Database Processing CS 451 / 551

Lecture 6: Hashing





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Assignment 1 is Out! Deadline: Oct 28, 2025 at 11:59pm

Start collaborating with your groups!

Term Paper for Graduate Students

- Select one area.
- **Select one paper published** in 2025 from the following 4 conferences:
 - No two students can select the same paper.
 - Your selected paper needs my approval.
- VLDB, SIGMOD, OSDI, SOSP.
- Describe the following in 4-page style ACM Sigmod double-column style.
 - What is the paper's goal?
 - How is it meeting its goal?
 - What are the disadvantages of the proposed design and advantages of the proposed design?
 - Explain how can you improve the proposed design?
 - What architectural changes you need to do?
 - How to provide support for queries, say Natural Join?
- Topics:
 - Federated Learning, Vector Databases, Graph Databases, Privacy-Preserving Databases

Unordered Indexing

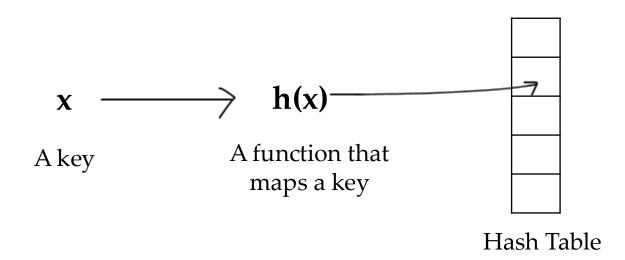
• Until now, we studied ordered indexes, such as clustered indexes and trees.

• Next, we will look at unordered indexes → Hash indexes.

Hashing

Hashing

- Three key components of a hash index:
 - A hash table, which stores all the keys.
 - A **function** that helps to map the key to hash table.
 - An hashing algorithm



Types of Hashing

Types of Hashing

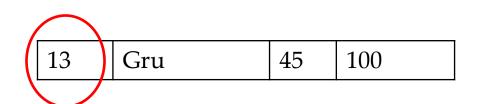
- Two types of hashing schemes:
 - **Static Hashing >** Size of hash map is fixed; cannot be increased.
 - **Dynamic Hashing** → Size of hash map can increase as needed.
 - Essentially as your databases increases over time, you can accommodate more data.

Complexity of Hashing

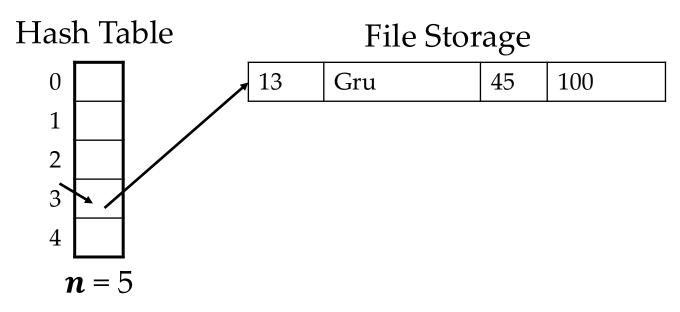
- As hashed indexes are unordered, they do not force maintaining any specific order.
- The position of a key in the hash table is dictated by the hash function.
- Average case complexity for insertion, deletion, and search $\rightarrow 0(1)$
 - But, there are constants, which matter.
- Worst case complexity, given $n \text{ keys} \rightarrow O(n)$
- Hash tables support **random access**, unlike earlier indexes, which support sequential access.

Static Hashing

- Say, we know that in our database there will be 5 records.
- So, we select a hash function and create a hash table (array) of size 5.

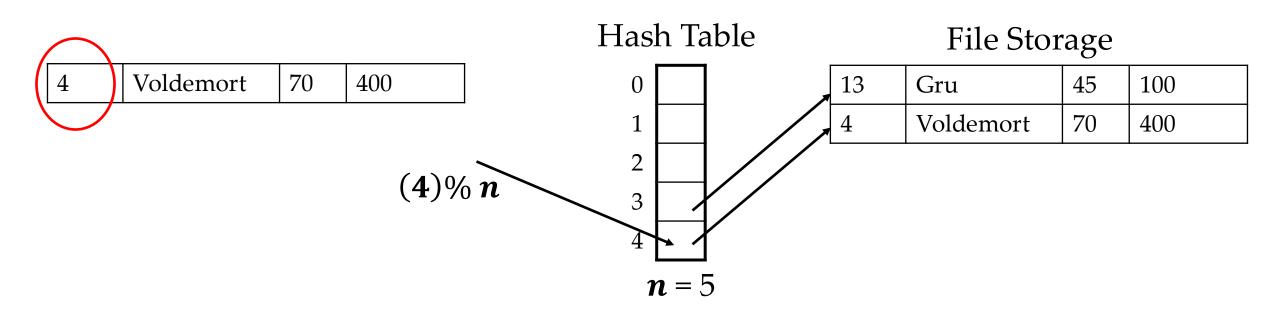


$$h(x) = key \% n = (13)\% n$$



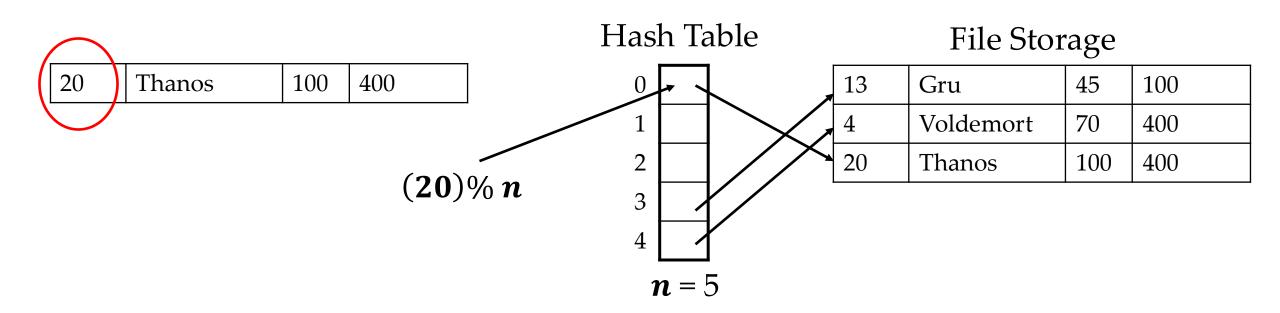
Static Hashing

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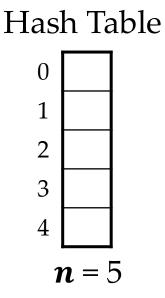
Static Hashing

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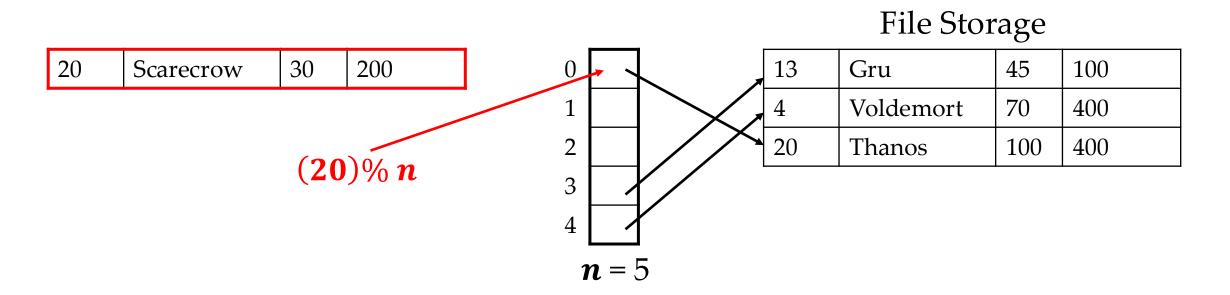


- Fixed number of Keys
- Duplicate Keys
- Collisions
- Disk Access Cost

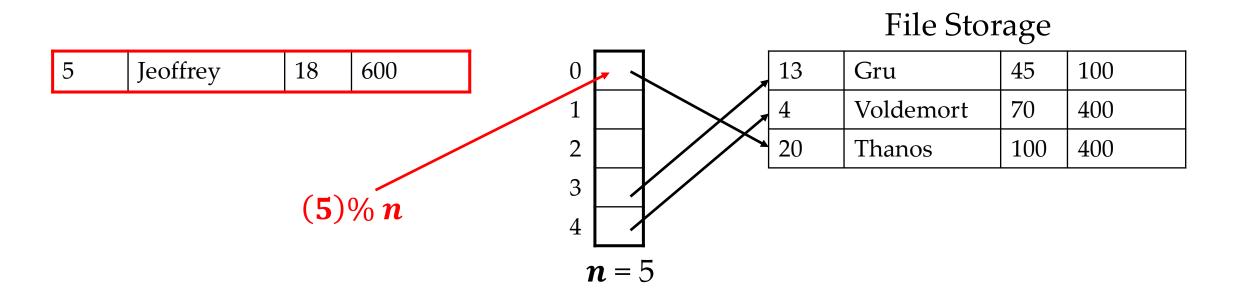
- **Fixed number of Keys** → You should know the total size of the database in the future, and it cannot grow any further!
- For example, this hash table can only store 5 keys and if in the future your database gets a 6^{th} record, you need to reorganize \rightarrow change hash table \rightarrow too expensive!



- Unique Keys → How to store and search for duplicate keys?
- Hash function would map duplicate keys to the same location.
 - Overwrite pointer to existing record?
 - How do you search for an existing record with duplicate keys?



- **No Collisions** \rightarrow Perfect hashing function that ensures there are no collisions.
- Hash function may end up assigning the same location to two or more records.



- Disk Access Cost and Lack of opportunities for Pre-fetching.
- Fetching a single record (point query) is fast. But, say I want to fetch a range of records. These records could be spread **across the disk \rightarrow multiple blocks!**
- No longer sequential access. Moreover, File Manager cannot even predict!

Design Decisions for Static Hashing

Design Decisions for Static Hashing

Good Hash Function:

- Maps a large set of keys to a small array.
- Dilemma b/w using a fast hash function vs. a hash function with low collisions.

• Hashing Algorithm:

- How to handle key collisions when they occur?
- Dilemma b/w allocating a large table to prevent collisions vs. setting up rules that allow storing duplicate and colliding keys!

Hash Functions

- Given an input key, it return an integer representation of that key.
 - Essentially, you can use hash function to convert an arbitrary byte array into a fixed-length code.
- We want a hash function that is both **fast** and has a **low collision rate**.
- Notice that we are allowing collisions as we desire fast hashing!
- Alternatively, you can use a cryptographic hash function, like SHA256.
 - No collisions!
 - Extremely secure → NIST recommended
 - Extremely slow!

Hash Functions

- Fortunately, we don't have to create a hash functions!
- <u>CRC-64</u> (1975)
 - Used in networks for error detection
- MurmurHash (2008)
 - Fast, general-purpose hash function.
- Google CityHash (2011)
 - Fast for keys of short length.
- Facebook XXHash (2012)
 - State-of-the-art
- Google FarmHash (2014)
 - Better version of CityHash; reduced collisions

Hash Schemes Performance

If you want to test the performance of various hash functions, or play with different hash functions, check out <u>SMHasher</u>.

