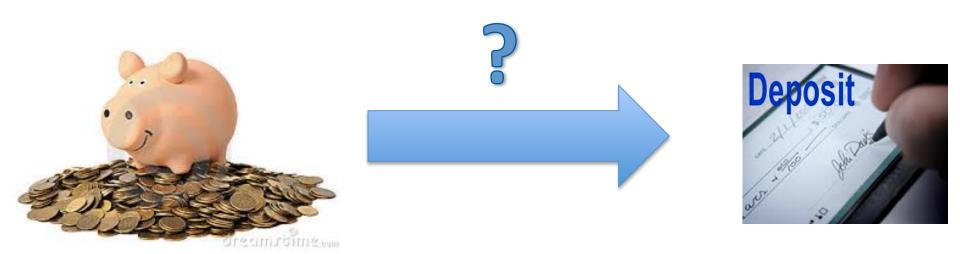
MapReduce

MapReduce Outline

- MapReduce Architecture
- MapReduce Internals
- MapReduce Examples
- JobTracker Interface

MapReduce: A Real World Analogy

Coins Deposit



MapReduce: A Real World Analogy

Coins Deposit



Coins Counting Machine

MapReduce: A Real World Analogy

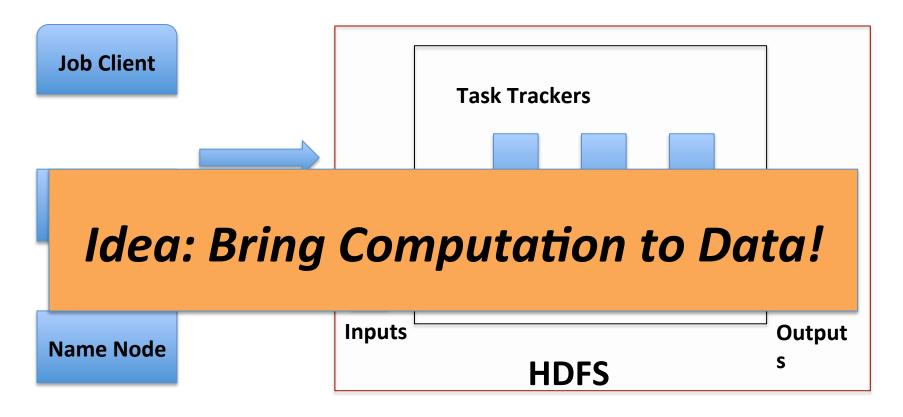
Coins Deposit



Mapper: Categorize coins by their face values

Reducer: Count the coins in each face value in parallel

MapReduce Architecture: Master-Slaves

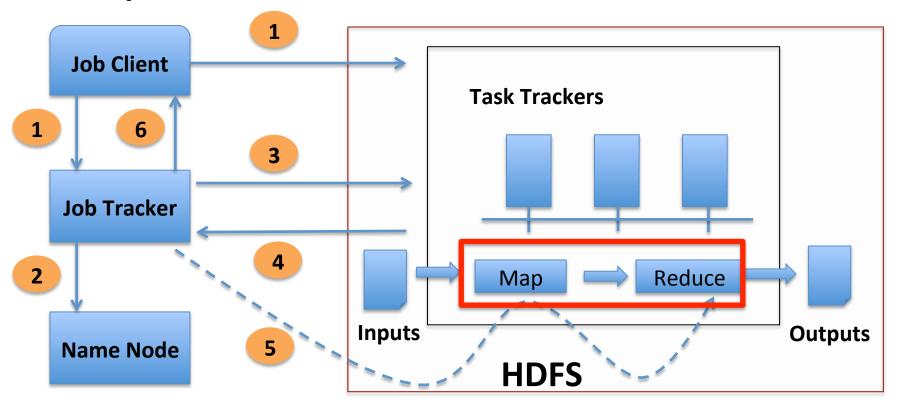


Job Client: Submit Jobs Task Tracker: Execute Jobs

Job Tracker: Coordinate Jobs Job: MapReduce Function+ Config

(Scheduling, Phase Coordination, etc.)

