## CSE 332

JULY $10^{\text {TH }}$ - HASHING

## EXAM FRIDAY

- Practice exam after class today
- Topics:
- Stacks and Queues
- BigO Notation and runtime Analysis
- Heaps
- Trees (BST and AVL)
- Design Tradeoffs


## EXAM FRIDAY

- Format


## EXAM FRIDAY

- Format
- No note sheet


## EXAM FRIDAY

- Format
- No note sheet
- One section of short answer


## EXAM FRIDAY

- Format
- No note sheet
- One section of short answer
- 4-5 Technical Questions


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- No note sheet
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- 4-5 Technical Questions
- 1 Design Decision Question


## EXAM FRIDAY

- Format
- No note sheet
- One section of short answer
- 4-5 Technical Questions
- 1 Design Decision Question
- Less than 10 minutes per problem


## EXAM FRIDAY

- No Java material on the exam


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- Looking for theoretical understanding


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- Explanations are important (where indicated)


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- If you get stuck on a problem, move on


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- No Java material on the exam
- Looking for theoretical understanding
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- If you get stuck on a problem, move on
- Any questions?


## TODAY'S LECTURE

- Hashing


## TODAY'S LECTURE

- Hashing
- Basic Concept


## TODAY'S LECTURE

- Hashing
- Basic Concept
- Hash functions


## TODAY'S LECTURE

- Hashing
- Basic Concept
- Hash functions
- Collision Resolution


## TODAY'S LECTURE

- Hashing
- Basic Concept
- Hash functions
- Collision Resolution
- Runtimes


## HASHING

- Introduction


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- Suppose there is a set of data M


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- Introduction
- Suppose there is a set of data $\mathbf{M}$
- Any data we might want to store is a member of this set. For example, $\mathbf{M}$ might be the set of all strings
- There is a set of data that we actually care about storing D, where $\mathbf{D} \ll \mathbf{M}$
- For an English Dictionary, D might be the set of English words


## HASHING

- What is our ideal data structure?


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- The data structure should use $O(D)$ memory


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- The data structure should use O(D) memory
- No extra memory is allocated
- The operation should run in O(1) time
- Accesses should be as fast as possible


## HASHING

- What are some difficulties with this?


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- Need to know the size of $\mathbf{D}$ in advance or lose memory to pointer overhead


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- What are some difficulties with this?
- Need to know the size of $\mathbf{D}$ in advance or lose memory to pointer overhead
- Hard to go from $\mathbf{M}$-> $\mathbf{D}$ in $\mathbf{O}(1)$ time


## HASHING

- Memory: The Hash Table


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- Consider an array of size c * D


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- Memory: The Hash Table
- Consider an array of size c * D
- Each index in the array corresponds to some element in $\mathbf{M}$ that we want to store.
- The data in D does not need any particular ordering.


## THE HASH TABLE

- How can we do this?



## THE HASH TABLE

- How can we do this?
- Unsorted Array



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## THE HASH TABLE

- How can we do this?
- Unsorted Array


| Apple |
| :---: |
| Pear |
| Orange |
| Durian |
|  |
|  |
|  |
|  |
|  |
|  |
|  |

## THE HASH TABLE

- How can we do this?
- Unsorted Array



## THE HASH TABLE

- What is the problem here?


| Apple |
| :---: |
| Pear |
| Orange |
| Durian |
| Kumquat |
|  |
|  |
|  |
|  |
|  |

## THE HASH TABLE

- What is the problem here?
- Takes $O(D)$ time to find the word in the list!



## THE HASH TABLE

- What is the problem here?
- Takes $O(D)$ time to find the word in the list
- Same problem with sorted arrays!



## THE HASH TABLE

- What is another solution?



## THE HASH TABLE

- What is another solution?
- Random mapping



## THE HASH TABLE

- What's the problem here?



## THE HASH TABLE

- What's the problem here?
- Can't retrieve the random variable, $O(D)$ search!



## THE HASH TABLE

- What about a pseudo-random mapping?



