

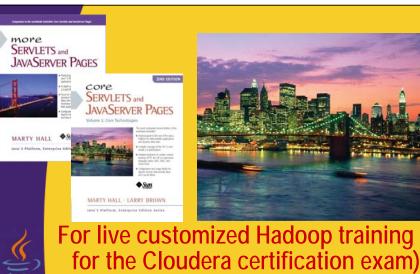
Map Reduce 2.0 Input and Output

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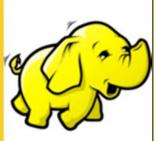
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Agenda

- MapReduce Theory
- Types of Keys and Values
- Input and Output Formats
- Discuss Anatomy of
 - Mappers
 - Reducers
 - Combiners
 - Partitioners

4

MapReduce Theory

- Map and Reduce functions produce input and output
 - Input and output can range from Text to Complex data structures
 - Specified via Job's configuration
 - Relatively easy to implement your own
- Generally we can treat the flow as

```
map: (K1,V1) \rightarrow list (K2,V2)
reduce: (K2,list(V2)) \rightarrow list (K3,V3)
```

- Reduce input types are the same as map output types

5

Map Reduce Flow of Data map: $(K1,V1) \rightarrow list(K2,V2)$ reduce: $(K2, list(V2)) \rightarrow list(K3, V3)$ Mapper Map Data **Split Task** Output Node #1 Reduce Reduce **Task** Output Node #X Mapper Map Data **Split** Task Output Node #N

Key and Value Types

- Utilizes Hadoop's serialization mechanism for writing data in and out of network, database or files
 - Optimized for network serialization
 - A set of basic types is provided
 - Easy to implement your own
- Extends Writable interface
 - Framework's serialization mechanisms
 - Defines how to read and write fields
 - org.apache.hadoop.io package

Key and Value Types

- Keys must implement WritableComparable interface
 - Extends Writable and java.lang.Comparable<T>
 - Required because keys are sorted prior reduce phase
- Hadoop is shipped with many default implementations of WritableComparable<T>
 - Wrappers for primitives (String, Integer, etc...)
 - Or you can implement your own

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WritableComparable<T> Implementations

Hadoop's Class	Explanation
BooleanWritable	Boolean implementation
BytesWritable	Bytes implementation
DoubleWritable	Double implementation
FloatWritable	Float implementation
IntWritable	Int implementation
LongWritable	Long implementation
NullWritable	Writable with no data

Implement Custom WritableComparable<T>

Implement 3 methods

- write(DataOutput)
 - Serialize your attributes
- readFields(DataInput)
 - De-Serialize your attributes
- compareTo(T)
 - · Identify how to order your objects
 - If your custom object is used as the key it will be sorted prior to reduce phase

10

BlogWritable – Implemenation of WritableComparable<T>

11

BlogWritable – Implemenation of WritableComparable<T>

```
@Override
public void readFields(DataInput input) throws IOException {
    author = input.readUTF();
    content = input.readUTF();
                                         1. How the data is read
}
@Override
public void write(DataOutput output) throws IOException {
    output.writeUTF(author);
    output.writeUTF(content);
                                         2. How to write data
                                                3. How to order
@Override
                                                BlogWritables
public int compareTo(BlogWritable other) {
     return author.compareTo(other.author); 
}
```

Mapper

- Extend Mapper class
 - Mapper<KeyIn, ValueIn, KeyOut, ValueOut>
- Simple life-cycle
 - 1. The framework first calls setup(Context)
 - 2. for each key/value pair in the split:
 - map(Key, Value, Context)
 - 3. Finally cleanup(Context) is called

InputSplit

- Splits are a set of logically arranged records
 - A set of lines in a file
 - A set of rows in a database table
- Each instance of mapper will process a single split
 - Map instance processes one record at a time
 - map(k,v) is called for each record
- Splits are implemented by extending InputSplit class