Static Hashing Algorithms

- We will be looking at two common algorithms:
 - Linear Probe Hashing
 - Cuckoo Hashing
- These algorithms are also termed as **open addressing**:
 - Essentially, the key may not be in the location where the hash function points.
- More advanced algorithms (not part of this course)
 - Robinhood hashing
 - Hopscotch hashing
 - Swiss Tables

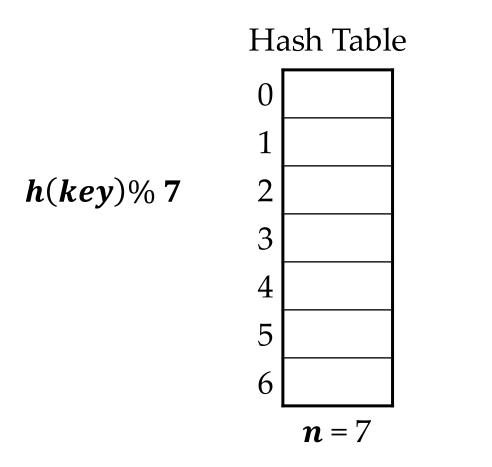
- Simplest hashing algorithm \rightarrow resolves collision by searching for next empty **slot**.
- Requires a fixed-size giant array (smaller the size, more collisions).
 - Hash table's load factor (like a threshold) determines when the table is too full.
 - No new key should be added, otherwise collisions → allocate new table!

• Inserting a key:

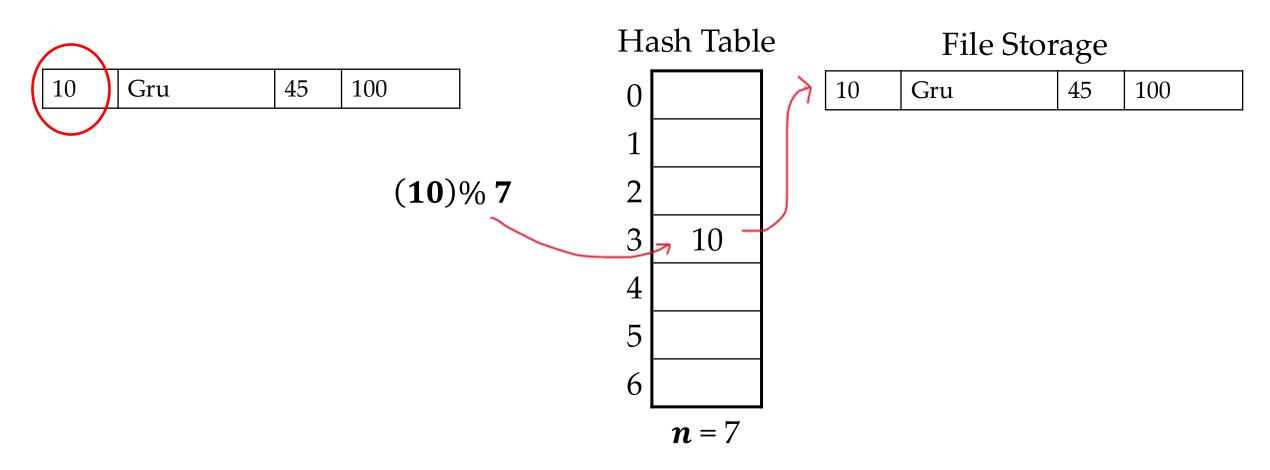
- Use your hash function to find a **slot** (position).
- If the location is empty, store the key in that slot.
- Otherwise, start sequential scanning from that location.
- When you find an empty slot, insert your key in that slot.

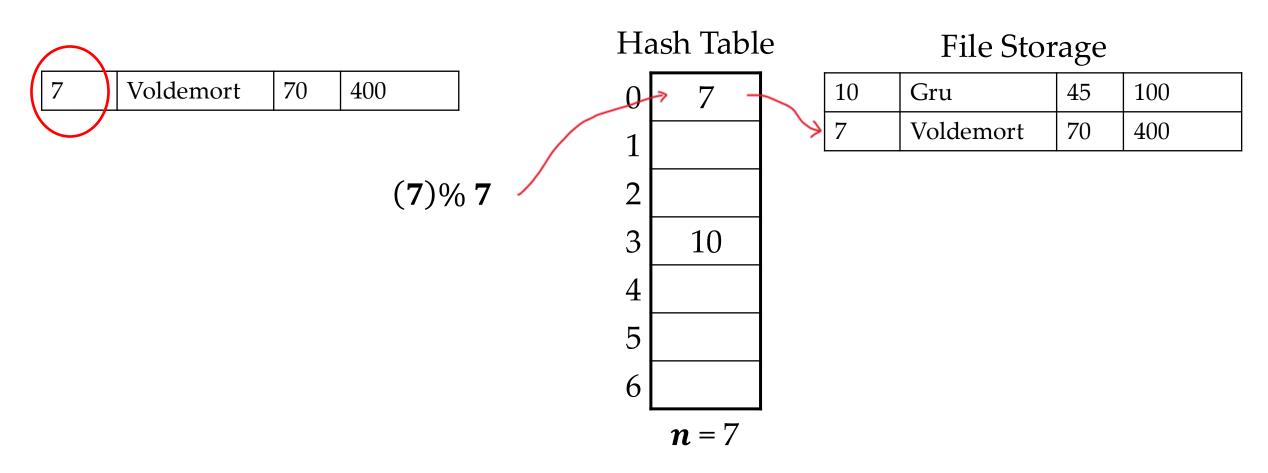
• Deletion and Search:

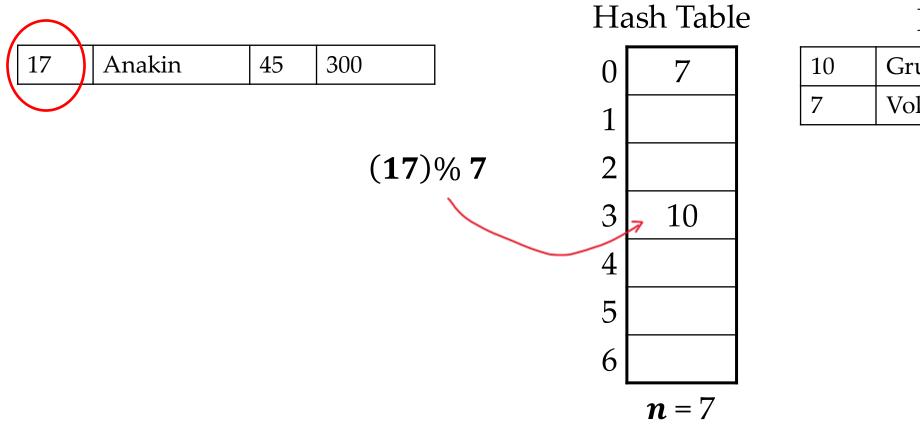
• Same as insertion.



