# Deep Learning in NLP: Summary / Takeaway

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# **Overview**



#### **2** Further exploration of DL in NLP

**3 AP Projects** 

### What was it all about?

- \* Introductory course to Deep Learning in NLP and applications with Keras.
- \* Now you are able to understand and build neural network models using Keras and Python.
- You can solve some NLP Problems using deep learning architectures: POS-tagging, named entity recognition, chunking, sentiment analysis, text classification, semantic role labeling, text generation.
- \* You can better understand and adapt code you find on GitHub according to your deep learning task (or write your own code based on some scientific article).

#### Deep Learning Projects: 6 Basic Steps

- Problem or task: machine translation, sentiment analysis, text classification, part-of-speech tagging, text generation etc.
- 2 Input data points (features): files of people speaking, text files, images etc.
- 3 Examples of the expected outputs (tags or labels): transcripts of sound files, positive/negative, cat/dog/fish etc.
- **4** Suitable algorithm to train on those data.
- Way to evaluate how good the algorithm and the model is (distance between the algorithm's output and expected output).
- 6 Present your results.

# Neural Network Models we used in class

- \* Feed-Forward Network (FFN) (Boston housing)
- \* Simple Recurrent Neural Network (RNN) (Name recognition)
- \* Long Short-Term Memory (LSTM) (POS-tagging)
- $\star$  Gated Recurrent Unit (GRU) + Bidirectional GRU (used for sequence labeling)
- \* Bidirectional Recurrent Neural Network (BiLSTM) (POS-tagging, chunking, named entity recognition)
- $\star\,$  Convolutional Neural Network (CNN) for character models  $\rightarrow\,$  capture morphological information
- \* Multichannel CNN (sentiment analysis)
- $\star$  Conditional Random Field (CRF) + BiLSTM (semantic role labeling, sequence labeling in general)

### Use embeddings

- \* Character-level embeddings (to encode characters in words).
- Word-level embeddings (one-hot, pretrained, own embeddings: Word2Vec).
- $\star$  Sentence-level embeddings + text-level embeddings.

### What else?

- $\star\,$  Use Keras and Keras functional API.
- $\star\,$  Prepare (vectorize) your data for training.
- \* Visualize your neural network model (using plt model).
- $\star$  Save and re-use pre-trained models.
- $\star$  Make predictions with the model.
- ★ Evaluate the model performance with the following metrics: accuracy, recall + precision + F1 score, confusion matrix.
- $\star\,$  Use early stopping to save the best model and stop training.
- $\star\,$  Use fit generator to avoid out-of-memory (OOM) issues.
- $\star\,$  Use LaTeX to describe your experiments and results.

# Books on Deep Learning in NLP

- \* Chollet (2017) Deep Learning with Python.
- \* Goldberg and Hirst (2017) Neural Network Methods for Natural Language Processing.
- \* Brownlee (2016). Deep Learning for Natural Language Processing. Develop Deep Learning Models for Natural Language in Python.
- \* Geron, Aurelien (2017) Hands-on machine learning with Scikit-Learn and TensorFlow : concepts, tools, and techniques to build intelligent systems.
- \* Harrison and Honnibal (2020) Deep Learning with Text: A Modern Approach to Natural Language Processing with Python and Keras.

# Some good tutorials

- $\star$  https://keras.io/
- $\star$  https://machinelearningmastery.com/

# **AP Projects: Evaluation Criteria**

- $\star$  Clarity of writing / presentation: 1-5
- $\star$  Research question / hypothesis: 1-5
- $\star$  Soundness / correctness: 1-5
- ★ Empirical evaluation: 1-5
- ★ Dataset / effort: 1-5

#### **References** I

- Brownlee, J. (2016). Deep Learning for Natural Language Processing. Develop Deep Learning Models for Natural Language in Python.
- Chollet, F. (2017). Deep Learning with Python. Manning Publications.
- Geron, Aurelien (2017). Hands-on machine learning with Scikit-Learn and TensorFlow : concepts, tools, and techniques to build intelligent systems. O'Reilly Media, Sebastopol, CA.
- Goldberg, Y. and Hirst, G. (2017). Neural Network Methods for Natural Language Processing. Synthesis Lectures on Human Language Technologies. Morgan & Claypool Publishers.
- Harrison, P. and Honnibal, M. (2020). Deep Learning with Text: A Modern Approach to Natural Language Processing with Python and Keras. O'Reilly Media.