14

InputSplit

- Framework provides many options for InputSplit implementations
 - Hadoop's FileSplit
 - HBase's TableSplit
- Don't usually need to deal with splits directly
 - InputFormat's responsibility

InputFormat

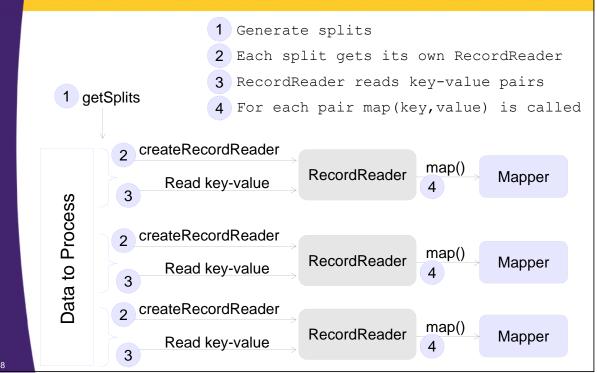
- Specification for reading data
- Creates Input Splits
 - Breaks up work into chunks
- Specifies how to read each split
 - Divides splits into records
 - Provides an implementation of RecordReader

16

Framework's Usage of InputFormat Implementation

- 1. Calculate splits by calling InputFormat.getSplits
- 2. For each split schedule a map task
 - Distributed between the cluster
 - Each Map executes in its own JVM
- 3. For each Mapper instance a reader is retrieved by InputFormat.createRecordReader
 - Takes InputSplit instance as a parameter
- 4. RecordReader generates key-value pairs
- map() method is called for each key-value pair

Framework's Usage of InputFormat Implementation



Hadoop's InputFormats

- Hadoop eco-system is packaged with many InputFormats
 - TextInputFormat
 - NLineInputFormat
 - DBInputFormat
 - TableInputFormat (HBASE)
 - StreamInputFormat
 - SequenceFileInputFormat
 - Etc...
- Configure on a Job object
 - job.setInputFormatClass(XXXInputFormat.class);

TextInputFormat

- Plaint Text Input
- Default format

Split: Single HDFS block (can be configured)

Record: Single line of text; linefeed or carriage-return used

to locate end of line

Key: LongWritable - Position in the file

Value: Text - line of text

20

NLineInputFormat

 Same as TextInputFormat but splits equal to configured N lines

Split: N lines; configured via *mapred.line.input.format* or

NLineInputFormat.setNumLinesPerSplit(job, 100);

Record: Single line of text

Key: LongWritable - Position in the file

Value: Text - line of text

^{**} Please see StartsWithCountJob for sample usage

^{**} Please see StartsWithCountJob_NLineInput for sample usage

Running TextInputFormat vs. NLineInputFormat

- Two separate runs StartsWithCountJob, one with TextInputFormatconfigured, next with NLineInputFormat configured
- Input is /training/playArea/hamlet.txt
 - 5159 lines
 - 206.3k

<pre>job.setInputForma</pre>	tClass(TextI	nputForm	at.class)	;	
Job ID *	Name \$	State M	ap Progress M	laps Total	
job_1338595987451_0003	StartsWithCount	RUNNING	1		
Showing 1 to 1 of 1 entries					
<pre>job.setInputFormatClass(NLineInputFormat.class); NLineInputFormat.setNumLinesPerSplit(job, 100);</pre>					
Job ID	Name	State 0	Map Progress	Maps Maps	Total
job_1338595987451_0002	StartsWithCount	RUNNING		52	
Showing 1 to 1 of 1 entries					

TableInputFormat

- Converts data in HTable to format consumable to MapReduce
- Mapper must accept proper key/values

Split: Rows in one HBase Region (provided Scan may

narrow down the result)

Record: Row, returned columns are controlled by a

provided scan

Key: ImmutableBytesWritable

Value: Result (HBase class)

