# Introduction to Machine Learning

Lecture 3: Regression

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#### Before we start

Would you be interested in a more advanced course? I can propose

- Machine learning from scratch (how to implement an ML algorithm with no library)
- ► A more advanced version of this course (with more theoretical technical details)
- Large-scale machine learning

# Regression in Machine Learning

This lecture is about regression in Machine learning.

**Reminder**: In regression, the output *y* is **continous**.

#### Example:

- **Price estimation**: y = price (e.g. 50000 BGN for a house)
- ▶ **Predicting the future** (*e.g.* weather forecast): *y* = temperature or amount of rain

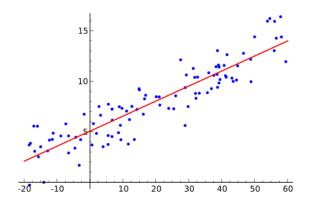
# Regression in Machine Learning: Applications

#### Domains of application:

- ▶ Price estimation/prediction
- Weather forecast
- Production quantity estimation
- Stock option price prediction
- ▶ Fit statistical model to data
- Physics & chemistry
- ... and others

# Linear and polynomial regression

Purpose of regression: **approximate solutions** of **overdetermined systems**.



In this course, we will see

- ► Linear regression
- ► Polynomial regression

# Linear regression

### Linear regression

#### Principal components:

- Old problem (least-squares method usually credited to Carl Friedrich Gauss in 1795)
- Several ways to approximate the data
  - Linear model
  - Polynomial model (remember kernels from SVMs)
  - ▶ Fit a distribution
  - **.** . . .
- Several ways to formulate the problem
  - Least Squares
  - Support Vector regression
  - •
- Several ways to solve the problem

# Linear regression with ordinary least-squares

**Linear** regression: Estimate y as a **linear** function of x:

$$\hat{y} = w^T x$$

• ,	••	' ' '	
50	1	30	
76	2	48	
26	1	12	
102	3	90	
	50 76 26	50 1 76 2 26 1	50 1 30   76 2 48   26 1 12

living area (m<sup>2</sup>) | **# bedrooms** | price (1000's euros)

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#### Variable standardisation

Variables have various magnitudes. Example:

- ► Living area: Up to a few hundreds m<sup>2</sup>
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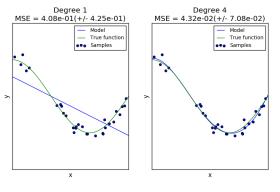
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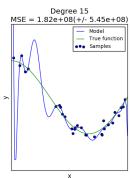
Another option: Scale between 0 and 1

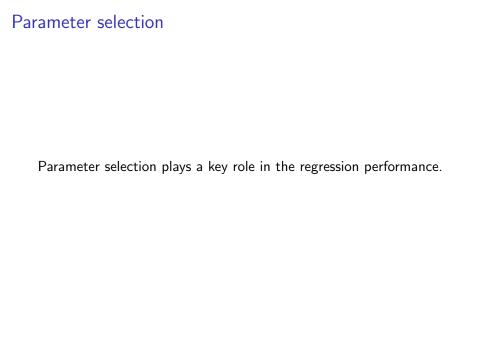
$$x = \frac{x - \min}{\max - \min}$$

# Overfitting and underfitting

$$y = \cos\left(\frac{3\pi}{2}x\right) + \text{noise}$$







# Fitting a distribution

# Thank you! Questions?