CSE373: Data Structures and Algorithms

Hashing II (Collisions)

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Winter 2016

This lecture material represents the work of multiple instructors at the University of Washington.
Thank you to all who have contributed!

Hash Tables: Review

- Aim for constant-time (i.e., $O(1)$) **find**, **insert**, and **delete**
  - "On average" under some reasonable assumptions

- A hash table is an array of some fixed size
  - But growable as we'll see

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Collision resolution

Collision:

    When two keys map to the same location in the hash table

We try to avoid it, but number-of-keys exceeds table size

So hash tables should support collision resolution
    – Ideas?

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Separate Chaining

Chaining:
    All keys that map to the same table location are kept in a list (a.k.a. a "chain" or "bucket")

As easy as it sounds

Example:
    insert 10, 22, 107, 12, 42
    with mod hashing
    and **TableSize** = 10

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Separate Chaining

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Separate Chaining

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    All keys that map to the same table location are kept in a list (a.k.a. a "chain" or "bucket")

As easy as it sounds

Example:
    insert 10, 22, 107, 12, 42
    with mod hashing
    and **TableSize** = 10

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Separate Chaining

```
0  ┌───┐→10 /
1  │ / │
2  ├───┤→22 /
3  │ / │
4  │ / │
5  │ / │
6  │ / │
7  ├───┤→107 /
8  │ / │
9  │ / │
   └───┘
```

Chaining:
All keys that map to the same table location are kept in a list (a.k.a. a "chain" or "bucket")

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Example:
insert 10, 22, 107, 12, 42
with mod hashing
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Separate Chaining

```
0  ┌───┐→10 /
1  │ / │
2  ├───┤→12 →22 /
3  │ / │
4  │ / │
5  │ / │
6  │ / │
7  ├───┤→107 /
8  │ / │
9  │ / │
   └───┘
```

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Separate Chaining

```
0  ┌───┐→10 /
1  │ / │
2  ├───┤→42 →12 →22 /
3  │ / │
4  │ / │
5  │ / │
6  │ / │
7  ├───┤→107 /
8  │ / │
9  │ / │
   └───┘
```

Chaining:
All keys that map to the same table location are kept in a list (a.k.a. a "chain" or "bucket")

As easy as it sounds

Example:
insert 10, 22, 107, 12, 42
with mod hashing
and **TableSize** = 10

Thoughts on chaining

- Worst-case time for **find**?
  - Linear
  - But only with really bad luck or bad hash function
  - So not worth avoiding (e.g., with balanced trees at each bucket)

- Beyond asymptotic complexity, some "data-structure engineering" may be warranted
  - Linked list vs. array vs. chunked list (lists should be short!)
  - Move-to-front
  - Maybe leave room for 1 element (or 2?) in the table itself, to optimize constant factors for the common case
    - A time-space trade-off…

Time vs. space (constant factors only here)

```
0  ┌───┐→10 /
1  │ / │
2  ├───┤→42 →12 →22 /
3  │ / │
4  │ / │
5  │ / │
6  │ / │
7  ├───┤→107 /
8  │ / │
9  │ / │
   └───┘
```

0	10	/
1	/	X
2	42	→12 →22 /
3	/	X
4	/	X
5	/	X
6	/	X
7	107	/
8	/	X
9	/	X

More rigorous chaining analysis

Definition: The load factor, λ, of a hash table is

$$\lambda = \frac{N}{TableSize} \qquad \leftarrow \textbf{ number of elements}$$

Under chaining, the average number of elements per bucket is ___

More rigorous chaining analysis

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Under chaining, the average number of elements per bucket is λ

So if some inserts are followed by *random* finds, then on average:
- Each unsuccessful **find** compares against _____ items

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More rigorous chaining analysis

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$$\lambda = \frac{N}{\text{TableSize}} \quad \leftarrow \textbf{number of elements}$$

Under chaining, the average number of elements per bucket is λ

So if some inserts are followed by *random* finds, then on average:
- Each unsuccessful **find** compares against λ items
- Each successful **find** compares against _____ items

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More rigorous chaining analysis

Definition: The load factor, λ, of a hash table is

$$\lambda = \frac{N}{\text{TableSize}} \quad \leftarrow \textbf{number of elements}$$

