

# Assignment 3: Scientific Area And Similarity Classifier (SAASC)

Advanced Information Retrieval 22/23

Group 23:

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# Introduction

## Research Questions:

## (Goals) Motivation:

- Get training data via Arxiv.
  - Automated retrieval of recent papers for various categories
- Implement Network classifying scientific belonging
  - Extract nouns via text processing (textblob)
  - Creation and training of Classification Network based on nouns
- Classify and Compare any Papers (URLs).
  - Input:
    - URL's of papers where the Network is applied on
  - Output:
    - Categories per URL
    - Plot of training/network statistic
    - Plots of Similarity

# Data + Methods TODO(optional theoretical background)

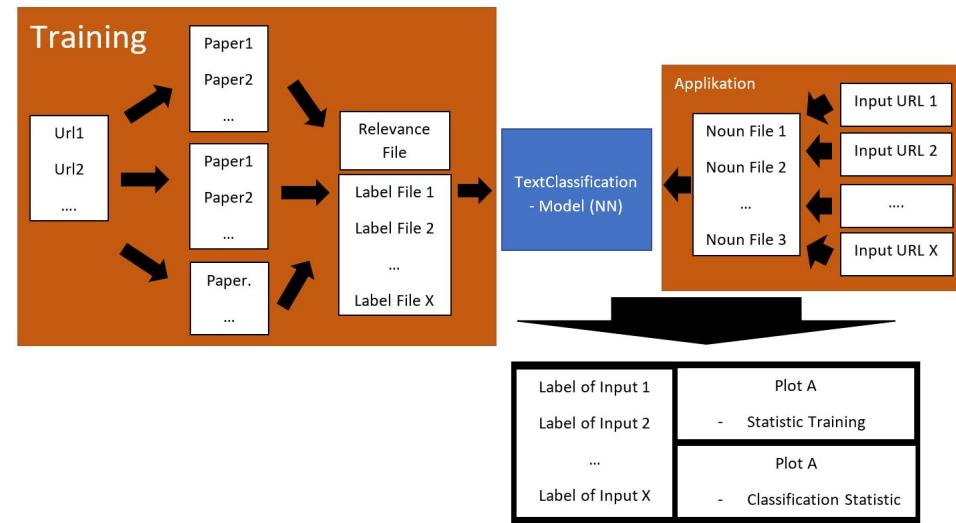
- papers from <https://arxiv.org/>
  - Retrieves automatically x newest papers of configured categories y categories.
  - Downloads content of URL and processes its nouns into labelled files which are stored via caching.
  - Creates labelled dataset for machine learning (Categorization)
- cs
  - [https://arxiv.org/list/cs/pastweek?  
show=1000](https://arxiv.org/list/cs/pastweek?show=1000)
- q-bio
  - [https://arxiv.org/list/q-bio/pastwe  
ek?show=1000](https://arxiv.org/list/q-bio/pastweek?show=1000)
- physics
  - [https://arxiv.org/list/physics/past  
week?show=1000](https://arxiv.org/list/physics/pastweek?show=1000)
- eess
  - [https://arxiv.org/list/eess/pastwe  
ek?show=1000](https://arxiv.org/list/eess/pastweek?show=1000)
- econ
  - [https://arxiv.org/list/econ/pastwe  
ek?show=1000](https://arxiv.org/list/econ/pastweek?show=1000)

# Data + Methods TODO(optional theoretical background)

- automatically retrieving papers from <https://arxiv.org/>
  - classified into five scientific areas
    - cs
      - <https://arxiv.org/list/cs/pastweek?show=1000>
    - q-bio
      - <https://arxiv.org/list/q-bio/pastweek?show=1000>
    - physics
      - <https://arxiv.org/list/physics/pastweek?show=1000>
    - eess
      - <https://arxiv.org/list/eess/pastweek?show=1000>
    - econ
      - <https://arxiv.org/list/econ/pastweek?show=1000>

