lab1-part2-ml

November 10, 2015

0.1 Part2 - Logistic Regression

```
In [1]: #notebook settings
        %matplotlib inline
        #import some useful libraries and utilities
        import sys
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.cross_validation import KFold
        #avoiding some warnings
        np.seterr(all='ignore')
        #data directory
        path = './credit/'
        #parameters to try
        params = np.array([0.2e-3, 0.4e-3, 0.6e-3, 0.8e-3, 1.0e-3])
In [2]: #helper functions
        def rescale(M, a=0., b=1.):
            """ Rescale features of M to [a,b] range """
            #max and min vectors
           maxv = np.max(M, axis=0)
           minv = np.min(M, axis=0)
            return (b-a)*M/(maxv-minv) + (a*maxv-b*minv)/(maxv-minv)
        def normalize(M):
            #mean and standard deviation vectors
           meanv = np.mean(M, axis=0)
            stdv = np.std(M, axis=0)
            return (M-meanv)/stdv
In [3]: """ Find the best learning parameter for algorithm, between
        parameters in params using k-fold cross validation """
        def cross_alpha(X, y, algorithm, params):
            #creating kfold
           m,n = X.shape
           kf = KFold(m, n_folds=5)
            tr_err = np.empty((5,5))
            ts_err = np.empty((5,5))
            i = 0 #index of fold
```

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for tr_index,ts_index in kf:
        j = 0 #index of parameter
        for param in params:
            X_train, X_test = X[tr_index], X[ts_index]
            y_train, y_test = y[tr_index], y[ts_index]
            beta,_ = algorithm(X_train, y_train, param)
            tr_err[i,j] = error_rate(X_train, y_train, beta)
            ts_err[i,j] = error_rate(X_test, y_test, beta)
            j += 1
        i += 1
    #arrays with mean costs for each alpha
    ts_mean = np.mean(ts_err, axis=0)
    return params[np.argmin(ts_mean)], tr_err, ts_err
11 11 11
Function to generate histogram of winners (alphas)
def make_hist(winners,params):
   winners = np.array(winners)
   freqs = np.zeros(5)
    for i in xrange(5):
        freqs[i] = np.sum(params[i]==winners)
    labels = ['alp1', 'alp2', 'alp3', 'alp4', 'alp5']
   pos = np.arange(len(labels))
   width = 1.0
    ax = plt.axes()
    ax.set_xticks(pos + (width / 2))
    ax.set_xticklabels(labels)
   plt.bar(pos, freqs, width, color='r')
   plt.title('Best alpha for each dataset')
   plt.show()
def solve_logistic(algorithm, params, data_func=None):
    winners = list()
    for i in xrange(20):
        #Loading dataset
        tr_file = path+'credit-tr-{0}.npy'.format(i)
        ts_file = path+'credit-ts-{0}.npy'.format(i)
        tr_data = np.load(tr_file)
        ts_data = np.load(ts_file)
        if data_func is not None:
            X_tr = data_func(tr_data[:,:-1])
        else:
            X_tr = tr_data[:,:-1]
        y_tr = np.ascontiguousarray(tr_data[:,-1])
        #Adding column of 1's
        m,n = X_t.shape
        X_tr = np.concatenate((np.ones((m,1)),X_tr),axis=1)
        if data_func is not None:
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X_ts = data_func(ts_data[:,:-1])
              else:
                 X_{ts} = ts_{data}[:,:-1]
              y_ts = np.ascontiguousarray(ts_data[:,-1])
              #Adding column of 1's
              m,n = X_ts.shape
              X_ts = np.concatenate((np.ones((m,1)),X_ts),axis=1)
              alpha,tr_err,ts_err = cross_alpha(X_tr, y_tr, algorithm, params)
              winners.append(alpha)
              beta,it = algorithm(X_tr, y_tr, alpha)
              print "Dataset: {0}".format(i)
              print 'Best alpha: {0}'.format(alpha)
              fig = plt.figure()
              fig.set_figheight(7)
              fig.set_figwidth(13)
              fig.add_subplot(121)
              plt.boxplot(tr_err, showmeans=True, meanline=True)
              fig.add_subplot(122)
              plt.boxplot(ts_err, showmeans=True, meanline=True)
              print 'Training error rate: {0}'.format(error_rate(X_tr,y_tr,beta))
              print 'Testing error rate: {0}'.format(error_rate(X_ts,y_ts,beta))
              print 'N° iterations: {0}'.format(it)
