Problem 1

d is the correct answer.
We have this because we can vary $\mathbf{w}$ and b. And since $\mathbf{w}$ has d variables, we have $d+1$ variables in total.

Problem 2

a is the correct answer.
I wrote some code to implement SVM with soft margin on the zip-code data by solving the equations laid out in the problem set. I used binary classification error. To help me write this code, I used the libsvm library. For this problem, I made sure to use the polynomial kernel as given in the problem. After running this code for 0 versus all, 2 versus all, 4 versus all, 6 versus all, and 8 versus all, I found that 0 versus all gave me the lowest accurary and thus the highest E_{in} (ran prediction code on the training data).

Problem 3

a is the correct answer.
I wrote some code to implement SVM with soft margin on the zip-code data by solving the equations laid out in the problem set. I used binary classification error. To help me write this code, I used the libsvm library. For this problem, I made sure to use the polynomial kernel as given in the problem. After running this code for 1 versus all, 3 versus all, 5 versus all, 6 versus all, and 9 versus all, I found that 1 versus all gave me the highest accurary and thus the lowest E_{in} (ran prediction code on the training data).

Problem 4

c is the correct answer.
Using the code I wrote for the above two problems, I got 386 support vectors for 1 versus all and 2180 support vectors for 0 versus all (with all the other parameters as specified in the problem). The difference between these two numbers is around 1800.

Problem 5

d is the correct answer.
I wrote some code that runs the 1 versus 5 classifier with $Q = 2$ and $C \in \{0.001, 0.01, 0.1, 1\}$. I then observed the number of support vectors, E_{in}, and E_{out} for each C value. I found that the only statement given in the problem that is true is that the maximum C (1) achieves the lowest E_{in} and the highest accuracy (around 99.68% accuracy). I found that all the other statements given in the problem are false.

Problem 6

b is the correct answer.
I wrote some code that runs the 1 versus 5 classifier with $Q = 2$ and $C \in \{0.0001, 0.001, 0.01, 0.1, 1\}$ and then runs the 1 versus 5 classifier with $Q = 5$ and $C \in \{0.0001, 0.001, 0.01, 0.1, 1\}$. I did this so I could compare values between all the runs. I found that the only statement given in the problem that is true is that when $C = 0.001$, the number of support vectors is lower at $Q = 5$ (76 versus 25).

Problem 7

b is the correct answer.
I wrote some code that runs 10-fold cross validation for the polynomial kernel. With this code, I considered the 1 versus 5 classifier with $Q = 2$. I used E_{cv} to select $C \in \{0.0001, 0.001, 0.01, 0.1, 1\}$. If there was a tie in E_{cv}, I selected the smaller C. I then ran this code for 100 random runs (so I tried 100 different partitions). With this, I discovered that $C = 0.001$ is selected most often.

Problem 8

c is the correct answer.
I basically used the same code I wrote for the above problem, and for each run, I added the value of E_{cv} for $C = 0.001$. I then took the average value of this list, and it was around 0.005.

Problem 9

e is the correct answer.
I wrote code that runs SVM with the RBF kernel for all the values of C given in the problem. I ran this code on the 1 versus 5 classifier. To help me write this code, I used the libsvm library. The value $C = 10^6$ gave me the lowest E_{in} value (ran prediction code on the training data).

Problem 10

c is the correct answer.
I wrote code that runs SVM with the RBF kernel for all the values of C given in the problem. I ran this code on the 1 versus 5 classifier. To help me write this code, I used the libsvm library. The value $C = 100$ gave me the lowest E_{out} value (ran prediction code on the test data).

