Introduction to Map Reduce

Sponsored by AWS

Continuum Analytics



MapReduce

• A story about **Big Data**?



MapReduce

- A story about **Big Data**?
- I Prefer Large Data
 - No Hype Machine
- Data of An Unsual Size
 - o C. Titus Brown





Can't fit into Excel



- Can't fit into Excel
 - Increase Memory



- Can't fit into Excel
 - Increase Memory
- Can't fit into R



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- Can't fit into Excel
 - Increase Memory
- Can't fit into R
 - Increase Memory
- Can't fit into Memory
 - Increase Memory
- Can't fit on a single disk
 - Distributed Filesystem: SAN, HDFS/DDFS, AWS: S3, Redshift, etc.



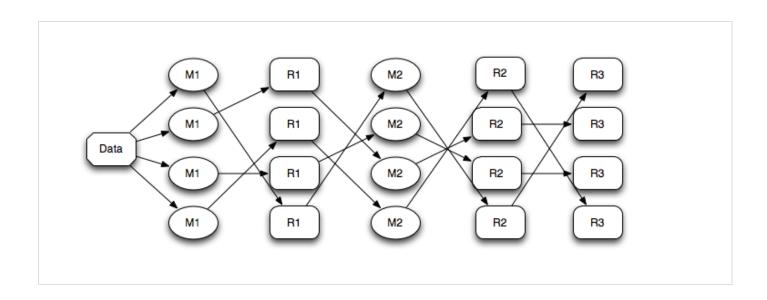
MapReduce

Framework to help solve the problem of distributed computation for distributed data

- A mass of data: records
- Split/Map records into key-values pairs
- Collect/Partition kv pairs (Optional Sort)
- Buckets are passed to Reduce function
- Result is returned



MapReduce Workflow



- Push Code to Data
- Lots of Network Traffic



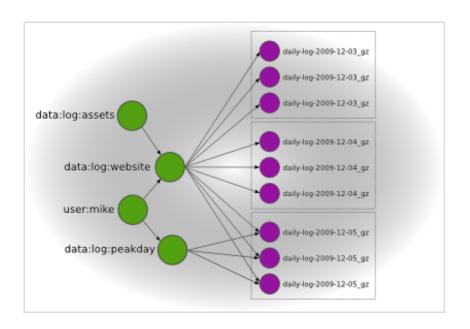
MR Implementations

- Disco: Python + Erlang
 - Distributed FileSystem: DDFS
- Hadoop: Java
 - Streaming with Python
 - Dumbo
 - MRJob
 - Hadoopy



DDFS

Tag based system



- ddfs push tag:mytag /the/path/data
- Replication spreads data across across distributed filesystem
- Comes with chunker



MapReduce: It's a Party



Batteries Included

- NumPy
- SciPy
- pandas
- scikits-learn
- OpenCV

• ...



Canonical Example

```
from disco.job import Job
    from disco.core import result iterator
    class WordCount(Job):
 6
        partitions = 3
        input=["sherlock.txt", "poirot.txt", "clouseau.txt"]
 7
 8
 9
        @staticmethod
10
        def map(line, params):
            import string
11
12
            for word in line.split():
13
                yield word, 1
14
15
        @staticmethod
16
        def reduce(iter, params):
17
            from disco.util import kvgroup
            for word, counts in kvgroup(sorted(iter)):
18
                 vield word, sum(counts)
19
20
21
    if __name__ == "__main__":
22
        from disco words import WordCount
23
24
        wordcount = WordCount().run()
25
26
        for (word, counts) in result_iterator(wordcount.wait(show=True)):
27
            print word, counts
28
```



Bitly Data

- URL shortening service
 - URLs with .gov/.mil are made public
- Typical Clickstream Data
 - JSON encoded
 - url
 - GeoLoc
 - Time
 - OS/Browser
 - 0 ...
- Thank you Wes McKinney



Demo 1



Questions

- What are the top 10 websites are for the day, month, quarter, etc.
- Most popular government institutions
 - NIH vs NASA vs CDC?
- Do most people look at links on mobile devices?
 - what is the break down of Windows vs OSX?
- Click through rate from Facebook vs twitter?
 Regional Questions
- NIH only popular in DC, NY, LA?
- what links are popular outside of the US



WikiLog Data

- Hourly summary statistics
- 4 fields: projectcode, pagename, pageviews, byte size

```
ca Casino_de_Manresa 1 8334
ca Caspar_David_Friedri 2 20242
ca Casquet_Glacial_Pata 1 8640
ca Castell d%27Eramprun 6 11885
```

