Model Performance

Model Name: LocationClassificationV1 Test Date: 17/03/2022 19:20:25 Creator: Tobias Rothlin



Overview

ML Principle: Multinomial Naive Bayes

References:

- NultinomialNB Explained
- Stanford NLP Course
- Stanford NLP Lecture
- Engilsh Stopwords

Algorithm Description:

The learning algorithm used in this classification is the Multinomial Naïve Bayes. This approach was chosen as it is easy to implement and is computational very efficient. The first step in the classification pipeline is removing all strop words for example 'i', 'me', 'my', 'myself', etc. A list of English stop word is provided by the nltk module. The stop words remover just removes every word that is in the list of stop words. Next the sentence is passed through the stemmer. Stemmers remove morphological affixes from words, leaving only the word stem. This is done with the PorterStemmer class from the nltk module. The final preprocessing step is to vectorize the sentence. This results in a bag of words representation of the sentence. First all the words must be tokenized and then counted. The result will be a numerical feature vector. To generate this vector the CountVectorizer class from sklearn is used. This class implements both tokenization and occurrence counting in a single class. With the sentence now represented in a vector the Naïve Bayes classifier can work with this vector. For the implementation of the Naïve Bayes classifier can work with this vector.



Classification Pipeline

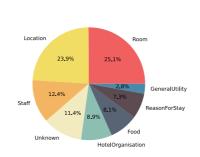
Metrics

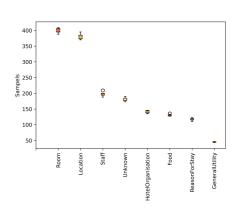
Data: ClassifiedDataSetV1.2 with 10 folds cross validation

Split seed: 4.989823

Training Dataset

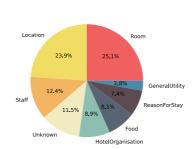
Classes	Number of samples		
Room	399		
Location	380		
Staff	197		
Unknown	181		
HotelOrganisation	141		
Food	129		
ReasonForStay	117		
GeneralUtility	45		

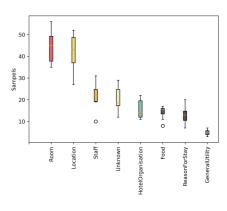


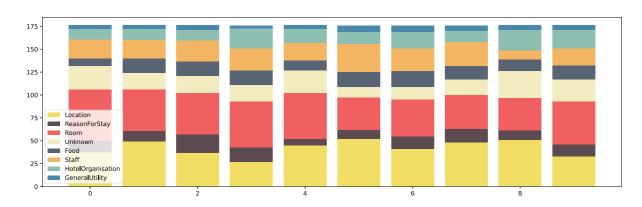


Test Dataset

Classes	Number of samples
	•
Room	44
Location	42
Staff	21
Unknown	20
HotelOrganisation	15
Food	14
ReasonForStay	13
GeneralUtility	5

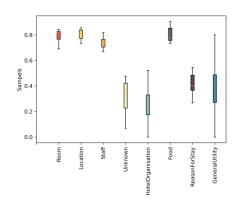






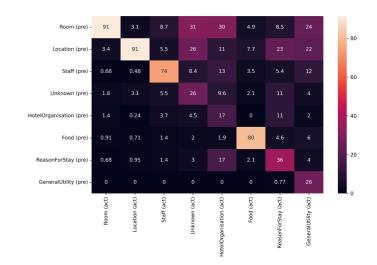
Classification Performance

Classes	Precision	Recall	F1 Score
Room	70.16%	91.16%	79.29%
Location	72.18%	91.43%	80.67%
Staff	72.52%	73.85%	73.18%
Unknown	43.70%	25.74%	32.40%
HotelOrganisation	40.00%	16.67%	23.53%
Food	81.43%	79.72%	80.57%
ReasonForStay	49.47%	36.15%	41.78%
GeneralUtility	92.86%	26.00%	40.62%
Accuracy			68.12%
Macro Average	65.29%	55.09%	56.51%
Weighted Average	65.26%	68.12%	64.77%



ConfusionMatrix:

Normalised ConfusionMatrix:



F1 Socre by split:

