

# Machine Learning Course Workbook

## Introduction

### **ML is everywhere!**

*Where (else) do you use ML in your everyday life incl. work?*

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### **ML history: Why now?**

*What accelerated the rise of ML in the last few years?*

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*What is the difference between ANI and AGI?*

## **What is ML?**

*Define ML:*

*What do you need to create an ML-powered product (i.e., value)?*

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*AI and ML Researchers, Statisticians, and Data Scientists all use a certain set of tools.*

*What is the difference between...*

- ML vs. AI?
- ML vs. Deep Learning?
- ML vs. Statistics?
- ML vs. Data Science?

## How do machines “learn”?

*Describe the different learning strategies:*

- Unsupervised Learning:
- Supervised Learning:
- Reinforcement Learning:

## When should you use ML?

*When should you not use ML?*

*For which kinds of problems does ML have a high chance of success and when is the outcome uncertain?*

*What distinguishes an ML project from a data science project (in terms of deliverables)?*

*In what ways can you create value with ML?*

## Solving problems with ML: Workflow

*What are the 3 main steps to create value with ML?*

- 1.
- 2.
- 3.

*What are the two main deployment possibilities for an ML model and when should you use which?*

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*Which tasks take up most of a Data Scientist’s time?*

# ML with Python

*What are the standard abbreviations used when importing the numpy and pandas libraries?*

```
import numpy as ...
import pandas as ...
```

## Data & Preprocessing

*What are “features” and what are “labels”?*

- Features:
- Labels:

*What does structured and unstructured data look like? Which of them is homogeneous and which (usually) heterogeneous?*

- Structured Data:
- Unstructured Data:

*What is the difference between feature extraction and feature engineering?*

- Feature Extraction:
- Feature Engineering:

*A feature matrix  $X$  has the shape  $(n \times d)$ . What do  $n$  and  $d$  stand for?*

- $n$ : number of ...
- $d$ :

## What constitutes 1 data point?

*You are given a dataset with time series data, consisting of measurements from  $d$  sensors for  $n$  time points. What would your feature matrix look like, if your task is...*

- ... to make a prediction for each time point?
- ... to categorize the different sensors?
- ... to predict the quality of each of the 100 products that were produced during this time span?

## Feature Extraction

*What is one way to transform categorical features into a meaningful numerical representation?*

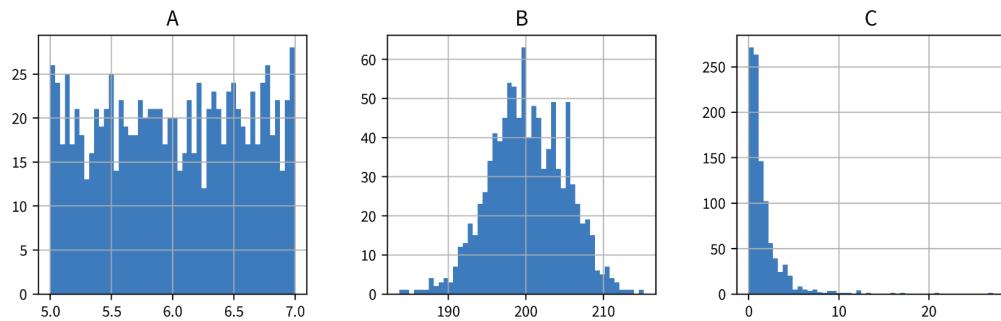
*What are the steps to transform a corpus (i.e., dataset with text documents) into a TF-IDF feature matrix?*

*What are the disadvantages of TF-IDF feature vectors?*

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## Feature Engineering & Transformations

*These are the histograms of 3 different variables A, B, and C:*



*How would you characterize their distributions (Gaussian, exponential, uniform) and which kind of transformation (StandardScaler, MinMaxScaler, PowerTransformer) would be best suited for which of the variables?*

- A:
- B:
- C:

## Computing Similarities

*What preprocessing steps can be helpful to compute a more meaningful similarity or distance between your data points' feature vectors (especially for heterogeneous data)?*

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## Garbage in, garbage out!

*Think about some of the datasets you've encountered in the past: In what ways were they messy?*

*Which concrete next steps should your organization take to improve their data quality?*

## ML Solutions: Overview

*What does the output of the different algorithm categories look like for one data point?*

- Dimensionality Reduction:
- Anomaly Detection:
- Clustering:
- Classification:
- Regression:

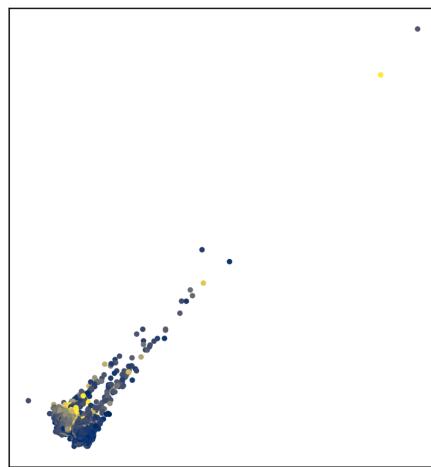
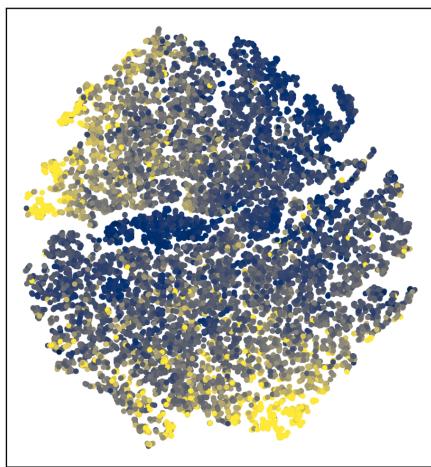
*What are the benefits of breaking down a complex input-output problem into simpler subproblems?*

