

STRATEGY USED

- Use frequency tables to detect the language
 - 1. Split text into words
 - 2. For each word, check if it is in the top N words in the language
 - 3. Add the number of "hits" for the words in the list
 - 4. Pick the language which results in the highest number

COMPETINGSTRATEGIES

IS THIS THE ONLY WAY TO DO IT?

- •Definitely not. For example the paper suggested using character unigrams, bigrams, or trigrams, and then computing the KL distance of a testing sequence with the histogram/probability function generated with a training set.
- Another approach may be to use something like an RNN to do the categorisation, probably at a character level
- There are several "older" Machine Learning approaches. For example create a vector of meta data (character unigrams, bigrams, word lengths, etc.), and classify that vector using SVM, random forest, etc.,
- There are likely ways of combining these approaches, and many others

COMPETINGSTRATEGIES

WHY DID I CHOOSE THIS APPROACH

- •When exploring the problem, I came across https://github.com/hermitdave/
 FrequencyWords, realised it had all of the frequency counts for the languages from the data set, and saw it as easy to implement.
- Had confidence that frequency counts could lead to accurate predictions
- Realised that the approach was simple, and would likely work for many situations

COMPETINGSTRATEGIES

STRATEGY DOWNSIDES

- The strategy needs at least a few "words" from the language to make an accurate prediction.
- Using a completely character-based approach (like character bigrams, trigrams, etc.),
 for many languages, may produce accurate results with less input data needed.
- Using a character-based approach may be less computationally complex.
 - Computing histograms is computationally fast
 - Computing KL distance, or projection of a vector into a multi-dimensional space is O(1), etc. for SVM, could be computationally fast.

CONTINUING DEVELOPMENT

WHAT WOULD I TRY NEXT

- I would likely try a character based approach (unigrams, then bigrams, followed by KL distance)
- I would try creating a meta-data vector, and using SVM, or random forest, etc. Simply because if you set up the data, training, testing, for one method, the consistency of the scikit-learn interface allows testing of several ML techniques easily.
- I would try an RNN or CNN for detection, it would probably work well, especially if there was additional data.
- Would imagine, the winning strategy may have a combination of techniques

HOW DID THE STRATEGY PERFORM?

Text sequences: 5, 15, and 15 word sequence

Overall average on all languages:

- 71.7% (5 words given)
- 93.7% (15 words given)
- 97.6% (30 words given)

HOW DID THE STRATEGY PERFORM?

Text sequences: 5, 15, and 15 word sequence (note there is variability given random choice of word sequences)

Strong performance on Cyrillic character sets (EL, BG):

• Easily get 100% with 30 words

Weaker performance on English, which can be mistaken for languages such as NL, because of the character set being the same.

HOW DID THE STRATEGY PERFORM?

Text sequences: 5, 15, and 15 word sequence (note there is variability given random choice of word sequences)

Strong performance on EL:

- 96% (5 words given)
- 100% (15 words given)
- 100% (30 words given)

HOW DID THE STRATEGY PERFORM?

Sentences: minimum 5 words, average sentence length 20.4 words

Overall average on all languages:

- 97.14%
- 9 languages had 100% accuracy, 50 trials for each language (random sentences chosen)
- Lowest accuracy 86%, RO, majority of errors are from sentences with low numbers of words.

ACCURACY - BIGRAM

HOW DID THE STRATEGY PERFORM?

Sentences: minimum 5 words

Average accuracy: 94.9%

bigramResuits

bigramResuits			
bg	100	100	
cs	100	97	
da	100	91	
de	100	92	
el	100	100	
en	100	8	
es	100	56	
et	100	80	
fi	100	92	
fr	100	57	
hu	100	100	
it	100	57	
lt	100	95	
lv	100	100	
nl	100	13	
pl	100	96	
pt	100	82	
ro	100	90	
sk	100	73	
sl	100	26	
SV	100	71	

ACCURACY-TRIGRAM

HOW DID THE STRATEGY PERFORM?

Sentences: minimum 5 words

Average accuracy: 94.9%

Topons	_trigitarii_teetiirig	
bg	100	98
CS	100	99
da	100	99
de	100	99
el	100	100
en	100	80
es	100	89
et	100	99
fi	100	100
fr	100	93
hu	100	100
it	100	75
It	100	95
Iv	100	100
nl	100	98
pl	100	100
pt	100	85
ro	100	97
sk	100	95
sl	100	95
SV	100	97