File Storage

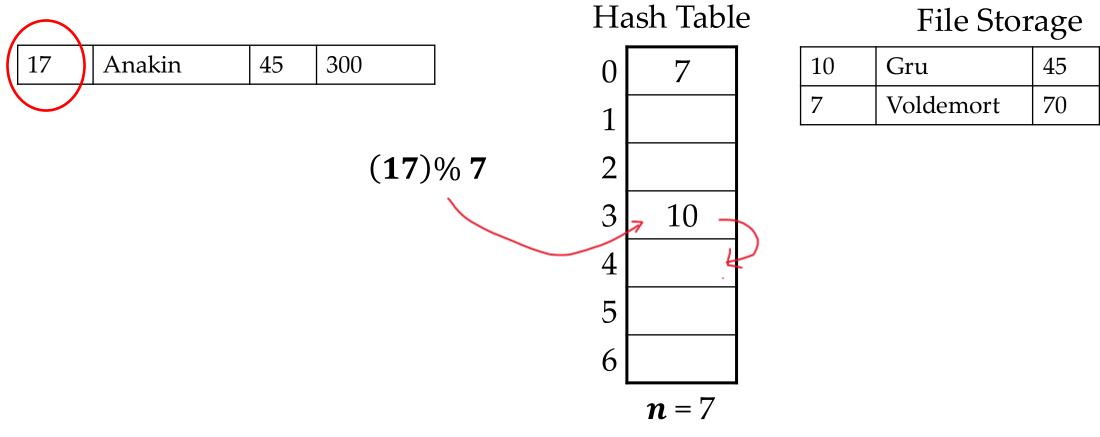




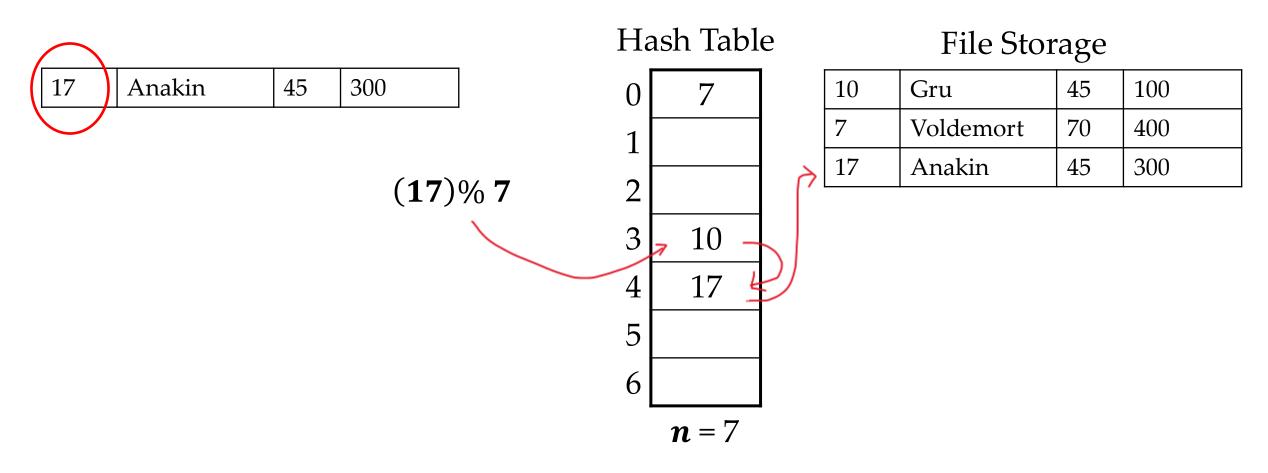


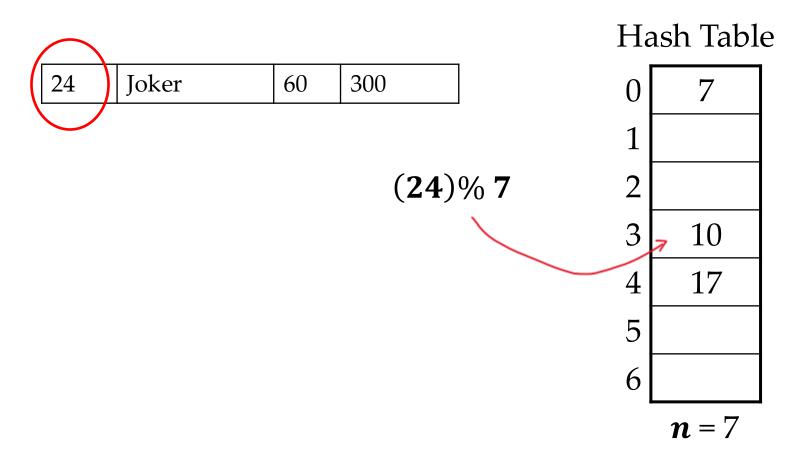
File Storage

10	Gru	45	100
7	Voldemort	70	400



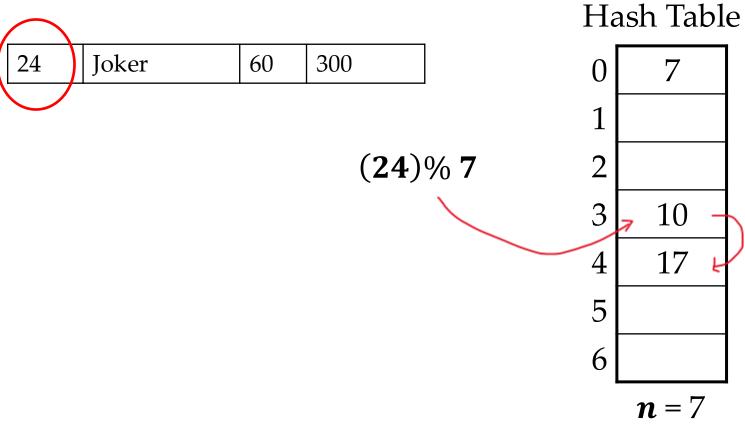
10	Gru	45	100
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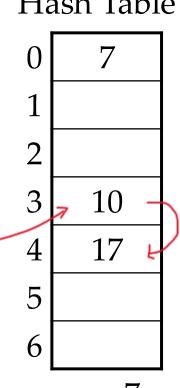




File Storage

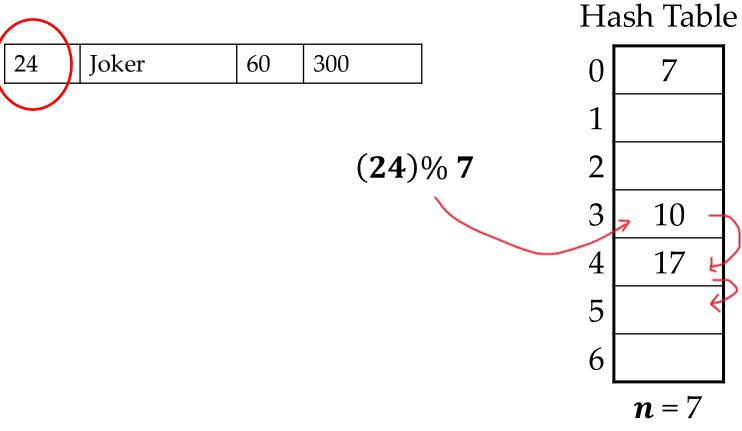
10	Gru	45	100
7	Voldemort	70	400
17	Anakin	45	300

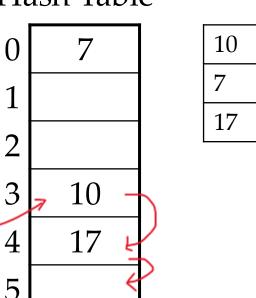




File Storage

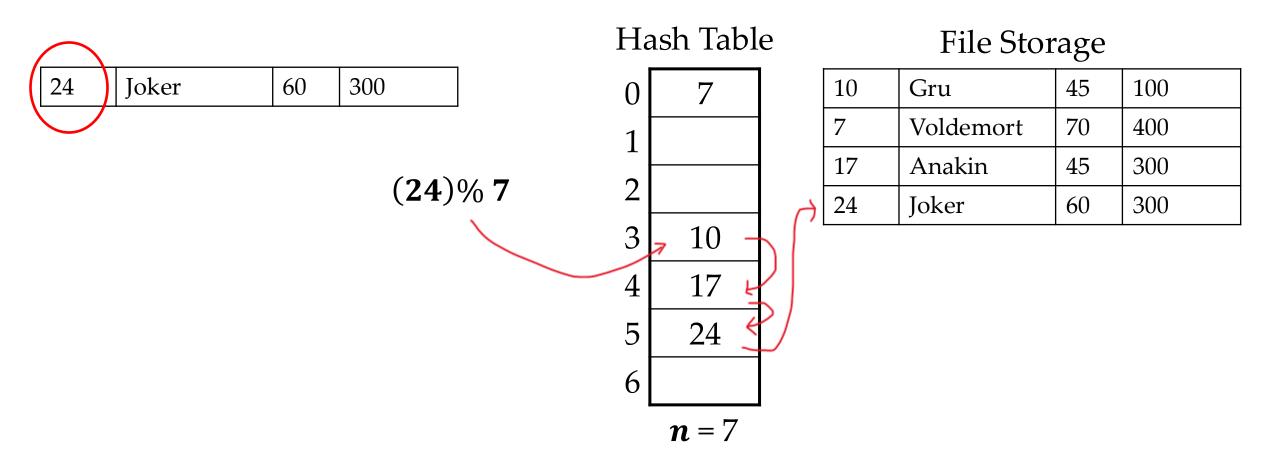
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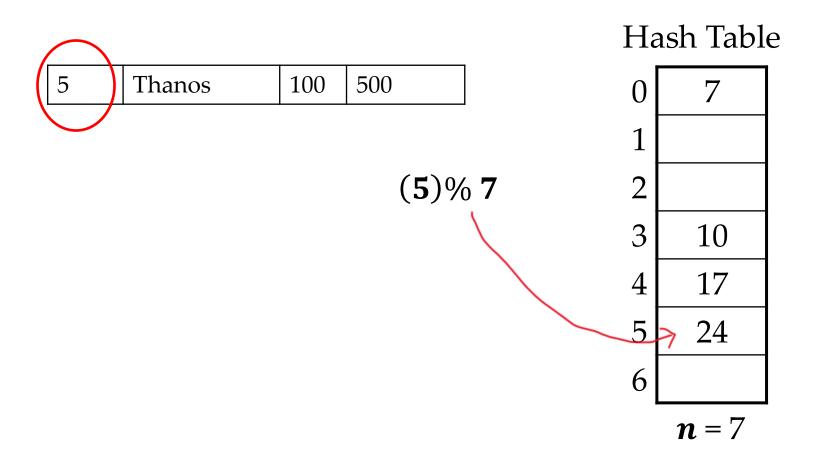


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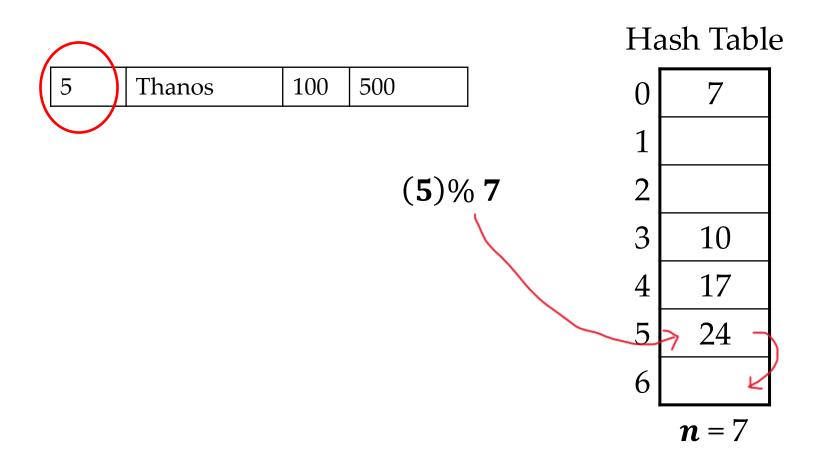
Linear Probe Hashing



File Storage

10	Gru	45	100
7	Voldemort	70	400
17	Anakin	45	300
24	Joker	60	300

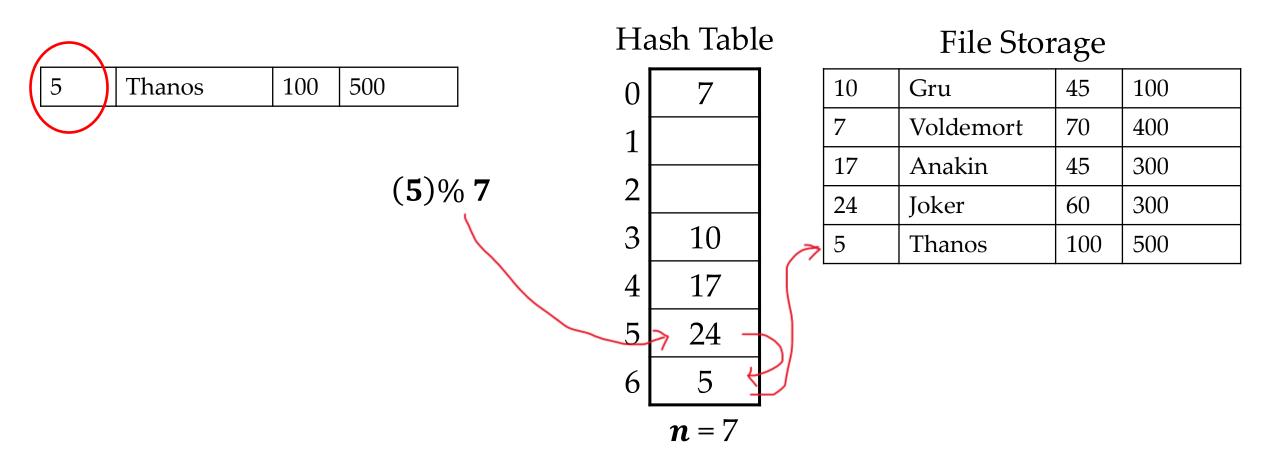
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Linear Probe Hashing

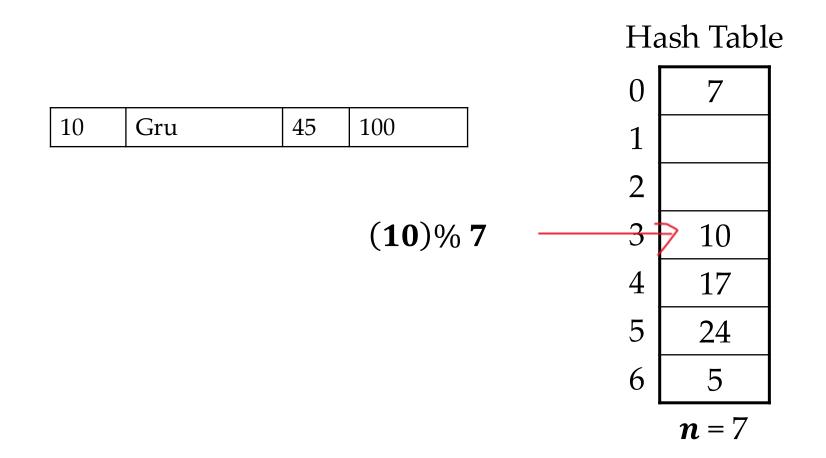


Searching in Linear Probe Hashing

- Follow the same algorithm as you are trying to insert.
 - If the slot is empty, key not found.
 - If the slot is full, then continue to next slot.
 - Stop when you reach an empty slot or have covered all the slots.

