Hashtables

Implementations for dictionaries and sets

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See https://github.com/parrt/msds692/blob/master/notes/hashtable.ipynb and https://github.com/parrt/msds692/blob/master/notes/dict.ipynb



How to search big collections quickly

- Hashtables are data structures that efficiently implement search/lookup operations for sets and dictionaries
- Sets and dictionaries are abstract data structures, hashtables are concrete implementations of those structures
- Simple lists of elements and lists of tuples work but are slow
- Hashtable's key idea: partition the search space into welldefined regions so we don't have to search linearly through the entire collection to find an element
- We use a (hash) function of the values to partition into buckets

Review: Sets

 A set is just an unordered, unique collection of elements; here is an example using integers:

```
ids = {100, 103, 121, 102, 113, 113, 113, 113}
```

We can do lots of fun set arithmetic:

```
{100,102}.union({109})

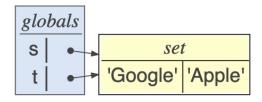
{100, 102, 109}

{100,102}.intersection({100,119})

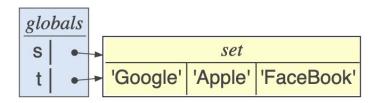
{100}
```

Watch out for aliasing!

```
s = {"Apple", "Google"}
t = s
```



```
s.add("FaceBook")
```



Review: Dictionaries map keys to values

- If we arrange two lists side-by-side and kind of glue them together, we get a dictionary (type is dict)
- Dictionaries map one value to another, just like a dictionary in the real world maps a word to a definition
- Here is a sample dictionary:

```
movies = { 'Amadeus ': 1984, 'Witness ': 1985} 

'Amadeus' \rightarrow 1984

'Witness' \rightarrow 1985
```

Index by key to get the value; e.g., movies['Amadeus']

List implementation for sets

```
import numpy as np
n = 5_000_000
A = list(np.random.randint(low=0,high=1_000_000,size=n))
A[0:10]
```

[509385, 571020, 998421, 173251, 567339, 229005, 614066, 89806, 878866, 496601]

```
def lsearch(A,x):
    for a in A:
        if a==x:
        return True
    return False
```

Searching linearly is pretty slow

```
%time lsearch(A, 999)

CPU times: user 362 ms, sys: 8.02 ms, total: 370 ms
Wall time: 378 ms

True

%time for a in range(50): lsearch(A, a)

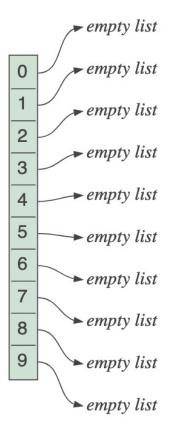
CPU times: user 8.65 s, sys: 102 ms, total: 8.75 s
Wall time: 8.9 s
```

Can we do better than linear search?

- Rather than search through every element, let's partition the set of numbers into 10 buckets so that, on average, we only need to search 1/10 of the elements
- We partition with a hash function that operates on set values
- We are effectively using something about the value to hint at the location
- E.g., where does Eric Erickson live in US?
- Imagine a hash function that gave the postal code given a name (set value)

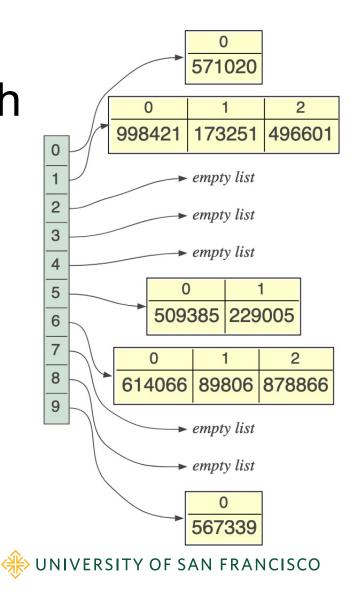
Partitioning requires a new data structure

- Rather than a list of set values, we need a list of regions called buckets where each bucket is a list of values
- The hash of a value leads to the bucket index
- For sets of integers, let's use the value modulo 10 to uniquely place values into one of 10 buckets, indexed 0..9



