Sravani Beeram CS6240 Parallel Data Processing using Mapreduce Homework 2

Pseudo Code - Question : 1

1 No-Combiner:

```
<u>Mapper</u>
       map(value) {
      //From value consider only records that contain either TMIN or TMAX
         key = stationId
          temp = temperature_value
         // If record contains TMIN
            value = {temp,0,true}
         // If record contains TMAX
            value = {0,temp,false}
          emit(key,value)
      }
Reducer
       reducer(key,value_list){
       for (value : value_list){
        // If minTemp
          minTempSum += value.minTemp()
          minCount++
       // else
         maxTempSum += value.maxTemp()
         maxCount++
      }
       Output = key + (minTempSum/minCount) + (maxTempSum/maxCount)
       emit(output,res)
                        //res is NullWritable
      }
```

2 Combiner:

```
<u>Mapper</u>
       map(value) {
      //From value consider only records that contain either TMIN or TMAX
         key = stationId
         temp = temperature_value
         // If record contains TMIN
            value = {temp,0,true}
         // If record contains TMAX
            value = {0,temp,false}
          emit(key,value)
      }
Combiner
       reduce(key,value_list){
       minTempSum = 0
       maxTempSum = 0
       minCount = 0
       maxCOunt = 0
      for (value : value_list){
       // If minTemp
         minTempSum += value.minTemp()
         minCount++
       // else
         maxTempSum += value.maxTemp()
         maxCount++
       data = {minTempSum,maxTempSum,minCount,maxCount}
       emit(key,data)
     }
```

Reducer

3 InMapper:

```
Mapper:
```

```
setup(){
//Initialize hashmap
}
map(value) {
//From value consider only records that contain either TMIN or TMAX
   key = stationId
   temp = temperature_value
   // If record contains TMIN
     //if hashmap weather_data already contains key stationId
        //update the temperature and count
     //else
        // value = (minTemp,maxTemp,minCount,maxCount)
        //insert key and value(temp,0,1,0) into hashmap
   // If record contains TMAX
     //if hashmap weather_data already contains key stationId
        //update the temperature and count
     //else
        //insert key and value(0,temp,0,1) into hashmap
  }
 cleanup(){
  for(s : hashmap){
   emit(s,hashmap(s))
 }
```

Reducer

Pseudo Code - Question: 2

Secondary Sort:

```
Mapper
    map(value) {

    //From value consider only records that contain either TMIN or TMAX

    temp = temperature_value
    CompositeKey = {stationId,year}

    // If record contains TMIN

    value = {temp,0,1,0}
    emit(CompositeKey,value)

    // If record contains TMAX

    value = {0,temp,0,1}
    emit(CompositeKey,value)
}
```

```
Partitioner
      getPartition(CompositeKey,numReduceTasks){
         return key.getStationId().hashCode() % numReduceTasks
     }
Key Comparator
    //Sorts in increasing order of stationId first and If the stationId is equal, sorts in decreasing
order of year next
  compare(CompositeKey k1,CompositeKey k2){
    res= key1.getStationId().compareTo(key2.getStationId())
    if(res == 0) {
          return key1.getYear().compareTo(key2.getYear())
     }
       return res
 }
Grouping Comparator
     //Sorts in increasing order of stationId and does not consider year
    compare(CompositeKey k1,CompositeKey k2){
      return key1.getStationId().compareTo(key2.getStationId())
    }
Reducer
  reduce(CompositeKey,value_list){
      for(value :value_list){
        if(current_year == prev_year){
              minTempSum += s.getMinTemp()
              minCount += s.getMinCount()
              maxTempSum += s.getMaxTemp()
              maxCount += s.getMaxCount()
        }else{
```

```
prev_year = current_year

minTempSum = s.getMinTemp()
minCount = s.getMinCount()

maxTempSum = s.getMaxTemp()
maxCount = s.getMaxCount();

}

Rec = year + (minTempSum/minCount) + (maxTempSum/maxCount)

// if arraylist already has a value with year same as in Rec then replace the
// value with Rec
// else insert the value into arraylist
emit(stationId+arraylist,res) //res is NullWritable
}
```

<u>Comments</u>::When we call reduce method we sort objects in order of its key(stationId,year).Output through this reduce call would be (stationid,year) values for each record. The output format expected is such that we need to have all stationid's listed for a single year. So, we use a grouping comparator which compares objects on the basis of stationId and groups them together to display year and value

Performance Comparison

Running Times:

Programs	Run 1(seconds)	Run 2(seconds)
NoCombiner	80	82
Combiner	76	80
InMapper	74	74
Secondary Sort	58	-

1. Was the Combiner called at all in program Combiner? Was it called more than once per Map task?

Yes, Combiner was called in program Combiner. Below data from syslog confirms the same.

Combine input records=8798241 Combine output records=223783

Hadoop doesn't guarantee on how many times a combiner function will be called for each map task key. So, we cannot tell if combiner was called more than once or not.

2. What difference did the use of a Combiner make in Combiner compared to NoCombiner?

In combiner, number of input records in reducer is very less compared to NoCombiner there by decreasing the overhead usage.

Below data from syslog confirms the same

No-Combiner: Map output records=8798241

Reduce input records=8798241

Combiner: Map output records=8798241

Reduce input records=223783

3. Was the local aggregation effective in InMapperComb compared to NoCombiner?

Yes, the number of map output records and reduce input records in InMapper decreased significantly compared to the NoCombiner map output records

InMapperCombiner: Map input records=30868726

Map output records=223783

Reduce input records=223783 Reduce output records=14135

No Combiner: Map input records=30868726

Map output records=8798241

Reduce input records=8798241 Reduce output records=14135

4. Which one is better, Combiner or InMapperComb? Briefly justify your answer.

InMapperCombiner: Map input records=30868726

Map output records=223783
Reduce input records=223783

Combiner: Map input records=30868726

Map output records=8798241 Reduce input records=223783

Runtime of InMapper is less compared to Combiner. But, InMapper uses Hashmap it needs more memory compared to Combiner. So if memory is not an issue then InMapper usage is better but if memory is an issue then using Combiner is better.

5. How do the running times and accuracy of these MapReduce programs compare to the sequential implementation of per-station mean temperature?

Sequential implementation took approximately 8.5 seconds to run which is significantly very less compared to the mapreduce programs. As current dataset is small (only 1 GB) and Map reduce programs have overhead as data transfers between map

and reduce which is not the case in sequential. So, sequential running time is very less. But this wouldn't be case if the data is very huge. If data is very huge then Mapreduce programs run faster compared to sequential.