# Software for cluster computing

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### **Abstract**

This presentation will cover some of the software systems for cluster computing with an emphasis on R libraries that can work with these systems.

# Common features of software systems for cluster computing

All of software systems described in this presentation are

- distributed under open source licenses,
- have lots of free documentation, and
- run on any reasonable Unix based system.

# Message Passing Interface (MPI)

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### Message Passing Interface (MPI)

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### **Table of Contents**

- Abstract
- 2. What is MPI?
- 3. LLNL MPI Implementations and Compilers
- 4. Getting Started
- 5. Environment Management Routines
- 6. Exercise 1
- 7. Point to Point Communication Routines
  - 1. General Concepts
  - 2. MPI Message Passing Routine Arguments
  - 3. Blocking Message Passing Routines
  - 4. Non-blocking Message Passing Routines
- 8. Exercise 2
- 9. Collective Communication Routines
- 10. Derived Data Types
- 11. Group and Communicator Management Routines
- 12. Virtual Topologies
- 13. A Brief Word on MPI-2 and MPI-3
- 14. Exercise 3
- 15. References and More Information
- 16. Appendix A: MPI-1 Routine Index

Screenshot from Blaise Barney's MPI tutorial

## MPI is a routine library

### MPI Routines used for communicating between nodes

The Message Passing Interface is a standard used for message passing. Typically it is used in conjunction with a C or C++ program to farm out computation to the nodes of a cluster. The implementation of MPI used in this project was the open-source MPICH library. Two types of MPI operations were used in this project, collective and non-collective operations. Only two non-collective operations were used to send data from one node to another, MPI\_Recv is used to receive data from a particular node. Both these operations are blocking, meaning that the node which calls the operation pauses until the operation is complete.

```
MPLSend(void *buf, int count, MPLDatatype datatype, int dest, int tag, MPLComm comm)

MPLRecv(void *buf, int count, MPLDatatype datatype, int source, MPLComm comm, MPLStatus *status)
```

The other operations used are all collective. The MPI\_Bcast operation broadcasts a message from the root node to all other nodes/processes in the specified group. This is used to broadcast the dimension of the matrix to all nodes, and also to broadcast an "exit" matrix to each node.

```
MPLBcast(void *buffer, int count, MPLDatatype datatype, int root MPLComm comm)
```

When the group of nodes that are to work on the matrix-vector multiplication has been set up, the root node must give out a portion of the matrix to each node. This can be achieve with MPI\_Sentd, but it is much more efficient to use the MPI\_Scatter operation. This operation farms out pieces of an array to different node. Thus, the decomposition of the matrix can be achieved in just one command!

```
MPLScatter(void *sendbuf, int sendcnt, MPLDatatype sendtype, void *recvbuf int recvcnt, MPLDatatype recvtype, int root, MPLComm comm)
```

There is also a function called MPI\_Gather that implements the opposite function of MPI\_Scatter. When called on the root node, it gathers in data of a fixed size from all the nodes in the specified group, into an array. This is used to gather in the newly calculated qubit vector from the nodes, when the calculation is finished.

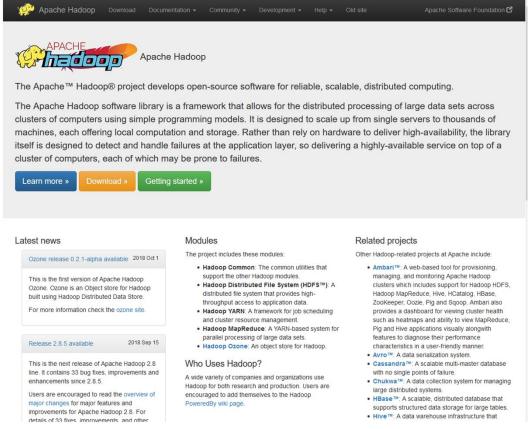
```
MPLGather(void *sendbuf, int sendcnt, MPLDatatype sendtype, void *recvbuf, int recvcount, MPLDatatype recvtype, int root, MPLComm comm)
```

None of the collective operations detailed above are blocking, even though they must operate at the same time on each node. To synchronize all the nodes, the MPI\_Barrier operation is called after a collective function. This ensures that all nodes in the group are operating in the correct place.

MPI\_Barrier(MPI\_Comm comm)

Colm Ó hÉigeartaigh;'s website listing of MPI routines

# Hadoop

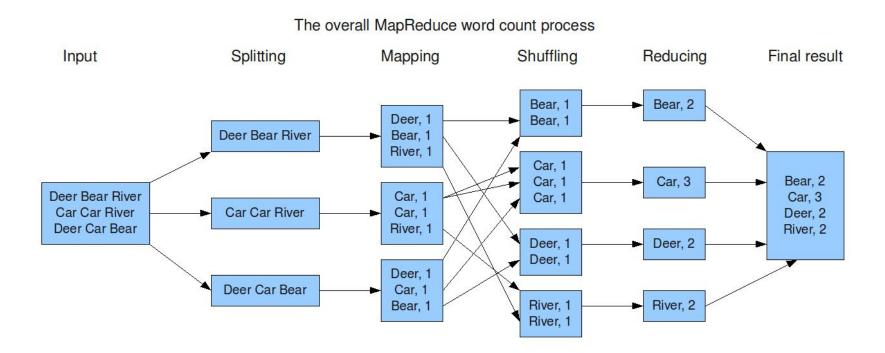


Screenshot from Apache Hadoop website

# Components of Hadoop

- MapReduce
- Hadoop Distributed File System (HDFS)
- Hive
- Pig

# MapReduce



Mapreduce applied to a simple word count example

# Hadoop Distributed File System (HDFS)

**HDFS Architecture** 

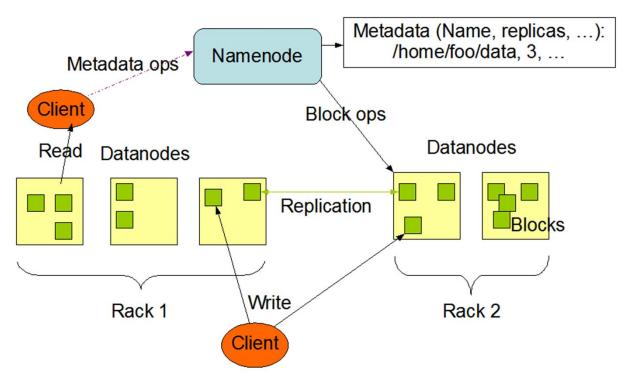


Illustration of HDFS architecture

# Pig

# Hive

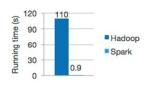
# Spark



### Speed

Run workloads 100x faster.

Apache Spark achieves high performance for both batch and streaming data, using a state-of-the-art DAG scheduler, a query optimizer, and a physical execution engine.



Developers -

Logistic regression in Hadoop and Spark

### Ease of Use

Write applications quickly in Java, Scala, Python, R, and SQL.

Spark offers over 80 high-level operators that make it easy to build parallel apps. And you can use it interactively from the Scala, Python, R, and SQL shells.



Spark's Python DataFrame API Read JSON files with automatic schema inference

### Latest News

Spark 2.3.2 released (Sep 24, 2018) Spark+Al Summit (October 2-4th,

Spark 2.4.0 released (Nov 02, 2018)

Apache Software Foundation •

2018, London) agenda posted (Jul 24,

Spark 2.2.2 released (Jul 02, 2018)



### **Download Spark**

### Built-in Libraries:

SQL and DataFrames Spark Streaming MLlib (machine learning) GraphX (graph)

Third-Party Projects

Screenshot of Apache Spark main web page

## Conclusion

### This talk has covered

- Message Passing Interface
- Hadoop
  - MapReduce
  - HDFS
  - Pig
  - Hive
- Spark