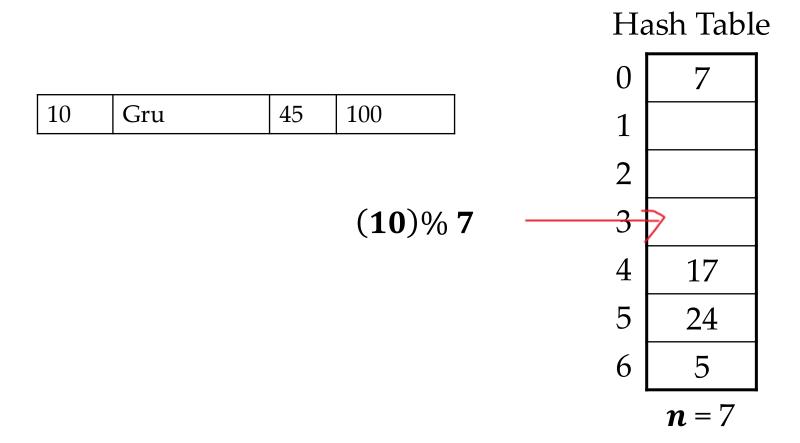
Deleting in Linear Probe Hashing

- How can we delete a record?
- Say, we want to delete the **record 10**, which maps to **slot 3**.



Deleting in Linear Probe Hashing

- How can we delete a record?
- Say, we want to delete the **record 10**, which maps to **slot 3**.
- Can we set slot 3 to **empty**?

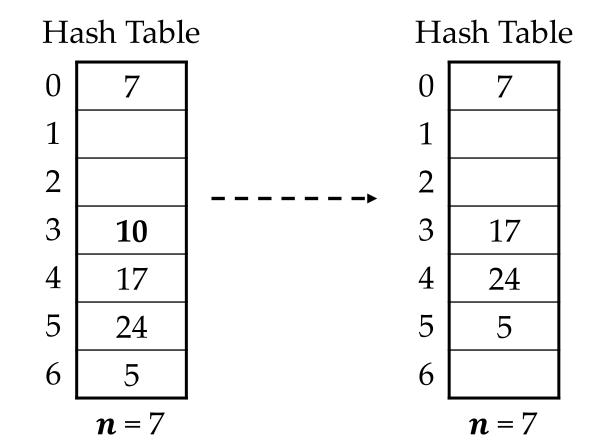


Deleting in Linear Probe Hashing

- On deleting a record, setting a slot to empty is dangerous!
 - Other keys could have also mapped to the same slot, but due to the slot being full, they were in subsequent locations.
 - By emptying the slot, you are indicating that other keys also do not exist!
- Two possible solutions:
 - Rearrangement
 - Tombstones

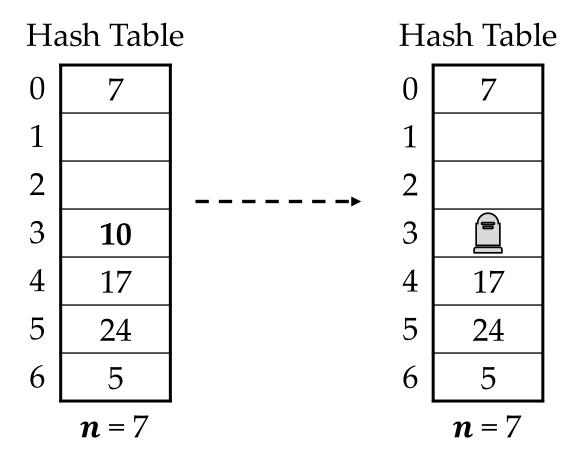
Deletions: Rearrangement

- Once a key is deleted, you can rehash all the keys again.
- Any key that was supposed to be mapped to the same slot can now take place.
- Too expensive! No database does this.



Deletions: Tombstones

- Once a key is deleted, you place a **tombstone for that key** in that slot.
- Tombstone informs any future query that the specific key does not exist.
- However, other keys may still exist!

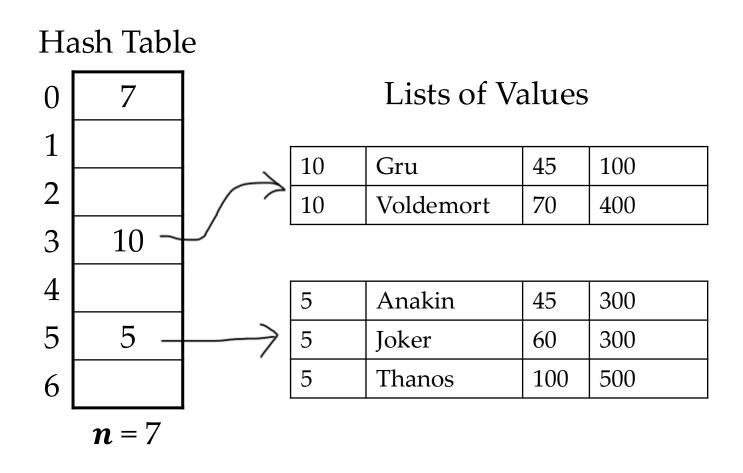


For each tombstone, you need to maintain the list of keys that have been deleted!

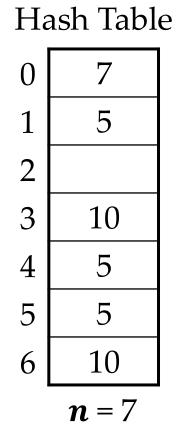
Duplicate Keys in Linear Probe Hashing

- How do you handle **duplicate** (non-unique) keys?
- Two ways:
 - Maintain a list of values
 - Just simply allow adding redundant keys

Duplicate Keys: List of Values

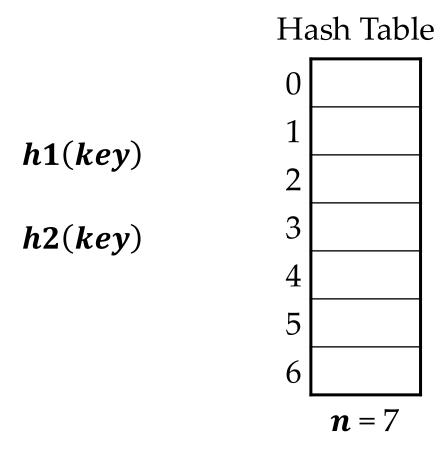


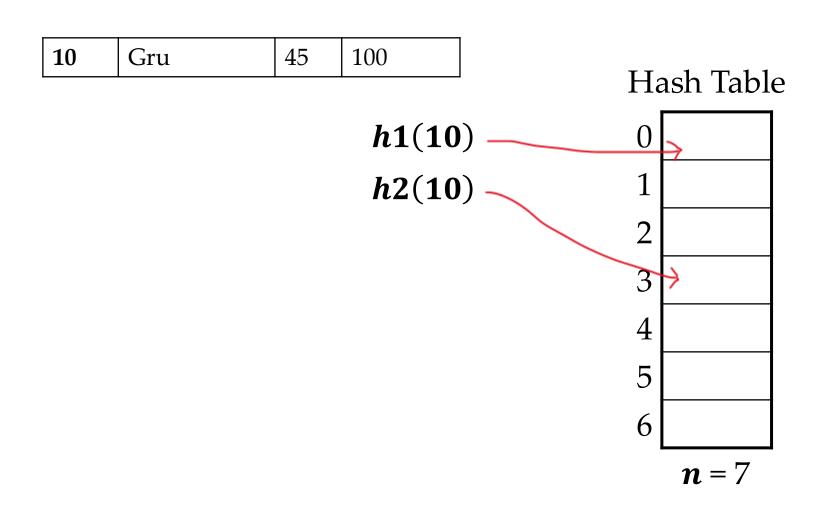
Duplicate Keys: Allow Redundant Keys



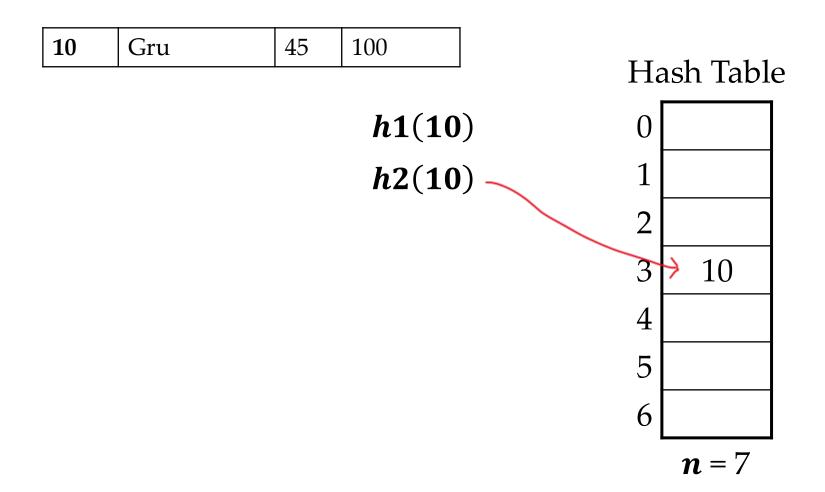
Cuckoo Hashing

- Why the name cuckoo?
- Like the bird cuckoo, if we do not find a free slot for a key, we may kick out an existing key!
- In cuckoo hashing, we use multiple **hash functions** to find free slots to store the key.
 - Each hash function may give us a slot to place and if any of those slots is free, we store the key!
- If no slot is free, evict an existing key!

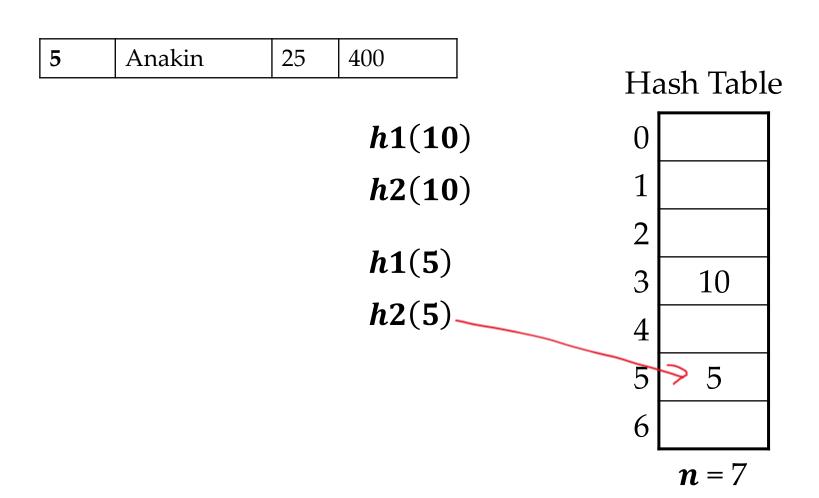




Randomly select a slot, say it selects slot 3 for storing key 10.

