

CSE 373

OCTOBER 18TH – HASHING

ADMINISTRIVIA

- **Written homework due individually tonight**
 - Broken into 5 problems
 - Please resubmit if you have already, this will make grading easier for us

ADMINISTRIVIA

- **Project 2 is out tonight**
 - This is a one week project (no checkpoint, so no opportunity for regrading)
 - Start early!
 - Implementing Hashtables and Hashsets
 - Canvas should be configured to allow you to make your own groups, hopefully this will make grade assignments easier

ADMINISTRIVIA

- **Project 1 EC and Part 1 regrades out Friday**
 - If you got a different grade than your partner, let me know
 - EC is calculated separately from the rest of the course

TODAY'S LECTURE

- **Hashtables**
 - Review of probing methods
 - Separate Chaining
 - Implementation considerations

HASHING

- **Introduction**

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

HASHING

- **What is our ideal data structure?**
 - 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

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

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: $H(x) = H(x)$ every run
 - Be equally distributed: For all y, z in D ,
 $P(H(y)) = P(H(z))$
 - Run in constant time: $H(x) = O(1)$

HASH FUNCTION

- **In reality, good hash functions are difficult to produce**
 - We want a hash that distributes our data evenly throughout the space
 - Usually, our hash function returns some integer, which must then be modded to our table size
 - Needs to incorporate all the data in the keys

HASH EXAMPLE

- **Possible solutions:**
 - Store in the next available space
 - Store both in the same space
 - Try a different hash
 - Resize the array

COLLISIONS

- **Hash table methods are defined by how they handle collisions**
- **Two main approaches**
 - Probing
 - Chaining

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- 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. **Primary clustering**

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 - Certain tables can cause **secondary clustering**

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 - 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

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 - Need to be careful, possible for infinite loops with a very empty array
 - If the secondary hash value and the table size are coprime (they share no factors), then secondary hashing will succeed if there is an open space

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 - Need to be careful, possible for infinite loops with a very empty array
 - If the secondary hash value and the table size are coprime (they share no factors), then secondary hashing will succeed if there is an open space
 - If table size is prime, only need to check if hash is a multiple

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- If the hash table size and the secondary hash value are coprime, then the search will succeed if there is space available

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- 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

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- 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 $O(n)$ runtime.

LOAD FACTOR

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- **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 (λ).**
 - We've discussed this a bit implicitly before
 - What are good load-factor (λ) values for each of our collision techniques?

LOAD FACTOR

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

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 - Access times?

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- **Linear Probing?** $0.25 < \lambda < 0.5$
- **Quadratic Probing?**
- **Secondary Hashing?**
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- **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 $O(n)$ runtime
- **Secondary Hashing?**
- **Chaining?**

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

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- **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 $O(n)$) because you have to resize the array and rehash every element into the new table.

DELETION

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 - Chaining: just remove the object from the underlying data structure
 - Probing: Must be able to follow the path in order to find elements that have been added later
 - Need to mark as deleted, but not treat as completely empty

LAZY DELETION

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 - When you delete, mark the element as deleted, but maintain the data structure as-is
 - Works well for AVL as well
 - Can insert values into place if reinserted, just cannot return the associated value on a call to find
 - Necessary for Probing (aka Open Addressing) collision methods

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 - Many implement with a simple linked list
 - If the load factor is λ , what is the expected number of elements in a single bin? λ
 - However, the expected **maximum** actually grows (roughly) logarithmically with table length
 - The more elements we add, the higher chance that there is one bad bin

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 - If we still want $1/N$ collision probability, how large is the table? N^2 but N is almost always a constant
 - Some constant number have $\log n$ memory, but this is $O(n)$ memory usage overall!