

MANAGEMENT AND PROCESSING OF BIG DATA

LEVEL-I

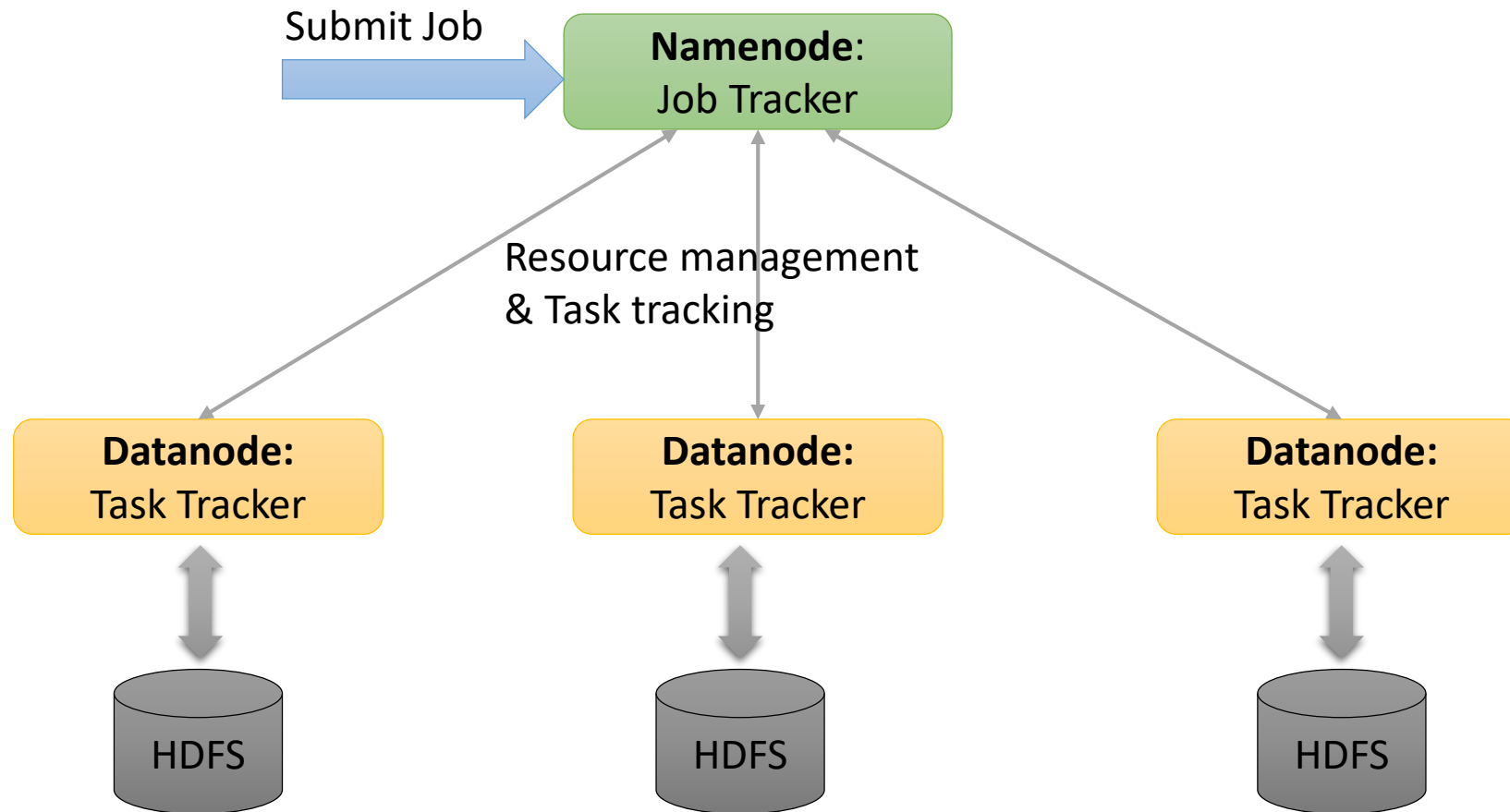
SESSION-2



Recap

- Data processing trend
- Big Data and its characteristics
- Applications of big data
- Various computing technologies
- History of Hadoop
- Major components of Hadoop cluster
- Unix commands
- HDFS

Hadoop Cluster Overview



HDFS

- HDFS is a file system designed for storing very large files with streaming data access patterns, running on clusters of commodity hardware

