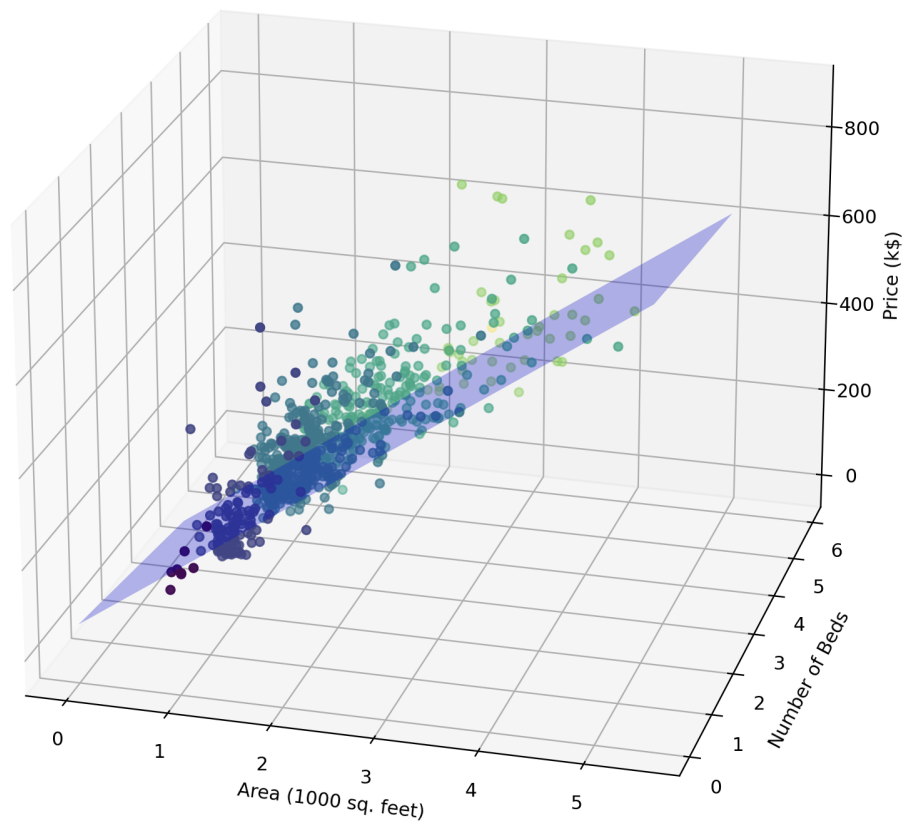
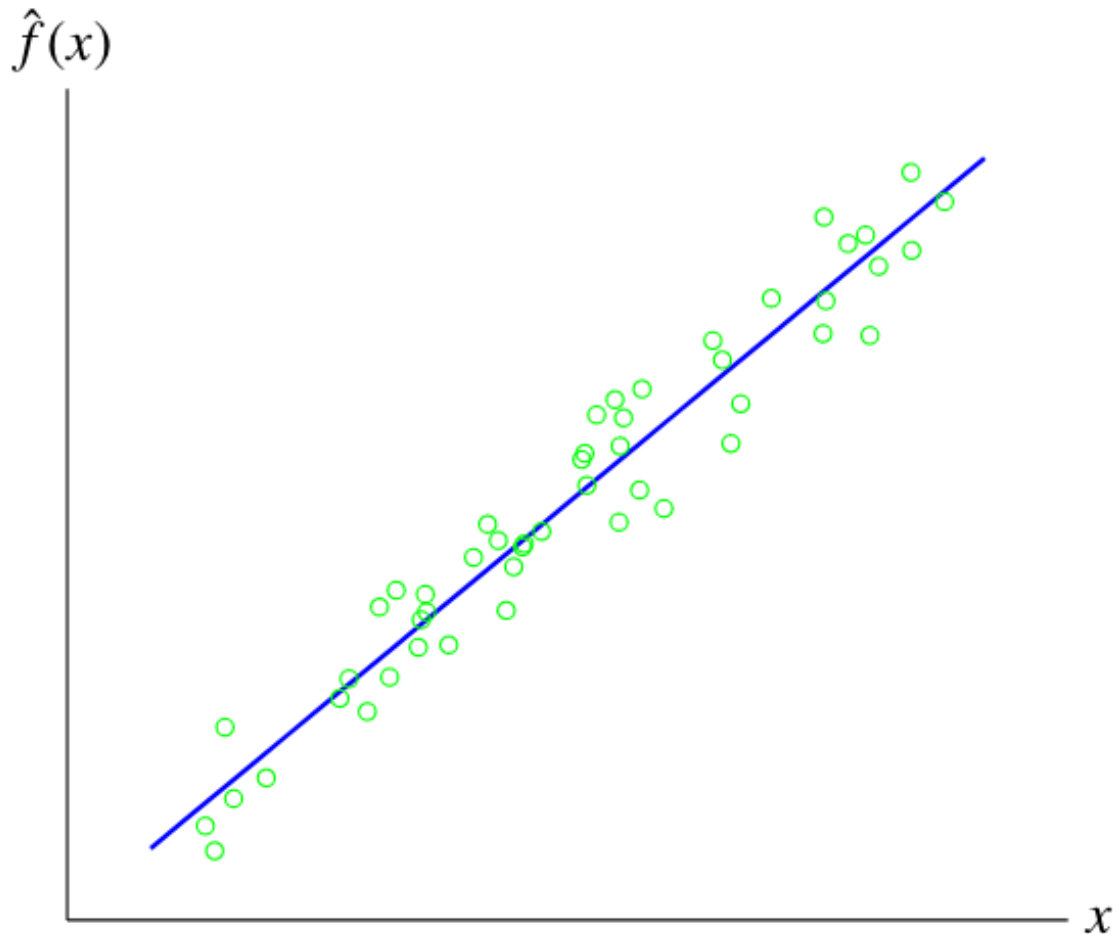


