ML Exercise 2: Neural Network

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1 Implementation

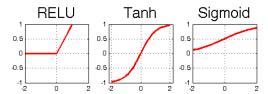
- 2 Hyperparameter optimization
- 3 Comparison with other methods

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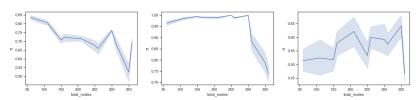
- Made use of object-oriented programming principles by encapsulating the layer and network behavior in Python classes
- Code structure inspired by Omar Aflak

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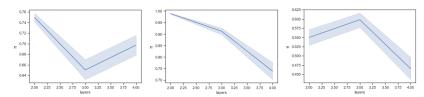
- Split into train, validation and test set
- Different hyperparameter settings
 - nodes per layer 10, 50, 100 or 200
 - hidden layers 2, 3 or 4
 - threshold 0.3, 0.5 or 0.9
 - activation functions for hidden layers Tanh or ReLU



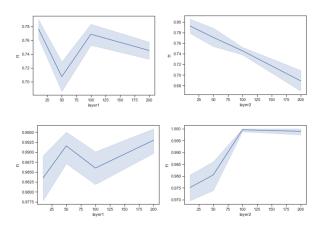
Total number of nodes; hepatitis, breast cancer, dry beans



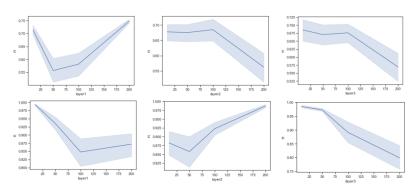
Number of layers; hepatitis, breast cancer, dry beans



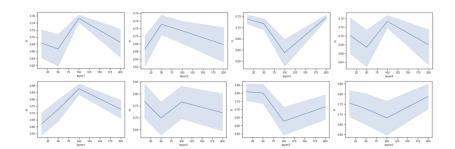
Number of nodes per layer - 2 layer NNs; hepatitis & breast cancer



Number of nodes per layer – 3 layer NNs; hepatitis & breast cancer



Number of nodes per layer – 4 layer NNs; hepatitis & breast cancer



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3 Comparison with other methods

Table: Comparison with other classification methods (5-fold cross-validation with regards to the mean of the F1-score)

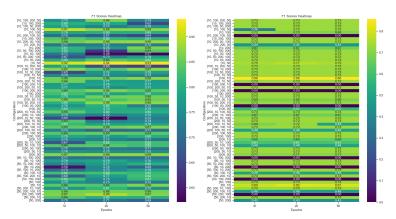
Dataset	Our NN	Keras
Hepatitis	0.832043	0.704000
Breast Cancer	0.965974	0.943478
Dry Beans	0.885843	0.915548

Table: Comparison with other classification methods from exercise 1 (5-fold cross-validation with regards to the mean of the F1-score)

	Our NN		SVM	DecTree
Hepatitis	0.832043	0.898341	0.917261	0.876667
Breast Cancer	0.965974	0.962698	0.967143	0.828402
Dry Beans	0.885843	0.923538	0.929239	0.893124

Appendix: Impact of Hyperparameters

Hepatitis



Appendix: Impact of Hyperparameters

Breast Cancer

