

# 面试常见的大数据问题之 TopK Question

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### **Outline**



- TopK on single node
- TopK on multiple nodes
- Realtime topK with low frequency
- Realtime topK with high frequency
- Bonus: Approx TopK
- Bonus: MapReduce



### Question:

Given a list of unsorted Integer, find top k based on value.



### Which data structure we will use?



# PriorityQueue



### **Priority Queue:**

Like a regular queue, but all elements are ordered based on their value.

### Function:

- offer()
- peak()
- poll()



# PriorityQueue:

- Ascending?
- Descending?



### Why Ascending?

list = 
$$[3, 10, 1000, -99, 98, 99], k = 3$$

Workflow: nextNum > smallest\_number\_in\_queue? replace: continue

- [3, 10, 1000]
- -99<3, continue ——— [3, 10, 1000]
- 98>3, replace [10, 98, 1000]
- 99>10, replace [98, 99, 1000]



#### Question:

Given a list of Twitter Hashtag, find hottest K hashtag.



8:12 AM - 27 Mar 13



#### Overview:

- Data preprocessor
- Priority Queue



Step1: Data Preprocessor

Which data structure to use?



```
Step1: Data Preprocessor
HashMap<String, Integer> counter = new HashMap<>();
for (String word : words) {
    if (counter.containsKey(word)) {
        counter.put(word, counter.get(word) + 1);
    } else {
        counter.put(word, 1);
```



Step 2: What data to be stored in PQ?

- Word?
- Frequency?

### **TopK on Single Node**



```
Key + Frequency:
class Pair {
    String key;
    int frequency;
    Pair(String key, int frequency) {
        this.key = key;
        this.frequency = frequency;
```

### **TopK on Single Node**



```
private Comparator<Pair> pairComparator = new Comparator<Pair>() {
    public int compare(Pair left, Pair right) {
        return left.frequency - right.frequency
};
```

### **TopK on Single Node**



http://www.jiuzhang.com/solutions/top-k-frequent-words/



### Time Complexity:

$$O(n + n*lgk) = O(nlgk)$$

**Space Complexity:** 

$$O(|n|+k)$$

(|n| = number of unique words)



### 场景一:

假设给一组10T文件,文件内容是10million用户,当天的微博搜索记录,求微博今日热搜?



# 能否继续用single node处理?



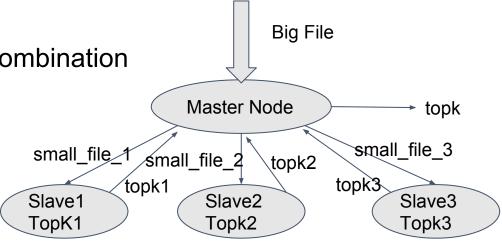
### 不可以:

- 文件太大, 单机无法处理
- 处理速度太慢



Overview: 分 & 合

- Cut into small files
- Dispatch small files to different machines(node)
- Get topk from each machine
- Calculate topk from the topk combination



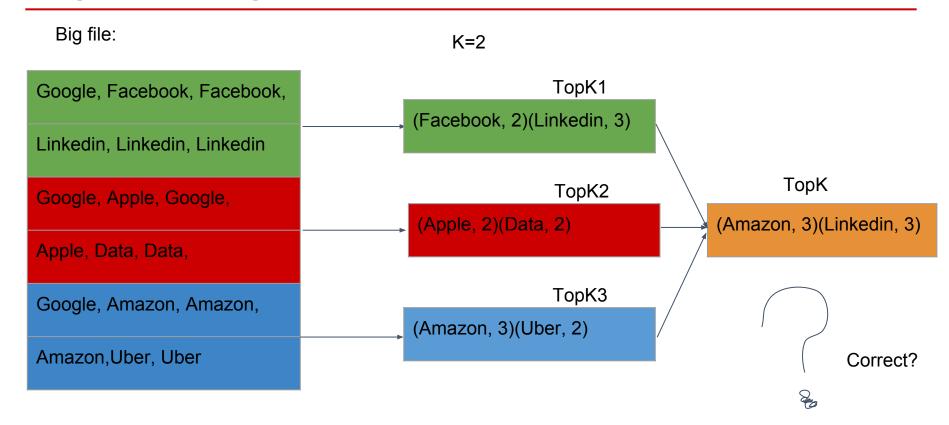


### **Divide Solution:**

Divide by file order?

