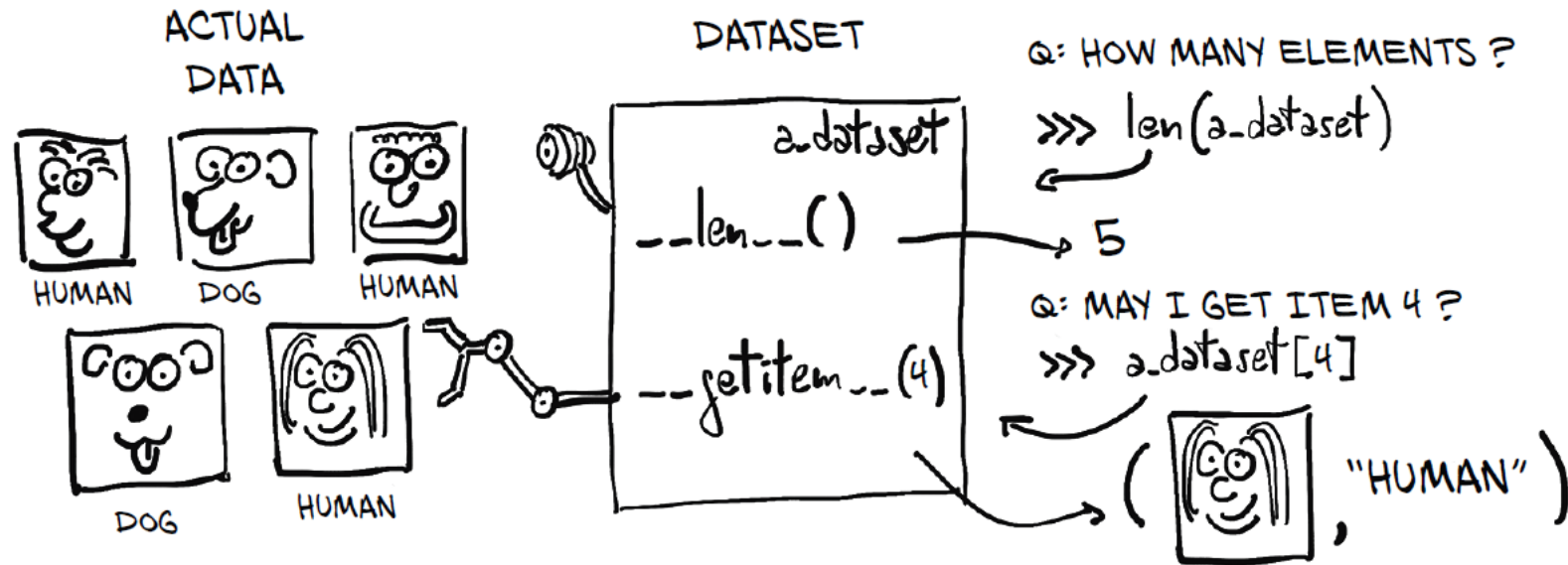


# Vorlesung 7

# CIFAR