# 12 Least Squares Data Fitting





Unit 1: Vectors, Book ILA Ch. 1-5

Unit 2: Matrices, Book ILA Ch. 6-11 + Book IMC Ch. 2

Unit 3: Least Squares, Book ILA Ch. 12-14

- 11 Least Squares
- **12 Least Squares Data Fitting: Regression and Classification**

## Outline: 12 Least Squares Data Fitting

- [Least Square Model Fitting](#)
- [Application to Regression and Classification](#)

True relationship:  $f$

**Definition:** When we believe that a scalar  $y$  and an  $n$ -vector  $x$  are related by model:

$$y \approx f(x),$$

we use the following vocabulary:

- $x$  is called the independent variable
- $y$  is called the outcome or response variable
- $f : \mathbb{R}^n \rightarrow \mathbb{R}$  represents the "true" relationship between  $x$  and  $y$ .

Generally, we do not know  $f$ , we just assume it exists. Our goal is to learn  $f$ , or a reasonable approximation of it, using data.

## Model: $\hat{f}$

**Definition:** Choosing a set of basis functions:  $f_j : \mathbb{R}^n \rightarrow \mathbb{R}$ , for  $j = 1 \dots p$ , we model a guess or approximation of  $f$  as:

$$\hat{f}(x) = \theta_1 f_1(x) + \dots + \theta_p f_p(x),$$

where:

- $\theta_j$  are parameters that we learn from data:  $x^{(1)}, \dots, x^{(N)}, \dots, y^{(1)}, \dots, y^{(N)}$ ,
- $\hat{y}^{(i)} = \hat{f}(x^{(i)})$  is (the model's) prediction of  $y^{(i)}$ , for  $i = 1, \dots, N$ .

**Remark:** If our model is good, then  $\hat{y}^{(i)} \approx y^{(i)}$ .

## Least Square Data Fitting

**Definition:** We define the prediction error, or residual for each  $i = 1, \dots, N$ :

$$r_i = y^{(i)} - \hat{y}^{(i)}.$$

**Definition:** The Least Square Data Fitting problem is the problem of choosing model's parameters  $\theta_1, \dots, \theta_n$  that minimize the RMS prediction error on the dataset:

$$\left( \frac{r_1^2 + \dots + r_N^2}{N} \right)^{1/2}.$$

**Proposition:** Define the  $N \times p$  matrix  $A$  with elements  $A_{ij} = f_j(x^{(i)})$ , such that  $\hat{y} = A\theta$ , where  $y = (y^{(1)}, \dots, y^{(N)})$  is vector of outcomes. The least square data fitting problem amounts to choose  $\theta$  that minimizes:

$$\|A\theta - y\|^2,$$

which shows that it can be written as a Least Square Problem. Assuming that the columns of  $A$  are independent, the solution is:

$$\hat{\theta} = (A^T A)^{-1} A^T y.$$

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# Classification and Regression

**Artificial Intelligence (AI):** Techniques that enable machines to mimic human intelligence.

**Machine Learning (ML):** Techniques that enable machines to learn from data.

**Supervised Learning:** Task of learning a function that maps an input to an output based on example input-output pairs.

Examples:

- **Regression:** maps an input to a quantitative value.
- **Classification:** maps an input to a categorical value.

**Unsupervised Learning:** Task of discovering any naturally occurring patterns in a data set.

Examples:

- **Clustering:** discover groups (clusters) within the data: today.
- **Dimension reduction:** later in this class.

## Regression of House Prices

- $x$ : house's area in 1000 sq feet,
- $y$ : house's price in k\$ (we do not consider the number of beds for simplicity).

Consider the model:  $\hat{f}(x) = \theta_1 f_1(x) + \theta_2 f_2(x)$  with  $f_1(x) = 1$  and  $f_2(x) = x$ , i.e.:

$$\hat{f}(x) = \theta_1 + \theta_2 x.$$

**Example:** What are  $A$ ,  $y$ ? Explain how you can find  $\hat{\theta}_1$  and  $\hat{\theta}_2$  with Python. At home: compute  $\hat{\theta}_1$  and  $\hat{\theta}_2$  manually (hard).

