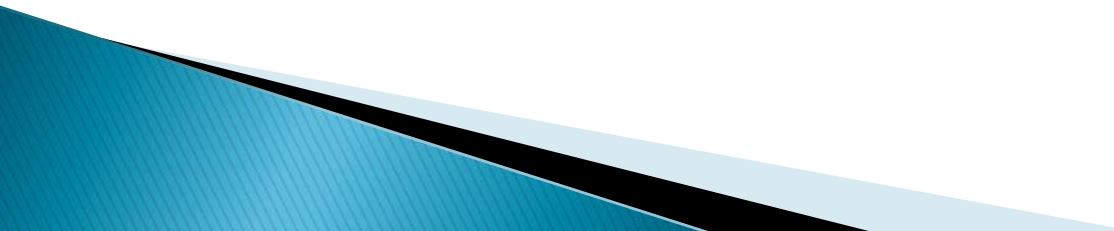


Hadoop MapReduce

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Some material adapted from slides by Christophe Bisciglia, Aaron Kimball, & Sierra Michels-Slettvet, Google Distributed Computing Seminar, 2007 (licensed under Creative Commons Attribution 3.0 License)

Outline

- ▶ MapReduce Overview
 - ▶ Hadoop–WordCount Demo
 - ▶ Hadoop–Blast Demo
 - ▶ Q&A
- 

MapReduce in Simple Terms

- ▶ Mapreduce – The Story of Tom
- ▶ By Saliya Ekanayake

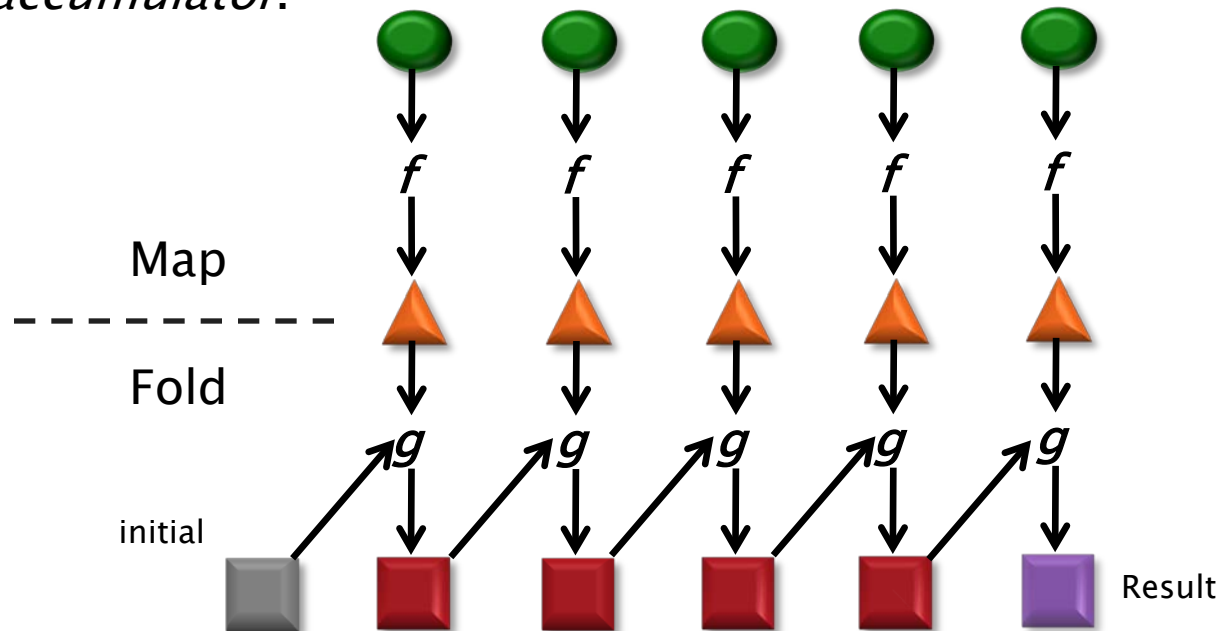
Mapreduce in Functional Programming

- ▶ Map

- Apply the same function f to all the elements of a list

- ▶ Fold

- Moves across a list, applying function g to each element plus an *accumulator*.

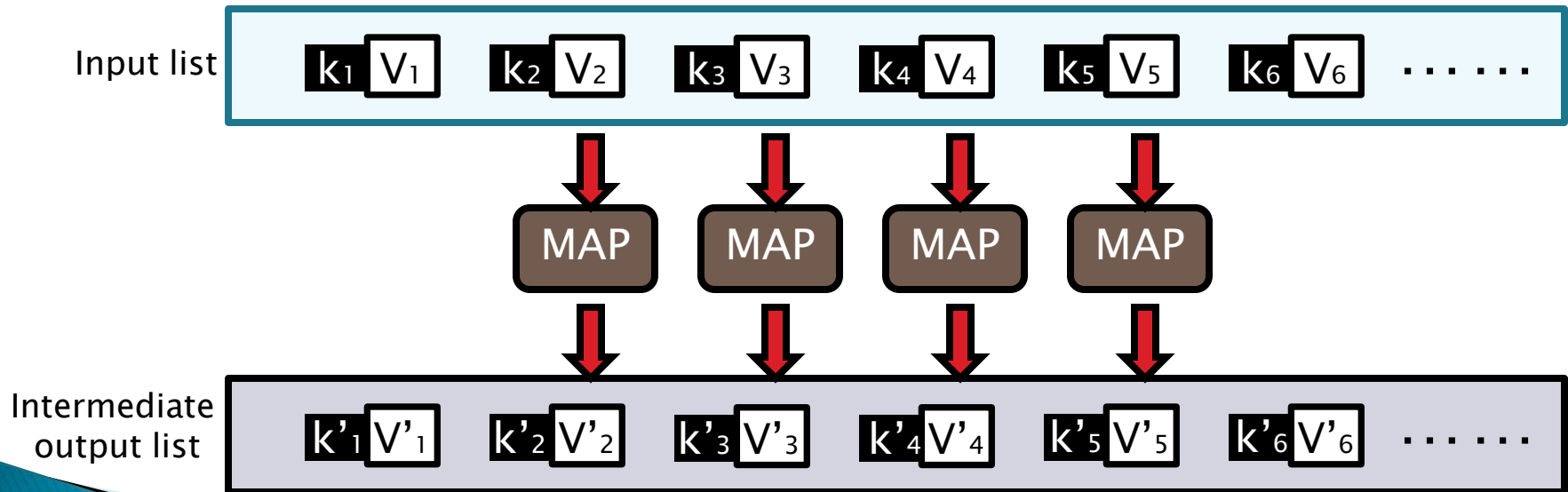


MapReduce Programming Model

- ▶ influenced by Functional Programming constructs
- ▶ Users implement interface of two functions:
 - `map (in_key, in_value) -> (out_key, intermediate_value) list`
 - `reduce (out_key, intermediate_value list) -> out_value list`

