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Introduction

Introduction

0.1 Programme of the course

Arguments treated:

- 1. **Motivation**: components of the learning problem and applications of Machine Learning. Supervised and unsupervised learning.
- 2. **Introduction**: the supervised learning problem, Data, Classes of models, Losses.
- 3. Probabilistic models and assumptions on the data: the regression function. Regression and Classification.
- 4. When is a model good?: model complexity, bias variance tradeoff/generalization (VC dimension, generalization error).
- 5. **Models for Regression**: linear Regression (scalar and multivariate), subset selection, linear-in-the-parameters models, regularization.
- 6. **Simple Models for Classification**: Logistic Regression, Perceptron, Naïve Bayes Classifier.
- 7. **Kernel Methods**: Support Vector Machines.
- 8. Neural Networks.
- 9. **Deep Learning**: Convolutional Neural Networks.
- 10. Validation and Model Selection: Generalization Error, Bias-Variance Tradeoff, Cross Validation. Model complexity determination.
- 11. **Unsupervised learning**: Cluster analysis, K-means Clustering, EM estimation.
- 12. Dimensionality reduction: Principal Component Analysis (PCA).

Bibliography