



Course Information

- MSIS 2696: Big Data
- Class hours: Tuesday & Thursday 5:45pm - 7:00pm
- Class room: Lucas Hall 310
- 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: Lucas Hall 221X
- Office Hours: TBDL

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
8. Basics of Indexing (Solr/Lucene/ElasticSearch)

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:

A	4.0
A-	3.7
B+	3.3
B	3.0
B-	2.7
C+	2.3
C	2.0
C-	1.7
F	0.0

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 absentee'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	March 31	<ul style="list-style-type: none">• 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	April 2	<ul style="list-style-type: none">• 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	April 7	<ul style="list-style-type: none">• Introduction to MapReduce• Pages 81-120 of BOOK-1• Chapters 3 of BOOK-2• Chapters 2 of BOOK-3
4	April 9	<ul style="list-style-type: none">• Introduction to MapReduce/Hadoop• Pages 121-187 of BOOK-1• Chapters 4 of BOOK-2• Introduction to Hadoop
5	April 14	<ul style="list-style-type: none">• 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	April 16	<ul style="list-style-type: none">• 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-Strata-2013.pdf pptx
7	April 21	<ul style="list-style-type: none">• MapReduce Algorithm Design• Chapter 3 of BOOK-2

8	April 23	<ul style="list-style-type: none"> • Language Models and Inverted Indexing • Chapter 4 of BOOK-2 • PAPER-3
9	April 28	<ul style="list-style-type: none"> • Midterm Exam • Exams are closed book, closed notes, closed computers • No makeup quizzes or exams will be given
10	April 30	<ul style="list-style-type: none"> • Similar Item Detection • Clustering and Classification
11	May 5	<ul style="list-style-type: none"> • Introduction to Search-1: Solr/Lucene
12	May 7	<ul style="list-style-type: none"> • Introduction to Search-2: Solr/Lucene
13	May 12	<ul style="list-style-type: none"> • NoSQL, MongoDB, Redis • Chapter 4 of BOOK-2 • PAPER-4
14	May 14	<ul style="list-style-type: none"> • NoSQL, MongoDB, Redis
15	May 19	<ul style="list-style-type: none"> • Frequent Itemsets • Implementing Recommendation Systems using Big Data • Chapter 6 of BOOK-3
16	May 21	<ul style="list-style-type: none"> • Redis Server role for Big Data • Chapter 6 of BOOK-3
17	May 26	<ul style="list-style-type: none"> • 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	May 28	<ul style="list-style-type: none"> • Regression Algorithms • Linear Regression • Ttest
19	June 2	<ul style="list-style-type: none"> • Overview of Big Data and MapReduce
	June 4	<ul style="list-style-type: none"> • Classes End
	June 8-11	<ul style="list-style-type: none"> • Spring Final Examinations