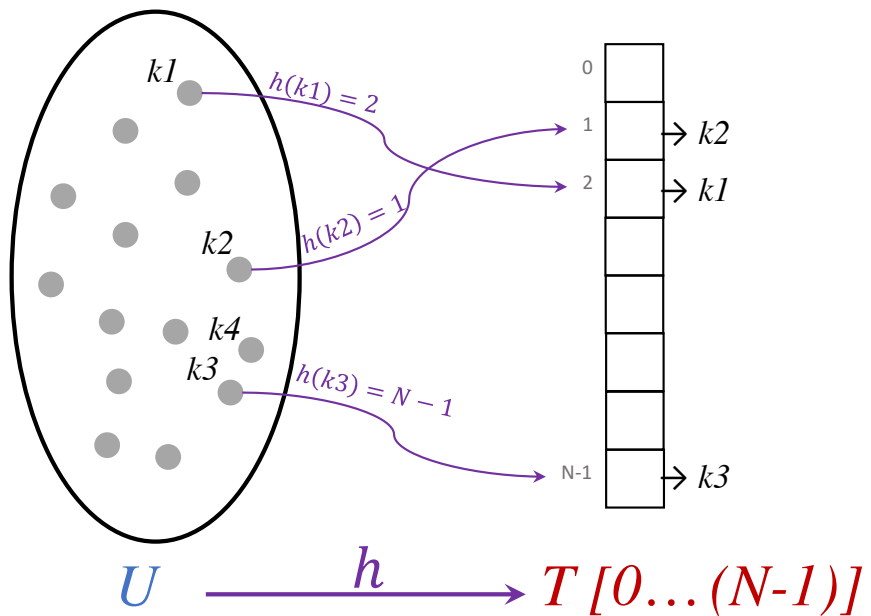
$i = h(\text{key})$

return $T[i]$

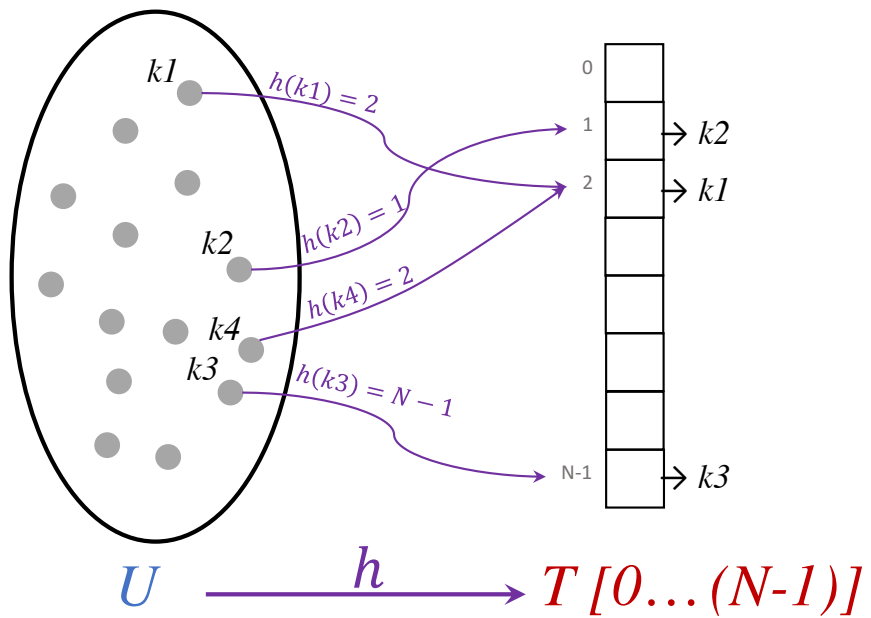
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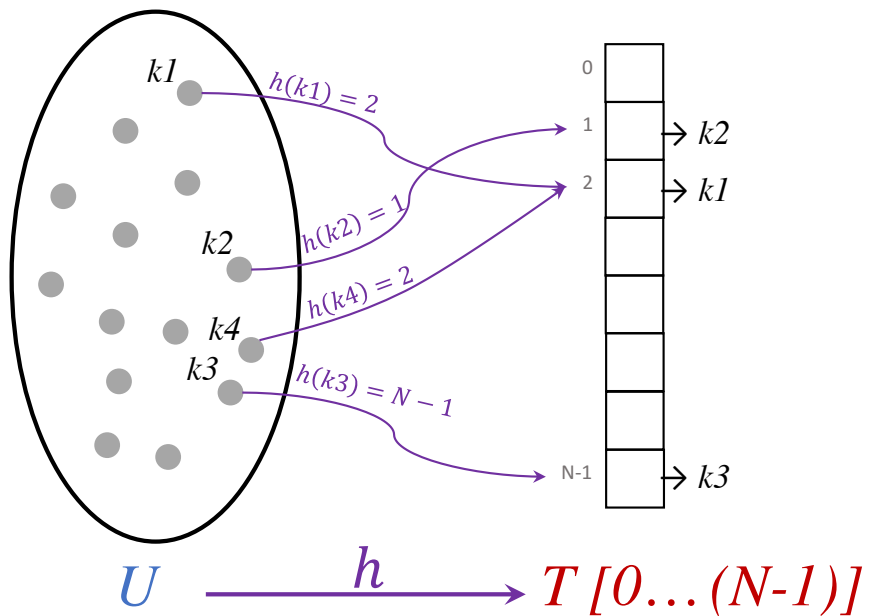
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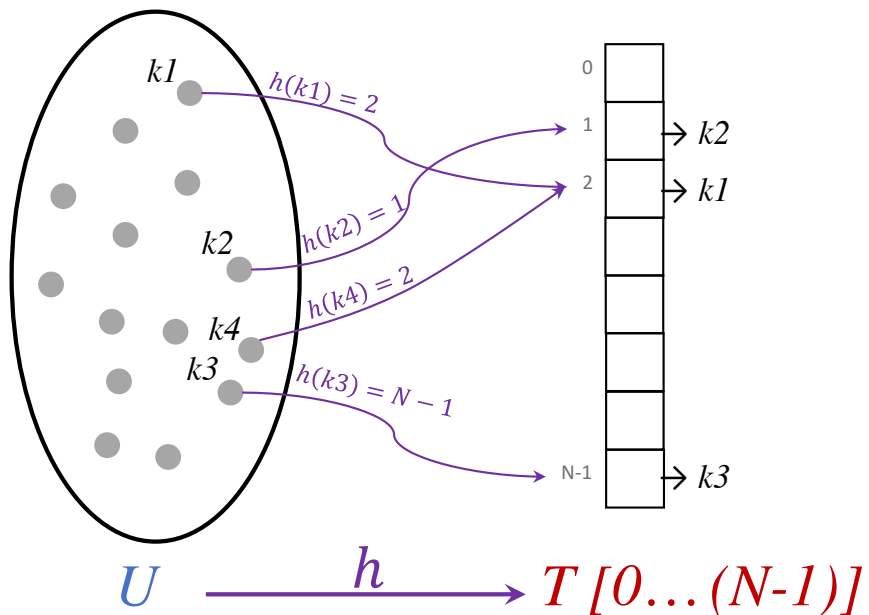
Hash Tables



The Collisions Problem:

multiple keys could be mapped to the same slot

Hash Tables



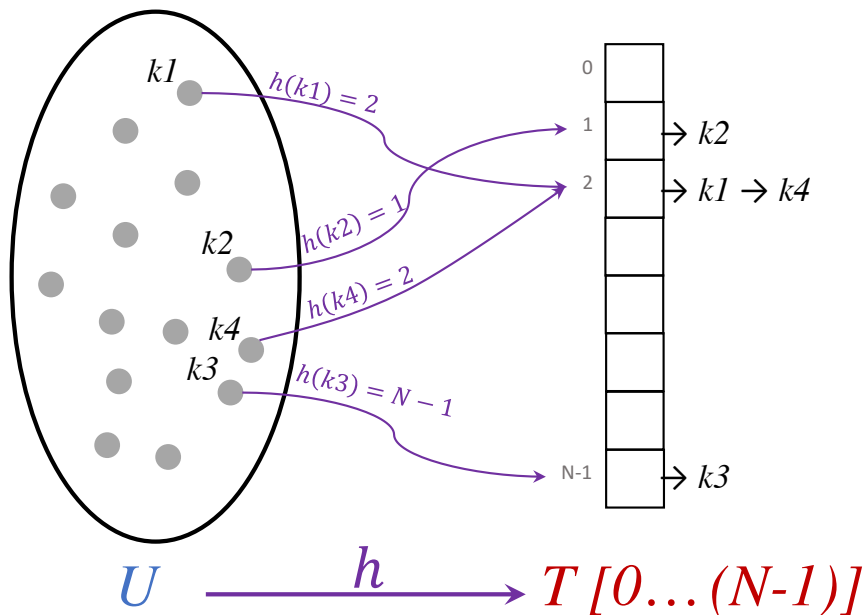
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An easy solution:

Chaining – store all entries that are mapped to the same slot, in some secondary collection