Large Files

**Streaming data
access**

**Commodity
hardware**

HDFS Blocks

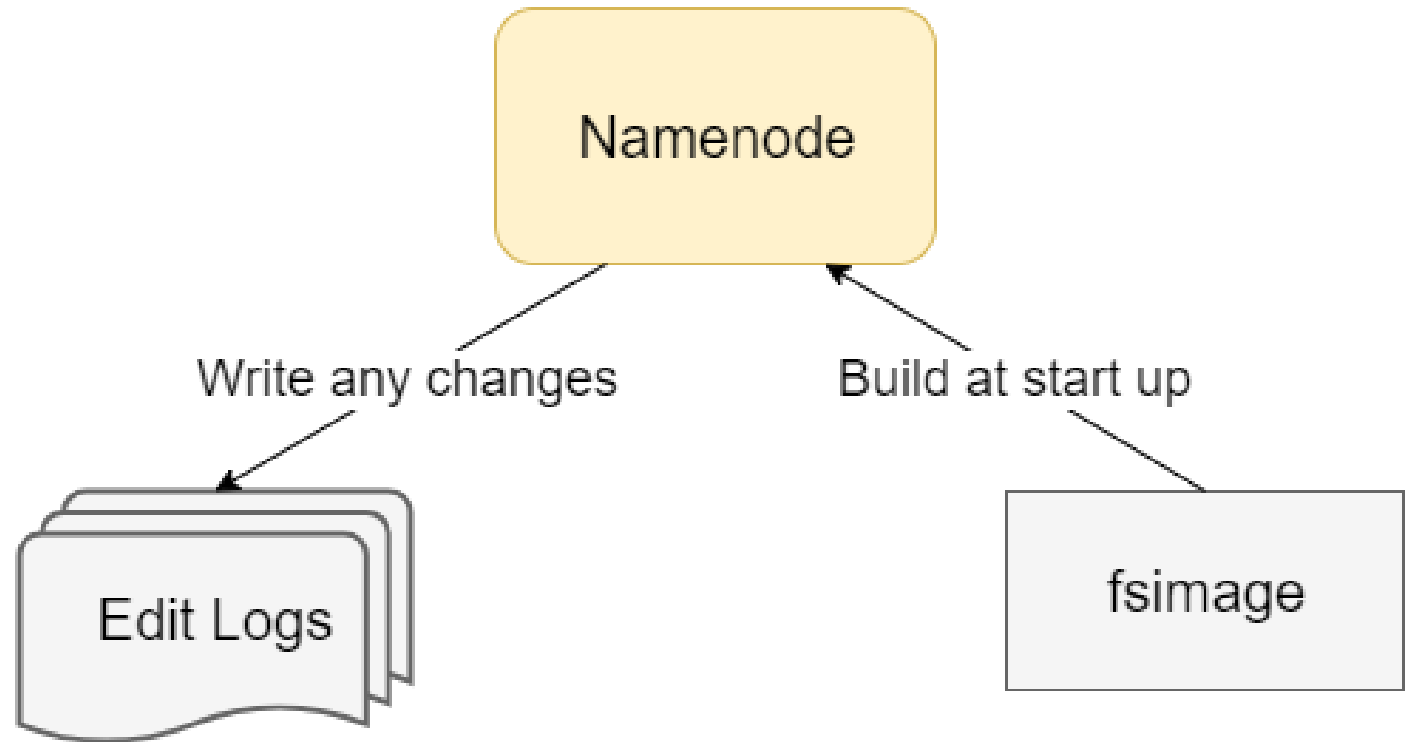
- Single unit of storage
- Default block size is 128MB
- Size of block will drive the ratio of time to read a block to the seek for a block

Benefits of blocks

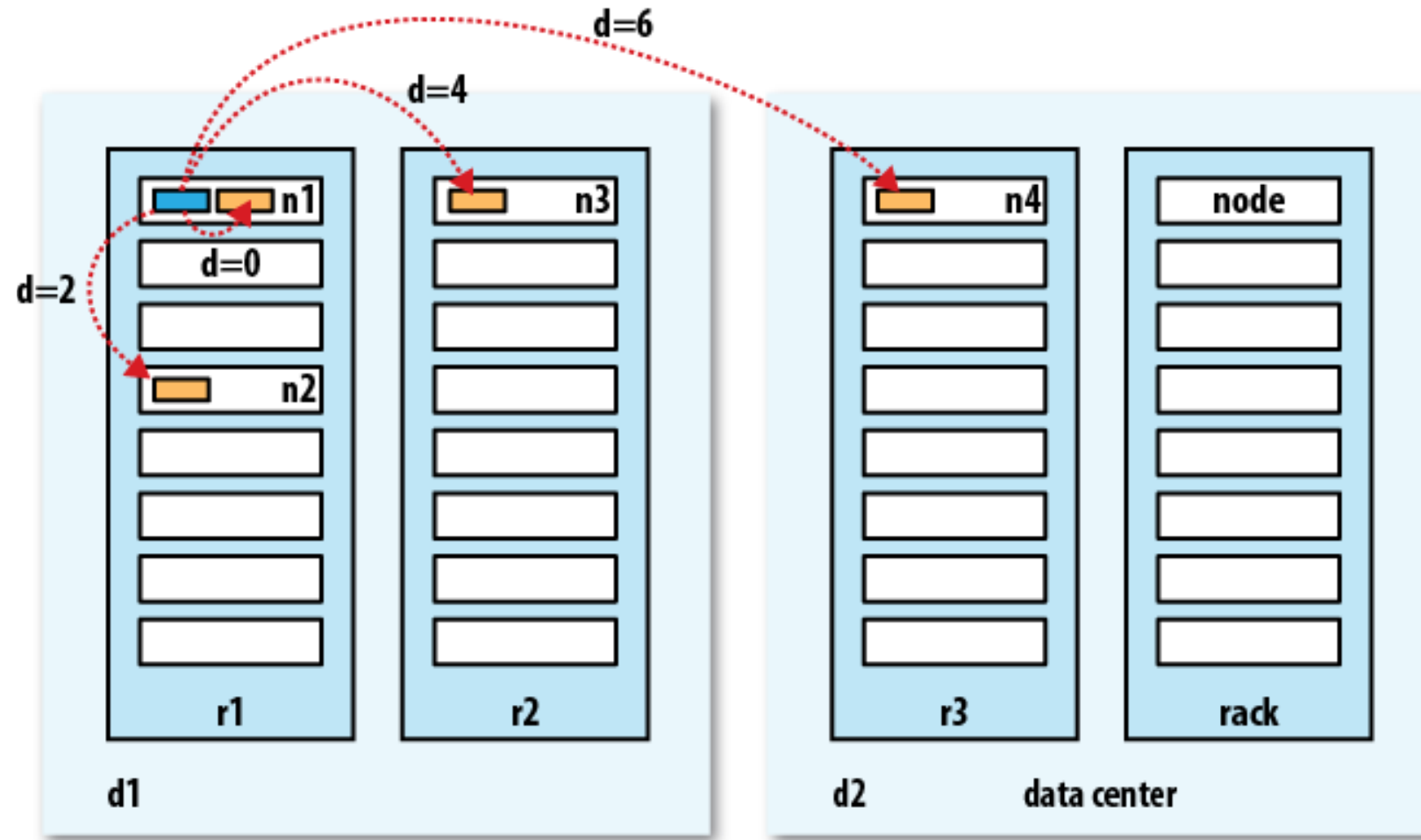
- Files can be larger than a single disk
- Simplicity at storage level as data node doesn't store any metadata
- Fault tolerance by replicating blocks