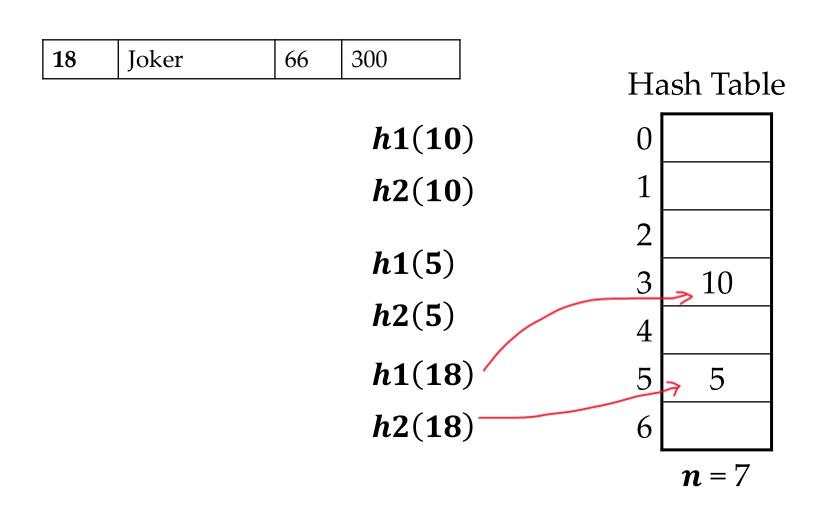


5	Anakin	25	400				_
				ı	Ha	sh Tab	le
			h1(10)))	0		
			h2(10)))	1		
			14(2)		2		
			h1(5)		3	10	
			h2(5)		4	/	
					5	>	
					6		
						n = 7	_



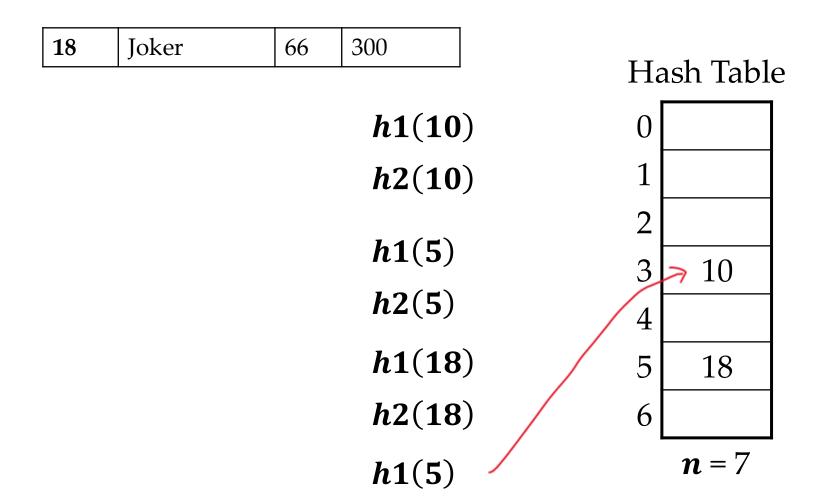
As slot 3 is occupied, we select slot 5 to store key 5.

18	Joker	66	300	
				Hash Table
			h1(10)	0
			h2(10)) 1
			.	2
			h1(5)	3 3 10
			h2(5)	4
			h1(18)	5 5
			h2(18)	6
				n=7

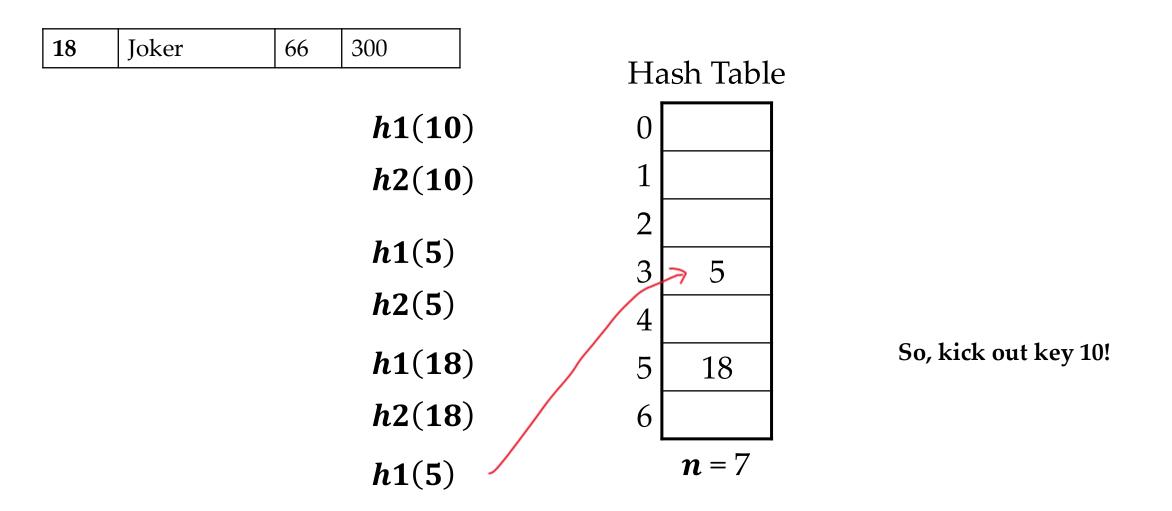


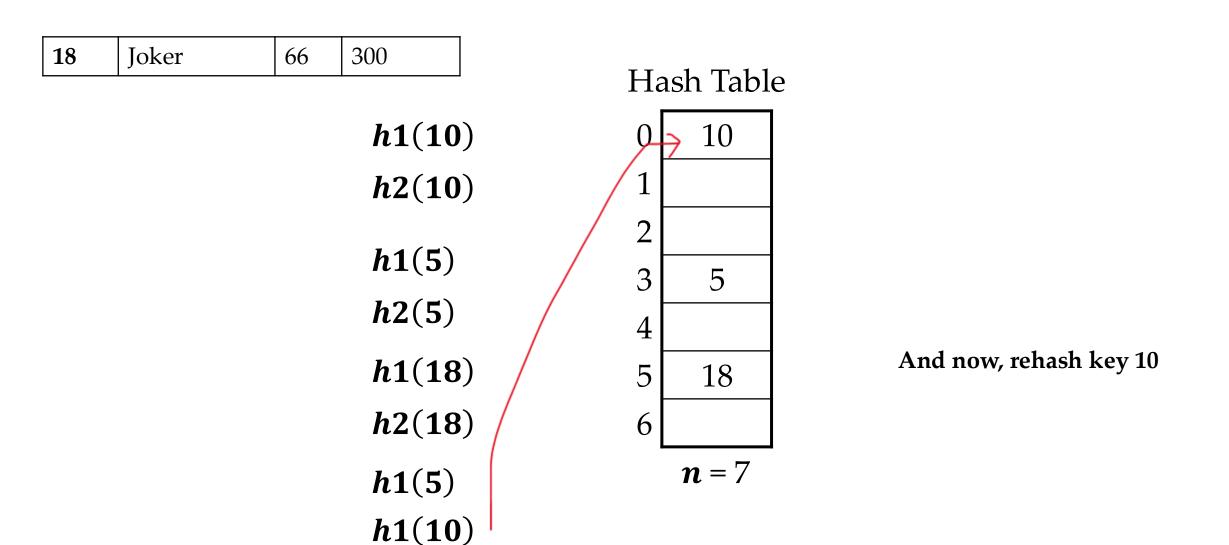
Say, it decides to kick key 5

18	Joker	66	300		Hash Table	
			h1(1	0)	0	
			h2(1)	0)	1	
			h1(5) h2(5)		2 3 10	
			h1(1	8)	4 5 7 18	Say, it decides to kick key 5
			h2(18)	8)	6 $n = 7$	



So, we need to rehash key 5, and only remaining slot is the slot occupied by key 10





Challenges with Cuckoo Hashing

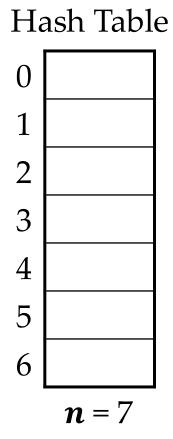
- So what are the challenges with cuckoo hashing?
- Insertions are expensive → We need to do rehashing!
- We can get stuck into an **infinite loop**.
 - To exit the infinite loop, add more hash functions, or increase size of table, or maintain some list.

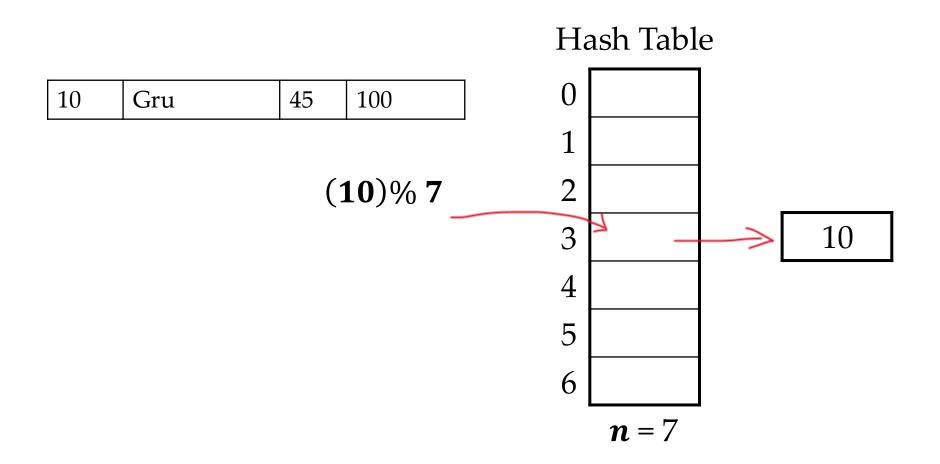
Dynamic Hashing

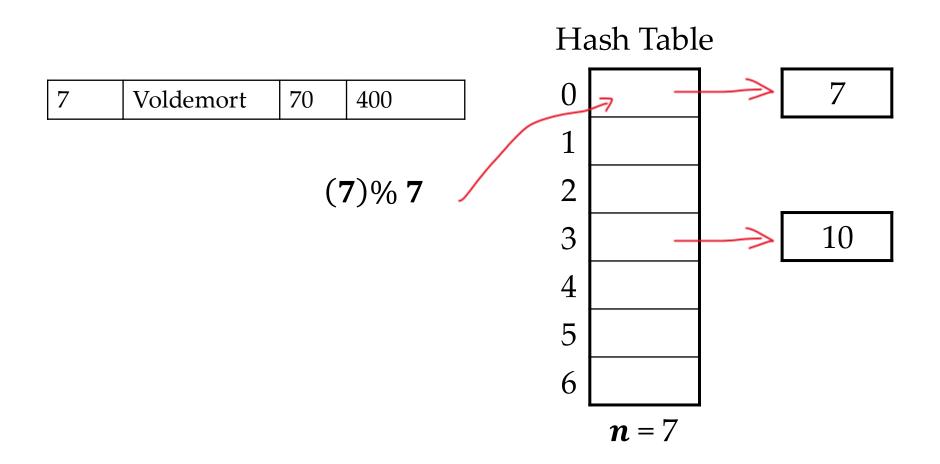
- The biggest challenge for static hashing remains to be **fixed size of hash table**.
- Alternatively, use dynamic hashing algorithms:
 - Chained Hashing
 - Extensible Hashing
 - Linear Hashing

