

## Announcements

## Today

- Finish up B-trees
- Attempt to clear up some (justified) confusion
- B-Tree Deletes
- Hashing
- Arrays for dictionary
- Key space to array space
- Dealing with collisions
- Hash functions
- Resizing and Load Factors
- Expected performance

10/19/2022
CSE 332

## B Tree: Example

- B+ Tree with $\boldsymbol{M}=4$ (\# pointers in internal node)
- and $L=5$
(\# data items in leaf)



## Operations

- Find(K)
- Return a pointer to the record of $K$
- Insert(K)
- Insert key K and return a pointer to the record of K
- Delete(K)
- Delete key K and associated data


## Node sizes

- Internal nodes
- 4096 bytes, 8 byte keys, 8 byte child pointer $-\mathrm{M}=256$
- Leaves
- 4096 bytes, 8 byte keys, 8 byte record pointer $-L=256$




## Deletion Algorithm

1. Remove the key from its leaf

- 2. If the leaf ends up with fewer than ${ }_{L / 2}$ items, underflow!
- Adopt data from a neighbor; update the parent
- If adopting won't work, delete node and merge with neighbor
- If the parent ends up with fewer than ${ }_{M / 2}$ I children, underflow!

Hashing

## Deletion Slide Two

- 3. If an internal node ends up with fewer than $\left\lceil_{M / 2}\right\rceil$ children, underflow!
- Adopt from a neighbor; update the parent
- If adoption won't work,
merge with neighbor
- If the parent ends up with fewer than $[m / 2]$ children, underflow!

4. If the root ends up with only one child, make the child the new root of the tree
5. Propagate keys up through tree.

This reduces the height of the tree!

| Hashing |
| :---: |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |

Dictionary

## Array for data lookup

- Store football players by jersey number

| 10 | Uchenna Nwosu |
| :--- | :--- |
| 11 | Marquise Goodwin |
| 12 |  |
| 13 | Josh Jones |
| 14 | DK Metcalf |
| 15 |  |
| 16 | Tyler Lockett |
| 17 |  |
| 18 |  |
| 19 | Penny Hart |
| 20 |  |
| 21 | Artie Burns |
| 22 |  |
| 23 | Sidney Jones IV |
| 24 | Isaiah Dunn |

## Arrays for dictionaries

- Index by key, O(1) insert and find



## Hash Tables

- Map keys to a smaller array called a hash table
- via a hash function h(K)
- Find, insert, delete: $\mathrm{O}(1)$ on average!



## Array for data lookup

- Store students by student ID number

| 2061129 |  |
| :--- | :--- |
| 2061130 |  |
| 2061131 |  |
| 2061132 |  |
| 2061133 |  |
| 2061134 |  |
| 2061135 |  |
| 2061136 |  |
| 2061137 |  |
| 2061138 |  |
| 2061139 |  |
| 2061140 | Artie Burns |
| 2061141 |  |
| 2061142 |  |
| 2061143 |  |

Hashing: Map large keyspace into small index space

- I(K) = hash(K)

Key space

Simple Integer Hash Functions

- key space K = integers
- TableSize = 10
- $\mathrm{h}(\mathrm{K})=$
- Insert: 7, 18, 41, 34


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## Simple Integer Hash Functions

- key space $\mathrm{K}=$ integers
- TableSize = 7
- $\mathrm{h}(\mathrm{K})=\mathrm{K} \bmod 7$
- Insert: 7, 18, 41, 34



## Aside: Properties of Mod

To keep hashed values within the size of the table, we will generally do:
$h(K)=$ function $(K) \bmod$ TableSize
(In the previous examples, function $(\mathrm{K})=\mathrm{K}$.)

