CS 378 – Big Data Programming

Lecture 21
Filtering Patterns

Review

- Assignment 8 User Session
 - Replicated join, multiple outputs

- Questions/issues:
 - DistributedCache
 - MultipleOutputs

Filtering Patterns

- Basic filtering
 - grep
 - Random sampling
 - Unbiased, biased
- Bloom filter
- Top N
- Distinct

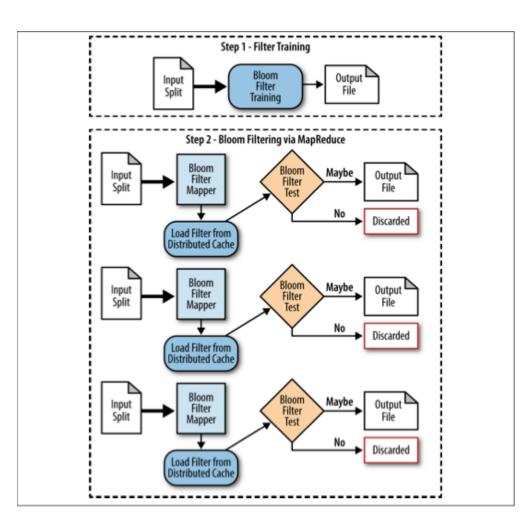
- Bloom filter like the basic filter
- But selection predicate is:
 - Does record contain a value from a predefined set?

This set may be too large to fit in memory

Bloom filter is fixed size, but has false positives

Bloom Filter – Data Flow

Figure 3-2 from MapReduce Design Patterns



- Bloom filter commonly used as map-only
 - Output files will have some false positives
 - Code examples in the book (pp. 53 57)
- We discussed how to combine Bloom filter with reduce-side join
 - Bloom filter represented user IDs with leads
 - Applied in the mapper
 - Reduced the data sent to reduce
 - Reduce eliminated false positives (non-lead sessions)

- Probabilistic data structure
 - Used to test whether something is in a predefined set
 - Can create "false positives"
 - Knows for sure that something is not a member of the set
 - Sometimes reports membership as true, when it is false
 - Never creates "false negatives"
 - Never reports "not a member" when it in fact it is a member
- Fixed size in memory
 - Train the filter using members of the set

- Can add members to the set (further training)
 - Can't remove members
 - There is a technique that allows removal
- Parameters of the filter
 - Number of bits in a bit array
 - Number of independent has functions
- These can be tuned to get a certain false positive rate

- We know that we want a specific number of outputs
 - Based on some evaluation/ranking criterion
- An obvious approach is to sort first
- But total sort is expensive for large data
 - In Hadoop, or in a database
- Output should be significantly smaller than the input

How might we accomplish this without sort?

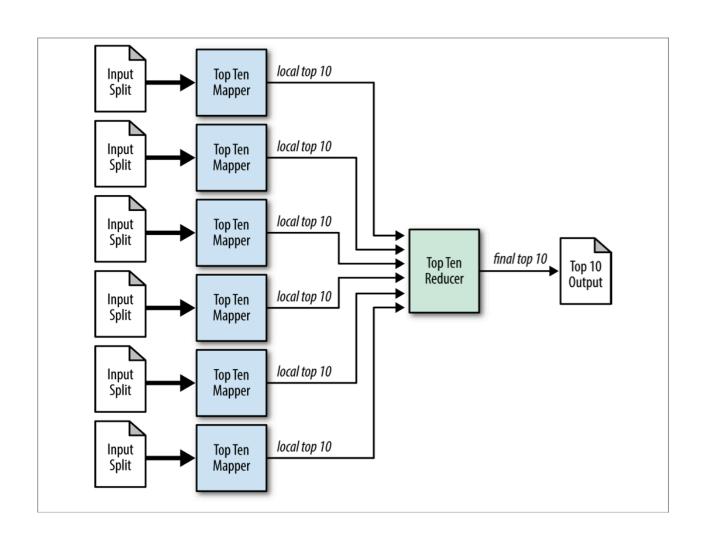
- Start with a comparison method
 - Given two records, which one is larger

- Each mapper finds the top ten from its data
- Each mapper sends it top ten to reduce
- Reduce finds the final top ten
 - How many reducers?

```
class mapper:
   setup():
      initialize top ten sorted list
   map(key, record):
      insert record into top ten sorted list
      if length of array is greater-than 10 then
         truncate list to a length of 10
   cleanup():
      for record in top sorted ten list:
         emit null, record
```

```
class reducer:
    setup():
        initialize top ten sorted list

reduce(key, records):
        sort records
        truncate records to top 10
        for record in records:
        emit record
```



- Remember to copy records retained in map ()
 - Why?
- What are the key/value output by the mappers?
- For top N, if N large, this pattern becomes inefficient
 - Single reducer
 - Data transferred to reduce
 - Reduce input is sorted (expensive for large data)
 - No parallel writes from reduce

Distinct

Want only one record when duplicate records exist

Map:

- Extract the data of interest (if not the entire record)
- Output this data as the key
- Make the value output by map() NullWritable

Reduce:

- Simply write out each unique key (the data of interest)
- Can use a large number of reducers

Distinct

Can we use a combiner?

- If duplicates are rare, combiner doesn't help much
- If duplicates are common, or co-located, a combiner can greatly reduce the data transferred

- Suppose we want all the data in the record, and
 - The compare method is complex
 - Can we approach this problem differently?

Assignment 9 - Filtering

Start with Assignment 8 - remove DistributedCache

- For our four session categories
 - Keep all submitters
 - Searchers by: output only those with Carfax actions
 - Browsers: random sample at %
 - Bouncers: random sample at % / 10

Pass in the % from the command line