# Introduction to Machine Learning

**Learning Tasks** 

Introduction: Supervised Learning &

#### **IDEA OF SUPERVISED LEARNING**

- **Goal:** Identify the fundamental functional relation in the data that maps an object's features to the target.
- Ideally, we would have full knowledge about the data-generating process and thus be able to specify this mapping function precisely.
- However, since this is basically impossible, we must try to learn the mapping function: for objects exhibiting certain patterns or properties, certain outcomes are much more likely.
  - $\rightarrow$  We call such an assumed mapping a **model** f.

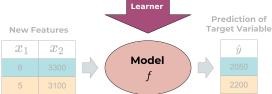
#### **IDEA OF SUPERVISED LEARNING**

- Supervised learning means we make use of labeled data, i.e., observations for which we already know the target outcome.
- We try to construct f automatically from an example set of such labeled objects.
  - $\rightarrow$  The algorithm for finding f is called **learner**.
- Using the thus learned model, we can make predictions based on the features of our data.
- Knowing the "truth" allows us to test how well we have grasped the nature of the underlying mapping: we just need to compare our predictions to the actually observed values.

# **IDEA OF SUPERVISED LEARNING**

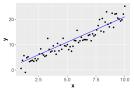
 Ultimately, we will use our model to compute predictions for new data whose target values are unknown.



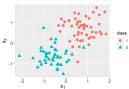


#### TASKS IN SUPERVISED LEARNING

- In general, supervised learning comes in two flavors we call **tasks**:
  - **Regression**: Given features  $\mathbf{x}$ , predict corresponding output from  $\mathcal{Y} \in \mathbb{R}^m$ .



• Classification: Assign an observation with features  $\mathbf{x}$  to one class of a finite set of classes  $\mathcal{Y} = \{C_1, ..., C_g\}, g \geq 2$  (details later).



#### **REGRESSION TASKS: EXAMPLE**

Imagine you want to investigate how salary and workplace conditions (features) affect productivity of employees (target) – a standard regression task. Therefore, you collect data about their worked minutes per week (productivity), how many people work in the same office as the employees in question, and the employees' salary.

	Feat	ures $x$	Target $y$		
	People in Office (Feature 1) $x_1$	Salary (Feature 2) $x_2$	Worked Minute (Target Vari		
	4	4300 €	2220		
$p=2$ $\Big\langle$	y 12	2700 €	1800		
$\downarrow$	5	3100 €	1920	×	
$oxed{x_1^{(2)}}$	n =	= 3	$x_2^{(1)}$	$y^{(3)}$	

# **REGRESSION TASKS: EXAMPLE**

- For our observed data we know which outcome is produced.
- For new employees can only observe the features but not the target.

	$x_1$	$x_2$	Functional Relationship	y
Already seen	4	4300 €	ř.	2200
	12	2700 €	1	1800
Julia	15	3100 €	f	1920
New Data ≺	6	3300 €	rico de	???
New Data	5	3100 €		???

# **MORE REGRESSION TASKS**

#### • House Prices

Aim: Predict the price for a house in a certain area

	Target $y$			
square footage of the house	number of bedrooms	swimming pool (yes/no)		house price in US\$
1,180	3	0		221,900
2,570	3	1		538,000
770	2	0		180,000
1,960	4	1		604,000



# MORE REGRESSION TASKS

# 2 Length-of-stay in a hospital

**Aim**: Predict the number of days a single patient has to stay in hospital at the time of admission

Features $x$					Target $y$
diagnosis category	admission type	gender	age		Length-of-stay in the hospital in days
heart disease	elective	male	75		4.6
injury	emergency	male	22		2.6
psychosis	newborn	female	0		8
pneumonia	urgent	female	67		5.5



# **CLASSIFICATION TASKS: EXAMPLE**

#### Risk category in life insurance

**Aim**: Predict one of five risk categories (**classification**) for a life insurance customer to determine the insurance premium

	Target $y$			
job type	age	smoker		risk group
carpenter	34	1		3
stuntman	25	0		5
student	23	0		1
white-collar worker	39	0		2