Useful properties of mod:
$(a+b) \bmod c=[(a \bmod c)+(b \bmod c)] \bmod c$ (ab) $\bmod c=[(a \bmod c)(b \bmod c)] \bmod c$ $a \bmod c=b \bmod c \rightarrow(a-b) \bmod c=0$

## Collision Resolutions

- Separate Chaining
- Open Addressing

|  | ate Cha | Insert: |
| :---: | :---: | :---: |
| 0 |  | 10 |
| 1 |  | 22 |
| 2 |  | 107 |
| 3 |  | 12 |
| 4 |  | 42 |
| 5 | All keys that map to the same hash value are kept in a list (or "bucket"). |  |
| 6 |  |  |
| 7 |  |  |
| 8 |  |  |
| 9 |  |  |
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## Analysis of Separate Chaining

The load factor, $\lambda$, of a hash table is $\lambda=\frac{\mathrm{N}}{\text { TableSize }}$ $\lambda=$ average \# of elements per bucket


Alternative: Use Empty Space in the Table Insert:

| 0 | 38 |
| :---: | :---: |
| 1 | 19 |
| 2 | 8 |
| 3 | 109 |
| 4 | 10 |
| 5 | Try h(K). |
| 6 | If full, try $h(K)+1$. |
| 7 | If full, try $h(K)+2$. |
| 8 | If full, try $\mathrm{h}(\mathrm{K})+3$. |
| 8 | Etc... |
| 9 |  |

## Open Addressing

After a collision, try "next" spot. If there's another collision, try another, etc.

Finding the next available spot is called probing:
$0^{\text {th }}$ probe $=h(k) \%$ TableSize
$1^{\text {th }}$ probe $=(h(k)+f(1)) \%$ TableSize
$2^{\text {th }}$ probe $=(h(k)+f(2)) \%$ TableSize
${ }^{\text {th }}$ probe $=(h(k)+f(i)) \%$ TableSize
$f(i)$ is the probing function. We'll look at a few...

## Linear Probing

$f(i)=i$

- Probe sequence:
$0^{\text {th }}$ probe $=h(K) \%$ TableSize
$1^{\text {th }}$ probe $=(h(K)+1) \%$ TableSize
$2^{\text {th }}$ probe $=(h(K)+2) \%$ TableSize
$i^{\text {th }}$ probe $=(\mathrm{h}(\mathrm{K})+\mathrm{i}) \%$ TableSize


## Linear Probing

|  |  |  | Insert: |
| :---: | :---: | :---: | :---: |
| 0 | 8 |  | 38 |
| 1 | 109 |  | 19 |
| 2 | 10 |  | 8 |
| 3 |  | Try h (K) | 109 |
| 4 |  | If full, try | 10 |
| 5 |  | If full, try |  |
| 6 |  | If full, tr |  |
| 7 |  | Etc... |  |
| 8 | 38 |  |  |
| 9 | 19 |  |  |

## Analysis of Linear Probing

- For any $\lambda<1$, linear probing will find an empty slot
- Expected \# of probes (for large table sizes) - unsuccessful search:

$$
\frac{1}{2}\left(1+\frac{1}{(1-\lambda)^{2}}\right)
$$

- successful search:

$$
\frac{1}{2}\left(1+\frac{1}{(1-\lambda)}\right)
$$

- Linear probing suffers from primary clustering
- Performance quickly degrades for $\lambda>1 / 2$




## Quadratic Probing Example



## Quadratic Probing

$$
f(i)=i^{2}
$$

Less likely to encounter Primary Clustering

- Probe sequence:

$$
\begin{aligned}
& 0^{\text {th }} \text { probe }=h(K) \% \text { TableSize } \\
& 1^{\text {th }} \text { probe }=(h(K)+1) \% \text { TableSize } \\
& 2^{\text {th }} \text { probe }=(h(K)+4) \% \text { TableSize } \\
& 3^{\text {th }} \text { probe }=(h(K)+9) \% \text { TableSize } \\
& \ldots \\
& i^{\text {th }} \text { probe }=\left(h(K)+i^{2}\right) \% \text { TableSize }
\end{aligned}
$$

