

# Hadoop

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## 1 Open Source Map-Reduce: Hadoop

- About
- Cluster Configuration

## 2 Programming Hadoop

## 3 Hadoop in Reality

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- **About**

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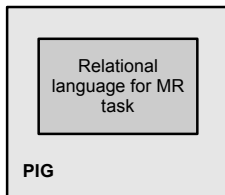
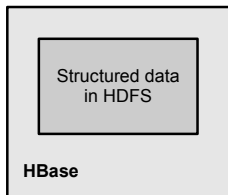
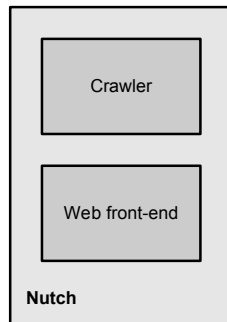
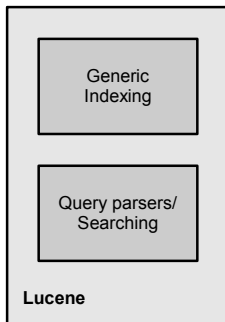
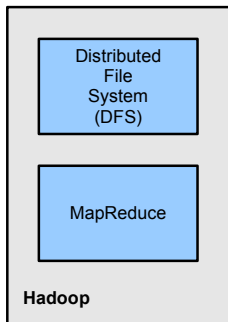
# The Hadoop project

- Mike Cafarella, Doug Cutting and others.
- Pronunciation not entirely clear. . .
- Project forked from Apache Nutch.
- Impressive, dynamic growth. Yahoo! involvement.
- Apache-license.



<http://lucene.apache.org/hadoop/>

# The open source MR landscape



# HDFS assumptions

HDFS is inspired by GFS (Google File System).

Design goals:

- expect hardware failures (processes, disk, nodes),
- streaming data access, large files (TB of data),
- simple coherence model (one writer, many readers),
- optimization of computation in map-reduce (data locality),
- single master (name node), multiple slaves (data nodes).

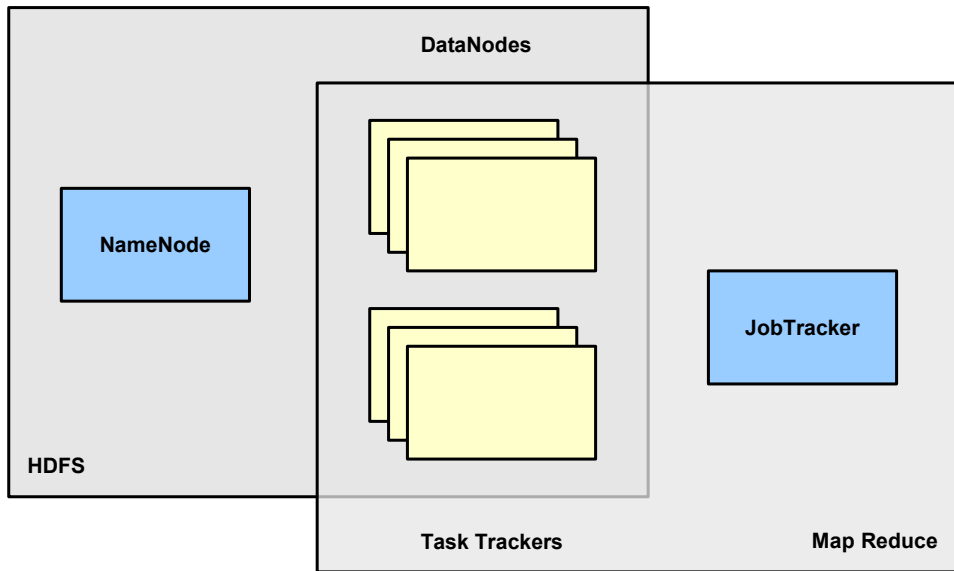
## 1 Open Source Map-Reduce: Hadoop

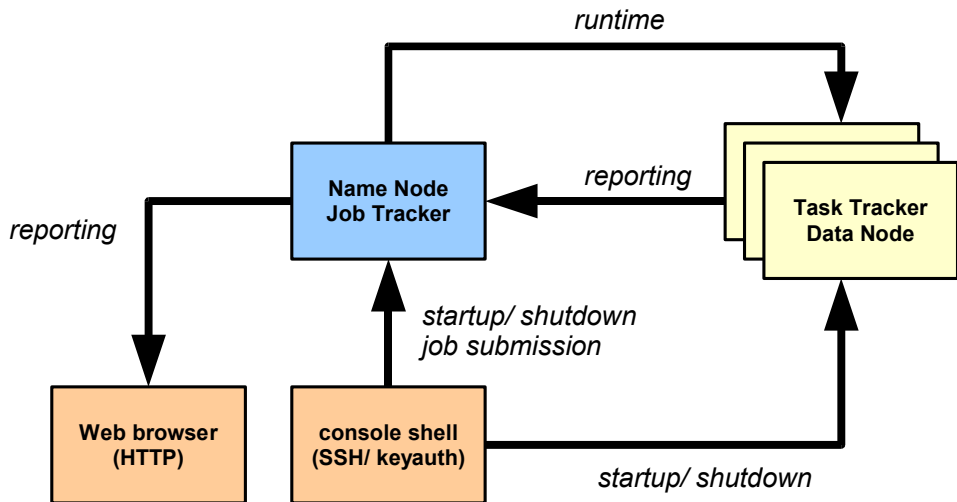
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# Prerequisites