apple, google, linkedin, airbnb, google, apple, amazon, google, groupon, databricks, airbnb, databricks, amazon, snapchat, uber, datastax, uber, snapchat, databricks, amazon, snapchat, oracle, cisco, amazon, airbnb uber, datastax, uber, snapchat,

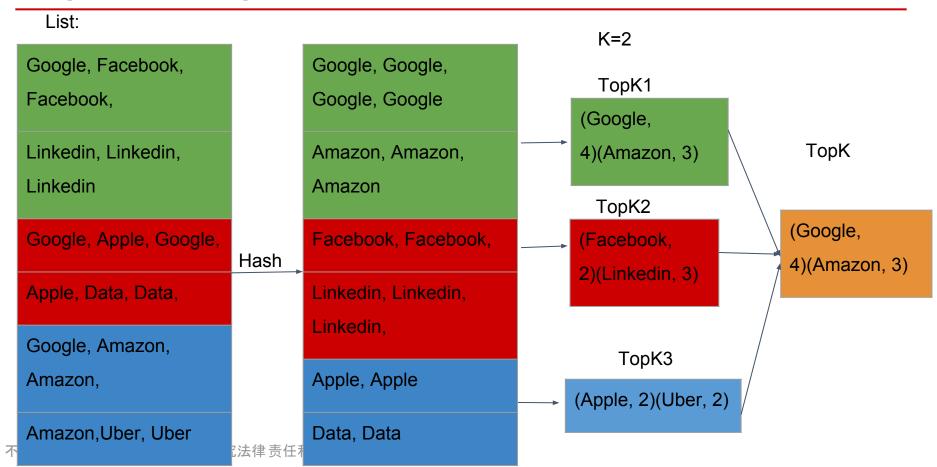






- Divide by hash value
  - SHA1
  - MD5







- Get list of topk: {topk1, topk2, topk3,...}
- Get final topk.



### 场景二:

有N台机器,每台机器各自存储单词文件,求所有单词出现频率的topK



### Solution:

- Get topK from each machine
- Merge {topk1, topk2...}, get final topK

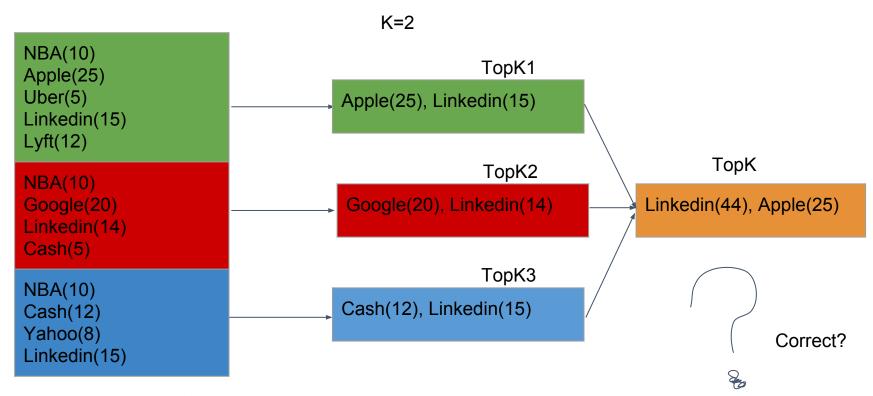


# Wrong!

### **TopK on multiple nodes: Divide**



File on each disk:





### Rehash!

### **TopK on multiple nodes: Divide**



List: K=2 NBA(10) **NBA(10) NBA(10)** Apple(25) NBA(10) Uber(5) Google(20) Linkedin(15) Lyft(12) Apple(25) Uber(5) NBA(10) Lyft(12) Google(20) Linkedin(14) Cash(5) Rehash Cash(12) Cash(5) Linkedin(15) **NBA(10)** Linkedin(14) Cash(12) Linkedin(15) Yahoo(8) Yahoo(8) Linkedin(15)



场景:有实时数据流进入,需要实时计算topK



#### Solution:

- When new data comes in, write it to disk file
- When server request for topK, run the algorithm on disk file
- Get topK



### Disadvantage:

Too slow!



