

# **Introduction to Machine Learning**

## **Introduction: Models & Parameters**

[compstat-lmu.github.io/lecture\\_i2ml](https://compstat-lmu.github.io/lecture_i2ml)

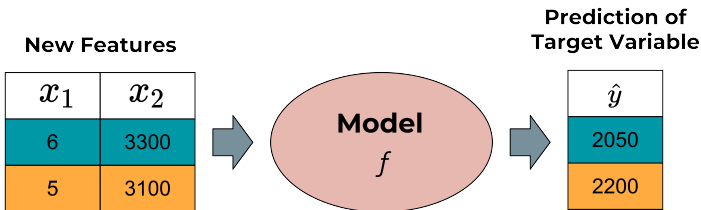
# WHAT IS A MODEL?

- A **model** (or **hypothesis**)

$$f : \mathcal{X} \rightarrow \mathbb{R}^g$$

is a function that maps feature vectors to predicted target values.

- In conventional regression we will have  $g = 1$ , for classification see later.



# WHAT IS A MODEL?

- We have already seen that machine learning typically requires constraining  $f$  to a certain class of functions.
- Otherwise, the task of finding a “good” model is basically impossible to solve.
- The set of functions defining a model class is called a **hypothesis space**  $\mathcal{H}$ .
- For example, the set of all constant functions

$$\mathcal{H} = \{f : f(\mathbf{x}) = c, c \in \mathbb{R}\}$$

forms a specific hypothesis space.

# PARAMETERS OF A MODEL

- Within one hypothesis space, models are typically all “alike” in a sense: they all share a certain structure.  
→ E.g., all Gaussian density functions exhibit a bell-like shape.
- The only (!) aspects in which they differ are described by **parameters**.  
→ Gaussians are solely determined by mean and variance.
- We typically subsume all parameters in a **parameter vector**  $\theta = (\theta_1, \theta_2, \dots)$  from a **parameter space**  $\Theta$ .
- $\theta$  is learned during training: finding a “good” model boils down to finding a suitable combination of parameters.
- We will in the next chapter how the “goodness” of a model can be determined.