Installation/ operation recommendations:

- Java 1.5.x or higher.
- Linux (or Windows under CygWin).
- Password-less SSH/ key cert.
- rsync, NFS and homogeneous paths make life easier.

## Setting up a cluster (in a pill)

- 1 Unpack Hadoop.
- 2 Configure: `hadoop-env.sh`, `hadoop-site.xml` and `slaves`.  

```
1 fs.default.name      -> masterhost:50010  
2 mapred.job.tracker  -> masterhost:50551
```
- 3 Prepare password-less SSH to the slaves.
- 4 Check firewall (!).
- 5 Distribute files.
- 6 Start the cluster.

Hadoop Web site contains more verbose instructions.

# Setting up local development environment

- 1 Unpack Hadoop.
- 2 Configure Hadoop as if you worked with a cluster.
- 3 Start Eclipse (Linux) or Eclipse from CygWin's shell (Windows).
- 4 Develop your map and reduce classes.
- 5 Set your job configuration to use local file system and in-process job tracker:

```
1 conf.set("mapred.job.tracker", "local");  
2 conf.set("fs.default.name", "local");
```

# NameNode '150.254.130.24:50010'

**Started:** Wed Nov 28 19:00:28 CET 2007  
**Version:** 0.15.0, r589881  
**Compiled:** Mon Oct 29 15:01:10 PDT 2007 by cutting

[Browse the filesystem](#)

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## Cluster Summary

**Capacity** : 476.85 GB  
**DFS Remaining** : 354.72 GB  
**DFS Used** : 260 KB  
**DFS Used%** : 0 %  
[Live Nodes](#) : 13  
[Dead Nodes](#) : 0

---

## Live Datanodes : 13

Node	Last Contact	Admin State	Size (GB)	Used (%)	Used (%)	Remaining (GB)	Blocks
lab-143-1	1	In Service	36.68	0	<input type="text"/>	27.27	0
lab-143-10	1	In Service	36.68	0	<input type="text"/>	27.29	0
lab-143-11	1	In Service	36.68	0	<input type="text"/>	27.26	0
lab-143-13	1	In Service	36.68	0	<input type="text"/>	27.28	0
lab-143-14	1	In Service	36.68	0	<input type="text"/>	27.28	0
lab-143-15	1	In Service	36.68	0	<input type="text"/>	27.29	0
lab-143-2	1	In Service	36.68	0	<input type="text"/>	27.28	0
lab-143-3	1	In Service	36.68	0	<input type="text"/>	27.28	0
lab-143-4	1	In Service	36.68	0	<input type="text"/>	27.28	0
lab-143-5	1	In Service	36.68	0	<input type="text"/>	27.29	0
lab-143-6	0	In Service	36.68	0	<input type="text"/>	27.29	0
lab-143-7	1	In Service	36.68	0	<input type="text"/>	27.3	0
lab-143-8	1	In Service	36.68	0	<input type="text"/>	27.32	0

**Dead Datanodes : 0**

## 1 Open Source Map-Reduce: Hadoop

## 2 Programming Hadoop

- Jobs
- Reading and splitting the input data
- Affecting shuffling: partitioning and sorting
- Tracking progress: counters
- Job launching

## 3 Hadoop in Reality

## 1 Open Source Map-Reduce: Hadoop

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### ■ Jobs

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## Map Reduce jobs:

- Classes (JARs) in Java.
- Hadoop Streaming (arbitrary shell commands).
- C/C++ APIs to HDFS.

This lecture focuses on Java jobs.

# Hadoop Java API: Fundamentals

```
1 public interface Mapper<
2     K1 extends WritableComparable, V1 extends Writable,
3     K2 extends WritableComparable, V2 extends Writable>
4     extends JobConfigurable, Closeable
5 {
6     void map(K1 key, V1 value, OutputCollector<K2, V2> output,
7             Reporter reporter)
8         throws IOException;
9 }
```

```
1 public interface Reducer<
2     K2 extends WritableComparable, V2 extends Writable,
3     K3 extends WritableComparable, V3 extends Writable>
4     extends JobConfigurable, Closeable
5 {
6     void reduce(K2 key, Iterator<V2> values, OutputCollector<K3, V3> output,
7             Reporter reporter)
8         throws IOException;
9 }
```

Note shared parameter names between the mapper and the reducer, these must be compatible types.

