CS 744: MAPREDUCE

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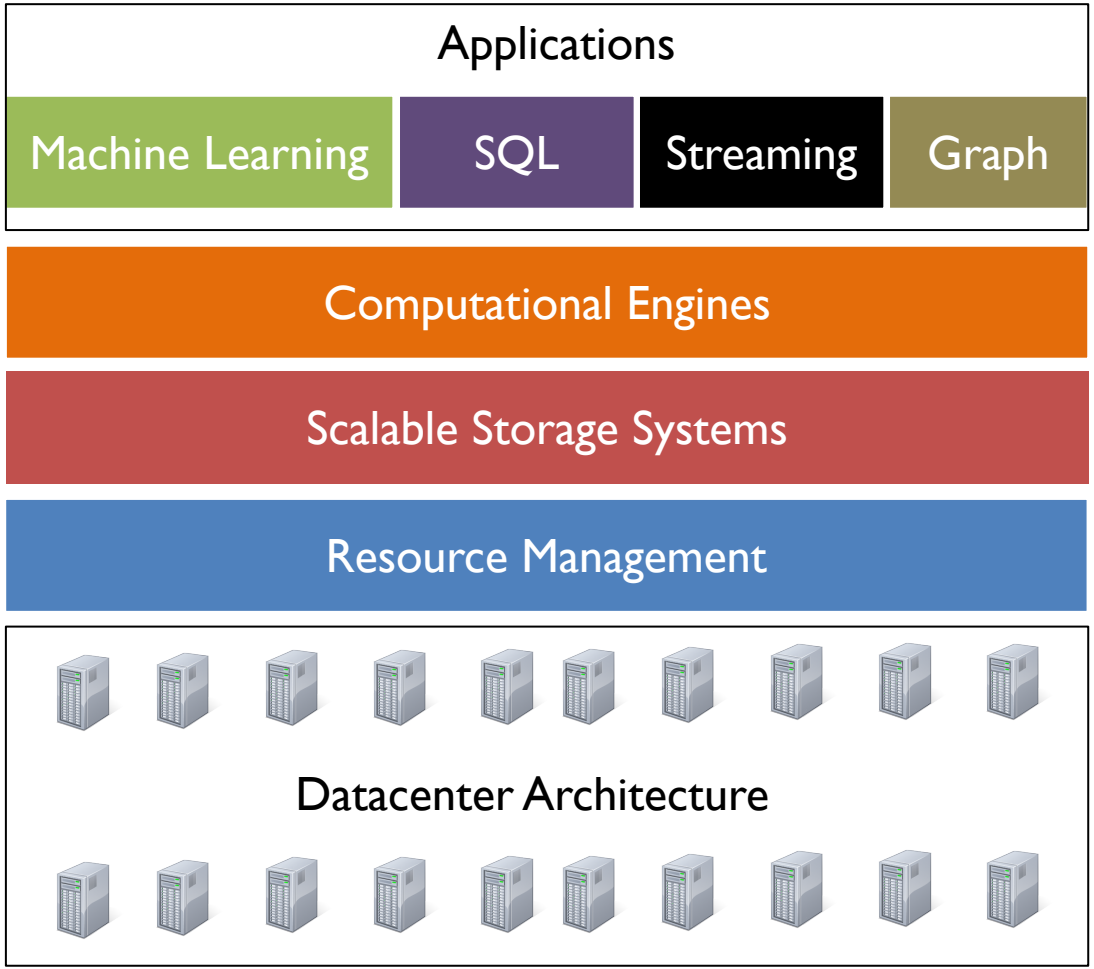
ANNOUNCEMENTS

- Assignment 1 deliverables
  - Code (comments, formatting)
  - Report
    - Partitioning analysis (graphs, tables, figures etc.)
    - Persistence analysis (graphs, tables, figures etc.)
    - Fault-tolerance analysis (graphs, tables, figures etc.)

- See Piazza for Spark installation


Applications
Machine Learning
SQL
Streaming
Graph
Computational Engines
Scalable Storage Systems
Resource Management
Datacenter Architecture

BACKGROUND: PTHREADS

```
void *myThreadFun(void *vargp)
{
    sleep(1);
    printf(“Hello World\n");
    return NULL;
}

int main()
{
    pthread_t thread_id_1, thread_id_2;
    pthread_create(&thread_id_1, NULL, myThreadFun, NULL);
    pthread_create(&thread_id_2, NULL, myThreadFun, NULL);
    pthread_join(thread_id_1, NULL);
    pthread_join(thread_id_2, NULL);
    exit(0);
}
```

BACKGROUND: MPI

```
int main(int argc, char** argv) {
    MPI_Init(NULL, NULL);

    // Get the number of processes
    int world_size;
    MPI_Comm_size(MPI_COMM_WORLD, &world_size);

    // Get the rank of the process
    int world_rank;
    MPI_Comm_rank(MPI_COMM_WORLD, &world_rank);

    // Print off a hello world message
    printf("Hello world from rank %d out of %d processors\n",
           world_rank, world_size);

    // Finalize the MPI environment.
    MPI_Finalize();
}
```

```
mpirun -n 4 -f host_file ./mpi_hello_world
```

MOTIVATION

Build Google Web Search

- Crawl documents, build inverted indexes etc.

Need for

- automatic parallelization
- network, disk optimization
- handling of machine failures

OUTLINE

- Programming Model
- Execution Overview
- Fault Tolerance
- Optimizations

PROGRAMMING MODEL

Data type: Each record is (key, value)

Map function:

$$(K_{in}, V_{in}) \rightarrow \text{list}(K_{inter}, V_{inter})$$

Reduce function:

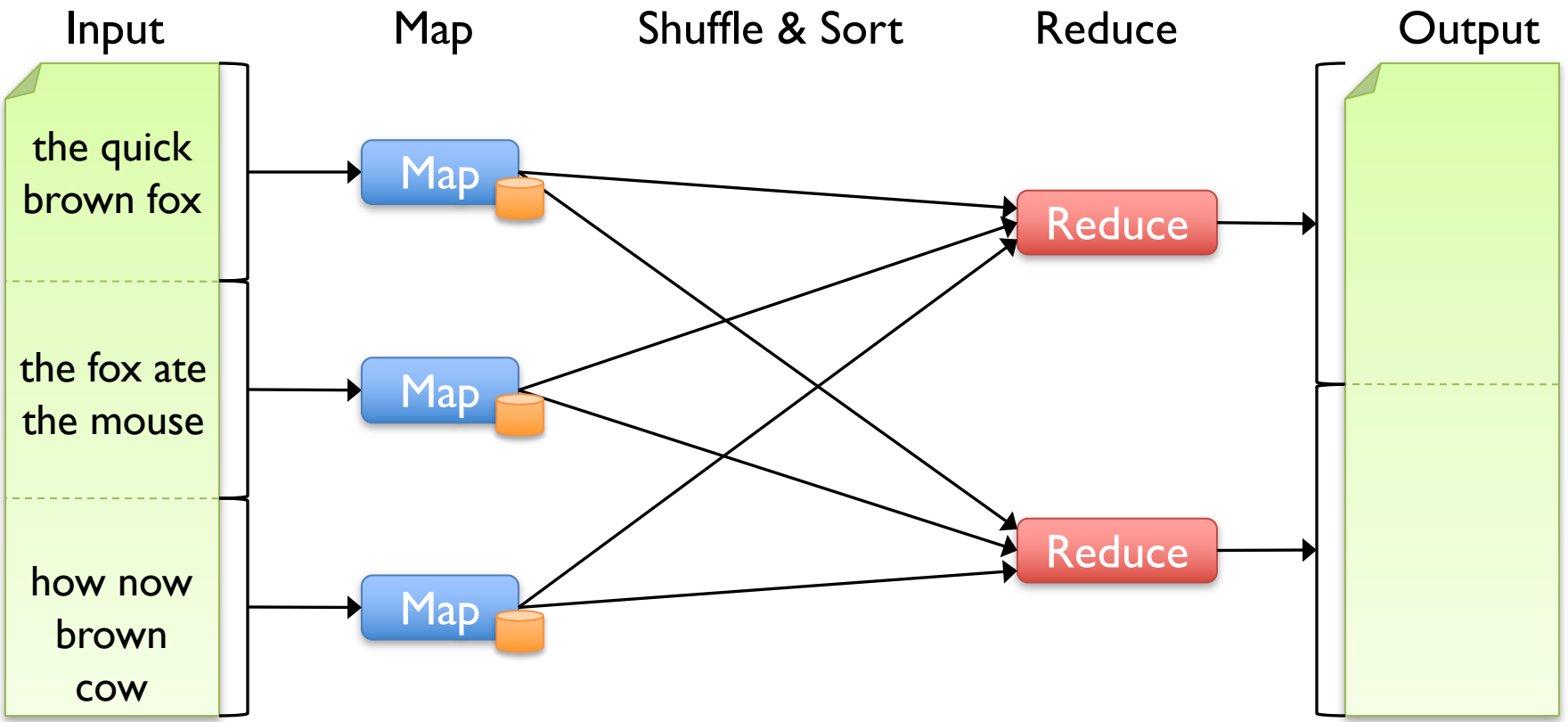
$$(K_{inter}, \text{list}(V_{inter})) \rightarrow \text{list}(K_{out}, V_{out})$$

EXAMPLE: WORD COUNT

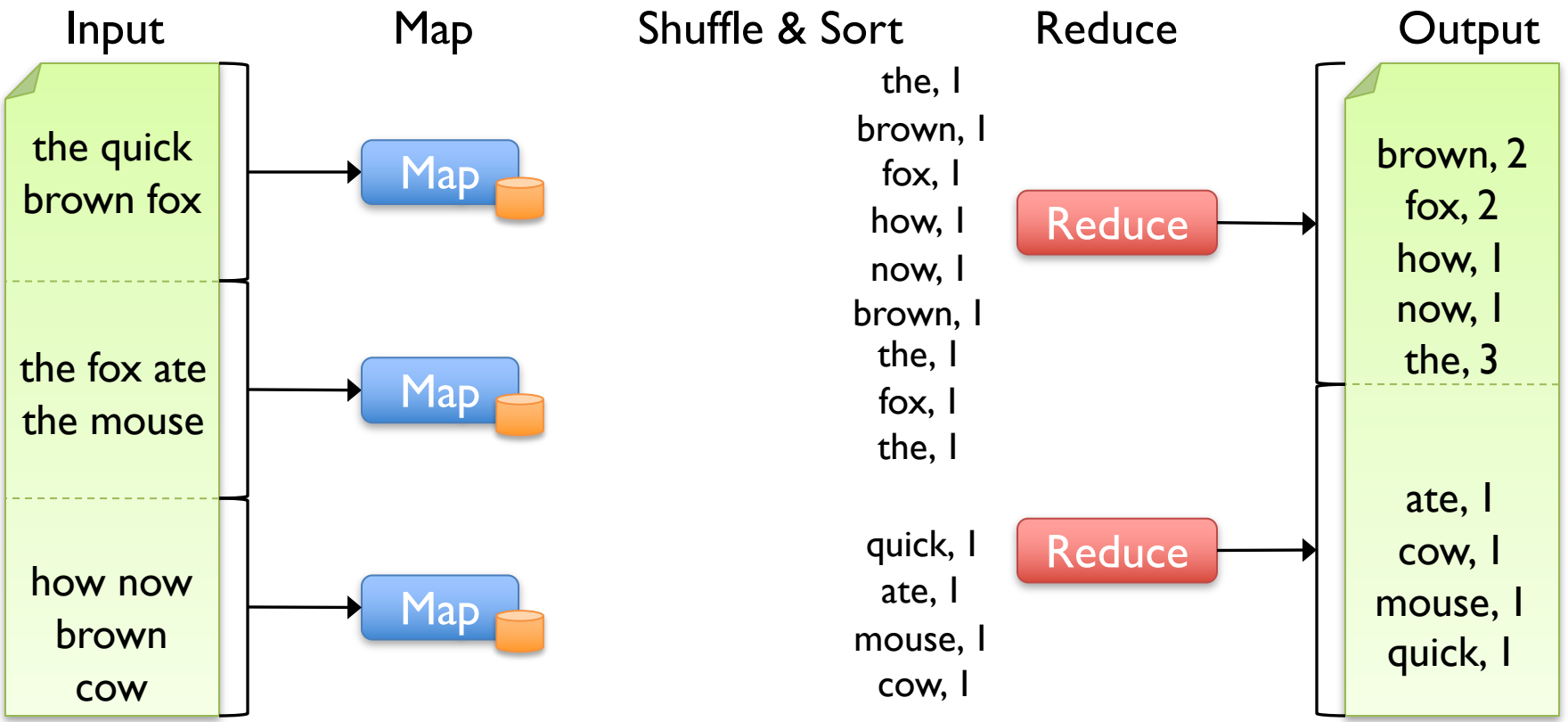
```
def mapper(line):
    for word in line.split():
        output(word, 1)


def reducer(key, values):
    output(key, sum(values))
```

WORD COUNT EXECUTION



WORD COUNT EXECUTION

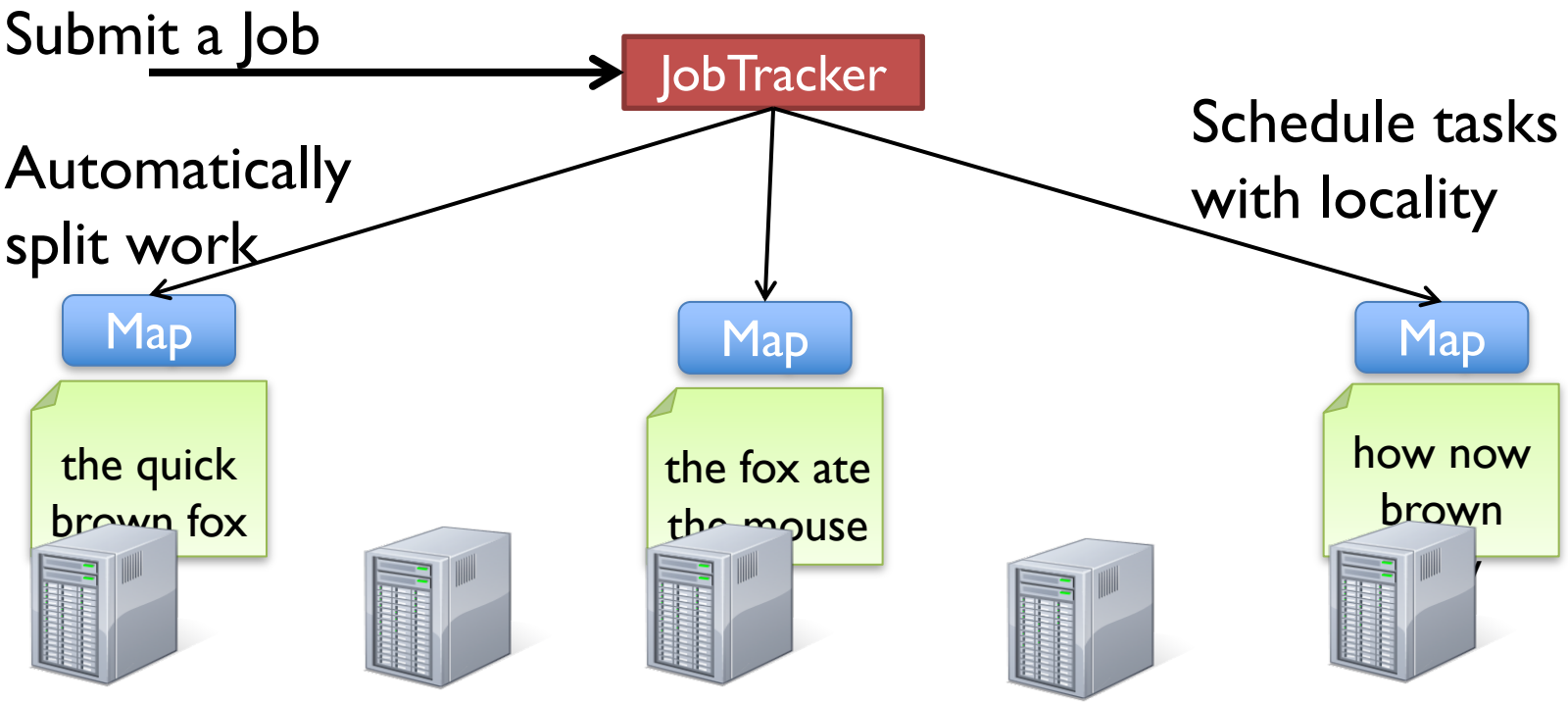


ASSUMPTIONS

ASSUMPTIONS

1. Commodity networking, less bisection bandwidth
2. Failures are common
3. Local storage is cheap
4. Replicated FS

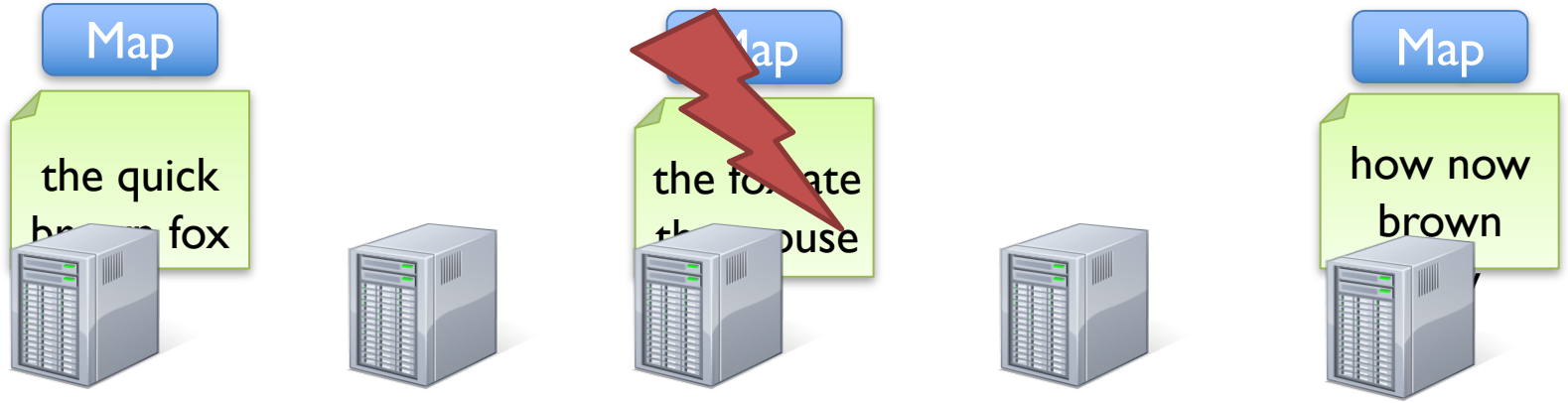
WORD COUNT EXECUTION



FAULT RECOVERY

If a task crashes:

- Retry on another node
- If the same task repeatedly fails, end the job

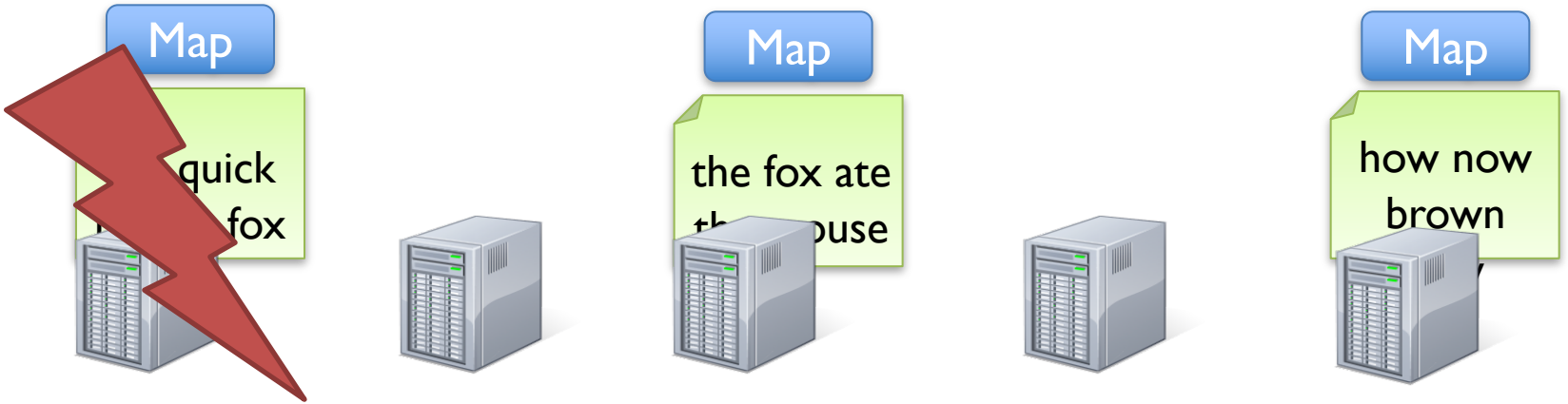


FAULT RECOVERY

If a node crashes:

