```
In [92]: import numpy as np
         import random
         import matplotlib.pyplot as plt
         import sklearn.svm as svm
         from sklearn.model selection import KFold
In [52]: #digit intensity symmetry
         train = {}
         for digit in range(10):
             train[digit] = []
         with open('features.train','r') as f:
             for line in f:
                 line = [float(i) for i in line.split()]
                 train[line[0]].append(line[1:])
         for digit in range(10):
             train[digit] = np.asarray(train[digit])
In [53]: | print(train[1].shape)
         (1005, 2)
In [54]: | test = {}
         for digit in range(10):
             test[digit] = []
         with open('features.test','r') as f:
             for line in f:
                 line = [float(i) for i in line.split()]
                 test[line[0]].append(line[1:])
         for digit in range(10):
             test[digit] = np.asarray(test[digit])
In [62]: print([test[i].shape for i in range(10)])
         [(359, 2), (264, 2), (198, 2), (166, 2), (200, 2), (160, 2), (170, 2),
         (147, 2), (166, 2), (177, 2)
In [55]: def get misclassified point idx(a,b):
             idx = []
             for i in range(len(a)):
                 if a[i] != b[i]:
                      idx.append(i)
             return idx
```

```
In [118]:
          def soft margin_svm(train_data,train_class,test_data,test_class,C,kernel
          ,degree=None):
              if kernel is 'poly':
                  clf = svm.SVC(C = C, kernel = kernel, degree=degree, gamma=1,coe
          f0 = 0)
              else:
                  clf = svm.SVC(C = C, kernel = kernel,gamma=1)
              clf.fit(train data, train class)
              train_results = clf.predict(train_data)
              idx = get_misclassified_point_idx(train_class,train_results)
              e in = len(idx)/float(len(train class))
              test results = clf.predict(test data)
              idx = get_misclassified_point_idx(test_class,test_results)
              e out = len(idx)/float(len(test class))
              n_support_vecs = len(clf.support_)
              return e_in,e_out, n_support_vecs
```

```
In [119]: def create train_test_data_one_v_all(digit,train,test):
              train_data = []
              train class = []
              test_data = []
              test_class = []
              for i in range(10):
                  if i == digit:
                       class id = -1
                  else:
                      class id = 1
                  train data.extend(train[i])
                  train class.extend([class id for j in train[i]])
                  test data.extend(test[i])
                  test class.extend([class id for j in test[i]])
              return np.asarray(train data), np.asarray(train class), np.asarray(te
          st data), np.asarray(test class)
```

```
In [120]: def create_train_test_data_one_v_one(digit_1,digit_2,train,test):
              train_data = []
              train_class = []
              test_data = []
              test_class = []
              for i in range(10):
                  if i == digit 1:
                      class_id = -1
                  elif i == digit_2:
                      class_id = 1
                  else:
                       continue
                  train_data.extend(train[i])
                  train_class.extend([class_id for j in train[i]])
                  test_data.extend(test[i])
                  test_class.extend([class_id for j in test[i]])
              return np.asarray(train_data),np.asarray(train_class), np.asarray(te
          st_data), np.asarray(test_class)
```

```
In [121]: #x vs all experiments for each digit x
          C = 0.01
          kernel = 'poly'
          degree = 2
          e_{ins} = []
          e_outs = []
          support_vecs = []
          for digit in range(10):
              print('Digit: '+str(digit))
              train_data,train_class,test_data,test_class = create_train_test_data
          _one_v_all(digit,train,test)
              e_in, e_out, n_support_vecs = soft_margin_svm(train_data,train_clas
          s,test_data,test_class,C,kernel,degree=degree)
              e ins.append(e in)
              e_outs.append(e_out)
              support_vecs.append(n_support_vecs)
              print('E_in: '+str(e_in))
              print('E_out: '+str(e_out))
              print('Support vectors: '+str(n_support_vecs))
```

Digit: 0 E in: 0.119050884652 E\_out: 0.128550074738 Support vectors: 2278 Digit: 1 E\_in: 0.0142641612947 E\_out: 0.0224215246637 Support vectors: 400 Digit: 2 E\_in: 0.100260595254 E\_out: 0.0986547085202 Support vectors: 1464 Digit: 3 E in: 0.0902482512687 E\_out: 0.0827105132038 Support vectors: 1320 Digit: 4 E in: 0.0894253188863 E out: 0.0996512207275 Support vectors: 1307 Digit: 5 E\_in: 0.0762584007681 E\_out: 0.079720976582 Support vectors: 1117 Digit: 6 E\_in: 0.0910711836511 E\_out: 0.0847035376183 Support vectors: 1330 Digit: 7 E in: 0.0884652311068 E out: 0.0732436472347 Support vectors: 1293 Digit: 8 E in: 0.0743382252092 E\_out: 0.0827105132038 Support vectors: 1091 Digit: 9 E in: 0.0883280757098

E\_out: 0.0881913303438
Support vectors: 1290

```
In [122]:
          #1 vs 5 experiments
          kernel = 'poly'
          degree = 2
          digit_1 = 1
          digit 2 = 5
          train_data,train_class,test_data,test_class = create_train_test_data_one
          _v_one(digit_1,digit_2, train,test)
          for C in [.0001,.001,.01,.1,1]:
              for degree in [2,5]:
                  print('C: '+str(C)+', Q: '+str(degree))
                  e in, e out, n support vecs = soft margin svm(train data,train
          class,test_data,test_class,C,kernel,degree=degree)
                  print('E_in: '+str(e_in))
                  print('E_out: '+str(e_out))
                  print('Support vectors: '+str(n_support_vecs))
          C: 0.0001, Q: 2
          E in: 0.0102498398463
          E out: 0.0165094339623
          Support vectors: 244
          C: 0.0001, Q: 5
          E_in: 0.00448430493274
          E out: 0.0165094339623
          Support vectors: 26
          C: 0.001, Q: 2
          E_in: 0.00448430493274
          E out: 0.0165094339623
          Support vectors: 80
          C: 0.001, Q: 5
          E in: 0.00448430493274
          E out: 0.0165094339623
          Support vectors: 26
          C: 0.01, Q: 2
          E in: 0.00448430493274
          E out: 0.0188679245283
          Support vectors: 34
          C: 0.01, Q: 5
          E in: 0.00512491992313
          E out: 0.0165094339623
          Support vectors: 27
          C: 0.1, Q: 2
          E in: 0.00448430493274
          E out: 0.0188679245283
          Support vectors: 24
          C: 0.1, Q: 5
          E in: 0.00448430493274
          E out: 0.0188679245283
          Support vectors: 24
          C: 1, Q: 2
          E in: 0.00384368994234
          E out: 0.0188679245283
          Support vectors: 24
          C: 1, Q: 5
          E in: 0.00448430493274
          E out: 0.0165094339623
```

Support vectors: 24

```
In [124]:
          #Cross validation in 1 vs. 5
          kernel = 'poly'
          degree = 2
          digit_1 = 1
          digit 2 = 5
          cv train data,cv train class,cv test data,cv test class = create train t
          est_data_one_v_one(digit_1,digit_2, train,test)
          kf = KFold(n splits=10, shuffle=True)
          kf.get_n_splits(train_data)
          all_errors = []
          selections = []
          Cs = [0.0001, 0.001, 0.01, 0.1, 1]
          for i in range(100):
              c errors = []
              for C in Cs:
                  done = False
                   errors = []
                  while len(errors) < 10:</pre>
                       errors = []
                       for train index, test index in kf.split(cv train data):
                           train data = cv train data[train index]
                           train_class = cv_train_class[train_index]
                           if len(list(set(train_class))) < 2:</pre>
                               continue
                           test data = cv train data[test index]
                           test_class = cv_train_class[test_index]
                           e_in, e_out, n_support_vecs = soft margin svm(train dat
          a,train class,test data,test class,C,kernel,degree=degree)
                           errors.append(e out)
                   c errors.append(np.mean(errors))
              selections.append(np.argmin(c errors))
              all errors.append(c errors)
          selected idx = int(np.median(selections))
          print('Selected C: '+str(Cs[selected idx]))
          print('Average error: '+str(np.mean([i[selected idx] for i in all errors
          1)))
          print(list(set(selections)))
```

```
Selected C: 0.01
Average error: 0.00461273068757
[1, 2, 3, 4]
```

```
In [125]: #1 vs 5 experiments
          kernel = 'rbf'
          train data, train class, test data, test class = create train test data one
           _v_one(digit_1,digit_2, train,test)
          for C in [.01,1,100,10**4, 10**6]:
              print('C: '+str(C))
              e_in, e_out, n_support_vecs = soft_margin_svm(train_data,train clas
          s, test data, test class, C, kernel)
              print('E_in: '+str(e_in))
              print('E_out: '+str(e_out))
              print('Support vectors: '+str(n_support_vecs))
          C: 0.01
          E in: 0.00384368994234
          E_out: 0.0235849056604
          Support vectors: 403
          C: 1
          E in: 0.00448430493274
          E out: 0.0212264150943
          Support vectors: 31
          C: 100
          E in: 0.00320307495195
          E out: 0.0188679245283
          Support vectors: 22
          C: 10000
          E in: 0.00256245996156
          E_out: 0.0235849056604
          Support vectors: 20
          C: 1000000
          E in: 0.000640614990391
          E out: 0.0235849056604
          Support vectors: 17
```

In [ ]: