

Figure 30-11. Iris dataset - neural network architecture

Using the Keras Sequential API

This code segment will construct a neural network model with the Sequential API using the method 'tf.keras.Sequential()' to stack layers on each other. The model creates a hidden layer with 32 neurons and an output layer with 3 output units because the Iris target contains 3 classes.

```
!pip install -q tensorflow==2.0.0-beta0

# import packages
import tensorflow as tf
import pandas as pd
from sklearn.preprocessing import OneHotEncoder

# dataset url
train_data_url = "https://storage.googleapis.com/download.tensorflow.org/
data/iris_training.csv"
test_data_url = "https://storage.googleapis.com/download.tensorflow.org/
data/iris_test.csv"
```

```
# define column names
columns = ['sepal length', 'sepal width', 'petal length', 'petal width', 'species']
# download and load the csv files
train data = pd.read csv(tf.keras.utils.get file('iris train.csv', train
data url),
                               skiprows=1, header=None, names=columns)
test data = pd.read csv(tf.keras.utils.get file('iris test.csv', test data url),
                               skiprows=1, header=None, names=columns)
# separate the features and targets
(X train, y train) = (train data.iloc[:,0:-1], train data.iloc[:,-1])
(X test, y test) = (test data.iloc[:,0:-1], test data.iloc[:,-1])
# apply one-hot encoding to targets
y train=tf.keras.utils.to categorical(y train)
y test=tf.keras.utils.to categorical(y test)
# create the sequential model
def model fn():
    model = tf.keras.Sequential()
    # Add a densely-connected layer with 32 units to the model:
    model.add(tf.keras.layers.Dense(32, activation='sigmoid', input dim=4))
    # Add a softmax layer with 3 output units:
    model.add(tf.keras.layers.Dense(3, activation='softmax'))
    # compile the model
    model.compile(optimizer=tf.keras.optimizers.SGD(),
                    loss='categorical crossentropy',
                    metrics=['accuracy'])
    return model
# parameters
batch size=50
```

```
# use tf.data to batch and shuffle the dataset
train ds = tf.data.Dataset.from tensor slices(
    (X train.values, y train)).shuffle(len(X train)).repeat().batch(batch size)
test ds = tf.data.Dataset.from tensor slices((X_test.values, y_test)).
batch(batch size)
# build train model
model = model fn()
# print train model summary
model.summary()
# train the model
history = model.fit(train ds, steps per epoch=5000)
# evaluate the model
score = model.evaluate(test ds)
print('Test loss: {:.2f} \nTest accuracy: {:.2f}%'.format(score[0],
score[1]*100))
'Output':
Test loss: 0.22
Test accuracy: 96.67%
```

Using the Keras Functional API

The general code pattern for the Functional API is structurally the same as the Sequential version. The only change here is in how the network model is constructed. We also demonstrated the Keras feature for printing the graph of the model in this example. The output is illustrated in Figure 30-12.

```
!pip install -q tensorflow==2.0.0-beta0

# import packages
import tensorflow as tf
import pandas as pd
from sklearn.preprocessing import OneHotEncoder
```