23

StartCountJob – Input from HBase

Let's re-write StartWithCountJob to read input from HBase table

- 'HBaseSamples' table, 'count:word' family-column qualifier

\$ hbase shell

```
hbase(main):005:0> scan 'HBaseSamples', {COLUMNS=>'count:word'}
ROW
          COLUMN+CELL
           column=count:word, timestamp=1338605322765, value=Elephant
count001
           column=count:word, timestamp=1338605412699, value=count
count002
           column=count:word, timestamp=1338605412729, value=Updating
count003
count004
           column=count:word, timestamp=1338605412757, value=all
count005
           column=count:word, timestamp=1338605412780, value=regions
count006
           column=count:word, timestamp=1338605412809, value=with
count007
           column=count:word, timestamp=1338605412835, value=the
count008
           column=count:word, timestamp=1338605412856, value=new
count009
           column=count:word, timestamp=1338605412888, value=updated
count010
           column=count:word, timestamp=1338605412910, value=Done
           column=count:word, timestamp=1338605412933, value=seconds
count011
count012
           column=count:word, timestamp=1338605414526, value=row
12 row(s) in 0.1810 seconds
```

StartCountJob – Input from HBase

1. Re-configure Job to use HBase as input

- Read from table 'HBaseSamples' and column 'count:word'
- Construct new Job 'StartWithCountJob_HBaseInput'
- Configure Job to use new Mapper
- New mapper now has to accept HBase Writables
 - ImmutableBytesWritable for key
 - Result for value
- Keep reducer and combiner the same

2. Implement a new Mapper

- Grab the value from Result and write-out Text and IntWritable
- Output is the same as in the original StartWithCountJob

1: Re-configure Job to use HBase as Input

1: Re-configure Job to use HBase as Input

1: Re-configure Job to use HBase as Input

```
Keep output the same - same
. . .
                                      reducer, combiner an output keys
     // configure mapper and reducer
       job.setCombinerClass(StartsWithCountReducer.class);
       job.setReducerClass(StartsWithCountReducer.class);
       // configure output
       TextOutputFormat.setOutputPath(job, new Path(args[0]));
       job.setOutputFormatClass(TextOutputFormat.class);
       job.setOutputKeyClass(Text.class);
       job.setOutputValueClass(IntWritable.class);
      return job.waitForCompletion(true) ? 0 : 1;
 public static void main(String[] args) throws Exception {
       int exitCode = ToolRunner.run(new
                     StartWithCountJob HBaseInput(), args);
       System.exit(exitCode);
```

2: Implement a New Mapper

```
public class StartsWithCountMapper HBase
              extends TableMapper<Text, IntWritable> {
Extends TableMapper which assumes types for input key and value
 private final static IntWritable countOne = new IntWritable(1);
 private final Text reusableText = new Text();
  @Override
  protected void map(ImmutableBytesWritable key,
              Result value, Context context)
              throws IOException, InterruptedException {
       byte[] bytes = value.getValue(toBytes(FAMILY),
                                             toBytes (COLUMN));
       String str = Bytes.toString(bytes);
                                                  Retrieve value from
                                                  input column
       reusableText.set(str.substring(0, 1));
       context.write(reusableText, countOne);
```

Run StartWithCountJob_HBaseInput

\$ yarn jar \$PLAY_AREA/HadoopSamples.jar mr.wordcount.**StartWithCountJob_HBaseInput** /training/playArea/wordCount