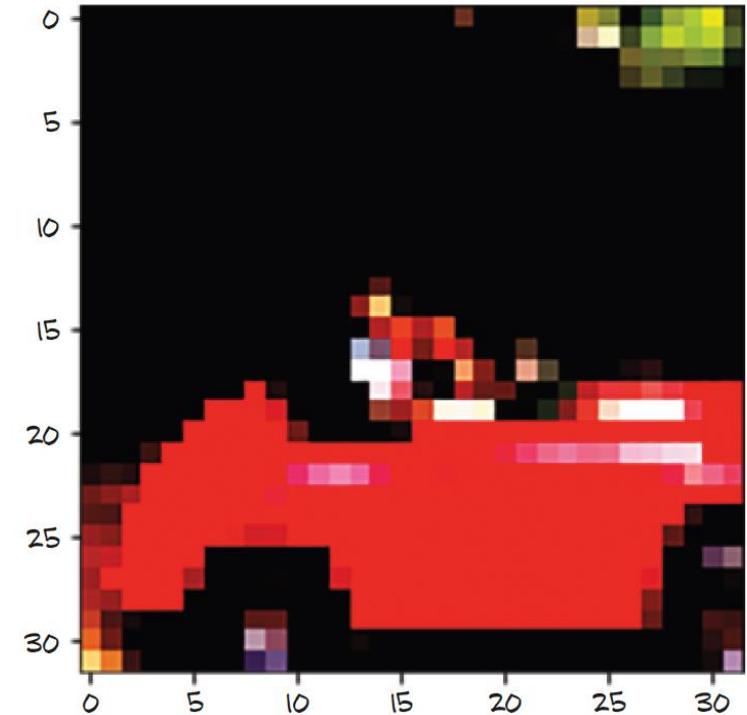
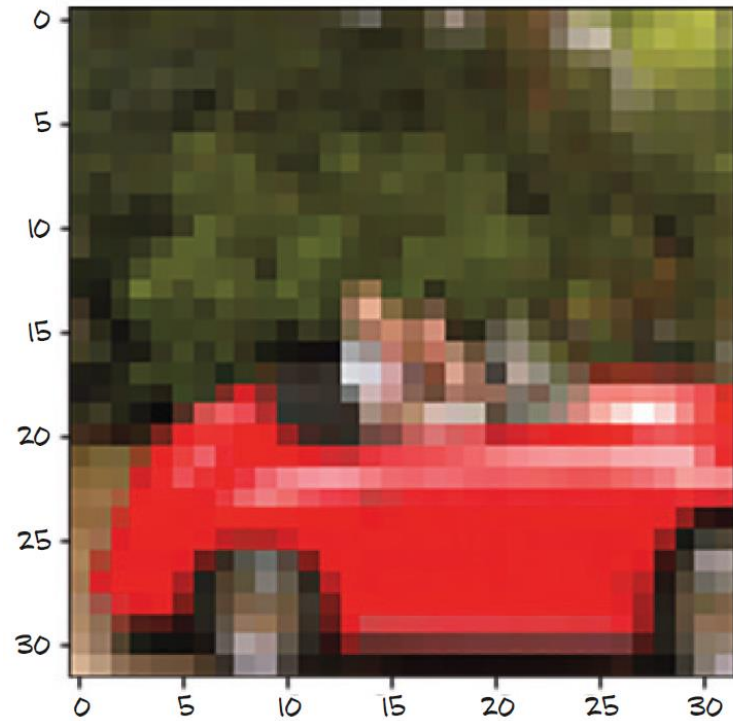