## Another Quadratic Probing Example

| 0 | $\begin{aligned} & \text { TableSize = } 7 \\ & \mathrm{~h}(\mathrm{~K})=\mathrm{K} \% 7 \end{aligned}$ |  |
| :---: | :---: | :---: |
| 2 | insert(76) | $76 \% 7=6$ |
| 3 | insert(40) | $40 \% 7=5$ |
| 4 | insert(48) | $48 \% 7=6$ |
|  | insert(5) | $5 \% 7=5$ |
| 5 | insert(55) | $55 \% 7=6$ |
| 6 | insert(47) | $47 \% 7=5$ |

TableSize $=7$
$h(K)=K \% 7$
insert(76) $76 \% 7=6$
insert(40) 40 \% $7=5$
insert(48) $48 \% 7=6$
insert(5) $5 \% 7=5$
insert(47) $47 \% 7=5$

## Quadratic Probing: Properties

- For any $\lambda<1 / 2$, quadratic probing will find an empty slot; for bigger $\lambda$, quadratic probing may find a slot.
- Quadratic probing does not suffer from primary clustering: keys hashing to the same area is ok
- But what about keys that hash to the same slot?
- Secondary Clustering!


## Double Hashing

Idea: given two different (good) hash functions $\mathrm{h}(\mathrm{K})$ and $\mathrm{g}(\mathrm{K})$, it is unlikely for two keys to collide with both of them.

So...let's try probing with a second hash function:

$$
f(i)=i^{*} g(K)
$$

- Probe sequence:
$0^{\text {th }}$ probe $=h(K) \%$ TableSize
$1^{\text {th }}$ probe $=(\mathrm{h}(\mathrm{K})+\mathrm{g}(\mathrm{K})) \%$ TableSize
$2^{\text {th }}$ probe $=\left(h(K)+2^{\star} g(K)\right) \%$ TableSize
$3^{\text {th }}$ probe $=\left(h(K)+3^{*} g(K)\right) \%$ TableSize
$\mathrm{i}^{\text {th }}$ probe $=\left(\mathrm{h}(\mathrm{K})+\mathrm{i}^{*} g(\mathrm{~K})\right) \%$ TableSize


## Double Hashing Example



> TableSize $=7$
> $\mathrm{~h}(\mathrm{~K})=\mathrm{K} \% 7$
> $\mathrm{~g}(\mathrm{~K})=5-(\mathrm{K} \% 5)$

Insert(76) $76 \% 7=6$ and $5-76 \% 5=$ Insert(93) $93 \% 7=2$ and $5-93 \% 5=$ Insert(40) $40 \% 7=5$ and $5-40 \% 5=$ Insert(47) $47 \% 7=5$ and $5-47 \% 5=$ Insert(10) $10 \% 7=3$ and $5-10 \% 5=$ Insert $(55), 55 \%_{3} \%_{0} 7_{8}=6$ and $5-55 \% 5=$ 43

## Deletion in Open Addressing

$$
\mathrm{h}(\mathrm{k})=\mathrm{k} \% 7
$$


Linear probing

Delete(23)
Find(59)
Insert(30)

Need to keep track of deleted items... leave a "marker"
$\qquad$

## Rehashing Picture

- Starting with table of size 2, double when load factor > 1 .

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## Deletion in Separate Chaining

How do we delete an element with separate chaining?

## Rehashing

When the table gets too full, create a bigger table (usually $2 x$ as large) and hash all the items from the original table into the new table.

- When to rehash?
- Separate chaining: full $(\lambda=1)$
- Open addressing: half full $(\lambda=0.5)$
- When an insertion fails
- Some other threshold
- Cost of a single rehashing?


## Amortized Analysis of Rehashing

- Cost of inserting $n$ keys is $<3 n$
- suppose $2^{k}+1 \leq n \leq 2^{k+1}$
- Hashes = n
- Rehashes $=2+2^{2}+\ldots+2^{k}=2^{k+1}-2$
- Total $=n+2^{k+1}-2<3 n$
- Example
$-\mathrm{n}=33$, Total $=33+64-2=95<99$