Under chaining, the average number of elements per bucket is λ

So if some inserts are followed by *random* finds, then on average:
- Each unsuccessful **find** compares against λ items
- Each successful **find** compares against $\lambda/2$ items

So we like to keep λ fairly low (e.g., 1 or 1.5 or 2) for chaining

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Alternative: Use empty space in the table

- Another simple idea: If **h(key)** is already full,
  - try **(h(key) + 1) % TableSize**. If full,
  - try **(h(key) + 2) % TableSize**. If full,
  - try **(h(key) + 3) % TableSize**. If full…
- Example: insert 38, 19, 8, 109, 10

0	/
1	/
2	/
3	/
4	/
5	/
6	/
7	/
8	38
9	/

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- Example: insert 38, 19, 8, 109, 10

0	/
1	/
2	/
3	/
4	/
5	/
6	/
7	/
8	38
9	19

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  - try **(h(key) + 1) % TableSize**. If full,
  - try **(h(key) + 2) % TableSize**. If full,
  - try **(h(key) + 3) % TableSize**. If full…
- Example: insert 38, 19, 8, 109, 10

0	8
1	/
2	/
3	/
4	/
5	/
6	/
7	/
8	38
9	19

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Alternative: Use empty space in the table

- Another simple idea: If `h(key)` is already full,
  - try `(h(key) + 1) % TableSize`. If full,
  - try `(h(key) + 2) % TableSize`. If full,
  - try `(h(key) + 3) % TableSize`. If full…

- Example: insert 38, 19, 8, 109, 10

0	8
1	109
2	/
3	/
4	/
5	/
6	/
7	/
8	38
9	19

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  - try `(h(key) + 2) % TableSize`. If full,
  - try `(h(key) + 3) % TableSize`. If full…

- Example: insert 38, 19, 8, 109, 10

0	8
1	109
2	10
3	/
4	/
5	/
6	/
7	/
8	38
9	19

Probing hash tables

Trying the next spot is called probing (also called open addressing)
- We just did linear probing
  - i^{th} probe was `(h(key) + i) % TableSize`
- In general have some probe function `f` and use
  `h(key) + f(i) % TableSize`

Open addressing does poorly with high load factor λ
- So want larger tables
- Too many probes means no more $O(1)$

Other operations

`insert` finds an open table position using a probe function

What about `find`?
- Must use same probe function to "retrace the trail" for the data
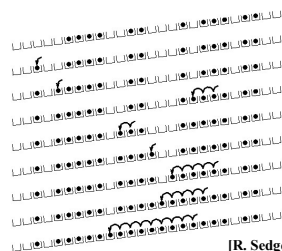- Unsuccessful search when reach empty position

What about `delete`?
- **Must** use "lazy" deletion. Why?
  - Marker indicates "no data here, but don't stop probing"
- Note: `delete` with chaining is plain-old list-remove

(Primary) Clustering

It turns out linear probing is a *bad idea*, even though the probe function is quick to compute (which is a good thing)

Tends to produce *clusters*, which lead to long probing sequences
- Called primary clustering
- Saw this starting in our example

[R. Sedgewick]

Analysis of Linear Probing

- Trivial fact: For any $\lambda < 1$, linear probing will find an empty slot
  - It is "safe" in this sense: no infinite loop unless table is full

- Non-trivial facts we won't prove:
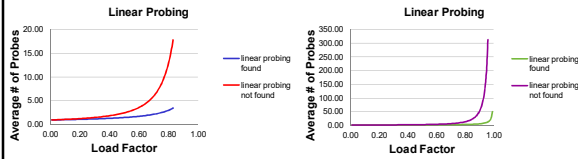 Average # of probes given λ (in the limit as `TableSize` $\rightarrow \infty$)
  - Unsuccessful search:
  $$\frac{1}{2}\left(1 + \frac{1}{(1-\lambda)^2}\right)$$

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

- This is pretty bad: need to leave sufficient empty space in the table to get decent performance (see chart)

In a chart

- Linear-probing performance degrades rapidly as table gets full
  - (Formula assumes "large table" but point remains)

- By comparison, chaining performance is linear in λ and has no trouble with $\lambda > 1$

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Quadratic probing

- We can avoid primary clustering by changing the probe function
  `(h(key) + f(i)) % TableSize`