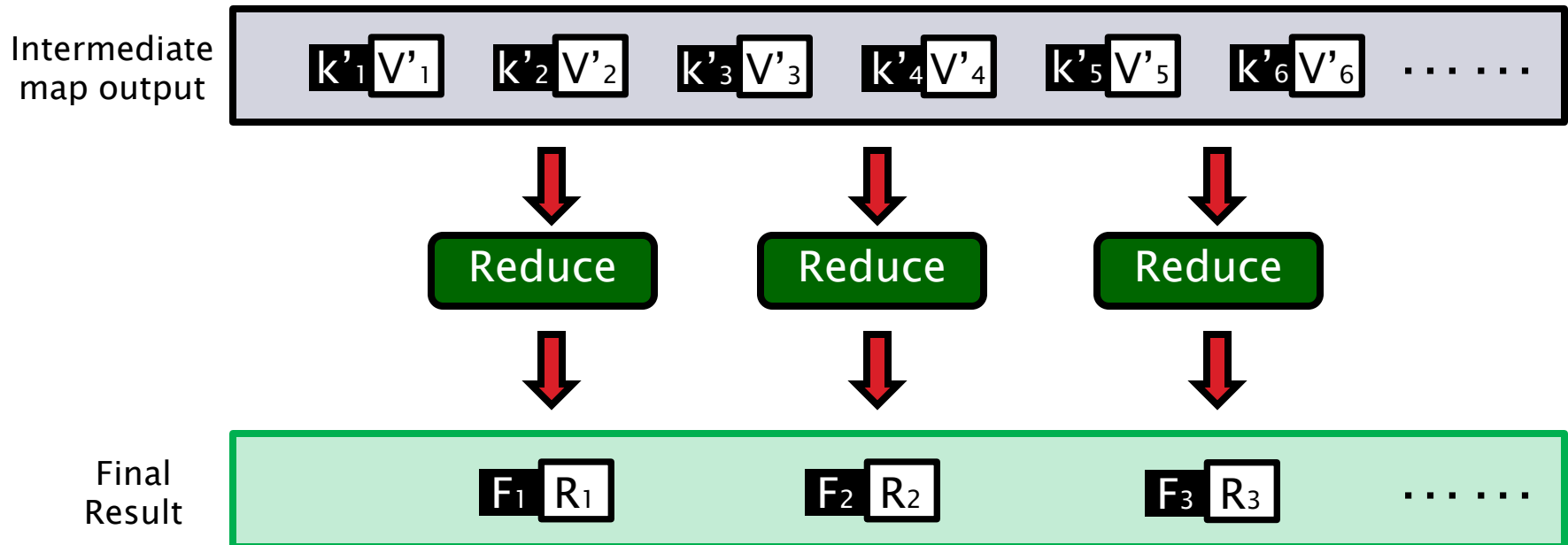
Map

- ▶ A list of data elements are passed, one at a time, to `map()` functions which transform each data element to an individual output data element.
- ▶ A `map()` produces one or more *intermediate* <key, values> pair(s) from the input list.



Reduce

- ▶ After map phase finish, those intermediate values with same output key are reduced into one or more *final values*