## From Regression to (Binary) Classification

- Model  $\hat{f}$  outputs a number.
- Binary classification wants a category: +1 or -1 only.

→ use  $\text{sign}(\hat{f})$  in place of  $\hat{f}$  to classify.

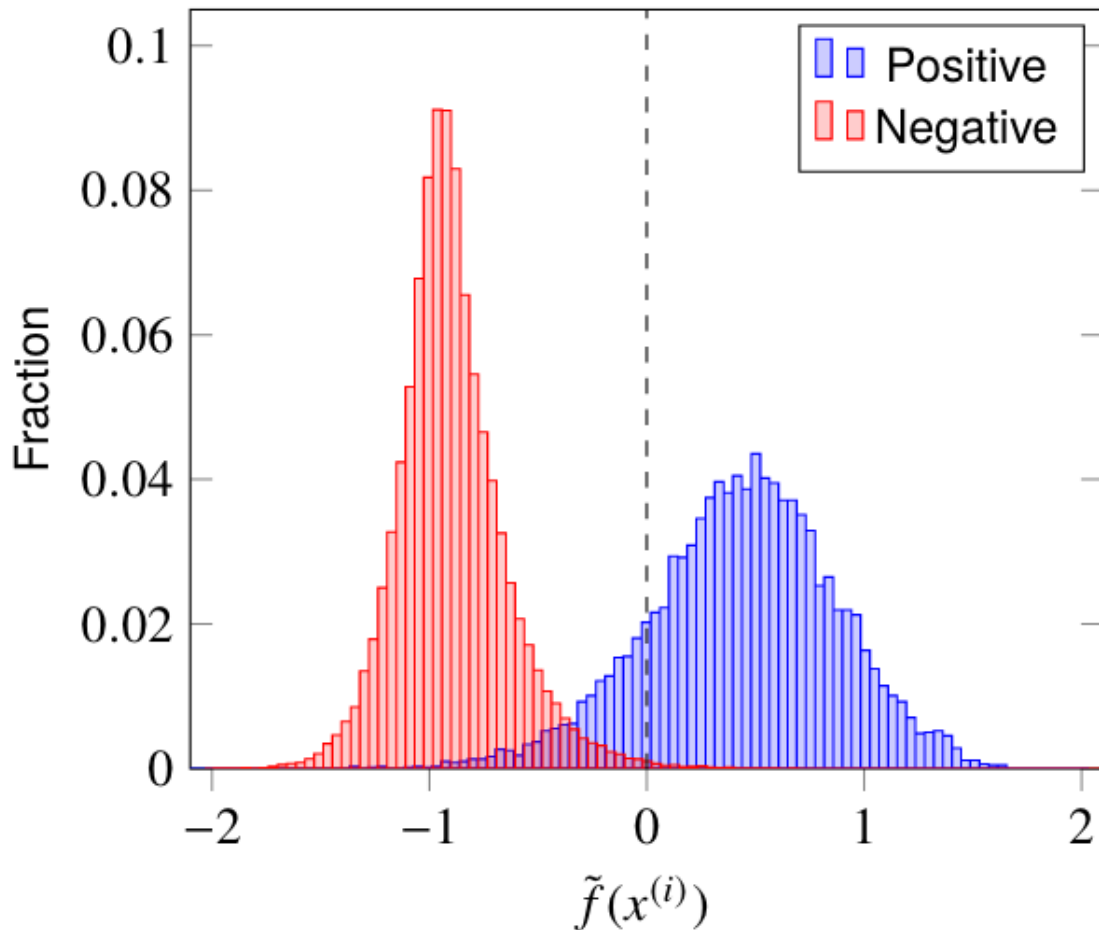
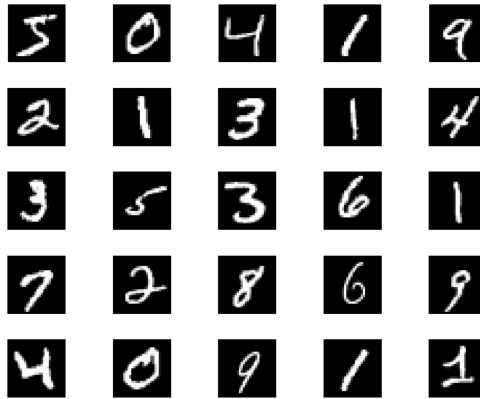
## Classification of MNIST

- $x$  image of size  $28 \times 28$  from the MNIST dataset,
- $y$ : whether the image shows a 0 digit or another digit.

Consider the model:  $\hat{f}(x) = \text{sign}(\theta_0 f_0(x) + \theta_1 f_1(x) + \dots + \theta_{784} f_{784}(x))$  with  $f_0(x) = 1$ , and  $f_p(x) = x_p$  for  $p = 1, \dots, 784$ , i.e.:

$$\hat{f}(x) = \text{sign}(\theta_0 + \theta_1 x_1 + \dots + \theta_{784} x_{784}).$$

**Example:** What are  $x, y, A$ ? Explain how you can find the  $\hat{\theta}$ s with Python.



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Resources: Book ILA Ch. 13-14.