              print 'Beta: {0}'.format(beta)
              print '\n'
          make_hist(winners,params)
0.1.1 Mean functions
        def sigmoid(z):
            return 1./(1.+np.exp(-z))
```

```
In [140]: #Sigmoid function
          #Overall cost function for logistic regression (log likelihood function)
          def 1(X, y, beta):
              y1_mask = y.astype(bool)
              y0_mask = np.logical_not(y1_mask)
              f = sigmoid(np.dot(X,beta))
              return (np.log(f[y1_mask])).sum() + (np.log(1-f[y0_mask])).sum()
          #Error rate for logistic regression
          def error_rate(X, y, beta):
              h = np.round(sigmoid(np.dot(X,beta)))
              h = h.astype(int)
              y = y.astype(int)
              m_{\bullet} = h.shape
              return np.logical_xor(h,y).sum()/np.float(m)
          #Stochastic gradient ascent for logistic regression
```

```
M,N = X.shape
              beta = np.zeros(N)
              11 = 1(X, y, beta)
              for i in xrange(max_iter):
                  10 = 11
                  for m in xrange(M):
                      beta += alpha*(y[m]-sigmoid(np.dot(X[m],beta)))*X[m]
                  11 = 1(X,y,beta)
                  if np.abs(11) < eps: break</pre>
              return (beta,i+1)
          #Newton-Raphson method for logistic regression
          def nr_logistic(X, y, eps=1e-5, max_iter=100000):
              M,N = X.shape
              beta = np.zeros(N)
              11 = 1(X, y, beta)
              for i in xrange(max_iter):
                  10 = 11
                  f = sigmoid(np.dot(X,beta))
                  W = np.diag(f*(1-f))
                  Hess = -1*np.dot(X.T, np.dot(W, X))
                  Dl = np.dot(X.T, y-f)
                  #when it converges, Hess became singular
                  try:
                      beta -= np.linalg.solve(Hess, D1)
                  except np.linalg.LinAlgError:
                      break
                  11 = 1(X, y, beta)
                  if np.abs(11) < eps: break</pre>
              return (beta, i+1)
In [135]: #working array
          wa = np.load(path+'credit-tr-1.npy')
          \#X = wa[:,:-1]
          X = rescale(wa[:,:-1], 0., 1.)
          \#X = normalize(wa[:,:-1])
          m,n = X.shape
          X = np.concatenate((np.ones((m,1)),X),axis=1)
          y = np.ascontiguousarray(wa[:,-1])
In [146]: beta1,it = nr_logistic(X, y, 1.0)
          print 'Beta:', beta1
          print 'N° Iter:', it
          print 'Error rate:', error_rate(X,y,beta1)
Beta: [ 386.22218941 -134.63700429 -1311.72058215 -264.81627176 2719.35544532
  -839.27080576 47.22642184]
N° Iter: 21
Error rate: 0.0
In [139]: beta2,it = gd_stochastic(X, y, 10)
          print 'Beta:', beta2
          print 'N° iter:', it
          print 'Error rate:', error_rate(X,y,beta2)
```

def gd_stochastic(X, y, alpha, eps=1e-5, max_iter=100000):

Beta: [219.09024529 -71.63364477 -789.74696886 -159.88784422 1690.73437923

-518.06907211 26.54939681]

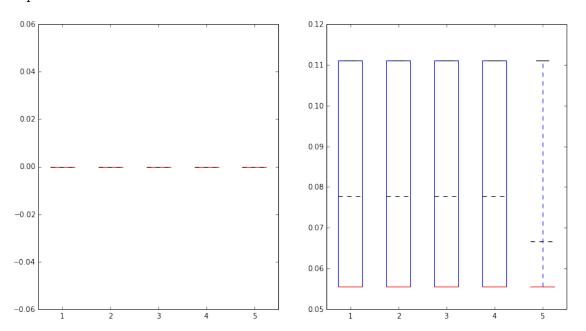
 N° iter: 100000 Error rate: 0.0

0.2 1)

0.2.1 Newton-Raphson with raw data

In [148]: solve_logistic(nr_logistic, params1)

Dataset: 0
Best alpha: 1.0



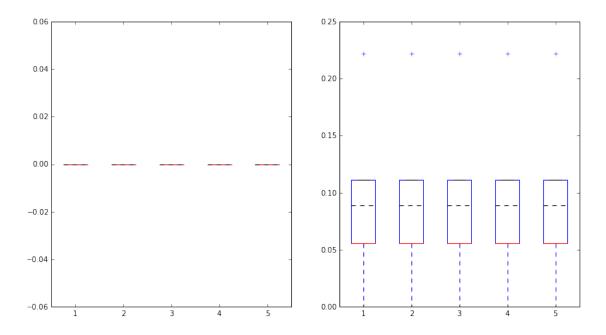
Training error rate: 0.0

Testing error rate: 0.0666666666667

 N° iterations: 21

-857.53711237 1.58287241]

Dataset: 1
Best alpha: 0.2

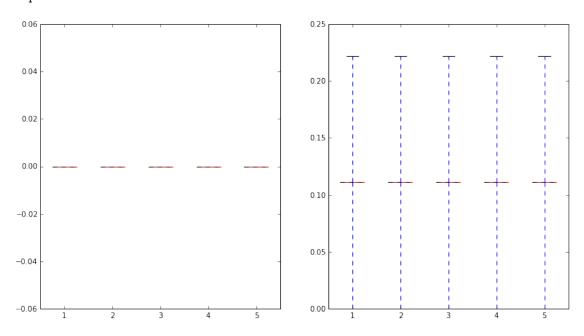


Testing error rate: 0.0666666666667

 N° iterations: 98

-829.36038411 2.05277711]

Dataset: 2
Best alpha: 0.2



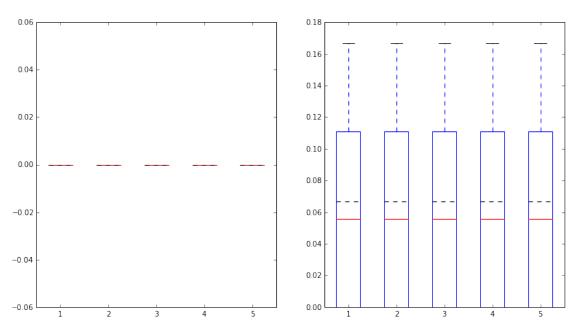
Testing error rate: 0.133333333333

 N° iterations: 92

Beta: [432.7528628 -1.18396963 -9.87466609 -0.6545947 101.62283621

-297.01969987 -3.37241181]

Dataset: 3
Best alpha: 0.2



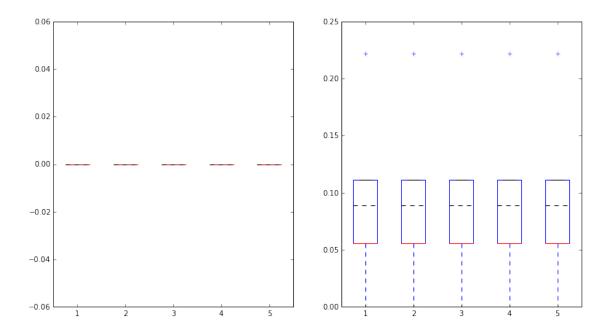
Training error rate: 0.0

Testing error rate: 0.0666666666667

 \mbox{N}° iterations: 94

-321.56787374 2.2425915]

Dataset: 4
Best alpha: 0.2

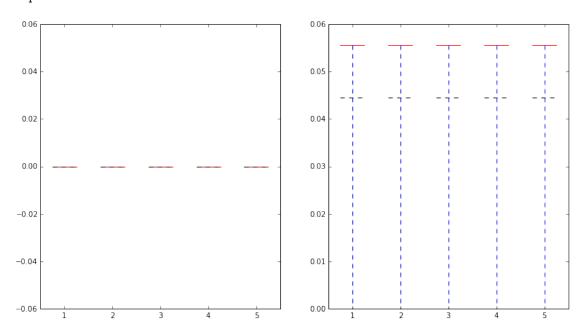


Testing error rate: 0.0666666666667

 N° iterations: 100

-1108.27998538 -16.01121636]

Dataset: 5
Best alpha: 0.2

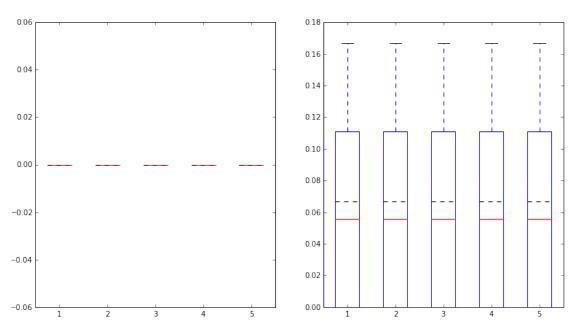


Testing error rate: 0.03333333333333

 N° iterations: 106

-3573.95725761 -7.48025182]

Dataset: 6
Best alpha: 0.2



Training error rate: 0.0

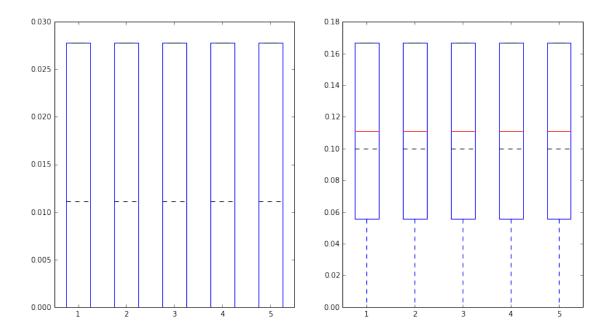
Testing error rate: 0.133333333333

 \mbox{N}° iterations: 95

Beta: [8.02749000e+02 -2.85393577e+00 -2.24002118e+01 -8.28612720e-01

1.81609705e+02 -4.41035636e+02 7.67654920e-03]

Dataset: 7
Best alpha: 0.2

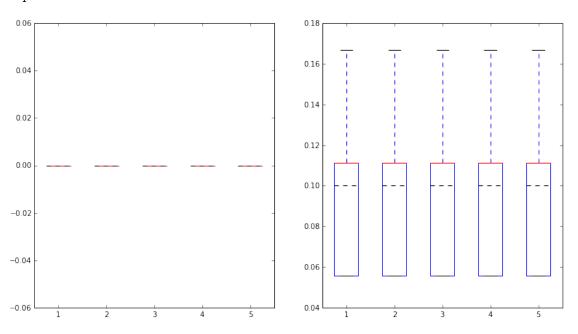


 N° iterations: 100000

Beta: [1.35700989e+02 8.90540535e-01 -4.87957578e+00 -1.94248638e-01

5.94531064e+01 -1.53417253e+02 -6.49440579e-02]

Dataset: 8
Best alpha: 0.2



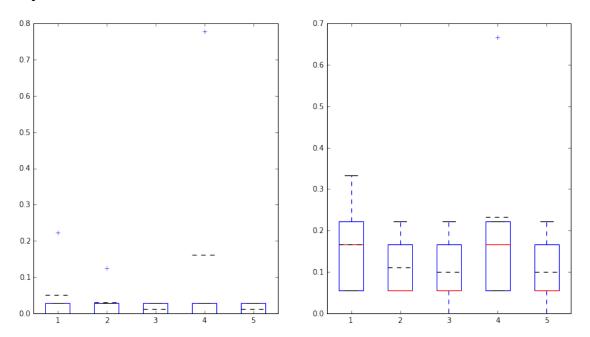
Testing error rate: 0.066666666667

 N° iterations: 97

Beta: [9.61887734e+02 4.31038788e-02 -3.06994970e+01 -1.30182851e+00

2.71252453e+02 -6.57098324e+02 1.82342588e+00]

Dataset: 9
Best alpha: 0.6



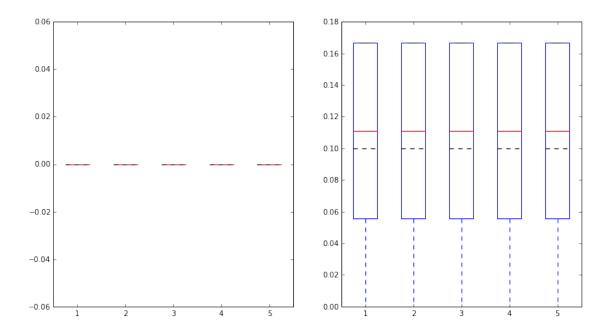
Training error rate: 0.02222222222

Testing error rate: 0.0

 N° iterations: 76

-337.56787276 -0.7495022]

Dataset: 10 Best alpha: 0.2

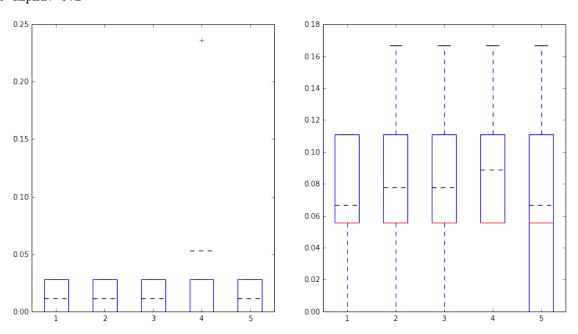


Testing error rate: 0.0666666666667

 N° iterations: 97

-653.93076892 -7.67298756]

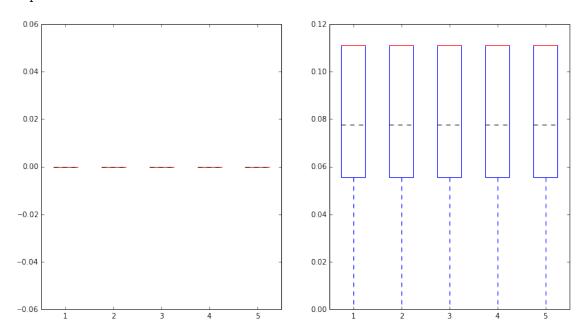
Dataset: 11
Best alpha: 0.2



Testing error rate: 0.0 N° iterations: 217

-342.33741555 -0.74804431]

Dataset: 12 Best alpha: 0.2

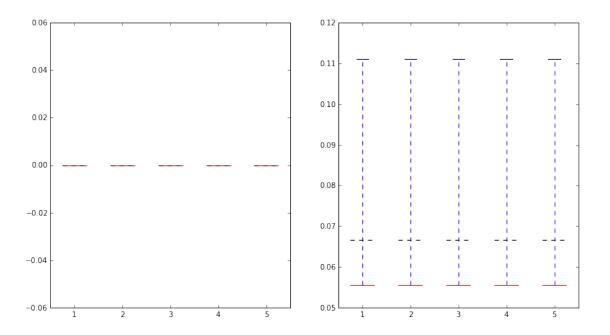


Training error rate: 0.0

 \mbox{N}° iterations: 101

-1300.00392446 -17.2658916]

Dataset: 13
Best alpha: 0.2



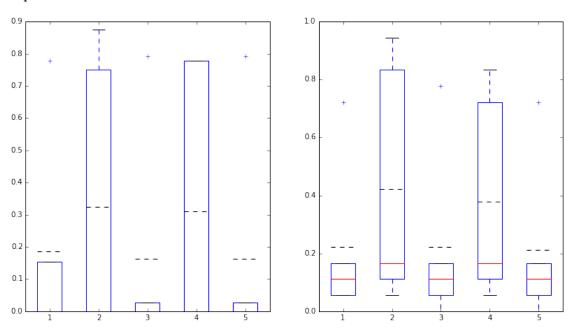
Testing error rate: 0.0666666666667

 N° iterations: 107

Beta: [9053.38570034 -60.04846597 -219.72261609 -9.8154918 1410.78796505

-3254.67460945 -10.87682624]

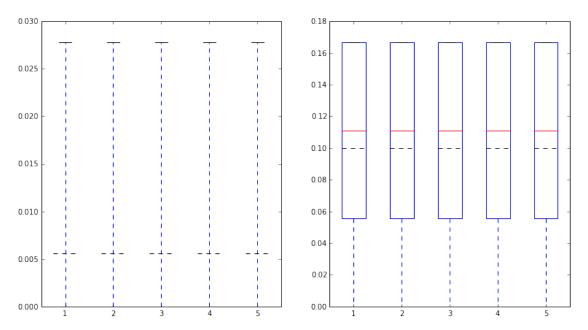
Dataset: 14
Best alpha: 1.0



Testing error rate: 0.0 N° iterations: 42

-335.46684295 -0.7495022]

Dataset: 15
Best alpha: 0.2

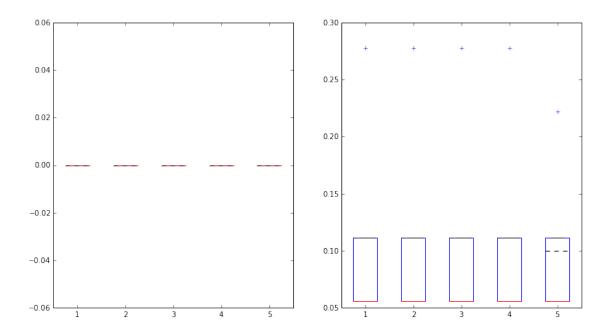


 N° iterations: 197

Beta: [360.40255443 1.7442343 -11.32366332 -0.47161439 131.15765626

-361.85187213 -0.74803991]

Dataset: 16
Best alpha: 1.0



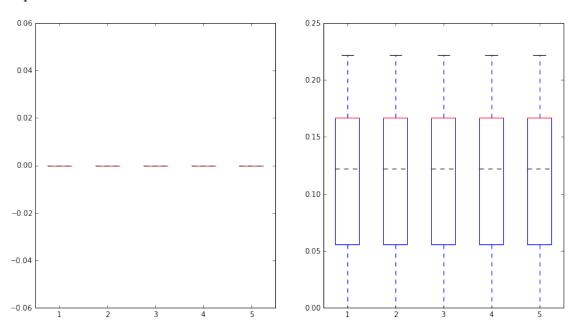
Testing error rate: 0.16666666667

 N° iterations: 20

Beta: [502.75858454 3.38561512 -13.91335666 -0.96227088 140.96090842

-310.64468185 -7.2485097]

Dataset: 17
Best alpha: 0.2

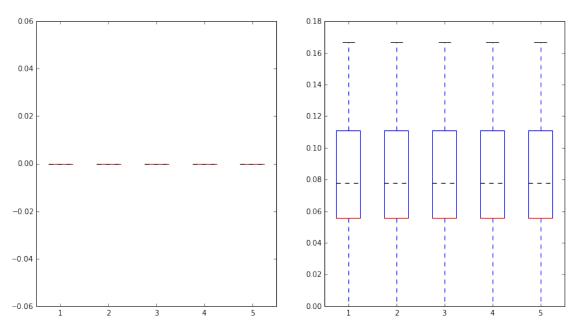


Testing error rate: 0.066666666667

 N° iterations: 100

-1102.65760889 -17.19515579]

Dataset: 18
Best alpha: 0.2



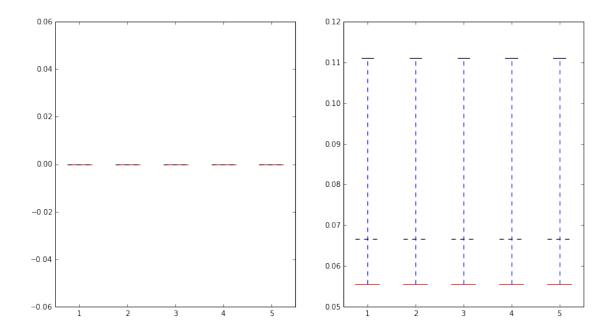
Training error rate: 0.0

Testing error rate: 0.0666666666667

 \mbox{N}° iterations: 101

-1185.02036141 -15.4608018]

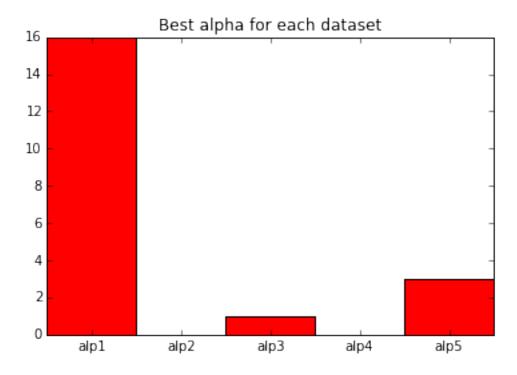
Dataset: 19
Best alpha: 0.2



Testing error rate: 0.133333333333

 \mbox{N}° iterations: 97

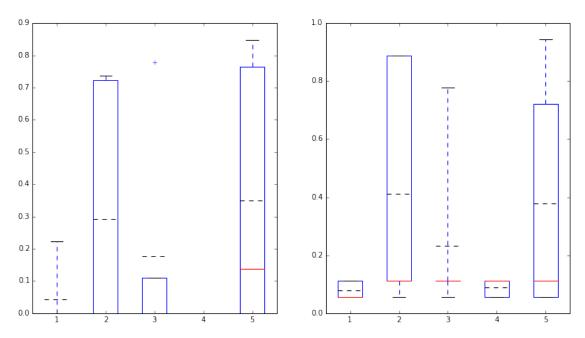
-625.21892347 1.78493656]



0.2.2 Newton-Raphson with rescaled data [0,1]

In [26]: solve_logistic(nr_logistic, params1, rescale)

Dataset: 0
Best alpha: 0.2



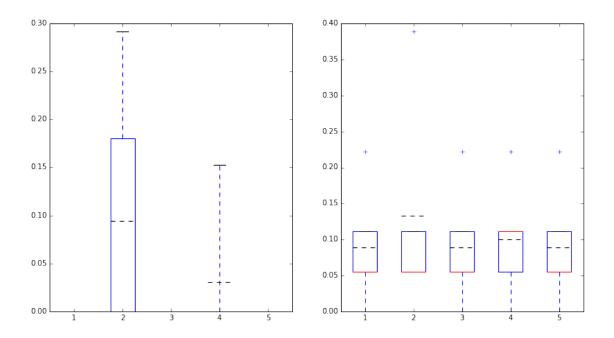
Training error rate: 0.0 Testing error rate: 0.1 \mbox{N}° iterations: 306

N Iterations. 500

Beta: [403.17533722 820.96644589 -4042.53731287 -909.2673169 9013.72824914

-2890.82982924 119.23537701]

Dataset: 1
Best alpha: 0.2



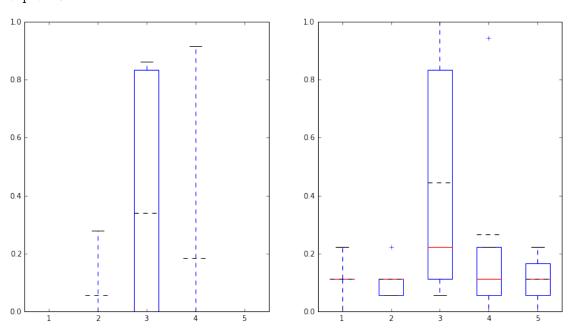
Testing error rate: 0.0666666666667

 \mbox{N}° iterations: 324

Beta: [1368.12235724 -493.17224455 -4650.84991261 -937.58950021 9608.21424165

-2963.62442416 174.00060381]

Dataset: 2
Best alpha: 0.2

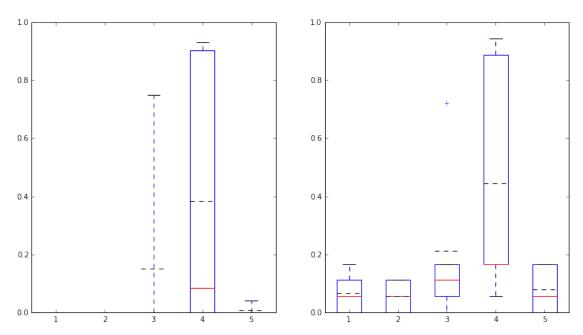


Training error rate: 0.0 Testing error rate: 0.2 \mbox{N}° iterations: 225

Beta: [511.30766707 -132.13826935 -1115.7161734 -385.5482107 2079.42654945

-859.68271936 -213.19909913]

Dataset: 3
Best alpha: 0.4



Training error rate: 0.1

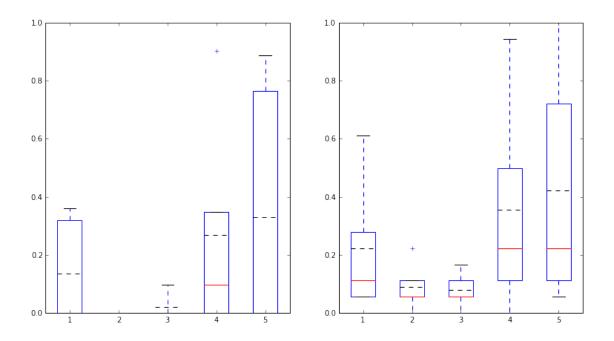
Testing error rate: 0.16666666667

 N° iterations: 105

Beta: [-1.57035286e+29 3.96672376e+31 -9.27562194e+31 -2.37402222e+31

2.48967772e+46 -9.33629145e+45 -8.11296384e+30]

Dataset: 4
Best alpha: 0.6



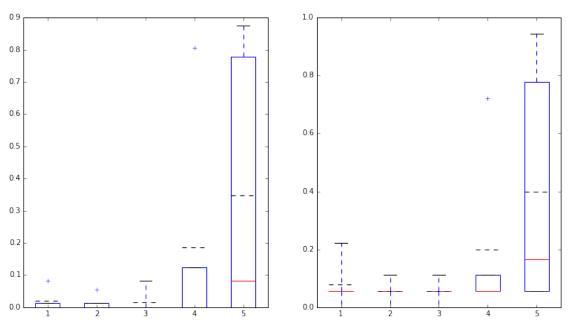
Testing error rate: 0.0666666666667

 \mbox{N}° iterations: 74

Beta: [874.28761684 955.1803553 -4678.56727714 -1352.57682369

10278.28262432 -3112.85045345 -952.99561226]

Dataset: 5
Best alpha: 0.4



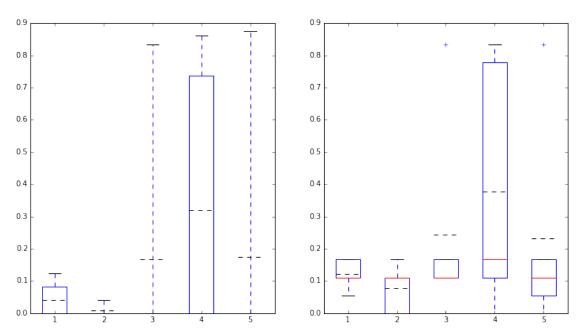
Testing error rate: 0.03333333333333

 N° iterations: 123

Beta: [3121.32974545 1939.88387458 -16715.2552242 -3650.177319

33635.54620703 -10646.86146648 -515.69312123]

Dataset: 6
Best alpha: 0.4

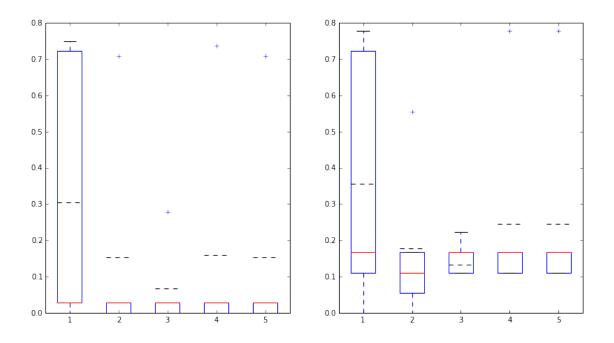


 N° iterations: 104

Beta: [-2.77145831e+40 -1.30191434e+40 1.26358761e+41 3.00151296e+40

3.71592495e+40 -3.09289170e+40 -4.15087276e+38]

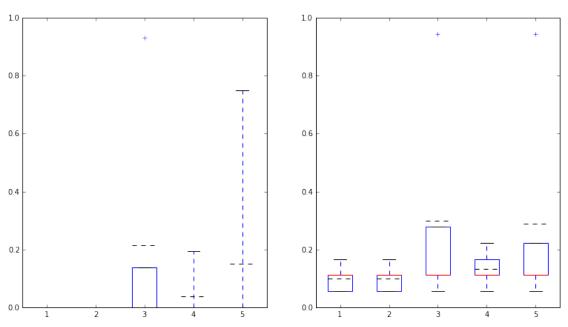
Dataset: 7
Best alpha: 0.6



 ${\tt N}^{\circ}$ iterations: 24

-139.8762308 -1.39994843]

Dataset: 8
Best alpha: 0.2

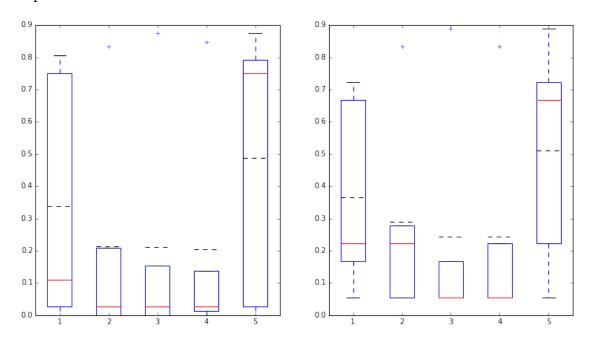


 N° iterations: 202

Beta: [-8.13734170e+15 -9.68815889e+14 3.98262971e+16 8.76511885e+15

-5.09709947e+16 1.65705267e+16 -1.91214795e+15]

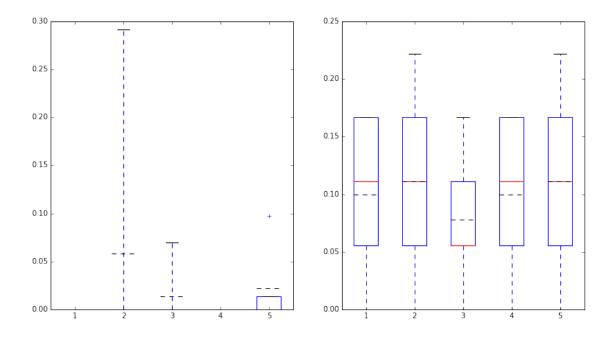
Dataset: 9
Best alpha: 0.6



 N° iterations: 68

5.51240451e+17 -2.06715169e+17 -4.51391668e+03]

Dataset: 10 Best alpha: 0.6



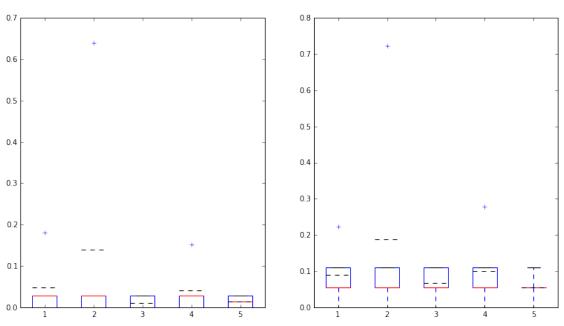
Testing error rate: 0.9

 \mbox{N}° iterations: 70

Beta: [5.31213725e+02 4.26657108e+02 -2.69686307e+03 -7.25991907e+02

-3.41567293e+14 1.46385983e+14 -4.62262281e+02]

Dataset: 11 Best alpha: 1.0

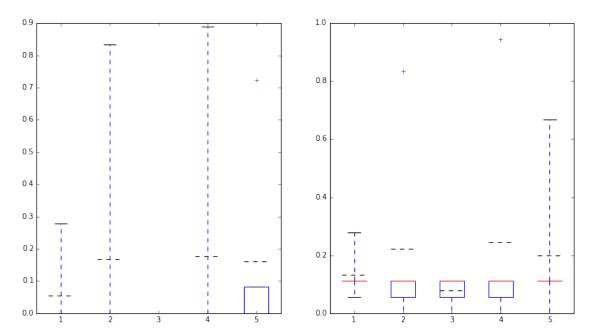


 N° iterations: 16

Beta: [72.54940286 64.84160581 -440.58204416 -98.54777607 957.10260405

-306.68916908 -16.21455461]

Dataset: 12 Best alpha: 0.6



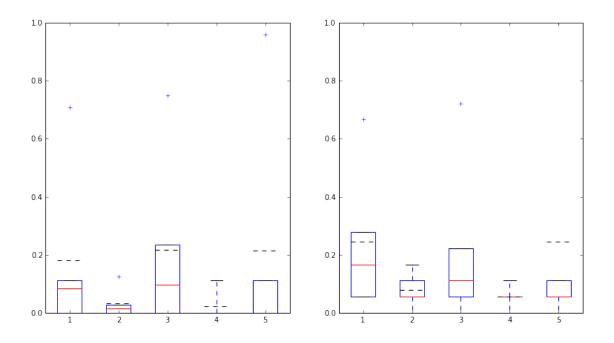
Training error rate: 0.27777777778

Testing error rate: 0.3

 ${\tt N}^{\circ}$ iterations: 66

9.99510622e+03 3.60594331e+13 -8.75620746e+02]

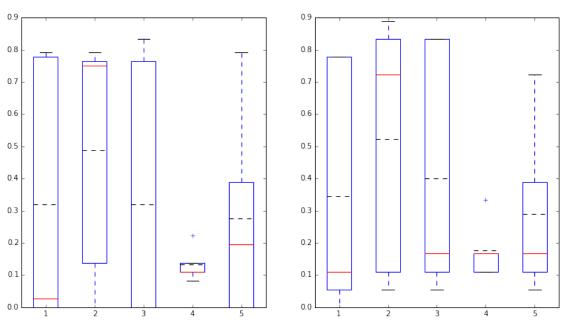
Dataset: 13
Best alpha: 0.8



Testing error rate: 0.1 N° iterations: 17

4.12375442e+19 -8.15442266e+18 -8.71048641e+17]

Dataset: 14
Best alpha: 0.8



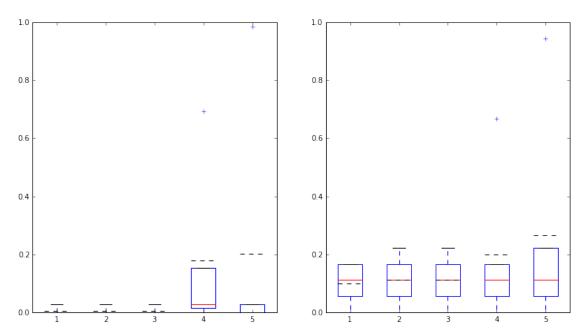
Training error rate: 0.7888888888889 Testing error rate: 0.566666666667

 N° iterations: 57

Beta: [1.96382055e+05 1.18958614e+05 -8.89383825e+05 -1.99386501e+05

-2.43394616e+17 9.12729809e+16 -3.77438072e+04]

Dataset: 15
Best alpha: 0.2

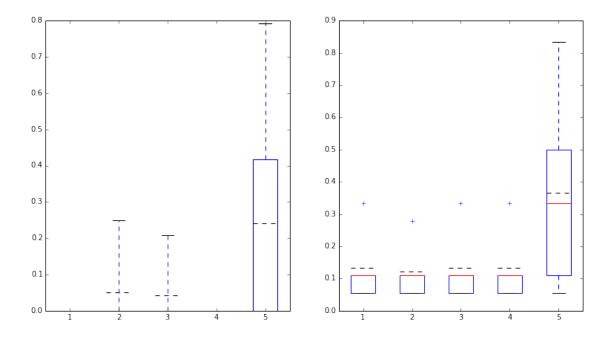


 N° iterations: 70

Beta: [68.02194047 64.34511143 -425.57134887 -97.73027239 942.50877792

-297.75309867 -14.36196829]

Dataset: 16
Best alpha: 0.4



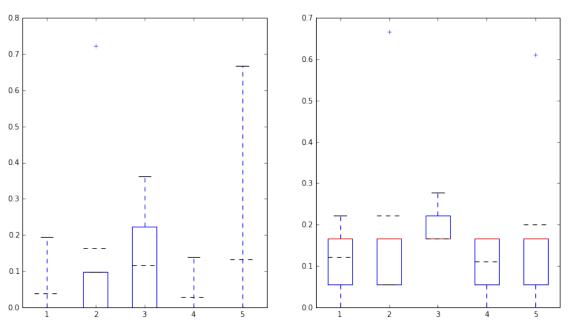
Testing error rate: 0.16666666667

 N° iterations: 123

Beta: [328.02015068 371.15795255 -1533.66017163 -578.03959615 3255.8185186

-897.39047035 -489.44739504]

Dataset: 17
Best alpha: 0.8



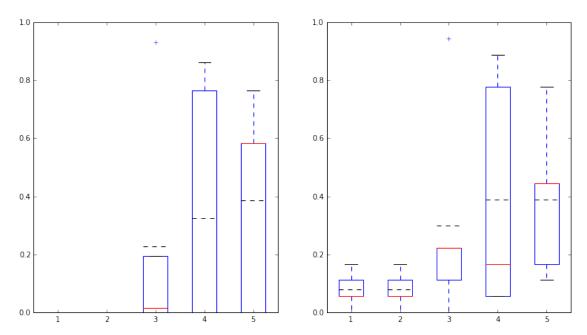
Training error rate: 0.911111111111
Testing error rate: 0.7333333333333

 N° iterations: 52

Beta: [-2.37438205e+41 -9.05630048e+28 4.51261867e+29 1.36691938e+29

-8.80933445e+29 2.37438205e+41 9.21927709e+28]

Dataset: 18
Best alpha: 0.2



Training error rate: 0.0

Testing error rate: 0.03333333333333

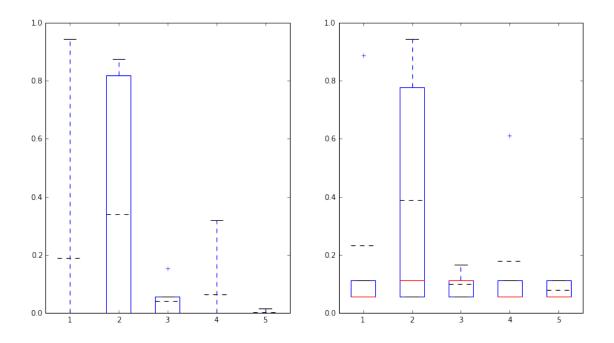
 N° iterations: 224

Beta: [554.59164122 874.96986471 -435

874.96986471 -4353.20082791 -1022.94638004

10565.64911993 -3271.59931233 -989.9647166]

Dataset: 19
Best alpha: 1.0

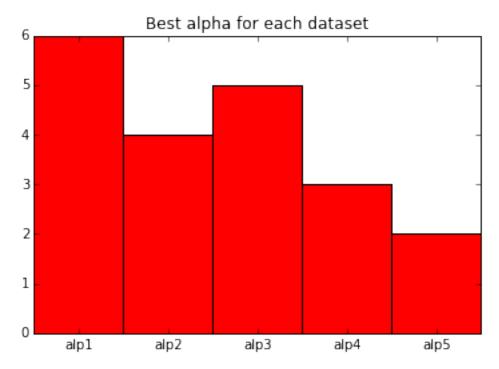


Training error rate: 0.855555555556 Testing error rate: 0.766666666667

 N° iterations: 43

Beta: [-1.43882276e+19 2.27702581e+19 -1.80704305e+18 3.11197634e+17

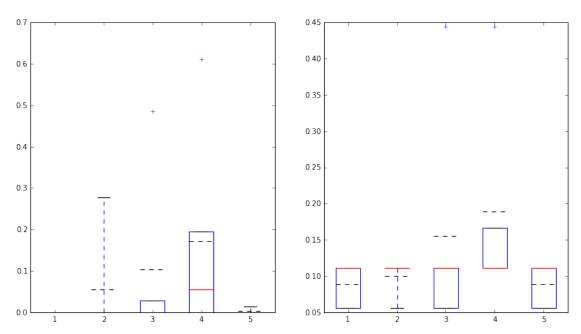
-2.53530120e+30 1.08655766e+30 -4.44988616e+17]



0.2.3 Newton-Raphson with normalized data

In [27]: solve_logistic(nr_logistic, params1, normalize)

Dataset: 0
Best alpha: 0.2



Training error rate: 0.0

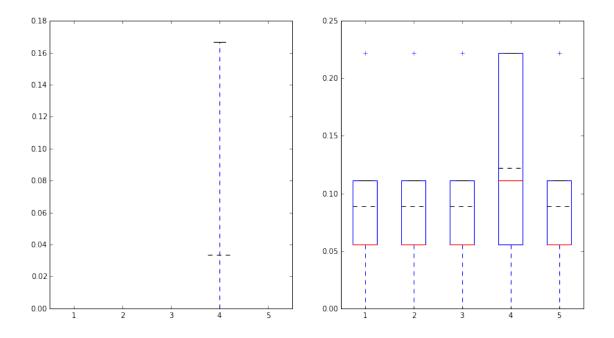
Testing error rate: 0.033333333333333

 \mbox{N}° iterations: 260

Beta: [-60.14111749 210.91643787 -995.83336414 -323.98326533 1944.18606247

-1254.46881236 25.67057134]

Dataset: 1
Best alpha: 0.2



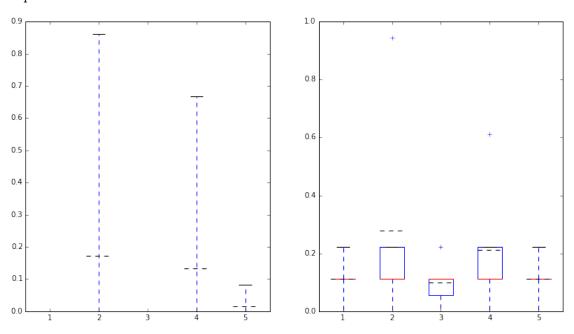
Testing error rate: 0.0666666666667

 ${\tt N}^{\circ}$ iterations: 285

Beta: [-42.80901837 -129.95201628 -1205.71573278 -349.37134083 2074.0937013

-1337.16263784 34.93461495]

Dataset: 2
Best alpha: 0.6

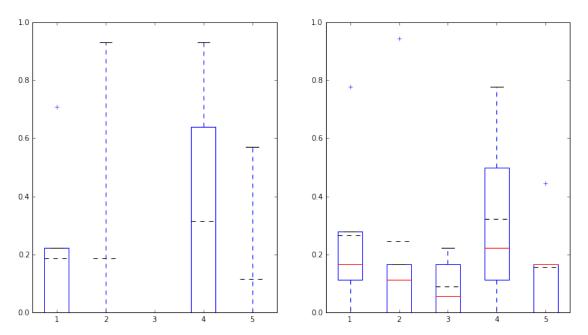


Testing error rate: 0.133333333333

 N° iterations: 94

-481.48139903 -50.25590159]

Dataset: 3
Best alpha: 0.6



Training error rate: 0.0

