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In [ ]: from numpy import loadtxt
       from tensorflow.keras.models import Sequential
       from tensorflow.keras.layers import Dense
In [ ]: | dataset = loadtxt(r"C:\Users\92341\Downloads\pima-indians-diabetes.data.csv", delimiter=',')
       # split into input (X) and output (y) variables
       X = dataset[:,0:8]
       y = dataset[:,8]
       print(X)
       print(y)
                           ... 33.6
                                       0.627 50.
      [[ 6.
              148.
                     72.
                                                  1
                          ... 26.6
      [ 1.
               85.
                     66.
                                       0.351 31.
                                                  -1
                         ... 23.3
              183.
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                     64.
                                                 ]
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                     72.
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                                       0.245 30.
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                           ... 30.1
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In [ ]: |# define the keras model
       model = Sequential()
       model.add(Dense(12, input_shape=(8,), activation='relu'))
       model.add(Dense(8, activation='relu'))
       model.add(Dense(1, activation='sigmoid'))
In [ ]: | model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
In [ ]: model.fit(X, y, epochs=10, batch_size=10)
      Epoch 1/10
      77/77 [===
                       Epoch 2/10
      77/77 [===
                        =========] - 0s 2ms/step - loss: 0.5473 - accuracy: 0.7044
      Epoch 3/10
      77/77 [==========] - 0s 1ms/step - loss: 0.5803 - accuracy: 0.6836
      Epoch 5/10
      77/77 [==========] - 0s 1ms/step - loss: 0.5631 - accuracy: 0.7188
      Epoch 6/10
      Epoch 7/10
      77/77 [==========] - 0s 1ms/step - loss: 0.5473 - accuracy: 0.7161
      Epoch 8/10
      Fnoch 9/10
      77/77 [==========] - 0s 1ms/step - loss: 0.5457 - accuracy: 0.7292
      Epoch 10/10
      Out[]: <keras.src.callbacks.History at 0x250e671ad50>
In [ ]: # evaluate the keras model
       _, accuracy = model.evaluate(X, y)
       print('Accuracy: %.2f' % (accuracy*100))
```

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Accuracy: 73.31
In [ ]: # make probability predictions with the model
       predictions = model.predict(X)
       # round predictions
       rounded = [round(x[0]) for x in predictions]
      24/24 [========] - 0s 1ms/step
In [ ]: # make class predictions with the model
       predictions = (model.predict(X) > 0.5).astype(int)
      24/24 [=======] - 0s 1ms/step
In [ ]: for i in range(5):
       print('%s => %d (expected %d)' % (X[i].tolist(), predictions[i], y[i]))
      [6.0, 148.0, 72.0, 35.0, 0.0, 33.6, 0.627, 50.0] => 1 (expected 1)
      [1.0, 85.0, 66.0, 29.0, 0.0, 26.6, 0.351, 31.0] => 0 (expected 0)
      [8.0, 183.0, 64.0, 0.0, 0.0, 23.3, 0.672, 32.0] => 1 (expected 1)
      [1.0, 89.0, 66.0, 23.0, 94.0, 28.1, 0.167, 21.0] => 0 (expected 0)
      [0.0, 137.0, 40.0, 35.0, 168.0, 43.1, 2.288, 33.0] => 0 (expected 1)
```