Hash Tables



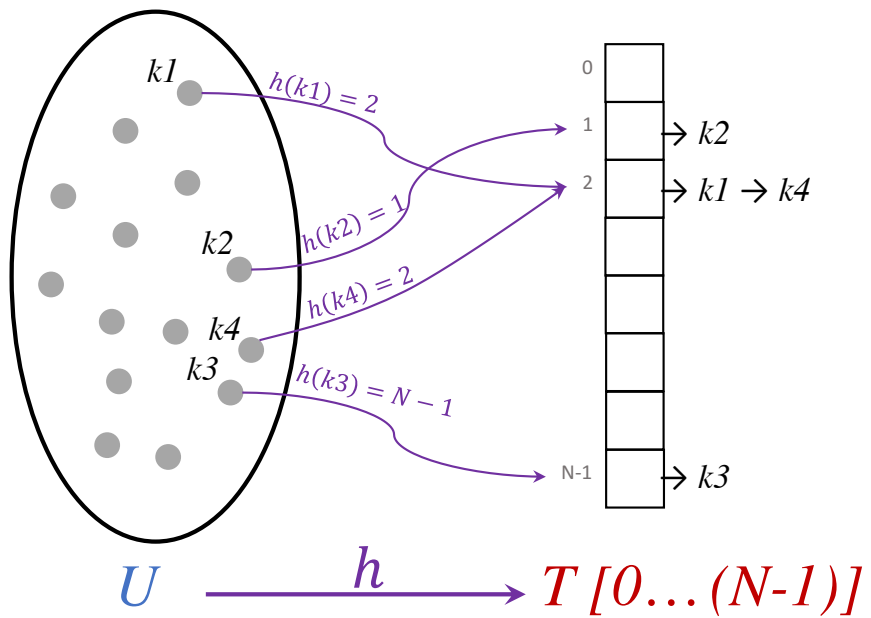
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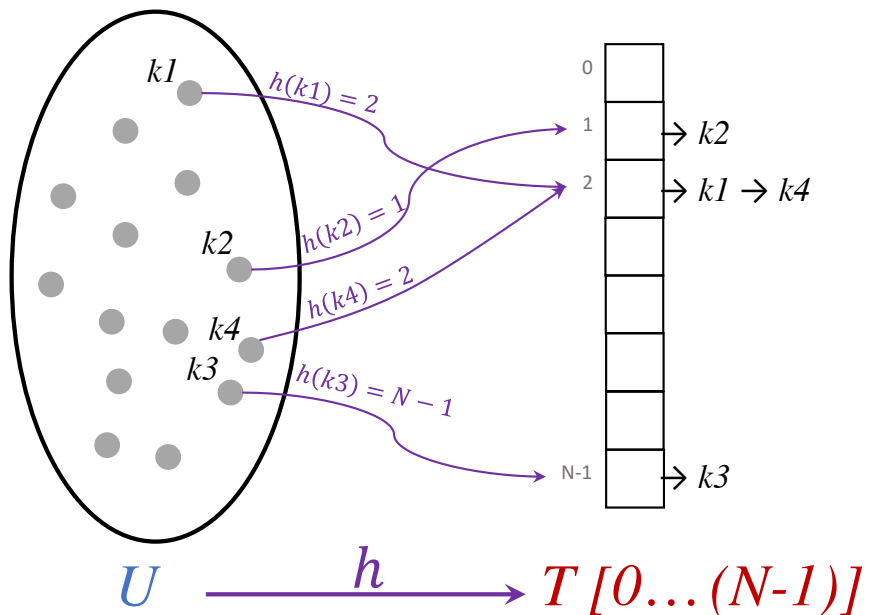
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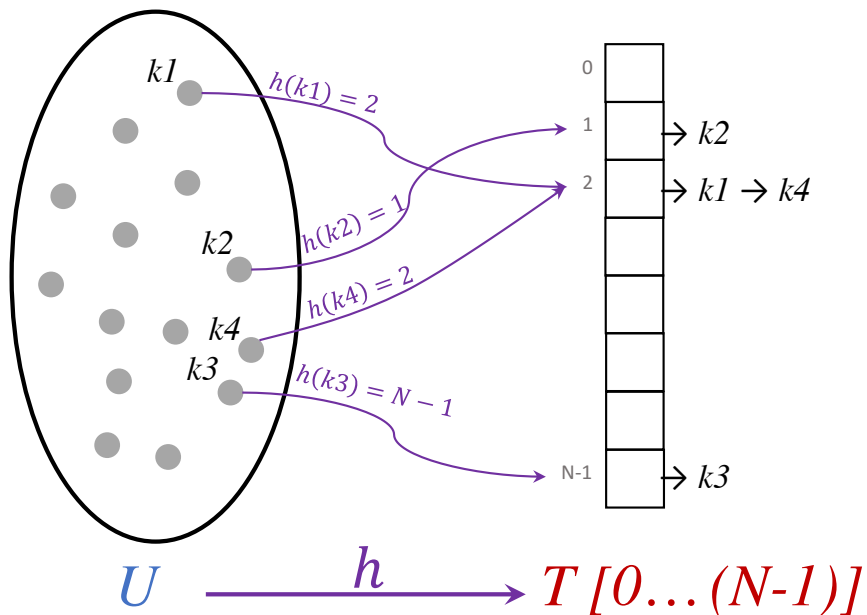


Hash Tables



Problem: When using chaining, a lot of keys could end up being stored at the same slot \Rightarrow bad performance

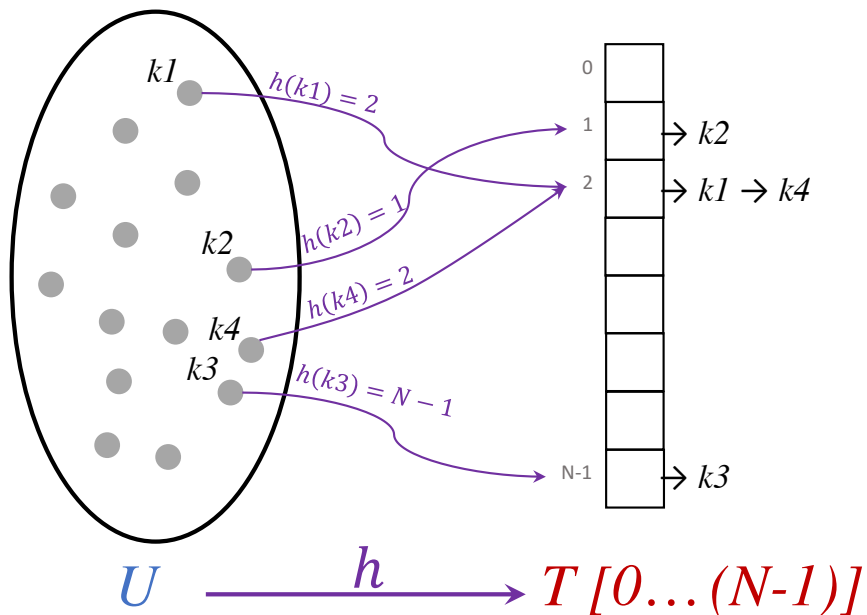
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Solution: Use a “good” hash function (a function that evens out the collisions)

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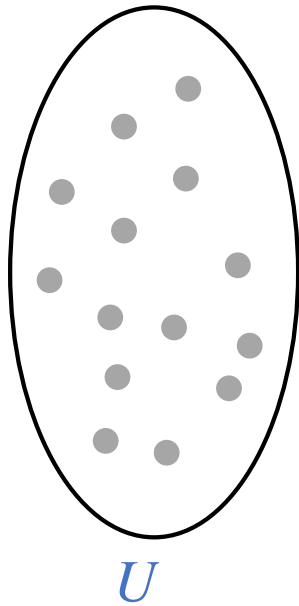
Solution: Use a “good” hash function (a function that evens out the collisions)

Uniform Hashing Function:

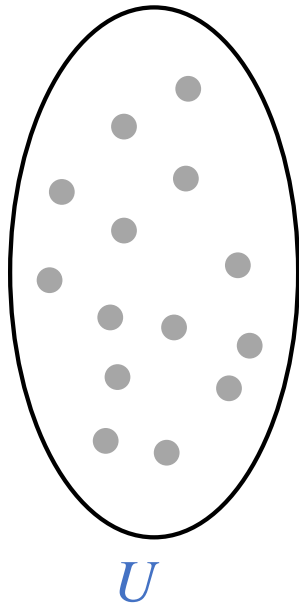
A function that when given a randomly chosen key, it will be equally likely mapped to any of the N slots of T , independently of where any other key has hashed to