MapReduce Architecture: Workflow



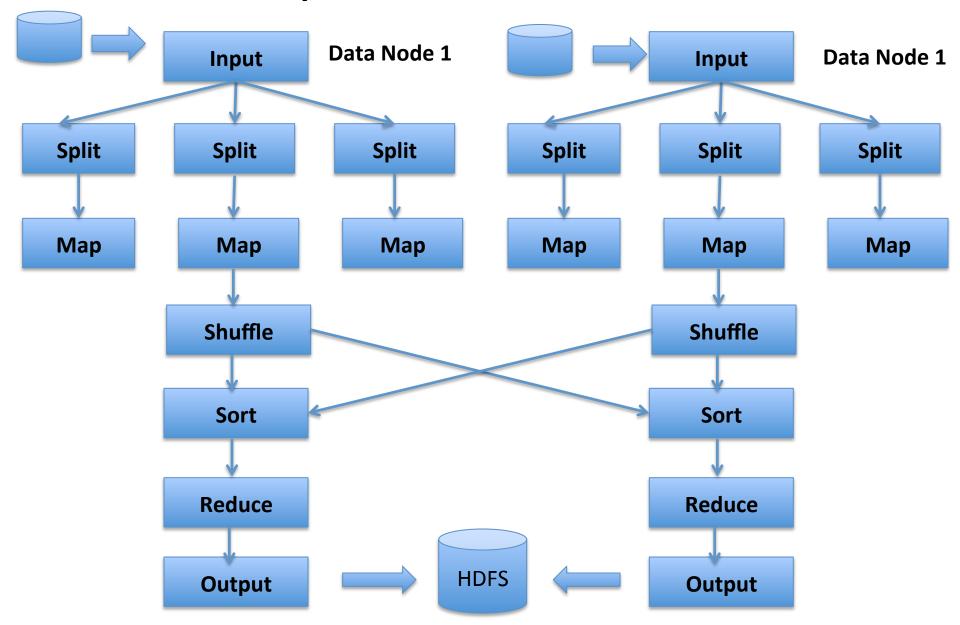
- 1. Client submits job to Job Tracker and copy code to HDFS
- 2. Job Tracker talks to NN to find data it needs
- 3. Job Tracker creates execution plan and submits work to Task Trackers

- 4. Task trackers do the job and report progress/status to Job Tracker
- 5. Job Tracker manages task phases
- 6. Job Tracker finishes the job and updates status

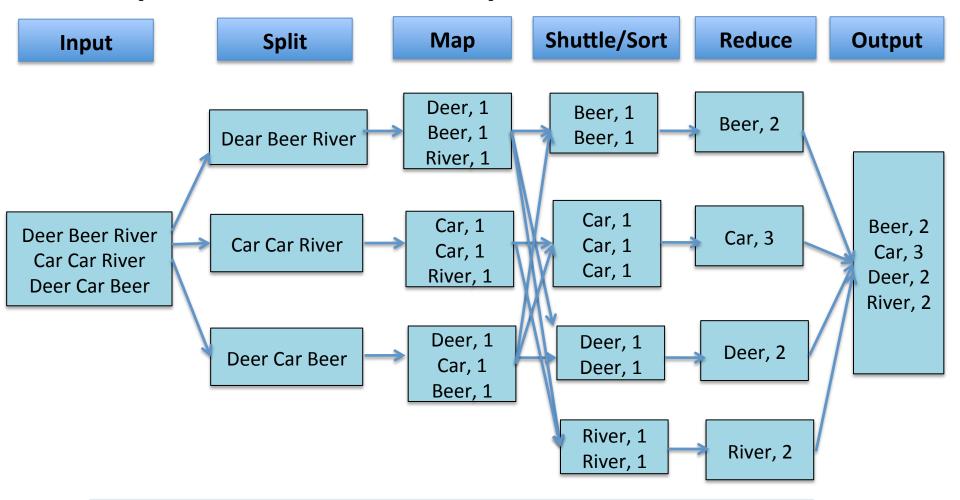
MapReduce Paradigm

- Implement two functions:
 - **Map** (k1,v1) -> list (k2, v2)
 - Reduce(k2, list(v2)) -> list (v3)
- Framework handles everything else
- Value with the same key go to the same reducer