30

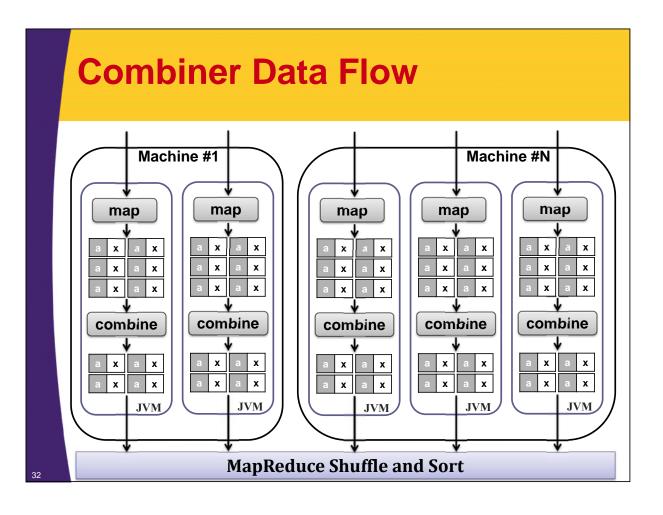
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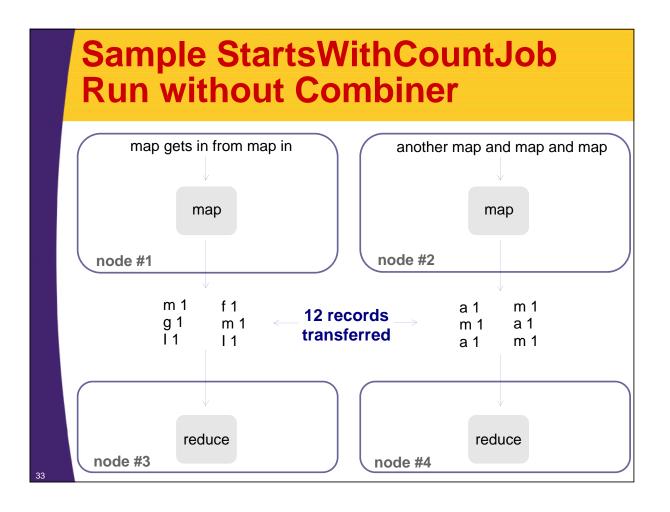
Combiner

- Runs on output of map function
- Produces output for reduce function

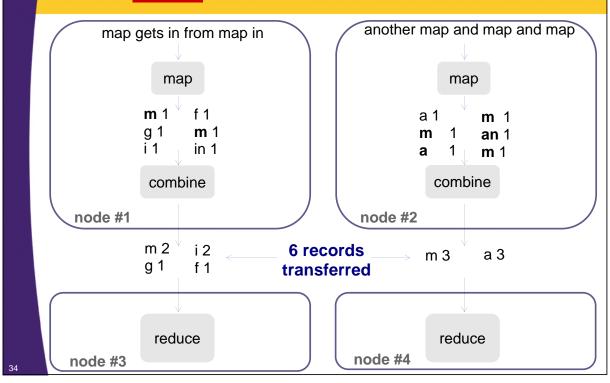
```
map: (K1,V1) \rightarrow list (K2,V2)
combine: (K2,list(V2)) \rightarrow list (K2,V2)
reduce: (K2,list(V2)) \rightarrow list (K3,V3)
```

- Optimization to reduce bandwidth
 - NO guarantees on being called
 - Maybe only applied to a sub-set of map outputs
- Often is the same class as Reducer
- Each combine processes output from a single split





Sample StartsWithCountJob Run with Combiner



Specify Combiner Function

- To implement Combiner extend Reducer class
- Set combiner on Job class
 - job.setCombinerClass(StartsWithCountReducer.class);

Reducer

Extend Reducer class

- Reducer<KeyIn, ValueIn, KeyOut, ValueOut>
- KeyIn and ValueIn types must match output types of mapper

Receives input from mappers' output

- Sorted on key
- Grouped on key of key-values produced by mappers
- Input is directed by Partitioner implementation

Simple life-cycle – similar to Mapper

- The framework first calls setup(Context)
- for each key \rightarrow list(value) calls
 - reduce(Key, Values, Context)
- Finally cleanup(Context) is called