Code

```
from __future__ import division
from svm import *
from svmutil import *
from collections import Counter
import sys
import random
import numpy as np

class MySvm:
    def __init__(self, file_train, file_test):
        self.training_labels = self.load_labels(file_train)
        self.training_data = self.load_data(file_train)
        self.training_data_curr = self.training_data
        self.scores_tr = []

        self.test_labels = self.load_labels(file_test)
        self.test_data = self.load_data(file_test)
        self.test_data_curr = self.test_data
        self.scores_test = []

        self.cross_data_curr_list = []
        self.scores_cross_list = []
        self.scores_tr_list = []
        self.training_data_curr_list = []

        self.model = None
        self.model_list = []
        self.num_support_vectors = 0

    def load_labels(self, file_name):
        file_obj = open(file_name)
        labels = []
        for line in file_obj:
            labels.append(float(line.split()[0]))
        return labels

    def load_data(self, file_name):
        file_obj = open(file_name)
        data = []
        for line in file_obj:
            item = []
            line_split = line.split()
            for i in range(1, len(line_split)):
                item.append(float(line_split[i]))
            data.append(item)
        return data

    def one_versus_all(self, num_one, kernel_type, error_const, poly_degree,
            is_cross, is_repeat = False):
        if not is_repeat:
            self.scores_tr, self.training_data_curr = self.get_one_versus_all_lists(
                    num_one, self.training_labels, self.training_data)
            self.scores_test, self.test_data_curr = self.get_one_versus_all_lists(
                    num_one, self.test_labels, self.test_data)
            if is_cross:
                self.scores_tr, self.training_data_curr, self.scores_cross, self.cross_data

        prob = svm_problem(self.scores_tr, self.training_data_curr)
        param_str = '-t %d -r 1 -g 1 -c %f -d %d' %(kernel_type, error_const, poly_degree)
        param = svm_parameter(param_str)
        model = svm_train(prob, param)
```

```
62          self.model = model
63          self.num_support_vectors = len(model.get_SV())
64
65      def one_versus_one(self, num_one, num_other, kernel_type, error_const,
66              poly_degree, is_cross, is_repeat = False):
67          if not is_repeat:
68              self.scores_tr, self.training_data_curr = self.get_one_versus_one_lists(
69                      num_one, num_other, self.training_labels, self.training_data)
70              self.scores_test, self.test_data_curr = self.get_one_versus_one_lists(
71                      num_one, num_other, self.test_labels, self.test_data)
72              if is_cross:
73                  self.scores_tr_list, self.training_data_curr_list, self.scores_cross_list,
74
75          if not is_cross:
76              prob = svm_problem(self.scores_tr, self.training_data_curr)
77              param_str = '-t %d -r 1 -g 1 -c %f -d %d' %(kernel_type, error_const, poly_degr
78              param = svm_parameter(param_str)
79              model = svm_train(prob, param)
80              self.model = model
81              self.num_support_vectors = len(model.get_SV())
82          else:
83              self.model_list = []
84              sum_num_support_vectors = 0
85              for i in range(0, len(self.scores_tr_list)):
86                  prob = svm_problem(self.scores_tr_list[i], self.training_data_curr_list[i])
87                  param_str = '-t %d -r 1 -g 1 -c %f -d %d' %(kernel_type, error_const, poly_
88                  param = svm_parameter(param_str)
89                  model = svm_train(prob, param)
90                  self.model_list.append(model)
91                  sum_num_support_vectors += len(model.get_SV())
92              self.num_support_vectors = sum_num_support_vectors / len(self.scores_tr_list)
93
94      def get_one_versus_all_lists(self, num_one, labels, data_items):
95          scores = []
96          for label in labels:
97              if label == num_one:
98                  scores.append(1)
99              else:
100                 scores.append(-1)
101         return (scores, data_items)
102
103     def get_one_versus_one_lists(self, num_one, num_other, labels, data_items):
104         scores = []
105         data_curr = []
106         for i in range(0, len(labels)):
107             label = labels[i]
108             item = data_items[i]
109             if label == num_one:
110                 scores.append(1)
111                 data_curr.append(item)
112             elif label == num_other:
113                 scores.append(-1)
114                 data_curr.append(item)
115         return (scores, data_curr)
116
117     def get_cross_val_lists(self):
118         z_list = list(zip(self.scores_tr, self.training_data_curr))
119         random.shuffle(z_list)
120         shuffled_scores_tr, shuffled_training_data_curr = zip(*z_list)
121         index = int(len(self.scores_tr) / 10)