MapReduce Internal



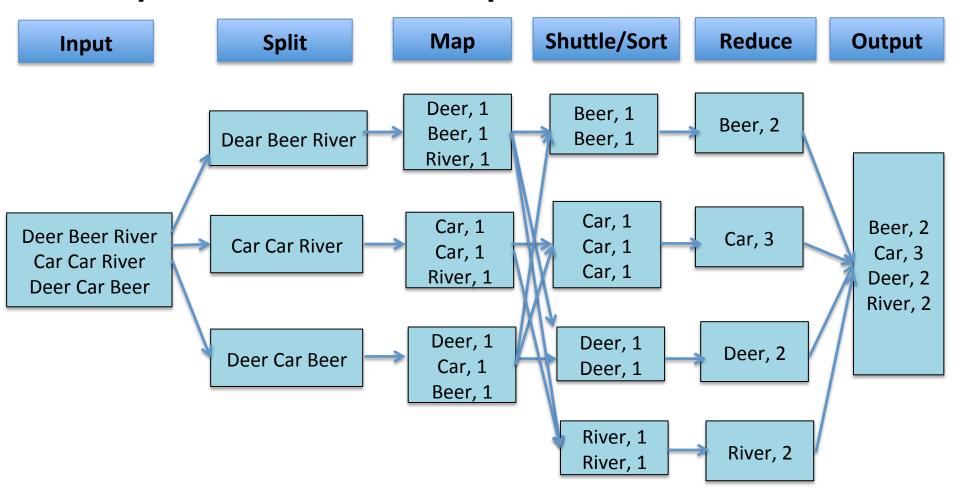
MapReduce Example: Word Count



Similar Flavor of Coins Deposit?



MapReduce Example: Word Count



Q: What are the Key and Value Pairs of Map and Reduce?

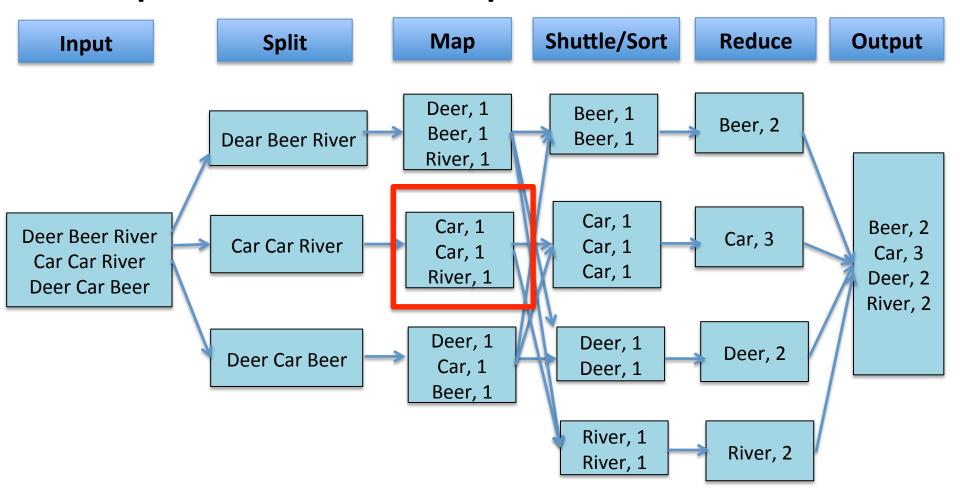
Map: Key=word, Value=1

Reduce: Key=word, Value=aggregated count

Mapper and Reducer of Word Count

```
Map(key, value){
   // key: line number
   // value: words in a line
   for each word w in value:
       Emit(w, "1");}
                                   Combiner is the same
  Reduce(key, list of values){
                                   as Reducer
   // key: a word
   // list of values: a list of counts
   int result = 0;
   for each v in values:
       result += ParseInt(v);
   Emit(key, result);}
```

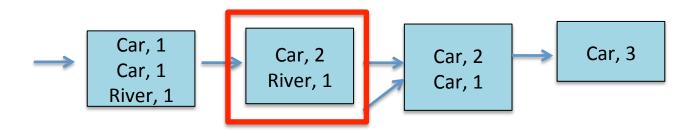
MapReduce Example: Word Count



Q: Do you see any place we can improve the efficiency? Local aggregation at mapper will be able to improve MapReduce efficiency.

