Machine Learning: Uncovering Hidden Structure in Data
ICPSR Summer Program in Quantitative Methods of Social Research
29 July-2 August, 2019, Berkeley, CA

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Course Description:
The field of machine learning is most commonly associated with “big data”: how we can use massive
datasets to make better predictions about things like credit card fraud, Netflix recommendations,
and the like. Though machine learning has been most influential in its commercial and medical
applications, a growing number of social scientists are taking advantage of these methods to: (1)
uncover patterns and structure embedded in data, (2) test and improve model specification and
predictions, and (3) perform data reduction. This course covers the mechanics underlying machine
learning methods and discusses how these techniques can be leveraged by social scientists to gain
new insight from their data. Specifically, the course will cover: decision trees, random forests,
boosting, k-means clustering and nearest neighbors, support vector machines, kernels, neural
networks, and ensemble learning. We will also discuss topics related to best practices, including
error rates, cross-validation, and the use of bootstrapping methods to develop uncertainty estimates.
The course will demonstrate how to estimate these models in both R and Python, as well as methods
for interpreting and presenting model output.

Recommended Texts:
1. Hastie, Trevor, Robert Tibshirani, and Jerome Friedman. 2009. The Elements of Statistical
   MA: MIT Press.
   New York: Springer.
   Information Processing Systems 27, 2672-2680.
   Adversarial Examples.” In Proceedings of the 2015 International Conference on Learning


12. **R**
   
   (a) James, Gareth, Daniela Witten, Trevor Hastie, and Robert Tibshirani. 2013. *An Introduction to Statistical Learning with Applications in R*. New York: Springer.

13. **Python**
   

**Course materials**: Course materials (including slides, code, and problem sets) will be available on a private Dropbox folder.

**Tentative Schedule**:
This schedule is subject to change:

- **Monday, July 29: Machine Learning: Theory and Concepts**
  - Computational Learning Theory and the Development of Machine Learning
  - The Bias-Variance Tradeoff and Error Rates
  - Model Validation and Tuning
  - Resampling Techniques
  - Predictions and Counterfactuals
  - Quick Review of Linear Regression Models
  - R and Python
  - Computing Performance and Practical Tips

- **Tuesday, July 30: Supervised and Semi-Supervised Learning**
  - Generalized Linear Models and Extensions
  - Shrinkage/Regularization Methods and the Lasso
  - Naive Bayes
  - Neural Networks and the Perceptron
  - Support Vector Machines and Relevance Vector Machines
  - k-nearest Neighbors
• **Wednesday, July 31: Tree-Based Methods**
  Classification and Regression Trees
  Ensemble Methods: Random Forests and Boosting
  Assessing Variable Importance and Effects

• **Thursday, August 1: Unsupervised Learning I**
  \(k\)-means Clustering
  PCA and Factor Analysis
  Manifold Learning and Multidimensional Scaling

• **Friday, August 2: Unsupervised Learning II**
  Generative Adversarial Networks
  Deep Learning
  Mixture Models and Latent Class Analysis