Hash Functions

Hash Functions

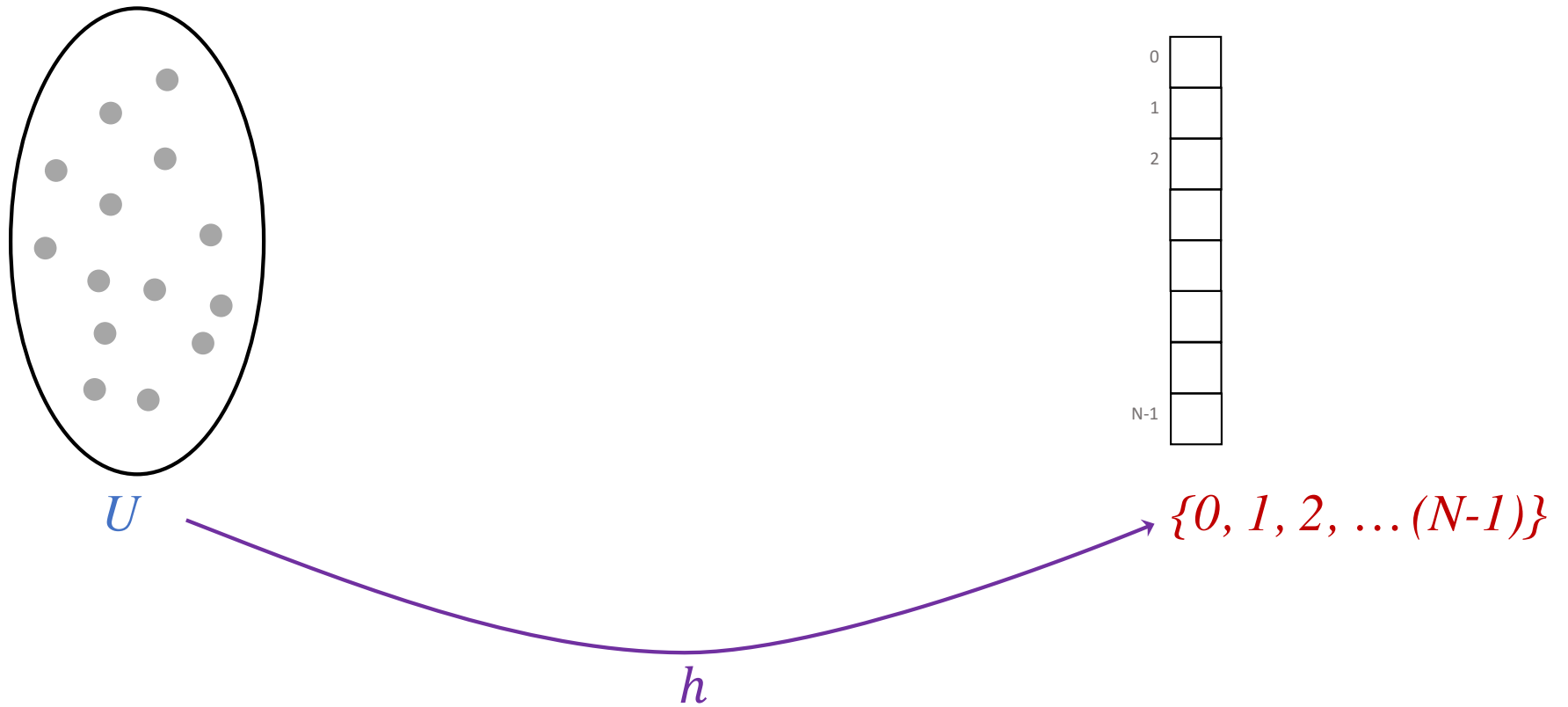


Hash Functions

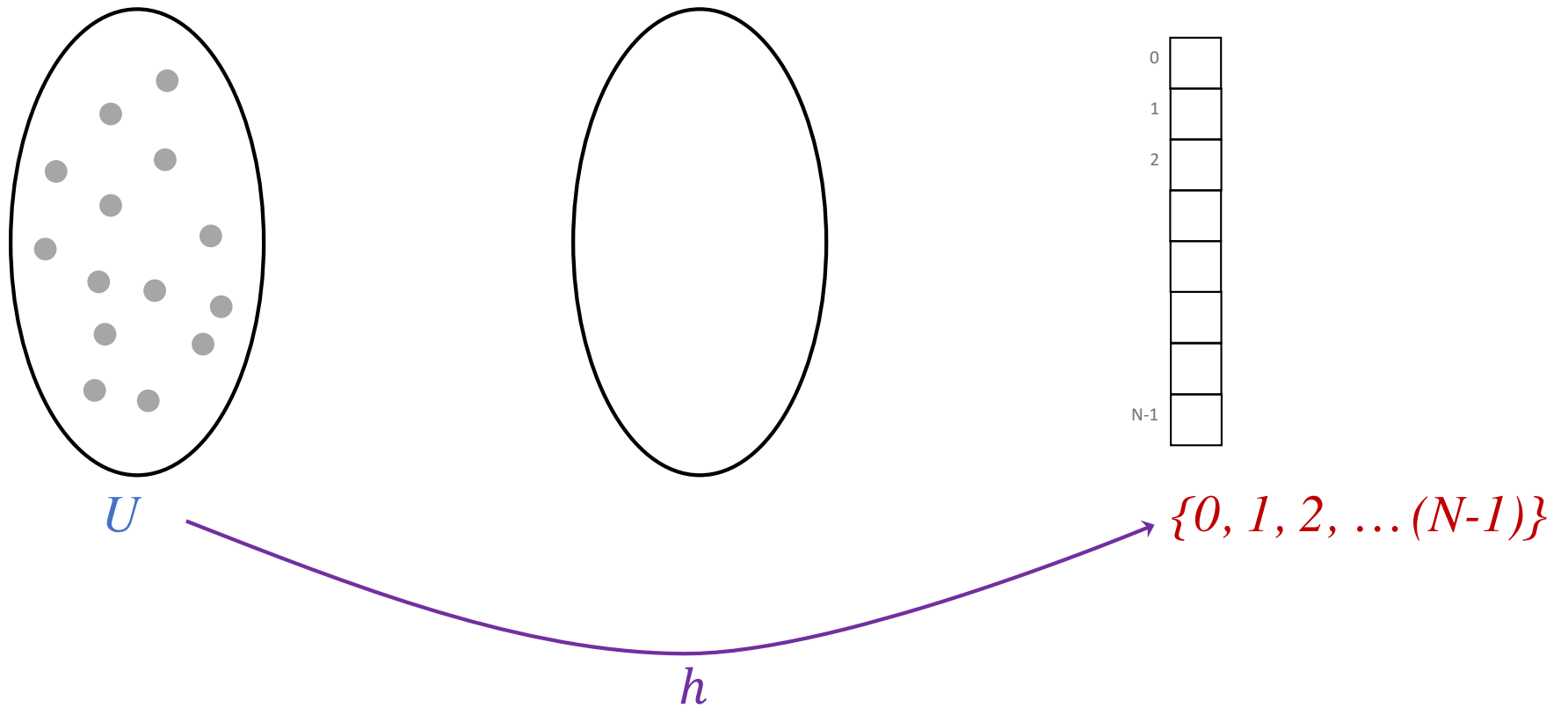


$\{0, 1, 2, \dots (N-1)\}$

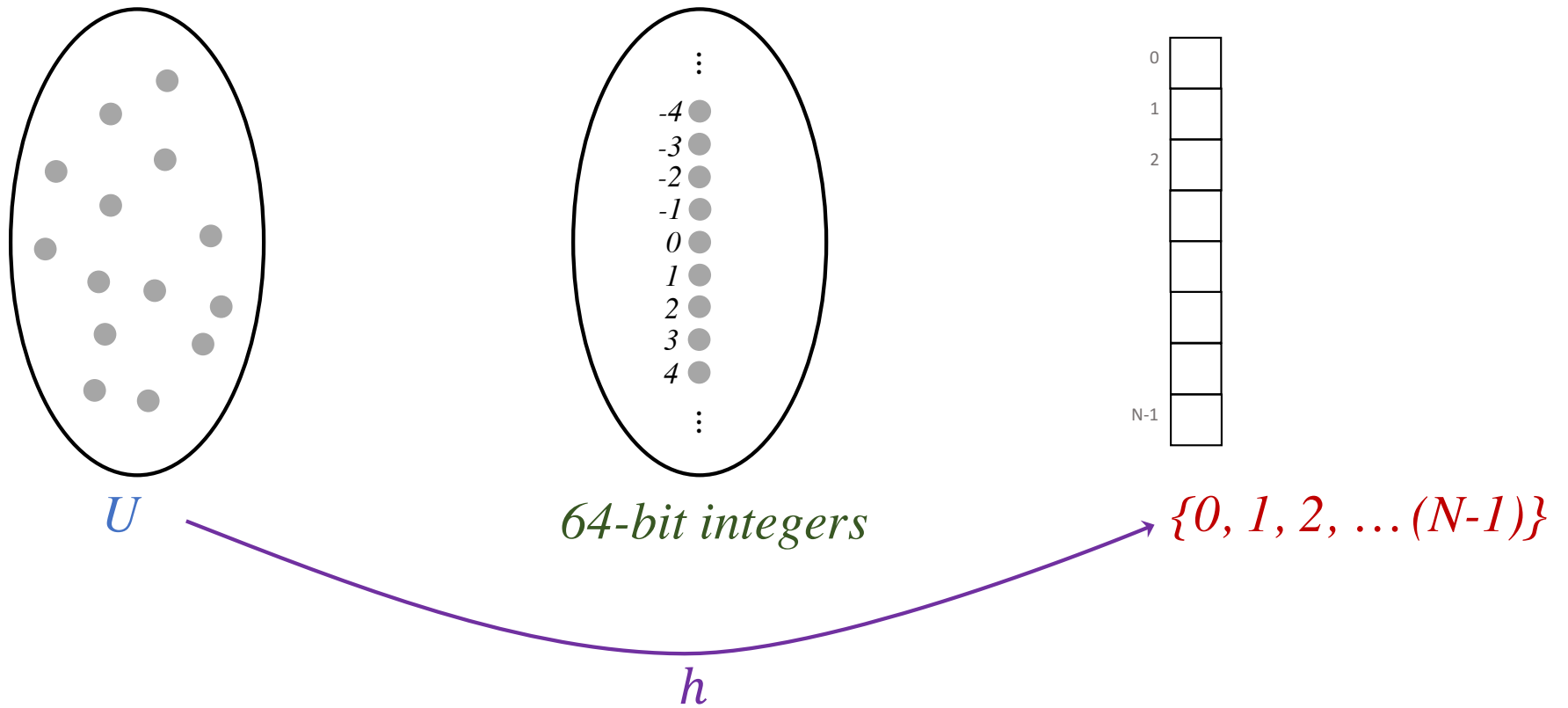
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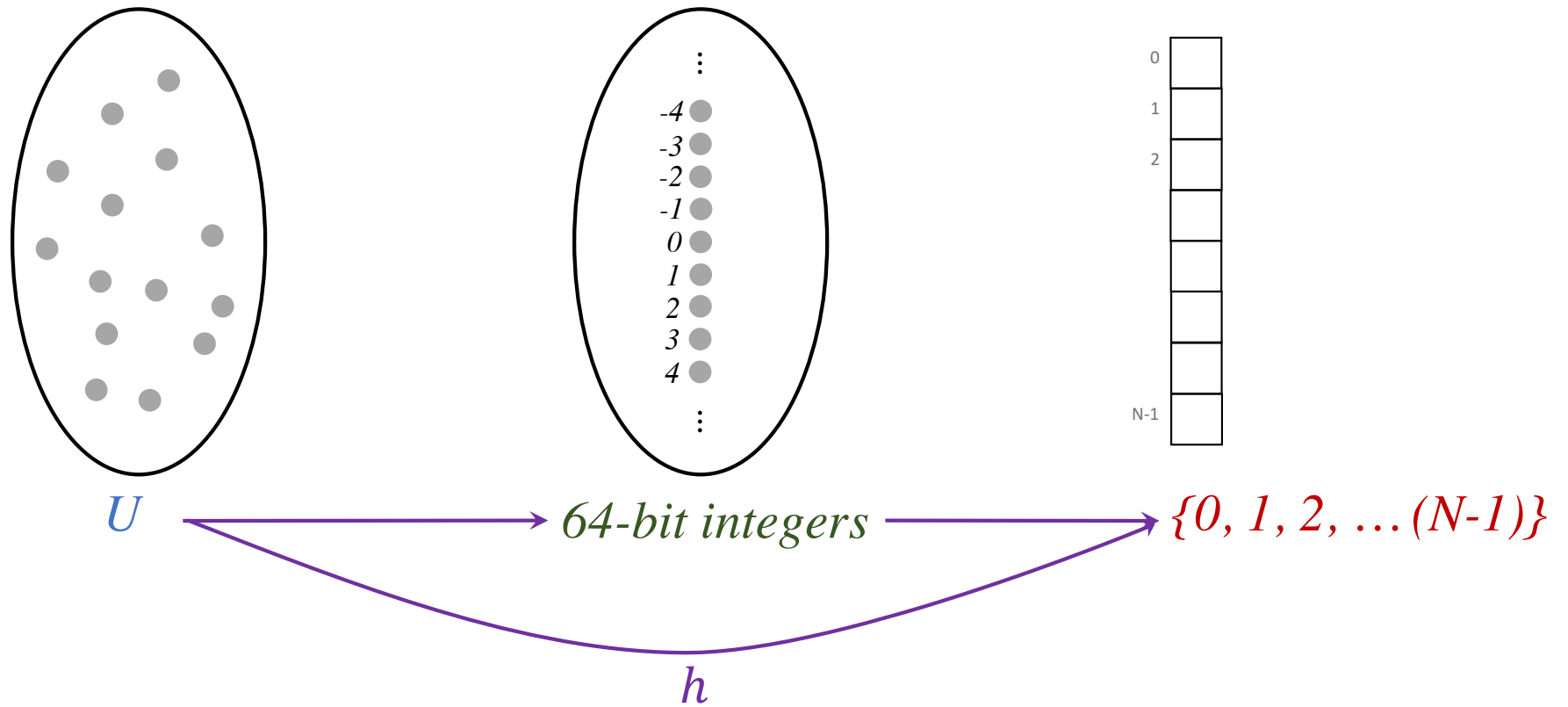
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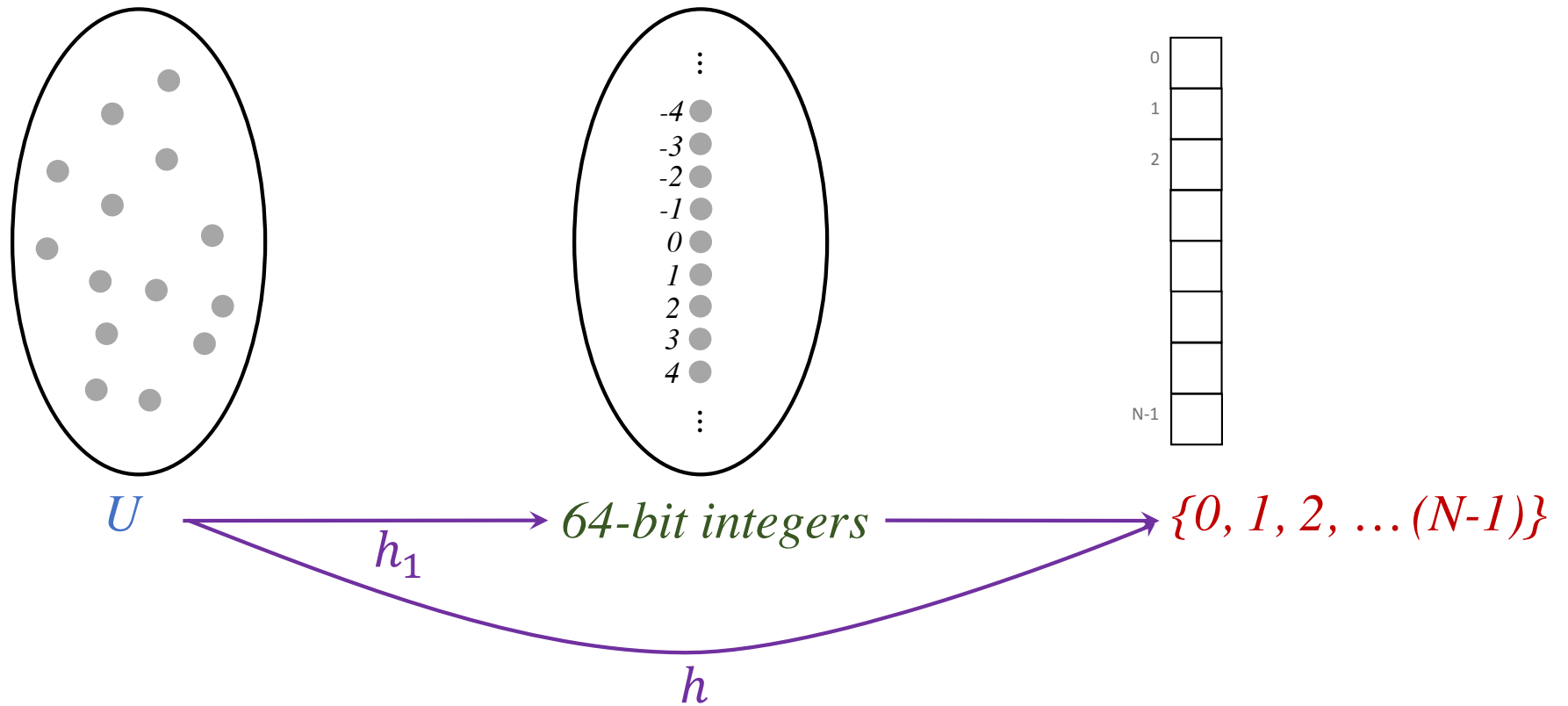
