MSE TSM Deep Learning

# $\begin{array}{c} {\rm Practical~Work~10-25/04/2024} \\ {\rm Recurrent~Nets} \end{array}$

### **Objectives**

The main objective of this PW is to build and train recurrent networks and experience some of the difficulties to do so. You can carry through this practical work with your framework of choice (PyTorch or Tensorflow/Keras).

#### Submission

— **Deadline**: Wednesdays 8th May, noon

— **Format** : ipynb

#### Exercise 1 Names Classification

In this exercise, you will develop a character-based model for classifying last names by language.

- a) Download the zip-file names\_classification.zip with a jupyter notebook and a data folder, unzip the file, open the jupyter notebook.
- b) The first cells contain some helper functions to load the training and test data. Complete the generation of the alphabet (characters to be represented) and the functionality to create the vector representation for the characters. Use a one-hot-vector representation.
- c) Implement a many-to-one model consisting of a single RNN-layer (Elman) and a linear layer (with or without softmax) for the classification. Train the model try different numbers of hidden units (dimension of hidden state vector) and play with different batch sizes. Analyse the accuracy and confusion matrix. Report about your findings in the notebook: Which number of hidden layers work best?
- d) Can you improve the results by stacking RNN layers? Do the comparison on the basis of the accuracy but also the confusion matrix.

- e) Figure out a treatment of class imbalance, implement it and apply it to the given example. What improvement can you achieve? Keywords: Minority over-sampling, class weights in loss. Make sure that you don't leak any information from the training set into the test set!
- f) Explain why a treatment of class imbalance is important.

## Exercise 2 Human Activity Recognition

In this example, you will study models for human activity recognition from accelerometer and gyroscope data recorded by smartphones (as presented in class). The goal is to explore the hyperparameter/model space to find the best suited model for the given problem.

- a) Download the zip-file activity\_recognition.zip with a jupyter notebook and a data folder, unzip the file and open the jupyter notebook.
- b) Load the training and test data by running the first cells with suitable helper functionality.
- c) Now implement and train at least 5 different many-to-one models models by choosing at least two types from {Elman RNN | LSTM | Conv1d | Dense | Linear} and varying number of layers and (hidden) units per layer. Make a decent choice of hyper-parameters (batchsize, nepochs, nhidden, clipnorm) until the training gets more or less stable.
- d) Compare the models quantitatively and qualitatively, on the basis of the learning curves and the achieved classification performance. Spot potential issues for specific settings / models. Make sure that the results are robust against different initialisation seeds.

## Exercise 3 Optional: Review Questions

- a) What are the different types of tasks that can be addressed with recurrent neural networks. Give for each type an example application.
- b) Compute the number of parameters to be trained for a two-layer Elman RNN and softmax with hidden state dimensions 32 and 64, respectively, 10 classes to classify in the softmax, inputs given by sequences of length 100 and each element a vector of dimension 30.
- c) Compute the number of parameters to be trained for a two-layer *LSTM* and *softmax* with hidden state dimensions 32 and 64, respectively, 10 classes to classify in the softmax, inputs given by sequences of length 100 and each element a vector of dimension 30.
- d) Why is gradient clipping rather needed in long than in short sentences or texts?
- e) Describe why Elman networks have problems in learning long-term dependencies.
- f) In what situations would you expect LSTMs (by its design) to perform better than Elman networks?
- g) In some sequence problems, convolutional approaches may be better suited and more robust to learn. How could you apply conv layers when learning a time series forecasting task? What do you need to pay attention to?

h)	What is the difference kernel size $(w, 1)$ ?	between	using	a Conv1d	with	kernel	size w	and a	Conv2d	with
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