

61A Lecture 36

Wednesday, November 28

MapReduce

MapReduce is a *framework* for batch processing of Big Data.

What does that mean?

- **Framework:** A system used by programmers to build applications.
- **Batch processing:** All the data is available at the outset, and results aren't used until processing completes.
- **Big Data:** A buzzword used to describe data sets so large that they reveal facts about the world via statistical analysis.

The MapReduce idea:

- Data sets are too big to be analyzed by one machine.
- When using multiple machines, systems issues abound.
- Pure functions enable an abstraction barrier between data processing logic and distributed system administration.

(Demo)

Systems

Systems research enables the development of applications by defining and implementing abstractions:

- **Operating systems** provide a stable, consistent interface to unreliable, inconsistent hardware.
- **Networks** provide a simple, robust data transfer interface to constantly evolving communications infrastructure.
- **Databases** provide a declarative interface to software that stores and retrieves information efficiently.
- **Distributed systems** provide a single-entity-level interface to a cluster of multiple machines.

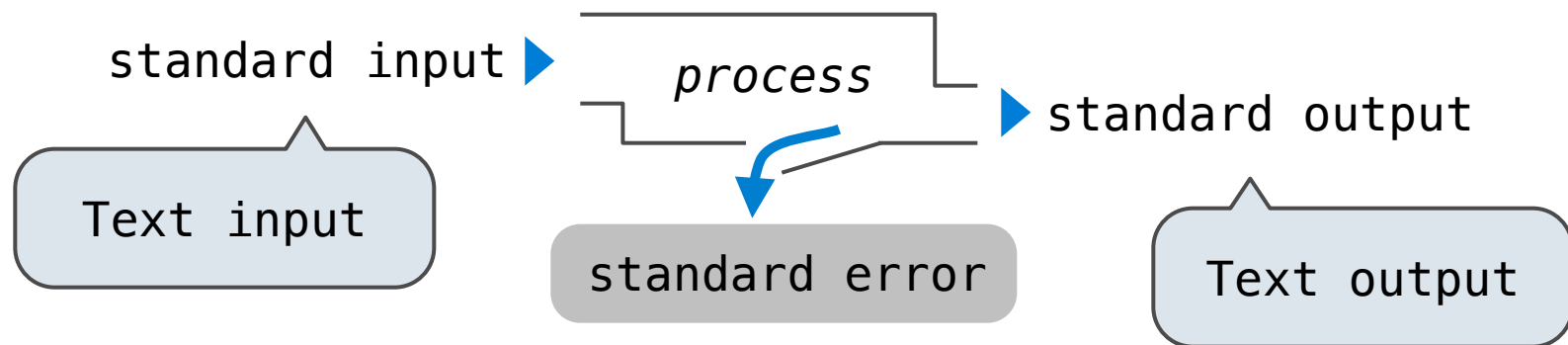
A unifying property of effective systems:

Hide *complexity*, but retain *flexibility*

The Unix Operating System

Essential features of the Unix operating system (and variants):

- **Portability:** The same operating system on different hardware.
- **Multi-Tasking:** Many processes run concurrently on a machine.
- **Plain Text:** Data is stored and shared in text format.
- **Modularity:** Small tools are composed flexibly via pipes.



The ***standard streams*** in a Unix-like operating system are conceptually similar to Python iterators.

(Demo)

Python Programs in a Unix Environment

The built-in `input` function reads a line from *standard input*.

The built-in `print` function writes a line to *standard output*.

(Demo)

The values `sys.stdin` and `sys.stdout` also provide access to the Unix *standard streams* as "files."

A Python "file" is an interface that supports iteration, read, and write methods.

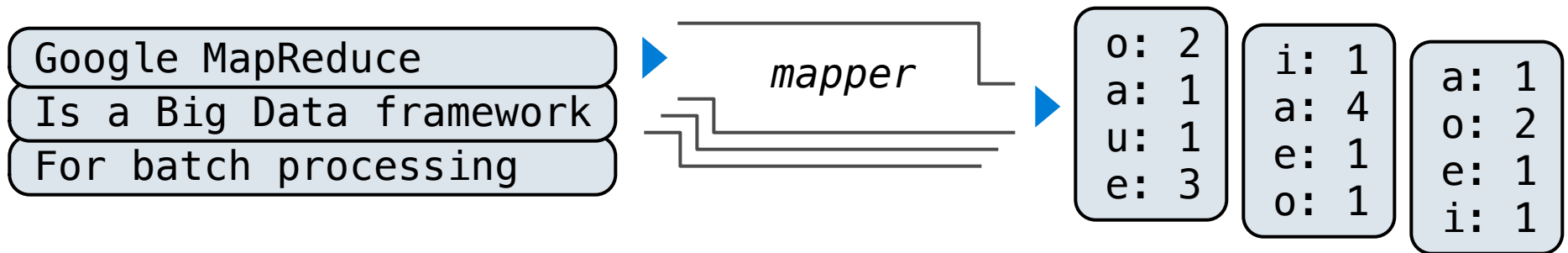
Using these "files" takes advantage of the operating system *standard stream* abstraction.