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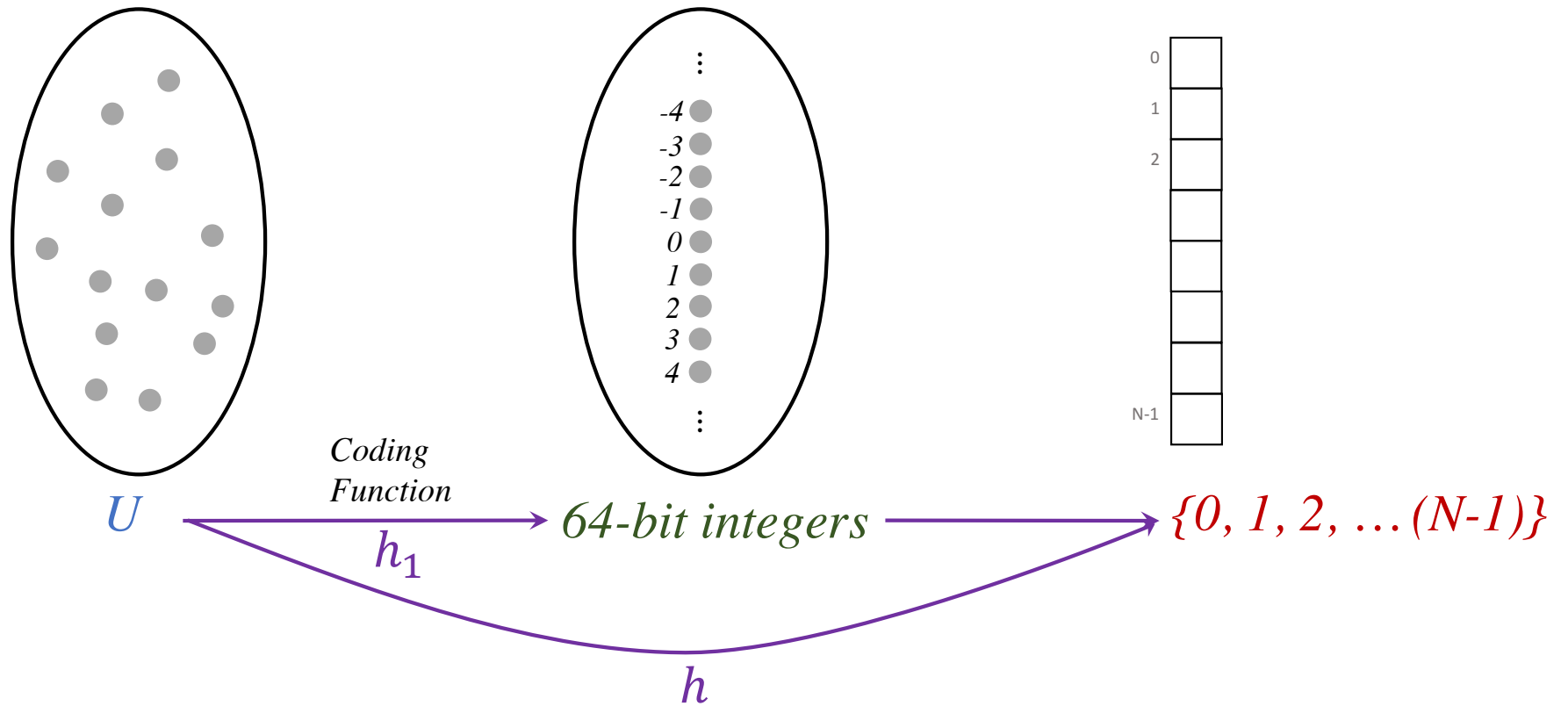
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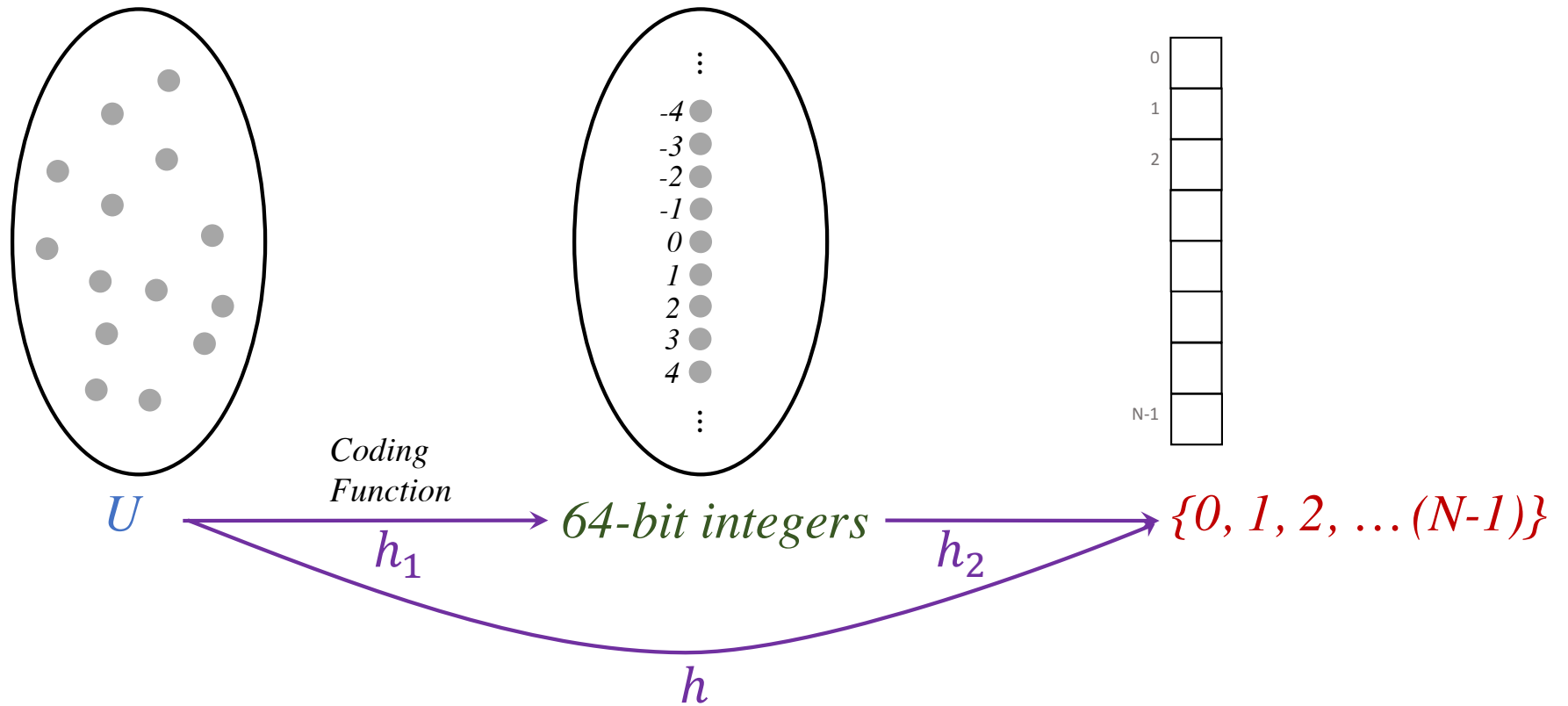
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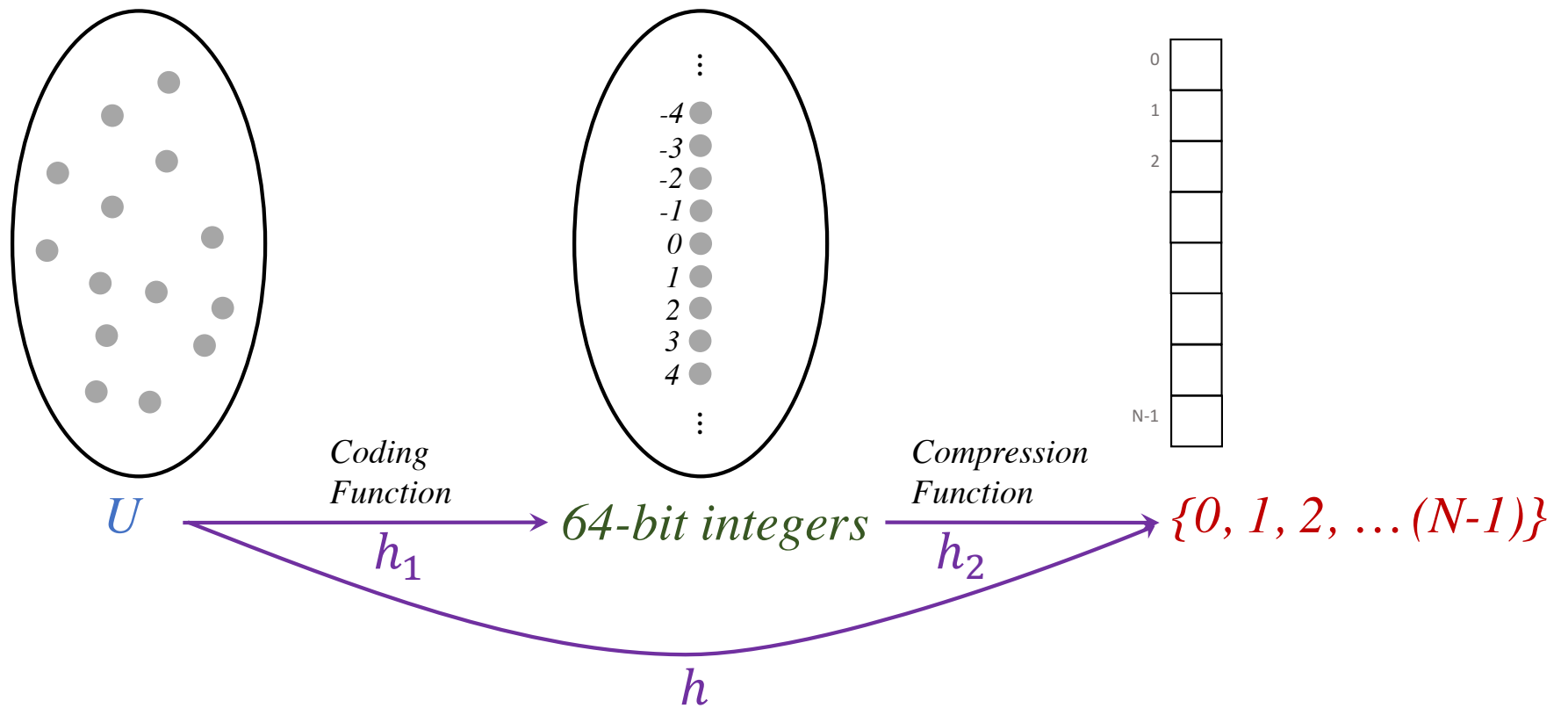
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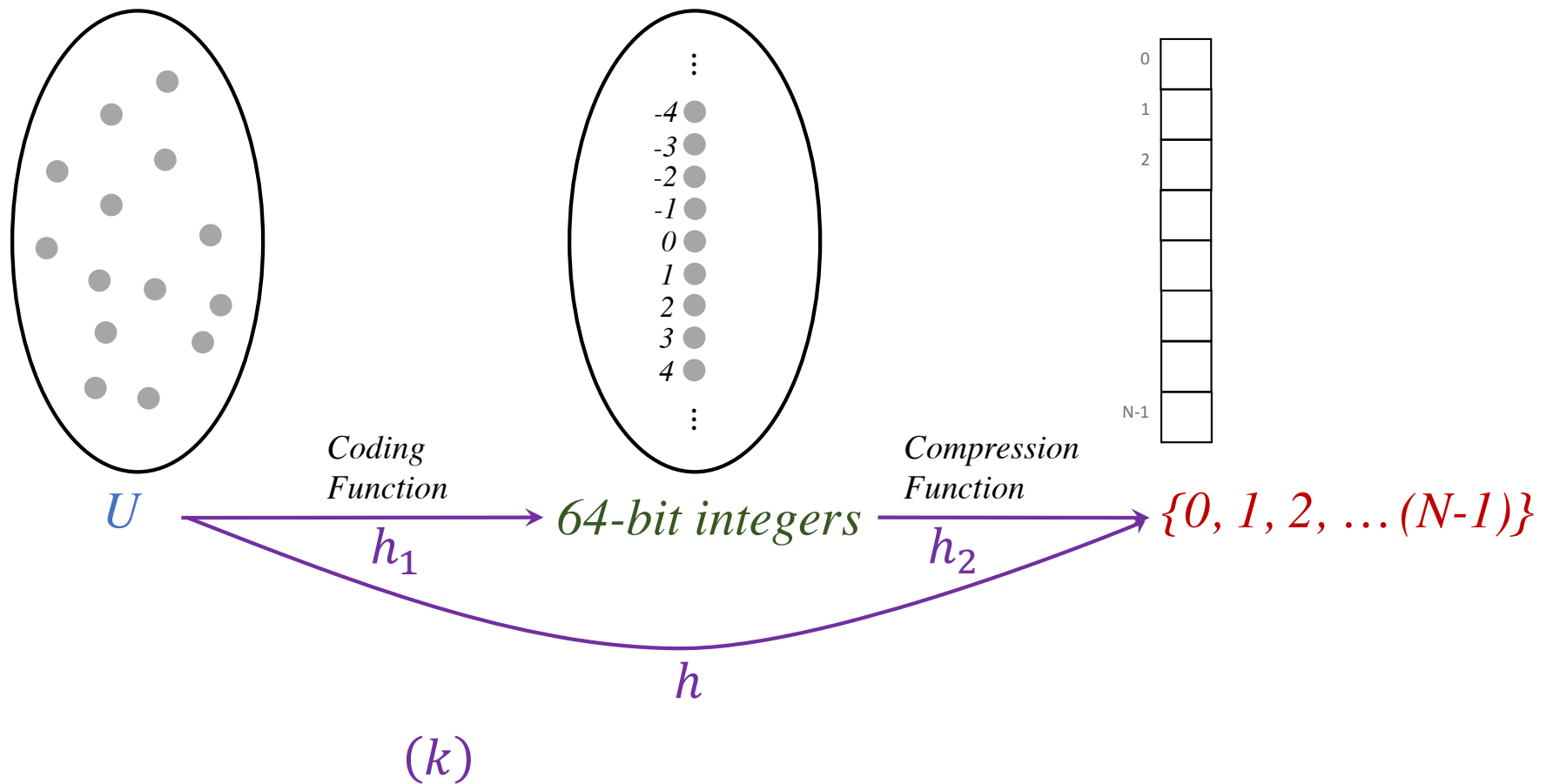