Filesystem metadata

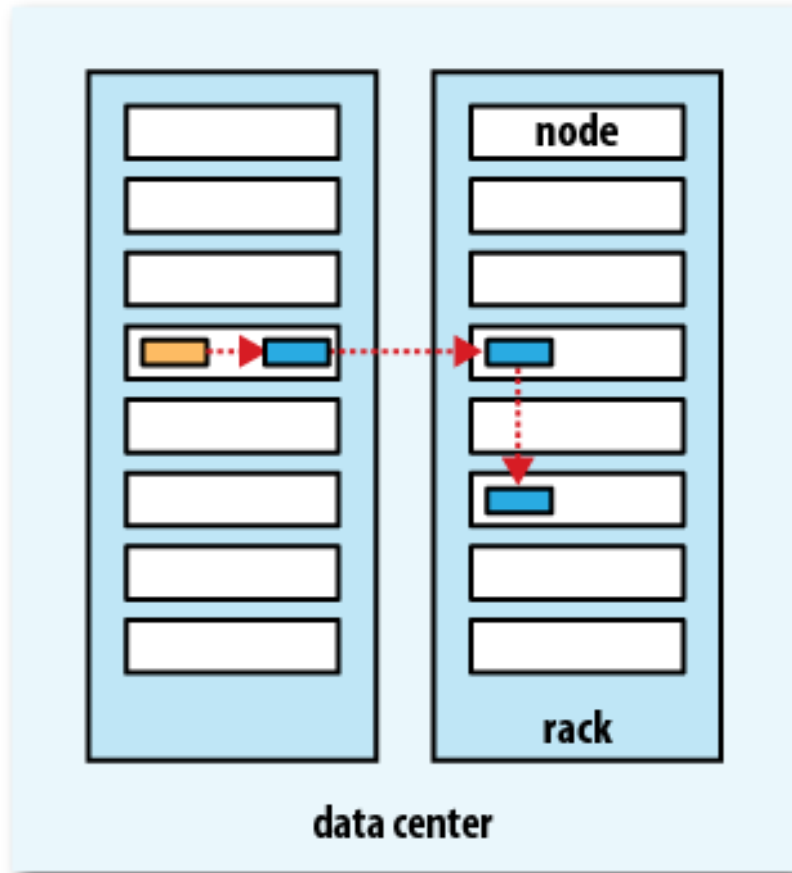
- Namenode stores the metadata
- Backup of metadata on secondary namenode



Network Topology



Rack awareness



HDFS CLI Read Commands

- Copy a file from local file system to HDFS

```
hadoop fs -copyToLocal <source> <target>
```

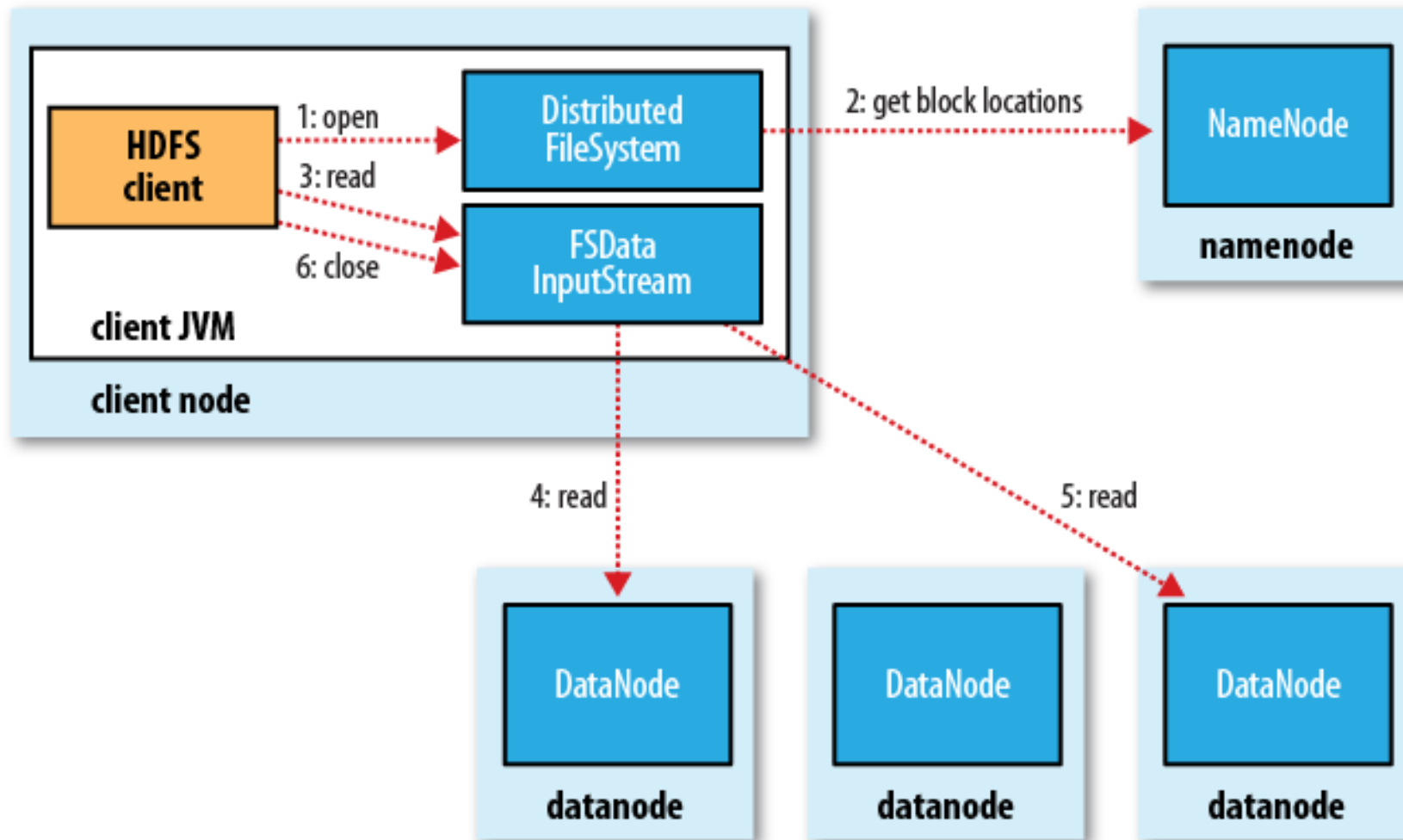
- Get merged file from HDFS to local file system

```
hadoop fs -getmerge <source> <target>
```