36

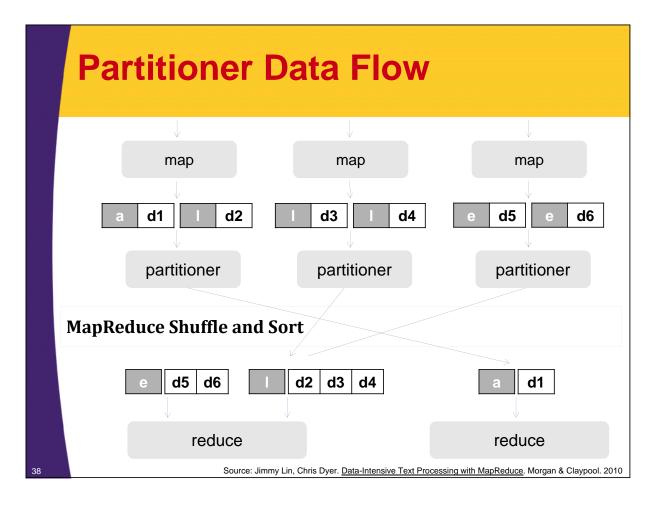
Reducer

Can configure more than 1 reducer

- job.setNumReduceTasks(10);
- mapreduce.job.reduces property
 - job.getConfiguration().setInt("mapreduce.job.reduces", 10)

Partitioner implementation directs key-value pairs to the proper reducer task

- A partition is processed by a reduce task
 - # of partitions = # or reduce tasks
- Default strategy is to hash key to determine partition implemented by HashPartitioner<K, V>



HashPartitioner

```
public class HashPartitioner<K, V> extends Partitioner<K, V> {
   public int getPartition(K key, V value, int numReduceTasks) {
     return (key.hashCode() & Integer.MAX_VALUE) % numReduceTasks;
   }
}
```

Calculate Index of Partition:

- Convert key's hash into non-negative number
 - Logical AND with maximum integer value
- Modulo by number of reduce tasks

In case of more than 1 reducer

- Records distributed evenly across available reduce tasks
 - Assuming a good hashCode() function
- Records with same key will make it into the same reduce task
- Code is independent from the # of partitions/reducers specified

Custom Partitioner

 All blogs with the same author will end up in the same reduce task

10

OutputFormat

- Specification for writing data
 - The other side of InputFormat
- Implementation of OutputFormat<K,V>
- TextOutputFormat is the default implementation
 - Output records as lines of text
 - Key and values are tab separated "Key /t value"
 - Can be configured via "mapreduce.output.textoutputformat.separator" property
 - Key and Value may of any type call .toString()

OutputFormat

- Validates output specification for that job
 - You may have seen annoying messages that output directory already exists
- Creates implementation of RecordWriter
 - Responsible for actually writing data
- Creates implementation of OutputCommitter
 - Set-up and clean-up Job's and Tasks' artifacts (ex. Directories)
 - Commit or discard tasks output

42

OutputFormat Interface

```
public abstract class OutputFormat<K, V> {
  public abstract RecordWriter<K, V>
    getRecordWriter(TaskAttemptContext context
                  ) throws IOException, InterruptedException;
Provide implementation of
RecordReader
  public abstract void checkOutputSpecs(JobContext context
                  ) throws IOException, InterruptedException;
                                        Validates output specification for
Provide
                                        that job
implementation of
OutputCommitter
  public abstract
  OutputCommitter getOutputCommitter(TaskAttemptContext context
                  ) throws IOException, InterruptedException;
```

Hadoop's OutputFormats

- Hadoop eco-system is packaged with many OutputFormats
 - TextOutputFormat
 - DBOutputFormat
 - TableOutputFormat (HBASE)
 - MapFileOutputFormat
 - SequenceFileOutputFormat
 - NullOutputFormat
 - Etc...
- Configure on Job object
 - job.setOutputFormatClass(XXXOutputFormat.class);
 - job.setOutputKeyClass(XXXKey.class);
 - job.setOutputValueClass(XXXValue.class);

. .

