

# Text Classification: 20newsgroups

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Natural Language Processing

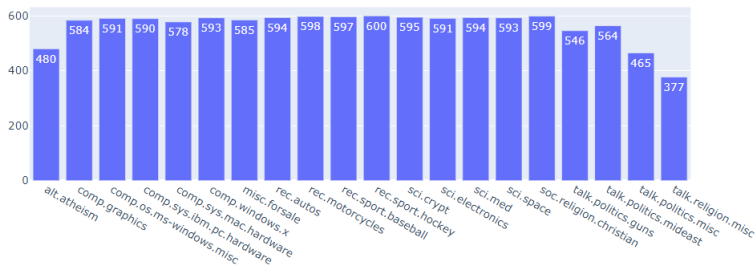
# 1. The Dataset: 20 newsgroups

- Training data: 11314 texts  
Test data: 7532 texts

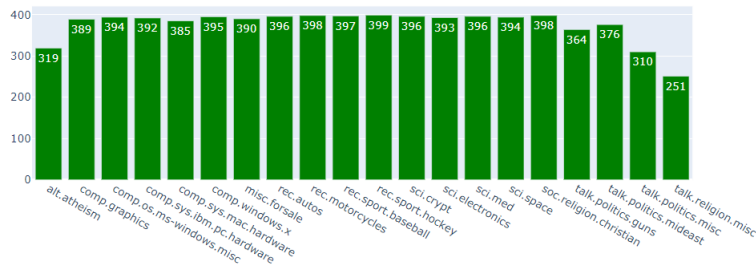
## Categories :

- *alt.atheism*
- *comp.graphics*
- *comp.os.ms-windows.misc*
- *comp.sys.ibm.pc.hardware*
- *comp.sys.mac.hardware*
- *comp.windows.x*
- *misc.forsale*
- *rec.autos*
- *rec.motorcycles*
- *rec.sport.baseball*
- *rec.sport.hockey*
- *sci.crypt*
- *sci.electronics*
- *sci.med*
- *sci.space*
- *soc.religion.christian*
- *talk.politics.guns*
- *talk.politics.mideast*
- *talk.politics.misc*
- *talk.religion.misc*

## Training set: Class distribution



## Test set: Class distribution



## 2. ML approach

- Preprocessing
  - Preprocess1: lower characters, nltk's word\_tokenize
  - Preprocess2: lower characters, nltk's word\_tokenize, remove small words, remove stopwords, nltk's PorterStemmer
- Tfidf Vectorizer
  - tokenizer: Preprocess1, Preprocess2
  - n-grams: uni-grams, uni-grams & bi-grams, bi-grams
  - norm: 'l1', 'l2'
- Classifiers
  - Support Vector Machine
  - Multinomial Naive Bayes
  - Random Forest

- Evaluation of 36 pipeline models wrt validation accuracy score
- Optimal: SVM with Preprocess2, uni-grams & l2 norm
- Overall test scores
  - accuracy: 0.66
  - precision (weighted): 0.68
  - recall (weighted): 0.66
  - f1-score (weighted): 0.66

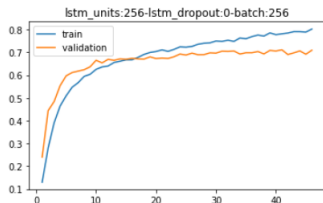
### 3. DL approach

- Kera's Tokenizer
  - Tokenize words and lower characters.
  - Learns 200-dim representations per text. Sequence of integers.
  - Consider 20K most common words and assign integers based on their frequency in descending order
- GloVe Embeddings
  - Pre-trained word vectors of dim 100
  - Matrix of shape (20K,100); eventually describes the weights of the Embedding layer of the NN
- The model: GloVe-based BiLSTM architecure
  - Hyper-params tuned: lstm nodes and dropout and training batch size
  - No need to tune: Adam(0.001), Categorical-Crossentropy Loss, Tanh activation fct

# Accuracy history and Summary

- Overall test scores

- accuracy: 0.67
- precision (weighted): 0.68
- recall (weighted): 0.67
- f1-score (weighted): 0.67



Layer (type)	Output Shape	Param #	Connected to
input_26 (InputLayer)	[(None, 200)]	0	[]
embedding_25 (Embedding)	(None, 200, 100)	2000000	['input_26[0][0]']
spatial_dropout1d_25 (Spatial1D dropout1D)	(None, 200, 100)	0	['embedding_25[0][0]']
bidirectional_25 (Bidirectional1)	(None, 200, 512)	731136	['spatial_dropout1d_25[0][0]']
global_average_pooling1d_25 (GlobalAveragePooling1D)	(None, 512)	0	['bidirectional_25[0][0]']
global_max_pooling1d_25 (GlobalMaxPooling1D)	(None, 512)	0	['bidirectional_25[0][0]']
concatenate_25 (Concatenate)	(None, 1024)	0	['global_average_pooling1d_25[0][0]', 'global_max_pooling1d_25[0][0]']
dropout_125 (Dropout)	(None, 1024)	0	['concatenate_25[0][0]']
dense_125 (Dense)	(None, 512)	524800	['dropout_125[0][0]']
dropout_126 (Dropout)	(None, 512)	0	['dense_125[0][0]']
dense_126 (Dense)	(None, 512)	262656	['dropout_126[0][0]']
dropout_127 (Dropout)	(None, 512)	0	['dense_126[0][0]']
dense_127 (Dense)	(None, 256)	131328	['dropout_127[0][0]']
dropout_128 (Dropout)	(None, 256)	0	['dense_127[0][0]']
dense_128 (Dense)	(None, 128)	32896	['dropout_128[0][0]']
dropout_129 (Dropout)	(None, 128)	0	['dense_128[0][0]']
dense_129 (Dense)	(None, 20)	2580	['dropout_129[0][0]']
Total params: 3,685,396			
Trainable params: 1,685,396			
Non-trainable params: 2,000,000			

## 4. Conclusion

- Summary:

Test Summary (Weighted metrics)				
Model	Accuracy	Precision(W)	Recall(W)	F1-score(W)
SVM	0.66	0.68	0.66	0.66
BiLSTM	0.67	0.68	0.67	0.67

Thank you!