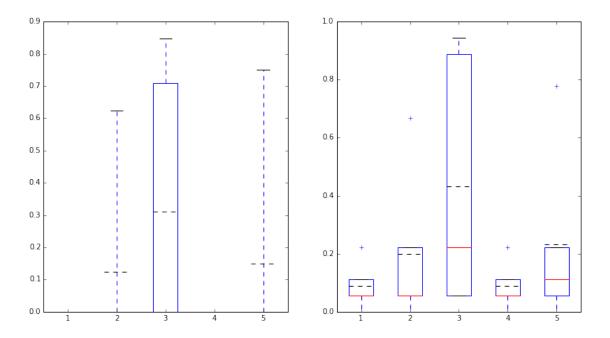
Testing error rate: 0.066666666667

 N° iterations: 124

Beta: [4.36463668 65.10222628 -517.91559563 -114.38409119 998.56879844

-589.54871087 35.28941142]

Dataset: 4
Best alpha: 0.2



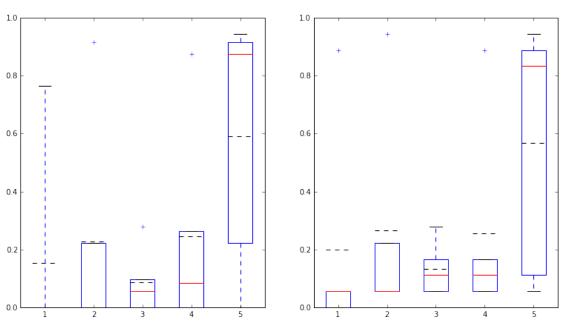
Testing error rate: 0.16666666667

 ${\tt N}^{\circ}$ iterations: 234

Beta: [-433.52931172 271.9301953 -1291.1347422 -570.52546171 2568.90930335

-1469.17296951 -219.05165906]

Dataset: 5
Best alpha: 0.6



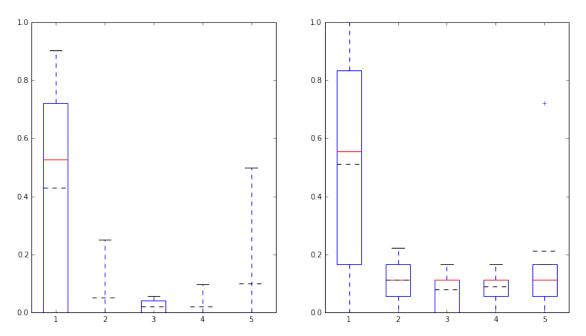
Training error rate: 0.077777777778
Testing error rate: 0.133333333333

 N° iterations: 72

Beta: [2.51848847e+29 1.03203056e+17 -8.57785898e+17 -2.98654118e+17

1.05222271e+30 -8.19210724e+29 -2.24860225e+16]

Dataset: 6
Best alpha: 0.6



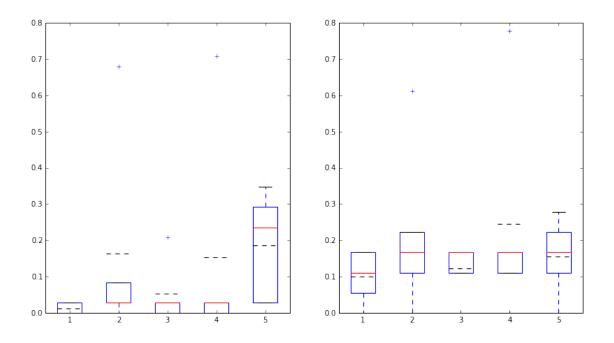
Training error rate: 0.388888888889

Testing error rate: 0.5 N° iterations: 101

Beta: [1.74672409e+26 1.18026073e+26 -3.47775473e+26 8.36594763e+25

2.90131401e+26 -2.79082871e+26 -6.27877655e+26]

Dataset: 7
Best alpha: 0.2

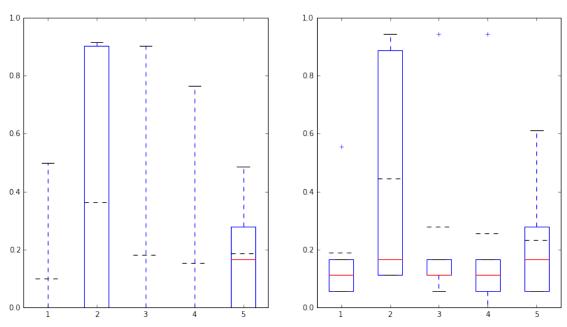


 N° iterations: 67

Beta: [-6.74149961 9.27585304 -52.71052459 -16.58532723 102.25711105

-66.92244913 -0.30369144]

Dataset: 8
Best alpha: 0.2



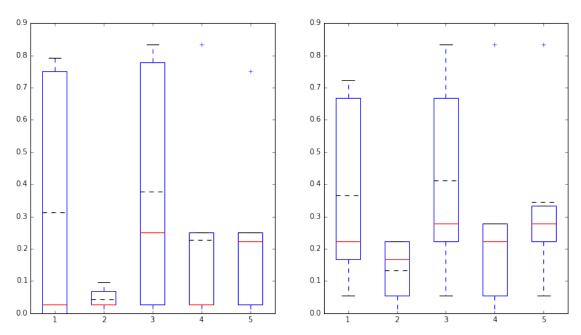
Training error rate: 0.0666666666667 Testing error rate: 0.0666666666667

 N° iterations: 223

Beta: [-1.51380877e+26 -1.52089068e+27 -8.57565589e+27 -2.99475370e+27

1.00417080e+28 -6.08338731e+27 1.64718150e+26]

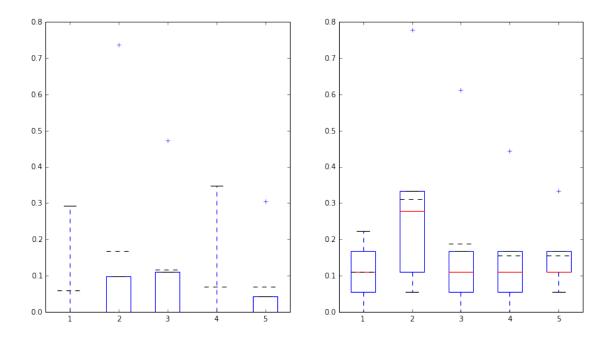
Dataset: 9
Best alpha: 0.4



 N° iterations: 110

Beta: [-1.29477667e+21 6.69349076e+20 5.00600126e+19 7.45498629e+19

Dataset: 10
Best alpha: 0.2



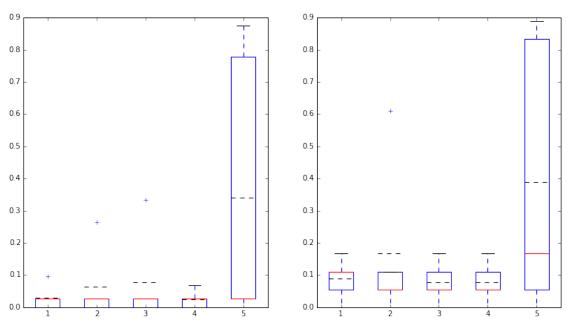
Training error rate: 0.0 Testing error rate: 0.1

 \mbox{N}° iterations: 346

Beta: [-137.62166984 171.14019211 -1074.43374367 -446.00940624 1928.00248824

-1180.98206299 -138.83050789]

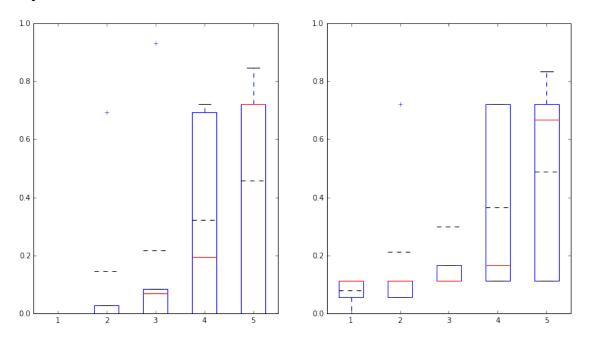
Dataset: 11 Best alpha: 0.6



 N° iterations: 25

-147.95097367 -3.73099628]

Dataset: 12 Best alpha: 0.2

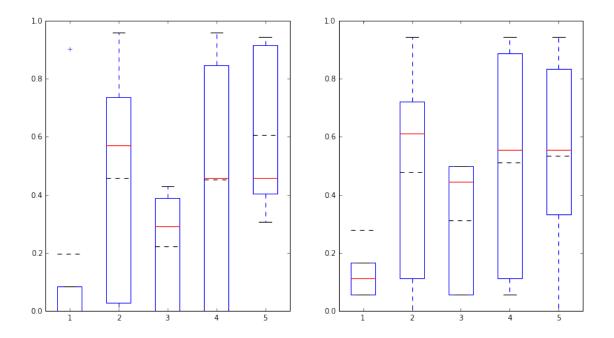


 N° iterations: 197

Beta: [-2.43120385e+42 -2.45921814e+29 1.27409267e+30 5.64538432e+29

-2.42039603e+30 3.27310835e+42 2.06891725e+29]

Dataset: 13
Best alpha: 0.2



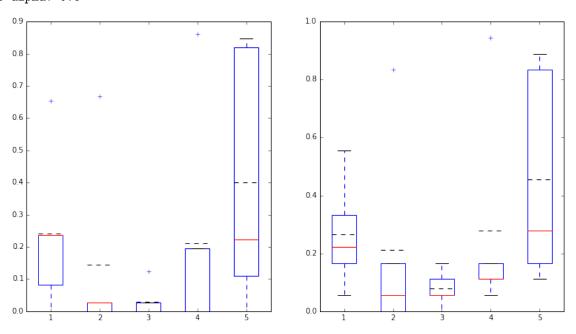
Testing error rate: 0.133333333333

 $\ensuremath{\text{N}^{\circ}}$ iterations: 209

Beta: [-865.77539564 -1525.89481516 -5925.05741319 -2155.1086108 6521.49412554

-4052.32268649 -143.88882892]

Dataset: 14
Best alpha: 0.6

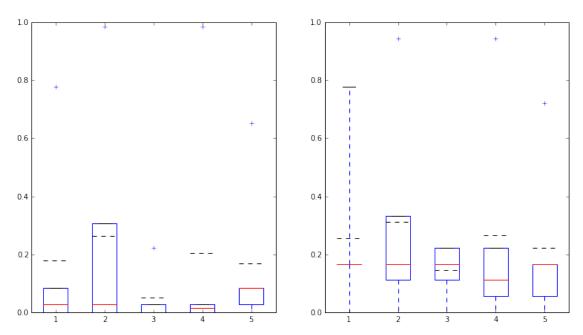


 N° iterations: 69

Beta: [16.96227501 18.18216981 -124.95847509 -40.73476592 262.3598964

-165.46726821 -3.77485352]

Dataset: 15
Best alpha: 0.6



Training error rate: 0.02222222222

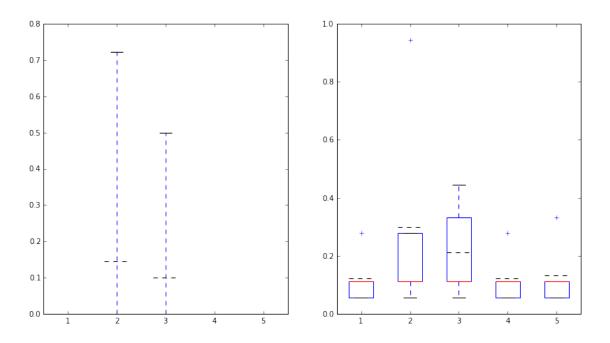
Testing error rate: 0.1

 N° iterations: 25

Beta: [-1.74335987 17.84144914 -118.43731658 -41.10808597 218.58570218

-148.04563625 -3.49285296]

Dataset: 16
Best alpha: 0.2

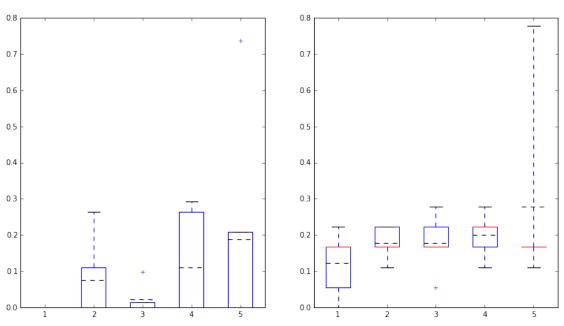


Training error rate: 0.0 Testing error rate: 0.2 \mbox{N}° iterations: 218

Beta: [8.65008245 92.12626372 -373.64109105 -221.1702541 750.04693164

-396.01611494 -102.25811851]

Dataset: 17
Best alpha: 0.2



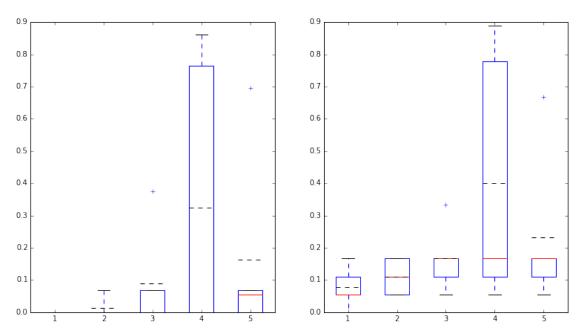
Training error rate: 0.9111111111111
Testing error rate: 0.7333333333333

 N° iterations: 199

Beta: [-5.56370127e+42 -1.32346168e+29 6.59239789e+29 3.00352842e+29

-1.14582378e+30 7.67554040e+42 1.18563962e+29]

Dataset: 18
Best alpha: 0.2



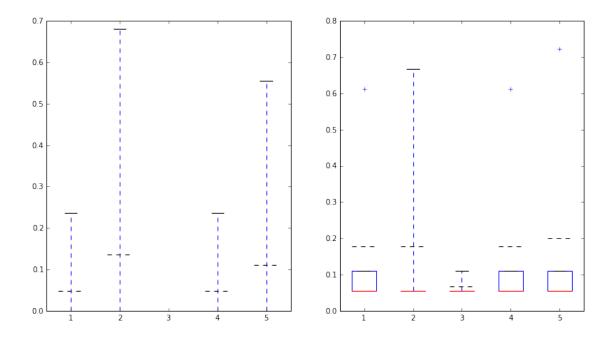
Training error rate: 0.77777777778

Testing error rate: 0.8 N° iterations: 205

Beta: [-5.79597525e+43 3.31140992e+30 -1.71173985e+31 -6.11268280e+30

-1.36031432e+44 1.03601057e+44 -2.84108169e+30]

Dataset: 19
Best alpha: 0.6

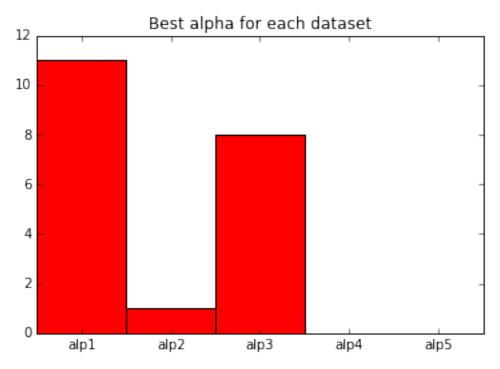


Testing error rate: 0.7

 \mbox{N}° iterations: 71

Beta: [3.61265136e+17 1.73692100e+14 6.86958778e+17 2.52165849e+17

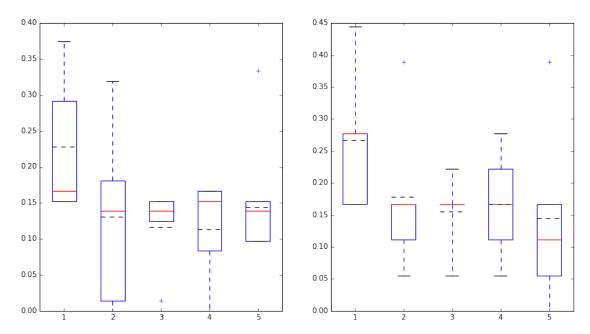
-1.60878823e+17 -4.84950962e+16 -1.03967275e+16]



0.2.4 Stochastic Gradient Ascent with raw data

In []: solve_logistic(gd_stochastic, params2)

Dataset: 0
Best alpha: 0.01

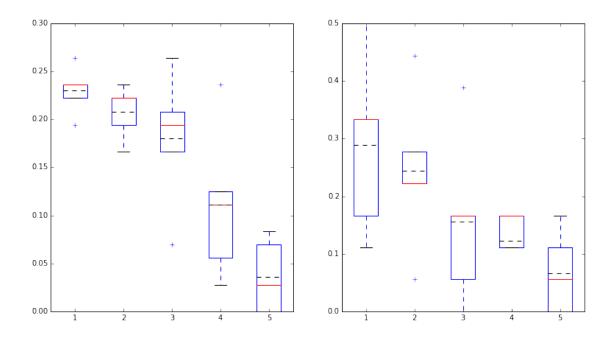