TextOutputFormat

- Outputs plain text
- Saves key-value pairs separated by tab
 - Configured via mapreduce.output.textoutputformat.separator property
- Set output path
 - TextOutputFormat.setOutputPath(job, new Path(myPath));

TableOutputFormat

- Saves data into HTable
- Reducer output key is ignored
- Reducer output value must be HBase's Put or Delete objects

46

TableOutputFormat

47

TableOutputFormat

```
Scan scan = new Scan();
   scan.addColumn(toBytes(FAMILY), toBytes(INPUT COLUMN));
   TableMapReduceUtil.initTableMapperJob(
          TABLE NAME,
          scan,
          StartsWithCountMapper HBase.class,
          Text.class,
                                              Utilize HBase's
          IntWritable.class,
                                              TableMapReduceUtil to
                                              setup the job; internally
                                              delegate to job.XXX
   TableMapReduceUtil.initTableReducerJob(
                                              methods
          TABLE NAME,
          StartsWithCountReducer HBase.class,
   job.setNumReduceTasks(1);
   return job.waitForCompletion(true) ? 0 : 1;
public static void main(String[] a) throws Exception {
   int code = ToolRunner.run(new StartWithCountJob HBase(), a);
   System.exit(code);
```

TableOutputFormat

```
public class StartsWithCountReducer HBase extends
       TableReducer<Text, IntWritable, ImmutableBytesWritable> {
  @Override
  protected void reduce (Text key, Iterable < IntWritable > counts,
       Context context)
              throws IOException, InterruptedException {
       int sum = 0;
       for (IntWritable count : counts) {
              sum+= count.get();
                                             Reducer must output
                                             either Put or Delete
                                             object
       Put put = new Put(key.copyBytes());
       put.add(toBytes(FAMILY), toBytes(RESULT COLUMN),
                      toBytes(Integer.toString(sum)));
       context.write(null, put);
```

TableOutputFormat

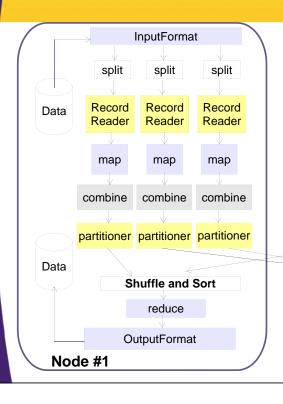
\$ yarn jar \$PLAY_AREA/HadoopSamples.jar mr.wordcount.StartWithCountJob_HBase \$ hbase shell

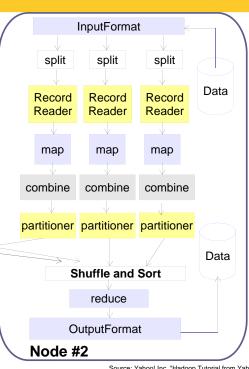
```
hbase(main):029:0> scan 'HBaseSamples', {COLUMN=>'count:result'}
ROW
              COLUMN+CELL
```

ICO W	COLOMINICELLE
D	column=count:result, timestamp=1338951024384, value=1
E	column=count:result, timestamp=1338951024384, value=1
U	column=count:result, timestamp=1338951024384, value=1
a	column=count:result, timestamp=1338951024384, value=1
c	column=count:result, timestamp=1338951024386, value=1
n	column=count:result, timestamp=1338951024386, value=1
r	column=count:result, timestamp=1338951024386, value=2
S	column=count:result, timestamp=1338951024386, value=1
t	column=count:result, timestamp=1338951024386, value=1
u	column=count:result, timestamp=1338951024386, value=1
W	column=count:result, timestamp=1338951024386, value=1

12 row(s) in 0.0530 seconds

Component Overview





Source: Yahoo! Inc. "Hadoop Tutorial from Yahoo!". 2012



Wrap-Up

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Summary

- In this lecture we learned about
 - MapReduce Theory
 - Types of Keys and Values
 - Input and Output Formats
 - Anatomy of
 - Mappers
 - Reducers
 - Combiners
 - Partitioners

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