## THE HASH TABLE

- What about a pseudo-random mapping?
- This is "the hash function"



## HASH FUNCTIONS

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## HASH FUNCTIONS

- The Hash Function maps the large space M to our target space D.
- We want our hash function to do the following:
- Be repeatable: $\mathrm{H}(\mathrm{x})=\mathrm{H}(\mathrm{x})$ every run
- Be equally distributed: For all $y, z$ in $D$, P(H(y)) = P(H(z))
- Run in constant time: $\mathrm{H}(\mathrm{x})=\mathrm{O}(1)$


## HASH EXAMPLE

- Let's consider an example. We want to save 10 numbers from all possible Java ints


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- What is a simple hash function?

|  |
| :---: |
|  |
|  |
|  |
|  |
|  |



| 0 |
| :--- |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |

## HASH EXAMPLE

- Let's consider an example. We want to save 10 numbers from all possible Java ints
- What is a simple hash function?
ints

| 0 |
| :--- |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |

## HASH EXAMPLE

- Let's insert(519) table


| 0 |
| :--- |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |

## HASH EXAMPLE

- Let's insert(519) table
- Where does it go?


| 0 |
| :--- |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |

## HASH EXAMPLE

- Let's insert(519) table
- Where does it go?
- $519 \% 10=$


| 0 |
| :--- |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |

## HASH EXAMPLE

- Let's insert(519) table
- Where does it go?
- $519 \% 10=9$



## HASH EXAMPLE

- Insert(214)



## HASH EXAMPLE

- Insert(214)



## HASH EXAMPLE

- insert(1001)



## HASH EXAMPLE

- insert(1001)



## HASH EXAMPLE

- Is there a problem here?



## HASH EXAMPLE

- Is there a problem here?
- insert(3744)



## HASH EXAMPLE

- Is there a problem here?
- insert(3744)



## HASH EXAMPLE

- Is there a problem here?
- insert(3744)
- This is called a collision!



## HASH FUNCTION

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- Needs to incorporate all the data in the keys


## HASH FUNCTION

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- You will not have to produce hash functions, but you should recognize good ones
- They run in constant time
- They evenly distribute the data
- They return an integer
- These hash functions are chosen in advance, you should not pick a hash function relative to your data


## HASH EXAMPLE

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- How to rectify collisions?
- Think of a strategy for a few minutes
- Possible solutions:
- Store in the next available space
- Store both in the same space
- Try a different hash
- Resize the array


## HASH EXAMPLE

- Consider the simplest solution


## HASH EXAMPLE

- Consider the simplest solution
- Find the next available spot in the array


## LINEAR PROBING

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- This is called clustering


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- If the cluster becomes too large, two things happen:
- The chances of colliding with the cluster increase
- The time it takes to find something in the cluster increases. This isn't O(1) time!


## CLUSTERING

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- Store multiple items in one location
- This is called chaining
- We'll discuss it later


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## COLLISIONS

- Probing
- Linear probing
- Try the appropriate hash table row first
- Increase the index by one until a spot is found
- Guaranteed to find a spot if it is available
- If the array is too full, its operations reach O(n) time


## COLLISIONS

- Probing
- Quadratic Probing


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## COLLISIONS

- Probing
- Quadratic Probing
- Rather than increasing by one each time, we increase by the squares
- k+1, k+4, k+9, k+16, k+25
- Certain tables can cause secondary clustering
- Can fail to insert if the table is over half full


## COLLISIONS

- Probing
- Secondary Hashing


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- Secondary Hashing
- If two keys collide in the hash table, then a secondary hash indicates the probing size


## COLLISIONS

- Probing
- Secondary Hashing
- If two keys collide in the hash table, then a secondary hash indicates the probing size
- Need to be careful, possible for infinite loops with a very empty array


## COLLISIONS

- Chaining


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- Chaining
- Rather than probing for an open position, we could just save multiple objects in the same position
- Some data structure is necessary here
- Commonly a linked list, AVL tree or secondary hash table.
- Resizing isn't necessary, but if you don't, you will get $\mathrm{O}(\mathrm{n})$ runtime.


