APPLIED DATA SCIENCE UG Minor

DSCI 353-453: Data Science Modeling, Prediction and Inference: for Energy & Manufacturing

Spring 2016  Tuesday, Thursday 11:30 am to 12:45 pm  Olin 303

**Prof. Roger H. French**, rxf131@case.edu, 3 Credits

This counts as a 5th level class in the **Applied Data Science UG Minor**

For more information see [http://datascience.case.edu/minor](http://datascience.case.edu/minor)

**Prerequisites:**

**Textbook:**  
- **OpenIntro Statistics**, D. M. Diez, C. D. Barr, M. Cetinkaya-Rundel  
- **An Introduction to Statistical Learning with Applications in R**, G. James, D. Witten, T. Hastie, R. Tibshirani  
- **The Art of Data Science**, R. D. Peng, E. Matsui  
- **Report Writing for Data Science in R**, R. D. Peng

**Course description:** Data science methods for inference, modeling and prediction.

In this course, we will use an open data science tool chain to develop reproducible data analyses useful for inference, modeling and prediction of the behavior of real energy and manufacturing systems. In addition to the standard data cleaning, assembly and exploratory data analysis steps essential to all data analyses, we will identify statistically significant relationships from datasets derived from population samples, and infer the reliability of these findings. We will use regression methods to model a number of both real-world and lab-based systems producing predictive models applicable in comparable populations. We will assemble and explore real-world datasets, use pair-wise plots to explore correlations, perform clustering, self-similarity, and logistic regression develop both fixed-effect and mixed-effect predictive models. We will also introduce machine-learning approaches for classification and tree-based methods. Results will be interpreted, visualized and discussed.

We will introduce the basic elements of data science and analytics using R Project for Statistical Computing. R Analytics will be applied to the case of energy systems (such as PV power plant degradation, and building energy efficiency) over time. And it will be applied to manufacturing systems to understand the principles of statistical process control and identify critical factors of variability and uniformity.
DSCI353/453: Data Science Modeling, Prediction and Inference: for Energy & Manufacturing

Roger French, Spring 2016
3 credit course, Undergraduate/Graduate
Instructor permission required
Format: Seminar

Course Description: Data Sci. Models & Prediction

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We will introduce the basic elements of data science and analytics using R Project for Statistical Computing. R is an open-source software project with broad abilities to access machine-readable open-data resources, data cleaning and assembly functions, and a rich selection of statistical packages, used for data analytics, model development, prediction, inference and clustering. This will include an introduction to R data types, reading and writing data, looping, plotting and regular expressions, so that one can start performing variable transformations for linear fitting and developing structural equation models, while exploring for statistically significant relationships.

R Analytics will be applied to the case of energy systems (such as PV power plant degradation, and building energy efficiency) over time, by analyzing system responses, combined with results of experiments to identify fundamental principles that are statistically significant in the observed system performance. And it will be applied to manufacturing systems to understand the principles of statistical process control and identify critical factors of variability and uniformity.
Learning Outcomes:
Familiarity with an open-data tool chain including
   R Statistics, scripting, functions, packages, automated data analysis,
   git versioning and Rmarkdown reproducible data science.
Familiarity with exploratory data analysis to guide data analysis

Familiarity with inference and significance of sample results to populations
Familiarity with regression and linear and non-linear statistical model building
   Including training, testing and validating dataset strategies
Applications of domain knowledge and statistical analytics
   To identify important predictors and develop initial predictive models
Familiarity with clustering, self-similarity methods
   For categorization by different distance metrics
Introduction to machine-learning approaches such as tree-based methods

Data types include:
   Time-series, spectral, image and higher order datatypes,
   And their assembly to produce augmented and derivative datasets.

Data set characteristics will include:
   Variety: Of types of information, including both structured and unstructured data,
   Volume: Data from human sources (vendors, suppliers, distributors, customers, etc.) and
   sensor networks of the energy system of factory, both small and large data volumes.
   Velocity: Energy system and manufacturing supply chains changes will be included.
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