# Datasets

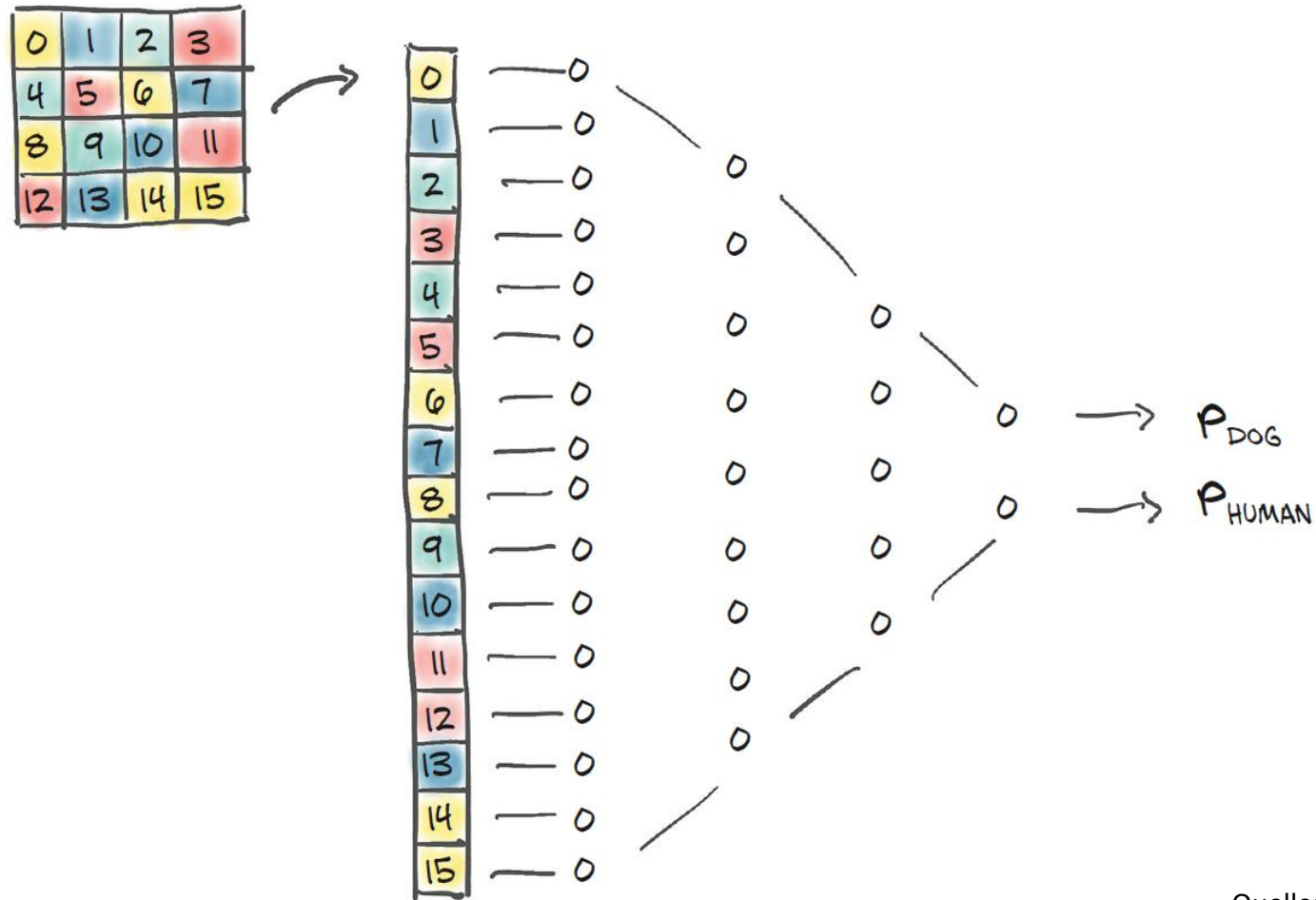


# Normalization

```
cifar10 = datasets.CIFAR10('data', train=True, download=False,  
                             transform=transforms.Compose([  
                                 transforms.ToTensor(),  
                                 transforms.Normalize((0.4915, 0.4823, 0.4468),  
                                                     (0.2470, 0.2435, 0.2616))  
                             ]))
```



# Classification Neural Network



# Softmax

$$0 \leq \frac{e^{x_1}}{e^{x_1} + e^{x_2}} \leq 1$$

EACH ELEMENT  
BETWEEN  
0 AND 1

$$\frac{e^{x_1}}{e^{x_1} + e^{x_2}} + \frac{e^{x_2}}{e^{x_1} + e^{x_2}} = \frac{e^{x_1} + e^{x_2}}{e^{x_1} + e^{x_2}} = 1$$

SUM OF ELEMENTS  
EQUALS 1

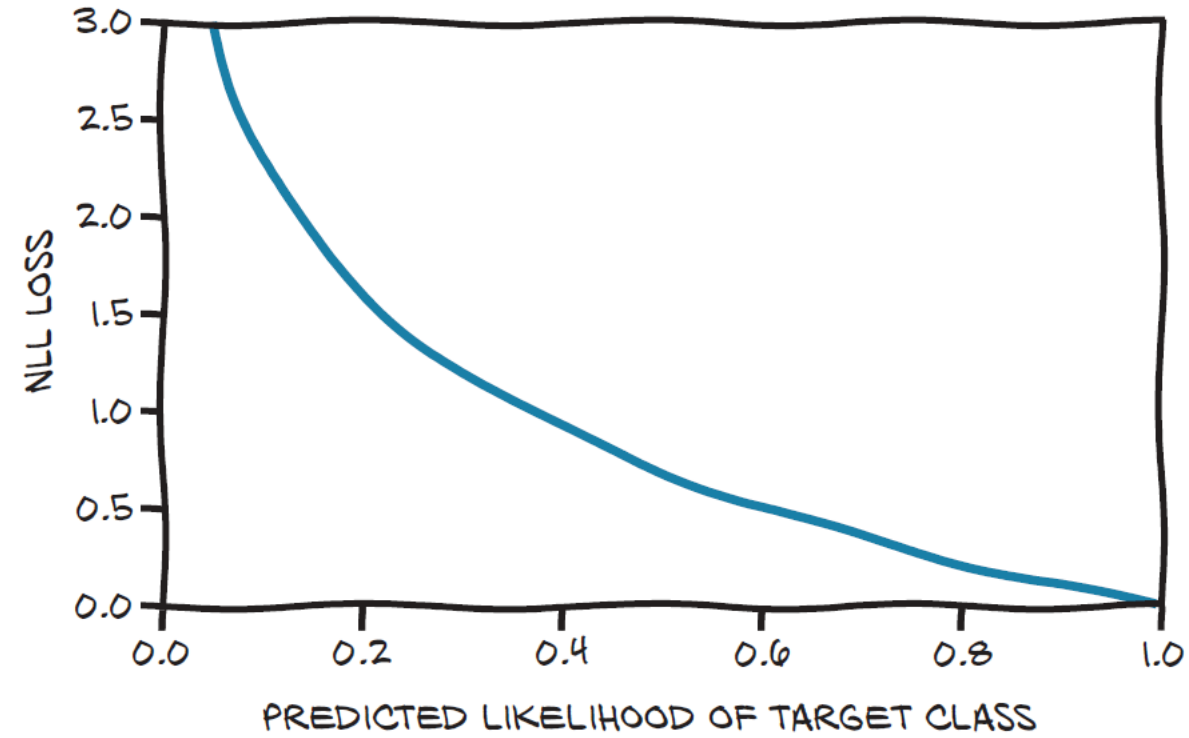
$$\text{softmax}(x_1, x_2) = \left( \frac{e^{x_1}}{e^{x_1} + e^{x_2}}, \frac{e^{x_2}}{e^{x_1} + e^{x_2}} \right)$$

$$\text{softmax}(x_1, x_2, x_3) = \left( \frac{e^{x_1}}{e^{x_1} + e^{x_2} + e^{x_3}}, \frac{e^{x_2}}{e^{x_1} + e^{x_2} + e^{x_3}}, \frac{e^{x_3}}{e^{x_1} + e^{x_2} + e^{x_3}} \right)$$

⋮

$$\text{softmax}(x_1, \dots, x_n) = \left( \frac{e^{x_1}}{e^{x_1} + \dots + e^{x_n}}, \dots, \frac{e^{x_n}}{e^{x_1} + \dots + e^{x_n}} \right)$$

# NLLLoss



# Minibatches

Ⓐ

FOR N EPOCHS:

WITH EVERY SAMPLE IN DATASET:

EVALUATE MODEL (FORWARD)

COMPUTE LOSS

ACCUMULATE GRADIENT OF LOSS  
(BACKWARD)

UPDATE MODEL WITH ACCUMULATED GRADIENT

Ⓒ

FOR N EPOCHS:

SPLIT DATASET IN MINIBATCHES

FOR EVERY MINIBATCH:

WITH EVERY SAMPLE IN MINIBATCH:

EVALUATE MODEL (FORWARD)

COMPUTE LOSS

ACCUMULATE GRADIENT OF LOSS (BACKWARD)

UPDATE MODEL WITH ACCUMULATED GRADIENT

Ⓑ

FOR N EPOCHS:

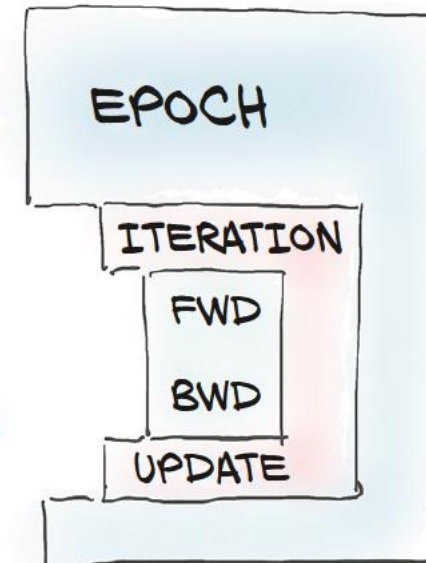
WITH EVERY SAMPLE IN DATASET:

EVALUATE MODEL (FORWARD)