- A common technique is quadratic probing:
  $f(i) = i^2$
  - So probe sequence is:
    - 0th probe: `h(key) % TableSize`
    - 1st probe: `(h(key) + 1) % TableSize`
    - 2nd probe: `(h(key) + 4) % TableSize`
    - 3rd probe: `(h(key) + 9) % TableSize`
    - …
    - i^{th} probe: `(h(key) + i²) % TableSize`

- Intuition: Probes quickly "leave the neighborhood"

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Quadratic Probing Example

0	
1	
2	
3	
4	
5	
6	
7	
8	
9	

TableSize=10
Insert:
89
18
49
58
79

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Quadratic Probing Example

0	
1	
2	
3	
4	
5	
6	
7	
8	
9	89

TableSize=10
Insert:
89
18
49
58
79

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Quadratic Probing Example

0	
1	
2	
3	
4	
5	
6	
7	
8	18
9	89

TableSize=10
Insert:
89
18
49
58
79

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Quadratic Probing Example

0	49
1	
2	
3	
4	
5	
6	
7	
8	18
9	89

TableSize=10
Insert:
89
18
49
58
79

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Quadratic Probing Example

0	49
1	
2	58
3	
4	
5	
6	
7	
8	18
9	89

TableSize=10
Insert:
89
18
49
58
79

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Quadratic Probing Example

0	49
1	
2	58
3	79
4	
5	
6	
7	
8	18
9	89

TableSize=10
Insert:
89
18
49
58
79

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Another Quadratic Probing Example

0	
1	
2	
3	
4	
5	
6	

TableSize = 7

Insert:
76   (76 % 7 = 6)
40   (40 % 7 = 5)
48   (48 % 7 = 6)
5   (5 % 7 = 5)
55   (55 % 7 = 6)
47   (47 % 7 = 5)

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Another Quadratic Probing Example

0	
1	
2	
3	
4	
5	
6	76

TableSize = 7

Insert:
76   (76 % 7 = 6)
40   (40 % 7 = 5)
48   (48 % 7 = 6)
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Another Quadratic Probing Example

0	
1	
2	
3	
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5	40
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TableSize = 7

Insert:
76   (76 % 7 = 6)
40   (40 % 7 = 5)
48   (48 % 7 = 6)
5   (5 % 7 = 5)
55   (55 % 7 = 6)
47   (47 % 7 = 5)

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Another Quadratic Probing Example

0	48
1	
2	
3	
4	
5	40
6	76

TableSize = 7

Insert:
76   (76 % 7 = 6)
40   (40 % 7 = 5)
48   (48 % 7 = 6)
5   (5 % 7 = 5)
55   (55 % 7 = 6)
47   (47 % 7 = 5)

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Another Quadratic Probing Example

TableSize = 7

0	48
1	
2	5
3	
4	
5	40
6	76

Insert:

76	(76 % 7 = 6)
40	(40 % 7 = 5)
48	(48 % 7 = 6)
5	(5 % 7 = 5)
55	(55 % 7 = 6)
47	(47 % 7 = 5)

Another Quadratic Probing Example

TableSize = 7

0	48
1	
2	5
3	55
4	
5	40
6	76

Insert:

76	(76 % 7 = 6)
40	(40 % 7 = 5)
48	(48 % 7 = 6)
5	(5 % 7 = 5)
55	(55 % 7 = 6)
47	(47 % 7 = 5)

Another Quadratic Probing Example

TableSize = 7

0	48
1	
2	5
3	55
4	
5	40
6	76

Insert:

76	(76 % 7 = 6)
40	(40 % 7 = 5)
48	(48 % 7 = 6)
5	(5 % 7 = 5)
55	(55 % 7 = 6)
47	(47 % 7 = 5)

Doh!: For all *n*, $((n*n) +5) \% 7$ is 0, 2, 5, or 6
- Excel shows takes "at least" 50 probes and a pattern
- Proof (like induction) using $(n^2+5) \% 7 = ((n-7)^2+5) \% 7$
  - In fact, for all *c* and *k*, $(n^2+c) \% k = ((n-k)^2+c) \% k$

From Bad News to Good News

- Bad news:
  - Quadratic probing can cycle through the same full indices, never terminating despite table not being full

