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# Purdue University Global
# IN400 - AI: Deep Learning and Machine Learning
# Unit 2 Assignment Part 1 / Module 2 Competency Assessment Part 1
# Exploring Machine Learning Tools With TensorFlow
# PyCharm Code
\mbox{\tt\#} NOTE: Ignore any "could not load dynamic library libcudart \dots "
# warnings produced during execution. The Codio environment does
# not have GPU capabilites.
# Import the applicable libraries and set the default parameters.
# If the libraries do not exist, they must be installed in the Python environment.
from pandas import read_csv
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras import Sequential
from tensorflow.keras.layers import Dense
# Data File Location
dataFile = '/home/codio/workspace/data/IN400/ionosphere.data'
# Output Header
print('\nUnit 2 Assignment Part 1 / Module 2 Competency Assessment Part 1\n')
from datetime import datetime
print(datetime.now().strftime("%m/%d/%Y %H:%M:%S"), '\n')
# Load the dataset from the specified location
df = read csv(dataFile, header=None)
# Peek at the data
print("DATA SNIPPET")
print(df.head())
print()
# Explore the data
print("DATA STATISTICS")
print(df.describe())
print()
# Split the data frame into input (features) and output (Class) columns
X = df.values[:, :-1]
y = df.values[:, -1]
# Make sure the values of all features are of type float
X = X.astype('float32')
# Change the Class column from strings to integer - good: g to 1 and bad: b to 0
y = LabelEncoder().fit_transform(y)
# Split the dataset into an 80% training set and a 20% test.
# The split happens for both the feature part and the class part
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
print("ARRAY SHAPES: TRAINING X-DATA / TEST X-DATA / TRAINING Y-DATA / TEST Y-DATA")
print(X_train.shape, X_test.shape, y_train.shape, y_test.shape)
print()
# Get the number of feature columns
n_features = X_train.shape[1]
# Define the parameters of the model
model = Sequential()
model.add(Dense(10, activation='relu', kernel_initializer='he_normal', input_shape=(n_features,)))
model.add(Dense(8, activation='relu', kernel initializer='he normal'))
model.add(Dense(1, activation='sigmoid'))
# Compile the model
model.compile(optimizer='adam', loss='binary crossentropy', metrics=['accuracy'])
# Fit the model
model.fit(X_train, y_train, epochs=150, batch_size=32, verbose=0)
# Evaluate the model
Accuracy = model.evaluate(X_test, y_test, verbose=0)
print()
print('Model Accuracy: %.3f' % Accuracy[1])
print()
# Test prediction by creating a new row that covers all the feature columns
\texttt{row} = [1,0,0.99539,-0.05889,0.85243,0.02306,0.83398,-0.37708,1,0.03760,0.85243,-0.17755,0.59755,-0.44945,0.60536,0.83398,-0.85243,0.03760,0.85243,-0.17755,0.59755,-0.44945,0.60536,0.83398,-0.85243,0.03760,0.85243,-0.17755,0.59755,-0.44945,0.60536,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.85243,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.03760,0.
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                   0.21266, -0.34090, 0.42267, -0.54487, 0.18641, -0.45300
TestPred = model.predict([row])
print('Test Prediction: %.3f' % TestPred)
```