## Writable and WritableComparable

Hadoop uses custom data type serialization/ RPC mechanism.  
Hence the two `Writable` and `WritableComparable` interfaces.

A number of default implementations are available:

- Primitive types:  
`IntWritable`, `Text`, `FloatWritable`, ...
- Arrays and dictionaries:  
`MapWritable`, `ArrayWritable`, ...
- Record compiler generated classes (`rcc`, later).

## Example: word count

```
1 /**
2  * Counts the words in each line.
3  *
4  * For each line of input, break the line into words and emit them as
5  * (<b>word</b>, <b>1</b>).
6  */
7 public class Map extends MapReduceBase
8     implements Mapper<LongWritable, Text, Text, IntWritable> {
9
10     private final static IntWritable one = new IntWritable(1);
11     private Text word = new Text();
12
13     public void map(LongWritable key, Text value,
14         OutputCollector<Text, IntWritable> output,
15         Reporter reporter) throws IOException
16     {
17         final String line = value.toString();
18         final StringTokenizer itr = new StringTokenizer(line);
19         while (itr.hasMoreTokens()) {
20             word.set(itr.nextToken());
21             output.collect(word, one);
22         }
23     }
24 }
```

## Example: word count

```
1 /**
2  * A reducer class that just emits the sum of the input values.
3  */
4 public class Reduce extends MapReduceBase
5     implements Reducer<Text, IntWritable, Text, IntWritable> {
6
7     public void reduce(Text key, Iterator<IntWritable> values,
8         OutputCollector<Text, IntWritable> output,
9         Reporter reporter) throws IOException
10    {
11        int sum = 0;
12        while (values.hasNext()) {
13            sum += values.next().get();
14        }
15        output.collect(key, new IntWritable(sum));
16    }
17 }
```

## Example: word count

```
1 public static void main(String[] args) throws IOException {
2     final JobConf conf = new JobConf(WordCount.class);
3     conf.setJobName("wordcount");
4
5     // The keys are words (strings).
6     conf.setOutputKeyClass(Text.class);
7
8     // The values are counts (ints).
9     conf.setOutputValueClass(IntWritable.class);
10
11    // Set mapper, reducer and combiner ('local' aggregation)
12    conf.setMapperClass(MapClass.class);
13    conf.setCombinerClass(Reduce.class);
14    conf.setReducerClass(Reduce.class);
15
16    // Set input and output paths. Note 'virtual' paths.
17
18    conf.setInputPath(new Path(input));
19    conf.setOutputPath(new Path(output));
20
21    // Set job properties here (to run locally for example).
22    // ...
23
24    JobClient.runJob(conf);
25 }
```

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- Jobs

- **Reading and splitting the input data**

- Affecting shuffling: partitioning and sorting

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## Data access abstraction layer

How does Hadoop split input files into keys and values?

## Data access abstraction layer

How does Hadoop split input files into keys and values?

Record access API:

```
InputFormat → InputSplit[] → RecordReader
```

Default implementation:

```
TextInputFormat → InputSplit[] → LineRecordReader
```

# InputFormat

```
1 public interface InputFormat<K extends WritableComparable,V extends Writable>
2 {
3     void validateInput(JobConf job);
4
5     InputSplit[] getSplits(JobConf job, int numSplits /* hint! */);
6
7     RecordReader<K,V> getRecordReader(
8         InputSplit split, JobConf job, Reporter reporter);
9 }
```

Note `InputFormat` is not tied to the filesystem, but implies the record format (returns a compatible `RecordReader`). Examples:

- `TextInputFormat` (file-based, K: line number, V: a single line, splits span many files),
- `KeyValueTextInputFormat` (file-based, K,V: lines split on a given byte separator),
- `SequenceFileInputFormat` (file-based, list of binary key/value pairs).

## Sequence files

- Sequence files store arbitrary `Writable` key/ value pairs.
- Sequence files can be compressed internally (block and record compression).
- Complex structures can be easily created/stored with a combination of `rcc` compiler and sequence files.

## Sequence file using local file system

```
1 // These would be normally provided by Hadoop's MR
2 final FileSystem fs = new RawLocalFileSystem();
3 fs.setConf(new Configuration());
4
5 // Output path for the sequence file.
6 final Path output = new Path("sequence-file.sf");
7
8 // Create a writer,
9 final SequenceFile.Writer writer = SequenceFile.createWriter(
10     fs, conf, output,
11     Text.class, IntWritable.class,
12     SequenceFile.CompressionType.BLOCK);
13
14 // Write to the writer.
15 writer.append(new Text("stary"), new IntWritable(0xbaca));
16
17 // Close it.
18 writer.close();
```

## Using sequence files in a Map-Reduce job

```
1 // ...
2 jobConf.setInputFormat(SequenceFileInputFormat.class);
3 jobConf.setOutputFormat(SequenceFileOutputFormat.class);
4 // ...
```

Note that key and value classes must be compatible with the mapper and the reducer.