#### Solution:

- When new data comes in, write it to hashmap
- When hashmap is updated, trigger the PQ at the same time
- Get topK



### Disadvantage:

- OOM(out of memory)
- Data loss when node failure or power off.

### Happens in which stage?

- Hashmap?
- Priority Queue?



### Replacement for HashMap:

- Store data in database
- Update counter in database

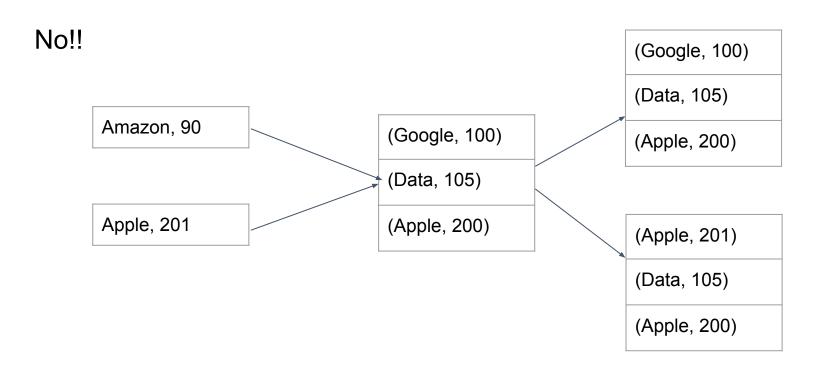
ID	Hashtag	Counter
1	Hello	2
2	World	5



Does PQ have the same function as before?

- Store topK
- When new data comes in, compare new data with smallest data
- Replace or continue







Solution → Use treemap to replace PQ

- Order by value
- Support all the functions in PQ
- Support find and delete by key



Follow-up:

What if the input stream has high frequency?

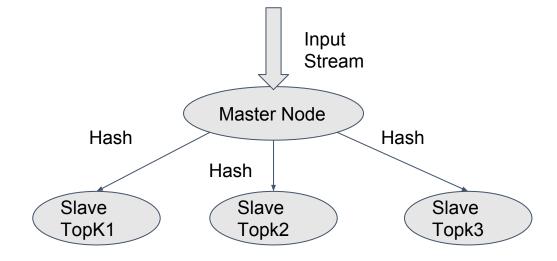


Qps too high → database couldn't respond immediately → high latency



### Solution:

分:





### 合:

- Get topK from each machine
- Merge {topk1, topk2...}, get final topK



Follow-up:

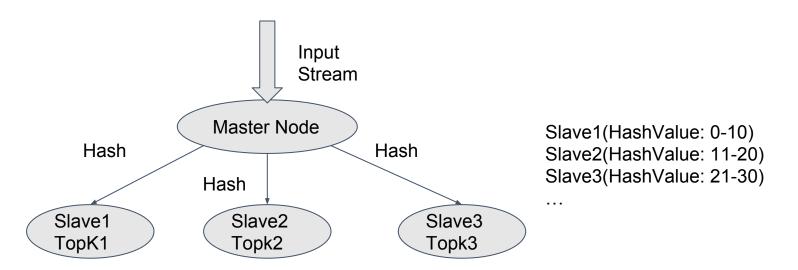
What if one key is too hot, writing frequency is very heavy on one node?

What will happen?

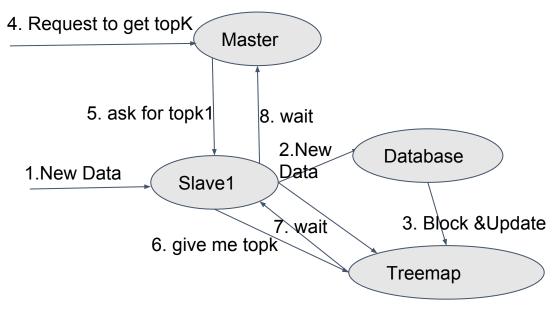


# **High Latency!**









Slave1 receive new data

Slave1 update counter in data base

Slave1 update topk inside PriorityQueue (PriorityQueue will be locked for writing)

Slave1 reading topK operation will be waiting



Confliction:

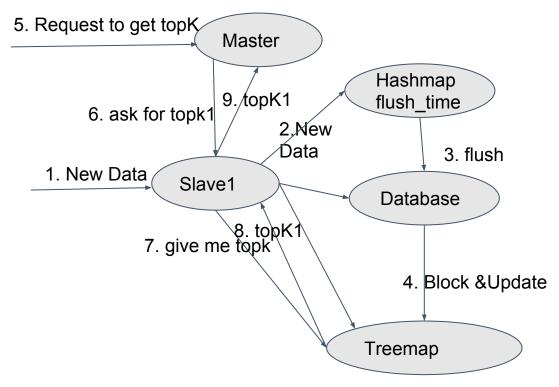
Acurracy VS Latency



#### Solution:

Cache





What to cache? Keyword-count pair.

Where to cache? Slave node.

Slave1 receive new data

Store new data into hashmap

Slave1 receive new data

Store new data into hashmap

After flush\_time, flush data into database

### Question



Store all words on disk:

Low frequency words take up so much space.



#### Solution:

Tradeoff: Sacrifice accuracy for space

- Flexible space!
- O(logK) time complexity
- 牛逼!



#### Overview:

- When a new word comes in, update its count in hashmap
- Update topK in treemap

## **Approx TopK Algorithm: Hashmap**



### Hashmap:

- Key = word\_hashvalue
- Value = frequency
- size can be customized



#### Inputstream

Apple(1)

Google(4)

Facebook(4)

Google(4)

Apple(1)

Linkedin(2)

Uber(6)

Google(4)

Uber(6)

Google(4)

Apple(1)

Amazon(6)

Snapchat(2)

Uber(6)

Apple(1)

#### Hashmap

Key	Value	
1	3	
2	2	
4	3	
66	23	

#### TreeMap

Key	Value
Apple	4
Google	8
<b>Appale</b> toorok	3

不允许录像与传播录像, 否则将追究法律责任和经济赔偿。



### Disadvantages:

- All low frequency words will be hashed to same value, which will result in incoerrect result. (low possibility)
- Some low frequency words will come later, which will have a great count, then replace other high frequency words.(bloom filter)

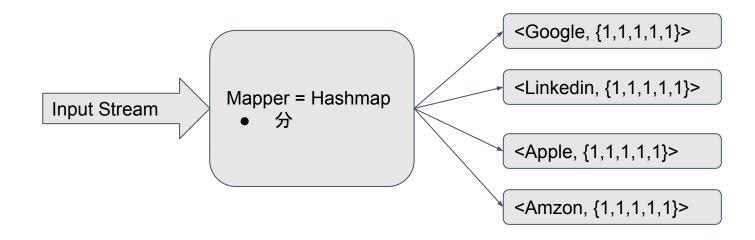


#### Bloom Filter:

- Hashmap will have 3 different hash functions
- Choose the lowest count from hashmap.
- <a href="https://en.wikipedia.org/wiki/Bloom\_filter">https://en.wikipedia.org/wiki/Bloom\_filter</a>

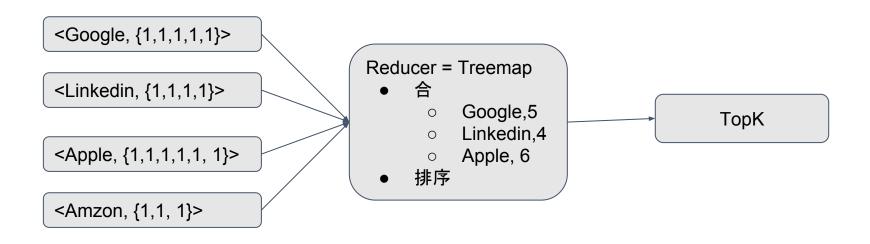
### **Use Mapreduce to Solve it: Mapper**





### **Use Mapreduce to Solve it: Reducer**





### **Use Mapreduce to Solve it**



http://www.lintcode.com/en/problem/top-k-frequent-words-map-reduce/

### What you learned



- Use PQ and Hashmap to calculate topK
- When the file is too big → 先分再合
- Real time data analytics system design
  - Latency vs Accuracy
  - Cache
- Approx TopK Algorithm
- Mapreduce to implement topK