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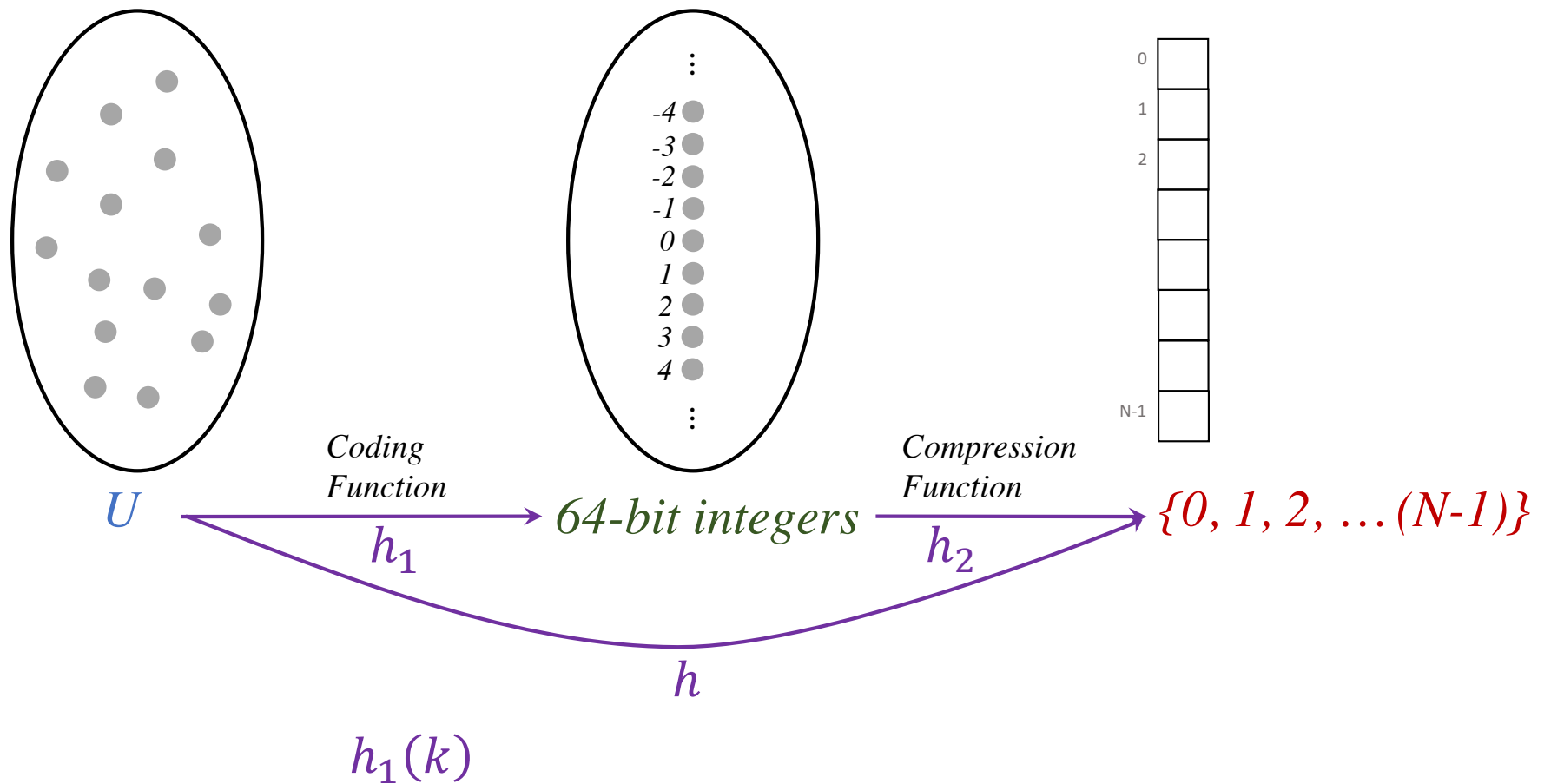
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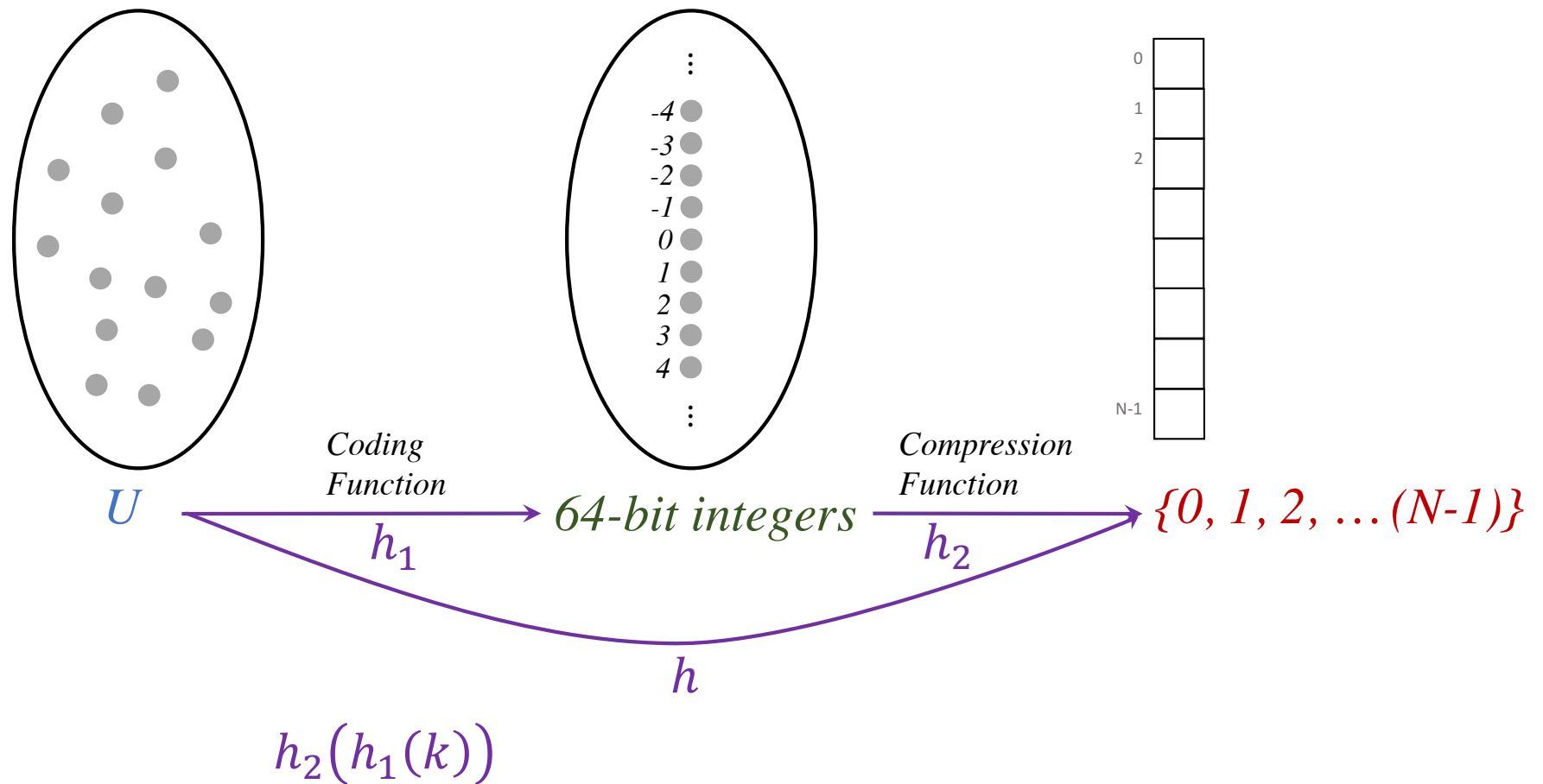
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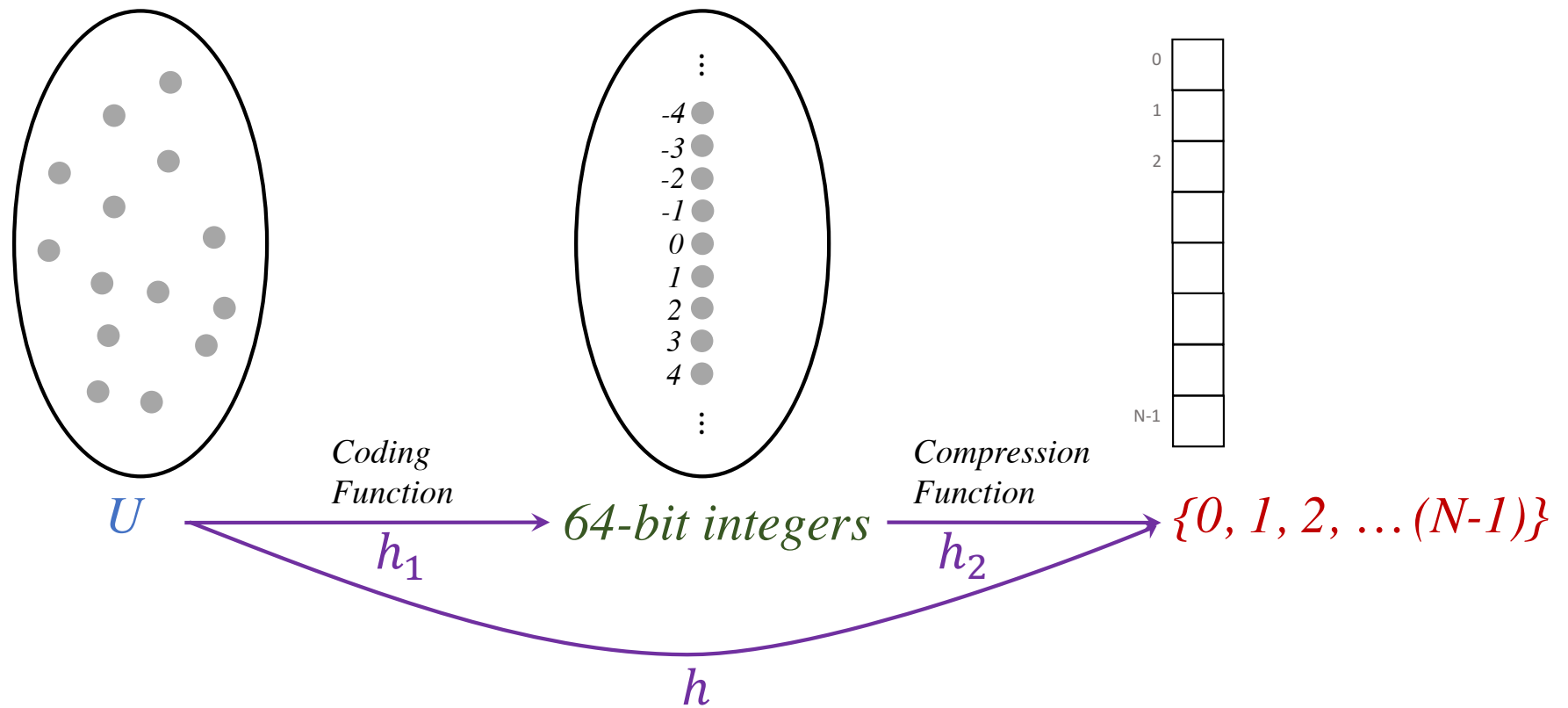
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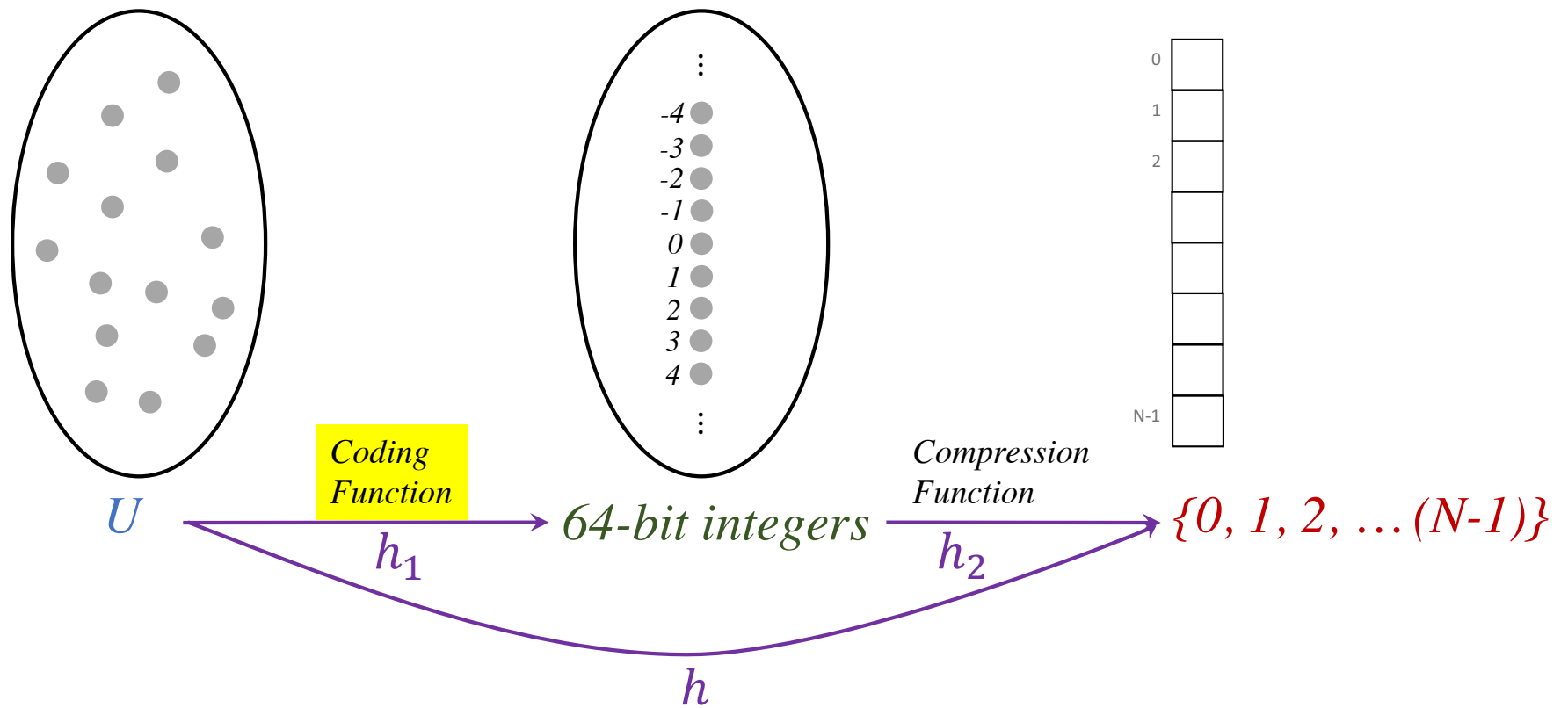