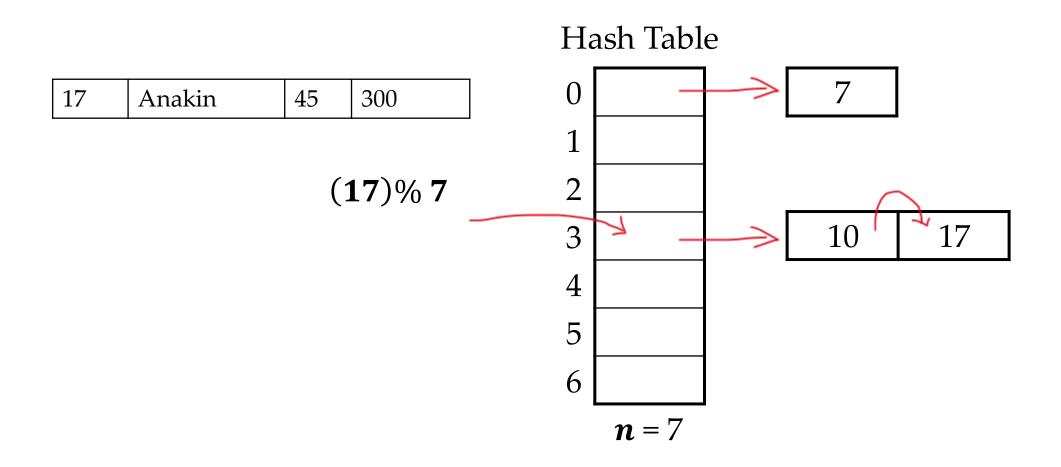
- For each slot in the hash table, there is a **linked list of buckets**.
- Essentially, collisions are resolved by **placing all keys with the same slot** into same linked list.
- Searching for a key requires scanning the linked list till you find the key or have reached end of the list.

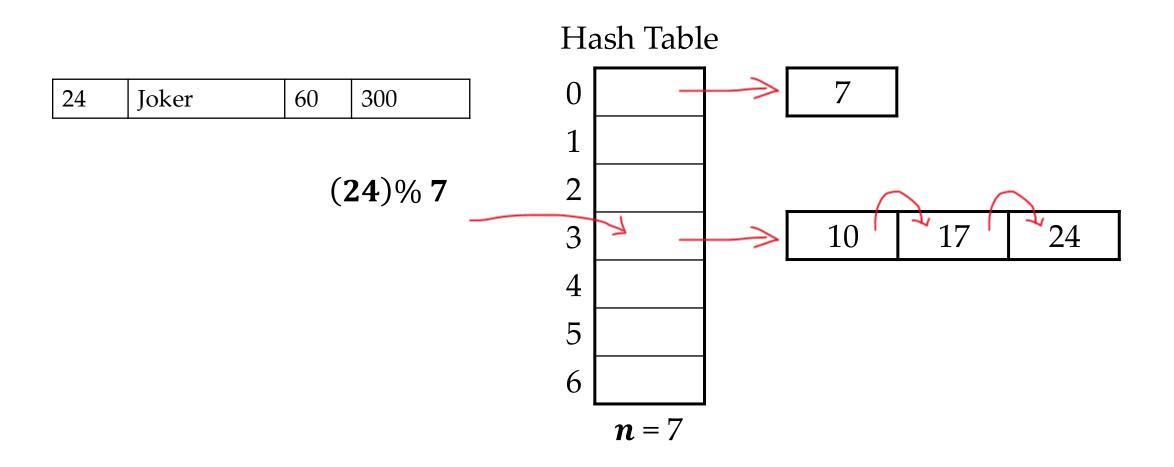
Simple hash function
$$h(key) = key \% n = (key) \% 7$$











Searching in Chained Hashing

• Use the hash function to reach the specific slot, and then scan the linked list till you find the key or have reached end of the list.

• For example, on searching 17, you would first reach slot 3, and then scan the list for slot 3 and find it is as the second entry in the linked list.

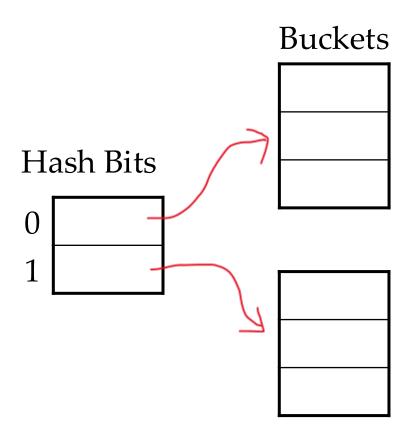
Challenges with Chained Hashing

- What is the key challenge with chained hashing?
- If a lot of keys are hashed to the same slot, then
 - You have a massively large linked list, and
 - Searching a key comes expensive \rightarrow same cost as linear scan.

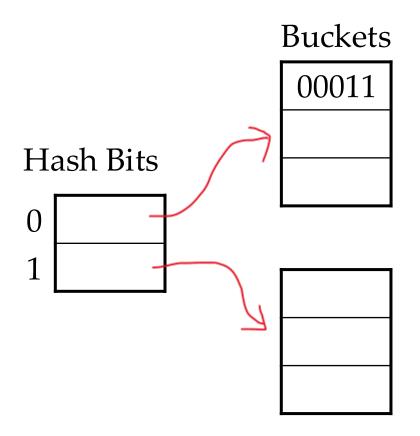
Extensible Hashing

- Solves the problem of massively large linked lists.
- Requires linked lists to be split, when size crosses a threshold.
- Requires observing each key in a bit format.
- When you hash a key, you get a numeric (base-10 or base-16) representation.
 - You can convert that base-10 to binary format (base-2).
- For example: 4 can be represented as 100 in a 3-bit representation.

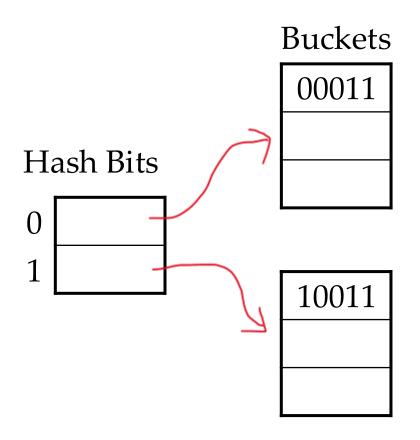
• Initially, your hash map is **1-bit**, and you have some fixed number of buckets for each bit → Say 3 buckets.



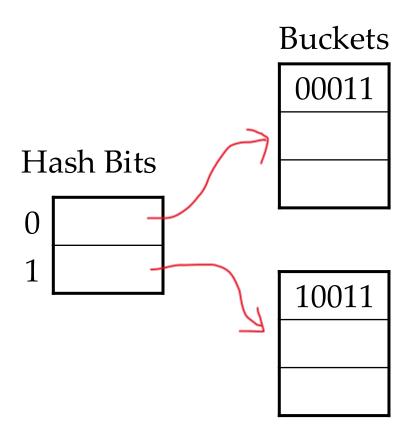
• Assume on passing the **key 13 through a hash function**, the binary representation is **00011**.



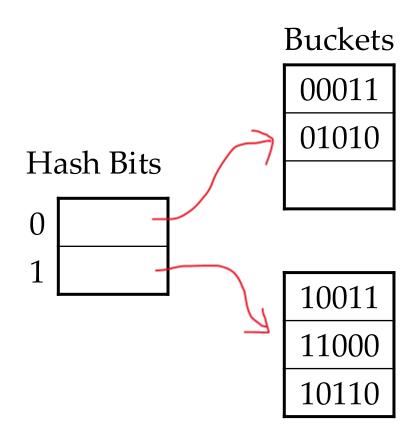
• Another **key 7**, after passing it **through a hash function**, let the binary representation be **10011**.



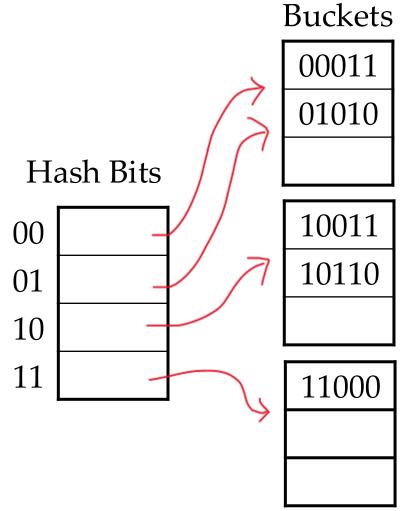
• This way, all keys with binary representation starting from 0 go to buckets for bit 0, and vice versa for buckets for bit 1.



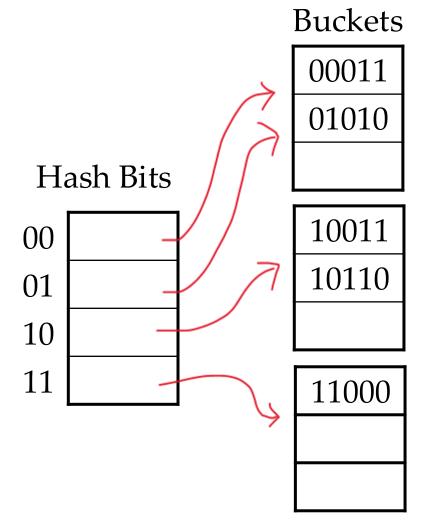
• Let's assume all the buckets for **bit 1** are full.



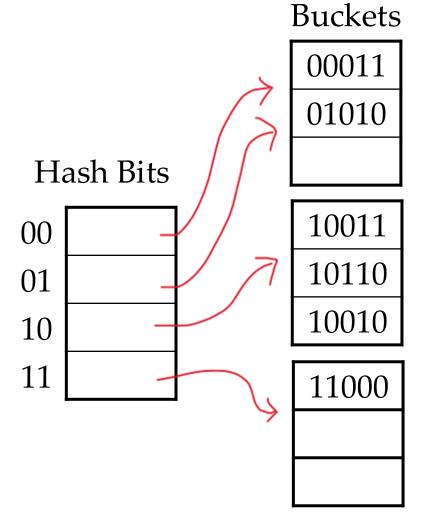
• So, now we need to **split the buckets** for bit 1. This will require expanding the bit representation from 1-bit to 2-bits.



• Notice that all the **2-bit** representations **starting with bit-0** continue pointing to the old buckets.

