SVM_spam_classifier

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- 0.0.1 Project and data are based on a free, online course of machine learning https://www.coursera.org/learn/machine-learning. I wholeheartedly recommend this!
- 0.1 I will show how do it in Python:

return re.sub('<[^<>]+>', ' ', text)

ن it '

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+ apply SVM (support vector machine) to spam classification,
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+ use regular expressions.

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[1]: import re
    import string
    import scipy.io
    import numpy as np
    import scipy.io
    from nltk.stem import PorterStemmer
    from nltk.tokenize import word_tokenize
    import warnings
    import sys
    # ignore warnings
    warnings.filterwarnings('ignore')
    # write packages and python version to file
    ! python -m pip list > packages_versions.txt
    # a append to file
    with open('packages_versions.txt', 'a') as f:
        f.write('Python version ' + str(sys.version))
    I \cap I \cap I
[2]: def remove_html_tags(text):
        111
        Removes html tags.
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assert remove_html_tags('deal with it') == ' deal with_

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def normalizing_URLs(text):
    Replaces URL to " httpaddr ".
    return re.sub('(http|https|ftp)://[^\s]*', ' httpaddr ', text)
assert normalizing_URLs('herehttps://en.wikipedia.org/wiki/URL') == 'here_
\hookrightarrowhttpaddr '
def normalizing_emails(text):
    Replaces email adress to " emailaddr ".
    return re.sub('[^\s]+0[^\s]+', ' emailaddr ', text)
assert normalizing_emails('contact awokado@gmail.com') == 'contact emailaddr '
def normalize_dolar(text):
   Replaces $ sign to " dolar ".
    return re.sub('[$]+', ' dollar ', text)
assert normalize_dolar('less$$$bargain$') == 'less dollar bargain dollar '
def normalize_numbers(text):
    Replaces $ sign to " number ".
    return re.sub('[\d]+', ' number ', text)
assert normalize numbers('for 3 day') == 'for number day'
def remove_punct(text):
    111
    Removes all punctuation marks.
    translator = str.maketrans(string.punctuation, ' '*len(string.punctuation))
    return text.translate(translator)
assert remove_punct('only#today!!!') == 'only today
def remove_multi_spaces(text):
    Replaces multi spaces to one.
   return re.sub(' +', ' ', text)
```

```
def proccess_email(text):
       Normalizes email.
       text = text.lower()
       text = remove_html_tags(text)
       text = normalizing_URLs(text)
       text = normalizing_emails(text)
       text = normalize_dolar(text)
       text = normalize numbers(text)
       text = remove_punct(text)
       text = text.replace('\n', '')
       text = remove_multi_spaces(text)
       return text
   def email_to_features(email_name, ps, vocab):
       Sets 1 at right position, when word from vocab occurs in email.
       with open(email_name, 'r') as file:
            email = file.read()
            email = proccess_email(email).split()
            email = [ps.stem(w) for w in email]
            email = sorted(set(email), key=email.index)
       ind list = []
       for word in email:
           try:
                ind_list.append(vocab_dict[word])
            except KeyError:
                pass
       init = np.zeros((1,1899))
       for ind in ind_list:
            init[0,ind] = 1
       return init
[3]: # use porterStemmer to treat program and programmer as 'program'
   ps = PorterStemmer()
   with open('vocab.txt', 'r') as file:
       vocab = file.read().replace('\n', '')
       vocab = vocab.replace('\t', ' ')
       vocab = remove_multi_spaces(vocab)
       vocab = vocab.split()
       vocab_dict = {}
       for i, ind in enumerate(range(1, len(vocab), 2)):
            vocab_dict[vocab[ind]] = i
```

```
rev_vocab_dict = dict(zip(vocab_dict.values(), vocab_dict.keys()))
[4]: # Train model
   data = scipy.io.loadmat('spamTrain.mat')
   # X indicates which words from spam vocabulary list are used in email.
   X = data['X']
   Y = data['y'].flatten()
   from sklearn.svm import SVC
   model = SVC(kernel='linear', C=0.1)
   model.fit(X, Y);
[5]: # Lets find words, which are most common in spam.
   coef = model.coef_.flatten()
   sorted_ind = np.argsort(coef)
   top_15 = sorted_ind[:-16:-1]
   print('word
                       weight')
   print('*' * 18)
   for top in top_15:
       print('{:10} {:.3f}'.format(rev_vocab_dict[top], coef[top]))
   word
               weight
   ******
               0.501
   our
               0.466
   click
   remov
               0.423
   guarante
             0.384
              0.368
   visit
   basenumb
              0.345
   dollar
             0.324
   will
               0.270
               0.267
   price
   pleas
               0.261
   most
               0.257
   nbsp
               0.254
   10
               0.253
               0.248
   ga
               0.246
   hour
[6]: for i, name in enumerate(['emailSample1.txt', 'emailSample2.txt', 'spamSample1.
     →txt', 'spamSample2.txt']):
       features_arr = email_to_features(name, ps, vocab)
       print('Predicted value for {} {}, it should be {}'.format(name[:-4], model.
     →predict(features_arr)[0], i//2))
```

Predicted value for emailSample1 0, it should be 0 Predicted value for emailSample2 0, it should be 0

```
Predicted value for spamSample1 1, it should be 1 Predicted value for spamSample2 1, it should be 1
```

Prediction on test set = 98.9%