- Cat a file from HDFS

```
hadoop fs -cat <filename>
```

HDFS Read operation



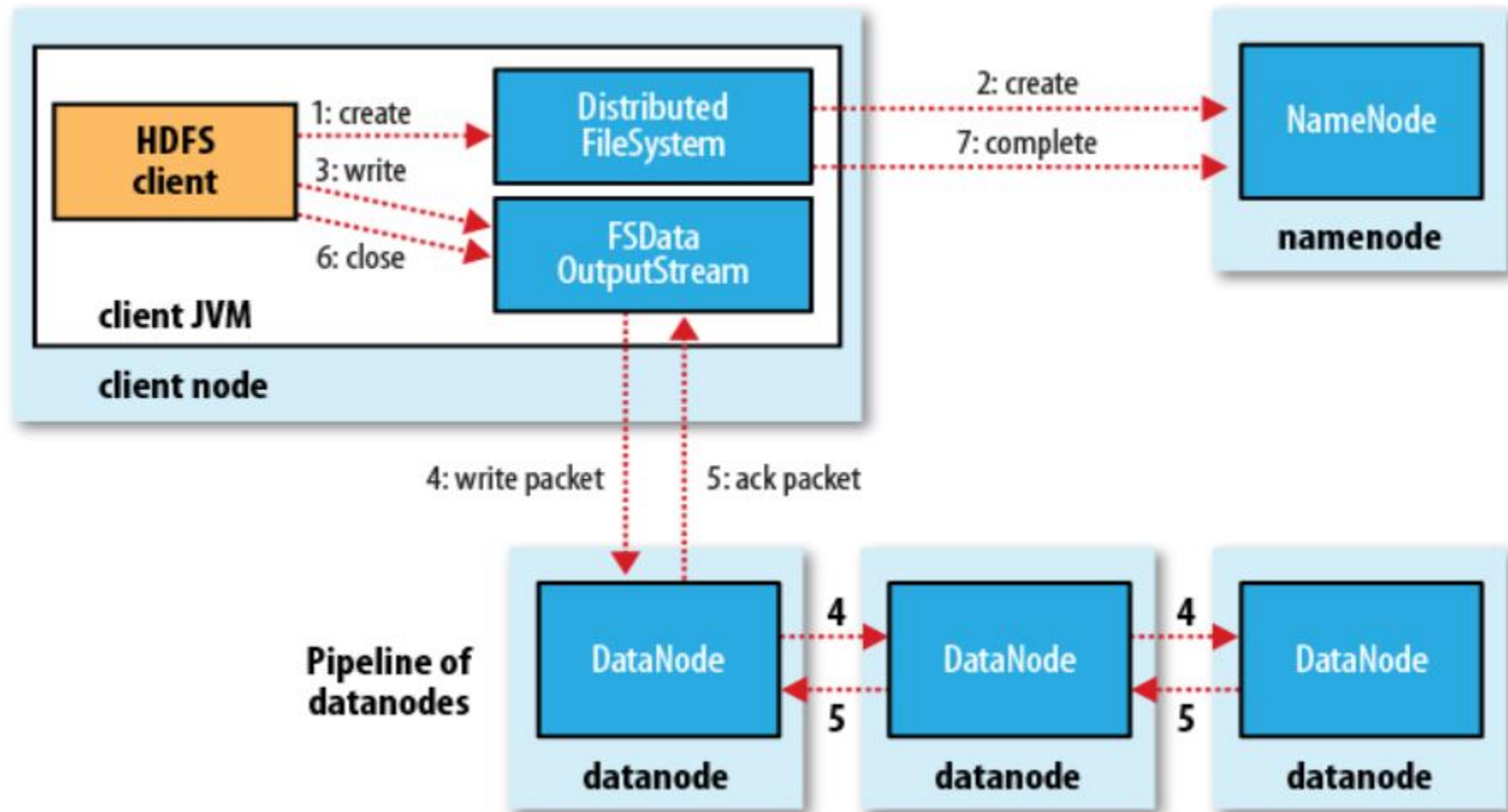
HDFS CLI Write Commands

- Write a file from local file system to HDFS

```
hadoop fs -copyFromLocal <source> <target>
```

```
hadoop fs -put <source> <target>
```

HDFS Write operation



HDFS not made for

- Low-latency data access
- Lots of small files
 - Each file | block | directory stores around 150 bytes of metadata.
Hence 1 million files each of one block will consume 300 MB of storage on Namenode
- Multiple writers, arbitrary file modifications

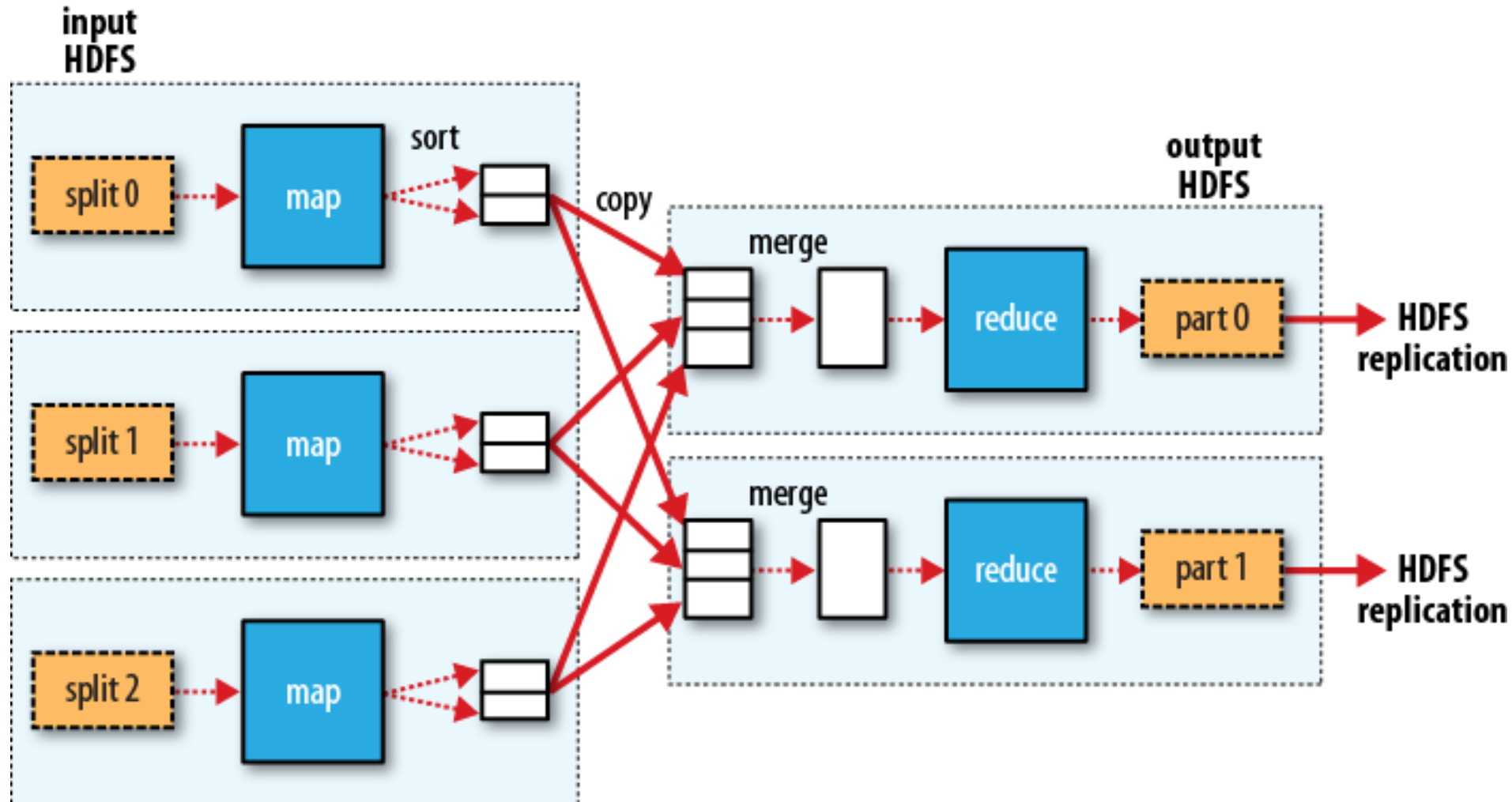
Agenda for today

- MapReduce detailed discussion
- Running your first MapReduce program
- MapReduce Job Chaining
- Hadoop 1 vs 2
- YARN