COMPUTE LOSS

COMPUTE GRADIENT OF LOSS  
(BACKWARD)

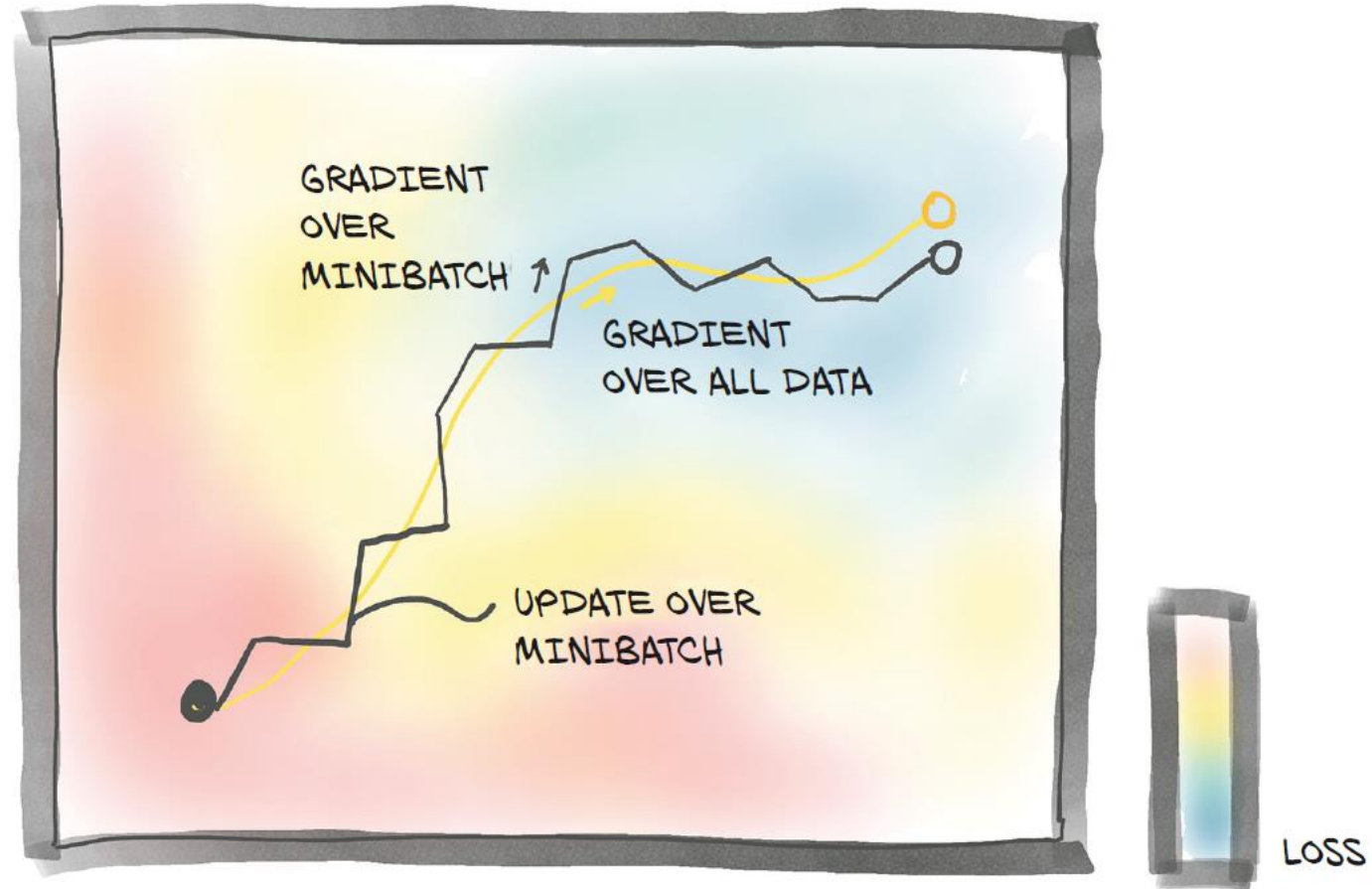
UPDATE MODEL WITH GRADIENT



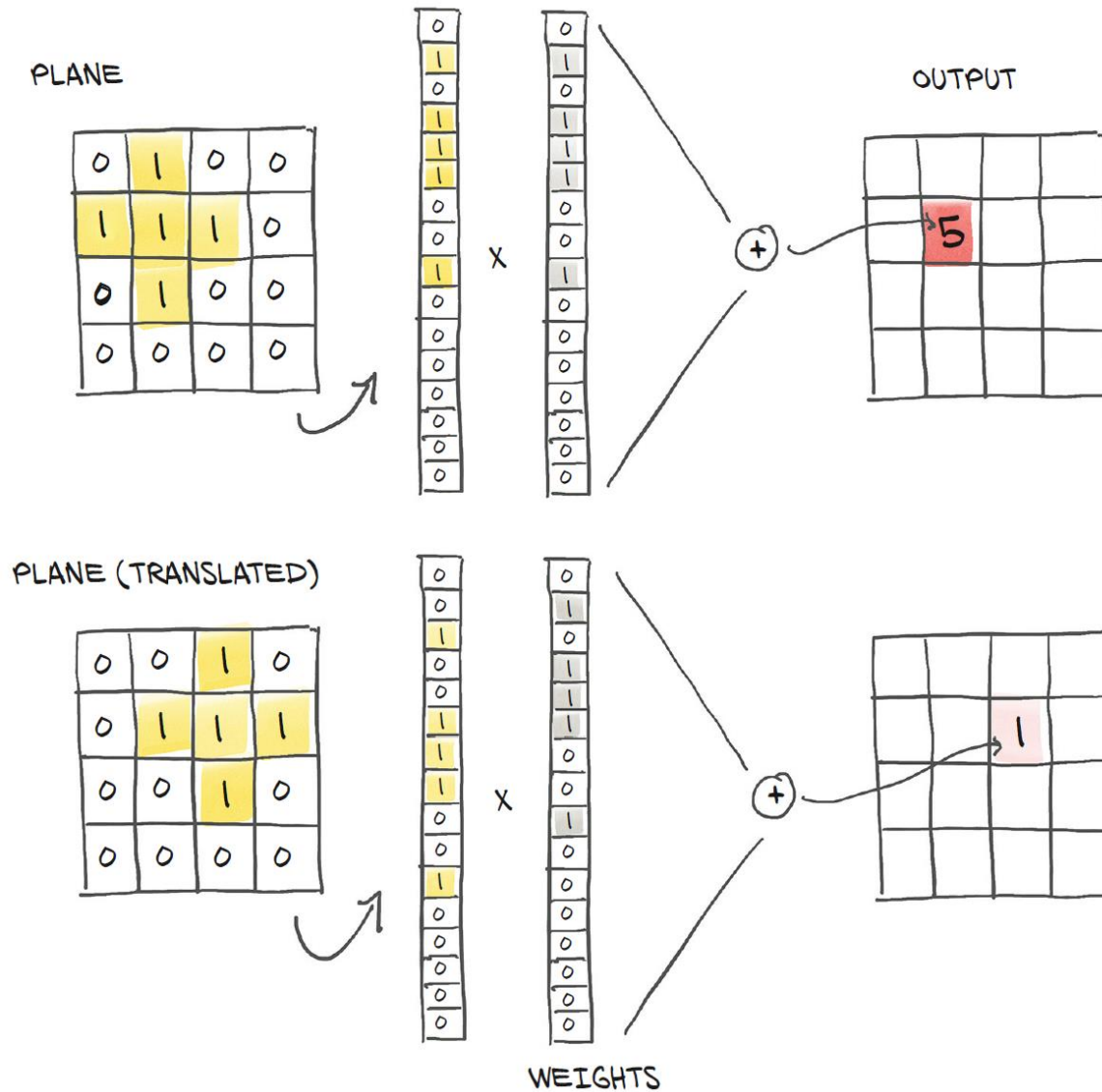
Quelle: Deep Learning with PyTorch



# Stochastic Gradient Descent



# Lack of Translation Invariance



# Aufgaben

- Verwenden Sie die MSELoss Function und vergleichen das Resultat mit der NLLLoss Function.
- Versuchen Sie durch verkleinern des Netzwerkes das Übertraining zu verringern.

Abgabe per Github bis zum 30.05.2023 23:59 Uhr