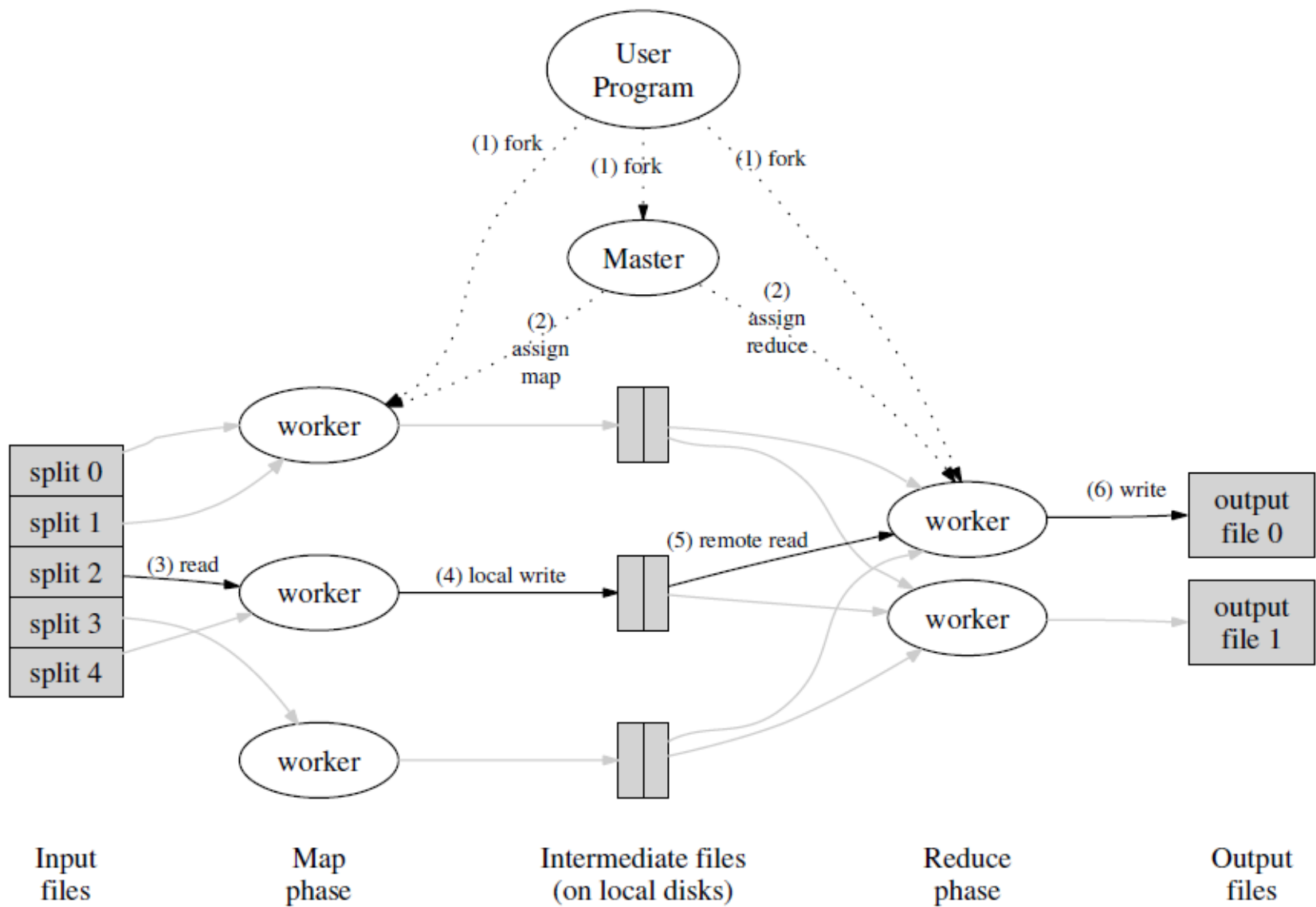
Parallelism

- ▶ map() functions run in parallel, creating different intermediate values from different input data elements
- ▶ reduce() functions also run in parallel, working with assigned output key
- ▶ All values are processed *independently*
- ▶ Reduce phase can't start until map phase is completely finished.

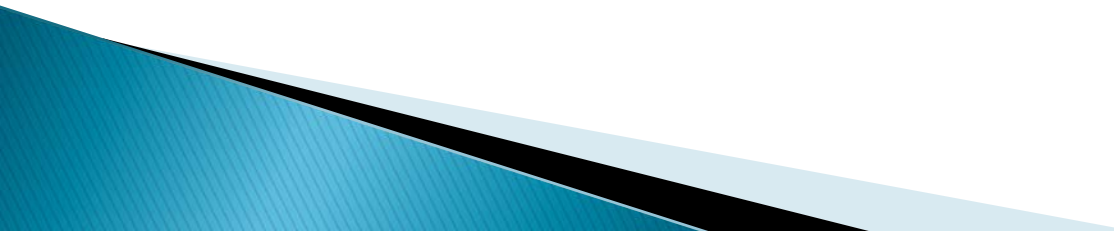
Apache Hadoop

- ▶ Open Source MapReduce framework implementation
- ▶ Widely used by
 - Amazon Elastic MapReduce
 - EBay
 - FaceBook
 - TeraSort in Yahoo!
 - ...
- ▶ Active community
- ▶ Many sub projects
 - HDFS
 - Pig Latin
 - Hbase
 - ...

Execution Overview



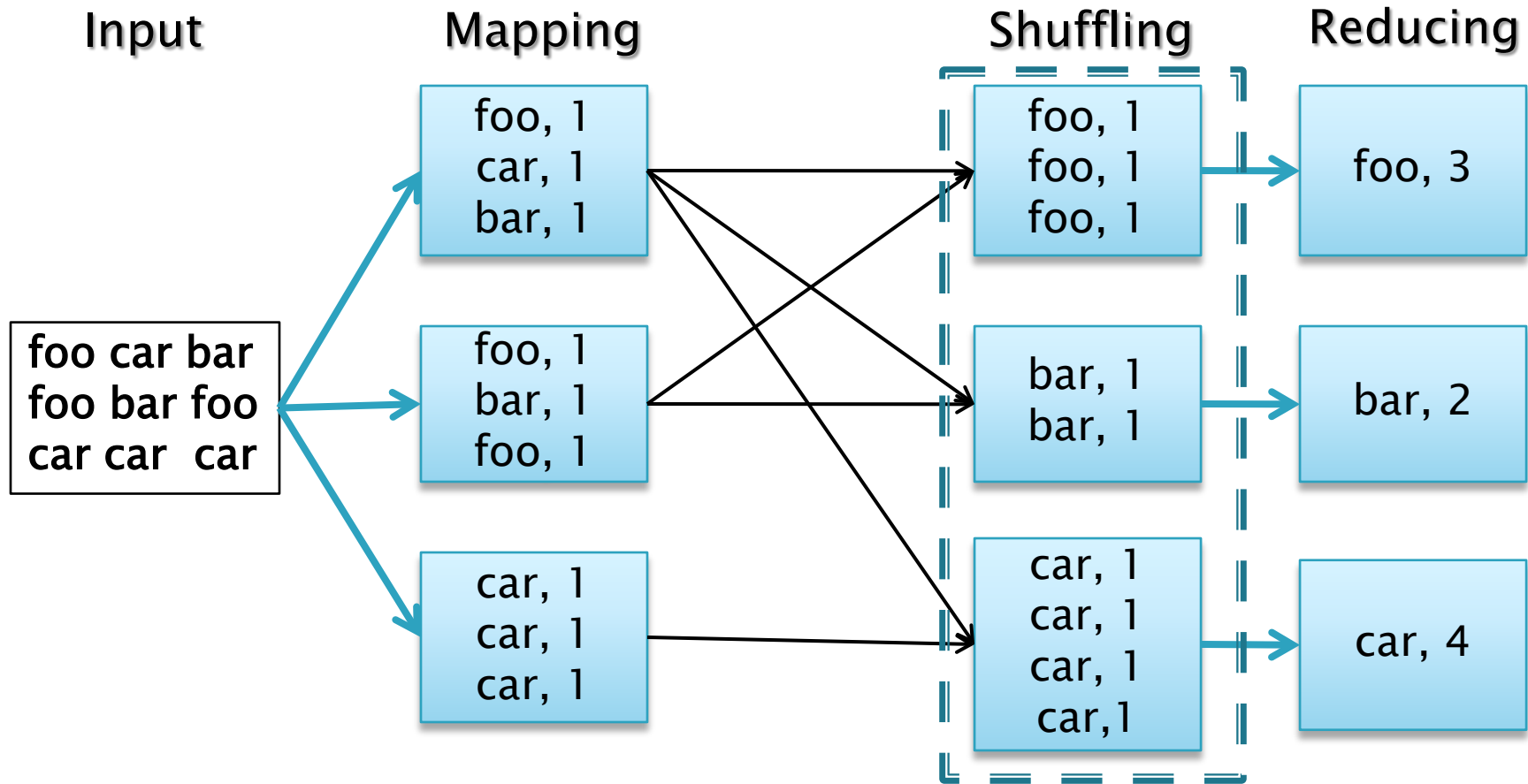
MapReduce in Apache Hadoop

- ▶ Autoparallelization
 - ▶ data distribution
 - ▶ Fault-tolerant
 - ▶ Status monitoring
 - ▶ Simple programming constructs
- 