## PRIMALITY

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## PRIMALITY

- Array sizes
- We normally choose our hash tables to have prime size
- This is because for any number we pick, so long as it is not a multiple of our table size, they must be coprime
- Two numbers $x$ and $y$ are coprime if they do not share any common factors.
- If the hash table size and the secondary hash value are coprime, then the search will succeed if there is space available


## PRIMALITY

- Array sizes
- We normally choose our hash tables to have prime size
- This is because for any number we pick, so long as it is not a multiple of our table size, they must be coprime
- Two numbers $x$ and $y$ are coprime if they do not share any common factors.
- If the hash table size and the secondary hash value are coprime, then the search will succeed if there is space available
- However, many primes cause secondary clustering when used with quadratic probing


## LOAD FACTOR

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## LOAD FACTOR

- When discussing hash table efficiency, we call the proportion of stored data to table size the load factor. It is represented by the Greek character lambda ( $\lambda$ ).
- We've discussed this a bit implicitly before
- What are good load-factor ( $\lambda$ ) values for each of our collision techniques?


## LOAD FACTOR

- Linear Probing?
- Quadratic Probing?
- Secondary Hashing?
- Chaining?


## LOAD FACTOR

- Linear Probing?
- Quadratic Probing?
- Secondary Hashing?
- Chaining?
- What are the tradeoffs?


## LOAD FACTOR

- Linear Probing?
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- What are the tradeoffs?
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- Linear Probing?
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## LOAD FACTOR

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- Quadratic Probing?
- Secondary Hashing?
- Chaining?
-What are the tradeoffs?
- Memory efficiency
- Failure rate
- Access times?


## LOAD FACTOR

- Linear Probing? $0.25<\lambda<0.5$
- Quadratic Probing?
- Secondary Hashing?
- Chaining?


## LOAD FACTOR

- Linear Probing? $0.25<\lambda<0.5$
- Quadratic Probing? $0.10<\lambda<0.30$
- Secondary Hashing?
- Chaining?


## LOAD FACTOR

- Linear Probing? $0.25<\lambda<0.5$
- Quadratic Probing? $0.10<\lambda<0.30$
- If it gets to 0.5 , then there is a chance of failure, and a high chance of $\mathrm{O}(\mathrm{n})$ runtime
- Secondary Hashing?
- Chaining?


## LOAD FACTOR

- Linear Probing? $0.25<\lambda<0.5$
- Quadratic Probing? $0.10<\lambda<0.30$
- Secondary Hashing? $0.25<\lambda<0.5$
- Chaining?


## LOAD FACTOR

- Linear Probing? $0.25<\lambda<0.5$
- Quadratic Probing? $0.10<\lambda<0.30$
- Secondary Hashing? $0.25<\lambda<0.5$
- But we've eliminated primary clustering
- Chaining?


## LOAD FACTOR

- Linear Probing? $0.25<\lambda<0.5$
- Quadratic Probing? $0.10<\lambda<0.30$
- Secondary Hashing? $0.25<\lambda<0.5$
- Chaining? $3.0<\lambda<10$


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- Because we allow multiple items in each space, we can increase memory efficiency by taking advantage


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- Quadratic Probing? $0.10<\lambda<0.30$
- Secondary Hashing? $0.25<\lambda<0.5$
- Chaining? $3.0<\lambda<10$
- Because we allow multiple items in each space, we can increase memory efficiency by taking advantage
- As long as there are a constant number in each space, we get $O(1)$ runtimes.


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- Here, these resizes are often for performance, rather than failure.
- Hash table maintenance is important
- Resizing is costly (but still $\mathrm{O}(\mathrm{n})$ ) because you have to resize the array and rehash every element into the new table.


## HASH TABLES

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- Can provide $O(1)$ access times
- Can be memory inefficient
- Probing can fail, and delete with probing mechanisms is difficult
- Chaining can be a good balance, but there is a lot of overhead maintaining all those data structures


## HASH TABLES

- Understand these tradeoffs and how these implementations work

HASH TABLES

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- Section tomorrow will provide practice problems for each of these hash table methods


## HASH TABLES

- Take-aways for the midterm


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- The hash function should run in constant time and should distribute among the indices in the target array


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- The array should be relative to the size of the data you want to keep
- The hash function should run in constant time and should distribute among the indices in the target array
- Linear probing is a solution for collisions, but only works when there is lots of free space
- Resizing is very costly


## NEXT CLASS

- Hash Tables
- Examples, examples, examples
- Finish discussion
- Exam review