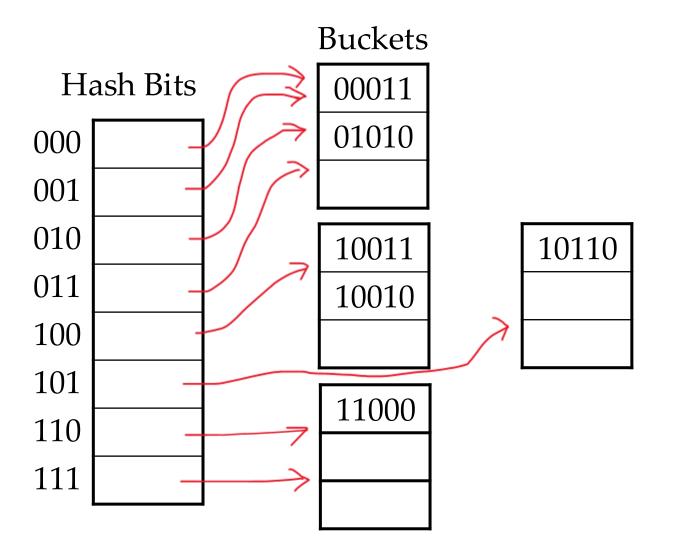


• Next, assume we received a **key 18**, and on passing it **through the hash function**, the binary representation is **10010**.



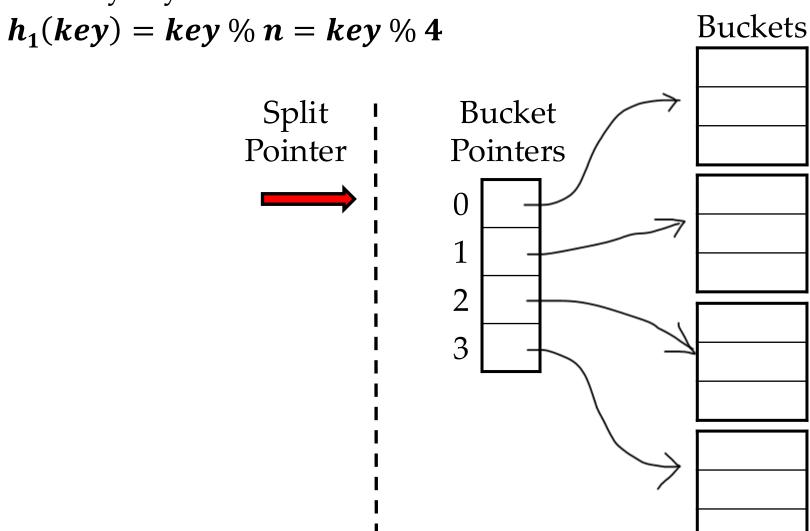
• Observe that all the buckets for bits 10 are full \rightarrow Need to split again buckets for 10.

• Now, **3-bits**.

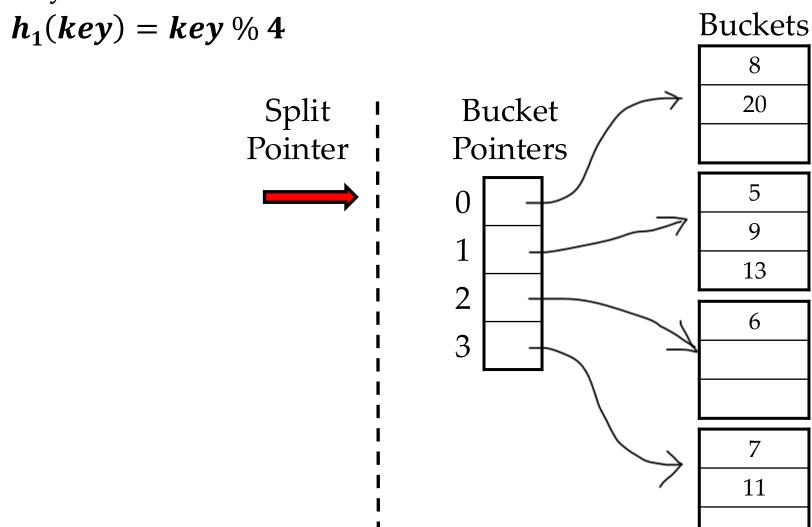


- Extensible hashing works well, but we perform the splitting **lazily** when the buckets for some bit(s) are full.
- What if we allow splitting to happen **eagerly** in the hope that in the future we would anyways need to split.
- Linear hashing performs eager and random splitting.
 - We call the splitting random because you may end up splitting empty buckets.
- Note: there is no longer tracking of buckets via binary representation.
- What we need is a **split pointer** that tells where did the last split took place.
 - Every **n-th split** introduces a new hash function.

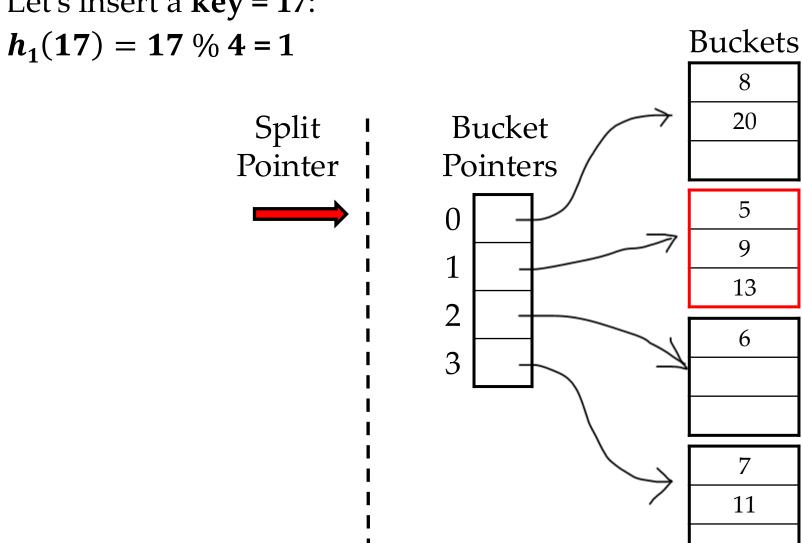
• Initially say our hash function is:



• Say our buckets look like this:

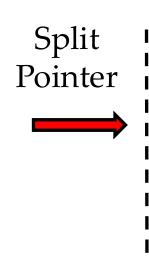


• Let's insert a **key = 17**:

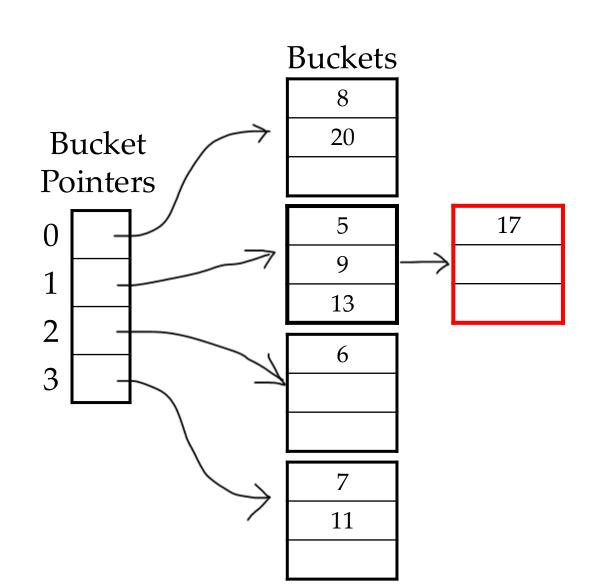


• Let's insert a **key = 17**:

$$h_1(17) = 17 \% 4 = 1$$

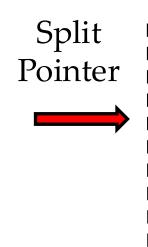


The bucket is full, so we create a new bucket and link.

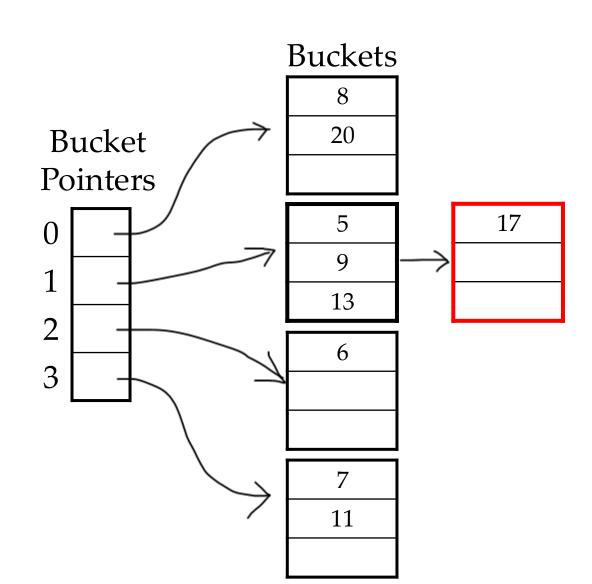


• Let's insert a **key = 17**:

$$h_1(17) = 17 \% 4 = 1$$



This situation has caused an overflow, so we need to split!

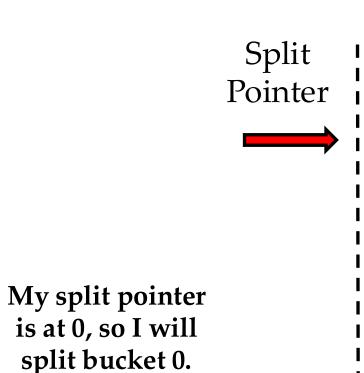


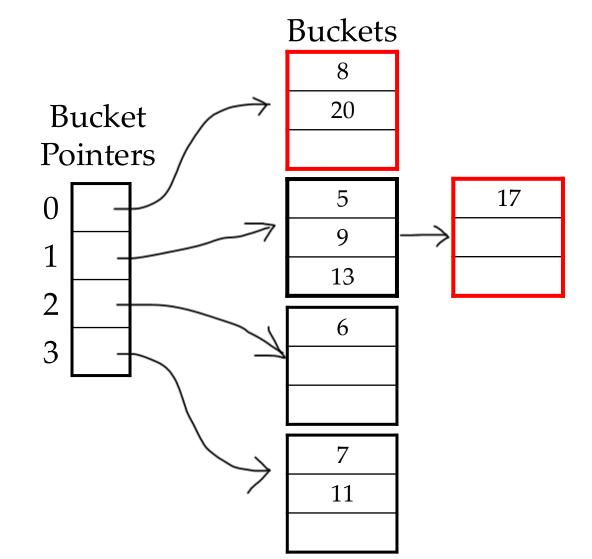
• Let's insert a **key = 17**:

$$h_1(17) = 17 \% 4 = 1$$

is at 0, so I will

split bucket 0.



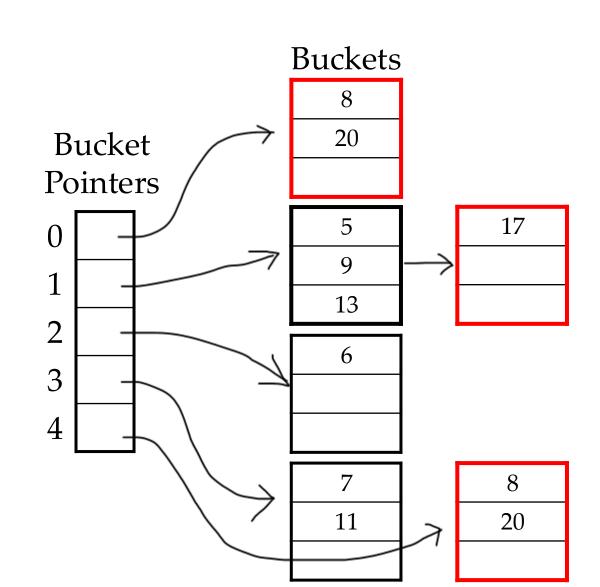


• Let's insert a **key = 17**:

$$h_1(17) = 17 \% 4 = 1$$



My split pointer is at 0, so I will split bucket 0, and add a new bucket pointer.



