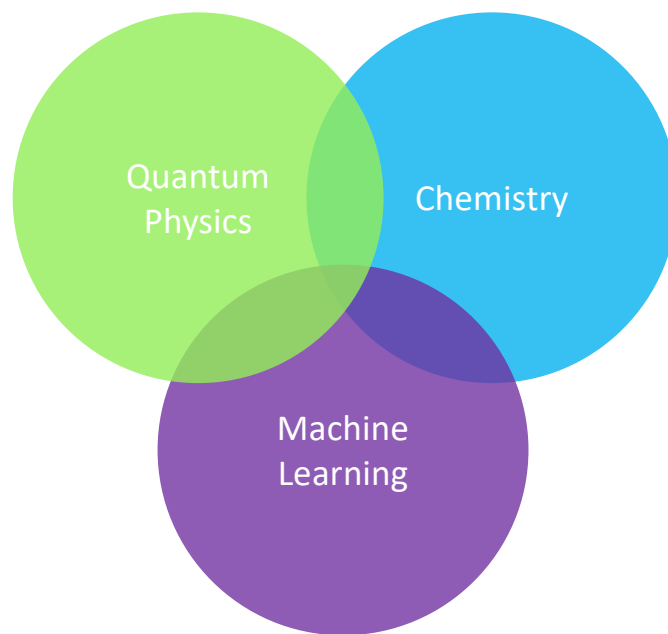


Chemistry 507B

Machine Learning for Chemistry

Professor Roman Krems



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Course outline

Major goals of this course:

- ⇒ Learn how machine learning (ML) works
- ⇒ Learn how ML can help solve Chemistry problems
- ⇒ Get hands-on experience with ML analysis
- ⇒ Apply ML to a research problem

This course will include three components:

I - Theoretical

- ⇒ Learn basic theory of ML
- ⇒ Discuss applications of ML to Chemistry

II - Practical

- ⇒ Learn to code ML models using python packages

III - Application of ML tools to a research problem

- ⇒ Use tools learned for a problem in your lab

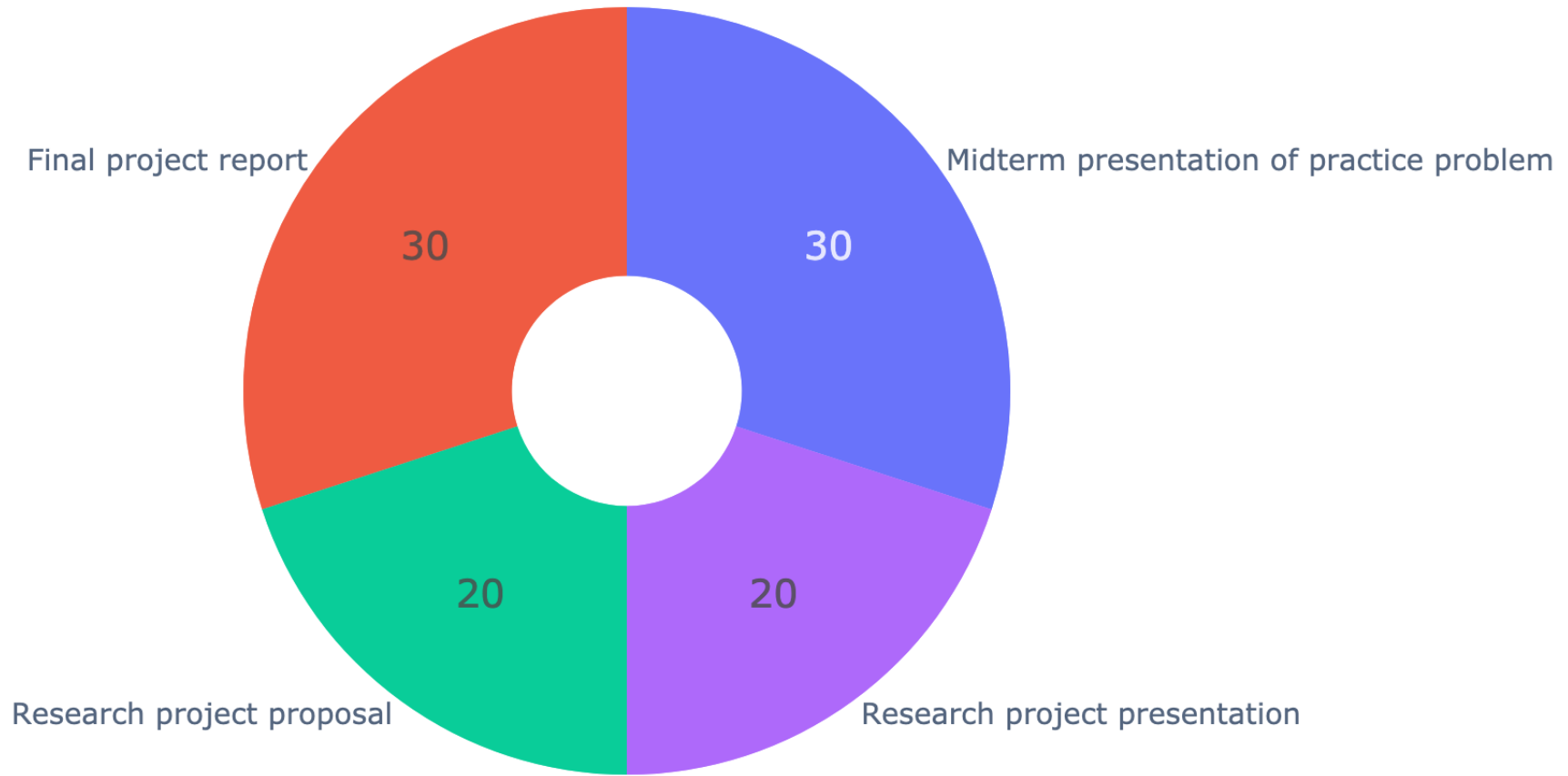
Summary of grading scheme:

