

CE 263N Scalable Spatial Analytics

Lecture: MW 11 – 12 pm, 212 O'Brien
Seminar: F 9 – 10 am, 212 O'Brien

Prerequisites

No formal prerequisites. (an undergraduate degree in engineering domain is expected, with knowledge of linear algebra and statistics, as well as some programming experience)

Primary goal: Provide theoretical background and hands-on experience in predictive modelling for designing scalable data-driven systems and location-based services based on data analytics.

Short Description: Introduction to modern methods of data analysis, spatial data handling and visualization technologies for engineers and data scientists. Theoretical coverage includes a selection of methods from spatial statistics, exploratory data analysis, spatial data mining, discriminative and generative approaches of machine learning. Projects and assignment tasks are targeted at real-world scalable implementation of systems and services based on data analytics in environmental remote sensing, transportation, energy, location-based services and the domain of “smart cities” in general.

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OH: M W F 4:00-5:30

Course Reader

The course readings are available on bCourses as pdf files and/or links to online materials.

Course Grading

1 or 2 Midterms	30%
Homework (6 problems)	50%
Final Project	20%

Course Schedule

LECTURE	DATE	TOPIC	THEME	PROBLEMS DUE
1	9/03	Introduction	General	
2	9/08	Exploratory data analysis	EDA	
3	9/10	Clustering	EDA	
4	9/15	Dimensionality reduction	EDA	
5	9/17	Spatial databases I	Data Handling	
6	9/22	Spatial databases II	Data Handling	
7	9/24	GIS and web mapping	Data Handling	
8	9/29	Trajectories and map matching	Data Processing	
9	10/01	Map matching and reconstruction	Data Processing	
10	10/06	Point Pattern Analysis	Spatial Statistics	
11	10/08	Geostatistics	Spatial Statistics	
12	10/13	Spatial Regression	Spatial Statistics	
13	10/15	Geographically weighted models	Spatial Statistics	
14	10/20	Intro to ML, model selection	ML discriminative	
15	10/22	Locally linear models	ML discriminative	
16	10/27	Kernel methods I, SVM	ML discriminative	
17	10/29	Kernel methods II, SVR	ML discriminative	
18	11/03	Elements of computer vision	ML discriminative	
19	11/05	Mixture models, EM	ML generative	
20	11/10	MC, HMM, CRF	ML generative	
21	11/12	Spatial PLSA/LDA	ML generative	
22	11/17	Recommender Systems	ML generative	
23	11/19	Complex networks	Complex Systems	
24	11/24	Agent-based models	Complex Systems	
25	11/26	Catch-up	General	
26	12/01	Final projects	General	
27	12/03	Final projects	General	
	5/5	Review (Reading Week)		
	5/7	Review (Reading Week)		

Tentative Reading Schedule

SESSION	DATE	READINGS
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1	09/05	Breiman L., Statistical Modeling: The Two Cultures, Statistical Science, Vol. 16, No. 3, pp. 199–23, 2001.
2	09/08	EDA online resources
3	09/10	Scully D., Web-Scale K-Means Clustering. WWW'2010. Bottou L., Bengio Y., Convergence Properties of the k-Means Algorithms, NIPS'95. Ester M, Kriegel H-P., Sander J., Xu X., A density-based algorithm for discovering clusters in large spatial databases with noise. KDD-96
4	09/15	Burges C., Dimension Reduction: A Guided Tour, 2009 (selected chapters)
5	09/17	Web resources: Postgre/PostGIS
6	09/22	Web resources: QGIS
7	09/24	Web resources: MongoDB
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10		

Seminar Schedule

09/05 Statistical modelling: the two cultures. Assignments problems overview.

09/12 Problem 1 introduction: clustering

09/19 Clustering and dimensionality reduction examples

09/26 Problem 2 introduction: PostgreSQL / PostGIS, MongoDB

10/03

10/10 Problem 3 introduction

10/17

10/24 Problem 4 introduction

10/31

11/07 Problem 5 introduction

11/14

11/21 Problem 6 introduction

11/28

12/01 Final project presentations

12/03 Final project presentations