

SIGEVO Summer School,  
10-12 July 2019 - Prague

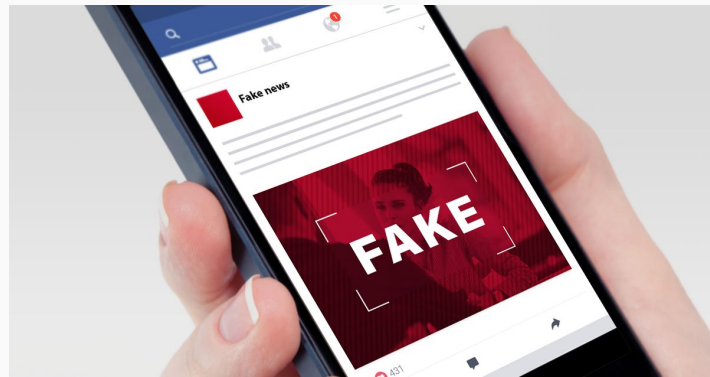


**FAKENEWS**

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# Initial thoughts on the topic “Fake News”

- The context:
  - **Textual** (as articles, tweets, etc.)
  - Images
  - Videos
- The approaches:
  - Network Science
  - **ML using classifiers**
- The target:
  - Containing the spread by the author
  - **Identify the fake news by the contents**



# Final Problem Formulation

- Main Goal: Try to discriminate Fake News from real ones
- Main Focus: Contents of articles
- Dataset: <https://www.kaggle.com/c/fake-news/data>
  - 20k articles



# General Classification Approach

1. Generate the features using the two different preprocessing approaches
  - 1.1. Topic Classification: apply best trained model to the real problem dataset
  - 1.2. doc2vec: transform dataset to vector representation
2. Apply the same 10 classifiers to the specified features
  - 2.1. Only Topic Classification Features
  - 2.2. Only doc2vec Features
  - 2.3. Both, Topic Classification and doc2vec features together
3. Evaluate the classifiers for their accuracy



# Topic Classification Preprocessing

- Data Basis for Training: Reuters-21578 News Article Dataset
  - 21578 instances of articles that are labelled with 135 categories (e.g. Business, Politics,..)
- 1. Train 10 different classifiers with partly different configurations [1]
  - LinearSVC, Decision Tree, Random Forest, kNearestNeighbour, SVM, Logistic Regression, Naive Bayes, AdaBoost, LDA
  - Features: each word with its number of appearances, that appears at least 3 times in the text and is no stop word like e.g. “and”, “it” etc → 26147 features in total
- 2. Choose best of the 10 classifiers and apply Hyperparameter Tuning to get the best combination of parameters
  - LinearSVC

# Topic Classification - Classifier Comparison

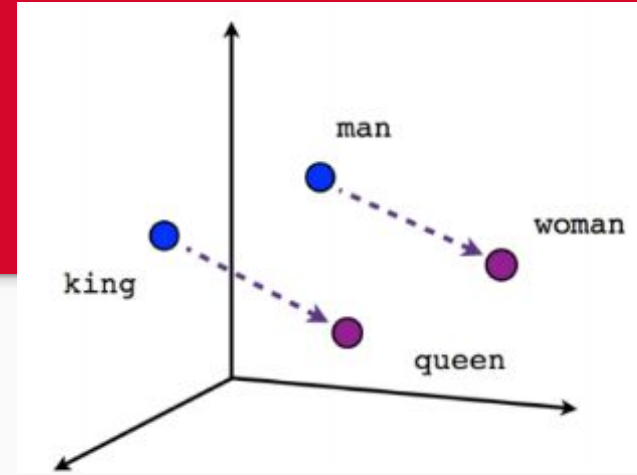
		Accuracy [%]	F1 [%]
	LinearSVC	81.05	84.04
	Logistic Regression (C=1000)	80.79	84.10
	kNN (n=5)	72.97	76.07
	kNN (n=3)	72.28	75.43
	Logistic Regression (C=1)	67.47	67.21
	Random Forest (200 trees)	65.75	64.36
	Random Forest (50 trees)	64.79	63.69
	Decision Tree	55.75	53.23
	Naive Bayes	43.86	47.98
	SVM linear	33.55	29.67

# Topic Classification - Hyperparameter Tuning

- Parameters to tune:
  - C = Penalty Parameter of the error term
    - [1,10,100,1000]
  - multi\_class = determines the multi-class strategy of the LinearSVC classifier
    - "ovr" trains n\_classes one-vs-rest classifiers
    - "crammer\_singer" optimizes a joint objective over all classes
- Accuracy: 82.15%
- Best Parameters found:
  - C = 1
  - multi\_class = crammer\_singer