Hash Functions



$$h(k) = h_2(h_1(k))$$

Hash Functions



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Look at the binary representation of *key*,

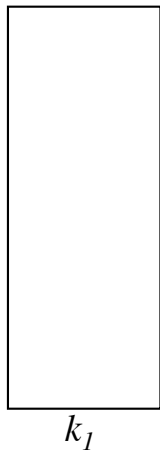
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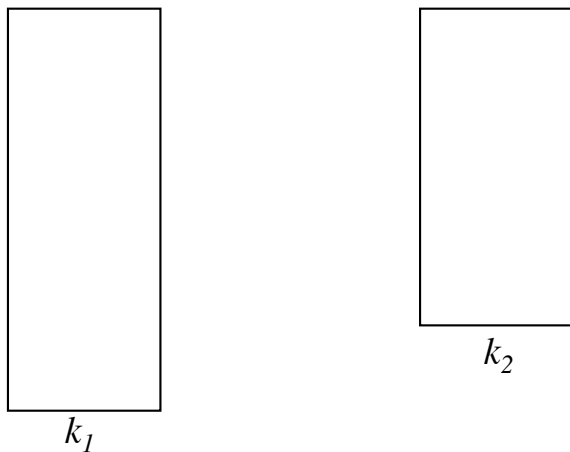
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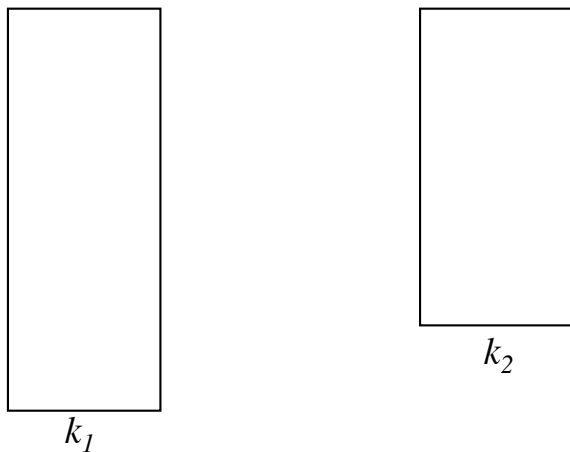
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Take the 8 least significant bytes, and interpret it as a 64-bit 2's complement number