MapReduce

- Two major phases: Map and Reduce
- Notion of <Key, Value> pairs
- Divides job into multiple tasks
- Map: extract important information from each record
- Reduce: Aggregate, Summarize, Filter, Transform

MapReduce Stages

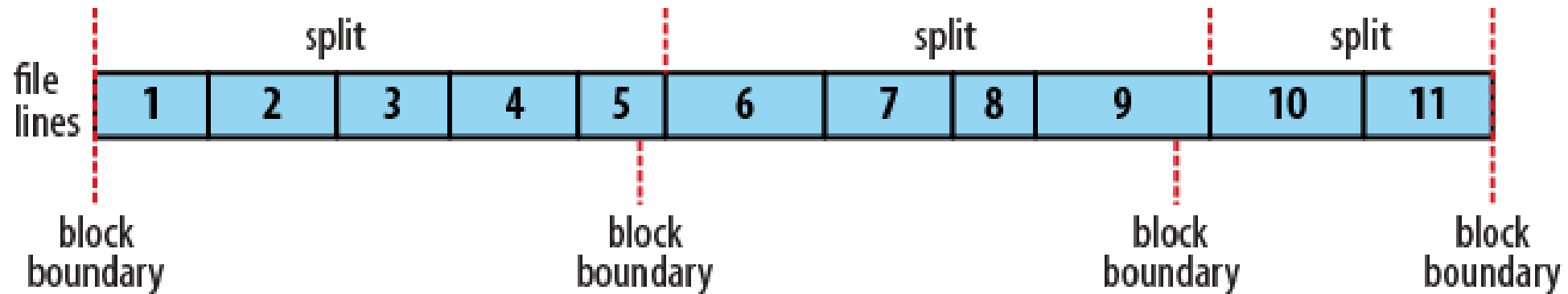


Map Tasks

- What is a good approach to decide how many map tasks a job should launch?
- Less number of big tasks
vs
higher number of small tasks
- Normally same as input data blocks

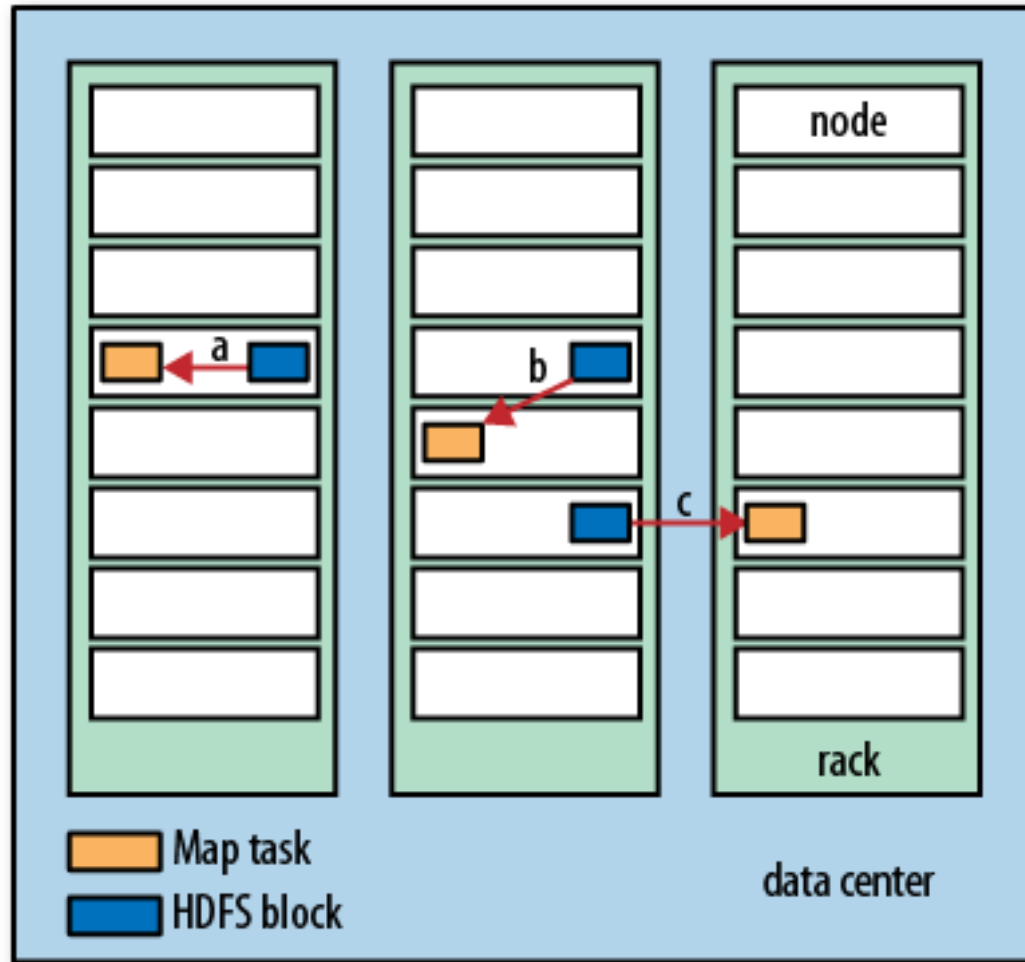
Input Splits

- Blocks are of fixed size
- Good chances of records being split between two block



Task to node mapping

- Notion of data locality



Reduce Tasks

- Can be configured by programmer
- Normally same as #datanodes participating in execution
- Input Key and Value type should be same as output type of mapper
- One output file per reducer under output directory
- Generates exception if output directory already exists.

