



Course Syllabus: Data Analytics - CS 220

Division	Computer, Electrical and Mathematical Sciences & Engineering
Course Number	CS 220
Course Title	Data Analytics
Event Details	Lecture CS 220
Academic Semester	Fall
Academic Year	2016/2017
Semester Start Date	08/21/2016
Semester End Date	12/13/2016
Class Schedule (Days & Time)	10:30 AM - 12:00 PM Mon Wed

Instructor(s)				
Name	Email	Phone	Office Location	Office Hours
Xin Gao	Xin.Gao@kaust.edu.sa		4217, 3, Ibn Sina (bldg. 3)	Time: Tuesday 1-2pm. Location: Building 3, Level 4, Room 4217 (sea side). Tel: 012-8080323.

Teaching Assistant(s)	
Name	Email
Yu Li; Fatima Smaili; Ibrahim Alabdulmohsin.	yu.li@kaust.edu.sa; fatimazohra.smaili@kaust.edu.sa; ibrahim.alabdulmohsin@kaust.edu.sa.

Course Information

Comprehensive Course Description	<p>Motivation for Establishing the Course: This course will provide an introduction to the basics and the principles of artificial intelligence, data mining and machine learning. These areas are well aligned with the research in CEMSE Division and the missions of KAUST. This course also fits well in the fourth paradigm of science: data-intensive scientific discovery.</p> <p>Course Description: This course covers the basic concepts and algorithms for artificial intelligence, data mining and machine learning. The main contents of the course are:</p> <ol style="list-style-type: none"> 1. Artificial intelligence: <ul style="list-style-type: none"> › Task environment › Performance measure › Problem solving by searching <ul style="list-style-type: none"> › Uninformed search › Informed search <p>-Constraint satisfaction problems</p> <ol style="list-style-type: none"> 2. Data mining: <ul style="list-style-type: none"> › Data and patterns › Summary statistics and visualization › Unsupervised feature selection <ul style="list-style-type: none"> › Clustering › Supervised feature selection <ul style="list-style-type: none"> › Individual feature ranking › Feature subset selection › Machine learning: <ul style="list-style-type: none"> › Cross validation › Supervised learning <ul style="list-style-type: none"> › K-nearest neighbors › Naïve Bayes <p>-Decision trees</p> <ol style="list-style-type: none"> 1. Support vector machines 2. Neural networks
Course Description from Program Guide	<p>The course covers basic concepts and algorithms for artificial intelligence, data mining and machine learning. The main contents are: artificial intelligence (task environment, performance measure, and problem solving by searching), data mining (data and patterns, summary statistics and visualization, unsupervised feature selection, and supervised feature selection), and machine learning (cross validation and supervised learning).</p>
Goals and Objectives	<p>The goal of this course is to introduce the main ideas, fundamental concepts, and key algorithms in the fields of artificial intelligence, data mining, and machine learning. The course will prepare students to adapt to the big data era, to facilitate them to conduct research in data science, and to be able to apply the introduced algorithms in their own research areas.</p>
Required Knowledge	<p>Students who take this course are assumed to be familiar with algorithm runtime analysis (e.g., big O notations), probability theory (e.g., Gaussian distribution and conditional probability), and programming language (e.g., MATLAB or C++).</p>
Reference Texts	<p>There will be no textbook required for this course. The lectures and slides are self-contained. The reference books are "Artificial Intelligence – A Modern Approach" by Stuart Russell and Peter Norvig, the Prentice Hall press, and "Pattern Recognition and Machine Learning" by Christopher M. Bishop, the Oxford University press.</p>
Method of evaluation (Percentages & Graded content such as Assignments, Oral quizzes, Projects, Midterm exam, Final Exam, Attendance and participation, etc.)	<p>40.00% - Final exam 20.00% - Research Project 10.00% - Quiz(zes) 20.00% - Midterm exam 10.00% - Homework /Assignments</p>
Nature of the assignments (assigned reading, case study, paper presentation, group project, written assignment, etc.)	<p>There will be written assignments with both theoretical and practical questions.</p>

Course Policies (Absences, Assignments, late work policy, etc.)	The final grade is composed of 10% assignment, 20% midterm, 10% in-class quiz, 20% semester-long project, and 40% final exam. The final exam of this course is also one of the three exams for the Ph.D. qualification exam in the CS program.
Additional Information	

Tentative Course Schedule
(Time, topic/emphasis & resources)

Week/Lecture	Topic
1	Introduction and task environment
2	Performance measure and problem solving by searching
3	Uninformed search
4	Informed search
5	Local search
6	Constraint satisfaction problems
7	Data mining and summary statistics
8	Data visualization and exploration
9	Unsupervised feature selection
10	Individual feature ranking
11	Feature subset selection
12	Density estimation and cross-validation
13	K-nearest neighbor and naive Bayes
14	Decision trees
15	Support vector machines
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Note

The instructor reserves the right to make changes to this syllabus as necessary.