- Midterm presentation of the practice problem: 30
- Research project proposal: 20
- Research project presentation: 20
- Final project report: 30

Important note

To receive full mark for the final report, students need to prove that the work presented may lead to a peer-reviewed publication. In other words, the ultimate goal is to obtain a research insight or develop a new research methodology by combining the tools learned in this course with the students' own research.

Final grade breakdown



List of discussion topics:

- General overview of ML problems
 - supervised learning, unsupervised learning, reinforcement learning, optimization
 - regression, classification, clustering, interpolation, generalization, dimensionality reduction
- General review of how ML works
 - feature space, feature-engineering, non-linear transformations, high-dimensional feature spaces, feature mapping
 - regression, underfitting and overfitting, the bias-variance trade-off, regularization, reproducing kernel Hilbert space, kernel trick, neural networks
- Linear regression as a neural network problem
- Bayes's theorem, Likelihood, Marginal Likelihood

- Linear classification (binary)

Logistic regression, Perceptron Learning Algorithm, Linear Discriminant Analysis

- Dimensionality reduction

Principle Component Analysis, Clustering

- Regularization

Bias-variance trade off, Regularization, Ridge Regression

- Reproducing Kernel Hilbert Space

Hilbert Space of Functions, Reproducing Kernels, RKHS, Kernel Ridge Regression

- Nonlinear Classification

Support Vector Machines, Hinge Loss

- Bayesian Machine Learning
 - Bayesian Neural Networks, Gaussian Processes, Gaussian Process Regression, Bayesian Optimization
- Machine Learning for Chemistry
 - Feature Spaces for Chemistry applications, Chemistry Optimization, Molecular Descriptors, Self-driving Labs, Inverse Problems, Interpolation in Chemical Space, Δ -Learning
- How to build better kernels
 - Model Selection Metrics, Marginal Likelihood, Bayesian Information Criterion, Extrapolation with Machine Learning, Bayesian Optimization of Time-consuming Experiments

- Multiclass classification

 - Reduction to Binary Classification, Linear Discriminant Analysis, Naive Bayes, Classification with Neural Networks

- Neural Networks

 - Neuron Activation Functions, Dense Neural Networks, Recurrent Neural Networks

- Convolutional Neural Networks

 - Convolution, Translational Equivariance, Translational Invariance, Pooling, Cats vs dogs

- Data Visualization

Lectures/discussions

What we will aim to do:

- ⇒ Discuss theory behind ML
- ⇒ Discuss useful mathematics/statistics concepts
- ⇒ Discuss examples of ML applications in Chemistry
- ⇒ Discuss examples of python code, if necessary

All python codes used in lectures will be posted on Canvas.

Download them and play with them before each class.

List of 5 applications for the coding excercises:

- Principle component analysis
- Kernel ridge regression
- Support vector classification
- Regression with a Deep Neural Network
- Gaussian process regression (interpolation)

List of optional but recommended applications for coding excercises:

- Bayesian optimization
- Convolutional Neural Networks
- Recurrent Neural Networks
- Gaussian process regression (extrapolation)
- Multi-output regression