

# 08\_Linear\_Model\_for\_Regression

## Noting Paradigm

- $x$  - Plain text: Scalar.
- $\mathbf{x}$  - Bold-Face lowercase: Vector of scalars.
  - e.g.,  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_D \end{bmatrix}$ , where  $\mathbf{x} \in \mathbb{R}^D$
- $\mathbf{X}$  - Bold-Face uppercase: Set of vectors.
  - e.g.,  $\mathbf{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N\}$ , where  $\mathbf{X} \subset \mathbb{R}^D$  and  $|\mathbf{X}| = N$ .

## 8.0 Regression

### 8.0.0 Why Regression?

#### Problem Setup

##### Given

- A set of inputs:
  - $\mathbf{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N\}$ , where, for each input:
    - $\mathbf{x}_i = \begin{bmatrix} x_{i1} \\ x_{i2} \\ \dots \\ x_{iD} \end{bmatrix} \in \mathbb{R}^D$
- A set of corresponding outputs:
  - $\mathbf{y} = \{y_1, y_2, \dots, y_N\}$ , where, for each output:
    - $y_i \in \mathbb{R}$
- A labelling relation:
  - $D = \{(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_N, y_N)\}$

##### Goal

- To learn a mapping
  - $f(x) : \mathbf{X} \rightarrow \mathbf{y}$ , where
    - Input: An arbitrary vector  $\mathbf{x} \notin \mathbf{X}$ ,
    - Output: A scalar  $y \notin \mathbf{y}$ .
- Such that we could make prediction

- about  $y_*$
- on an unseen input  $\mathbf{x}_*$  is encountered.

## Different Kinds

- Parametric Regression 参数性回归
  - Assume a functional form for  $f(x)$ .
- Nonparametric Regression 非参数性回归
  - Does not assume a functional form for  $f(x)$ .


## To sum up

- From given relations, learn a function that
  - Takes a vector as an input
  - Output a real number that
    - "Fits" the given pattern

## 8.0.1 Definition

 What is Regression?

- Regression aims at modelling the dependence of:
  - a response  $Y$ ,
  - on a covariate  $X$ .
- That is, to predict the value of one or more continuous target variables  $y$  given the value of input vector  $x$ .

 The regression model is described by

$$y = f(\mathbf{x}) + \epsilon$$

- where the dependence of a response  $y$  on a covariate  $\mathbf{x}$  is captured via:
  - $p(y|\mathbf{x})$ ,
  - i.e., a conditional probability distribution.

## 8.1 Linear Regression