Coding Functions

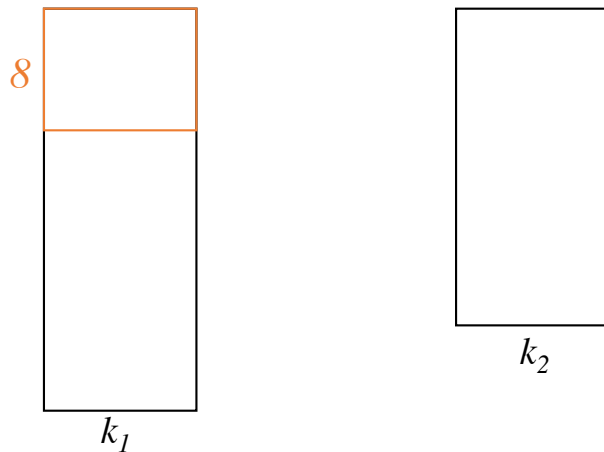
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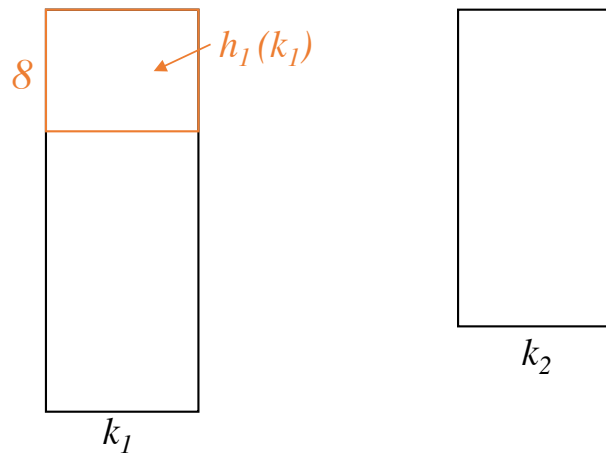
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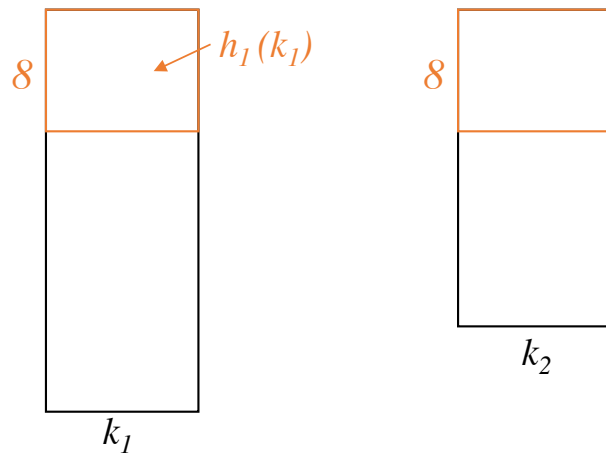
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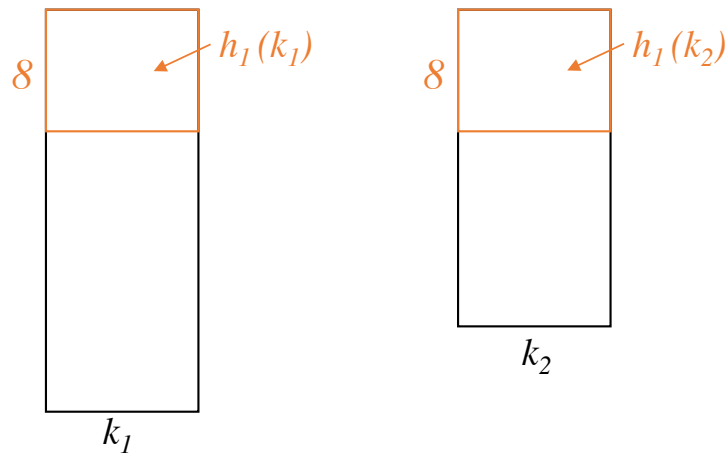
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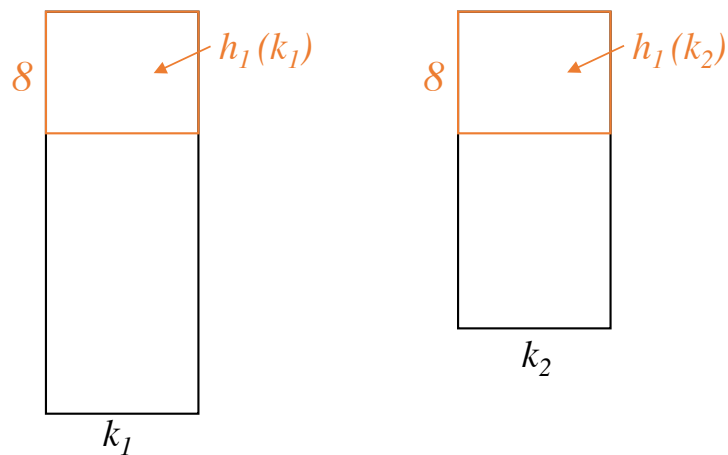
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Problem:

This approach ignores part of the data. In a biased set of keys, these parts could be where the keys differ

Coding Functions

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i. Common Approaches:

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Coding Functions

$h_1: U \rightarrow (64\text{-bit integers})$

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Break key to its components: $key = (k_0, k_1, k_2, \dots, k_{m-1})$.

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Break key to its components: $key = (k_0, k_1, k_2, \dots, k_{m-1})$.

The coding function h_1 would add all the components of key .

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That is: $h_1(key) = k_0 + k_1 + \dots + k_{m-1}$

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“stop”

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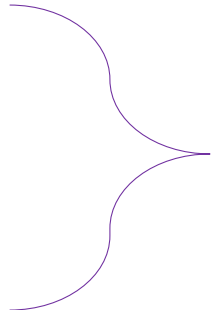
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all coded to the
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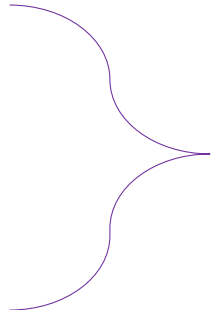
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Problem:

This approach doesn't take the positions of the components into account

Coding Functions

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i. Common Approaches:

- Integer Casting
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Coding Functions

$h_1: U \rightarrow (64\text{-bit integers})$

i. Common Approaches:

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Fun Fact:

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Fun Fact: If we take $z = 33$

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Fun Fact: If we take $z = 33$, When coding 50,000 English words, have at most 6 collisions

Coding Functions

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- Can be called with built-in **immutable** types (`int`, `str`, `float`, `tuple`)

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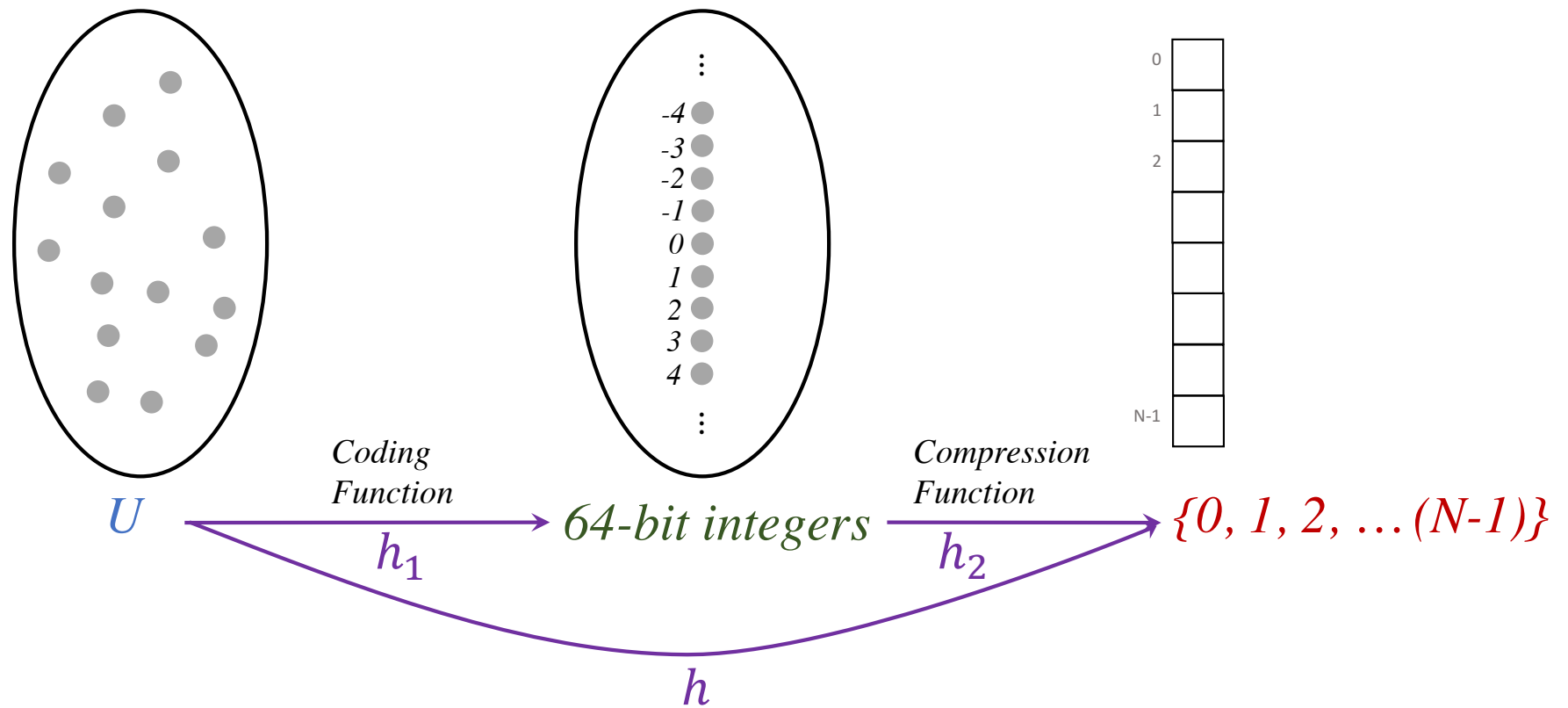
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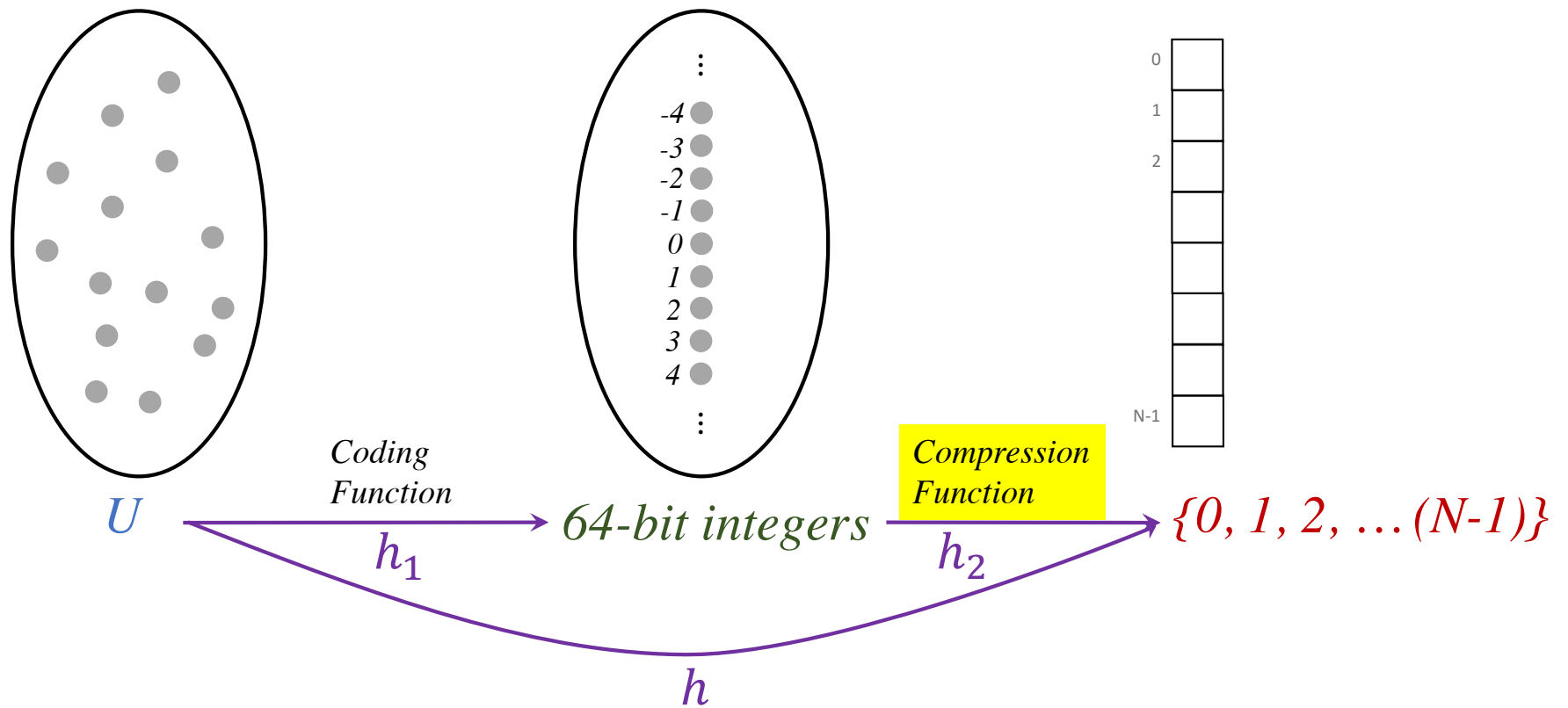
- Can be called with built-in **immutable** types (`int`, `str`, `float`, `tuple`)
- User defined classes are unhashable, unless they overload the `__hash__` method
- Make sure that $(x = y) \rightarrow \text{hash}(x) = \text{hash}(y)$

