

1. Download all the data in this folder <https://drive.google.com/open?id=1Z4TyI7FcFVEx8qdl4j09qxvxaqLSqoEu>. it contains two file both images and labels. The label file list the images and their categories in the following format:

path/to/the/image.tif,category

where the categories are numbered 0 to 15, in the following order:

- 0 letter**
- 1 form**
- 2 email**
- 3 handwritten**
- 4 advertisement**
- 5 scientific report**
- 6 scientific publication**
- 7 specification**
- 8 file folder**
- 9 news article**
- 10 budget**
- 11 invoice**
- 12 presentation**
- 13 questionnaire**
- 14 resume**
- 15 memo**

2. On this image data, you have to train 3 types of models as given below. You have to split the data into Train and Validation data.

3. Try not to load all the images into memory, use the generators that we have given the reference notebooks to load the batch of images only during the training data.

or you can use this method also

<https://medium.com/@vijayabhaskar96/tutorial-on-keras-imagedatagenerator-with-flow-from-dataframe-8bd5776e45c1> (<https://medium.com/@vijayabhaskar96/tutorial-on-keras-imagedatagenerator-with-flow-from-dataframe-8bd5776e45c1>).

<https://medium.com/@vijayabhaskar96/tutorial-on-keras-flow-from-dataframe-1fd4493d237c> (<https://medium.com/@vijayabhaskar96/tutorial-on-keras-flow-from-dataframe-1fd4493d237c>).

4. You are free to choose Learning rate, optimizer, loss function, image augmentation, any hyperparameters. but you have to use the same architecture what we are asking below.

5. Use tensorboard for every model and analyse your gradients. (you need to upload the screenshots for each model for evaluation)

Note: `fit_generator()` method will have problems with the tensorboard histograms, try to debug it, if you could not do use `histgrams=0` i.e don't include histograms, check the documentation of tensorboard for more information.

6. You can check about Transfer Learning in this link - <https://blog.keras.io/building-powerful-image-classification-models-using-very-little-data.html> (<https://blog.keras.io/building-powerful-image-classification-models-using-very-little-data.html>).

Model-1

1. Use [VGG-16](https://www.tensorflow.org/api_docs/python/tf/keras/applications/VGG16) (https://www.tensorflow.org/api_docs/python/tf/keras/applications/VGG16) pretrained network without Fully Connected layers and initialize all the weights with Imagenet trained weights.
2. After VGG-16 network without FC layers, add a new Conv block (1 Conv layer and 1 Maxpooling), 2 FC layers and a output layer to classify 16 classes. You are free to choose any hyperparameters/parameters of conv block, FC layers, output layer.
3. Final architecture will be **INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> Maxpool Layer --> 2 FC layers --> Output Layer**
4. Train only new Conv block, FC layers, output layer. Don't train the VGG-16 network.



Model-2

1. Use [VGG-16](https://www.tensorflow.org/api_docs/python/tf/keras/applications/VGG16) (https://www.tensorflow.org/api_docs/python/tf/keras/applications/VGG16) pretrained network without Fully Connected layers and initialize all the weights with Imagenet trained weights.
2. After VGG-16 network without FC layers, don't use FC layers, use conv layers only as Fully connected layer. any FC layer can be converted to a CONV layer. This conversion will reduce the No of Trainable parameters in FC layers. For example, an FC layer with K=4096 that is looking at some input volume of size 7x7x512 can be equivalently expressed as a CONV layer with F=7,P=0,S=1,K=4096. In other words, we are setting the filter size to be exactly the size of the input volume, and hence the output will simply be 1x1x4096 since only a single depth column “fits” across the input volume, giving identical result as the initial FC layer. You can refer [this](http://cs231n.github.io/convolutional-networks/#convert) (<http://cs231n.github.io/convolutional-networks/#convert>) link to better understanding of using Conv layer in place of fully connected layers.
3. Final architecture will be VGG-16 without FC layers(without top), 2 Conv layers identical to FC layers, 1 output layer for 16 class classification. **INPUT --> VGG-16 without Top layers(FC) --> 2 Conv Layers identical to FC --> Output Layer**
3. Train only last 2 Conv layers identical to FC layers, 1 output layer. Don't train in the VGG-16 network.