 ${\tt N}^{\circ}$ iterations: 5597

Beta: [-2.77201882 3.38715113 -3.35672344 0.29379831 72.6531521

-189.58276094 0.38774768]

Dataset: 1

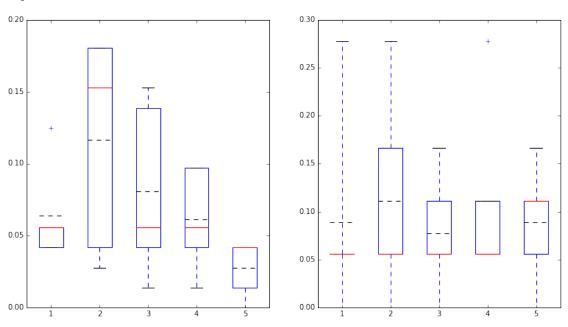


 N° iterations: 100000

Beta: [2.96220868e+02 5.94310552e+00 -1.62573263e+01 -3.97416950e-01

2.73784075e+02 -7.30271374e+02 3.56037168e+00]

Dataset: 2



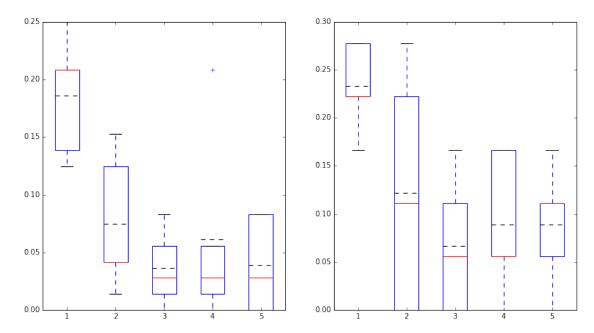
Testing error rate: 0.233333333333

 N° iterations: 1594

-46.01375111 2.38068629]

Dataset: 3

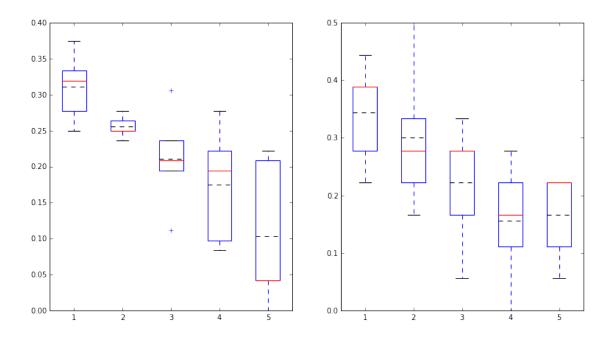
Best alpha: 0.0055



 N° iterations: 36061

-238.25081696 -2.6841496]

Dataset: 4

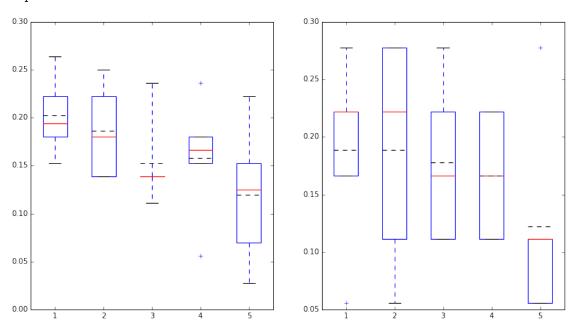


 N° iterations: 100000

Beta: [5.50725896e+01 1.23452509e+01 -1.83093594e+01 6.56583251e-01

2.71031080e+02 -7.13888691e+02 9.08417475e+00]

Dataset: 5
Best alpha: 0.01



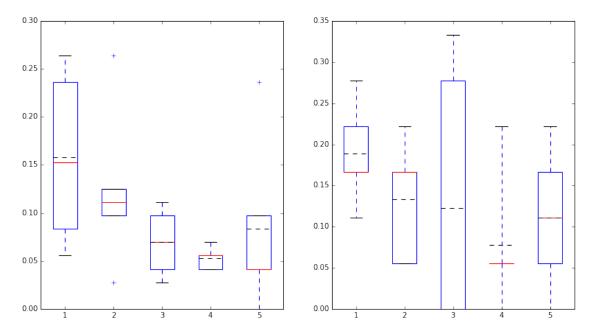
 N° iterations: 100000

Beta: [1.98637878e+02 4.59190459e+00 -1.76701364e+01 1.13880414e-01

2.12302462e+02 -4.96390153e+02 1.02183057e+01]

Dataset: 6

Best alpha: 0.00775

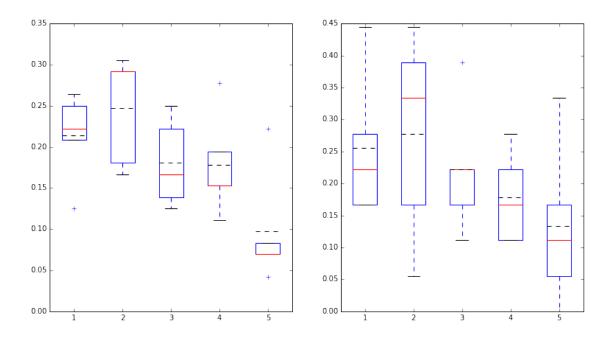


 ${\tt N}^{\circ}$ iterations: 24426

Beta: [66.28943327 2.62815811 -4.44461828 -0.31044347 83.28410451

-251.7378331 -0.33811518]

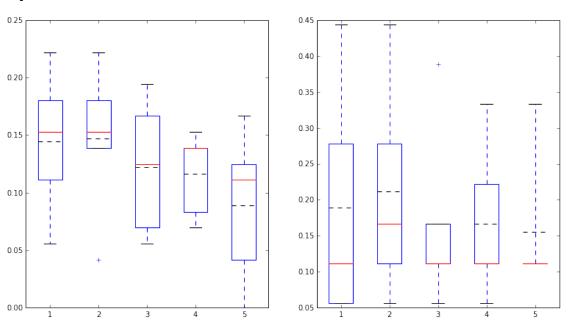
Dataset: 7
Best alpha: 0.01



Training error rate: 0.1 Testing error rate: 0.0 $\ensuremath{\mathbb{N}}^\circ$ iterations: 5517

-204.74675816 4.3879038]

Dataset: 8
Best alpha: 0.01

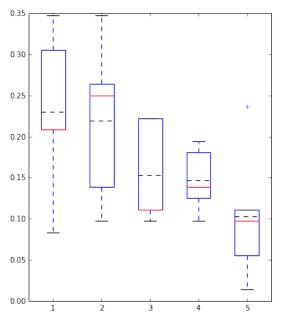


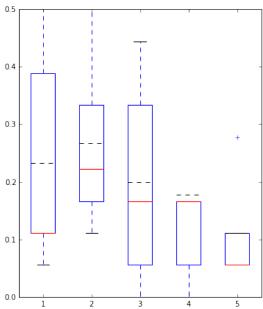
Testing error rate: 0.1 N° iterations: 7437

Beta: [15.82949665 2.33619918 -4.12898442 -0.27931539 81.40187178

-248.93266015 3.48212139]

Dataset: 9
Best alpha: 0.01



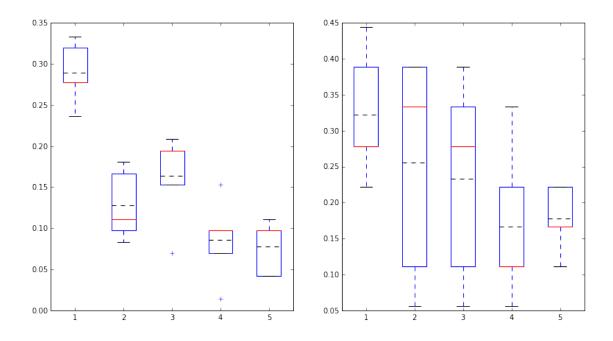


 N° iterations: 100000

Beta: [6.47049595e+01 1.73374504e+01 -1.38824100e+01 -6.41161445e-01

2.95320384e+02 -9.17307229e+02 -1.07515629e+00]

Dataset: 10



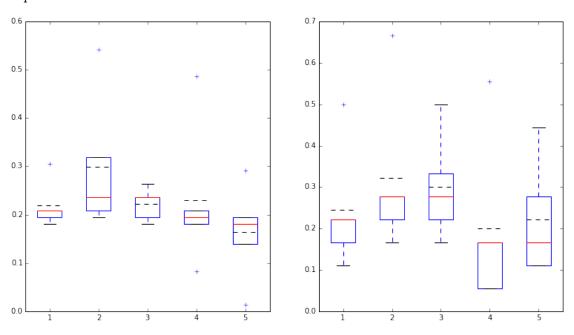
Training error rate: 0.077777777778
Testing error rate: 0.0333333333333

 ${\tt N}^{\circ}$ iterations: 15573

Beta: [34.19251485 4.32565401 -7.31939695 -0.65673664 101.94890151

-240.34929982 3.18664288]

Dataset: 11

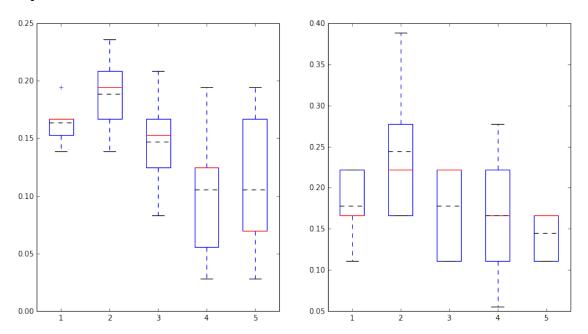


Testing error rate: 0.3 N° iterations: 8440

Beta: [7.95196568 4.38307616 -2.05560182 1.33837748 59.56178386

-198.48172868 -1.78307151]

Dataset: 12 Best alpha: 0.01



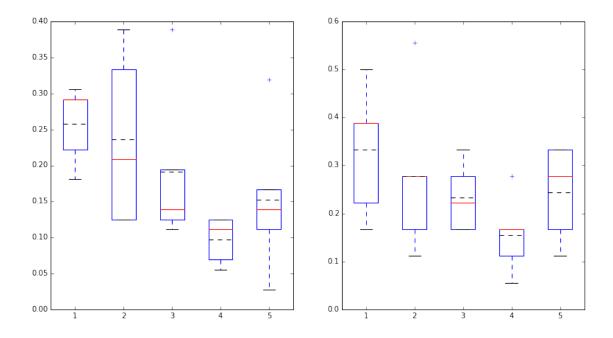
Training error rate: 0.0666666666667

Testing error rate: 0.0 $\ensuremath{\text{N}}^\circ$ iterations: 8075

Beta: [37.28120688 4.77510427 -3.04491408 -0.55979802 93.73152904

-295.60391698 -4.06865679]

Dataset: 13

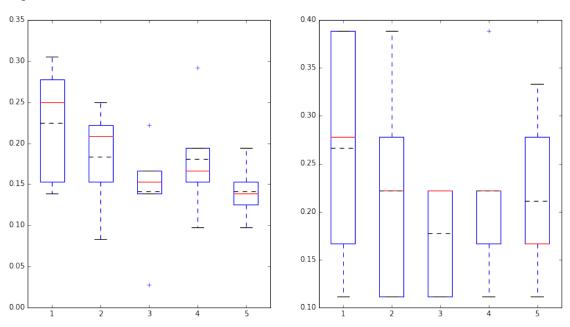


Training error rate: 0.1 Testing error rate: 0.1 N° iterations: 8295

Beta: [16.72232918 2.14718009 -4.74198117 -1.39808779 80.07311116

-206.35093256 2.50869161]

Dataset: 14



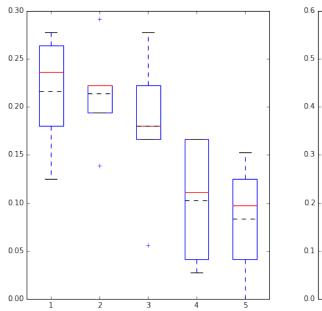
Training error rate: 0.166666666667 Testing error rate: 0.233333333333

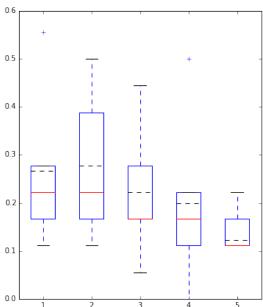
 N° iterations: 4379

Beta: [5.53236866 1.99840753 -3.42166633 0.87815441 48.18881

-115.00761954 2.59538129]

Dataset: 15
Best alpha: 0.01



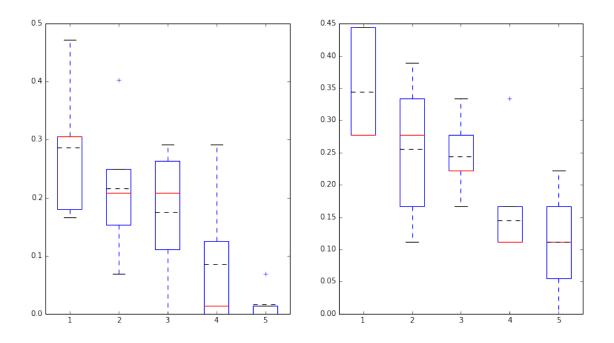


 ${\tt N}^{\circ}$ iterations: 9090

Beta: [4.81095935e+01 4.25893090e+00 -9.47334747e+00 5.62223680e-02

1.19851884e+02 -3.25760262e+02 7.85554049e+00]

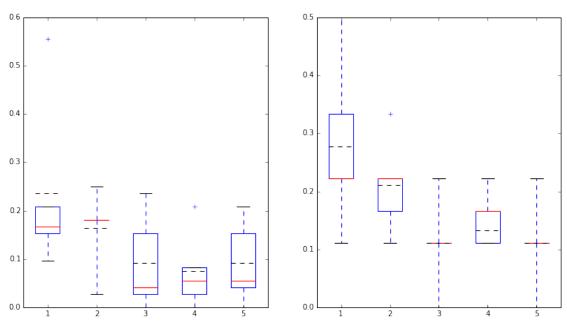
Dataset: 16
Best alpha: 0.01



 N° iterations: 100000

-740.97221574 -7.08799705]

Dataset: 17



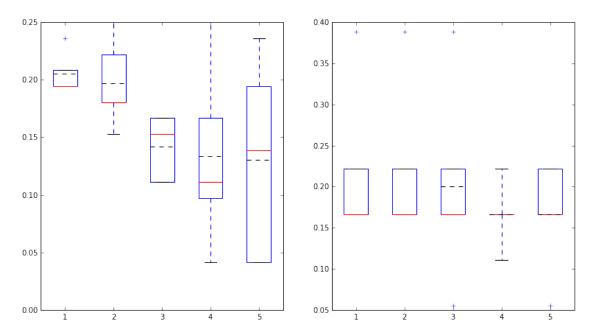
 N° iterations: 8505

Beta: [1.41167539e+00 3.25496018e-01 -3.43224585e+00 8.04290874e-02

5.53515028e+01 -1.25886642e+02 3.91139290e+00]

Dataset: 18

Best alpha: 0.00775

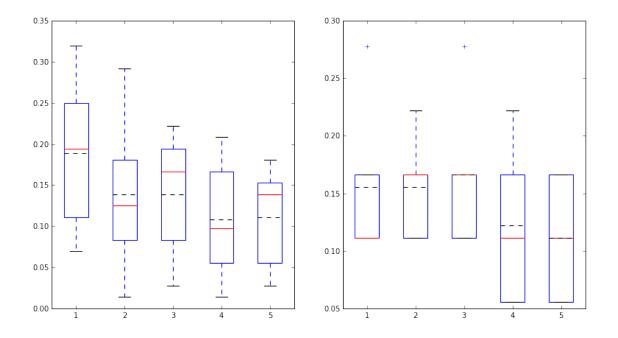


 N° iterations: 13747

Beta: [21.87934726 4.98441172 -4.05983966 0.63331549 100.51241416

-246.97168278 -5.14995137]

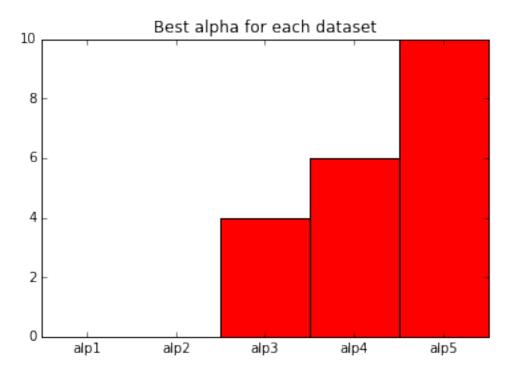
Dataset: 19 Best alpha: 0.01



 $\ensuremath{\text{N}^{\circ}}$ iterations: 100000

Beta: [3.00010214e+02 8.60435730e+00 -1.06446263e+01 -4.34057356e-01

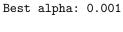
1.99552508e+02 -6.79706111e+02 -6.38398862e+00]

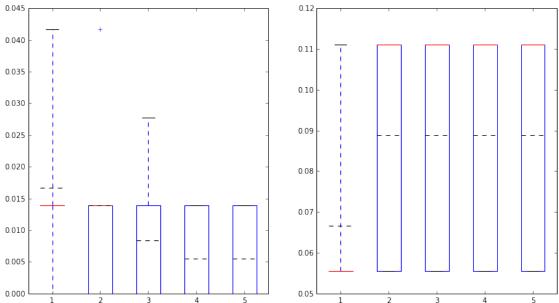


0.2.5 Stochastic Gradient Ascent with rescaled data [0,1]

In [7]: solve_logistic(gd_stochastic, params2, rescale)

Dataset: 0



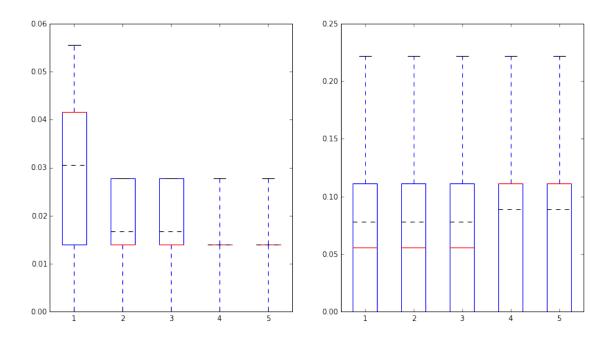


 N° iterations: 30094

Beta: [-1.48226982 3.10845433 -6.56795702 -0.74596999 20.94421037

-7.39747949 1.47998218]

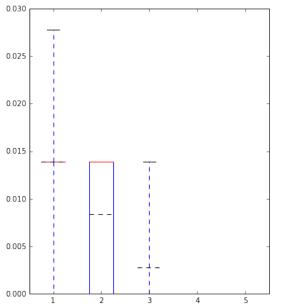
Dataset: 1

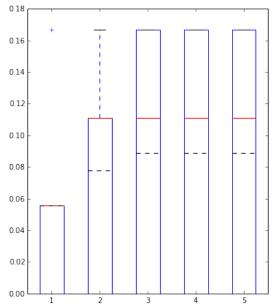


 N° iterations: 30362

-7.27964003 0.10261293]

Dataset: 2
Best alpha: 0.001



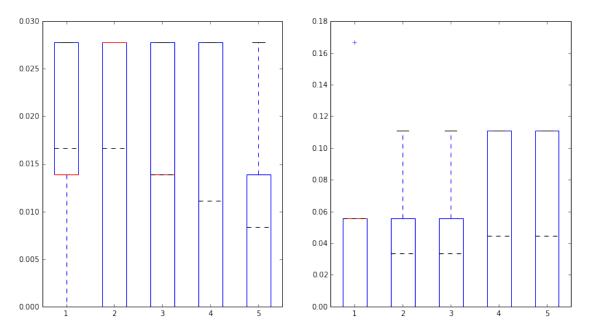


 N° iterations: 33692

-9.31836999 -0.62494963]

Dataset: 3

Best alpha: 0.00325

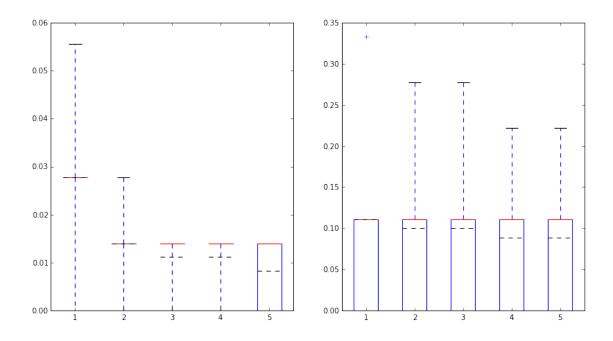


Training error rate: 0.022222222222

Testing error rate: 0.2 \mathbb{N}° iterations: 33377

-11.32724045 3.9459928]

Dataset: 4

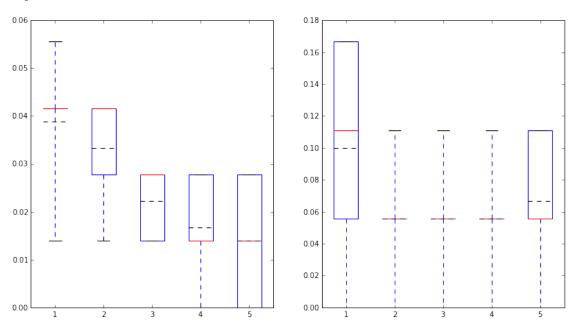


 \mbox{N}° iterations: 22589

Beta: [1.37777031 2.92424144 -15.17935656 -2.77520176 39.89058245

-13.26955394 -0.3819851]

Dataset: 5



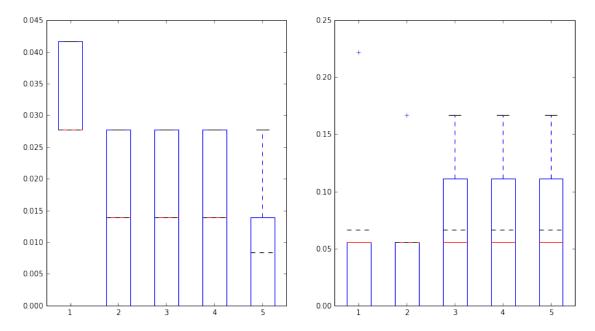
 N° iterations: 22923

Beta: [0.24919718 2.277642 -11.05563371 -1.50702535 28.99501356

-9.47242442 0.84556142]

Dataset: 6

Best alpha: 0.00325



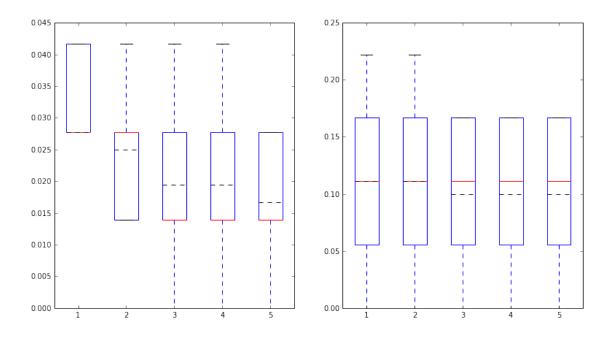
Training error rate: 0.022222222222

Testing error rate: 0.1 N° iterations: 30493

Beta: [2.75727723 0.27848703 -13.81133473 -2.4019636 29.99318716

-11.43199973 0.09017219]

Dataset: 7

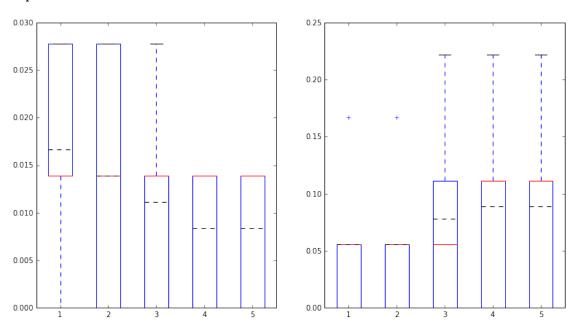


 N° iterations: 22254

Beta: [-1.89327687 3.43563265 -12.83768893 -0.43677254 35.49873424

-11.50203978 3.65426577]

Dataset: 8



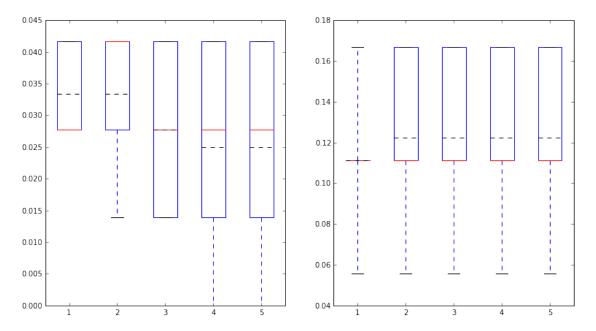
 N° iterations: 29501

Beta: [-0.08223237 2.11576518 -7.58685974 -1.69763745 20.04931902

-8.50629683 0.95290131]

Dataset: 9

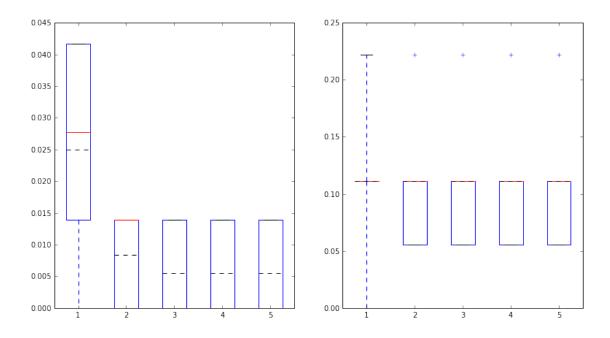
Best alpha: 0.001



 ${\tt N}^{\circ}$ iterations: 24527

-6.58761284 0.65722461]

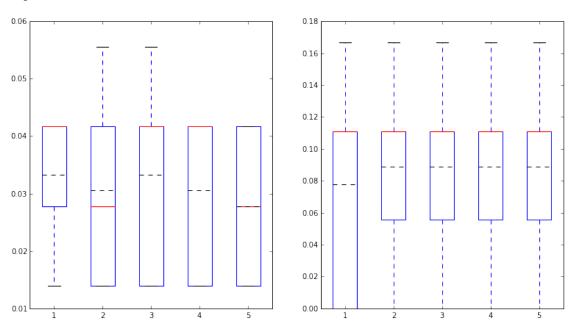
Dataset: 10



Testing error rate: 0.1 \mbox{N}° iterations: 28419

-7.58313334 0.84540945]

Dataset: 11
Best alpha: 0.001

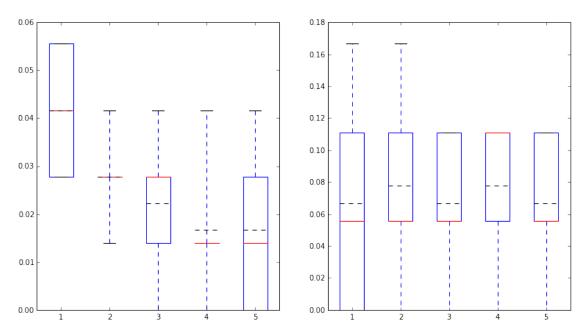


 N° iterations: 26769

Beta: [-1.10786796 2.33844479 -5.27238146 -0.12312292 20.34605662

-7.47816117 0.75113069]

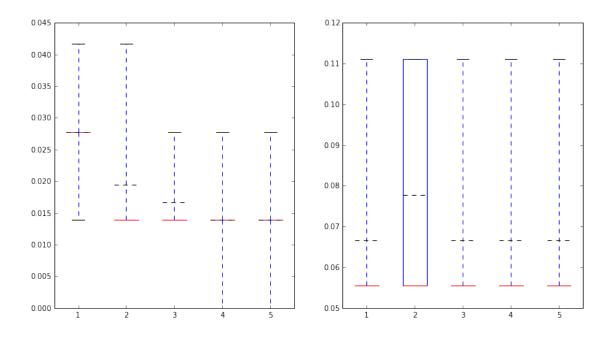
Dataset: 12 Best alpha: 0.001



 \mbox{N}° iterations: 26266

-7.03030324 0.71270764]

Dataset: 13 Best alpha: 0.001

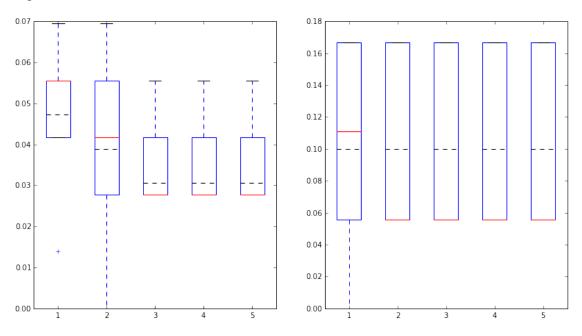


Testing error rate: 0.0 $\ensuremath{\text{N}^{\circ}}$ iterations: 28551

Beta: [0.7605753 1.66421907 -8.25068576 -1.86717674 19.96618115

-7.07100621 0.0964316]

Dataset: 14
Best alpha: 0.001

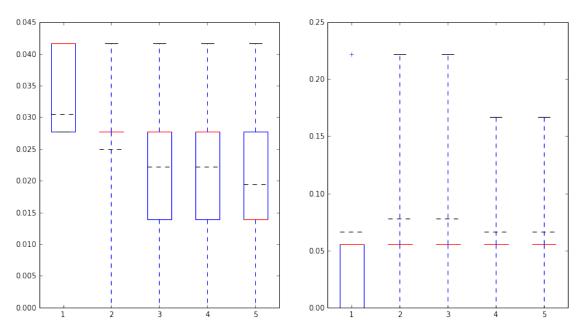


 N° iterations: 25697

Beta: [-1.42772421 3.50726093 -6.07024427 -1.06006468 20.35926098

-6.98091889 0.2542705]

Dataset: 15
Best alpha: 0.001

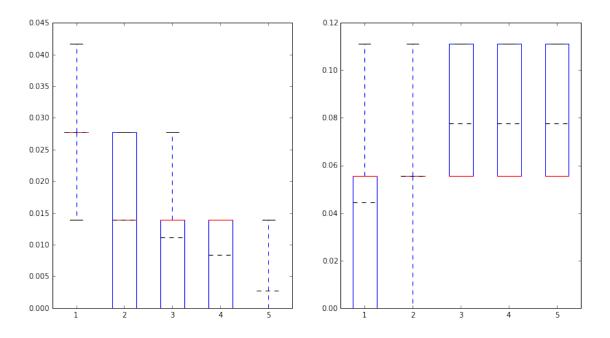


 ${\tt N}^{\circ}$ iterations: 27432

Beta: [0.57046464 1.59463087 -7.43514893 -1.12446697 20.10535806

-7.5073541 1.33498818]

Dataset: 16

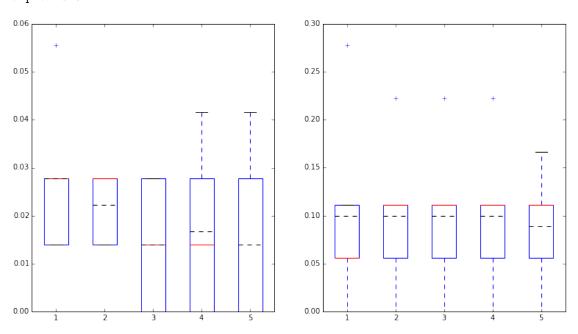


Testing error rate: 0.2 N° iterations: 33010

Beta: [-2.77940767 3.89934021 -5.64189862 -0.83647005 22.61962827

-7.2236446 1.03948318]

Dataset: 17
Best alpha: 0.01

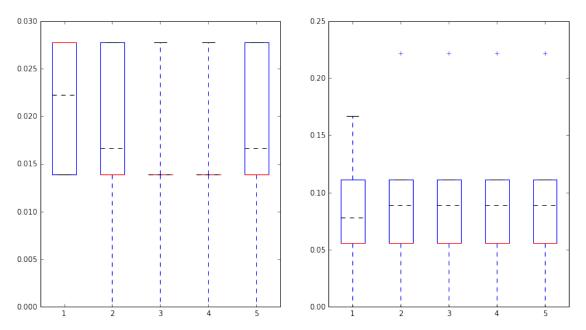


 N° iterations: 22359

Beta: [2.51850734 2.66217391 -16.82498501 -4.55875256 41.23397366

-12.51769947 -3.720826]

Dataset: 18
Best alpha: 0.001

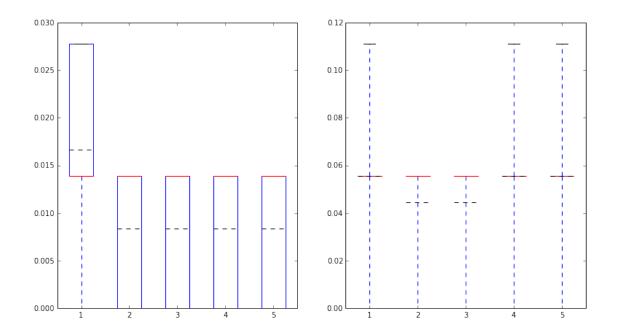


 ${\tt N}^{\circ}$ iterations: 28333

Beta: [-1.93638098 2.48464656 -4.92888944 0.1708657 20.42787817

-7.23494195 -0.14271104]

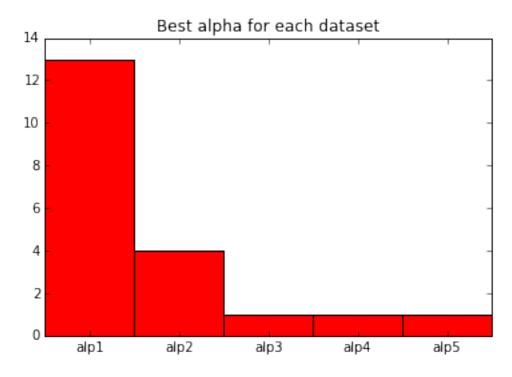
Dataset: 19



 ${\tt N}^{\circ}$ iterations: 27880

Beta: [3.31790124 1.5579711 -12.09740063 -3.83640828 26.5917787

-11.23145584 -1.83157801]

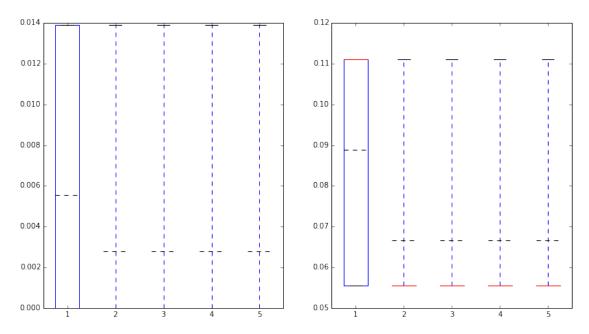


0.2.6 Stochastic Gradient Ascent with normalized data

In [8]: solve_logistic(gd_stochastic, params2, normalize)

Dataset: 0

Best alpha: 0.00325

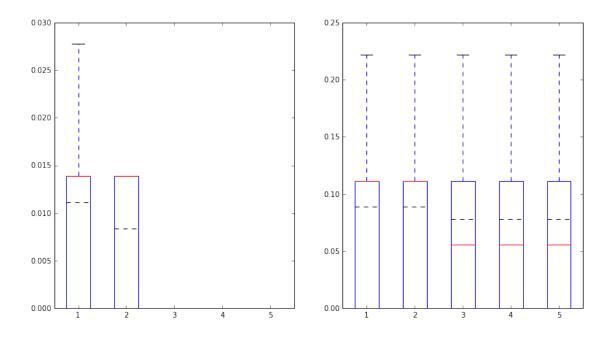


 N° iterations: 22853

Beta: [0.28212691 2.51599989 -6.84871026 -2.01439963 17.09542325

-11.65905654 0.02334595]

Dataset: 1

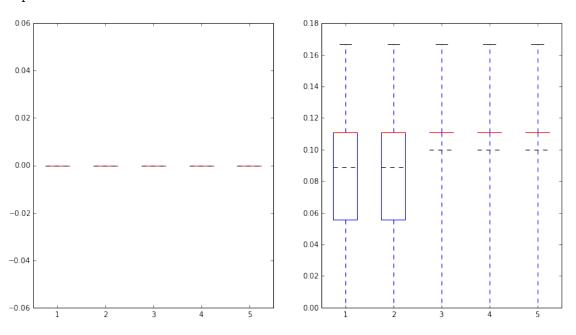


Training error rate: 0.0 Testing error rate: 0.1 $\ensuremath{\mathbb{N}}^\circ$ iterations: 23074

