CS 378 – Big Data Programming

Fall 2018 Lecture 1 Introduction

Class Logistics

- Class meets TTh, 9:30 AM 11:00 AM, PAR 203
- Office Hours GDC 6.402
 - T Th 11:00 12:00 AM
 - By appointment
 - Email: <u>dfranke@cs.utexas.edu</u>
 - Web page: cs.utexas.edu/~dfranke/courses/2018fall/cs378-BDP.htm
- TA: Vivek Pradhan
 - Office hours: TBD

Course Content

- Programming in Hadoop (map-reduce) and Spark
- Use ElasticMapReduce (EMR) on Amazon Web Services (AWS)
 - You an use a different Hadoop cloud service (like Azure)
 - You can us a local install of Hadoop if you want
- Local install of Spark (easiest)
 - Cloud based service from DataBricks (if you want)
 - Requires an AWS account

Textbooks

- MapReduce Design Patterns
 - Main content for Hadoop assignments
- Hadoop The Definitive Guide 4th Edition
 Recommended for your understanding, not required
- Learning Spark
 - Main content for Spark assignments
- All textbooks are available as ebooks from O'Reilly

Lectures

- PDF of (some) lecture notes accessible via syllabus
 - For your note taking, review, or whatever
 - cs.utexas.edu/~dfranke/courses/2018fall/cs378-BDP.htm

Assignments

- Assignments will be programming assignments
 - All work can be done using Java
 - Scala, Python are options for Spark
- IDE for developing code recommended
 - Eclipse, IntelliJ IDE (community edition) are free
 - Use maven to build "uber" JAR to upload to the cloud
 - I'll provide the pom.xml file used by maven

Assignments

- I'll review a solution in class on the due date
 - Work submitted after the start of class considered late
 - 25% penalty for late submission
 - Can be submitted until the next assignment is due
 - After that deadline, no credit is given
 - Will consider these in determining final grade
- I encourage you to keep pace with the assignments
 - Most assignments will build on previous work

Assignments

- We'll be using Hadoop for the first assignments
- You'll need to do one of:
 - Create a personal account for cloud-based Hadoop
 - AWS (Amazon Web Services), MicroSoft Azure, ...
 - Install Hadoop locally on your personal machine
 - I recommend you start on this now
- Instructions for cloud-based Hadoop account options or local Hadoop install are available in Assignment 0

• Questions?

Sources of "Big Data"

Taken from "How Much Data Do We Create Every Day? The Mind-Blowing Stats Everyone Should Read", by Bernard Marr, Forbes, May 21, 2018 www.forbes.com/sites/bernardmarr/2018/05/21/how-much-data-do-we-create-every-day-the-mind-blowing-stats-everyone-should-read Sourced from: www.domo.com/learn/data-never-sleeps-5

- 2.5 Quintillion bytes created every day (2017)
- 40K Google searches every second (3.5 billion/day)
- Social media (*every minute*):
 - Snapchat: users share >500K photos
 - LinkedIn: 120 members added
 - Youtube: > 4 million videos viewed
 - Twitter: >450K tweets sent
 - Instagram: ~47K photos posted
 - Facebook: 510K comments, 293K statuses updated

Sources of "Big Data"

Taken from "How Much Data Do We Create Every Day? The Mind-Blowing Stats Everyone Should Read", by Bernard Marr, Forbes, May 21, 2018 www.forbes.com/sites/bernardmarr/2018/05/21/how-much-data-do-we-create-every-day-the-mind-blowing-stats-everyone-should-read Sourced from: www.domo.com/learn/data-never-sleeps-5

Communication (<u>every minute</u>):

- 16 million text messages
- Tinder: 990K swipes
- 156 million emails sent
 - ~103 million spam emails
- 154K calls on Skype
- Services (*every minute*):
 - Spotify adds 13 new songs
 - Uber riders take ~46K trips
 - 600 pages edits on Wikipedia
- IoT (Internet of Things)
 - Fastest growing source of data

- Apart from just collecting and storing the data, we are also interested in processing that data.
- What can we do when the data gets big?
 - Too big for the CPU memory of any single machine
 - Larger than the disk storage of a single machine
- Big data is spread across a network of machines

Need to bring distributed storage and distributed processing to bear to handle big data

- Issues:
 - Distributing computation across many machines
 - Maximizing performance
 - Minimize I/O to disk, minimize transfers across the network
 - Combining the results of distributed computation
 - Recovering from failures

- We'll look at two popular tools/approaches
 - Hadoop
 - Spark
- Basic concepts of each
- How they address the aforementioned issues
- How to solve various problems with these two data processing/programming paradigms

- When writing a program with these tools ...
 - You don't know the size of the data
 - You don't know the extent of the parallelism
- Both try to collocate the computation with the data
 - Parallelize the I/O
 - Make the I/O local (versus across network)
- Data is often unstructured (vs. relational model)

Big Data vs. Relational

- RDBMS normalization
 - Goal is to remove redundancy and retain/insure integrity
- Big data apps want reads to be local
 - Send the code to the data, as it is much smaller (Jim Gray)
 - Normalization makes read non-local
- Processing examines one input record at a time
 Minimal state in programs it's in the data

Big Data Tools

- This all sounds great. What are the issues?
 - Coordinating the distributed computation
 - Handling partial failures
 - Combining the results of distributed computation
- Tools offer a programming model that abstracts
 - Disk read and write
 - Parallelization (computation and I/O)
 - Combining data (keys and values)

MapReduce Design Patterns

- Summarization
- Filtering
- Data Organization
 - Partitioning/binning, sorting, shuffle
- Joins
 - Merging data sets
- Meta-patterns
 - Optimizing map-reduce chains (data pipelines)

Resources for Hadoop

- *Hadoop: The Definitive Guide, 4th Edition,* by Tom White
 - O'Reilly Media
 - Print ISBN: 978-1-4919-0163-2 | ISBN 10: 1-4919-0163-2
 - Ebook ISBN: 978-1-4919-0162-5 | ISBN 10: 1-4919-0162-4
- *MapReduce Design Patterns,* by Donald Miner and Adam Shook
 - O'Reilly Media
 - Print ISBN: 978-1-4493-2717-0 | ISBN 10: 1-4493-2717-6
 - Ebook ISBN: 978-1-4493-4197-8 | ISBN 10: 1-4493-4197-7
- <u>http://hadoop.apache.org/</u>
- Several vendors provide Hadoop distributions
 - Cloudera, HortonWorks, MapR, ...
- Cloud-based Hadoop examples:
 - Amazon Web Services, MicroSoft Azure, ...

Resources for Spark

- Learning Spark, by Holden Karau, Andy Konwinsky, Patrick Wendell, Matei Zaharia
 - O'Reilly Media
 - Print ISBN: 978-1-4493-5862-4 | ISBN 10: 1-4493-5862-4
 - Ebook ISBN: 978-1-4493-5860-0 | ISBN 10: 1-4493-5860-8
- <u>http://spark.apache.org/</u>
 - Can download a version that runs on your local machine
- Cloud services
 - Spark on AWS, Azure
 - DataBricks offers a cloud service