MapReduce: Combiner

Combiner: do local aggregation/combine task at mapper



- Q: What are the benefits of using combiner:
 - Reduce memory/disk requirement of Map tasks
 - Reduce network traffic
- Q: Can we remove the reduce function?
 - No, reducer still needs to process records with same key but from different mappers
- Q: How would you implement combiner?
 - It is the same as Reducer!

- New Goal: output all words sorted by their frequencies (total counts) in a document.
- Question: How would you adopt the basic word count program to solve it?

Solution:

- Sort words by their counts in the reducer
- Problem: what happens if we have more than one reducer?

- New Goal: output all words sorted by their frequencies (total counts) in a document.
- Question: How would you adopt the basic word count program to solve it?

Solution:

- Do two rounds of MapReduce
- In the 2nd round, take the output of WordCount as input but <u>switch key and value pair</u>!
- Leverage the sorting capability of shuffle/sort to do the global sorting!

- New Goal: output the top K words sorted by their frequencies (total counts) in a document.
- Question: How would you adopt the basic word count program to solve it?

Solution:

- Use the solution of previous problem and only grab the top K in the final output
- Problem: is there a more efficient way to do it?

- New Goal: output the top K words sorted by their frequencies (total counts) in a document.
- Question: How would you adopt the basic word count program to solve it?
- Solution:
 - Add a sort function to the *reducer* in the first round and only output the top K words
 - Intuition: the <u>global top K must be a local top K</u> in any reducer!

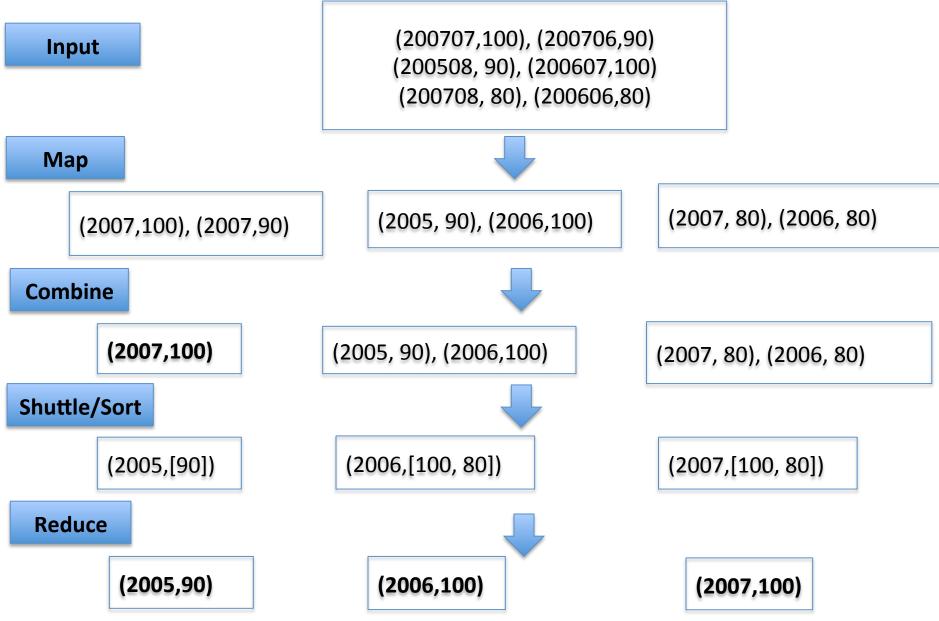
MapReduce In-class Exercise

- Problem: Find the maximum monthly temperature for each year from weather reports
- Input: A set of records with format as:
 - <Year/Month, Average Temperature of that month>
 - (200707,100), (200706,90)
 - (200508, 90), (200607,100)
 - (200708, 80), (200606,80)
- Question: write down the Map and Reduce function to solve this problem
 - Assume we split the input by line

