

# OR 750. Advanced Data Analytics

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**Department of Systems Engineering and Operations Research**  
**George Mason University**  
**Fall 2017**

The class will consist of 7-8 lectures given by the instructor on several advanced topics in data analysis. The rest of the semester (another 6-7 lectures) students will present on the topic of their choice.

## Announcements

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- 5/3/2017: First class is on Aug 30 at 4:30pm

## List of topics

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- Probabilistic models for Machine Learning
  - Conjugate distributions, exponential family
  - Model choice
  - Hierarchical linear and generalize linear models (regression and classification)
  - Models for missing data (EM-algorithm)
  - LDA, Normal mixtures, Bayes PCA
  - Bayes computations (MCMC, Variational Bayes)
  - Graphical Models
  - Probabilistic modeling with Stan
- Deep learning
  - Optimization
  - Architectures (CNN, LSTM, MP, VAE)
  - Bayesian DL
  - Generative models (GANs)
  - Modeling with TensorFlow
- Filtering
  - Kalman Filter
  - Extended KF and ensemble KF
  - Particle filter
  - Modeling with DLM package in R
- Probability and Statistics methods for Decision Making

- Real-time hypothesis testing
- Brownian motion
- Bayesian methods for optimal stopping time detection
- Bayesian Optimization
  - Tuning machine learning algorithms
  - Engineering model calibration
  - Modeling with spearmint

## Course staff

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**Instructor:** [Vadim Sokolov](#)

**Office:** [Engineering Building](#), Room 2242

[vsokolov\(at\)gmu.edu](mailto:vsokolov(at)gmu.edu)

**Tel:** 703-993-4533

## Other Reading

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- Lectures on “Probability in High Dimension” ([pdf](#))
- Book on “High-Dimensional Probability” ([pdf](#))
- Blog post on SGD ([link](#))

## Office hours

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Vadim Sokolov: By appointment (at Engineering 2242)

## Lectures

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**Location:** [Nguyen Engineering Building](#) 1109

**Times:** 4:30-7:10pm on Wednesday

## Grades

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**Grade composition:** No in-class examination. Grade is based entirely on participation in class and homework assignments.

## Deep Learning

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- Normalization Propagation: A Parametric Technique for Removing Internal Covariate Shift in Deep Networks ([paper](#))
- Why does Monte Carlo Fail to Work Properly in High-Dimensional Optimization Problems? ([paper](#))
- Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift ([paper](#))
- Auto-Encoding Variational Bayes ([paper](#))

- A Central Limit Theorem for Convex Sets ([paper](#), [slides](#))
- Learning Deep Architectures for AI [monograph](#)
- Generative Adversarial Networks ([presentation](#))
- GANs at OpenAI ([blog](#))
- Tuning CNN architecture ([blog](#))
- Unsupervised learning ([blog](#))
- Deep Energy ([blog](#))
- DL Summer school 2015 ([videos](#))
- DL Representations ([blog](#))

## Decision Making

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- Multivariate Industrial Time Series with Cyber-Attack Simulation: Fault Detection Using an LSTM-based Predictive Data Model ([paper](#))

## Filtering

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- Curse-of-dimensionality revisited: Collapse of the particle filter in very large scale systems [paper](#)
- Can local particle filters beat the curse of dimensionality? [paper](#)

## Bayes

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- The Markov Chain Monte Carlo Revolution ([paper](#))
- Graphical Models, Exponential Families, and Variational Inference ([monograph](#))
- Variational Inference: A Review for Statisticians ([paper](#))

## Calibration

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- Hyperband ([demo](#))

## Tools

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- [Stan](#)
- [TensorFlow](#)
- [dlm \(R\)](#)
- [Keras](#)

## Misc Links

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- [Independent learning](#)

