Course Description
CSCE 587 – Big Data Analytics (3) Prereq: STAT 509 or STAT 515 or STAT 513. Foundational techniques and tools required for data science and big data analytics. Concepts, principles, and techniques applicable to any technology and industry for establishing a baseline that can be enhanced by future study.

Course Overview
This course introduces the student to concepts of big data management, database management, data mining techniques and the underlying statistics that support big data analytics. In this course we will use the programming language R as the primary tool for analysis.

Learning Outcomes
By the end of the course the student will be able to:
1. deploy a structured lifecycle approach to data science and big data analytics projects
2. select visualization techniques and tools to analyze big data and create statistical models
3. use tools such as R and RStudio, and MapReduce/Hadoop.

Required Texts, Other Materials, Suggested Readings
This course does not have a required text. However, ad hoc readings from the field will be assigned. In addition, material from “Data Science and Big Data Analytics Student Guide” distributed by EMC Education Services will be provided to the students.

Course Delivery Structure
The course will be delivered in a computer-equipped classroom. Approximately 50% of the time will be devoted to lecture and the other 50% devoted to the supervised working through of exercises.

Course Requirements
Readings: Students will read lecture material assigned for each class prior to the class.
Homework: Students will complete assignments demonstrating mastery of material. These will be due at the beginning of class.

Course Outline/Schedule
Lecture 1: Introduction to Big Data Analytics
Lecture 2: DBMS Overview
Lecture 3: Introduction to R and RStudio
Lecture 4: Basic analysis in R
Lecture 5: Intermediate R
Lecture 6: Intermediate analysis in R
Lecture 7: Visualization and Data Exploration
Lecture 8: K-means Clustering.
Lecture 9: Independent Sample Tests
Lecture 10: Basic Association Analysis
Lecture 11: Association Rule Speedup
Lecture 12: Linear regression part 1
Lecture 13: Linear regression part 2
Lecture 14: Logistic regression
Lecture 15: Naïve Bayes
Lecture 16: Decision trees part 1
Lecture 17: Decision trees part 2
Lecture 18: Review for Midterm Exam
Lecture 19: Midterm Exam
Lecture 20: Introduction to Hadoop and HDFS
Lecture 21: Using R with Hadoop
Lecture 22: First R/Hadoop program
Lecture 23: Intermediate R/Hadoop programming
Lecture 24: Pig, Hive, and HBase
Lecture 25: Discussion of rmr2 Project
Lecture 26: Support Vector Machines Part 1
Lecture 27: Support Vector Machines Part 2
Lecture 28: Review for Final Exam

Assignments

Readings: Students will read lecture material assigned for each class prior to the class. Readings will be assigned at the end of the preceding class.

Homework: Students will complete assignments demonstrating mastery of material. These will be due at the beginning of class. Graded written evaluations will be returned one week after submission.

HW 1: R Homework assignment
HW 2: K-means homework assignment
HW 3: Association Analysis homework assignment
HW 4: Linear and logistic regression homework assignment
HW 5: Naïve Bayes homework assignment
HW 6: Decision tree homework assignment.
HW 7: rmr2 Project
Midterm exam: Covers lectures 1 – 17: In-class exam as well as take-home applied-exam
Final exam: Covers entire semester: In-class exam as well as take-home applied-exam

Graduate students will be given a special project homework that involves the analysis of a sample dataset.
Grading Scheme
Final grade: 90 <= A, 87 <= B+ <90, 80<= B < 87, 77 <= C+ < 80, 65<= C < 77, 60 <= D+ <65, 50 <= D < 60, F < 50

Grades for undergraduates will be calculated from homework (50%), midterm (20%), final exam (30%).

Grades for graduate students will be calculated from homework (40%), midterm (20%), final exam (30%), graduate project (10%).

Difference between Undergraduate and Graduate Work:
Graduate students are assigned additional problems, of greater depth and difficulty, in both homework and exams.

Course Policies
Attendance: Attendance is mandatory. Students will be expected to have read the material for each lecture prior to the lecture and to be able to actively participate in discussions during class.

Tardiness, late assignments: homework is due at the beginning of class. Late assignments will be charged 20% per day.

Violations of academic honesty: Assignments and examination work are expected to be the sole effort of the student submitting the work. Students are expected to follow the University of South Carolina Honor Code and should expect that every instance of a suspected violation will be reported. Students found responsible for violations of the Code will be subject to academic penalty under the Code in addition to whatever disciplinary sanctions are applied.

Policy on disabilities or special needs: Any student with a documented disability should contact the Office of Student Disability Services at 803-777-6142 to make arrangements for appropriate accommodations.