# **Machine Learning Lab 5**

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```
In [2]: import os
    import numpy as np
    import scipy.optimize as opt
    import scipy.io
    import matplotlib.pyplot as plt
    import matplotlib.image as mpimg
```

#### 1. read ex5data1.mat

```
In [3]: data = scipy.io.loadmat('../Data/lab5/ex5data1.mat')
    X = data['X']
    y = np.squeeze(data['y'])
    X.shape, y.size
Out[3]: ((51, 2), 51)
```

# 2. plot data

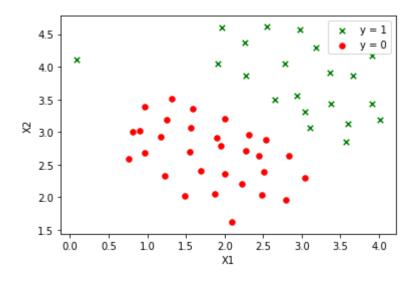
```
In [4]: def plot_data(X, y, axes=None):
    if axes == None:
        axes = plt.gca()

axes.scatter(X[y == 1, 0], X[y == 1, 1], marker='x', c='g', s=30, lataxes.scatter(X[y == 0, 0], X[y == 0, 1], c='r', s=30, label='y = 0')
    axes.set_xlabel('X1')
    axes.set_ylabel('X2')
    axes.legend(frameon= True, fancybox = True)

    return axes

plot_data(X, y)
```

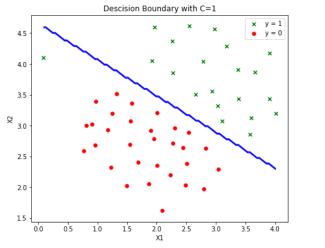
Out[4]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7f56878393c8>

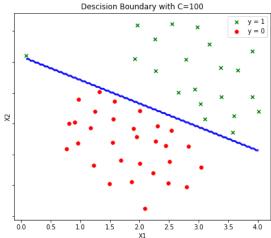


#### 3. sym classifier

# 4. descision boundary with c=1, c=100

verbose=0)





При C=100 можно наблюдать переобученую модель: граница захватывает случайные выбросы. При C=1 граница выглядит правильно.

# 5. svm gaussian kernel

```
In [22]: def gaussian_kernel(x1, x2, sigma):
    return np.exp(-np.sum((x1 - x2) ** 2) / (2 * (sigma ** 2)))

def gaussian_kernel_gram_matrix(X, L, sigma, K_function=gaussian_kernel)
    gram_matrix = np.zeros((X.shape[0], L.shape[0]))
    for i, x in enumerate(X):
        for j, l in enumerate(L):
            gram_matrix[i, j] = K_function(x, l, sigma)
    return gram_matrix
```

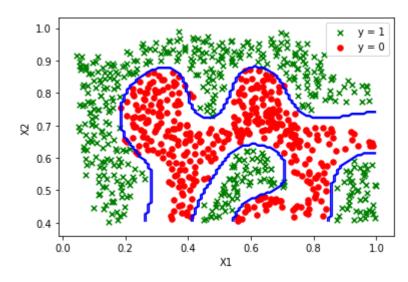
# 6-8. ex5data2.mat with svm gaussian kernel

# 9. plot data

```
In [30]: ax = plot_data(X, y)

X_1, X_2 = get_meshgrid(X)
X_plot = np.array([X_1.ravel(), X_2.ravel()]).T
gram_plot = gaussian_kernel_gram_matrix(X_plot, L, sigma)
Z = clf.predict(gram_plot).reshape(X_1.shape)
ax.contour(X_1, X_2, Z, 1, colors='b')
```

Out[30]: <matplotlib.contour.QuadContourSet at 0x102ef2910>



# 10. read ex5data3 data

```
In [33]: data = scipy.io.loadmat('../Data/lab5/ex5data3.mat')
    X = data['X']
    y = np.squeeze(data['y'])
    X_val = data['Xval']
    y_val = np.squeeze(data['yval'])

