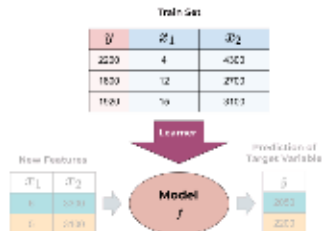


Introduction to Machine Learning

ML-Basics Learner



Learning goals

- Know formal definition of learner
- Understand that a learner receives training data and outputs the best model from \mathcal{H}

SUPERVISED LEARNING EXAMPLE

Imagine we want to investigate how working conditions affect productivity of employees.

- It is a **regression** task since the target *productivity* is continuous.
- We collect data about worked minutes per week (*productivity*), how many people work in the same office as the employee in question, and the employee's salary.

	Features x		Target y
	People in Office (Feature 1) x_1	Salary (Feature 2) x_2	Worked Minutes Week (Target Variable)
$n = 3$	4	4300 €	2220
	12	2700 €	1800
	5	3100 €	1920

$p = 2$

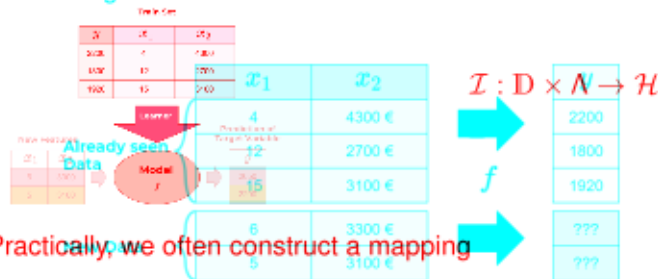
$x_1^{(2)}$ $x_2^{(1)}$ $y^{(3)}$



LEARNER DEFINITION LEARNING

Goal: Automatically identify the fundamental functional relation in the data that maps an object's features to the target.

- The learner is our means of picking the best element from the hypothesis space \mathcal{H} for given training data.
- Formally, it maps training data $D \in \mathbb{D}$ (plus a vector of hyperparameter control settings $\lambda \in \Lambda$) to a model:
- Ultimately, we use our model to compute predictions for new data whose target values are unknown.



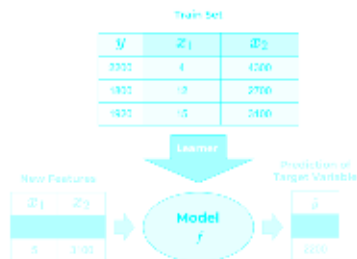
- Practically, we often construct a mapping

$$\mathcal{I} : \mathbb{D} \times \Lambda \rightarrow \Theta$$

LEARNER DEFINITION

- The algorithm for finding our f is called **learner**. It is also called **learning algorithm** or **inducer**.
- We prescribe a certain hypothesis space, the learner is our means of picking the best element from that space for our data set.
- Formally, it maps training data $\mathcal{D} \in \mathbb{D}$ (plus a vector of **hyperparameter** control settings $\lambda \in \Lambda$) to a model:

$$\mathcal{I} : \mathbb{D} \times \Lambda \rightarrow \mathcal{H}$$



LEARNER DEFINITION

As pseudo-code template it would work like this:

- Learner has a defined model space of parametrized functions \mathcal{H} .
- User passes data set $\mathcal{D}_{\text{train}}$ and control settings λ .
- Learner sets parameters so that model matches data best.
- Optimal parameters $\hat{\theta}$ or function \hat{f} is returned for later usage.

