Count min sketch with max heap (25pts)

In the class, we studied count-min sketch, an elegant algorithm to estimate the frequency of elements. Here we want to find the top-k frequent elements with one pass over an insert-only stream, using this algorithm. Note that to get top-k elements with only one pass, we must maintain an additional data structure, e.g., a heap, to store the current top-k frequent elements in the stream.

Let w be the width (num of columns) and d be the depth (num of rows) of the count min sketch. The high-level idea of the algorithm is the following:

- 1. Initialize a heap and a matrix of size $w \times d$.
- 2. For each element e coming from the stream, you should update the count min sketch matrix using d different hash functions (as taught in the class). Then, check if the heap size is already k. If no, insert a tuple (the count of e, e) into the heap; Else, replace the tuple of the least frequent element in the heap, say h_0 , with a tuple of (estimated count of e, e), if e has an estimated frequency larger than h_0 .

Note the heap will have size at most k, and it stores the (estimated) top-k frequent elements. We need only O(wd + k) words of memory to implement the whole idea. Please implement this algorithm as a class in the cell below.

What is heap and how to use it in Python:

A heap is a specialized tree-based data structure which is essentially an almost complete tree that satisfies the heap property: in a max heap, for any given node C, if P is a parent node of C, then the key (the value) of P is greater than or equal to the key of C. A brief introduction can be found here (https://en.wikipedia.org/wiki/Heap (data structure)).

In python, the most common implementation of heap can be found heapq.html). The heapq module implement min heap by storing the data in a list, and exposing all heap-operation APIs needed. You can also find examples of heapq heapq here (https://www.geeksforgeeks.org/heap-queue-or-heapq-in-python/). Note when the data stored in the list are tuples, the first element of each tuple will be considered as the "key".

```
In [1]:
        import heapq
        import time
        import matplotlib.pyplot as plt
        class CountMinSketchHeap:
            The CountMinSketchHeap class implements the Count-Min Sketch algorit
                of item frequencies. It also maintain a heap of size k to store
            def __init__(self, width, depth, k, hashfunc):
                Following variables are initialized here.
                width: an integer representing the number of columns in the has
                depth: an integer representing the number of rows in the hash to
                k: the number of top frequent elements to store.
                hashfunc: a callable hash function.
                counts: a two-dimensional list representing the hash table.
                keys: a list that stores the current top k frequent elements.
                heap: a max heap of size at most k, it stores (count, elment) po
                hash execution time: an integer used to compute the efficiency
                self.width = width
                self.depth = depth
                self.k = k
                self.hashfunc = hashfunc
                self.counts = [[0] * width for _ in range(depth)]
                self.keys = []
                self.heap = []
                self.hash execution time = 0
            def hash(self, key):
                0.00
                DONT CHANGE THIS!
                This function takes a key and generates in total self.depth has
                    pairwise independent hash functions.
                hashes = []
                st = time.time()
                for i in range(self.depth):
                    h = self.hashfunc((i, key)) % self.width
                    hashes.append(h)
                self.hash_execution_time += (time.time()-st)
                return hashes
            def update(self, key, count=1):
                """ This function updates both heap and count min sketch when we
                Please implement below. """
                hashes = self.hash(key)
                for i, h in enumerate(hashes):
                    self.counts[i][h] += count
                estimate = self.query(key)
                if len(self.keys) < self.k:</pre>
                     if key not in self.keys:
                        self.keys.append(key)
                        heapq.heappush(self.heap, (estimate, key))
```

```
else:
            # if key is already stored, check whether to update the
            for i, tup in enumerate(self.heap):
                e, k = tup
                if k == key and estimate > e:
                    self.heap[i] = (estimate, key)
                    heapq.heapify(self.heap)
    else: # cases when the heap is already full
        if key not in self.keys:
            least count, least key = self.heap[0]
            if estimate > least count:
                heapq.heapreplace(self.heap, (estimate, key))
                self.keys.remove(least key)
                self.keys.append(key)
        else:
            # if key is already stored, check whether to update the
            for i, tup in enumerate(self.heap):
                e, k = tup
                if k == key and estimate > e:
                    self.heap[i] = (estimate, key)
                    heapq.heapify(self.heap)
def query(self, key):
    """ This function returns the estimated frequency of an element
    Please implement below. """
    hashes = self.hash(key)
    min count = float('inf')
    for i, h in enumerate(hashes):
        min count = min(min count, self.counts[i][h])
    return min count
def topk(self):
    return sorted(self.heap, reverse=True)
```