Demo 1 – WordCount

- ▶ “Hello World” in MapReduce programming style
- ▶ Fits well with the MapReduce programming model
 - count occurrence of each word in given files
 - each file (may be split) run in parallel
 - need reduce phase to collect final result
- ▶ Three main parts:
 - Mapper
 - Reducer
 - Driver

Word Count



WordCount Map

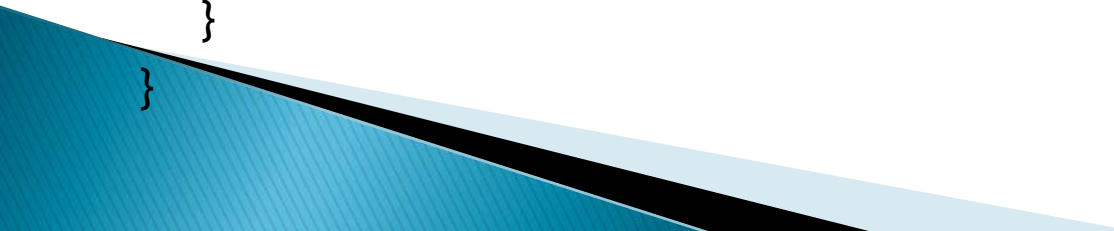
```
void Map (key, value) {  
    for each word x in value:  
        output.collect(x, 1);  
}
```

WordCount Map

```
public static class Map
    extends Mapper<LongWritable, Text, Text, IntWritable>{
    private final static IntWritable one = new IntWritable(1);
    private Text word = new Text();

    public void map(LongWritable key, Text value, Context context )
        throws IOException, InterruptedException {
        StringTokenizer itr = new StringTokenizer(value.toString());

        while (itr.hasMoreTokens()) {
            word.set(itr.nextToken());
            context.write(word, one);
        }
    }
}
```



WordCount Reduce

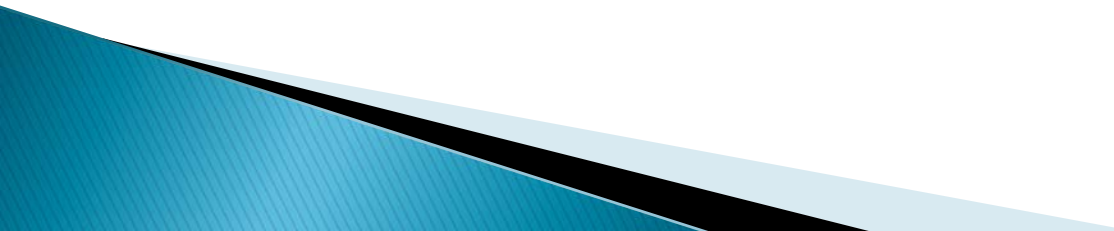
```
void Reduce (keyword, <list of value>){  
    for each x in <list of value>:  
        sum+=x;  
    output.collect(keyword, sum);  
}
```

WordCount Reduce

```
public static class Reduce
    extends Reducer<Text,IntWritable,Text,IntWritable> {
    private IntWritable result = new IntWritable();

    public void reduce(Text key, Iterable<IntWritable> values,
        Context context) throws .....{
        int sum = 0; // initialize the sum for each keyword
        for (IntWritable val : values) {
            sum += val.get();
        }
        result.set(sum);

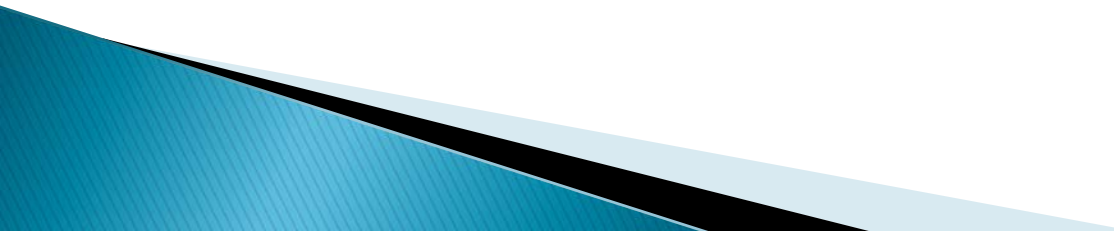
        context.write(key, result);
    }
}
```



Log in to FutureGrid

- ▶ `qsub -l`
 - Get a FutureGrid Work node
- ▶ `hostname`
 - e.g. `i51`
 - Public hostname name should be
 - `i51r.idp.iu.futuregrid.org`
 - `http://salsahpc.indiana.edu/tutorial/futuregrid\_access.html`

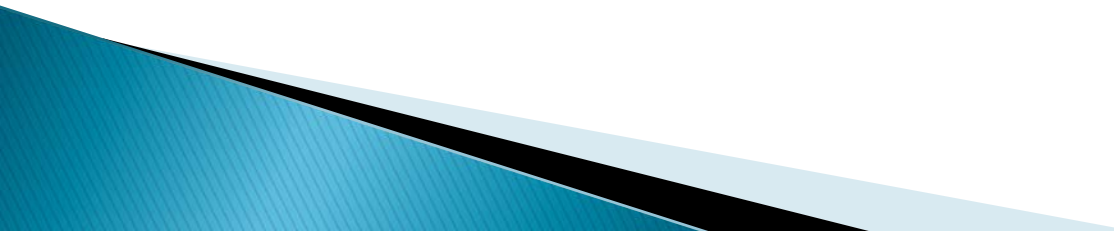
Building WordCount program

- ▶ `cd Hadoop-WordCount`
 - ▶ `cat WordCount.java`
 - ▶ `cat build.sh`
 - ▶ `./build.sh`
 - ▶ `ls -l`
- 