    X.shape, X_val.shape
Out[33]: ((211, 2), (200, 2))
```

### 11. cross validation

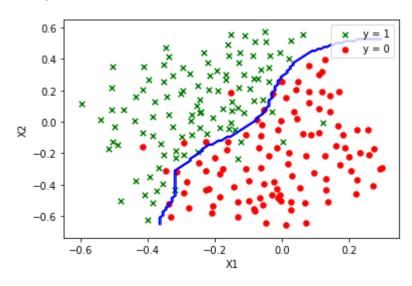
```
In [34]: def cross_validate(X, y, X_val, y_val):
             C array = np.array([0.01, 0.03, 0.1, 0.3, 1, 3, 10, 30])
             sigma array = np.array([0.01, 0.03, 0.1, 0.3, 1, 3, 10, 30])
             err array = np.zeros([C array.size, sigma array.size])
             for i in np.arange(C_array.size):
                  for j in np.arange(sigma array.size):
                      sigma = sigma array[j]
                      C = C \operatorname{array}[i]
                      gram = gaussian_kernel_gram_matrix(X, X, sigma)
                      clf = svm.SVC(C=C, kernel='precomputed')
                      clf.fit(gram, y)
                      predictions = clf.predict(gaussian kernel gram matrix(X val)
                      pred_error = np.mean(predictions != y_val)
                      err array[i, j] = pred error
             idx = np.unravel index(np.argmin(err array, axis=None), err array.s⊦
             C = C \operatorname{array}[idx[0]]
             sigma = sigma_array[idx[1]]
              return C, sigma
In [35]: C, sigma = cross validate(X, y, X val, y val)
         print(f'Found:\nC:\t{C}\nsigma:\t{sigma}')
         Found:
         C:
                  1.0
         sigma: 0.1
In [36]: gram = gaussian_kernel_gram_matrix(X, X, sigma)
         clf = svm.SVC(C=C, kernel='precomputed')
         clf.fit(gram, y)
Out[36]: SVC(C=1.0, cache size=200, class weight=None, coef0=0.0,
              decision_function_shape='ovr', degree=3, gamma='auto_deprecated',
              kernel='precomputed', max iter=-1, probability=False, random_state
         =None,
             shrinking=True, tol=0.001, verbose=False)
```

## 12. plot data

```
In [37]: ax = plot_data(X, y)

X_1, X_2 = get_meshgrid(X, num=50)
X_plot = np.array([X_1.ravel(), X_2.ravel()]).T
gram_plot = gaussian_kernel_gram_matrix(X_plot, X, sigma)
Z = clf.predict(gram_plot).reshape(X_1.shape)
ax.contour(X_1, X_2, Z, 1, colors='b')
```

Out[37]: <matplotlib.contour.QuadContourSet at 0x12903e7d0>



# 13-15. read spamTrain.mat & spamTest.mat data

```
In [38]: spam_train = scipy.io.loadmat('../Data/lab5/spamTrain.mat')
X = spam_train['X']
y = np.squeeze(spam_train['y'])

spam_test = scipy.io.loadmat('../Data/lab5/spamTest.mat')
X_test = spam_test['Xtest']
y_test = np.squeeze(spam_test['ytest'])
```

#### 14-16. svm + cross validation

```
In [39]: def spam_cross_validation(X, y, X_val, y_val):
             C array = np.array([0.01, 0.1, 0.3, 1, 10])
             err_array = np.zeros(C_array.size)
             for i in np.arange(C_array.size):
                 C = C_array[i]
                 clf = svm.SVC(C=C, kernel='linear')
                 clf.fit(X, y)
                 predictions = clf.predict(X val)
                 pred_error = np.mean(predictions != y_val)
                 err array[i] = pred error
             idx = np.unravel_index(np.argmin(err_array, axis=None), err_array.sf
             return C_array[idx[0]]
In [40]: C = spam_cross_validation(X, y, X_test, y_test)
         print(f'Found:\nC:\t{C}\n')
         Found:
                 0.1
         C:
In [42]:
         clf = svm.SVC(C=C, kernel='linear')
         clf.fit(X, y)
         train accuracy = clf.score(X, y) * 100
         test_accuracy = clf.score(X_test, y_test) * 100
         print(f'Train accuracy:\t{train_accuracy}%')
         print(f'Test accuracy:\t{test accuracy}%')
         Train accuracy: 99.825%
         Test accuracy: 98.9%
```

Линейное ядро быстрее и точнее чем Гаусовское

# 17. text preprocessing

```
In [44]:
          import re
          from nltk.stem import PorterStemmer
          ps = PorterStemmer()
          HTML REGEX = r'<.*?>'
          URL REGEX = r'[http|https]://[^\s]*'
          EMAIL REGEX = r'[^\s]+@[^\s]+'
          NUMBER REGEX = r'[0-9]+'
          DOLLAR REGEX = r'[\$]+'
          def preprocess data(data):
              result = data.lower()
              result = re.sub(HTML_REGEX, '', result)
              result = re.sub(URL_REGEX, 'httpaddr', result)
              result = re.sub(EMAIL REGEX, 'emailaddr', result)
              result = re.sub(NUMBER_REGEX, ' number ', result)
result = re.sub(DOLLAR_REGEX, ' dollar ', result)
              result = re.sub(r'[^a-zA-Z\s]+', '', result)
              result = result.replace('\n', ' ')
              result = [ps.stem(token) for token in result.split(' ')]
              result = ' '.join(result)
              return result
```

#### 18. read vocab.txt