• Let's insert a **key = 17**:

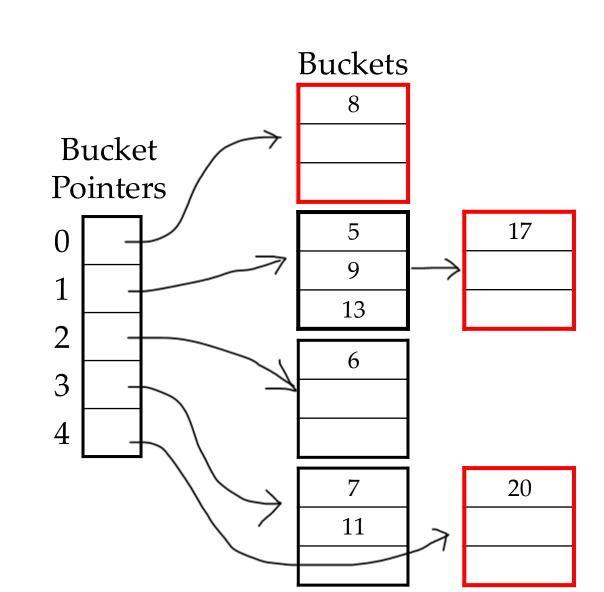
$$h_1(key) = key \% 4 = 1$$

 $h_2(key) = key \% 8 = 1$
Split
Pointer

Introduce a new hash function and Rehash the keys in original bucket 0.

$$8 \% 8 = 0$$

20 % $8 = 4$

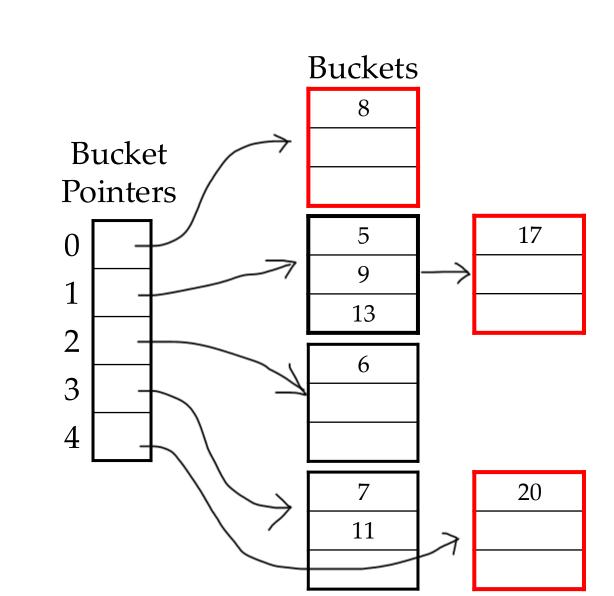


• Let's insert a **key = 17**:

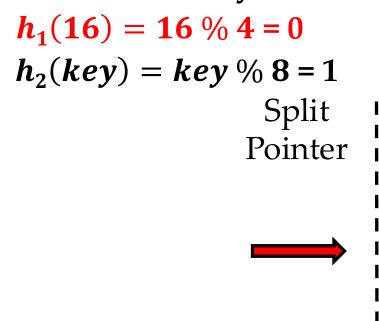
$$h_1(key) = key \% 4 = 1$$

 $h_2(key) = key \% 8 = 1$
Split
Pointer

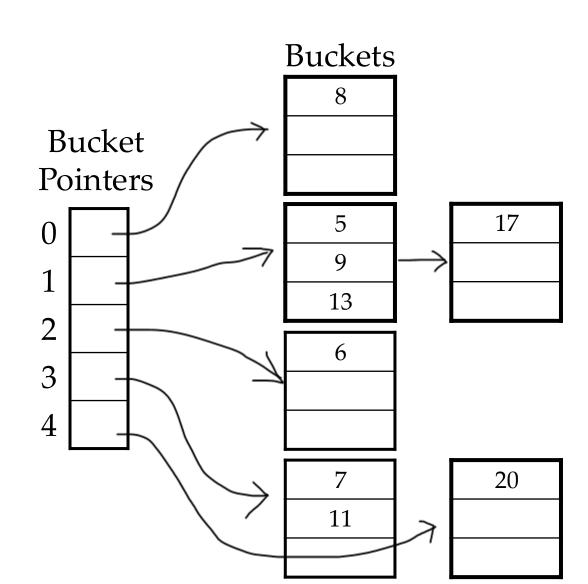
Move the split pointer



• Let's insert a **key = 16**:



First try the hash function h_1 (key).



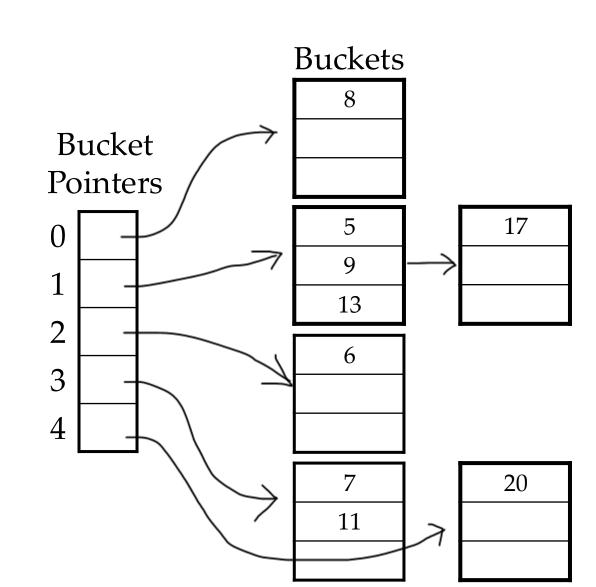
• Let's insert a **key = 16**:

$$h_1(16) = 16 \% 4 = 0$$
 $h_2(16) = 16 \% 8 = 0$
Split

Split Pointer



As 0 is above the split pointer, so we need to run the next hash function.



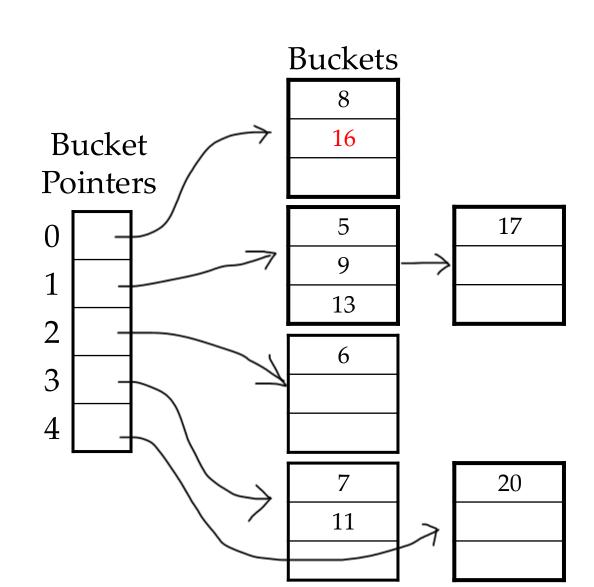
• Let's insert a **key = 16**:

$$h_1(16) = 16 \% 4 = 0$$
 $h_2(16) = 16 \% 8 = 0$
Split

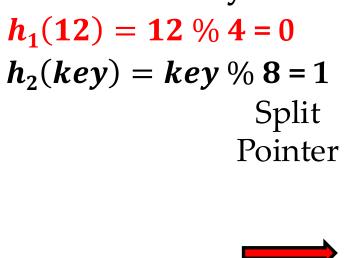
Split Pointer



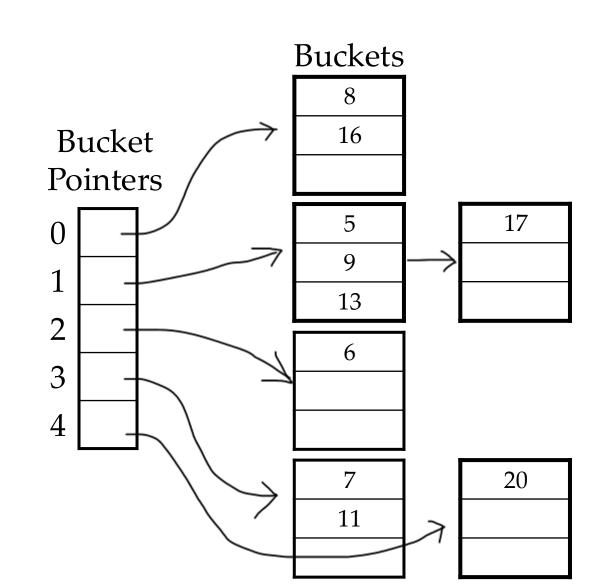
As 0 is above the split pointer, so we need to run the next hash function.



• Let's insert a **key = 12**:



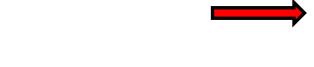
First try the hash function $h_1(key)$.



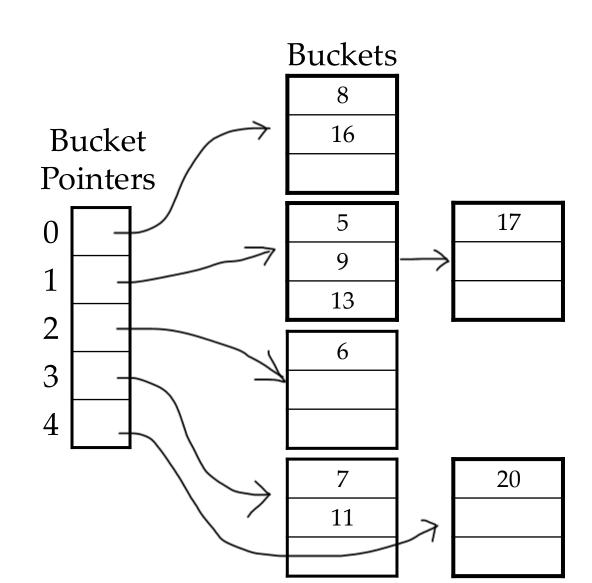
• Let's insert a **key = 12**:

$$h_1(12) = 12 \% 4 = 0$$
 $h_2(12) = 12 \% 8 = 4$

Split Pointer



As 0 is above the split pointer, so we need to run the next hash function.



• Let's insert a **key = 12**:

$$h_1(12) = 12 \% 4 = 0$$
 $h_2(12) = 12 \% 8 = 0$

Split Pointer



As 0 is above the split pointer, so we need to run the next hash function.

