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Keras Model Import into DeepLearning4J

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Getting Help

- Join us on Gitter Chat
 - <https://gitter.im/deeplearning4j/deeplearning4j>
 - Email Tom@skymind.io
-

Training Information

- <https://skymind.ai/academy>

LABS

- The Rest of the PDF is Lab work that you can complete at home.
 - Questions:
 - tom@skymind.io
-

Lab Prep

1. Step One

```
git clone https://github.com/tomthetrainer/KerasWorkshop.git
```

1. Step two

Import the project into IntelliJ by following instructions at <https://deeplearning4j.org/quickstart>

1. Additional Resources

Two saved Keras Model for VGG-16 and the DL4J version are too large to put into the main github repo. I have included them on github by creating a release for the project.

Although there is no specific Lab to load those models, there are examples in the Solution project that show loading these large useful models for image recognition.

Get the saved models from the following url

<https://github.com/tomthetrainer/KerasWorkshop/releases>

vgg16.zip

vgg_combined_save.h5

The Labs expect those to be in the /tmp directory. Either place them there or modify the java code that refers to the file path.

1. Structure of the Lab Repo.

Navigate to src/main/java/ai.skymind.training and you will find 3 projects

- demos
- labs
- solutions

labs

You will work through the Lab detailed later in this document in the labs project. A stub file is prepared for you and the instructions will walk you through completing the code.

demos

Some demos are available in the demo project. These are complete code examples.

1. ImageDemo demonstrates simple image into INDArray using NativeImageLoder
2. IrisNoImport is a working example of classification of iris flower data
3. Simplest network is a very simple network that takes a single value input and trains towards a single value output. It also shows the Web based User Interface.
4. VGG16SparkJavaWebApp, takes network that was imported from Keras into DL4J and builds web app to run inference on user selected image.

solutions

Working solutions to the labs and a few other included examples.

Keras Model Import Lab

Introduction

- Keras
 - Popular Python tool for Deep Learning
 - DeepLearning4J
 - Production worthy Java based Deep Learning
 - Keras Model Import
 - Best of both worlds
-

A Keras Model

- [resources/Keras/iris.py](#)

```
import numpy
import pandas
from keras.models import Sequential
from keras.layers import Dense
from keras.wrappers.scikit_learn import KerasClassifier
from keras.utils import np_utils
from sklearn.cross_validation import cross_val_score, KFold
from sklearn.preprocessing import LabelEncoder
from sklearn.pipeline import Pipeline

# fix random seed for reproducibility
seed = 7
numpy.random.seed(seed)

# load dataset
dataframe = pandas.read_csv("iris.csv", header=None)
dataset = dataframe.values
X = dataset[:,0:4].astype(float)
Y = dataset[:,4]
print(X)
print(Y)

#encode class values as integers
encoder = LabelEncoder()
encoder.fit(Y)
encoded_Y = encoder.transform(Y)

# convert integers to dummy variables (hot encoded)
dummy_y = np_utils.to_categorical(encoded_Y)
print(dummy_y)

# define baseline model
#def baseline_model():
# create model
model = Sequential()
model.add(Dense(4, input_dim=4, activation='relu'))
model.add(Dense(3,activation='sigmoid'))

# Compile model
model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

```
#    return model
#model.fit
model.fit(X, dummy_y, nb_epoch=200, batch_size=5)
prediction = model.predict(numpy.array([[4.6,3.6,1.0,0.2]]));
print(prediction);

# To casve just the weights
model.save_weights('/tmp/iris_model_weights')

# To save the weights and the config
# Note this is what is used for this demo
model.save('/tmp/full_iris_model')

# To save the Json config to a file
json_string = model.to_json()
text_file = open("/tmp/iris_model_json", "w")
text_file.write(json_string)
text_file.close()
```

Iris DataSet

- Flower data
 - resources/Keras/iris.csv

```
5.1,3.5,1.4,0.2,Iris-setosa
4.9,3.0,1.4,0.2,Iris-setosa
6.1,2.8,4.0,1.3,Iris-versicolor
6.3,2.5,4.9,1.5,Iris-versicolor
7.1,3.0,5.9,2.1,Iris-virginica
6.3,2.9,5.6,1.8,Iris-virginica
```

Saving a Keras Model

```
model.save('/tmp/full_iris_model')
``
```

```
<div style="page-break-after: always;"></div>
```

```
# Lab Step 1
```

```
In this lab we have run the python Keras code and the saved model and weights are
stored in ``` resources/Keras/full_iris_model
```

Create a String for the filepath to the full_iris_model file.

We use DataVec's ClassPathResource here to get information on files in our Classpath by providing a name of the file. The String for the classpath could be provided directly, we use ClassPathResource so we can drop a file in the resources directory and have a portable demonstration environment.

```
String kerasModelfromKerasExport = new ClassPathResource("Keras/full_iris_model").
getFile().getPath();
```

Lab Step 2

Create a MultiLayerNetwork by using

`KerasModelImport.importKerasSequentialModelAndWeights` and providing a String for the FilePath to the saved Keras Model.

```
MultiLayerNetwork model = KerasModelImport.importKerasSequentialModelAndWeights(k
erasModelfromKerasExport);
```

The required import for this step is:

```
import org.deeplearning4j.nn.modelimport.keras.KerasModelImport;
```

Lab Step 3

Add some code to verify the model import was successful. In this step you create the Input Array to test with.

The Keras model was trained and then tested with a single input.

```
prediction = model.predict(numpy.array([[4.6,3.6,1.0,0.2]]))
```

The output showed high probability for Class 1, less for class 2, and even less for class 3.

```
[[ 0.92084521  0.13397516  0.03294737]]
```

Add this code to the class.

```
INDArray myArray = Nd4j.zeros(1, 4); // one row 4 column array
myArray.putScalar(0,0, 4.6); // first value sepal length
myArray.putScalar(0,1, 3.6); // second value sepal width
myArray.putScalar(0,2, 1.0); // third value petal length
myArray.putScalar(0,3, 0.2); // fourth value petal width
```

Lab Step 4

Validate the model is working and results match the Keras model.

Add this to the class.

```
INDArray output = model.output(myArray);
System.out.println("First Model Output");
System.out.println(myArray);
System.out.println(output);
```

Lab Step 5

Run your code and verify the output.

You should see on the console output the following values.

```
[4.60, 3.60, 1.00, 0.20]  
[0.92, 0.13, 0.03]
```

The first line is our input INDArray The second line is our output.

Does it match the prediction of the model when run in Keras?

Congratulations you have imported a Keras Model into DeepLearning4J.
