#### **Fundamentals of Neural Networks**

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#### Introduction



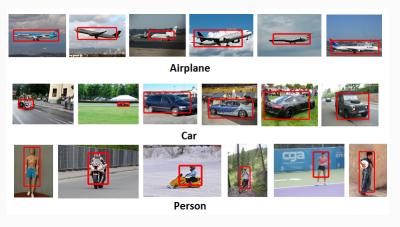
**Figure 1:** A self-driving car. Credit: Marc van der Chijs / CC BY-ND 2.0

#### Introduction



**Figure 2:** A digital assistant. Credit: Kārlis Dambrāns / CC BY 2.0

#### Introduction



**Figure 3:** Object detection in images. Credit: Lu et. al<sup>1</sup>

 $<sup>^{1}</sup>$  '1-HKUST: Object Detection in ILSVRC 2014" ,  $\it CoRR$  , vol. abs/1409.6155, 2014

## **Outline**

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- 1. The Perceptron
- 2. Feedforward Neural Networks

Architecture

Mathematical formulation

3. Training Feedforward Neural Networks

Cost functions

Stochastic Gradient Descent

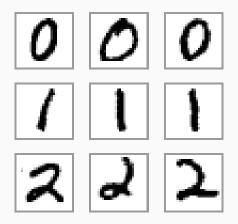
Back-propagation

4. Extensions

# The Perceptron

 Predict whether an input image of a handwritten digit shows a zero or another digit

## **MNIST Data Sample**



**Figure 4:** Examples from the MNIST database. Credit: Josef Steppan / CC BY-SA 4.0

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- $\bullet$  The image is represented as a flattened vector of pixel intensities  $\textbf{x} \in \mathbb{R}^{784}$
- ullet The output should be 1 if the image shows a zero, otherwise it should be -1

### **Model Specification**

The perceptron accepts n input values and computes an output value  $\hat{y}$ :

$$\hat{y} = \operatorname{sign}\left(\sum_{i=1}^{n} w_i x_i\right)$$

$$\equiv \hat{y} = \operatorname{sign}(\mathbf{w}^{\top} \mathbf{x})$$
(1)

**Feedforward Neural Networks** 

**Training Feedforward Neural** 

**Networks** 

## **Extensions**