Mapper and Reducer of Max Temperature

```
Map(key, value){
   // key: line number
   // value: tuples in a line
   for each tuple t in value:
                                          Combiner is the same
       Emit(t->year, t->temperature);}
                                          as Reducer
Reduce(key, list of values){
   // key: year
   //list of values: a list of monthly temperature
   int max_temp = -100;
   for each v in values:
       max_temp= max(v, max_temp);
   Emit(key, max temp);}
```

MapReduce Example: Max Temperature



MapReduce In-class Exercise

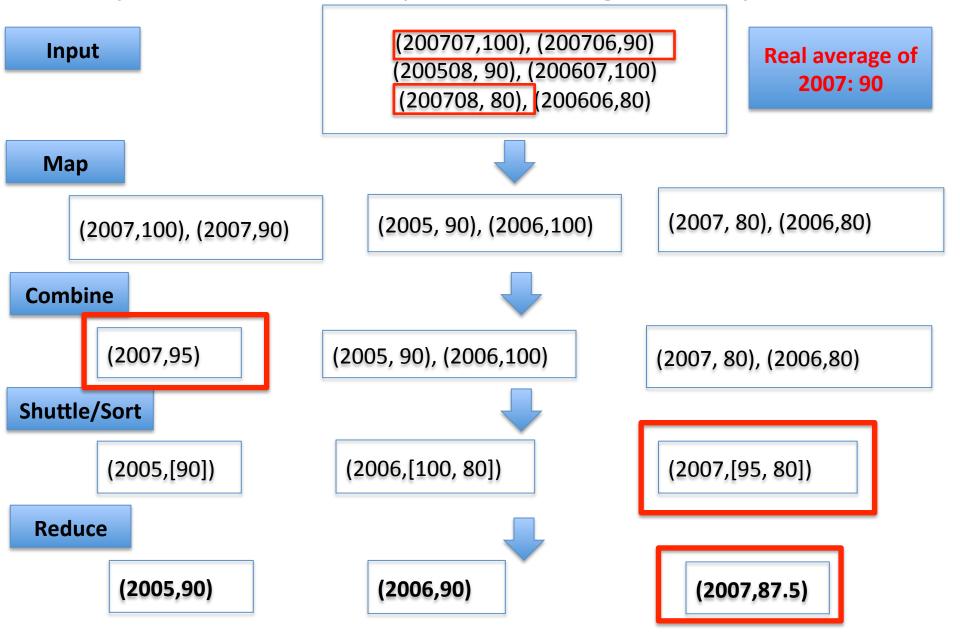
- Key-Value Pair of Map and Reduce:
 - Map: (year, temperature)
 - Reduce: (year, maximum temperature of the year)

 Question: How to use the above Map Reduce program (that contains the combiner) with slight changes to find the average monthly temperature of the year?

Mapper and Reducer of Average Temperature

```
Map(key, value){
 // key: line number
 // value: tuples in a line
 for each tuple t in value:
     Emit(t->year, t->temperature);}
                                        Combiner is the same
Reduce(key, list of values){
                                        as Reducer
 // key: year
 // list of values: a list of monthly temperatures
 int total_temp = 0;
 for each v in values:
     total_temp= total_temp+v;
 Emit(key, total_temp/size_of(values));}
```