Project codes can almost be tied to country/region



Demo 2

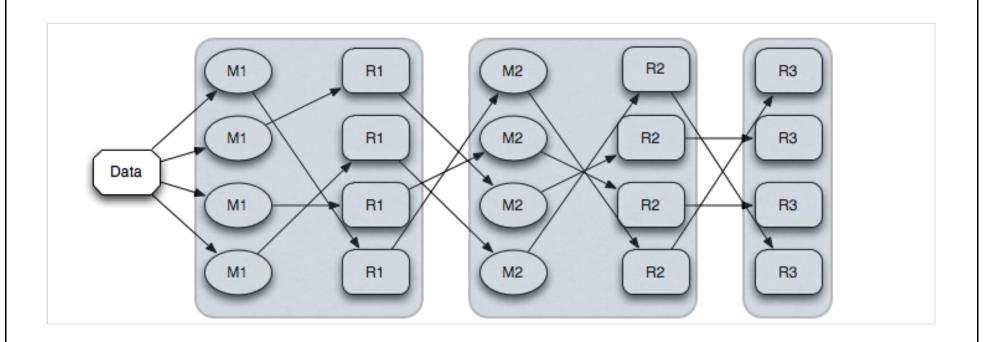


ETL Thoughts

- Try not to use MapReduce for ETL for every job
- Often not tied to SQL
- Be conscience of file storage
 - JSON
 - CSV
 - HDF5
 - AVRO
 - NETCDF4
 - NPZ -Compression
 - IOPS are expensive, GBs/TBs/PBs are cheap



Chaining Jobs



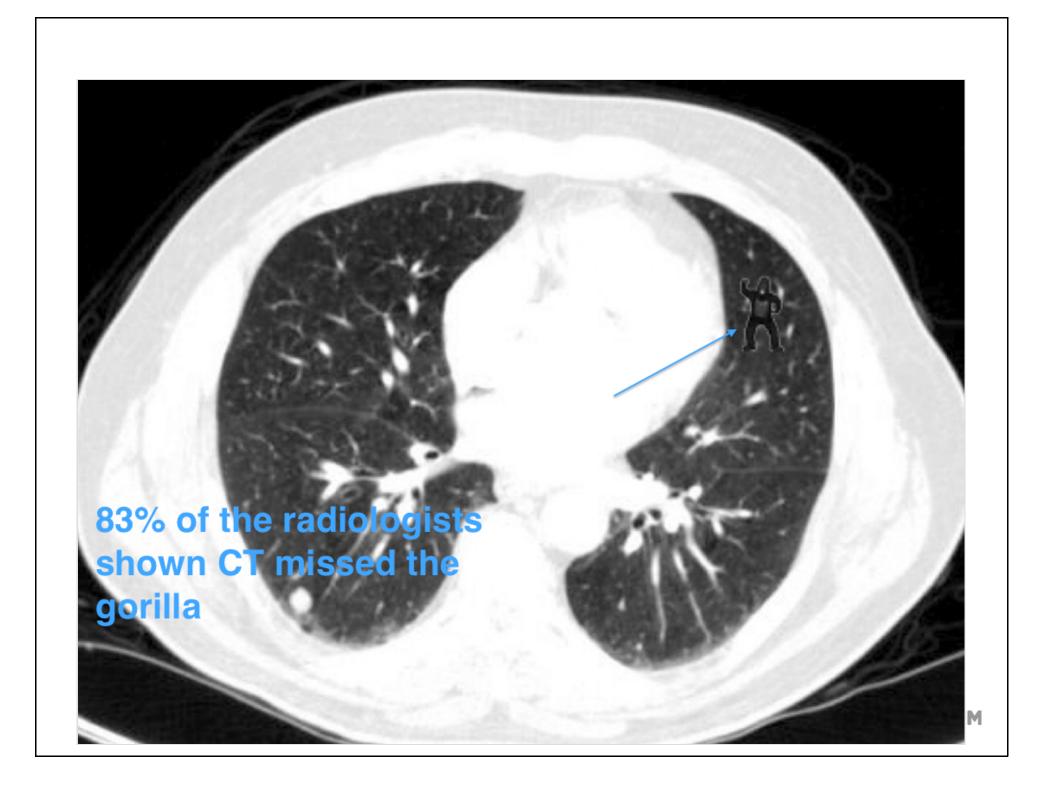


MapReduce Thoughts

- Data Cleansing
 - Everyone's pain point
- Task Deconstruction
 - Good for code management
 - Hides -- in a good way -- data management
- Can Be Inefficient
 - Network traffic
 - Job organization







Google Flu

Explore flu trends - United States

We've found that certain search terms are good indicators of flu activity. Google Flu Trends uses aggregated Google search data to estimate flu activity. Learn more »



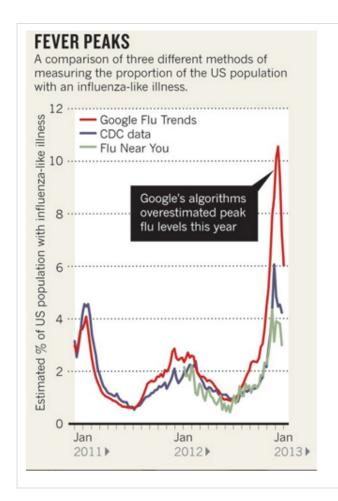
States | Cities (Experimental)



- Data Mining
- Faster than CDC



Google Gets It Wrong



- Typically, prediction is great!
- This year not so much
- Google: No comment!
- Feedback mechanism from hype-up media



Data Philosophy

- Invisible Gorillas will stay Invisible
 - Inattentional Blindess
- Machine Learning without Oversight
 - Turnkey analytics is dangerous
- Good Analysis
 - Requires iterative exploration
 - Peer review and collaboration



Thank you!

And thank you to the SciPy Sponsors and Organizing Committee