MapReduce: Mapper code

```
public class WebHitCounterMapper extends
Mapper<Input Key, Input Value, Output Key, Output Value>
{

    public void map(Input Key, Input Value, Context
context) throws IOException, InterruptedException {

        <MAP Logic goes here>

        context.write(Output Key, Output Value)
    }
}
```

MapReduce: Reducer code

```
public class WebHitCounterReducer extends
Reducer<Input Key, Input Value, Output Key, Output Value>
{
    public void reduce(Input Key, Iterable<Value Data type>
values, Context context) throws IOException, InterruptedException {

        <REDUCE logic goes here>

        context.write(Output Key, Output Value);
    }
}
```

MapReduce: Driver Code

```
public class WebHitCounterMain {  
    public static void main(String[] args) throws Exception {  
  
        Configuration conf = new Configuration();  
        Job job = Job.getInstance(conf, "Daily Web Hit Counter");  
  
    }  
}
```

MapReduce: Driver Code

```
public class WebHitCounterMain {  
    public static void main(String[] args) throws Exception {  
  
        Configuration conf = new Configuration();  
        Job job = Job.getInstance(conf, "Daily Web Hit Counter");  
  
        job.setJarByClass(main.WebHitCounterMain.class);  
        job.setMapperClass(mapper.WebHitCounterMapper.class);  
        job.setReducerClass(reducer.WebHitCounterReducer.class);  
  
    }  
}
```

MapReduce: Driver Code

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        job.setReducerClass(reducer.WebHitCounterReducer.class);  
  
        job.setOutputKeyClass(Text.class);  
        job.setOutputValueClass(IntWritable.class);  
  
    }  
}
```


MapReduce: Driver Code

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public class WebHitCounterMain {  
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        job.setReducerClass(reducer.WebHitCounterReducer.class);  
  
        job.setOutputKeyClass(Text.class);  
        job.setOutputValueClass(IntWritable.class);  
  
        FileInputFormat.addInputPath(job, new Path(args[0]));  
        FileOutputFormat.setOutputPath(job, new Path(args[1]));  
  
    }  
}
```

MapReduce: Driver Code

```
public class WebHitCounterMain {  
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        FileInputFormat.addInputPath(job, new Path(args[0]));  
        FileOutputFormat.setOutputPath(job, new Path(args[1]));  
  
        System.exit(job.waitForCompletion(true) ? 0 : 1);  
    }  
}
```

Additional Conf Properties

- Configure maximum retries for failed tasks
 - `setMaxMapAttempts(int n)`
 - `setMaxReduceAttempts(int n)`
- Configure reduce tasks explicitly
 - `setNumReduceTasks(int tasks)`

Input formats

Input format	Description
TextInputFormat	Read Text file line by line. Key is offset and value is record text
KeyValueTextInputFormat	Tab separated key values from a text file
SequenceFileInputFormat<K,V>	Hadoop's file format

Output Formats

Output Format	Description
TextOutputFormat<K,V>	Tab separated key value pairs in plain text format. One record per key value pair
SequenceFileOutputFormat<K,V>	Hadoop's Sequence file format
NullOutputFormat<K,V>	Nothing. Helps in map only job

Programming Exercise

Performance tuning

- Cluster configuration
- Use compression technique
- Tuning # mappers and reducers
- Use combiner
- Appropriate data type
- Reuse objects
- Profiling

MapReduce Job chaining

- Two separate jobs
- Multiple mappers/reducers within same job

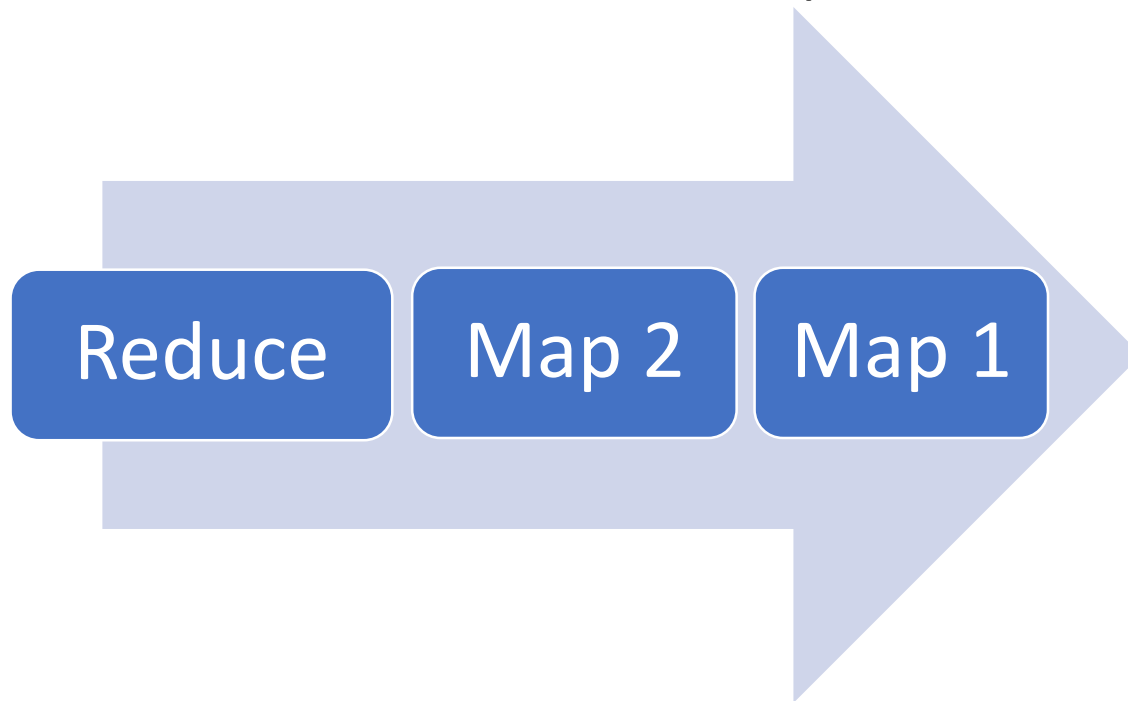
MapReduce Job chaining

- Two separate jobs
1. Configure multiple job objects and run sequentially.