Beta: [-0.83064519 -0.04761609 -11.36013897 -3.3839981 20.45804098

-13.26422916 -0.05551133]

Dataset: 2

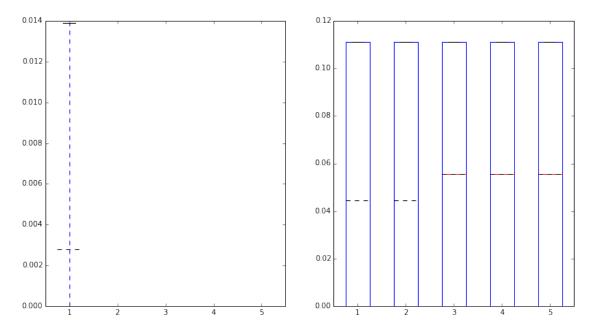


Training error rate: 0.0 Testing error rate: 0.1 $\ensuremath{\mathbb{N}}^\circ$ iterations: 42727

-10.45973575 -0.50162077]

Dataset: 3

Best alpha: 0.001

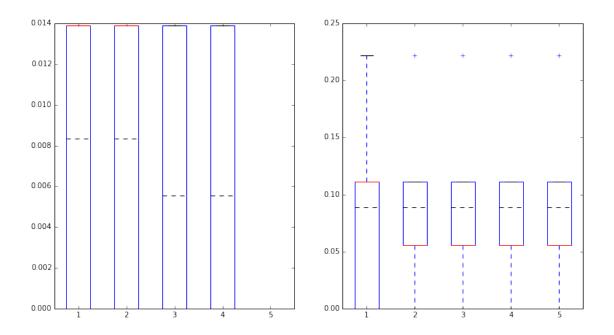


 N° iterations: 30368

Beta: [0.76931706 1.2793493 -5.14824189 0.5091976 13.10813001

-8.03395793 1.39673957]

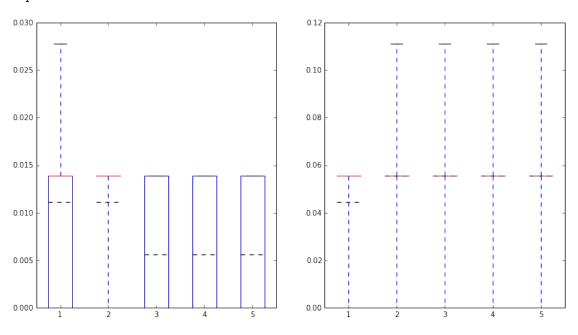
Dataset: 4



 N° iterations: 21107

-7.05492595 -0.27537794]

Dataset: 5



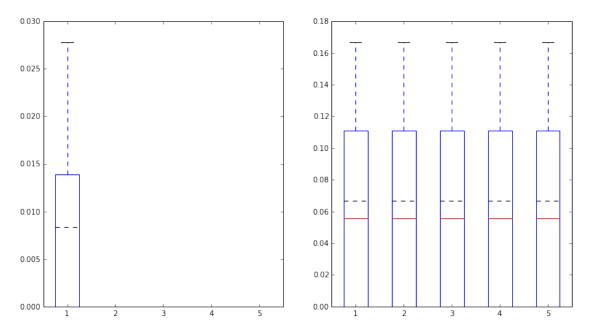
Training error rate: 0.0111111111111

Testing error rate: 0.0 N° iterations: 19506

-6.98938989 0.23329645]

Dataset: 6

Best alpha: 0.001

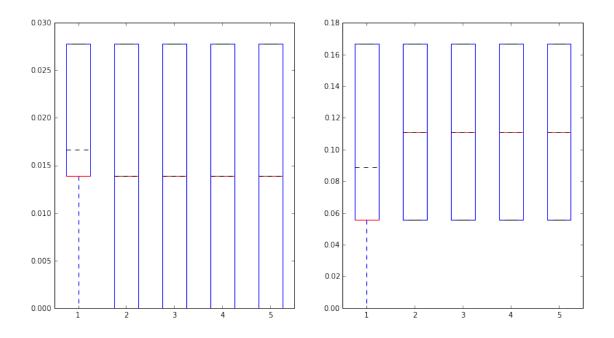


 ${\tt N}^{\circ}$ iterations: 27785

Beta: [0.25973306 -0.43656727 -6.84416509 -1.92523736 11.72510341

-7.99440143 -0.04086211]

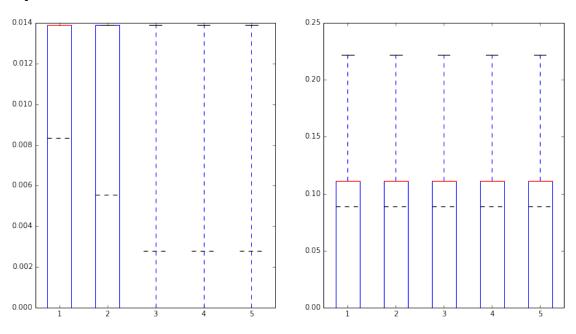
Dataset: 7



 \mbox{N}° iterations: 19145

-7.08214331 0.79262915]

Dataset: 8



Training error rate: 0.0111111111111

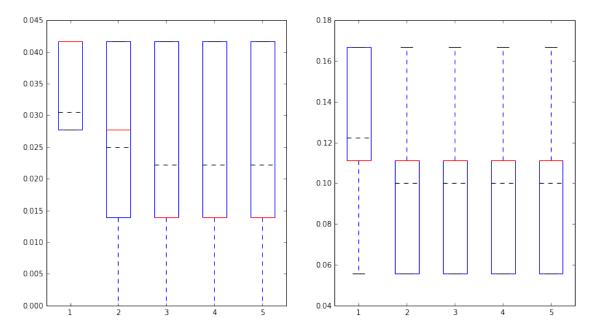
Testing error rate: 0.1 N° iterations: 26466

Beta: [0.98512574 0.20449691 -6.18565185 -2.61863488 11.52072476

-8.55935312 -0.19801622]

Dataset: 9

Best alpha: 0.00325

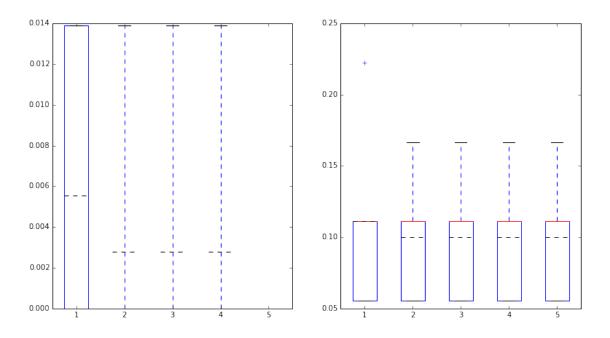


 ${\tt N}^{\circ}$ iterations: 13573

Beta: [1.18086384 1.29180006 -6.2545281 -2.19180191 14.36792718

-8.8891139 -0.10662136]

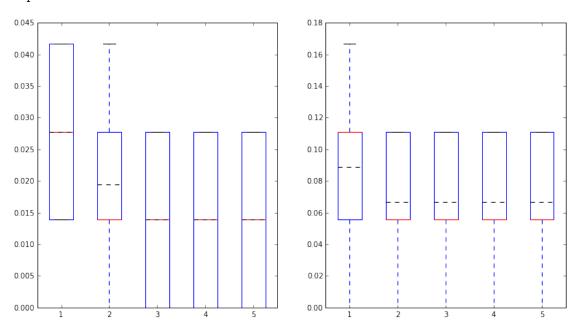
Dataset: 10



 N° iterations: 28782

-11.92779315 -0.57515176]

Dataset: 11

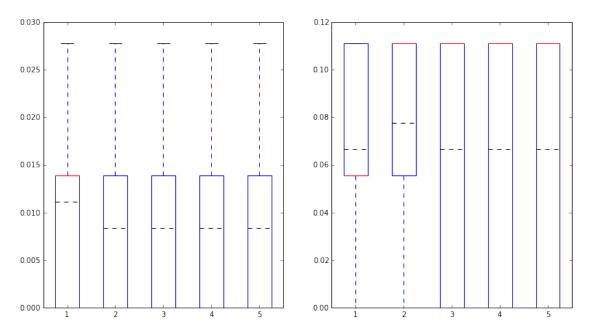


 N° iterations: 12480

-9.02566211 -0.60104034]

Dataset: 12

Best alpha: 0.0055

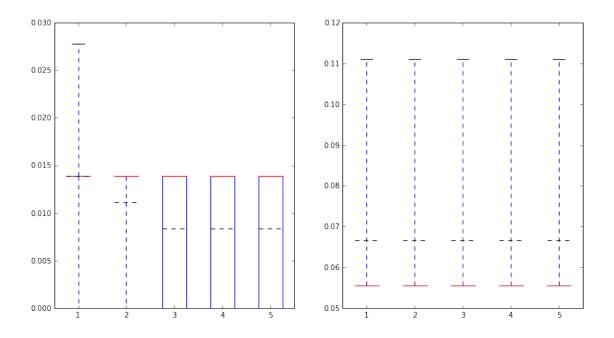


 ${\tt N}^{\circ}$ iterations: 21089

Beta: [1.17501047 1.25021376 -10.52537946 -4.79187833 20.17626635

-13.03576152 -2.04678605]

Dataset: 13

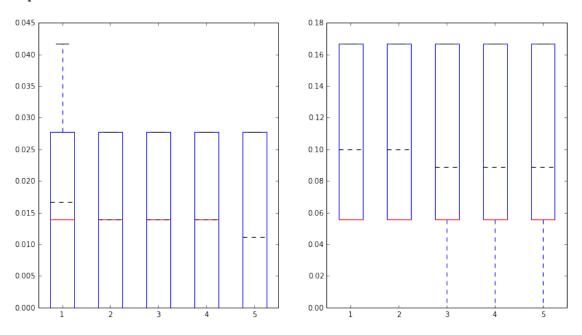


N° iterations: 19371

Beta: [-0.83157681 -0.22984747 -6.32680414 -2.41278289 9.81267399 -6.74299817

-0.35219639]

Dataset: 14

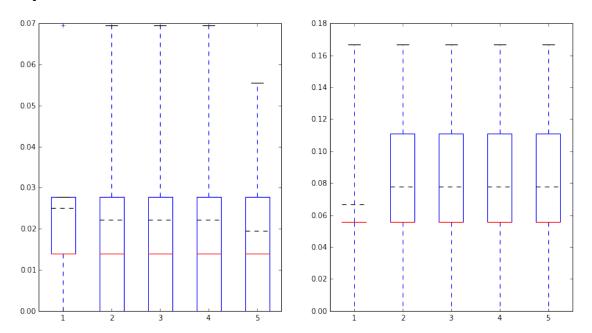


 N° iterations: 12311

Beta: [1.06846176 1.51230049 -7.746944 -2.52735609 16.54763725

-10.51471904 -0.15137984]

Dataset: 15
Best alpha: 0.001

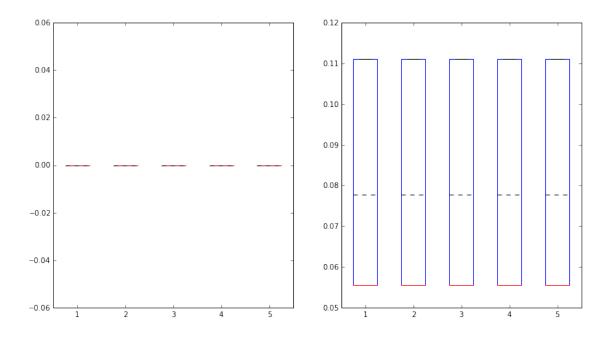


 \mbox{N}° iterations: 17110

 $\texttt{Beta:} \ [\ \texttt{0.59735747} \ \ \texttt{0.3925426} \ \ -4.97515802 \ -1.49478748 \ \ \texttt{9.4856989} \ \ -7.05047378 \\$

0.26428835]

Dataset: 16
Best alpha: 0.001

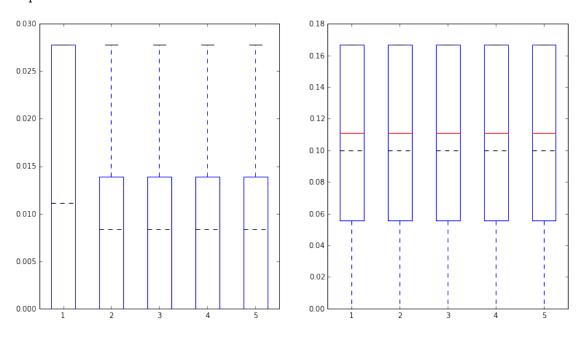


 N° iterations: 32435

Beta: [0.53147909 1.63587046 -5.40451023 -2.65320364 13.0548019

-7.36149183 -0.71556737]

Dataset: 17
Best alpha: 0.001



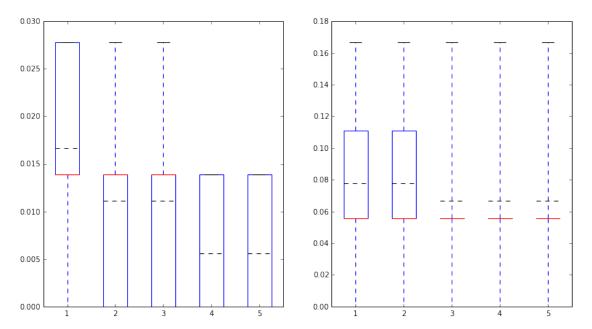
 N° iterations: 20752

Beta: [0.36508413 0.64407219 -5.21690358 -2.25339424 10.55314001

-6.35046728 -1.08978961]

Dataset: 18

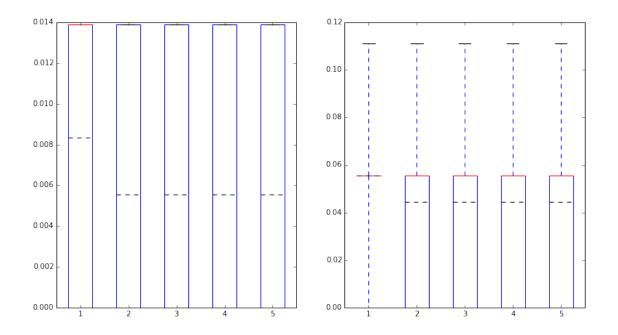
Best alpha: 0.0055



 N° iterations: 15138

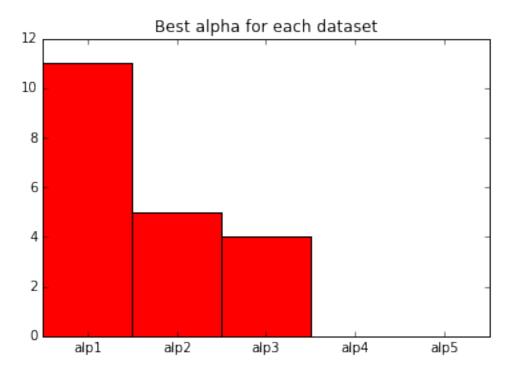
-10.7422816 -1.27091608]

Dataset: 19



 \mbox{N}° iterations: 18115

-9.7774349 -0.67099967]



```
In [89]: #Newton-Raphson method for logistic regression
         def nr_logistic(X, y, alpha, eps=1e-10, max_iter=100000):
             M,N = X.shape
             beta = np.zeros(N)
             11 = 1(X, y, beta)
             for i in xrange(max_iter):
                 10 = 11
                 h = sigmoid(np.dot(X,beta))
                 #print 'beta:',beta
                 print '1:',11
                 #print 'h:',np.max(h)
                 W = np.diag(h*(1-h))
                 Hess = -1.*np.dot(X.T, np.dot(W, X))
                 Dl = np.dot(X.T, y-h)
                 \#when\ it\ converges,\ Hess\ became\ singular
                 try:
                     beta -= alpha*np.linalg.solve(Hess, Dl)
                 except np.linalg.LinAlgError:
                     print 'singular'
                     break
                 11 = 1(X, y, beta)
                 if np.abs(11) < eps:</pre>
                     print 'stopping criterion'
                     break
             print '1:',11
             print 'h', h*(1-h)
             return (beta,i+1)
```