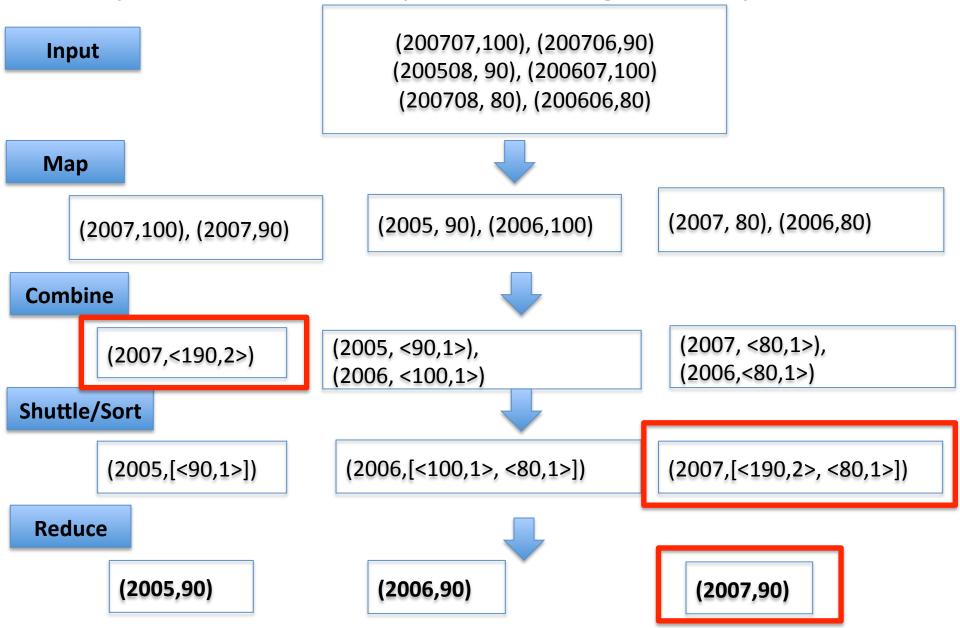
MapReduce Example: Average Temperature



MapReduce In-class Exercise

- The problem is with the combiner!
- Here is a simple counterexample:
 - (2007, 100), (2007,90) -> (2007, 95) (2007,80)->(2007,80)
 - Average of the above is: (2007,87.5)
 - However, the real average is: (2007,90)
- However, we can do a small trick to get around this
 - Mapper: (2007, 100), (2007,90) -> (2007, <190,2>)
 (2007,80)->(2007,<80,1>)
 - Reducer: (2007,<270,3>)->(2007,90)

MapReduce Example: Average Temperature



Mapper and Reducer of Average Temperature

```
Map(key, value){
   // key: line number
   // value: tuples in a line
   for each tuple t in value:
       Emit(t->year, t->temperature);}
   Reduce (key, list of values){
    // key: year
    // list of values: a list of <temperature
sums, counts> tuples
    int total_temp = 0;
    int total count=0;
    for each v in values:
        total temp= total temp+v->sum;
         total_count=total_count+v->count;
  Emit(key,total_temp/total_count)|
```

```
• Combine(key, list of values){
    // key: year
    // list of values: a list of monthly
temperature
    int total_temp = 0;
    for each v in values:
        total_temp= total_temp+v;
Emit(key, <total_temp, size_of(values)>);}
```

MapReduce In-class Exercise

- Functions that can use combiner are called distributive:
 - Distributive: Min/Max(), Sum(), Count(), TopK()
 - Non-distributive: Mean(), Median(), Rank()

Gray, Jim*, et al. "Data cube: A relational aggregation operator generalizing group-by, cross-tab, and subtotals." Data Mining and Knowledge Discovery 1.1 (1997): 29-53.

*Jim Gray received Turing Award in 1998

- Problem 1: Find Word Length Distribution
- Statement: Given a set of documents, use Map-Reduce to find the length distribution of all words contained in the documents

Question:

– What are the Mapper and Reducer Functions?

This is a test data for the word length distribution problem

MapReduce
7: 1
6: 1
4: 4
3: 2
2: 1
1: 1

Mapper and Reducer of Word Length Distribution

```
Map(key, value){
 // key: document name
 // value: words in a document
 for each word w in value:
     Emit(length(w), w);}
Reduce(key, list of values){
 // key: length of a word
 // list of values: a list of words with the same length
  Emit(key, size of(values));}
```

- Problem 1: Find Word Length Distribution
- Mapper and Reducer:
 - Mapper(document)
 { Emit (Length(word), word) }
 - Reducer(output of map)
 - { Emit (Length(word), Size of (List of words at a particular length))}

- Problem 2: Indexing & Page Rank
- Statement: Given a set of web pages, each page has a page rank associated with it, use Map-Reduce to find, for each word, <u>a list of</u> <u>pages (sorted by rank)</u> that contains that word

Question:

— What are the Mapper and Reducer Functions?



