

# Machine Learning for Agricultural Applications

## Assignment 4

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### Task 1 – Keras

[20 points]

In this exercise, we use Keras to train a multilayer perceptron. Load the `fashion_mnist` data set, which is included in Keras. Use the sequential API in Keras to create a multilayer perceptron, that is, a neural network consisting of multiple fully connected layers. You can freely choose the number of hidden layers and neurons per layer. Train the model on the training part of the `fashion_mnist` data set and evaluate it on the test part of the data set. Try to improve the model to reach a test accuracy around 90%. Use only dense layers and play with the number of hidden layers, nodes and number of training epochs.

If you worked on your computer locally, also try to run the program on **Colab** or a similar cloud platform that allows to run your code on GPUs.

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Solution:     `ex4_fashion_mnist.py`

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### Task 2 – AutoDiff in TensorFlow

[15 points]

Write a python script that uses the `GradientTape` function of TensorFlow for calculating the gradient of the function

$$f(x_0, x_1, w_0, w_1, w_2) = \left[ \tanh \left( \frac{1}{w_0 x_0 + w_1 x_1 + w_2} \right) - 1 \right]^2, \quad x_0, x_1, w_0, w_1, w_2 \in \mathbb{R}.$$

Also implement the forward pass as in Assignment 3 and use `GradientTape` to calculate all intermediate (downstream) gradients that you computed in the backward pass of Assignment 3.

Use the values  $x_0 = 1.2$ ,  $x_1 = 3.1$ ,  $w_0 = 1.5$ ,  $w_1 = -1.5$ ,  $w_2 = 2.0$  and compare the outputs with the results from Assignment 3.

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Solution:     `ex4_GradTape.py`

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### Task 3 – Regression with TensorFlow

[15 points]

Use TensorFlow to perform a linear regression with squared loss on 3D random data, i.e. fit a plane

$$f(x, y) = z = w_1x + w_2y + b, \quad \text{with } x, y, w_1, w_2, b \in \mathbb{R}$$

through a cloud of points. Pick arbitrary "true" values for the parameters  $w_1$ ,  $w_2$ ,  $b$  and generate artificial data by drawing  $x$  and  $y$  values randomly from `tf.random.uniform`, computing  $z$  from your function and adding noise using `tf.random.normal` to the  $z$ -labels.

Use the `GradientTape` function of TensorFlow to implement a gradient descent algorithm for training (fitting) the model to your artificially generated data.

Plot the convergence of the parameters as well as the point cloud and the resulting plane.

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Solution:     `ex4_tf_linReg2D.py`

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