- Relaunch its current tasks on other nodes

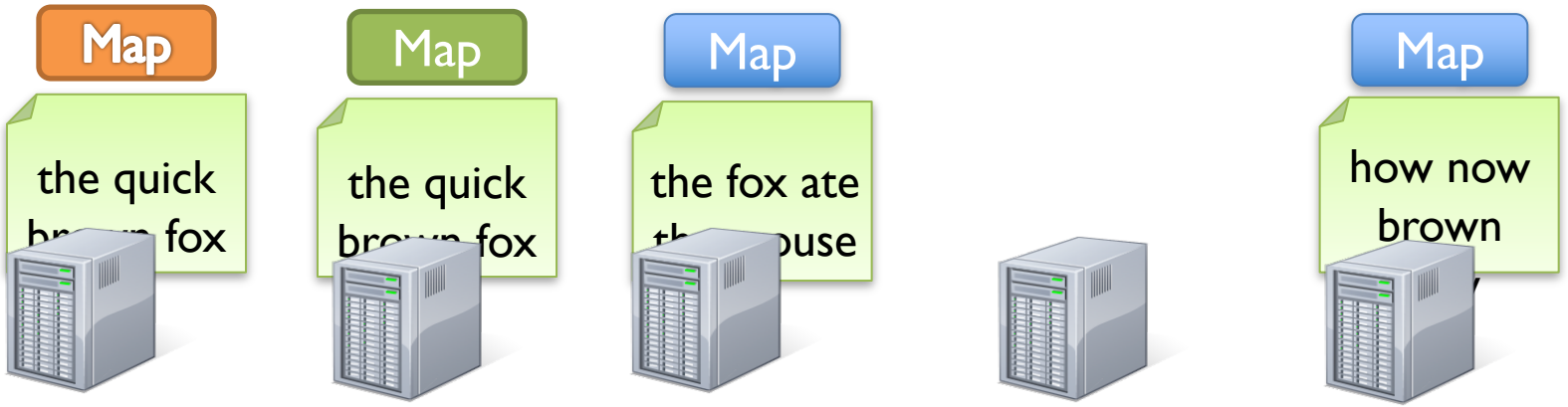
What about task inputs ? File system replication



FAULT RECOVERY

If a task is going slowly (straggler):

- Launch second copy of task on another node
- Take the output of whichever finishes first



MORE DESIGN

Master failure

Locality

Task Granularity

MAPREDUCE: SUMMARY

- Simplify programming on large clusters with frequent failures

- Limited but general functional API
  - Map, Reduce, Sort
  - No other synchronization / communication

- Fault recovery, straggler mitigation through retries

DISCUSSION

https://forms.gle/mAHD4QuMXko7vnjB6

DISCUSSION

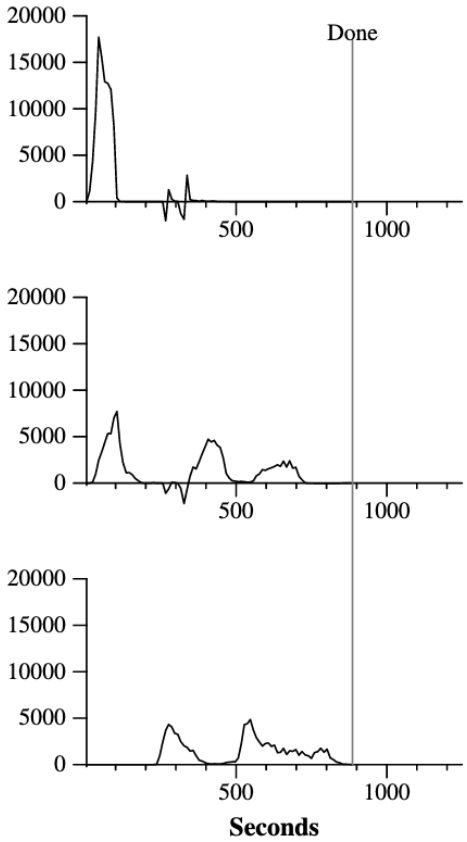
List one similarity and one difference between MPI and MapReduce

DISCUSSION

Indexing pipeline where you start with HTML documents. You want to index the documents after removing the most commonly occurring words.

1. Compute most common words.
2. Remove them and build the index.

What are the main shortcomings of using MapReduce to do this?



(c) 200 tasks killed

MapReduce Usage Statistics Over Time

	Aug, ‘04	Mar, ‘06	Sep, '07	Sep, ’09
Number of jobs	29K	171K	2,217K	3,467K
Average completion time (secs)	634	874	395	475
Machine years used	217	2,002	11,081	25,562
Input data read (TB)	3,288	52,254	403,152	544,130
Intermediate data (TB)	758	6,743	34,774	90,120
Output data written (TB)	193	2,970	14,018	57,520
Average worker machines	157	268	394	488

Jeff Dean, LADIS 2009

NEXT STEPS

- Next lecture: Spark
- Assignment 1: Use Piazza!