## Machine Learning

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## 1 PRACTICA 4: ENTRENAMIENTO DE REDES NEURONALES

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In [3]: import matplotlib.pyplot as pl
        import numpy as np
        from pandas.io.parsers import read_csv
        import scipy.optimize as opt
        #from sklearn.preprocessing import PolynomialFeatures as pf
        from scipy.io import loadmat
        import checkNNGradients
In [4]: def sigmoide(z):
            return 1 / (1 + np.exp(-1*z))
In [5]: def derivadaSigmoide(z):
            return sigmoide(z)*(1-sigmoide(z))
In [6]: def pesosAleatorios(L_in, L_out):
            return np.random.rand(L_out, 1 + L_in) * 0.24 - 0.12
In [7]: def backprop(params_rn, num_entradas, num_ocultas , num_etiquetas , X, y, reg):
            theta1 = np.reshape(params_rn[:num_ocultas*(num_entradas + 1)], (num_ocultas, (num_
            theta2 = np.reshape(params_rn[num_ocultas*(num_entradas + 1):], (num_etiquetas, (num_etiquetas))
            m = len(X)
           #Input
            ones_columns_input = np.array(np.ones(m))
            a1 = np.insert(X, 0,ones_columns_input, axis = 1)
            #hidden_layer
            z2 = np.dot(theta1, a1.transpose())
            a2 = sigmoide(z2)
            one_columns_hidden = np.array(np.ones(m))
            a2 = np.insert(a2, 0, one_columns_hidden, axis = 0)
            #Output_layer
            z3 = np.dot(theta2, a2)
            h = sigmoide(z3)
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y_converted = np.zeros(h.shape)
            for i in range (0,len(X)):
                y_{\text{converted}}[np.ravel(y)[i]-1][i] = 1
             #Cost
            regulation = (reg/(2*m)) * (np.sum(theta1**2) + np.sum(theta2**2))
            J = np.sum(-y\_converted * np.log(h) - (1 - y\_converted)*np.log(1 - h)) * (1/m)
            J_regulated = J + regulation
            # Retro-Propagation
            d3 = h - y_converted
            z2 = np.insert(z2, 0, np.ones(m), axis = 0)
            z2prima = derivadaSigmoide(z2)
            d2 = (np.dot(theta2.transpose(), d3))*z2prima
            #Gradient
            delta2 = np.dot(d3,a2.transpose())
            delta1 = np.dot(d2, a1)
            #Regularization
            D1 = (delta1[1:,:]/m + theta1*reg/m).ravel()
            D2 = (delta2/m + theta2*reg/m).ravel()
            gradient = np.r_[D1, D2]
            return J_regulated, gradient
In [8]: def test():
            num_entradas = 400
            num_ocultas = 25
            num_etiquetas = 10
            reg = 1
            data = loadmat('ex4data1.mat')
            y = data ['y']
            X = data ['X']
            weights = loadmat( 'ex4weights.mat' )
            theta1, theta2 = weights[ 'Theta1' ], weights[ 'Theta2' ]
            coste, gradiente = backprop(np.concatenate((np.ravel(theta1), np.ravel(theta2))), :
            checkNNGradients.checkNNGradients(backprop, reg)
In [9]: test()
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