Partitioning with modulo hash

```
def hash(x):
    return x % 10
[(a,hash(a)) for a in A[0:10]]
[(509385, 5),
 (571020, 0),
 (998421, 1),
 (173251, 1),
 (567339, 9),
 (229005, 5),
 (614066, 6),
 (89806, 6),
 (878866, 6),
 (496601, 1)]
```



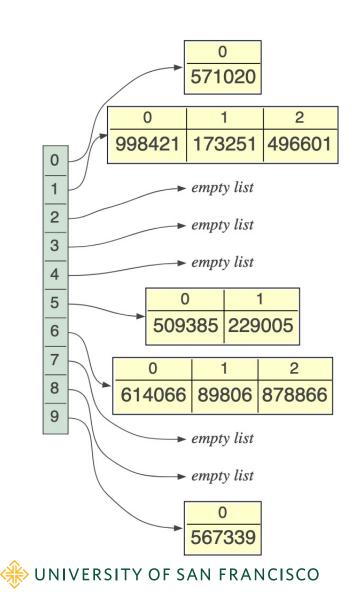
Hashtable construction

- Make 10 empty buckets (lists)
- Add each set element to correct bucket
- Amounts to appending element to one of 10 lists

```
def hash(x):
    return x % 10

"Build hashtable for integer values"
    buckets = [[] for i in range(10)]
    for a in A:
        buckets[hash(a)].append(a)
    return buckets

buckets = htable(A)
```



Searching hashtable set implementation

To find x, look in the bucket indicated by hash(x)

Linear def lsearch(A,x): for a in A: if a==x: return True return False

```
%time lsearch(A, 999)

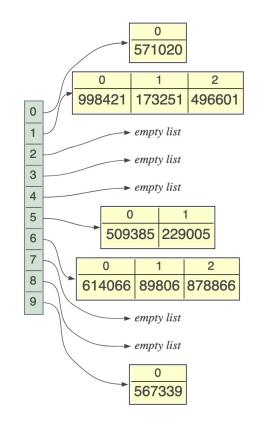
CPU times: user 190 ms, sys:
Wall time: 209 ms
```

Hashtable

```
def hsearch(buckets,x):
    i = hash(x)
    for a in buckets[i]:
        if a==x:
        return True
    return False
```

```
buckets = htable(A)
%time hsearch(buckets, 999)
```

```
CPU times: user 18 ms, sys: Wall time: 18.7 ms
```





Speed difference is dramatic

Linear

```
%time for a in range(50): lsearch(A, a)

CPU times: user 6.97 s, sys: 93.9 ms, total: 7.06 s
Wall time: 7.38 s

Hashtable
%time for a in range(50): hsearch(buckets, a)

CPU times: user 823 ms, sys: 14.3 ms, total: 837 ms
Wall time: 861 ms
```

Exercise

• What would happen if we used a hash function like these?

```
def hash(x):
    return x % 2

def hash(x):
    return 0
```

Sets of strings

- Hashtables work for any value for which we can define a good hash function, one that partitions the space evenly
- What hash function am I using here?

Hash is distance of char code from 'a's code

```
► empty list
'clarksville'
              'corona'
                         'cape coral
                'davie'
                'elgin'
     'fairfield'
                'fresno' 'frisco'
     'greeley'
                'garden grove'
              'hillsboro'
              empty list
              - empty list
               empty list
              empty list
              ► emntv list
```

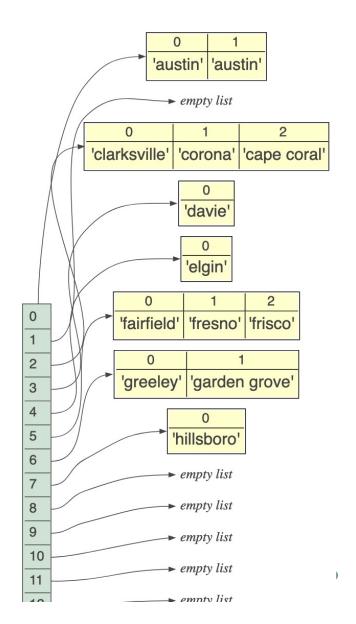
'austin' 'austin'