- Good news:
  - If **TableSize** is *prime* and $\lambda < \frac{1}{2}$, then quadratic probing will find an empty slot in at most **TableSize/2** probes
  - So: If you keep $\lambda < \frac{1}{2}$ and **TableSize** is *prime*, no need to detect cycles

  - Optional: Proof is available at
  - `http://courses.cs.washington.edu/courses/cse373/14au/uwnetid/quadraticProbingProof.txt`
    - Key fact: For prime **T** and $0 < i,j < T/2$ where $i \neq j$,
      $(k + i^2) \% T \neq (k + j^2) \% T$ (i.e., no index repeat)

Clustering reconsidered

- Quadratic probing does not suffer from primary clustering: no problem with keys initially hashing to the same neighborhood

- But it's no help if keys initially hash to the same index
  - Called secondary clustering

- Can avoid secondary clustering with a probe function that depends on the key: double hashing…

Double hashing

Idea:
- Given two good hash functions *h* and *g*, it is very unlikely that for some *key*, `h(key) == g(key)`
- So make the probe function `f(i) = i*g(key)`

Probe sequence:
- 0th probe: `h(key) % TableSize`
- 1st probe: `(h(key) + g(key))  % TableSize`
- 2nd probe: `(h(key) + 2*g(key)) % TableSize`
- 3rd probe: `(h(key) + 3*g(key)) % TableSize`
- …
- i[th] probe: `(h(key) + i*g(key)) % TableSize`
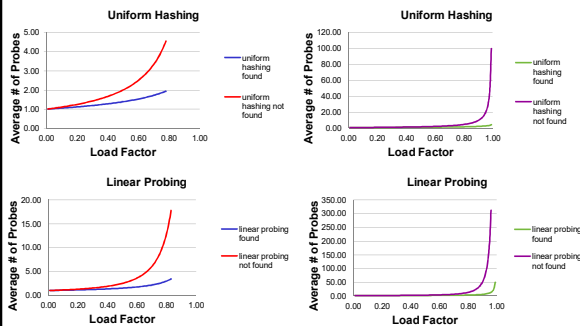
Detail: Make sure `g(key)` cannot be `0`

Double-hashing analysis

- Intuition: Because each probe is "jumping" by `g(key)` each time, we "leave the neighborhood" *and* "go different places from other initial collisions"

- But we could still have a problem like in quadratic probing where we are not "safe" (infinite loop despite room in table)
  - It is known that this cannot happen in at least one case:
    - `h(key) = key % p`
    - `g(key) = q – (key % q)`
    - `2 < q < p`
    - `p` and `q` are prime

More double-hashing facts

- Assume "uniform hashing"
  - Means probability of `g(key1) % p == g(key2) % p` is `1/p`

- Non-trivial facts we won't prove:
  Average # of probes given λ (in the limit as `TableSize` $\rightarrow \infty$)
  - Unsuccessful search (intuitive):
  $$\frac{1}{1-\lambda}$$

  - Successful search (less intuitive):
  $$\frac{1}{\lambda}\log_e\left(\frac{1}{1-\lambda}\right)$$

- Bottom line: unsuccessful bad (but not as bad as linear probing), but successful is not nearly as bad

Charts

Rehashing

- As with array-based stacks/queues/lists, if table gets too full, create a bigger table and copy everything

- With chaining, we get to decide what "too full" means
  - Keep load factor reasonable (e.g., < 1)?
  - Consider average or max size of non-empty chains?

- For probing, half-full is a good rule of thumb

- New table size
  - Twice-as-big is a good idea, except that won't be prime!
  - So go *about* twice-as-big
  - Can have a list of prime numbers in your code since you won't grow more than 20-30 times

Hashtable Scenarios

- For each of the scenarios, answer the following questions:
  - Is a hashtable the best-suited data structure?
  - If so, what would be used at the keys? Values?
  - If not, what data structure would be best-suited?
  - What other assumptions, if any, about the scenario must you make to support your previous answers?

- Catalog of items (product id, name, price)
- Bookmarks in a web browser (favicon, URL, bookmark name)
- IT support requests (timestamp, ticket id, description)
- Character frequency analysis (character, # of appearances)
- Activation records for nested function calls (return addresses, local variables, etc.)