Model-3

1. Use same network as Model-2 'INPUT --> VGG-16 without Top layers(FC) --> 2 Conv Layers identical to FC --> Output Layer' and train only Last 6 Layers of VGG-16 network, 2 Conv layers identical to FC layers, 1 output layer.

In []:

```
1 import matplotlib.pyplot as plt # importing the libraries
2 import pandas as pd
3 import numpy as np
4 import seaborn as sns
5 import tensorflow as tf
6 import datetime, os
7 from tensorflow import keras
8 from keras.models import Model
```

/usr/local/lib/python3.6/dist-packages/statsmodels/tools/_testing.py:19: FutureWarning: pandas.util.testing is deprecated. Use the functions in the public API at pandas.testing instead.
import pandas.util.testing as tm
Using TensorFlow backend.

In []:

```
1 !wget --header="Host: doc-0k-3k-docs.googleusercontent.com" --header="User-Agent: Mozilla/5.0 (X11; Linux x86_64; rv:68.0) Gecko/20100101 Firefox/68.0"
```

```
--2020-06-28 05:06:02-- https://doc-0k-3k-docs.googleusercontent.com/docs/securesc/8h8uvuh5ifib89027b7ihknt22nu5vgc/pb2qojp66b5mdedm2vf2dtj92nda0jem/1593320700000/00484516897554883881/05866892802988797180/1Z4TyI7FcFVEx8qd14j09qxvxaqLSqEu?e=download&authuser=0&nonce=cvcfvs0ct36a2&user=05866892802988797180&hash=ilt33fv0qrj4td17qaeas607uq6g796v (https://doc-0k-3k-docs.googleusercontent.com/docs/securesc/8h8uvuh5ifib89027b7ihknt22nu5vgc/pb2qojp66b5mdedm2vf2dtj92nda0jem/1593320700000/00484516897554883881/05866892802988797180/1Z4TyI7FcFVEx8qd14j09qxvxaqLSqEu?e=download&authuser=0&nonce=cvcfvs0ct36a2&user=05866892802988797180&hash=ilt33fv0qrj4td17qaeas607uq6g796v)
Resolving doc-0k-3k-docs.googleusercontent.com (doc-0k-3k-docs.googleusercontent.com)... 74.125.204.132, 2404:6800:4008:c04::84
Connecting to doc-0k-3k-docs.googleusercontent.com (doc-0k-3k-docs.googleusercontent.com)|74.125.204.132|:443... connected.
HTTP request sent, awaiting response... 200 OK
Length: unspecified [application/rar]
Saving to: 'rvl-cdip.rar'
```

```
rvl-cdip.rar [====>] 4.34G 39.5MB/s in 85s
```

```
2020-06-28 05:07:27 (52.2 MB/s) - 'rvl-cdip.rar' saved [4660541790]
```

In []:

```
1 get_ipython().system_raw("unrar x rvl-cdip.rar") # extracting the uploaded file
```

In []:

```
1 df=pd.read_csv("labels_final.csv")
```

In []:

```

1 labels_dict={ 0 : "letter",
2               1 : "form",
3               2 : "email",
4               3 : "handwritten",
5               4 : "advertisement",
6               5 : "scientific report",
7               6 : "scientific publication",
8               7 : "specification",
9               8 : "file folder",
10              9 : "news article",
11             10 : " budget",
12             11 : "invoice",
13             12 : " presentation",
14             13 : "questionnaire",
15             14 : "resume",
16             15 : "memo"}

```

In []:

```
1 df['label']=df['label'].apply(lambda x:labels_dict[x])
```

In []:

```
1 df.head(5)
```

Out[7]:

	path	label
0	imagesv/v/o/h/voh71d00/509132755+-2755.tif	handwritten
1	imagesl/l/x/t/lxt19d00/502213303.tif	handwritten
2	imagesx/x/e/d/xed05a00/2075325674.tif	email
3	imageso/o/j/b/ojb60d00/517511301+-1301.tif	handwritten
4	imagesq/q/z/k/qzk17e00/2031320195.tif	specification

In []:

```

1 from keras_preprocessing.image import ImageDataGenerator
2 datagen = ImageDataGenerator(rescale=1/255., validation_split=0.2) #image generator

```

In []:

```

1 print("-----TRAIN DATA-----") # train data
2 train_generator = datagen.flow_from_dataframe(dataframe=df, directory="/content/data_f",
3                                               x_col='path',
4                                               y_col='label', # using flow from data frame
5                                               target_size=(256,256),
6                                               class_mode='categorical',
7                                               batch_size=32,
8                                               subset='training',
9                                               seed=7)

```

-----TRAIN DATA-----

Found 38400 validated image filenames belonging to 16 classes.

