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In [261]: import pandas as pd import numpy as np import matplotlib.pyplot as plt from cvxopt import matrix, solvers %matplotlib inline # Load the wines data into a pandas dataframe fname = r'wines.csv' wines = pd.read_csv(fname,delimiter=';',header=0) wines.head()

Out[261]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	1
0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	ξ
1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	0.68	ć
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	ί
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	ί
4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	Ę

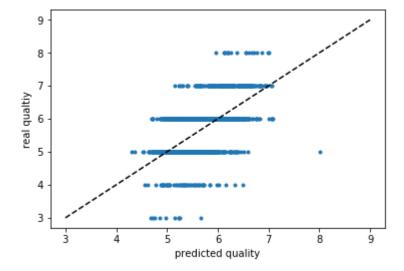
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```
In [281]: # dimension of the data
          n = wines.shape[0]
          m = wines.shape[1]-1
          # augment the X and parse Y
          min max scaler = preprocessing.MinMaxScaler()
          X = np.hstack((wines.iloc[:,:-1],np.ones((n,1))))
          Y = wines.iloc[:,-1].astype('double')
          # c is (m+1+n)
          # setting the coefficients in c corresponding to a j and b to zeros and averag
          e the z's
          c = matrix(np.hstack([np.zeros(m+1),np.ones(n)/n]))
          # A is 2n by (m+1+n)
          XX = np.vstack([-X,X])
          I = np.vstack([-np.eye(n),-np.eye(n)])
          A = np.hstack([XX,I])
          # we need to format this for the solver to include the constraints on z's
          G = matrix(np.vstack([A,np.hstack([np.zeros((n,m+1)),-np.eye(n)])]))
          # b is 2n
          b = np.hstack([-Y,Y])
          # again we format the b to include constraints on the z's
          h = matrix(np.hstack([b,np.zeros(n)]))
          # pass the matrices to the solver
          sol = solvers.lp(c, G, h,solver='splk')
```

```
k/t
               dcost
                                 pres
                                       dres
    pcost
                          gap
    0.0000e+00 6.9562e-16 4e+00
                                8e-01 8e-15
                                             1e+00
0:
1:
    2.7051e-01 1.4091e-01 1e+00
                                2e-01
                                      5e-15 9e-02
2:
    4.0476e-01 3.6896e-01 2e-01 4e-02
                                       7e-13
                                             1e-02
3: 4.6439e-01 4.5069e-01 8e-02 1e-02 4e-12 4e-03
4: 4.8118e-01 4.7506e-01 3e-02 6e-03
                                       2e-11
                                             1e-03
5: 4.8776e-01 4.8474e-01 2e-02 3e-03
                                      5e-12 4e-04
6: 4.9221e-01 4.9144e-01 4e-03 7e-04 1e-11 8e-05
7: 4.9331e-01 4.9310e-01 1e-03
                                2e-04 3e-12 1e-05
8: 4.9357e-01 4.9349e-01 4e-04 7e-05 3e-11 4e-06
9: 4.9369e-01 4.9367e-01 1e-04
                                      2e-11 9e-08
                                2e-05
10: 4.9374e-01 4.9373e-01 3e-05
                                5e-06
                                      3e-11 2e-08
    4.9375e-01 4.9375e-01 1e-06
                                       8e-12 7e-10
11:
                                2e-07
12:
    4.9375e-01 4.9375e-01 1e-08 2e-09 3e-11 7e-12
Optimal solution found.
```

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```
In [283]: d = np.asarray(sol['x'][:12])
   plt.plot(X.dot(d),Y,'.')
   plt.plot([3,9],[3,9],'k--')
   plt.xlabel('predicted quality')
   plt.ylabel('real qualtiy')
   plt.show()
```



```
In [287]: print('The average L1 loss is:')
    sol['primal objective']
```

Out[287]: 0.4937487285454359