Hash Functions



$$h(k) = h_2(h_1(k))$$

Hash Functions



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Compression Functions

$h_2: (64\text{-bit integers}) \rightarrow \{0, 1, \dots, N - 1\}$

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For $N=10$:

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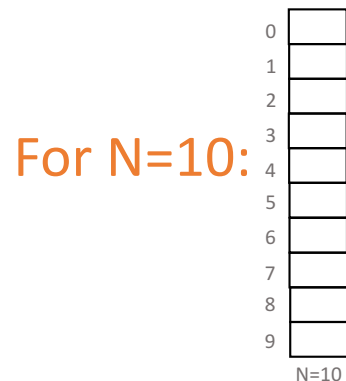
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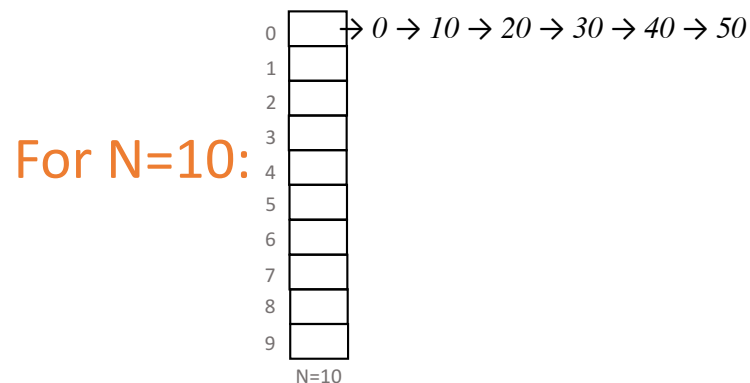
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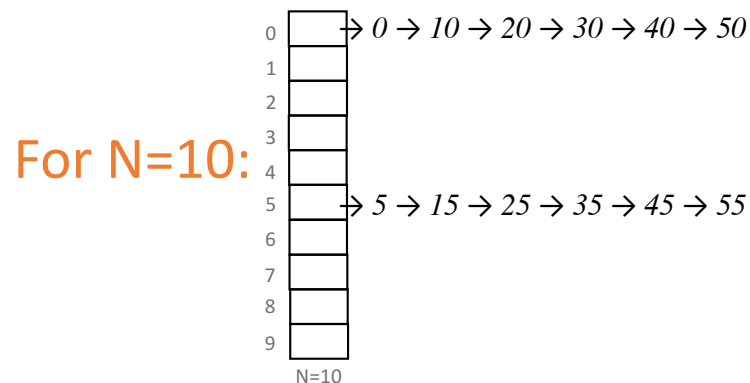
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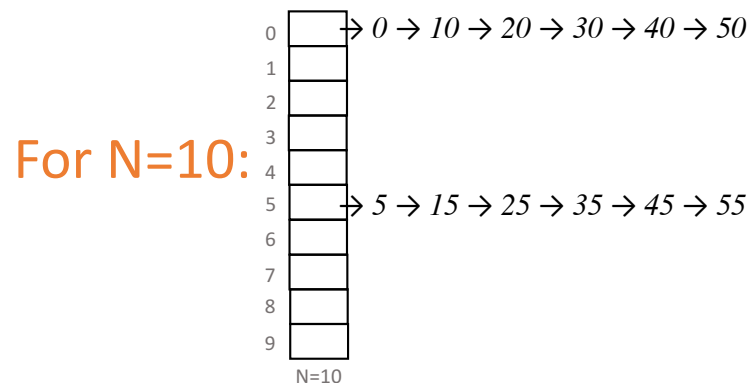
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Example: if our keys are: $\{0, 5, 10, 15, 20, 30, 35, 40, 45, 50, 55\}$:



A common heuristic is to choose N to be prime.

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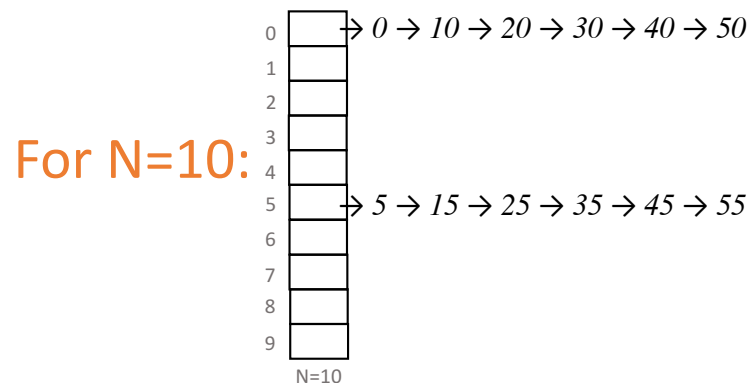
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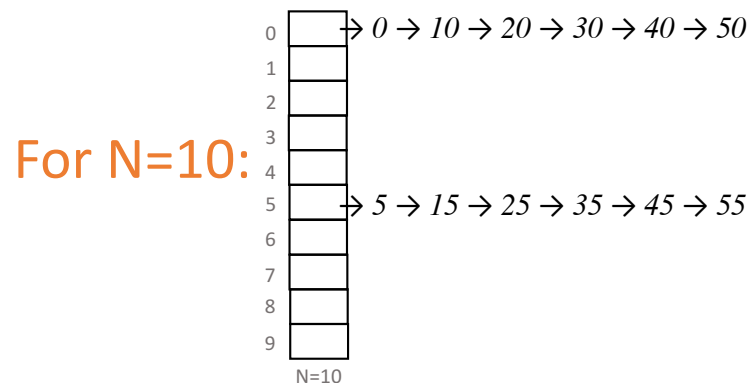
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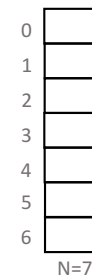
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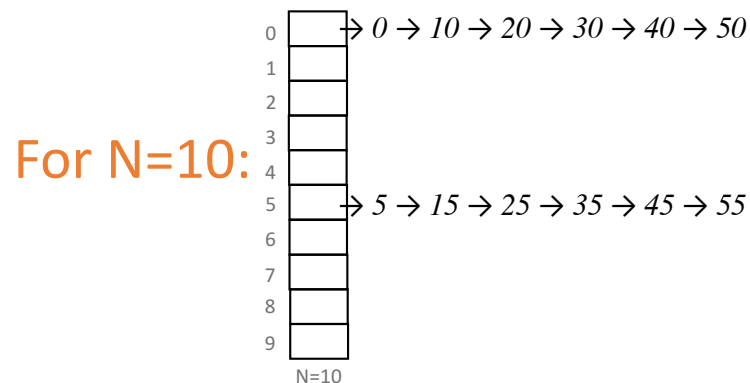
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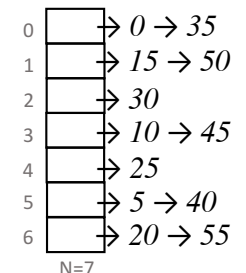
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Common Approaches:

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- The Multiply-Add-and-Divide (MAD) Method

Let p be a prime number, $p > |U|$

Let a be a random number from $\{1, 2, \dots, p-1\}$

Let b be a random number from $\{0, 1, 2, \dots, p-1\}$

We define: $h_2(k) = [(ak + b) \bmod p] \bmod N$

Compression Functions

$h_2: (64\text{-bit integers}) \rightarrow \{0, 1, \dots, N - 1\}$

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0	
1	
2	
3	
4	
5	
6	
7	
8	
9	

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0		→ 5 → 20
1		→ 35 → 50
2		
3		→ 10
4		→ 25 → 40
5		→ 55
6		→ 0
7		→ 15 → 30
8		→ 45
9		

N=10

Time Analysis of *Find*

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Worst-Case:

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Worst-Case: all keys are mapped to the same slot, it takes $\theta(n)$ to scan that chain

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Calculate the hash function and Access slot

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If we always maintain $n \leq N$

Time Analysis of *Find*

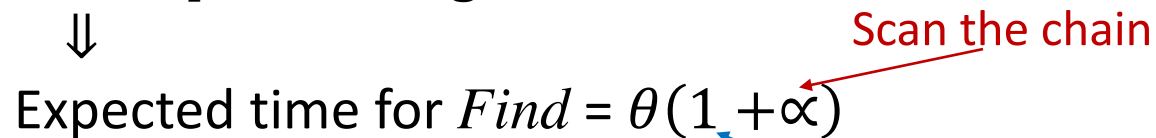
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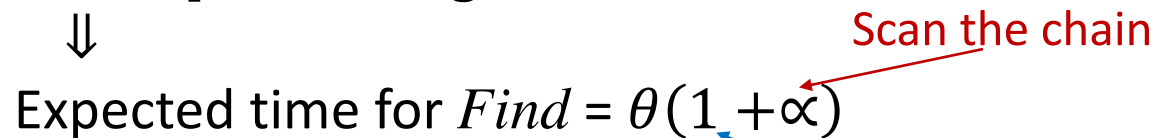
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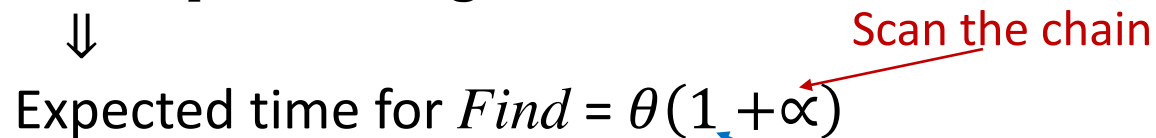
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If we always maintain $n \leq N \Rightarrow \alpha \leq 1 \Rightarrow$ Expected time for *Find* = $\theta(1)$

Implementation Points

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- Collision resolution:

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- Performance: always keep $n < N$