In []:

```

1 print("-----CROSS VALIDATION DATA-----") # cross validation data
2 validation_generator = datagen.flow_from_dataframe(dataframe=df, directory="/content/d
3                                     x_col='path',
4                                     y_col='label',
5                                     target_size=(256,256),
6                                     class_mode='categorical',
7                                     batch_size=32,
8                                     subset='validation',
9                                     seed=7)

```

-----CROSS VALIDATION DATA-----

Found 9600 validated image filenames belonging to 16 classes.

In []:

```

1 from keras.layers import Input, Lambda, Dense, Flatten
2 from keras.models import Model
3 from keras.applications.vgg16 import VGG16
4 from keras.applications.vgg16 import preprocess_input
5 from keras.preprocessing import image
6 from keras.layers import Dense, Conv2D, MaxPool2D , Flatten
7 from keras.callbacks import Callback
8 from keras.callbacks import TensorBoard

```

In []:

```

1 %load_ext tensorboard

```

In []:

```

1 logdir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S") # tensorboard
2 tensorboard_callback = TensorBoard(log_dir=logdir, histogram_freq=1)

```

In []:

```

1 IMAGE_SIZE = [256, 256] #pre trained vgg16 model
2 model = VGG16(input_shape=IMAGE_SIZE + [3], weights='imagenet', include_top=False)

```

Downloading data from https://github.com/fchollet/deep-learning-models/releases/download/v0.1/vgg16_weights_tf_dim_ordering_tf_kernels_notop.h5 (http s://github.com/fchollet/deep-learning-models/releases/download/v0.1/vgg16_weights_tf_dim_ordering_tf_kernels_notop.h5)

58892288/58889256 [=====] - 5s 0us/step

In []:

```
1 model.summary() #pre trained vgg16 model
```

Model: "vgg16"

Layer (type)	Output Shape	Param #
=====		
input_1 (InputLayer)	(None, 256, 256, 3)	0
block1_conv1 (Conv2D)	(None, 256, 256, 64)	1792
block1_conv2 (Conv2D)	(None, 256, 256, 64)	36928
block1_pool1 (MaxPooling2D)	(None, 128, 128, 64)	0
block2_conv1 (Conv2D)	(None, 128, 128, 128)	73856
block2_conv2 (Conv2D)	(None, 128, 128, 128)	147584
block2_pool1 (MaxPooling2D)	(None, 64, 64, 128)	0
block3_conv1 (Conv2D)	(None, 64, 64, 256)	295168
block3_conv2 (Conv2D)	(None, 64, 64, 256)	590080
block3_conv3 (Conv2D)	(None, 64, 64, 256)	590080
block3_pool1 (MaxPooling2D)	(None, 32, 32, 256)	0
block4_conv1 (Conv2D)	(None, 32, 32, 512)	1180160
block4_conv2 (Conv2D)	(None, 32, 32, 512)	2359808
block4_conv3 (Conv2D)	(None, 32, 32, 512)	2359808
block4_pool1 (MaxPooling2D)	(None, 16, 16, 512)	0
block5_conv1 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
block5_pool1 (MaxPooling2D)	(None, 8, 8, 512)	0
=====		
Total params: 14,714,688		
Trainable params: 14,714,688		
Non-trainable params: 0		

MODEL_1(INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> Maxpool Layer --> 2 FC layers --> Output Layer)

In []:

```
1 #model_1
2 for layer in model.layers:
3     layer.trainable = False
4 #Adding custom Layers
5 x = model.output
6 x = Conv2D(filters=512, kernel_size=(3,3), padding="same", activation="relu")(x)
7 x = MaxPool2D(2,2)(x)
8 x = Flatten()(x)
9 x = Dense(256, activation="relu")(x)
10 x = Dense(128, activation="relu")(x)
11 output = Dense(16, activation="softmax")(x)
12 # creating the final model
13 model_1 = Model(inputs = model.input, outputs = output)
14 # compile the model
15 model_1.compile(loss = "categorical_crossentropy", optimizer = 'Adam', metrics=["accuracy"])
```

