

# Machine Learning Course Workbook

– Part 1 –

## 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 and what their requirements (in terms of data) are:*

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

*What is the drawback of unsupervised learning methods?*

*What is the goal of a supervised learning algorithm and how is it accomplished?*

## **When should you use ML?**

*In what ways can ML create value?*

*When should you not use ML?*

*Which kind of ML problems have a high chance of success and when is the outcome uncertain?*

## **Solving problems with ML: Workflow**

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

- 1.
- 2.
- 3.

*What should you check before starting an ML project?*

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*What are the two deployment options 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 one 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 produced during this time span?

### Feature Extraction

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

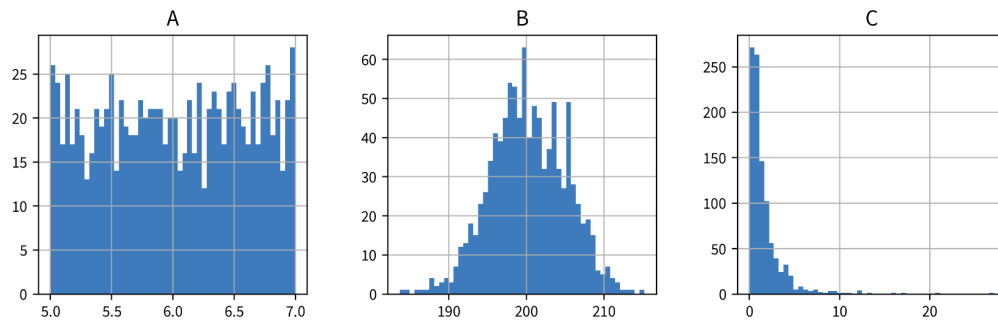
*What are the steps to transform a corpus (i.e., dataset with text documents) into a TF-IDF feature matrix?*

What are the disadvantages of TF-IDF feature vectors?

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## Feature Engineering & Transformations

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



How would you characterize their distributions (Gaussian, exponential, uniform) and which kind of transformation (StandardScaler, MinMaxScaler, PowerTransformer) might 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 the 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:
- Regression:
- Classification:
- Recommender Systems/Information Retrieval:

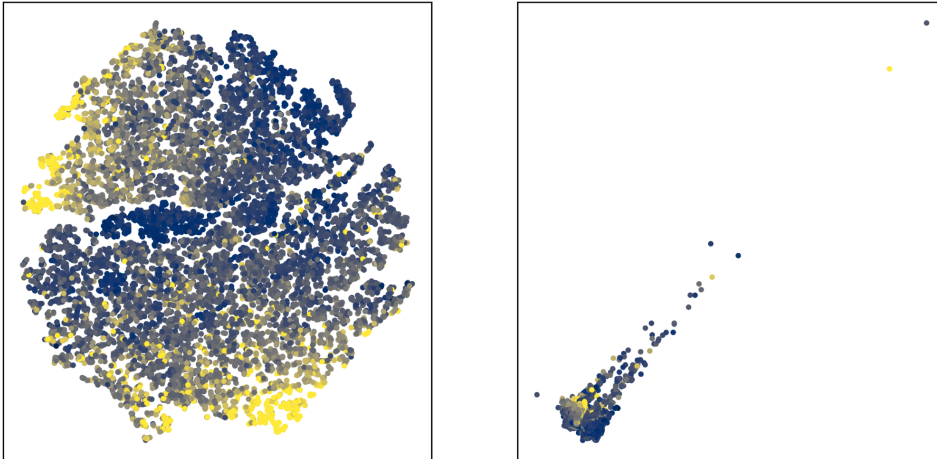
*What are the benefits of breaking down a complex input-output problem into simpler subproblems?*

– Part 2 –

## 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 nonlinear (e.g., with one input  $x$ )?*

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

## **Model Evaluation**

*With which stupid baseline should you compare regression and classification models respectively?*

*Name three regression evaluation metrics:*

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*Name two classification evaluation metrics:*

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*When is it a really bad idea to evaluate a classification model with the 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 in `sklearn`)?*

### 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

*What are the different strategies for creating an ensemble model?*

*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 kernel ridge regression (KRR) model 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 machine (SVM) model compared to KRR?*

## 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 nonlinear 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 could you use to forecast how many pretzels a bakery will sell tomorrow?*

*What is the difference between stateless and stateful models and which 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 can you circumvent it?*

## Avoiding Common Pitfalls

*What are some common pitfalls that you should avoid?*

*In what ways can domain knowledge help you arrive at a better model?*

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

*What could be reasons for data or concept drift in your domain / next project?*

### **Model does not generalize**

*How can you tell whether a model over- or underfits the data?*

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

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

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

### **Model abuses spurious correlations**

*Why can a model still be wrong, even though it generates correct predictions for data points from the testset?*

*What are “Adversarial Attacks”?*

### **Model discriminates**

*In what ways can a biased model negatively affect users?*

*Why can it happen that a model discriminates?*

*How can you check whether a model discriminates?*

*What can you do to get a fair model?*

### **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?*

## 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?*

*What is a Q-value and 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

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

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

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

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

*Where do you think your organization stands in this AI transformation process?*