

Course Information

• MSIS 2696: Big Data

• Class hours: Tuesday & Thurday 5:45pm

• Class room: Lucas 210

• Adjunct Lecturer: Mahmoud Parsian

Ph.D in Computer Science

LinkedIn profile: http://www.linkedin.com/in/mahmoudparsian

• Email: mparsian@yahoo.com

• GitHub: https://github.com/mahmoudparsian/BigData-MapReduce-Course

• Office Location: TBD

• Office Hours: TBD

Course Description:

http://www.scu.edu/business/msis/program/msis-specialization.cfm#faq8

Texts and Papers:

• BOOK-1: Big Data Now http://www.oreilly.com/data/free/files/bigdatanow2013.pdf

- BOOK-2: Data-Intensive Text Processing with MapReduce, Jimmy Lin and Chris Dyer http://lintool.github.io/MapReduceAlgorithms/ed1n/MapReduce-algorithms.pdf
- BOOK-3: Mining of Massive Datasets, Jure Leskovec, Anand Rajaraman, Jeffrey D. Ullman http://infolab.stanford.edu/~ullman/mmds/book.pdf

- PAPER-1: MapReduce: Simplified Data Processing on Large Clusters, Jeffrey Dean and Sanjay Ghemawat
 - http://lintool.github.io/MapReduce-course-2013s/material/Dean_Ghemawat_OSDI2004.pdf
- PAPER-2: Ullman. (2012) Designing Good Mapreduce Algorithms http://lintool.github.io/MapReduce-course-2013s/material/Ullman_2012.pdf
- PAPER-3: Large Language Models in Machine Translation, Thorsten Brants et. al. https://www.aclweb.org/anthology/D/D07/D07-1090.pdf
- PAPER-4: Bigtable: A Distributed Storage System for Structured Data http://lintool.github.io/MapReduce-course-2013s/material/ChangFay_etal_OSDI2006.pdf

Course Objectives:

At the completion of this course, students will be able to understand:

- 1. Elements of Big Data
- 2. Introduction to MapReduce, DAG
- 3. MapReduce algorithms and some design patterns
- 4. NoSQL Databases
- 5. Fundamentals of Hadoop framework
- 6. Fundamentals of Spark framework
- 7. Scale out vs. Scale up

Grade Distribution:

Assignment-1	10%
Assignment-2	10%
Assignment-3	10%
Assignment-4	10%
Assignment-5	10%
Midterm Exam	20%
Final Exam	30%

Letter Grade Distribution:

Course Policies:

- General
 - Quizzes and exams are closed book, closed notes.
 - No makeup quizzes or exams will be given.

• Grades

Grades in the C range represent performance that meets expectations; Grades in the
 B range represent performance that is substantially better than the expectations;
 Grades in the A range represent work that is excellent.

• Labs and Assignments

- Students are expected to work independently. Offering and accepting solutions from others is an act of plagiarism, which is a serious offense and all involved parties will be penalized according to the Academic Honesty Policy. Discussion amongst students is encouraged, but when in doubt, direct your questions to the professor.
- No late assignments will be accepted under any circumstances.

• Attendance and Absences

- Attendance is expected.
- Students are responsible for all missed work, regardless of the reason for absence. It is also the absence's responsibility to get all missing notes or materials.

• SANTA CLARA UNIVERSITY Academic Integrity Protocol:

http://www.scu.edu/provost/policy/academicpolicy/upload/revised-Academic-Integrity-protocol-6-17-12.pdf

• Instructor's Intended Purpose

The student's work must match the instructor's intended purpose for an assignment. While the instructor will establish the intent of an assignment, each student must clarify outstanding questions of that intent for a given assignment.

• Unauthorized/Excessive Assistance

The student may not give or get any unauthorized or excessive assistance in the preparation of any work.

• Authorship

The student must clearly establish authorship of a work. Referenced work must be clearly documented, cited, and attributed, regardless of media or distribution.

Tentative Course Outline:

The weekly coverage might change as it depends on the progress of the class. However, you must keep up with the reading assignments.

Session	Date	Content
1	January 6	 Introduction to Big Data Pages 1-40 of BOOK-1 Chapter 1 of BOOK-2 A Very Brief Introduction to MapReduce http://hci.stanford.edu/courses/cs448g/a2/files/map_reduce_tutorial.pdf Introduction to MapReduce http://lsd.ls.fi.upm.es/lsd/nuevas-tendencias-en-sistemas-distribuidos/IntroToMapReduce_2.pdf
2	January 8	 Introduction to MapReduce Pages 41-80 of BOOK-1 Chapters 2 of BOOK-2 Introduction to MapReduce and Hadoop by Matei Zaharia http://www.cs.berkeley.edu/~matei/talks/2010/amp_mapreduce.pdf
3	January 13	 Introduction to MapReduce Pages 81-120 of BOOK-1 Chapters 3 of BOOK-2 Chapters 2 of BOOK-3
4	January 15	 Introduction to MapReduce/Hadoop Pages 121-187 of BOOK-1 Chapters 4 of BOOK-2 Introduction to Hadoop
5	January 20	 Practice MapReduce/Spark/Hadoop Making Big Data Simple: by Matei Zaharia: https://www.youtube.com/watch?v=Nev1s6fHwMI Introduction to Spark: http://www.slideshare.net/jeykottalam/spark-sqlamp-camp2014?qid=a9e4b467-26b7-49a9-a361-8425876e9904&v=qf1&b=&from_search=4
6	January 22	 Spark/Hadoop Nuts and Bolts Making Big Data Simple: by Matei Zaharia: https://www.youtube.com/watch?v=Nev1s6fHwMI Introduction to Spark: http://www.slideshare.net/jeykottalam/spark-sqlamp-camp2014?qid=a9e4b467-26b7-49a9-a361-8425876e9904&v=qf1&b=&from_search=4 Parallel-Programming-With-Spark-Matei-Zaharia: http://ampcamp.berkeley.edu/wp-content/uploads/2013/02/Parallel-Programming-With-Spark-Matei-Zaharia-Strat.
7	January 27	 MapReduce Algorithm Design Chapter 3 of BOOK-2

8	January 29	 Language Models and Inverted Indexing Chapter 4 of BOOK-2 PAPER-3
9	February 3	 Midterm Exam Exams are closed book, closed notes, closed computers No makeup quizzes or exams will be given
10	February 5	• Similar Item Detection
11	February 10	• Clustering and Classification
12	February 12	• Clustering and Classification
13	February 17	 NoSQL, MongoDB, Redis Chapter 4 of BOOK-2 PAPER-4
14	February 19	• NoSQL, MongoDB, Redis
15	February 24	 Frequent Itemsets Chapter 6 of BOOK-3
16	February 26	 MapReduce Design Patterns Chapter 6 of BOOK-3
17	March 3	 Relational Algebra and MapReduce Chapter 6 of BOOK-2 Relational Algebra and MapReduce: http://www.eurecom.fr/~michiard/teaching/slides/clouds/tutorial-high_level.pdf MapReduce examples: http://courses.cs.washington.edu/courses/cse344/11sp/sections/section8/section8-mapreduce-solution.pdf MapReduce and relational algebra: http://www.cs.kent.edu/~jin/Cloud12Spring/DatabaseMapReduce.pptx
18	March 5	 Regression Algorithms Linear Regression Ttest
19	March 10	Overview of Big Data and MapReduce
	March 11	• Classes End
	March 16-19	Winter Final Examinations