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## Unsupervised Learning

### Dimensionality Reduction

*Guess: Which plot was generated with PCA and which with t-SNE?*



*How does PCA work?*

*Is PCA using the original input features for the computation or does it first compute a similarity matrix for the data points? What about Kernel PCA?*

*How does t-SNE work?*

*Is t-SNE using the original input features for the computation or does it first compute a similarity matrix for the data points?*

*When would you use PCA and when would you use t-SNE?*

*In the notebook, what did you observe about the PCA eigenvalue spectrum for the data with and without outliers? How do you interpret this?*

## **Outlier/Anomaly Detection**

*What factors should you consider when choosing an outlier detection method?*

*How does the  $\gamma$ -index work?*

*How could you set the parameter  $k$  of the  $\gamma$ -index to detect a cluster of outliers?*

## **Clustering**

*How does the k-means algorithm work?*

*True or False: One disadvantage of k-means is that it assumes spherical clusters?*

*K-means: What would happen if you set  $k$  to a very large value, e.g., the number of data points?*

*How does the DBSCAN algorithm work?*

*What are the advantages of DBSCAN?*

# Supervised Learning Basics

## Different types of models

*What is the difference between a regression and a classification problem?*

*How can you tell if a classification or regression dataset is linear or non-linear?*

*When should you use a features-based and when a similarity-based model and what are their respective drawbacks?*

## Model Evaluation

*Name three regression evaluation metrics:*

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*Name two classification evaluation metrics:*

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*What is the stupid baseline you should always compare your regression models against?*

*What is the stupid baseline you should always compare your classification models against?*

*When should you absolutely evaluate your models with the balanced accuracy metric?*

*How does a cross-validation work? What are the advantages and disadvantages compared to using a fixed validation set?*

# Supervised Learning Models

## Linear Models

*How does a linear model compute the prediction for a new data point?*

*What happens when you use a regularized model and set the regularization parameter to a high value (e.g., alpha for a linear ridge regression model)?*

## Decision Trees

*How does a decision tree compute the prediction for a new data point?*

*For a decision tree with max\_depth=2, how many different features can be used at most for the prediction?*

## Ensemble Methods

*How does a random forest compute the prediction for a new data point?*

## k-Nearest Neighbors (kNN)

*How does a kNN model compute the prediction for a new data point?*

*Why is it better to use an odd number of nearest neighbors for kNN for a binary classification problem?*

## Kernel Methods

*How does a SVM compute the prediction for a new data point?*

*Why is it more efficient to compute the prediction for a new data point using a support vector regression (SVR) model compared to a kernel ridge regression model?*

## Deep Learning & more

### Information Retrieval (Similarity Search)

*What is the most important (and difficult) step when trying to solve an information retrieval task?*

### Deep Learning (Neural Networks)

*How does a feed forward neural network (FFNN) compute the prediction for a new data point?*

*How could a multi-layer FFNN be simplified, if it did not contain any non-linear activation functions between its layers?*

*In what way could you manipulate the parameters (i.e., weight matrices) of an existing FFNN without changing its predictions?*

*What type of neural network architecture would be a natural choice for sequential data like text or time series data?*

*What type of neural network architecture would be a natural choice for image data?*

*How does self-supervised learning work (e.g., using text data)?*

*How does transfer learning work and when can it help?*

### Time Series Forecasting

*What kind of input features can you use in a time series forecasting problem?*

*What conditions need to be fulfilled so it makes sense to use a stateless time series forecasting model?*

## **Recommender Systems**

*What kind of problems (in terms of inputs and outputs) can you solve with recommender systems?*

*What is the “cold start problem” and how could you circumvent it?*

## Avoiding Common Pitfalls

*What are some things you can do to make sure the learned model is not completely wrong?*

*What is the difference between data and concept drift?*

### **Interpolation: Does the model generalize?**

*How can you determine if a model over- or underfits the data?*

*What can you do to improve the performance in case of underfitting?*

*What can you do to improve the performance in case of overfitting?*

*Why does the performance on the training set get worse as the size of the training set increases?*

*Why should you not use a univariate feature selection approach? What are better alternatives?*

*Why can the performance get worse if you (aggressively) reduce the dimensionality of the data with PCA?*

## **Extrapolation: Correlation vs. Causation**

*Why do ML models often fail to extrapolate, i.e., do not make reliable predictions for data points outside the training domain?*

*What are “Adversarial Attacks”?*

*What can you do to try to catch and prevent systematic bias?*

## **Explainability & Interpretable ML**

*What is the difference between local and global explainability?*

*Name two intrinsically interpretable models:*

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*How can you explain an individual prediction of a linear model?*

*How is the permutation feature importance computed?*

*How is a partial dependence plot generated?*

*How can an intrinsically interpretable surrogate model be used to explain an individual prediction of a more complex model?*

*How can you generate optimal inputs and counterfactual examples for a neural network (e.g., for adversarial attacks)?*

## Reinforcement Learning

*For which kinds of tasks does it make sense to use reinforcement learning and when does a normal optimization suffice?*

*How does the Epsilon-Greedy Policy manage the trade-off between exploration and exploitation?*

*How does Q-learning for tabular RL work?*

*How can Q-learning be extended to work with an infinite number of states?*

*Which factors can complicate the use of reinforcement learning?*

## Conclusion

### **AI Transformation of a Company**

*What can you do if you have “big data”?*

*According to Andrew Ng, what are the 5 steps for a successful AI transformation of a company?*

- 1.
- 2.
- 3.
- 4.
- 5.