Neural Networks

```
Homework 4
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Delta Rule

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In [82]:
         def hardlim(y_in,train=False):
              if train:
                  return y_in
              else:
                      y_out = [[],[]]
                      for x in range(len(y_in)):
                          if y_in[x] >= 0:
                              y out[x] = 1
                          if y_in[x] < 0:
                              y_out[x] = -1
                      return y_out
         def yin(x vec):
              y in = np.zeros(bias.shape) + bias
              for j in range(len(bias)):
                  y_{in[j]} += np.dot(x_{vec}, w[j])
              return y_in
```

Alpha = 0.001

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In [83]: w=np.zeros(shape=(X.shape[1], y.shape[1]))
          bias=np.zeros(shape=y.shape[1])
          alpha=0.001
          y = y.reshape(y.shape[0],-1)
           if y.shape[1] == 1:
               w = w.reshape(1, w.shape[0])
               bias = np.reshape(bias, -1)
           for i in range(1,10000):
               for x_vec, y_vec in zip(X, y):
                   y_{in} = yin(x_{vec})
                   yy = []
                   for j in range(len(bias)):
                       yy.append(hardlim(y_in[j],train=True))
                       w[j] = w[j] - (2*alpha*np.outer(np.subtract(yy[j],y_vec[j]),x_vec[j])
                       bias[j] = bias[j] - (2*alpha*np.subtract(yy[j],y_vec[j]))
 In [84]: print("Weights: \n", w.transpose())
          print("Bias: \n",bias)
          Weights:
           [[-0.59367939 \quad 0.16801311]
           [-0.05095887 -0.6659139 ]]
          Bias:
           [ 0.01325352  0.1679855 ]
In [101]: count = 0
          target = []
           for i in X:
               yy = i.dot(w) + bias
               y out = hardlim(yy)
               target.append(y_out)
          print("Predicted Output: ",target)
          Predicted Output: [[-1, -1], [-1, -1], [-1, 1], [-1, 1], [1, -1], [1, -
          1], [1, 1], [1, 1]]
In [114]: | if((y==target).all()):
               print("All inputs classified correctly")
          All inputs classified correctly
```

Alpha = 0.1

```
In [87]: w=np.zeros(shape=(X.shape[1], y.shape[1]))
          bias=np.zeros(shape=y.shape[1])
          alpha=0.1
           y = y.reshape(y.shape[0],-1)
           if y.shape[1] == 1:
               w = w.reshape(1, w.shape[0])
               bias = np.reshape(bias, -1)
           for i in range(1,10000):
               for x_vec, y_vec in zip(X, y):
                   y_{in} = yin(x_{vec})
                   yy = []
                   for j in range(len(bias)):
                       yy.append(hardlim(y_in[j],train=True))
                       w[j] = w[j] - (2*alpha*np.outer(np.subtract(yy[j],y vec[j]),x vec[j])
                       bias[j] = bias[j] - (2*alpha*np.subtract(yy[j],y_vec[j]))
 In [88]: print("Weights: \n", w.transpose())
          print("Bias: \n",bias)
          Weights:
           [[-0.39953017 \quad 0.35521879]
           [ 0.06679871 -0.60212805]]
           [ 0.06760018  0.38030401]
In [102]: count = 0
          target = []
           for i in X:
              yy = i.dot(w) + bias
               y out = hardlim(yy)
               target.append(y out)
          print("Predicted Output: ",target)
          Predicted Output: [[-1, -1], [-1, -1], [-1, 1], [-1, 1], [1, -1], [1, -
          1], [1, 1], [1, 1]]
In [105]: | if((y==target).all()):
               print("All inputs classified correctly")
          All inputs classified correctly
```

Part 2: Gradually Decreasing alpha

Alpha = 1/Iteration

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In [91]: w=np.zeros(shape=(X.shape[1], y.shape[1]))
                                bias=np.zeros(shape=y.shape[1])
                                y = y.reshape(y.shape[0],-1)
                                if y.shape[1] == 1:
                                            w = w.reshape(1, w.shape[0])
                                            bias = np.reshape(bias, -1)
                                for i in range(1,10000):
                                            for x_vec, y_vec in zip(X, y):
                                                        alpha=1/i;
                                                        y_{in} = yin(x_{vec})
                                                        yy = []
                                                         for j in range(len(bias)):
                                                                     yy.append(hardlim(y_in[j],train=True))
                                                                     w[j] = w[j] - (2*alpha*np.outer(np.subtract(yy[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j],y_vec[j]),x_vetarrow(yv[j],y_vec[j],y_vec[j],y_vec[j],y_vetarrow(yv[j],y_vec[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vetarrow(yv[j],y_vet
                                                                     bias[j] = bias[j] - (2*alpha*np.subtract(yy[j],y_vec[j]))
  In [97]: print("Weights: \n", w.transpose())
                                print("Bias: \n",bias)
                               Weights:
                                   [[-0.59466015 0.16680423]
                                   [-0.0521512 -0.66659059]]
                               Bias:
                                   [ 0.01308995  0.16680646]
In [103]: | count = 0
                                target = []
                                for i in X:
                                            yy = i.dot(w) + bias
                                            y_out = hardlim(yy)
                                            target.append(y_out)
                                print("Predicted Output: ",target)
                               Predicted Output: [[-1, -1], [-1, -1], [-1, 1], [-1, 1], [1, -1], [1, -
                                1], [1, 1], [1, 1]]
In [104]: if((y==target).all()):
                                            print("All inputs classified correctly")
                               All inputs classified correctly
```

Alpha = alpha*0.9

```
In [111]: | alpha=0.1
          w=np.zeros(shape=(X.shape[1], y.shape[1]))
          bias=np.zeros(shape=y.shape[1])
          y = y.reshape(y.shape[0],-1)
           if y.shape[1] == 1:
               w = w.reshape(1, w.shape[0])
               bias = np.reshape(bias, -1)
           for i in range(1,10000):
               for x_vec, y_vec in zip(X, y):
                   alpha=alpha*0.9;
                   y_{in} = yin(x_{vec})
                   yy = []
                   for j in range(len(bias)):
                       yy.append(hardlim(y_in[j],train=True))
                       w[j] = w[j] - (2*alpha*np.outer(np.subtract(yy[j],y_vec[j]),x_vec[j])
                       bias[j] = bias[j] - (2*alpha*np.subtract(yy[j],y_vec[j]))
          print("Weights: \n", w.transpose())
          print("Bias: \n",bias)
          Weights:
           [[-0.59392236 \quad 0.17147829]
           [-0.02947102 -0.64811857]
          Bias:
           [-0.04044308 \quad 0.16911988]
In [112]: count = 0
          target = []
           for i in X:
              yy = i.dot(w) + bias
               y out = hardlim(yy)
               target.append(y out)
          print("Predicted Output: ",target)
          Predicted Output: [[-1, -1], [-1, -1], [-1, 1], [-1, 1], [1, -1], [1, -
          1], [1, 1], [1, 1]]
In [113]: if((y==target).all()):
               print("All inputs classified correctly")
          All inputs classified correctly
```

Comment on the results:

As we can see, we get better results for weight and bais, hence we get less error with value of alpha nearer to zero.

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In [ ]:
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