122
123         scores_tr_list, training_data_curr_list, scores_cross_list, cross_data_curr_list =
124         start = 0
```

```
125         end = index
126         while end < len(self.scores_tr):
127             if end + index >= len(self.scores_tr):
128                 end = len(self.scores_tr)
129             scores_cross_list.append(list(shuffled_scores_tr[start:end]))
130             cross_data_curr_list.append(list(shuffled_training_data_curr[start:end]))
131             scores_tr_list.append(list(shuffled_scores_tr[0:start]) + list(shuffled_scores_
132             training_data_curr_list.append(list(shuffled_training_data_curr[0:start]) + lis
133             start += index
134             end += index
135         return (scores_tr_list, training_data_curr_list, scores_cross_list, cross_data_curr
136 
137     def get_error_in(self):
138         p_labels, p_acc, p_vals = svm_predict(self.scores_tr,
139                 self.training_data_curr, self.model)
140         return p_acc[0]
141 
142     def get_error_out(self):
143         p_labels, p_acc, p_vals = svm_predict(self.scores_test,
144                 self.test_data_curr, self.model)
145         return p_acc[0]
146 
147     def get_error_cv(self):
148         sum_acc = 0
149         for i in range(0, len(self.scores_cross_list)):
150             p_labels, p_acc, p_vals = svm_predict(self.scores_cross_list[i],
151                     self.cross_data_curr_list[i], self.model_list[i])
152             sum_acc += p_acc[0]
153         print '========LENGTH = ', len(self.scores_cross_list)
154         return sum_acc / len(self.scores_cross_list)
155 
156 if __name__ == '__main__':
157     prob1_3 = False
158     prob5 = False
159     prob6 = False
160     prob7_8 = True
161     prob9_10 = False
162     print 'Kshit is gay'
163     my_svm = MySvm('features.train', 'features.test')
164 
165     if prob1_3:
166         for i in range(1, 11, 2):
167             print 'i = ', i
168             my_svm.one_versus_all(i, 1, .01, 2, False)
169             p_acc = my_svm.get_error_in()
170             print 'Num support vectors = ', my_svm.num_support_vectors
171             print '\n\n'
172 
173     if prob5:
174         print 'Q = ', 2
175         for c in [.0001, .001, .01, .1, 1]:
176             print 'C = ', c
177             my_svm.one_versus_one(1, 5, 1, c, 2, False)
178             print 'E in'
179             my_svm.get_error_in()
180             print 'E out'
181             my_svm.get_error_out()
182             print 'Num support vectors = ', my_svm.num_support_vectors
183             print '\n'
184         print '\n\n\n'
185 
186     if prob6:
187         print 'Q = ', 5
```

```
188         for c in [.0001, .001, .01, .1, 1]:
189             print 'C = ', c
190             my_svm.one_versus_one(1, 5, 1, c, 5, False)
191             print 'E in'
192             my_svm.get_error_in()
193             print 'E out'
194             my_svm.get_error_out()
195             print 'Num support vectors = ', my_svm.num_support_vectors
196             print '\n'
197 
198     if prob7_8:
199         c_list = []
200         e_dict = {.0001 : [], .001 : [], .01 : [], .1 : [], 1 : []}
201         for i in range(0, 100):
202             min_error = sys.maxint
203             c_val = 1
204             repeat = False
205             for c in [.0001, .001, .01, .1, 1]:
206                 my_svm.one_versus_one(1, 5, 1, c, 2, True, repeat)
207                 repeat = True
208                 # Get error, NOT accurary
209                 error = (100 - my_svm.get_error_cv()) / 100
210                 e_dict[c].append(error)
211                 if error < min_error:
212                     min_error = error
213                     c_val = c
214             c_list.append(c_val)
215         c_data = Counter(c_list)
216         print 'Counts = ', c_data.most_common()
217         print 'Mode = ', c_data.most_common(1)
218         for key, val in e_dict.iteritems():
219             new_val = []
220             val_max = max(val)
221             val_min = min(val)
222             val_mean = sum(val) / len(val)
223             new_val = [val_mean, val_max, val_min]
224             e_dict[key] = new_val
225         print e_dict
226 
227     if prob9_10:
228         min_error_in = sys.maxint
229         min_error_out = sys.maxint
230         c_val_in = 0
231         c_val_out = 0
232         for c in [.01, 1, 100, 10000, 1000000]:
233             my_svm.one_versus_one(1, 5, 2, c, 2, False, False)
234             error_in = (100 - my_svm.get_error_in()) / 100
235             if error_in < min_error_in:
236                 min_error_in = error_in
237                 c_val_in = c
238             error_out = (100 - my_svm.get_error_out()) / 100
239             if error_out < min_error_out:
240                 min_error_out = error_out
241                 c_val_out = c
242         print 'Min error in = ', min_error_in
243         print 'C value in = ', c_val_in
244         print 'Min error out = ', min_error_out
245         print 'C value out = ', c_val_out
```