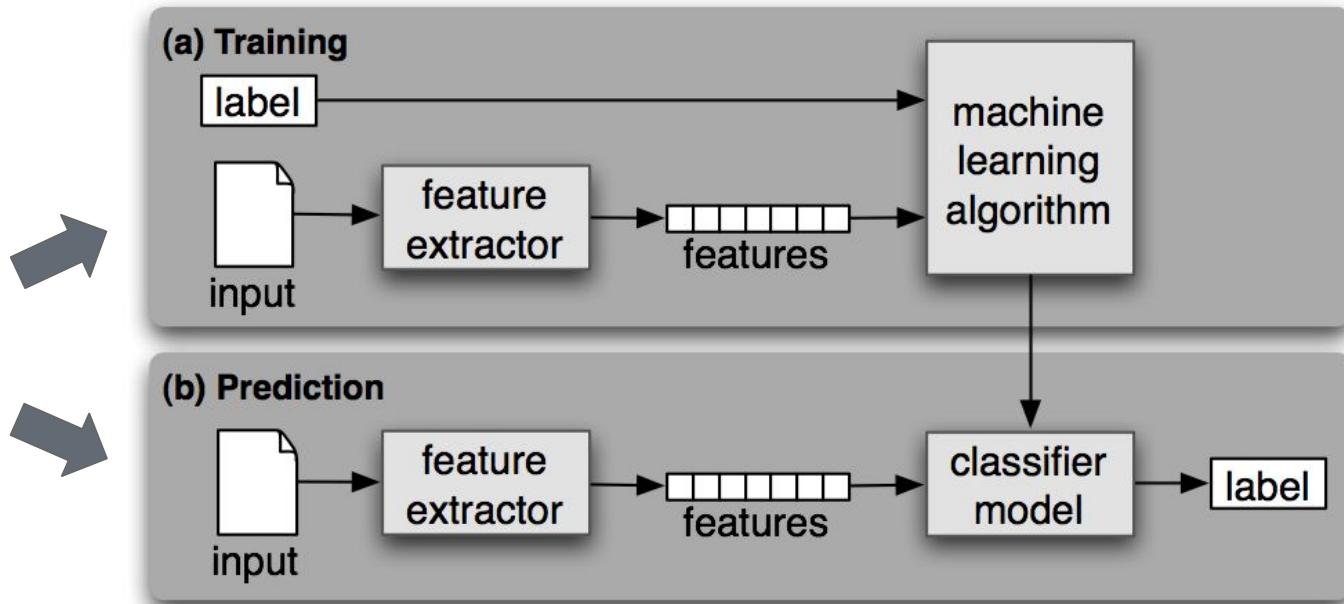
# Architecture (TODO adapt to AIR)

- Technologies
  - Python (Pycharm)
  - Neural Network
- Algorithms
  - Noun extraction
  - Neural Network
    - Loss Function: CrossEntropyLoss
    - Optimizer: SGD
- Libraries
  - Textblob
  - PyPdf2
  - torch
  - urllib.request
  - BeautifulSoup



# Design

Arxiv  
(URL's)



# Results - Analysis and Interpretation (TODO Add Results!!)

## Literature comparison:

- ❖ [How I achieved 90% accuracy on a text classification problem with ZERO preprocessing](#)
  - BERT sentence embeddings
  - used Spark NLP
  - 4 categories
  - Accuracy: 90%
- ❖ [Text Classification with TF-IDF, LSTM, BERT: a comparison of performance](#)
  - 5 categories
  - TF-IDF (97.9%)
  - Recurrent Neural Networks (94.6%)
  - Bert Language Model (96.6%)

- ❖ Test Accuracy
  - > 75%
- ❖ Comparison with literature
  -
- ❖ Comparison with DL
  -

# Literatur Vergleich Neurales Netzwerk

- Wie wurde ähnliches in Literatur umgesetzt + Welche Accuracy wurde erreicht?
  - <https://towardsdatascience.com/how-i-achieved-90-accuracy-on-a-text-classification-problem-with-zero-preprocessing-6acfa96e8d2e>
  - torch.NN Optimizer Adama Cross Entropy 90% accuracy by 5 Classees
  - <https://medium.com/@claude.feldges/text-classification-with-tf-idf-lstm-bert-a-quantitative-comparison-b8409b556cb3>
  - -- TODO christina -> tabelle

# Conclusion (incl. limitations/biases) TODO

bias: computerscience - da training sich an computerscience lehnt