MapReduce Job chaining

- Multiple mappers/reducers within same job
 - ChainMapper API: to add multiple mappers
 - ChainReducer API: to add multiple reducers



Challenges with Hadoop 1

- Applications were limited to MapReduce implementations only
- Namenode machine crash or maintenance activity
- Namespace scaling
- Backup and Recovery
- Batch oriented architecture
- Support for various file formats
- Dual responsibilities of Job tracker

Hadoop 2

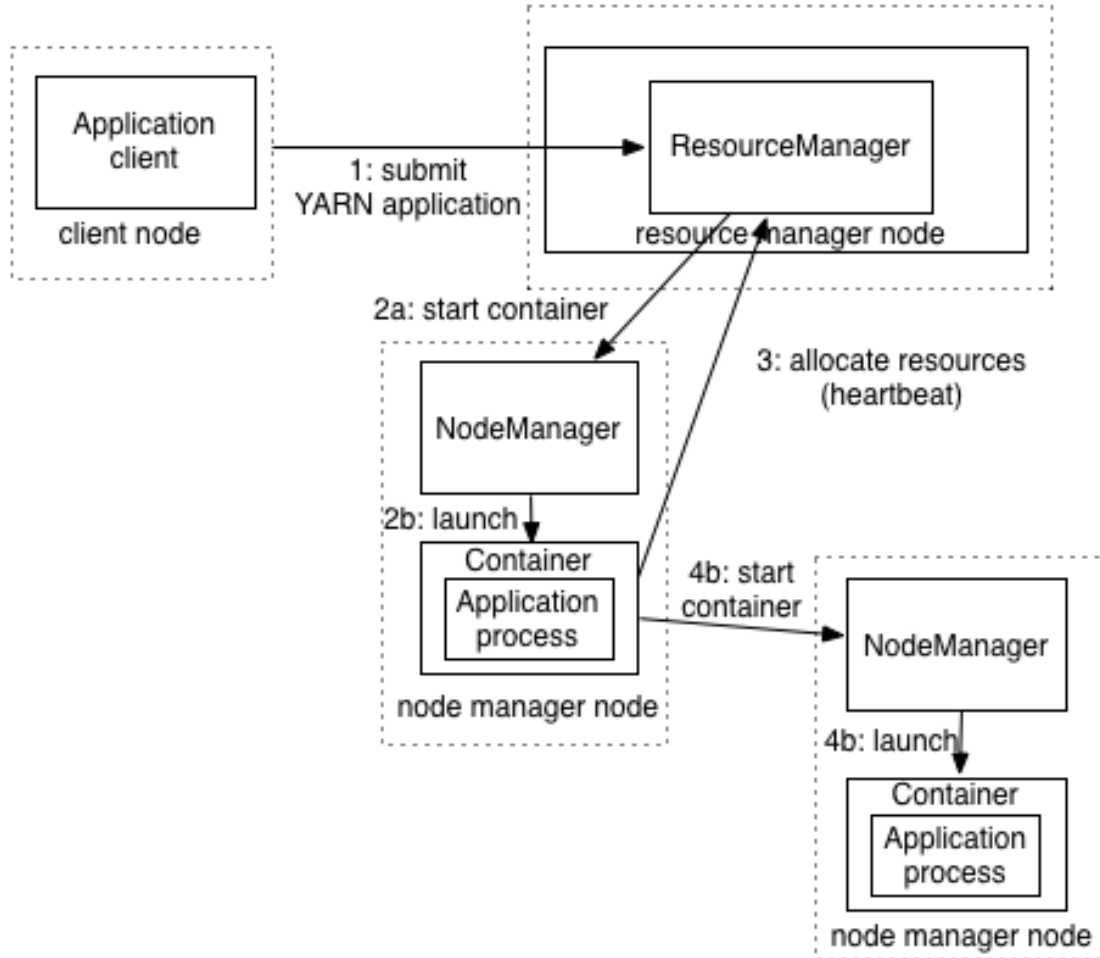
- Support for other data processing engines
- High Availability
- HDFS Federation
- HDFS Snapshot
- Introduced Streaming and Interactive analysis
- Support for various file formats
- YARN

YARN

- Yet Another Resource Negotiator

MapReduce 1	YARN
Job Tracker	Resource Manager, Application Master and Timeline server
Task Tracker	Node Manager
Slot	Containers

YARN model



Pros of YARN

- Scalability
 - 4,000 node and 40,000 tasks to 10,000 nodes and 100,000 tasks
- Availability
 - High Availability (HA) feature
- Utilization
 - No fixed slots for map and reduce.
- Multitenancy
 - Running other distributed system along Hadoop

References

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