(Demo)

MapReduce Evaluation Model

Map phase: Apply a *mapper* function to inputs, emitting a set of **intermediate key-value pairs**.

- The *mapper* takes an iterator over inputs, such as text lines.
- The *mapper* yields zero or more **key-value pairs** per input.

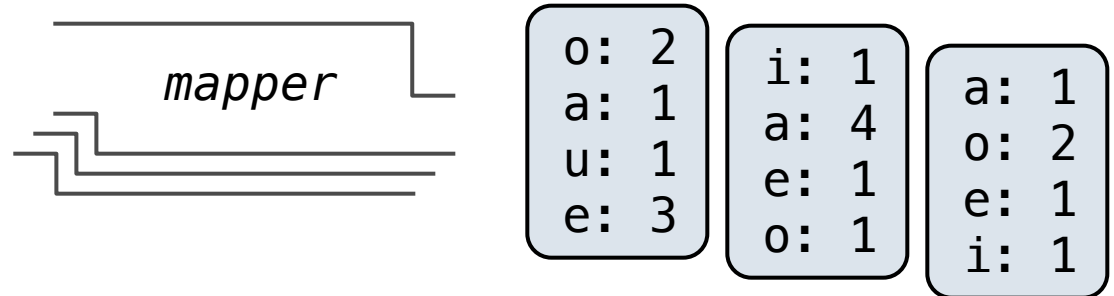


Reduce phase: For each **intermediate key**, apply a *reducer* function to accumulate all values associated with that key.

- The *reducer* takes an iterator over **key-value pairs**.
- All pairs with a given key are consecutive.
- The *reducer* yields 0 or more values, each associated with that **intermediate key**.

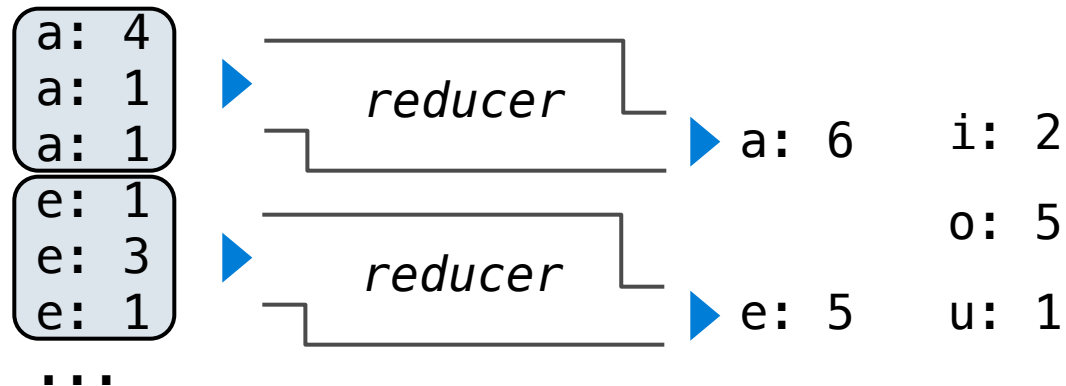
MapReduce Evaluation Model

Google MapReduce
Is a Big Data framework
For batch processing

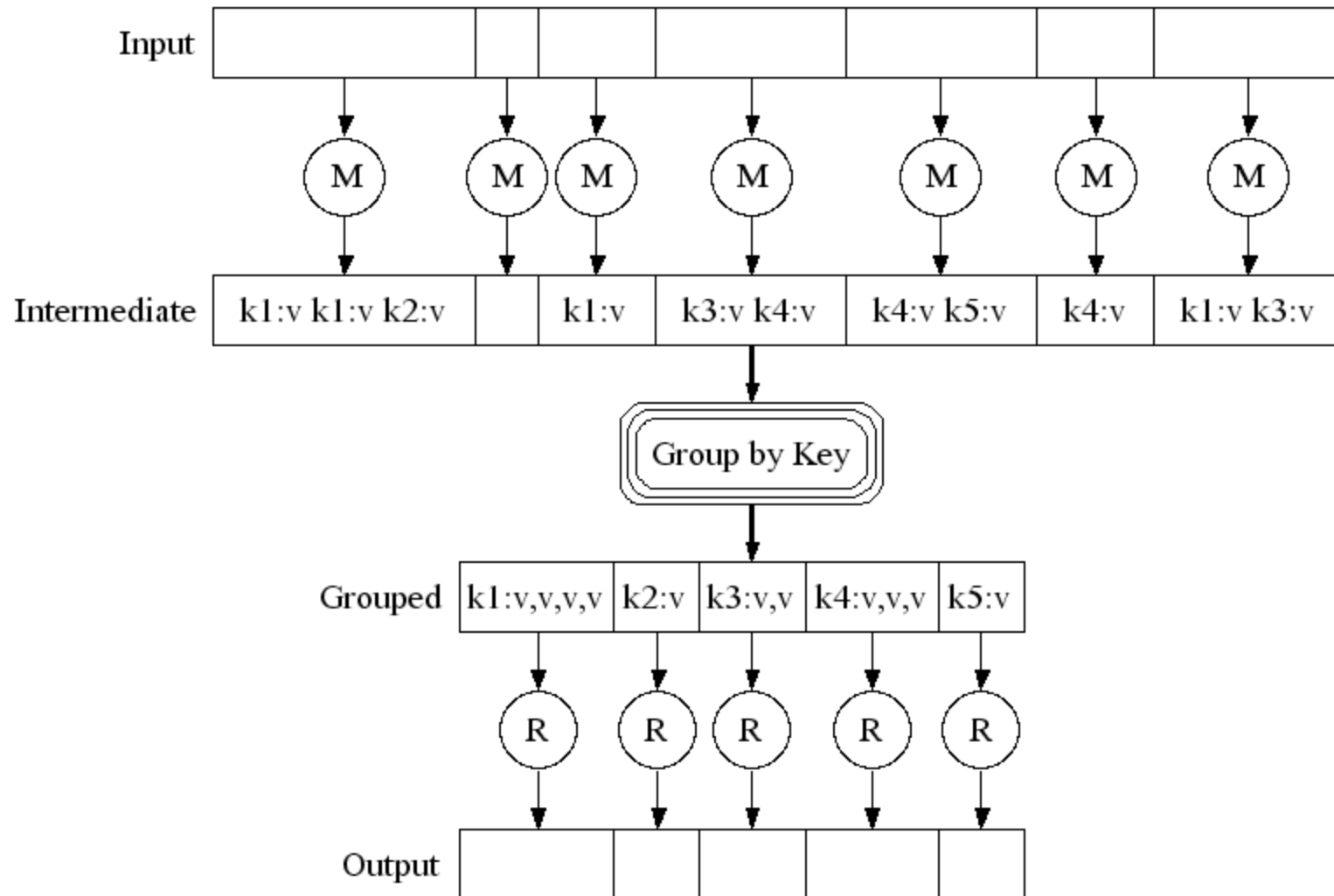


Reduce phase: For each *intermediate key*, apply a *reducer* function to accumulate all values associated with that key.