```
In [45]: vocab_data = open('../Data/lab5/vocab.txt', 'r').read().split('\n')[:-1]
vocab = {}
for elem in vocab_data:
   index, word = elem.split('\t')[:]
   vocab[word] = index
```

#### 19. word to code

```
In [46]: def map_to_vocabulary(data, vocab):
    result = []
    for word in data.split():
        if len(word) > 1 and word in vocab:
            result.append(int(vocab[word]))

    return result
```

#### 20. text to feature vector

```
In [47]: def map to feature(data, vocab):
             n = len(vocab)
             features = np.zeros((n,))
             for i in data:
                 features[i] = 1
             return features
In [50]: def generate_feature(data, vocab):
             preprocessed = preprocess data(data)
             word indexes = map to vocabulary(preprocessed, vocab)
             feature = map to feature(word indexes, vocab)
             return feature
         21. test classifier
In [53]: def predict_from_files(files, vocab, clf):
             features = []
             for file in files:
                 feature = generate_feature(open(f'../Data/lab5/{file}', 'r').rea
                 features.append(feature)
             features = np.array(features)
             result = clf.predict(features)
             return zip(files, result)
         FILES = ['emailSample1.txt', 'emailSample2.txt', 'spamSample1.txt', 'spa
In [54]:
         predicts = predict from files(FILES, vocab, clf)
         for file, predict in predicts:
             res = 'spam' if predict == 1 else 'not spam'
             print(f'{file} - {res}')
         emailSample1.txt - not spam
         emailSample2.txt - not spam
         spamSample1.txt - spam
         spamSample2.txt - spam
```

22. get dataset from spamassassin.apache.org/old/publiccorpus (http://spamassassin.apache.org/old/publiccorpus/)

```
In [67]: HAM_DIR = '../Data/lab5/easy_ham'
         SPAM_DIR = '../Data/lab5/spam'
         def read data(data dir, out):
             files = os.listdir(data dir)
             for file_name in files:
                 if file name == '.DS Store':
                      continue
                 with open(f'{data_dir}/{file_name}', 'r') as f:
                          out.append(f.read())
                     except:
                          continue
             return out
         X_{data} = []
         X data = read data(HAM DIR, X data)
         y_ham = np.zeros(len(X_data))
         X data = read data(SPAM DIR, X data)
         y_spam = np.ones(len(X_data) - len(y_ham))
         y = np.concatenate((y_ham, y_spam))
         len(X_data), y.size
Out[67]: (2827, 2827)
```

24. build vocabulary

```
In [68]:
         from sklearn.feature_extraction.text import CountVectorizer
         from sklearn.model_selection import train_test_split
         X \text{ vocab} = []
         for data in X_data[100:]:
             X_vocab.append(preprocess_data(data))
         vectorizer = CountVectorizer()
         vectorizer.fit(X vocab)
         n = 1000
         vocab = \{\}
         index = 0
         for word in vectorizer.vocabulary :
             vocab[word] = index
             index += 1
             if index >= n:
                  break
```

#### 25. test new dataset

```
In [69]: X = []
         for data in X data:
             feature = generate feature(data, vocab)
             X.append(feature)
         X = np.array(X)
In [70]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
In [71]: C = 0.1
         clf = svm.SVC(C=C, kernel='linear')
         clf.fit(X_train, y_train)
         train accuracy = clf.score(X, y) * 100
         test_accuracy = clf.score(X_test, y_test) * 100
         print(f'Train accuracy:\t{train accuracy}%')
         print(f'Test accuracy:\t{test_accuracy}%')
         Train accuracy: 100.0%
         Test accuracy: 100.0%
In [73]:
         FILES = ['emailSample1.txt', 'emailSample2.txt', 'spamSample1.txt', 'spa
         predicts = predict_from_files(FILES, vocab, clf)
         for file, predict in predicts:
             res = 'spam' if predict == 1 else 'not spam'
             print(f'{file} - {res}')
         emailSample1.txt - not spam
         emailSample2.txt - not spam
         spamSample1.txt - spam
         spamSample2.txt - spam
```

In [ ]: