

ML in Fundamental ~~Methods~~ Physics - Summary / Overview lecture

→ Basics of Machine Learning:

- × how to approach problem in ML (split into training / test set)
- × Bias vs variance tradeoff

→ Optimisers:

- × gradient descent
- × Newton's method vs. gradient descent
- × Learning rate: choices
- × SGD
- × Modifications to gradient descent
 - × What's meant with momentum?
 - × ~~Heuristics~~ 2nd moment methods (RMS-Prop, Adam)
 - × What's the idea behind modifications?

→ Simple regression:

- × What's a regression problem
- × Least square regression
- × Ridge / Lasso regularisation
- × Bayesian formulation

Logistic / multi-class classification

- x Logistic binary classification with logistic sigmoid
- x MLE
- x Multi-class classification (softmax)
- x Perceptron as a binary classifier & perceptron capabilities
- x Shannon information content

Other classifiers / methods

- x SVM (kernel trick)
- x Decision trees (How do they work)
- x Bagging
- x Boosting
- x Random Forests

Neural Networks

- x Activation Functions
- x Dense Layers
- x Cost functions
- x Universal Approximation Theorem
- x Backpropagation - aka how to optimise a neural network
 - ↳ what can happen with gradients
- x Regularisation methods: l_1 , l_2 , dropout, batch-normalisation, training with mini-batches
- x ~~NN~~ Practicalities: How do you set up NN
- x Convolutional layer
 - ↳ Motivation
 - ↳ ~~Realisation~~ Implementation
 - ↳ Connection to dense layers
- x Pooling layers
- x Standard datasets: MNIST, ImageNet
- x Training strategies, hyperparameter tuning
- x RNNs (idea, problems, LSTM)

Unsupervised learning

→ What's that?

- PCA (what's interpretation of first component) [important]
- Multi-dim scaling
- K-means clustering
- Variations of clustering (agglomerative clustering, density based clustering)
- Autoencoders
- t-SNE

~~Advanced~~ Advanced topics

a) Variational methods

- x KL-divergence
- x variational free energy minimisation (def, why)
- x Example: var free energy for spin systems (mean field equations)
- x Hopfield networks (def, training, capacity)
- x Boltzmann-machines (defn.) / energy based models
→ Negative log-likelihood

- Example: Restricted Boltzmann Machines
(how to train them?)
- Defn. Deep Boltzmann machine
- Variational autoencoder (reparametrisation trick)
implementation

b) Generative Adversarial Networks

- Idea
- Minimisation objective
→ limit of JS-divergence
- Variation: Wasserstein-GAN (Idea?)
- Implementation (→ potential problems)

c) Reinforcement learning

- Vocabulary (agent/worker, environment, actions, policy, reward, return)
- General procedure
- Policies: ϵ -greedy, greedy
on-policy, off-policy (defn.)
- SARSA, Q-learning
- How do you deal with large state spaces?
→ (e.g. Atari games)

Applications of ML in fundamental Physics

- * ML in string landscape, model building, string geometries
- * ML & holography
- * ML & differential equations
- * ML and phase classification (supervised & unsupervised)