Below is a set of hash functions implemented in hashlib package.

```
In [2]: # CODE FOR HASHING, DONT CHANGE
        import hashlib
        import mmh3
        def md5 toint(key):
            a = hashlib.md5()
            a.update(bytes(str(key), 'utf-8'))
            return int.from_bytes(a.digest(), "big")
        def sha256_toint(key):
            a = hashlib.sha256()
            a.update(bytes(str(key), 'utf-8'))
            return int.from bytes(a.digest(), "big")
        def blake2b_toint(key):
            a = hashlib.blake2b()
            a.update(bytes(str(key), 'utf-8'))
            return int.from_bytes(a.digest(), "big")
        def murmur_toint(key):
            return mmh3.hash(str(key))
```

Test and plot (5pts)

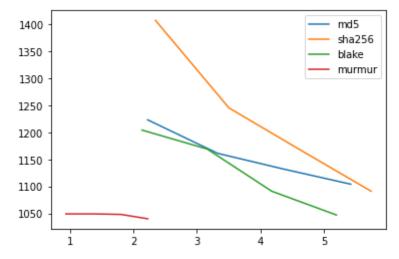
In the folder, we include a text file of the book "blue castle." The next cell use the CountMinSketchHeap implemented above to estimate the top 10 frequent words in this book. Set width=200, range depth in [10,15,20,25], and try different hash functions listed above. The test code collects for each hash function two performance metrics:

- The execution time of the hash function, which equals to hash_execution_time (a variable defined in CountMinSketchHeap).
- The sum of the difference between estimated and real frequencies of top-10 words.

In the last cell, please generate one plot that contains a curve for each hash function, with x-axis representing the execution time and y-axis representing the sum of the difference. Please add markers and legends properly.

```
In [3]: # TEST CODE, DONT CHANGE
        def count min test(depth, hashf):
            c = CountMinSketchHeap(200, depth, 10, hashf)
            f = open("bluecastle.txt","r",encoding="UTF-8")
            for line in f:
                for word in line.split():
                    if word not in d:
                        d[word]=0
                    d[word] += 1
                    c.update(word)
            return sum([i[0] for i in c.topk()])-sum(sorted([d[i] for i in d])[-
        md5_time, md5_diff = [], []
        sha256_time, sha256_diff = [], []
        blake_time, blake_diff = [], []
        murmur time, murmur_diff = [], []
        for depth in [10,15,20,25]:
            diff, timet = count min test(depth, md5 toint)
            md5_time.append(timet)
            md5_diff.append(diff)
            diff, timet = count_min_test(depth, sha256_toint)
            sha256 time.append(timet)
            sha256 diff.append(diff)
            diff, timet = count_min_test(depth, blake2b_toint)
            blake time.append(timet)
            blake diff.append(diff)
            diff, timet = count min test(depth, murmur toint)
            murmur time.append(timet)
            murmur diff.append(diff)
```

```
In [4]: # Plot here
plt.plot(md5_time, md5_diff, label = 'md5')
plt.plot(sha256_time, sha256_diff, label = 'sha256')
plt.plot(blake_time, blake_diff, label = 'blake')
plt.plot(murmur_time, murmur_diff, label = 'murmur')
plt.legend()
plt.show()
```



```
In [ ]:
```