Deep Learning Assignment 2

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1 Introduction

This assignment was completed using the Keras library by constructing a perceptron with two hidden layers, each containing 16 neurons. The model is able to reliably classify its training data set with perfect accuracy and performs well on unfamiliar data. The form of the perceptron function can be seen in Figure 1; its outline was generated by plotting a selection of points from the X/Y plane on a contour plot.

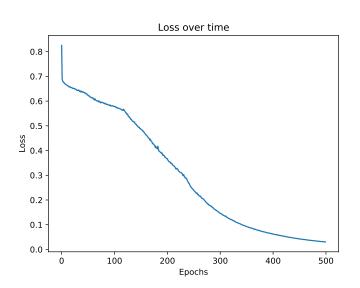


Figure 2: Loss over training iterations

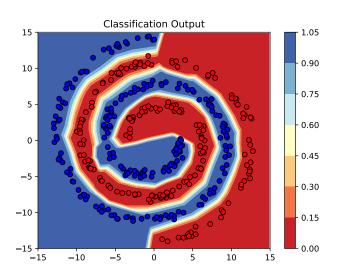


Figure 1: Output of multilayer perceptron classifier

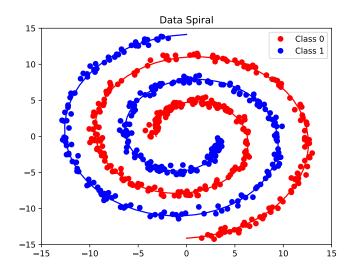


Figure 3: Starting Data

2 Code

```
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
import numpy as np
import matplotlib.pyplot as plt
def pol2cart(r,th):
    x = r * np.cos(th)
    y = r * np. sin(th)
    return x, y
\mathbf{def} \ \mathbf{gen}_{-}\mathbf{data}(\mathbf{n}, \ \mathbf{noise} = 0.0):
    th = np.random.uniform(-np.pi, -9/2 * np.pi, size = [n, 1])
    cl = np.random.randint(0,2,size=[n,1])
    fuzz = np.random.normal(scale=noise, size=[n,2])
    return np. hstack (pol2cart (th*(2*cl-1), th)) + fuzz, cl
#generate a clean spiral
th = np.linspace(-np.pi, -9/2 * np.pi, num=100)
plt.plot(*pol2cart(th,th), 'b')
plt.plot(*pol2cart(-th,th), 'r')
#generate and plot some noisy data
[x_{test}, y_{test}] = gen_{data}(600, .3)
y_test = y_test.T[0]
plt.plot(x_test[y_test==0,0], x_test[y_test==0,1], 'ro', label='Class_0')
plt.plot(x_test[y_test==1,0], x_test[y_test==1,1], 'bo', label='Class_1')
plt.title("Data_Spiral")
plt.xlim([-15,15]); plt.ylim([-15,15])
plt.legend()
plt.savefig('data.pdf', format='pdf')
\#plt.show()
plt.clf()
#tunable parameters
num_data = 1000
noise\_level = 0.3
num_{-epochs} = 500
#create training data
[x,y] = gen_data(num_data, noise_level)
#generate keras model
model = keras. Sequential (
        layers. Dense (16, input_dim=2, activation="relu", name="hidden1"),
        layers. Dense (16, input_dim=2, activation="relu", name="hidden2"),
        layers.Dense(1, activation='sigmoid', name="out", kernel_regularizer='12'),
```

```
\#run \mod el
model.compile(loss='binary_crossentropy', optimizer='Adam', metrics=['accuracy'])
fit_data = model.fit(x,y,epochs=num_epochs,batch_size=10,verbose=0)
#loss statistics
model.summary()
plt.plot(fit_data.history['loss'])
plt.title("Loss_over_time")
plt.xlabel('Epochs'); plt.ylabel('Loss')
plt.savefig('loss.pdf', format='pdf')
plt.show()
plt.clf()
#Check that training data is classified correctly
results = model.evaluate(x, y, batch_size=10)
print("training loss, test acc:", results)
\#result plot
[x_{check}, y_{check}] = gen_{data}(400, noise_{level})
y_{thresh} = np.round(model.predict(x_check).T[0])
plt.plot(x_check[y_thresh==0,0], x_check[y_thresh==0,1], 'ro', markeredgecolor='black')
plt.plot(x_check[y_thresh==1,0], x_check[y_thresh==1,1], 'bo', markeredgecolor='black',)
\#plt.show()
x = np.arange(-15, 15, 0.1)
y = np.arange(-15, 15, 0.1)
xx, yy = np.meshgrid(x, y)
xxx = xx. reshape(-1,1)
yyy = yy. reshape(-1,1)
inp = np.hstack((xxx, yyy))
z = model.predict(inp)
plt.contourf(xx,yy,z.reshape(xx.shape), cmap='RdYlBu')
plt.colorbar()
plt.title("Classification_Output")
plt. xlim ([-15,15]); plt. ylim ([-15,15])
plt.savefig('result.pdf', format='pdf')
plt.show()
```