

FOUNDATIONS OF MACHINE LEARNING (AI60203)



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Course Logistics

Teachers: Mahesh Mohan M R

References

- [Introduction](#) to Machine Learning, Ethem Alpaydin, MIT Press ([available online](#))
- Machine Learning, Tom M Mitchell, Mc Graw Hill Education ([available online](#))

Theory

- Monday – 11:00-11:55 AM, Tuesday – 8:00-9:55 AM
- Evaluation: Class Test-1, Midsem, Class Test-2, and Endsem

Mini-Project

- Open ended (will be announced)

Office Hours: Fridays 4-5 PM based on [appointment](#)

[Course logistics](#)

[Introducton and Motivation](#)

Intro to ML

Motivations of ML. Why ML is trending now?

Classification of ML: Supervised, Unsupervised and Reinforcement Learning

Algorithm and its ineffectiveness for problems intuitive to humans (e.g., digit classification)

Basics of Supervised Learning: Labelled Data, Model, Loss, and Parameter Optimization

[Introduction](#)

Reading Exercise: Intro to ML by Ethem Alpaydin (first reference): Sec 1.1, Secs. 1.2.1 - 1.2.3.

[K Nearest Neighbor](#)

Instance based vs Model based Learning

Local vs Global Approximation of Target

K Nearest Neighbor for Classification and Regression

Effect of K (less K -- > sensitive to noise and large K --> sensitive to possibly irrelevant inputs)

Weighted [KNN](#) for Classification and Regression

Issues of [KNN](#)

Measurement Scales (Solution: min-max or z-score normalization)

Curse of Dimensionality

Expensive and Storage Need

[KNN](#)[KNN Study](#)

[Linear and Logistic Regression](#)

Motivation of Linear Models
Prerequisites of Linear Algebra
Vector Inner Product (for linear models with a single scalar output)
Matrix-vector multiplication (for linear models with multidimensional outputs)
Inner Product perspective --> for a single op value
Outer product perspective --> for the entire op
Perceptron
History
Pre-inner product Interpretation
Post-inner product Interpretation
Optimization
Gradient descent
Batch Gradient Descent
Stochastic Gradient Descent
Minibatch Gradient Descent
Closed Form Solution
Prerequisites of Probability
Probability and Random Variables
Probability Distribution (Joint, Marginal, and Conditional)
Bayes' Theorem
Independence
Mean, Variance, and Covariance
[Logistic Regression](#)
Linear vs [Logistic Regression](#)

[linear_reg](#)[Logistic Regression](#)

[Naive Bayes](#)

Naive Bayes' Model
Main assumption: Independence. Why?
Naive Bayes' Method for Categorical Inputs
Naive Bayes' Method for Continuous Inputs
Numerical Stability
Merits and Demerits

[naive_bayes](#)

[Decision Tree](#)

Decision Tree for Classification and Regression

Motivation for Decision Tree: Interpretability
Elements of Decision Tree: Root, Nodes, and Leaves
When to Split a Node?
Concept of Impurity -- Which feature to consider in a given node
Entropy Measure and other measures
Dealing with categorical as well as numerical features
Rule Extraction from Decision Trees

Regression Trees
Pruning to address overfitting

 [decision tree](#)

Bias Variance Tradeoff

Bias-variance Tradeoff
Graphical Illustration
Mathematical derivation
Addressing Bias-variance issues
Underfitting and Overfitting in various models
Training and Validation Sets
K-fold Cross validation (to eliminate the drawback of the above)

 [Bias Variance Tradeoff](#)

 [Proof](#)

SVMs

Motivation of SVMs
Hard-margin Hyperplane SVMs
Soft-margin Hyperplane SVMs
Kernel SVMs
Optimization Problems for different SVMs

 [SVM](#)

Feed Forward NN

1. XOR problem: Solution with Multi-layer Neural Networks
2. Biological Motivation of Feedforward Neural Network
3. Machine Learning Vs Deep Learning
 - Case study: Face detection
 - Advantages of Deep Learning over Machine Learning
 - Advantages of Machine Learning over Deep Learning
4. Feed Forward Neural network
 - Overview
 - Terminologies (input, output, and hidden layers, and learnable parameters)
 - Impact of Bias (similar to the threshold in Perceptron)
 - Extracting low-level to high-level features, and Connectionism
 - Forward Pass
 - Matrix multiplication based mapping
 - The need for Non-linearities between layers
 - Backpropagation for optimizing weights
 - Vanishing/Exploding Gradient
 - Gradients for MSE, Linear Non-linear Activations, and Linear Layer
 - Reuse of component Gradients for different parameters

 [Feedforward NN](#)

Convolutional and Recurrent NN

1. Biological Motivation

2. CNNs from MLPs
3. Different Layers of CNNs
4. Applications of CNNs for Classification and Regression
5. Standard [CNN](#) Architectures
 - AlexNet
 - VGGNet
 - GoogleNet
 - ResNet
6. Feedforward vs Recurrent Networks
7. Motivation for RNNs
8. Different [RNN](#) types and Applications
 - One-to-Many
 - Many-to-One
 - Many-to-Many
9. [RNN](#) model and parameters
10. Optimizing RNNs (Backpropagation through time)
11. Applications of RNNs

[CNN](#)[CNN Visualization](#)[RNN](#)

[Class Test 1 Evaluation](#)

[CT1](#)

[Ensemble Learning](#)

Motivation of [Ensemble](#) Learning
Insights on How Bagging and Boosting reduce Variance and Bias, respectively.
Random Forests
Adaboost

[Ensemble](#)

[Dimensionality Reduction](#)

Role of Dimensionality Reduction in Machine Learning
Feature Selection Vs Feature Extraction
Feature Selection
Forward Search
Backward Search
Feature Extraction
Principal Component Analysis (PCA)
Linear Discriminant Analysis (LDA)

[DimRed](#)

[Unsupervised Learning: Clustering](#)

[Clustering](#) -- Unsupervised Learning
Flat [Clustering](#)
K Nearest Neighbor --> Compactness

Spectral [Clustering](#) --> Connectivity
Mixture of Gaussian --> Soft [Clustering](#)
Hierarchical [Clustering](#)
Divisive --> top-down
Agglomerative --> bottom-up

 [Clustering](#)

[Course Project](#)

 [Course Project Slide](#)

 [Course Project Google Collab](#)

 [Results](#)

[Class Test 2 Evaluations](#)

 [Class Test 2](#)

 [Moodle Docs for this page](#)

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[Data retention summary](#)