## Writing a custom InputFormat?

- Start from one of the defined base classes:  
FileInputFormat, MultiFileInputFormat.
- Implement a custom RecordReader, look and extensively reuse existing code (LineRecordReader):

```
1 boolean next(K key, V value) throws IOException;
2 K createKey();
3 V createValue();
```

## Writing a custom InputFormat?

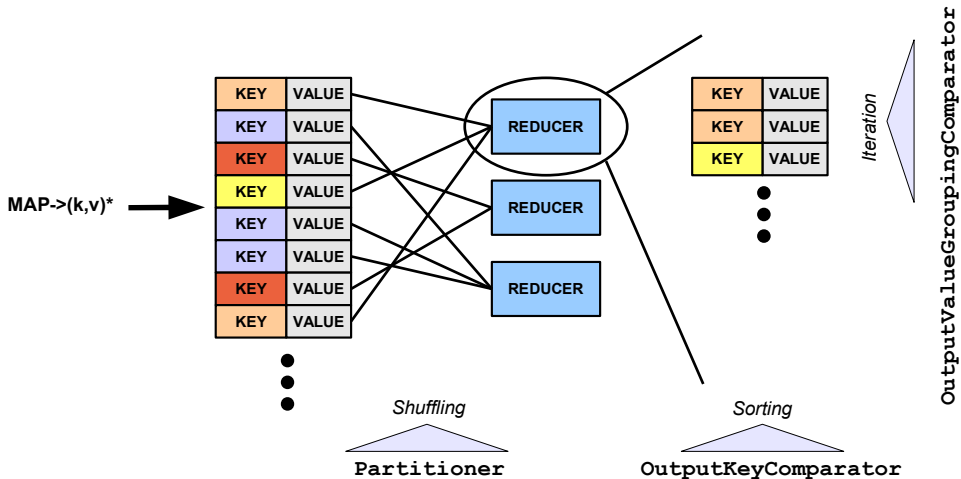
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```

What happens if a split is not on a record boundary?

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# Data shuffling and sorting



## Shuffling and sorting

- Partitioning is performed immediately, unless a combiner function has been specified.
- For the combiner, data is sorted locally, the combiner is executed, the result is partitioned.
- Sorting uses merge sort. The in-memory buffer size is configurable (larger buffer means fewer passes).

## Overriding the defaults

Possible via properties on JobConf:

```
1 conf.setOutputKeyComparatorClass(MyComparator.class);
```

- Partitioner should **balance** keys among reducers.
- Partitioner may work on a subset of the key (i.e., a domain of the entire URI).
- The comparator above is not `java.util.Comparator`!

## Custom comparator example

```
1 public class CustomTextComparator extends WritableComparator {
2     private Collator collator;
3
4     public CustomTextComparator() {
5         super(Text.class);
6
7         final Locale locale = new Locale("pl");
8         collator = Collator.getInstance(locale);
9     }
10
11    public int compare(WritableComparable a, WritableComparable b) {
12        synchronized (collator) {
13            return collator.compare(
14                ((Text) a).toString(), ((Text) b).toString());
15        }
16    }
17 }
```

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# Counters

Maps and reduces receive a `Reporter` object, which has the following methods:

```
1 void setStatus(String status);
2 void incrCounter(Enum key, long amount);
3 // ... (and other methods, not relevant here)
```

Note that counters are subclasses of `Enum`. This is quite convenient:

```
1 public enum MyStatistics {
2     COUNTER_NAME, ANOTHER_COUNTER,
3 }
```

```
1 public void map(...) throws IOException {
2     reporter.incrCounter(MyStatistics.COUNTER_NAME, 1);
3     reporter.incrCounter(MyStatistics.ANOTHER_COUNTER, 5);
4     // ...
5 }
```

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- Pack your Java job as a jar file.
- Place any extra jars under /lib folder inside the jar.
- Submit the job from command line:

```
1 hadoop jar [job-file.jar] my.package.MyMainClass param1 param2 ...
```

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## The trickery of Hadooping...

- Very fast development.
- Windows installation often broken (scripts, paths).
- Not source-level documentation scarce and not really up-to-speed with the code.
- Real setup of a distributed cluster requires some initial work (account setup, moving distributions around).

# Power of Hadoop

- Writing map reduce jobs relatively easy.
- It really works.
- Compatible with Amazon's paid cluster infrastructure.