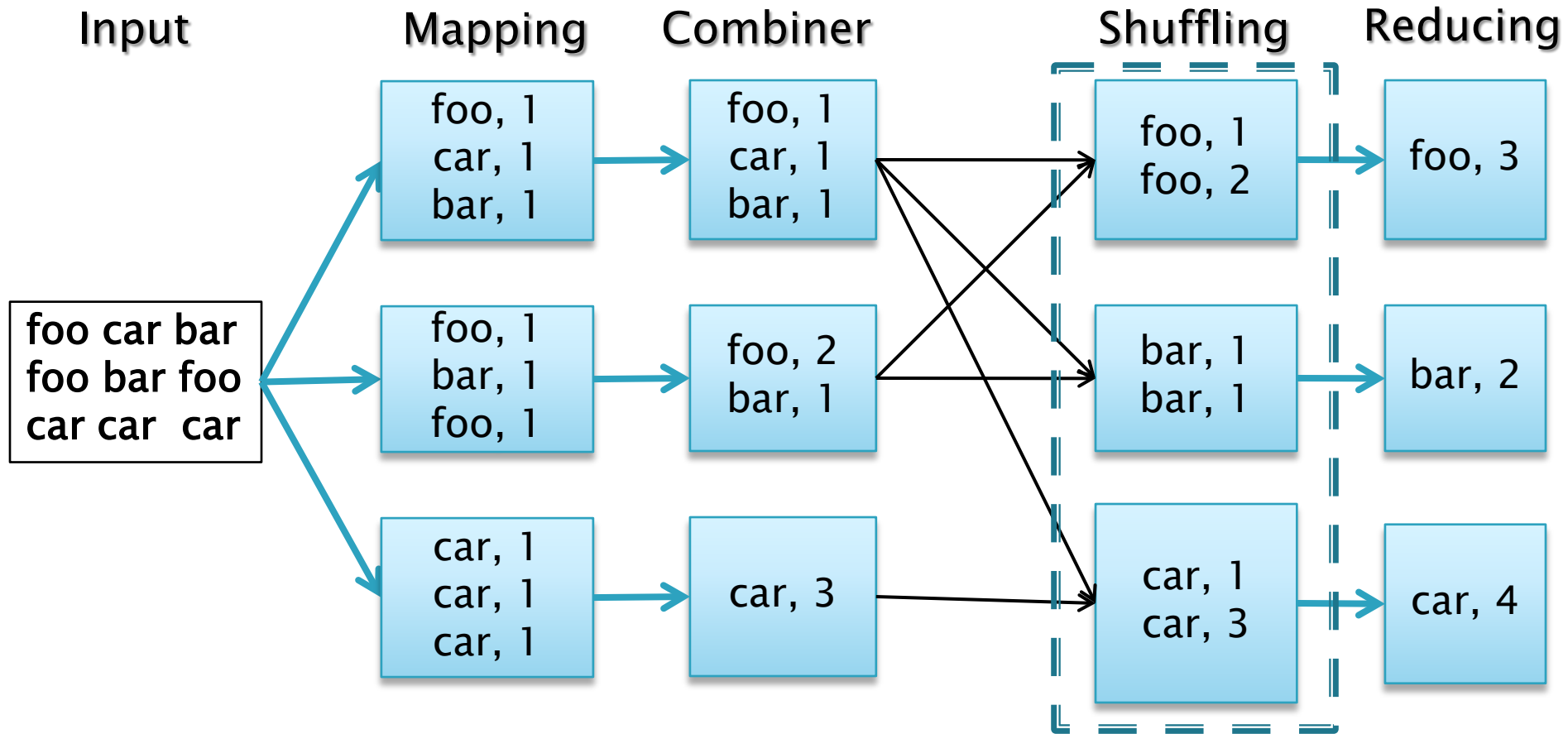
Running WordCount using Hadoop standalone

- ▶ `cd ~/hadoop-0.20.2-standalone/bin`
- ▶ `cp ~/Hadoop-WordCount/wordcount.jar
~/hadoop-0.20.2-standalone/bin`
- ▶ `./hadoop jar wordcount.jar WordCount ~/Hadoop-
WordCount/input ~/Hadoop-WordCount/output`
 - Simulate HDFS using the local directories
- ▶ `cd ~/Hadoop-WordCount/`
- ▶ `cat output/part-r-00000`

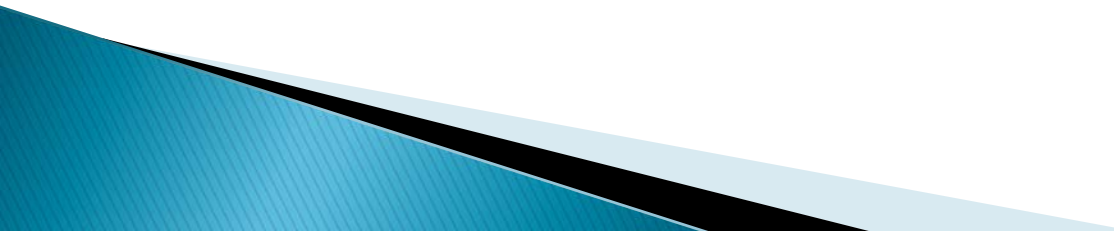
Optimizations

- ▶ Reduce phase can only be activated until all map tasks finish
 - It will be a waste if there is a extremely long map task
 - ▶ “Combiner” functions can run on same machine as a mapper
 - ▶ Causes a mini-reduce phase to occur before the real reduce phase, to save bandwidth
- 