Mapper and Reducer of Indexing and PageRank

```
Map(key, value){
   // key: a page
   // value: words in a page
   for each word w in value:
       Emit(w, <page_id, page_rank>);}
Reduce(key, list of values){
   // key: a word
   // list of values: a list of pages containing that word
    sorted pages=sort(values, page_rank)
   Emit(key, sorted pages);}
```

- Problem 2: Indexing and Page Rank
- Mapper and Reducer:

```
- Mapper(page_id, <page_text, page_rank>)
{ Emit (word, <page_id, page_rank>) }
```

– Reducer(output of map)

{ Emit (word, List of pages contains the word sorted by their page_ranks)}

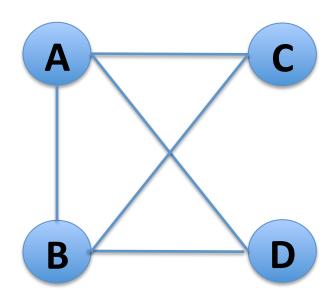
- Problem 3: Find Common Friends
- Statement: Given a group of people on online social media (e.g., Facebook), each has a list of friends, use Map-Reduce to find <u>common</u> <u>friends</u> of any two persons who are friends

Question:

— What are the Mapper and Reducer Functions?



- Problem 3: Find Common Friends
- Simple example:



Input:

A -> B,C,D B-> A,C,D

C-> A,B

D->A,B

Output:

(A,B) -> C,D

(A,C) -> B

(A,D) -> ..

•••

MapReduce

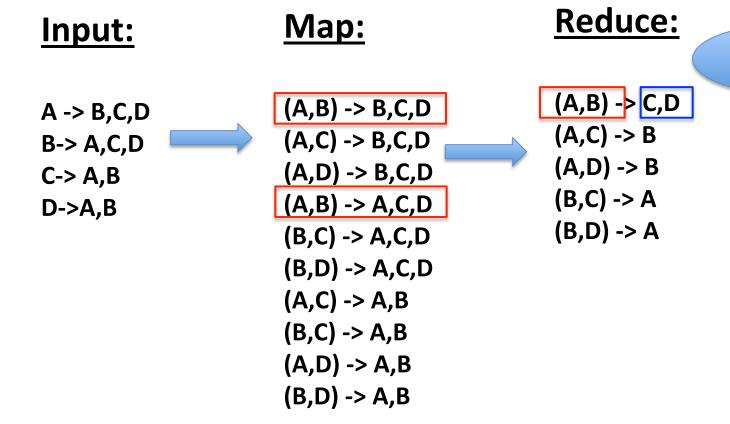
Mapper and Reducer of Common Friends

```
Map(key, value){
   // key: person_id
   // value: the list of friends of the person
   for each friend f_id in value:
       Emit(<person_id, f_id>, value);}
Reduce(key, list of values){
   // key: <friend pair>
   // list of values: a set of friend lists related with the friend pair
   for v1, v2 in values:
       common friends = v1 intersects v2;
   Emit(key, common_friends);}
```

- Problem 3: Find Common Friends
- Mapper and Reducer:
 - - { Emit (<friend pair>, Intersection of two (i.e, the one in friend pair) friend lists)}

Suggest Fiends ©

- Problem 3: Find Common Friends
- Mapper and Reducer:



JobTracker UI

Tools to see running/completed/failed jobs

disc01 Hadoop Map/Reduce Administration

State: RUNNING

Started: Mon Sep 08 12:26:02 EDT 2014

Version: 0.21.0, 985326

Compiled: Tue Aug 17 01:02:28 EDT 2010 by tomwhite from branches/branch-0.21

Identifier: 201409081224

Cluster Summary (Heap Size is 177.5 MB/1.74 GB)

| Queues | | Running Reduce Tasks | Total Submissions | Nodes | Occupied Map Slots | Occupied Reduce Slots | Reserved Map Slots | Reserved Reduce Slots | Map Slot Capacity | Reduce Slot Capacity | Avg. Slots/Node | Blacklisted Nodes | Excluded Nodes |
|--------|---|----------------------------|----------------------|-----------|--------------------------|-----------------------------|--------------------------|-----------------------------|----------------------|----------------------------|--------------------|----------------------|-------------------|
| 1 | 0 | 0 | 186 | <u>25</u> | 0 | 0 | 0 | 0 | 100 | 100 | 8.00 | <u>0</u> | <u>0</u> |

Filter (Jobid, Priority, User, Name)

