The Perceptron

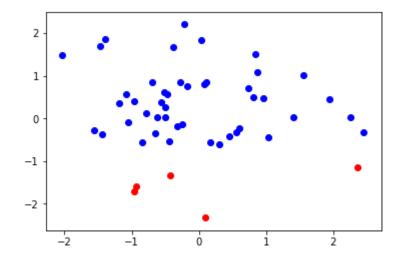
```
In [1]: import mxnet as mx
 from mxnet import nd, autograd
 import matplotlib.pyplot as plt
 import numpy as np
 mx.random.seed(1)
```

A Separable Classification Problem

```
In [2]:
     # generate fake data that is linearly separable with a margin epsilon given the da
     ta
     def getfake(samples, dimensions, epsilon):
         wfake = nd.random normal(shape=(dimensions)) # fake weight vector for separa
     tion
         bfake = nd.random normal(shape=(1))
                                              # fake bias
                                                        # rescale to unit length
         wfake = wfake / nd.norm(wfake)
         # making some linearly separable data, simply by chosing the labels accordingl
         X = nd.zeros(shape=(samples, dimensions))
         Y = nd.zeros(shape=(samples))
         i = 0
         while (i < samples):</pre>
             tmp = nd.random normal(shape=(1,dimensions))
             margin = nd.dot(tmp, wfake) + bfake
             if (nd.norm(tmp).asscalar() < 3) & (abs(margin.asscalar()) > epsilon):
                 X[i,:] = tmp[0]
                 Y[i] = 1 if margin.asscalar() > 0 else -1
                 i += 1
         return X, Y
```

```
In [3]: # plot the data with colors chosen according to the labels
     def plotdata(X,Y):
         for (x,y) in zip(X,Y):
              if (y.asscalar() == 1):
                  plt.scatter(x[0].asscalar(), x[1].asscalar(), color='r')
              else:
                  plt.scatter(x[0].asscalar(), x[1].asscalar(), color='b')
     # plot contour plots on a [-3,3] x [-3,3] grid
     def plotscore(w,d):
         xgrid = np.arange(-3, 3, 0.02)
         ygrid = np.arange(-3, 3, 0.02)
         xx, yy = np.meshgrid(xgrid, ygrid)
         zz = nd.zeros(shape=(xgrid.size, ygrid.size, 2))
         zz[:,:,0] = nd.array(xx)
         zz[:,:,1] = nd.array(yy)
         vv = nd.dot(zz,w) + d
         CS = plt.contour(xgrid,ygrid,vv.asnumpy())
         plt.clabel(CS, inline=1, fontsize=10)
```

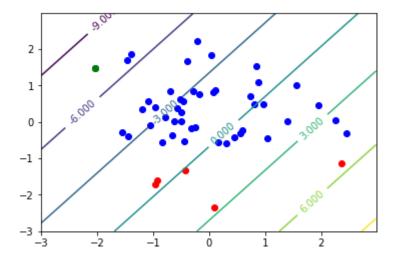
```
In [4]: X, Y = getfake(50, 2, 0.3)
 plotdata(X,Y)
 plt.show()
```



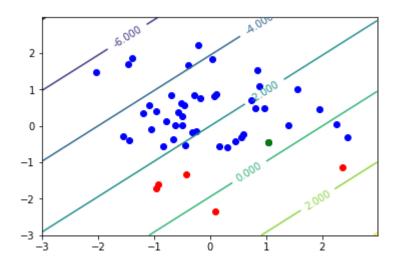
Perceptron Implementation

```
In [5]:
    def perceptron(w,b,x,y):
         if (y * (nd.dot(w,x) + b)).asscalar() <= 0:
             w += v * x
             b += y
             return 1
         else:
             return 0
     w = nd.zeros(shape=(2))
     b = nd.zeros(shape=(1))
     for (x,y) in zip(X,Y):
         res = perceptron(w,b,x,y)
         if (res == 1):
             print('Encountered an error and updated parameters')
             print('data {}, label {}'.format(x.asnumpy(),y.asscalar()))
             print('weight {}, bias {}'.format(w.asnumpy(),b.asscalar()))
             plotscore(w,b)
             plotdata(X,Y)
             plt.scatter(x[0].asscalar(), x[1].asscalar(), color='g')
             plt.show()
```

```
Encountered an error and updated parameters data [-2.0401056 \ 1.482131 \ ], label -1.0 weight [2.0401056 \ -1.482131 \ ], bias -1.0
```

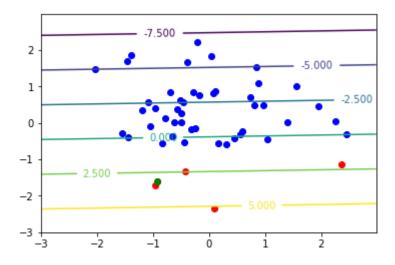


Encountered an error and updated parameters data [1.040828 -0.45256865], label -1.0 weight [0.9992776 -1.0295624], bias -2.0

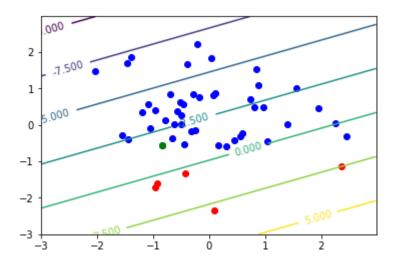


Encountered an error and updated parameters data [-0.934901 -1.5937569], label 1.0

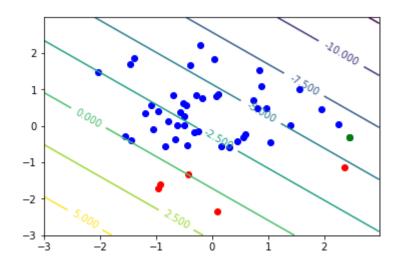
weight [0.06437659 - 2.6233191], bias -1.0



Encountered an error and updated parameters data [-0.84411263 -0.55587363], label -1.0 weight [0.9084892 -2.0674455], bias -2.0



Encountered an error and updated parameters data [2.4434173 -0.31752375], label -1.0 weight [-1.5349281 -1.7499218], bias -3.0



Perceptron Convergence in Action

```
In [7]: plt.plot(Eps, Err, label='average number of updates for training')
plt.legend()
plt.show()
```

