hash functions

```
In [1]: import os
   import re
   import string
   from collections import Counter
   import matplotlib.pyplot as plt
   import pandas as pd
   import time
   import time
```

hash functions

Topic today: how different hash functions affect the computation time?

last week we've talked about how hashtable can still work when collisions happened, But it still is something we want to avoid. and today we'll talk about how to avoid collisions, using different hash functions.

- First, we will set up the data we need
- secondly, we will hash some real life textual data, into our hashtable, and see the
 collisions distribution. Then, we will do the same thing, but with different hash
 functions, and see how it affects the collisions distribution.
- lastly, we will simulate the hash function computation time, and see how different hash functions affect the computation time.

STEP 1: data preparation

```
In [2]:
        def filelist(root: str) -> list[str]:
            """Return a fully-qualified list of filenames under root directory
            traversing subdirectories recursively and return all the paths of files
            return [os.path.join(root, f) for root, _, files in os.walk(root) for f
        def get text(fileName: str) -> str:
            f = open(fileName, encoding='latin-1')
            s = f.read()
            f.close()
            return s
        def words(text: str) -> list[str]:
            0.00
            Given a string, return a list of words normalized as follows.
            Split the string to make words first by using regex compile() function
            and string.punctuation + '0-9\ to replace all those
            char with a space character.
            Split on space to get word list.
            Ignore words < 3 char long.
            Lowercase all words
```

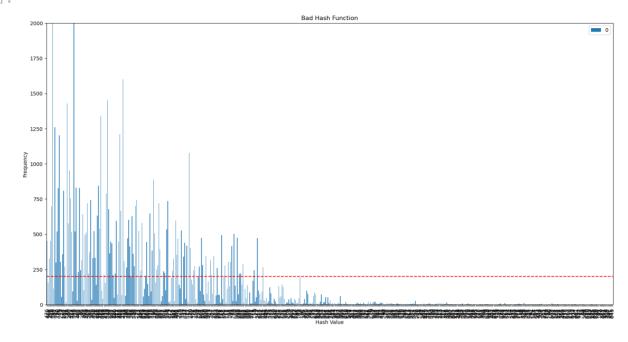
```
regex = re.compile('[' + re.escape(string.punctuation) + '0-9\\r\\t\\n]'
             # delete stuff but leave at least a space to avoid clumping together
             nopunct = regex.sub(" ", text)
             words = nopunct.split(" ")
             words = [w \text{ for } w \text{ in words if } len(w) > 2] # ignore a, an, to, at, be, ...
             words = [w.lower() for w in words]
             # print words
             return words
         def text_data_set_up(path: str) -> list[str]:
             files = filelist(path)
             texts = [get text(f) for f in files]
             word list = [words(t) for t in texts]
             word list = [w for wl in word list for w in wl]
             return list(set(word list))
In [3]: texts = text data set up('./slate')
         len(texts)
         texts[:10]
Out[3]: ['siphons',
          'smacks',
          'gettler',
          'gabbing',
          'scooted',
          'comb',
          'upat',
          'corp',
          'prowled',
          'custom']
```

STEP 2: comparing the hash functions in terms of collisions distribution

```
In [4]: def bad hash(word: str, size: int = 26*26) -> int:
            a = ord(word[0]) - 97
            b = ord(word[1]) - 97
            return (a * 26 + b) % size
        def good_hash(key, size: int = 26*26) -> int:
            key = str(key)
            h = 0
            for i in key:
                h += 31* h + ord(i)
            return h % size
In [5]: good hashes = [good hash(w) for w in texts]
        good hash counter = Counter(good hashes)
        hashes = [bad hash(w) for w in texts]
        bad_hash_counter = Counter(hashes)
In [6]: df = pd.DataFrame.from dict(bad hash counter, orient='index')
        fig, ax = plt.subplots()
        df.plot(kind='bar', ax = ax, figsize=(20, 10), xlim=(0, 26*26), ylim=(0, 200)
        ax.set xlabel('Hash Value')
```

```
ax.set_ylabel('Frequency')
ax.set_title('Bad Hash Function')
ax.hlines(200, 0, 26*26, colors='r', linestyles='dashed')
```

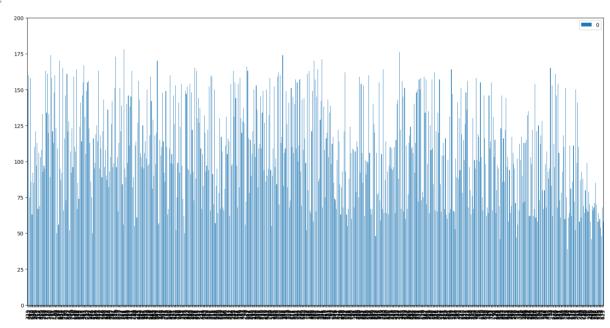
Out[6]: <matplotlib.collections.LineCollection at 0x1437a1190>



```
In [7]: df_g = pd.DataFrame.from_dict(good_hash_counter, orient='index')
    df_g.sort_values(by=0, ascending=False).tail(10)

df_g.plot(kind='bar', figsize=(20, 10), xlim=(0, 26*26), ylim=(0, 200))
```

Out[7]: <AxesSubplot:>



STEP 3: comparing the hash functions in terms of computation time

3-1 Hash table class

```
In [8]: from typing import Callable
```

```
class Hashtable:
   def init (self, hash func: Callable[[str,int],int], size: int) -> Non
       self.size = size
        self.table = [[] for in range(size)]
        self. hash = hash func
    def setitem (self, key: str, value:str) -> None:
        index = self. hash(key, self.size)
        self.table[index].append((key, value))
         getitem (self, key: str) -> str:
        index = self. hash(key, self.size)
        for k, v in self.table[index][::-1]:
            if k == kev:
                return v
    def __delitem__(self, key: str) -> None:
        index = self. hash(key, self.size)
        for k, v in self.table[index]:
            if k == key:
                self.table[index].remove((k, v))
```

```
In [9]: bad_dict = Hashtable(bad_hash, 26*26)
good_dict = Hashtable(good_hash, 26*26)
```

3-2 timer function & testing tasks set up

```
In [10]:
    def timer(f):
        def wrapper(*args, **kwargs):
            start = time.time()
            result = f(*args, **kwargs)
            end = time.time()
            return result, end - start
        return wrapper

    @timer
    def test_speed(my_dict: Hashtable, word_list: list[str]) -> None:
        for i, w in enumerate(word_list):
            my_dict[w] = i

        for w in word_list:
            x = my_dict[w]

        for w in word_list:
            del my_dict[w]
```

3-3 run test

bad_time=2.6138880252838135 seconds

good_time=0.6940028667449951 seconds

In []:	
In []:	