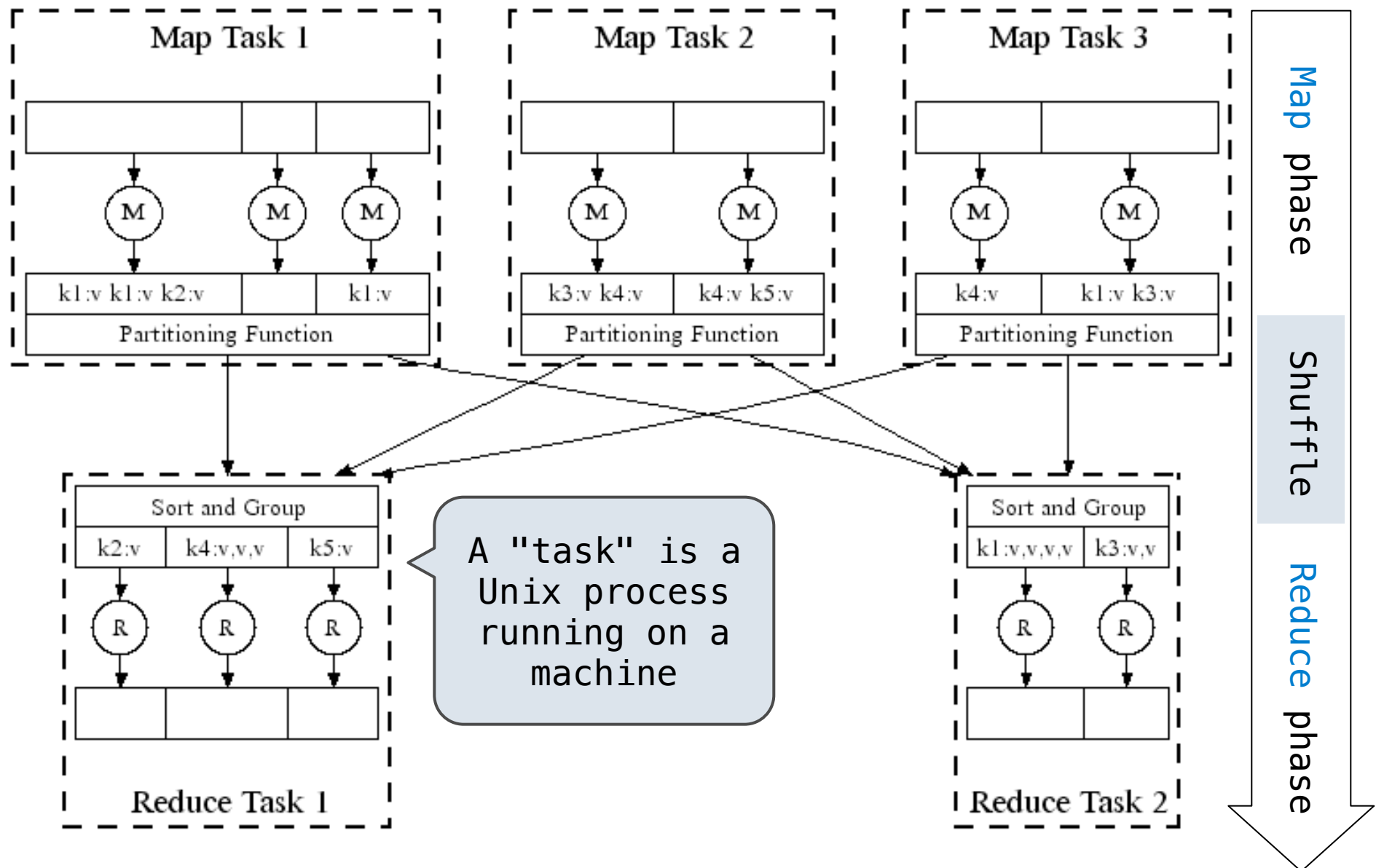
- The *reducer* takes an iterator over *key-value pairs*.
- All pairs with a given key are consecutive.
- The *reducer* yields 0 or more values, each associated with that *intermediate key*.



Above-the-Line: Execution model



Below-the-Line: Parallel Execution



MapReduce Assumptions

Constraints on the *mapper* and *reducer*:

- The *mapper* must be equivalent to applying a pure function to each input independently.
- The *reducer* must be equivalent to applying a pure function to the sequence of values for a key.

Benefits of functional programming:

- When a program contains only pure functions, call expressions can be evaluated in any order, lazily, and in parallel.
- Referential transparency: a call expression can be replaced by its value (or *vis versa*) without changing the program.

In MapReduce, these functional programming ideas allow:

- Consistent results, however computation is partitioned.
- Re-computation and caching of results, as needed.

Python Example of a MapReduce Application

The *mapper* and *reducer* are both self-contained Python programs.

- Read from *standard input* and write to *standard output*!

Mapper

```
#!/usr/bin/env python3
```

```
import sys
from ucb import main
from mapreduce import emit

def emit_vowels(line):
    for vowel in 'aeiou':
        count = line.count(vowel)
        if count > 0:
            emit(vowel, count)
```

```
for line in sys.stdin:
    emit_vowels(line)
```

Tell Unix: this is Python

The emit function outputs a key and value as a line of text to standard output

Mapper inputs are lines of text provided to standard input

Python Example of a MapReduce Application

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Reducer

```
#!/usr/bin/env python3
```

```
import sys
from ucb import main
from mapreduce import emit, group_values_by_key
```

Takes and returns iterators

Input: lines of text representing key-value pairs, grouped by key

Output: Iterator over (key, value_iterator) pairs that give all values for each key

```
for key, value_iterator in group_values_by_key(sys.stdin):
    emit(key, sum(value_iterator))
```

What Does the MapReduce Framework Provide

Fault tolerance: A machine or hard drive might crash.

- The MapReduce framework automatically re-runs failed tasks.

Speed: Some machine might be slow because it's overloaded.

- The framework can run multiple copies of a task and keep the result of the one that finishes first.

Network locality: Data transfer is expensive.

- The framework tries to schedule map tasks on the machines that hold the data to be processed.

Monitoring: Will my job finish before dinner?!?

- The framework provides a web-based interface describing jobs.

(Demo)