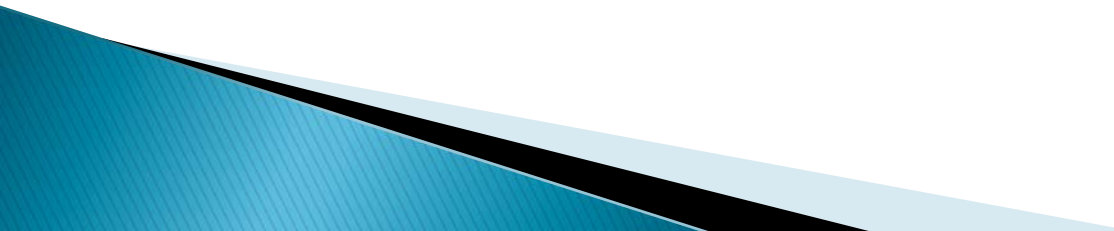
Word Count with Combiner



Terminology

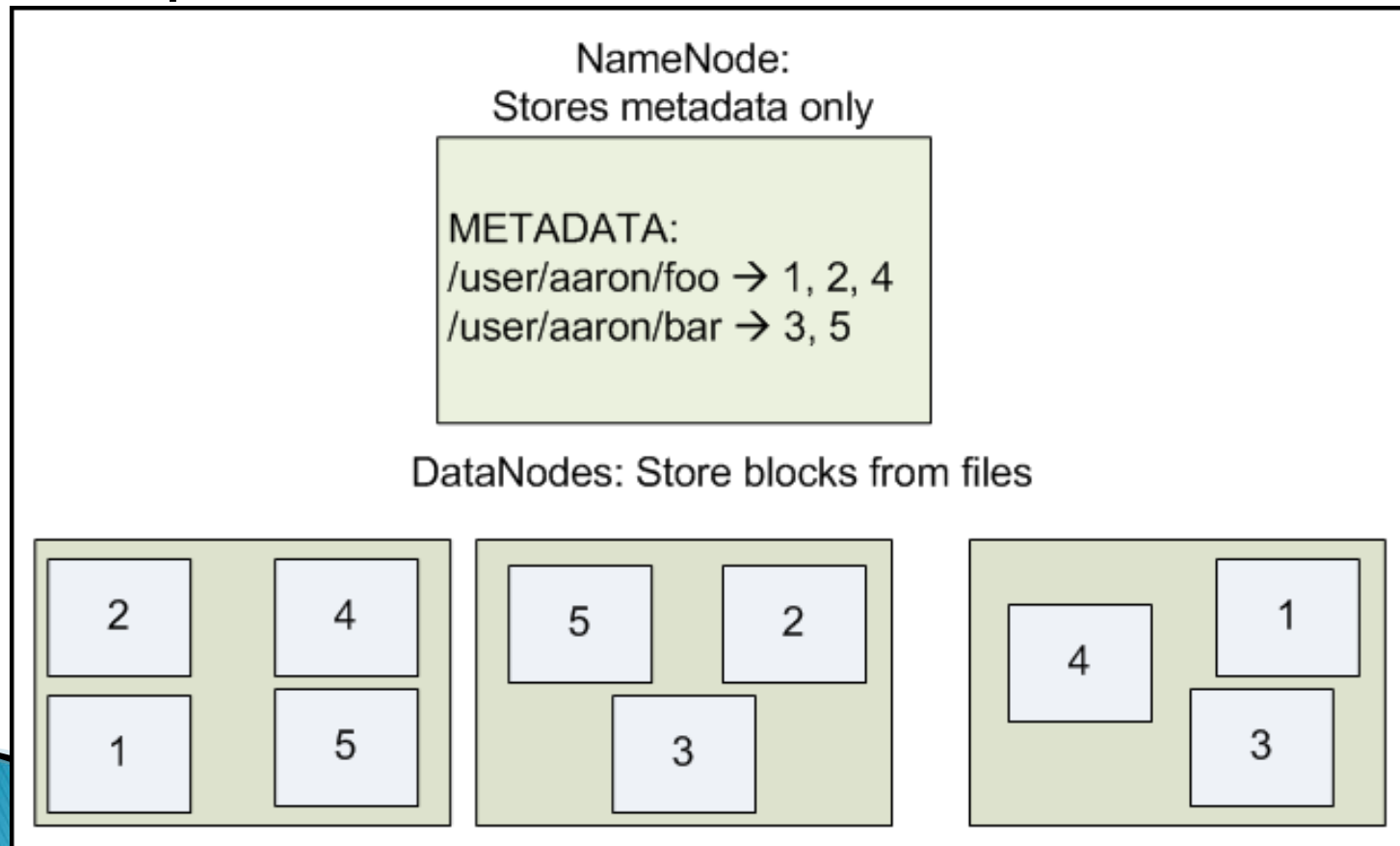
- ▶ Job
 - ▶ Jobtracker
 - ▶ Task
 - ▶ TaskTracker
 - ▶ Shard
- 

HDFS

- ▶ Based on Google File System
 - ▶ Run on commodity low-cost hardware
 - ▶ Fault tolerance
 - Redundant storage
 - ▶ High throughput access
 - ▶ Suitable for large data sets
 - ▶ Large capacity
 - ▶ Integration with MapReduce
- 

Architecture

- ▶ Single NameNode
- ▶ Multiple Data Nodes



Configuring HDFS

- ▶ **dfs.name.dir**
 - Dir in the namenode to store metadata
- ▶ **dfs.data.dir**
 - Dir in data nodes to store the data blocks
 - Must be in a local disk partition

conf/hdfs-site.xml

```
<property>
  <name>dfs.name.dir</name>
  <value>/tmp/hadoop-test/name</value>
</property>
<property>
  <name>dfs.data.dir</name>
  <value>/tmp/hadoop-test/data</value>
</property>
```

Configuring HDFS

- ▶ **fs.default.name**
 - URI of the namenode

conf/core-site.xml

```
<property>
```

```
  <name>fs.default.name</name>
```

```
  <value>hdfs://149.79.89.113:9000/</value>
```

```
</property>
```

- ▶ **conf/masters**
 - ip address of the master node
- ▶ **conf/slaves**
 - ip address of the slave nodes
- ▶ Should have password-less SSH access to all the nodes

HDFS Cluster

- ▶ Starting HDFS
 - `cd ~/hadoop-0.20.2/bin`
 - `./hadoop namenode -format`
 - `./start-dfs.sh`
- ▶ NameNode logs
 - `cd ~/hadoop-0.20.2`
 - `logs/hadoop-<username>-namenode-<nodeid>.log`
- ▶ Monitoring web interface
 - `http://<public_ip>:50070`

HDFS Commands

- ▶ `./bin/hadoop fs -[command]`
 - `put`
 - `get`
 - `ls`
 - `cp`
 - `mkdir`
 - `rm`
- ▶ http://hadoop.apache.org/common/docs/current/hdfs_shell.html
- ▶ Programmatic API

Configuring Hadoop MapReduce

- ▶ `mapred.job.tracker`
- ▶ `mapred.local.dir`
- ▶ `mapred.tasktracker.map.tasks.maximum`

`conf/mapred-site.xml`

```
<property>
  <name>mapred.job.tracker</name>
  <value>149.79.89.113:9001</value>
</property>
<property>
  <name>mapred.local.dir</name>
  <value>/tmp/hadoop-test/local</value>
</property>
<property>
  <name>mapred.tasktracker.map.tasks.maximum</name>
  <value>8</value>
</property>
```

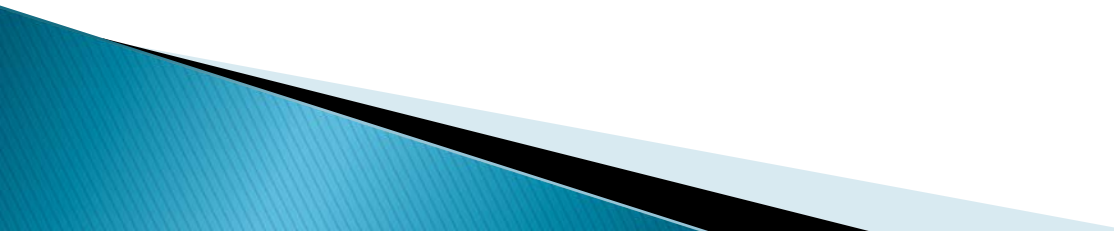
Starting Hadoop MapReduce Cluster

- ▶ Starting Hadoop MapReduce
 - `cd ~/hadoop-0.20.2/bin`
 - `./start-mapred.sh`
- ▶ JobTracker logs
 - `cd ~/hadoop-0.20.2`
 - `logs/hadoop-<username>-tasktracker<nodeid>.log`
- ▶ Monitoring web interface
 - `http://<public_ip>:50030`