Hashing strings

```
def hash(s):
    # convert first char to int in [0,25]
    return ord(s[0]) - ord('a')

[(c,hash(c)) for c in cities[0:10]]

[('elgin', 4),
    ('tyler', 19),
    ('austin', 0),
    ('hillsboro', 7),
    ('greeley', 6),
    ('davie', 3),
    ('rockford', 17),
    ('orange', 14),
    ('sandy springs', 18),
    ('garden grove', 6)]
```



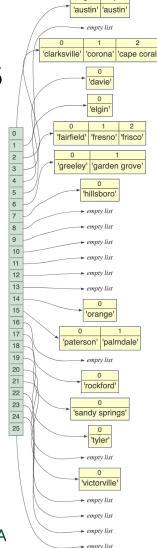
Hashtable impl. differs only in buckets

```
def htable(A):
    buckets = [[] for i in range(26)]
    for a in A:
        buckets[hash(a)].append(a)
    return buckets
```

```
def hsearch(buckets,x):
    i = hash(x)
    for a in buckets[i]:
        if a==x:
            return True
    return False
```

```
buckets = htable(cities)
%time hsearch(buckets, "austin")
```

```
CPU times: user 5 \mus, sys: 0 ns, total: 5 \mus Wall time: 7.15 \mus
```





An important implementation detail

- The hash function does not directly give the bucket index: it converts values to integers, then we make sure that the hash value fits into the table by doing modulo the number of buckets
- For our int sets, the hash(x) is just x; the modulo 10 just puts it in one of 10 buckets
- Same for strings; the hash(x) is in 0..25 but we could stick it into 10 buckets by taking modulo 10

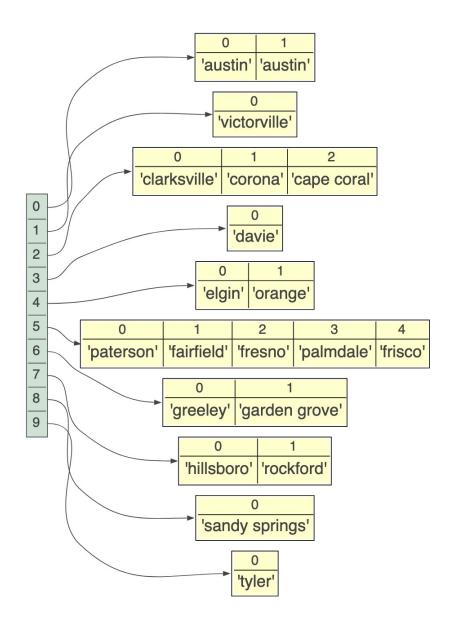
Hash vs bucket index

 Compute the hash and mod with the number of buckets we have

```
def hash(s):
    # convert first char to int in [0,25]
    return ord(s[0]) - ord('a')

def htable(A):
    buckets = [[] for i in range(10)]
    for a in A:
        # fit in 10 buckets
        b = hash(a) % 10
        buckets[b].append(a)
    return buckets
```

Previously: buckets[hash(a)].append(a)



Dictionary implementations

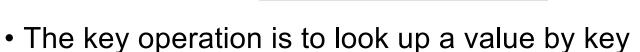


A simple dictionary implementation

Let's represent the set of key
 →value pairs as

a list of tuples:

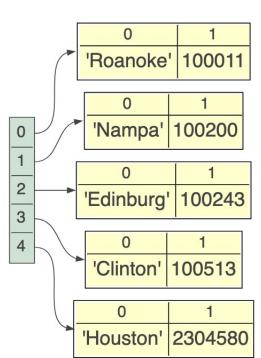
pop = [
 ('Roanoke', 100011),
 ('Nampa', 100200),
 ('Edinburg', 100243),
 ('Clinton', 100513),
 ('Houston', 2304580)
]



How would you implement this?