# Doc2Vec based classification

- using Doc2Vec (Mikilov and Le, 2014)
- based on Word2Vec
- training dataset - <https://www.kaggle.com/c/fake-news/data>
- k-fold cross validation used for evaluation



dataset

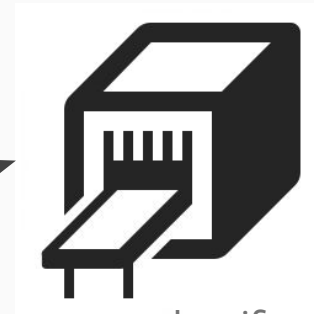


doc2vec model

$(x_1, x_2, \dots, x_n)$



class. learning



classifier



# Adapting the articles.

crisis prevention, and verifiable nuclear disarmament should be substituted for continuing counterproductive wars. Therefore, we, as signers of this petition, call for the immediate cancellation of the F-35 program as a whole, and the immediate cancellation of plans to base any such dangerous and noisy jets near populated areas.

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disarmament should be substituted continuing  
counterproductive wars therefore we **signers**  
**petition** **call immediate** cancellation program  
whole **immediate** cancellation plans any such  
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## Adapting the articles.

crisis prevention verifiable nuclear disarmament  
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# Adapting the articles.

['crisis', 'prevention', 'verifiable', 'nuclear', 'disarmament', 'should', 'be',  
'substituted', 'continuing', 'counterproductive', 'wars', 'therefore', 'we',  
'cancellation', 'program', 'whole', 'cancellation', 'plans', 'any', 'such', 'dangerous',  
'noisy', 'jets', 'we', 'replacing', 'any', 'basing', 'any', 'locations', 'we', 'further',  
'demand', 'money', 'human', 'needs', 'us', 'customer', 'world', 'including', 'climate',  
'change', 'student', 'debt', 'education', 'healthcare', 'housing', 'add', 'your',  
'swanson', 'is', 'an', 'author', 'journalist', 'host', 'he', 'is', 'director', 'coordinator',  
'rootsactionorg', 'books', 'war', 'is', 'lie', 'he', 'blogs', 'at', 'he', 'hosts', 'nation', 'he', 'is',  
'peace', 'prize', 'follow', 'him', 'on', 'twitter', 'support', 'clicking', 'here']

# doc2vec/preprocess - class. comparison

	Accuracy [%]	F1 [%]
LinearSVC	87.48	87.48
Logistic Regression (C=1000)	87.67	87.67
kNN (n=5)	81.60	81.39
kNN (n=3)	81.99	81.82
Logistic Regression (C=1)	87.72	87.72
Random Forest (200 trees)	87.02	87.01
Random Forest (50 trees)	87.07	87.06
Decision Tree	76.19	76.18
SVM adjusted	89.50	89.50
SVM linear	87.62	87.62

# doc2vec/no preprocess - class. comparison

	Accuracy [%]	F1 [%]
LinearSVC	74.93	74.89
Logistic Regression (C=1000)	74.81	74.79
kNN (n=5)	76.50	76.49
kNN (n=3)	75.82	75.81
Logistic Regression (C=1)	74.79	74.76
Random Forest (200 trees)	78.35	78.34
Random Forest (50 trees)	77.94	77.94
Decision Tree	71.13	71.12
Naive Bayes	59.45	57.33
SVM linear	74.79	74.72



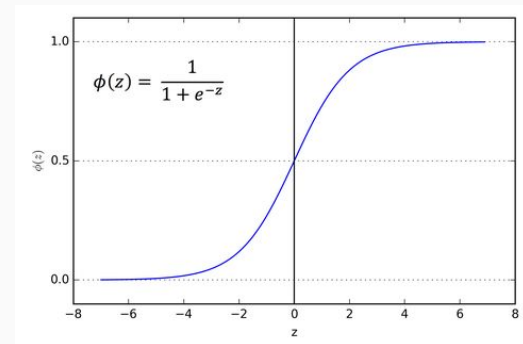
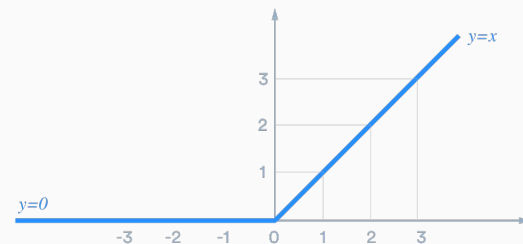
# Results - doc2vec/preprocess + deep learning

Accuracy on a training set: 97.45%

Accuracy on a test set: 89.23%

Activation function: ReLU, Sigmoid

Binary classification



# Overall Results - Preprocessing Approach Comparison

- Scripts are set up & ready.. but algorithms didn't run through so far (they take some time...)
- → So no results available for comparing the preprocessing approaches

# Outlook

- Finish the final and overall comparison of the preprocessing approaches
- Set up some sort of framework to be able to continuously/regularly and automatically train the models for continuous improvement/adaptation (language is changing!)
- Apply the trained models in real life e.g. on some sort of news website or similar
- Take satire into account

Thank you!

# References

[1] <https://martin-thoma.com/nlp-reuters/>