Running WordCount on Distributed Hadoop Cluster

- ▶ Upload input to HDFS
 - `cd ~/hadoop-0.20.2/bin`
 - `./hadoop fs -put ~/Hadoop-WordCount/input/input`
 - `./hadoop fs -ls input`
- ▶ Run word count
 - `./hadoop jar wordcount.jar WordCount input output`
 - `./hadoop fs -ls output`
 - `./hadoop fs -cat output/*`

Demo 2 : Hadoop-Blast

- ▶ An advance MapReduce implementation
 - ▶ Use Blast (*Basic Local Alignment Search Tool*), a well-known bioinformatics application written in C/C++
 - ▶ Utilize the Computing Capability of Hadoop
 - ▶ Use Distributed Cache for the Blast Program and Database
 - ▶ No Reducer in practice
- 

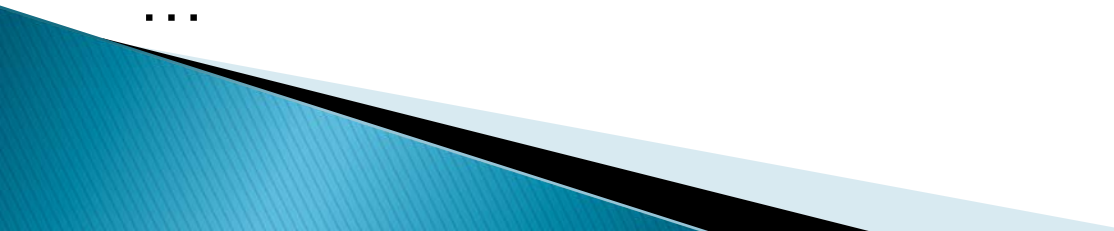
Map

```
public void map(String key, String value, Context context)
    throws IOException, InterruptedException {
    ...
    // download the file from HDFS
    ...
    fs.copyToLocalFile(inputFilePath, new Path(localInputFile)); // Prepare the arguments to the executable
    ...
    execCommand = this.localBlastProgram + File.separator + execName + " " + execCommand + " -db " + this.localDB;

    //Create the external process
    Process p = Runtime.getRuntime().exec(execCommand);
    ...
    p.waitFor();
    //Upload the results to HDFS
    ...
    fs.copyFromLocalFile(new Path(outFile),outputFileName);
    ...
} // end of overriding the map
```

Driver

```
...  
// using distributed cache  
DistributedCache.addCacheArchive(new URI(BlastPr  
ogramAndDB), jc);  
...  
// other parts  
...  
// input and output format  
job.setInputFormatClass(DataFileInputFormat.class  
);  
job.setOutputFormatClass(SequenceFileOutputFor  
mat.class);  
...
```



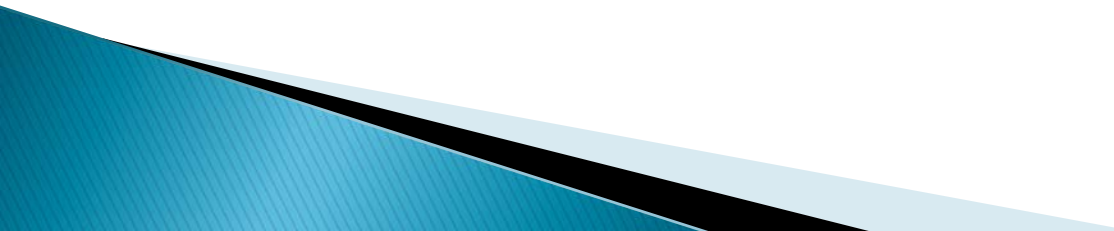
Preparation

- ▶ Make sure the HDFS and Map-Reduce daemon start correctly
- ▶ Put Blast queries on HDFS
 - `cd ~/hadoop-0.20.2/bin`
 - `./hadoop fs -put ~/hadoop-0.20.2/apps/Hadoop-Blast/input HDFS_blast_input`
 - `./hadoop fs -ls HDFS_blast_input`
- ▶ Copy Blast Program and DB to HDFS
 - `./hadoop fs -copyFromLocal $BLAST_HOME/BlastProgramAndDB.tar.gz BlastProgramAndDB.tar.gz`
 - `./hadoop fs -ls BlastProgramAndDB.tar.gz`

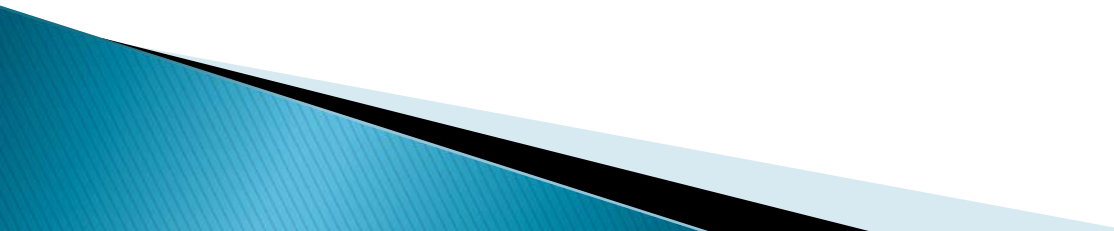
Execution & result

- ▶ Run the Hadoop-Blast program
 - `cd ~/hadoop-0.20.2/bin`
 - `./hadoop jar ~/hadoop-0.20.2/apps/Hadoop-Blast/executable/blast-hadoop.jar BlastProgramAndDB.tar.gz bin/blastx /tmp/hadoop-test/ db nr HDFS_blast_input HDFS_blast_output '-query #_INPUTFILE_# -outfmt 6 -seg no -out #_OUTPUTFILE_#'`
- ▶ Check the Result
 - `cd ~/hadoop-0.20.2/bin`
 - `./hadoop fs -ls HDFS_blast_output`
 - `./hadoop fs -cat HDFS_blast_output/pre_1.fa`