Example: 'user:smith 3200' will filter by 'smith' only in the user field and '3200' in all fields

Running Jobs

none

JobTracker UI

Tools to see running/completed/failed jobs

Hadoop Job job_201409081224_0217 on History Viewer

User: czheng2

JobName: InvertedIndex

JobConf: hdfs://disc01.crc.nd.edu:8020/users/czheng2/.staging/job_201409081224_0217/job.xml

Submitted At: 25-Sep-2014 17:15:47 Launched At: 25-Sep-2014 17:15:47 (0sec) Finished At: 25-Sep-2014 17:16:06 (18sec)

Status: SUCCEEDED Analyse This Job

| Kind | Total Tasks(successful+failed+killed) | Successful tasks | Failed tasks | Killed tasks | Start Time | Finish Time |
|---------|---------------------------------------|------------------|--------------|--------------|----------------------|-----------------------------|
| Setup | 1 | 1 | <u>0</u> | <u>0</u> | 25-Sep-2014 17:15:48 | 25-Sep-2014 17:15:48 (0sec) |
| Мар | 2 | 2 | <u>0</u> | <u>0</u> | 25-Sep-2014 17:15:51 | 25-Sep-2014 17:15:52 (1sec) |
| Reduce | 1 | 1 | <u>0</u> | <u>0</u> | 25-Sep-2014 17:15:54 | 25-Sep-2014 17:16:00 (6sec) |
| Cleanup | 1 | 1 | <u>0</u> | <u>0</u> | 25-Sep-2014 17:16:03 | 25-Sep-2014 17:16:04 (1sec) |

| | Counter | Мар | Reduce | Total |
|--|--|---------|--------|---------|
| | FILE_BYTES_READ | 0 | 60,224 | 60,224 |
| FileSystemCounters | FILE_BYTES_WRITTEN | 60,294 | 60,224 | 120,518 |
| disc01.crc.nd.edu:8033/iobconf history.isp?logFile=file:/data/hadoop1/lo | LIDEC DVTEC DEAD as/history/done/iob 201409081224 0217 czhena2 | 232,242 | 0 | 232,242 |

JobTracker UI

Tools to see running/completed/failed jobs

Task Logs: 'attempt_201409081224_0217_m_000000_0'

2014-09-25 17:15:51,991 INFO org.apache.hadoop.mapred.MapTask: bufstart = 0; bufvoid = 104857600 2014-09-25 17:15:51,991 INFO org.apache.hadoop.mapred.MapTask: kvstart = 26214396; length = 6553600

2014-09-25 17:15:52,152 INFO org.apache.hadoop.mapred.MapTask: bufstart = 0; bufend = 38416; bufvoid = 104857600

2014-09-25 17:15:52,152 INFO org.apache.hadoop.mapred.MapTask: Starting flush of map output

2014-09-25 17:15:52,152 INFO org.apache.hadoop.mapred.MapTask: Spilling map output

2014-09-25 17:15:52,195 INFO org.apache.hadoop.mapred.MapTask: Finished spill 0

```
syslog logs

2014-09-25 17:15:51,802 INFO org.apache.hadoop.metrics.jvm.JvmMetrics: Initializing JVM Metrics with processName=MAP, sessionId=
2014-09-25 17:15:51,809 WARN org.apache.hadoop.conf.Configuration: user.name is deprecated. Instead, use mapreduce.job.user.name
2014-09-25 17:15:51,891 WARN org.apache.hadoop.conf.Configuration: mapred.task.id is deprecated. Instead, use mapreduce.task.attempt.id
2014-09-25 17:15:51,991 INFO org.apache.hadoop.mapred.MapTask: (EQUATOR) 0 kvi 26214396(104857584)
2014-09-25 17:15:51,991 INFO org.apache.hadoop.mapred.MapTask: mapreduce.task.io.sort.mb: 100
2014-09-25 17:15:51,991 INFO org.apache.hadoop.mapred.MapTask: soft limit at 83886080
```

2014-09-25 17:15:52,152 INFO org.apache.hadoop.mapred.MapTask: kvstart = 26214396(104857584); kvend = 26203324(104813296); length = 11073/6

Enjoy MR and Hadoop @

