CSE 373: Open addressing

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Warmup:

With your neighbor, discuss and review:

- ► How do we implement **get**, **put**, and **remove** in a hash table using separate chaining?
- What about in a hash table using open addressing with linear probing?
- ► Compare and contrast your answers: what do we do the same? What do we do differently?

In both implementations, for all three methods, we start by **finding** the initial index to consider:

```
index = key.hashCode() % array.length
```

If we're using separate chaining, we then search/insert/delete from the bucket:

```
IDictionary<K, V> bucket = array[index]
bucket.get(key) // or .put(...) or .remove(...)
```

...and resize when $\lambda \approx 1$.

(When exactly to resize is a tuneable parameter)

If we're using linear probing, search until we find an array element where the key is equal to ours or until the array index is null:

```
while (array[index] != null
          && array[index].hashcode != key.hashCode()
          && !array[index].equals(key)) {
    index = (index + 1) % this.array.length
}
if (array[index] == null)
          // throw exception if implementing get
          // add new key-value pair if implementing put
else
          // return or set array[index]
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How do we delete? (complicated, see section 04 handouts)

When do we resize?

Strategy: Linear probing

If we collide, checking each next element until we find an open slot.

```
So, h'(k, i) = (h(k) + i) \mod T, where T is the table size
```

```
i = 0
while (index in use)
    try (hash(key) + i) % array.length
    i += 1
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Assume internal capacity of 10, insert the following keys:

38, 19, 8, 109, 10

 0	1	2	3	4	5	6	7	8	9

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What's the problem? Lots of keys close together: a "cluster". We ended up having to probe many slots!

Primary clustering

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Happens when λ is large, or if we get unlucky

In linear probing, we expect to get $\mathcal{O}(\lg(n))$ size clusters.

Questions:

► When is performance good? When is it bad?

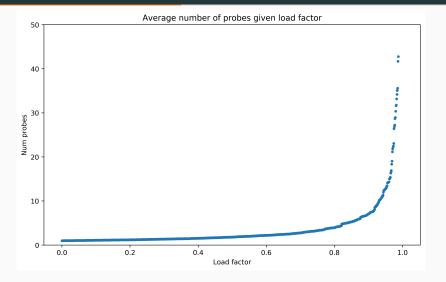
► What is the maximum load factor?

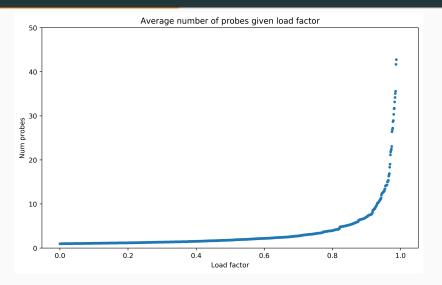
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- ▶ When is performance good? When is it bad? Runtime is bad when table is nearly full. Runtime is also bad when we hit a "cluster"
- ► What is the maximum load factor? Load factor is at most $\lambda = 1.0!$
- ▶ When do we resize?





Punchline: clustering can be potentially bad, but in practice, it tends to be ok as long as λ is small

Question: when do we resize?

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Nifty equations:

► Average number of probes for successful probe:

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

► Average number of probes for unsuccessful probe:

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

^{*}These equations aren't important to know

Problem: We can still get unlucky/somebody can feed us a malicious series of inputs that causes several slowdown

Can we pick a different collision strategy that minimizes clustering?

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Exercise: assume internal capacity of 10, insert the following:

89, 18, 49, 58, 79

0 1 2 3 4 5 6 7 8 9

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Strategy: Quadratic probing

```
If we collide: h'(k,i) = (h(k) + i^2) \mod T, where T is table size i = 0 while (index in use) try (hash(key) + i * i) % array.length i += 1
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What problems are there?

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Problem 2: Still can get clusters (though not as badly)

Secondary clustering

When using quadratic probing, we sometimes need to probe a sequence of table cells (that are not necessary next to each other). This problem is known as "secondary clustering".

Ex: inserting 19, 39, 29, 9:

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Ex: inserting 19, 39, 29, 9:

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Secondary clustering can also be bad, but is generally milder then primary clustering

Note: let
$$s = h(k)$$

► Linear probing:

$$s + 0$$
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► Linear probing:

$$s+0$$
, $s+1$, $s+2$, $s+3$, $s+4$, ...
Basic pattern: try $h'(k,i)=(h(k)+i) \bmod T$

▶ Quadratic probing: s + 0, s + 1, $s + 2^2$, $s + 3^2$, $s + 4^2$, ... Basic pattern: try $h'(k, i) = (h(k) + i^2) \mod T$

Observation: For both probing strategies, there are just $\mathcal{O}\left(T\right)$ different "probe sequences" – distinct ways we can probe the array.

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Observation: For both probing strategies, there are just $\mathcal{O}\left(T\right)$ different "probe sequences" – distinct ways we can probe the array.

Idea: Can we increase the number of distinct probe sequences to decrease odds of collision?

Open addressing: double-hashing

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 $s + 0j$, $s + 1j$, $s + 2j$, $s + 3j$, $s + 4j$, ...

Basic pattern: try $h'(k, i) = (h(k) + i \cdot g(k)) \mod T$

In pseudocode:

```
i = 0
while (index in use)
    try (hash(key) + i * jump_hash(key)) % array.length
    i += 1
```

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Ways we can do this:

- ▶ If T is a power of two, make g(k) return any odd integer
- ▶ If T is a prime, make g(k) return any smaller, non-zero integer (e.g. $g(k) = 1 + (k \mod (T 1))$)

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Result: in practice, double-hashing is very effective and commonly used "in the wild".

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No obvious answer: both implementations are common.

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No obvious answer: both implementations are common.

Separate chaining:

- ▶ Don't have to worry about clustering
- ▶ Potentially more "compact" (λ can be higher)

Open addressing:

- ► Managing clustering can be tricky
- ▶ Less compact (we typically keep $\lambda < \frac{1}{2}$)
- Array lookups tend to be a constant factor faster then traversing pointers

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Important: Depending on the application, we might want our hash function to have different properties.

How would you implement the following using hash functions?

For each application, also discuss what properties you want your hash function to have.

- ► Suppose we're sending a message over the internet. This message might become mildly corrupted. How can we detect if corruption probably occurred?
- ► Suppose you have many fragments of DNA and want to see where they appears in a (significantly longer) segment of DNA. How can we do this efficiently?

Same question as before:

- ► Suppose you're designing an video uploading site and want to detect if somebody is uploading a pirated movie. A naive way to do this is to check if the movie is byte-for-byte identical to some movie. How can we do this more efficiently?
- ▶ Suppose you're designing a website with a user login system. Directly storing your user's passwords is dangerous what if they get stolen? How can you store password in a safe way so that even if they're stolen, the passwords aren't compromised?

Same question as before:

- ➤ You are trying to build an image sharing site. Users upload many images, and you need to assign each image some unique ID. How might you do this?
- ▶ Suppose we have a long series of financial transactions stored on some (potentially untrustworthy) computer. Somebody claims they made a specific transaction several months ago. Can you design a system that lets you audit and determine if they're lying or not? Assume you have access to just the very latest transaction, obtained from a different trustworthy source.