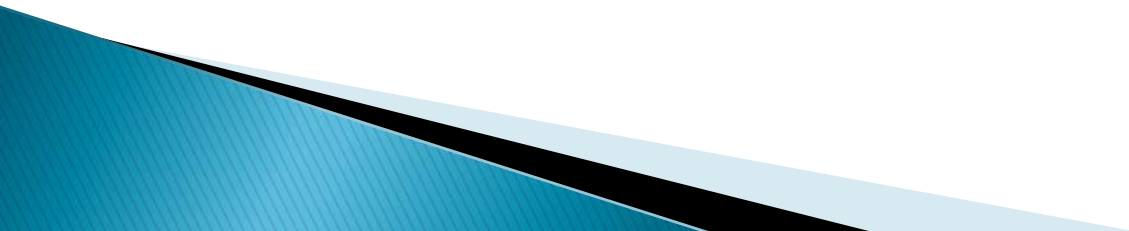
Performance Tips at the end

- ▶ More Map tasks
 - Much more than the Map task capacity of the cluster
 - ▶ Number of reduce tasks
 - 95% of the Reduce task capacity
 - Interleave computation with communication
 - ▶ Data locality
 - ▶ Map/Reduce task capacity of a worker
 - ▶ Logs
 - Don't store in NFS. They grow pretty fast.
- 

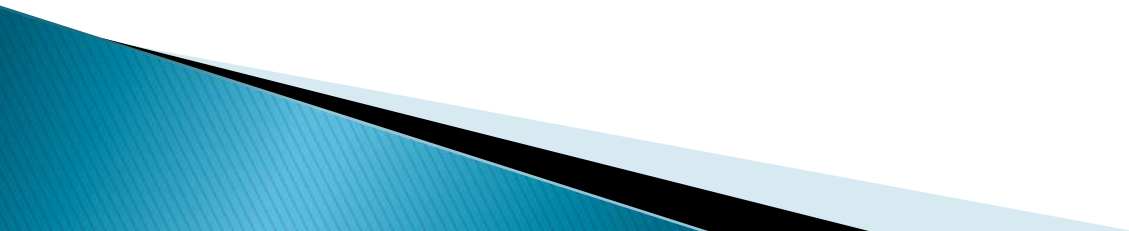
References

- ▶ [Big Data for Science Tutorial](#)
 - ▶ [Yahoo! Hadoop Tutorial](#)
 - ▶ [Hadoop wiki](#)
 - ▶ [Apache Hadoop Map/Reduce Tutorial](#)
 - ▶ [Google: Cluster Computing and MapReduce](#)
- 

Questions?



Thank You!!



WordCount

Input

Mapping

Shuffling

Sorting

Reducing

foo car bar
foo bar foo
car car car

foo, 1
car, 1
bar, 1

foo, 1
bar, 1
foo, 1

car, 1
car, 1
car, 1

foo, 1
car, 1
bar, 1
foo, 1
bar, 1
foo, 1
car, 1
car, 1
car, 1

bar, <1, 1>
car, <1, 1, 1, 1>
foo, <1, 1, 1>

bar, 2
car, 4
foo, 3

Execute 3 times

Framework level

MapReduce Fault Tolerance

- ▶ MasterNode handles failures
 - detects worker failures
 - rerun completed & in-progress map() tasks
 - rerun in-progress reduce() tasks
 - If particular input key/values cause crashes in map phase, and skips those values on re-execution.
 - Partly output

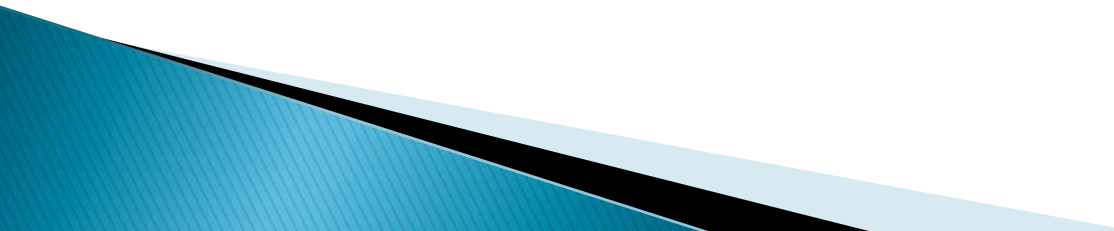
MapReduce Locality

- ▶ Master program tries to have map() tasks on same machine as physical file data, or at least on same rack
- ▶ map task inputs are spilt into blocks with default sizes as 64MB.

HDFS Assumptions

- ▶ High component failure rates
 - Inexpensive commodity components fail all the time
- ▶ “Modest” number of HUGE files
 - Just a few million
 - Each is 100MB or larger; multi-GB files typical
- ▶ Files are write-once, read many times
- ▶ Large streaming reads
- ▶ High sustained throughput favored over low latency

HDFS Design Decisions

- ▶ Files stored as blocks
 - Blocks stored across cluster
 - Default block size 64MB
 - ▶ Reliability through replication
 - *Default is 3*
 - ▶ Single master (NameNode) to coordinate access, keep metadata
 - Simple centralized management
 - ▶ No data caching
 - Little benefit due to large data sets, streaming reads
 - ▶ Focused on distributed applications
- 

Advanced Hadoop

- ▶ Input formats
- ▶ Output formats
- ▶ Data types
- ▶ Distributed Cache

Input Formats

- ▶ Defines how to break the inputs
- ▶ Provides RecordReader objects to read the files
 - Read data and converts to <key,value> pairs
- ▶ It's possible to write custom input formats
- ▶ Hadoop Built-in input formats
 - TextInputFormat
 - Treats each line as value
 - KeyValueInputFormat
 - Each line as key-value (tab delimited)
 - SequenceFileInputFormat
 - Hadoop specific binary representation

Output Formats

- ▶ How the result `<key,value>` pairs are written in to files
- ▶ Possible to write custom output formats
- ▶ Hadoop default OutputFormats
 - TextOutputFormat
 - `<key TAB value>`
 - SequenceFileOutputFormat
 - Hadoop specific binary representation

Data Types

- ▶ Implement Writable interface
 - Text, IntWritable, LongWritable, FloatWritable, BooleanWritable, etc...
- ▶ We can write custom types which Hadoop will serialize/deserialize as long as following is implemented.

```
public interface Writable {  
    void readFields(DataInput in);  
    void write(DataOutput out);  
}
```

Distributed Cache *

- ▶ Distributes a given set of data to all the mappers
 - Common data files, libraries, etc..
 - Similar to a broadcast
 - Available in the local disk
 - Automatically extract compressed archives
- ▶ First upload the files to HDFS
- ▶ `DistributedCache.addCacheFile()`
- ▶ `DistributedCache.getLocalCacheFiles()`