Linearly search through the list of tuples and compare the first value and the tuple to the key of interest; return the associated value if key is found

we university of san francisco



Linear lookup

- Looking for a key is a simple matter of examining the first element of every tuple stored in the list
- Return None if the key is not found

```
def llookup(A,x):
    for k,v in A:
        if k==x:
        return v
    return None
```

```
llookup(pop, 'Clinton')
```

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```
llookup(pop, 'SF')
```



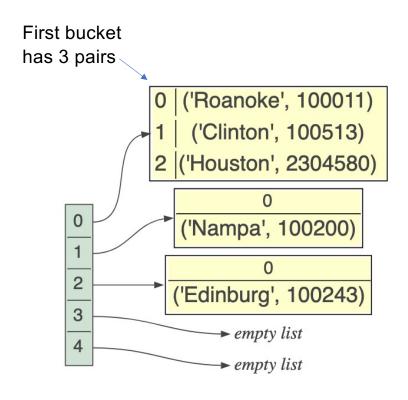
Hashtable dictionary implementation

 Split pairs into, say, 5 buckets (use our string hash function)

```
def hash(s):
    # convert first char to int in [0,25]
    return ord(s[0]) - ord('a')

def htable_dict(A,nbuckets):
    buckets = [[] for i in range(nbuckets)]
    for k,v in A:
        b = hash(k) % nbuckets
        buckets[b].append((k,v))
    return buckets
```

```
buckets = htable_dict(pop, 5)
```





Hashtable key look up

- Compute the hash, modulo a number of buckets, to get the bucket index
- Linear search within the bucket
- If key found, return value
- Else return None

```
def hlookup(buckets,x,nbuckets):
    i = hash(x) % nbuckets
    for k,v in buckets[i]:
        if k==x:
            return v
    return None
```

```
buckets = htable_dict(pop, 3)
hlookup(buckets, 'Clinton', nbuckets=3)
100513
hlookup(buckets, 'SF', nbuckets=3)
```

Degenerate case of one bucket

 With only one bucket, all pairs hash to the same bucket, which means doing a linear search of all elements to look up a key

```
buckets = htable_dict(pop, 1)

| 0 | ('Roanoke', 100011) |
| 1 | ('Nampa', 100200) |
| 2 | ('Edinburg', 100243) |
| 3 | ('Clinton', 100513) |
| 4 | ('Houston', 2304580) |
```

Some details relevant to the search project

Review: tuples

- A tuple is an *immutable* list and uses parentheses rather than square brackets for notation
- Tuples are often used to group related elements:

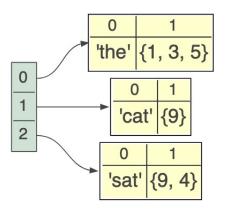
```
me = ('parrt',607)
userid,office = me
print(userid)
print(office)
print(me[0], me[1])
```

```
parrt
607
parrt 607
```

Values can be anything including sets

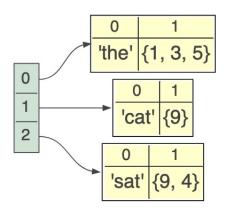
- The tuples used to represent key and value pairs are immutable, but the pair's value can point at a mutable data structures such as a set
- Consider a simple list of tuples implementation that maps words to sets of integers

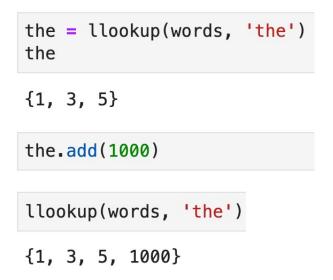
```
words = [('the', {3,1,5}), ('cat', {9}), ('sat', {4,9})]
```

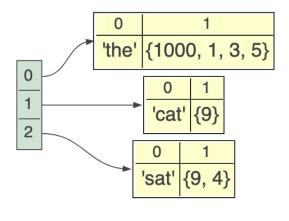


Modifying dictionary set values

 If you extract a mutable value from a data structure, you can modify it without having to delete and add an updated version









Summary

- Dictionaries and sets are typically implemented with a form of hashtable because the key lookup operation is so much faster
- The speed comes from a partitioning of the search space into a large number of small regions, which are searched linearly
- If we make enough buckets so that at most there are three keys in each bucket, lookup takes three operations no matter how many keys have been added to the dictionary