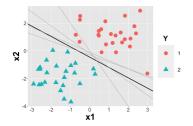
Introduction to Machine Learning

ML-Basics: In a Nutshell



Learning goals

- Understand fundamental goal of supervised machine learning
- Know concepts of task, model, parameter, learner, loss function, and empirical risk minimization



WHAT IS ML?

"A computer program is said to learn from experience E with respect to some task T and some performance measure P, if its performance on T, as measured by P, improves with experience E."

Tom Mitchell, Carnegie Mellon University, 1998

 \Rightarrow 99 % of this lecture is about **supervised learning**:



Training



Prediction

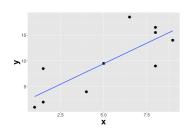


TASKS

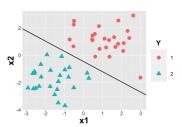
- Supervised tasks are labeled data situations where the goal is to learn the functional relationship between inputs (features) and output (target)
- We distinguish between regression and classification tasks, depending on whether the target is numerical or categorical

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Regression: Target is **numerical**, e.g., predict days a patient has to stay in hospital

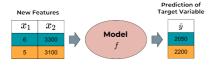


Classification: Target is **categorical**, e.g., predict one of two risk categories for a life insurance customer



MODELS AND PARAMETERS

• A model is a function that maps features to predicted targets





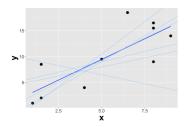
- For finding the model that describes the relation between features and target best, one needs to restrict the set of all possible functions
- This restricted set of functions is called **hypothesis space**. E.g., one could consider only simple linear functions as hypothesis space
- Functions are fully determined by parameters. E.g., in the case of linear functions, $y = \theta_0 + \theta_1 x$, the parameters θ_0 (intercept) and θ_1 (slope) determine the relationship between y and x
- Finding the optimal model means finding the optimal set of parameters

LEARNER

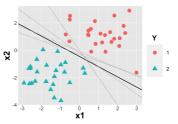
- Learns automatically the relation between features and target given a set of training data
- Learner picks the best element of the hypothesis space, i.e., the function that fits the training data best



Regression:



Classification:



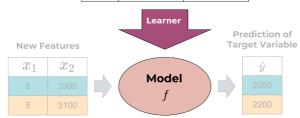
LEARNER / 2

• Learner uses labeled training data to learn a model f. This model is applied to new data for predicting the target variable

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Train Set

y	x_1	x_2
2200	4	4300
1800	12	2700
1920	15	3100



LOSS AND RISK MINIMIZATION

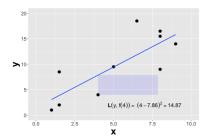
• Loss: Measured pointwise for each observation, e.g., *L*₂-loss

$$L(y, f(\mathbf{x})) = (y - f(\mathbf{x}))^2$$

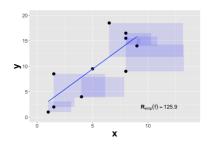
• Risk: Measured for entire model. Sums up pointwise losses.

$$\mathcal{R}_{emp}(f) = \sum_{i=1}^{n} L\left(y^{(i)}, f\left(\mathbf{x}^{(i)}\right)\right)$$

Squared **loss** of one **observation**.



Empirical risk of entire model





EMPIRICAL RISK MINIMIZATION

- \bullet The risk surface visualizes the empirical risk for all possible parameter values of the parameter vector θ
- Minimizing the empirical risk is usually done by numerical optimization

$$\hat{\theta} = \arg\min\nolimits_{\theta \in \Theta} \mathcal{R}_{\text{emp}}(\theta).$$