In []:

```

1 # summary of the model_1
2 model_1.summary()

```

Model: "model_1"

Layer (type)	Output Shape	Param #
=====		
input_1 (InputLayer)	(None, 256, 256, 3)	0
block1_conv1 (Conv2D)	(None, 256, 256, 64)	1792
block1_conv2 (Conv2D)	(None, 256, 256, 64)	36928
block1_pool (MaxPooling2D)	(None, 128, 128, 64)	0
block2_conv1 (Conv2D)	(None, 128, 128, 128)	73856
block2_conv2 (Conv2D)	(None, 128, 128, 128)	147584
block2_pool (MaxPooling2D)	(None, 64, 64, 128)	0
block3_conv1 (Conv2D)	(None, 64, 64, 256)	295168
block3_conv2 (Conv2D)	(None, 64, 64, 256)	590080
block3_conv3 (Conv2D)	(None, 64, 64, 256)	590080
block3_pool (MaxPooling2D)	(None, 32, 32, 256)	0
block4_conv1 (Conv2D)	(None, 32, 32, 512)	1180160
block4_conv2 (Conv2D)	(None, 32, 32, 512)	2359808
block4_conv3 (Conv2D)	(None, 32, 32, 512)	2359808
block4_pool (MaxPooling2D)	(None, 16, 16, 512)	0
block5_conv1 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
block5_pool (MaxPooling2D)	(None, 8, 8, 512)	0
conv2d_1 (Conv2D)	(None, 8, 8, 512)	2359808
max_pooling2d_1 (MaxPooling2D)	(None, 4, 4, 512)	0
flatten_1 (Flatten)	(None, 8192)	0
dense_1 (Dense)	(None, 256)	2097408
dense_2 (Dense)	(None, 128)	32896
dense_3 (Dense)	(None, 16)	2064
=====		
Total params: 19,206,864		
Trainable params: 4,492,176		

Non-trainable params: 14,714,688

In []:

```
1 train_steps = train_generator.n//train_generator.batch_size
2 validation_steps = validation_generator.n//validation_generator.batch_size
```

In []:

```
1 #fitting the model_1
2 model_1.fit_generator(train_generator,steps_per_epoch=train_steps, epochs=5,
3                       validation_data=validation_generator,validation_steps=va
4
```

Epoch 1/5

1200/1200 [=====] - 715s 596ms/step - loss: 1.2791
- accuracy: 0.6035 - val_loss: 0.6011 - val_accuracy: 0.7000

Epoch 2/5

1200/1200 [=====] - 703s 586ms/step - loss: 0.8746
- accuracy: 0.7310 - val_loss: 0.9771 - val_accuracy: 0.7421

Epoch 3/5

1200/1200 [=====] - 703s 586ms/step - loss: 0.7227
- accuracy: 0.7766 - val_loss: 0.7426 - val_accuracy: 0.7505

Epoch 4/5

1200/1200 [=====] - 703s 586ms/step - loss: 0.6244
- accuracy: 0.8055 - val_loss: 0.4625 - val_accuracy: 0.7480

Epoch 5/5

1200/1200 [=====] - 704s 587ms/step - loss: 0.5377
- accuracy: 0.8326 - val_loss: 0.6681 - val_accuracy: 0.7444

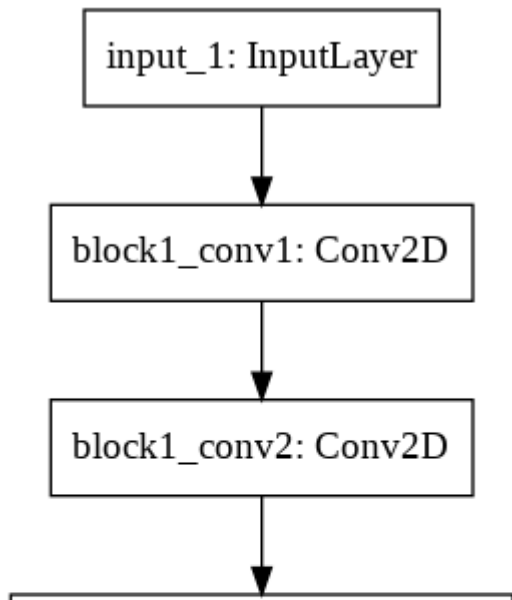
Out[19]:

<keras.callbacks.callbacks.History at 0x7f6a6041f2e8>

In []:

```
1 # model graphs
2 tf.keras.utils.plot_model(
3     model_1, to_file='model_1.png', show_shapes=False, show_layer_names=True,
4     rankdir='TB', expand_nested=False, dpi=96
5 )
```

Out[20]:



In []:

```
1 %tensorboard --logdir logs
```

<IPython.core.display.Javascript object>

observations

- out of total parameters only 1/4 are trainable.
- after running only 5 epochs we get accuracy of .75 ,that is if we increase the epoch number we can increase accuracy and reduce loss.
- from histogram we can understand that after 5 epoch added layers forms same distribution as pretrained layers as the number of neurons in the layers decreases range of dsistribution(mean=0) of weight increases *after every epochs weight distribution reduces its ranges centered towards mean(standard deviation reduces).

In []:

```
1 !rm -rf ./logs/
```

MODEL_2 (INPUT --> VGG-16 without Top layers(FC) --> 2 Conv Layers identical to FC --> Output Layer)

In []:

```
1 #model_2
2 for layer in model.layers:
3     layer.trainable = False
4 #Adding custom Layers
5 x = model.output
6 x = Conv2D(filters=4096,kernel_size=8 ,strides=1,activation="relu")(x)
7 x = Conv2D(filters=4096,kernel_size=1 ,strides=1,activation="relu")(x)
8 x = Flatten()(x)
9 # creating the final model
10 output= Dense(16, activation="softmax")(x)
11 model_2 = Model(inputs = model.input, outputs = output)
12 # compile the model
13 model_2.compile(loss="categorical_crossentropy",optimizer = 'Adam',metrics=['accuracy'])
```

In []:

```
1 # summary of the model_2
2 model_2.summary()
```

Model: "model_4"

Layer (type)	Output Shape	Param #
=====		
input_1 (InputLayer)	(None, 256, 256, 3)	0
block1_conv1 (Conv2D)	(None, 256, 256, 64)	1792
block1_conv2 (Conv2D)	(None, 256, 256, 64)	36928
block1_pool (MaxPooling2D)	(None, 128, 128, 64)	0
block2_conv1 (Conv2D)	(None, 128, 128, 128)	73856
block2_conv2 (Conv2D)	(None, 128, 128, 128)	147584
block2_pool (MaxPooling2D)	(None, 64, 64, 128)	0
block3_conv1 (Conv2D)	(None, 64, 64, 256)	295168
block3_conv2 (Conv2D)	(None, 64, 64, 256)	590080
block3_conv3 (Conv2D)	(None, 64, 64, 256)	590080
block3_pool (MaxPooling2D)	(None, 32, 32, 256)	0
block4_conv1 (Conv2D)	(None, 32, 32, 512)	1180160
block4_conv2 (Conv2D)	(None, 32, 32, 512)	2359808
block4_conv3 (Conv2D)	(None, 32, 32, 512)	2359808
block4_pool (MaxPooling2D)	(None, 16, 16, 512)	0
block5_conv1 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
block5_pool (MaxPooling2D)	(None, 8, 8, 512)	0
conv2d_6 (Conv2D)	(None, 1, 1, 4096)	134221824
conv2d_7 (Conv2D)	(None, 1, 1, 4096)	16781312
flatten_4 (Flatten)	(None, 4096)	0
dense_6 (Dense)	(None, 16)	65552
=====		
Total params: 165,783,376		
Trainable params: 151,068,688		
Non-trainable params: 14,714,688		

In []:

```

1 #fitting model_2
2 model_2.fit_generator(train_generator, steps_per_epoch=train_steps, epochs=5, verbose=1,
3                       validation_data=validation_generator, validation_steps=va
4

```

Epoch 1/5
 1200/1200 [=====] - 1410s 1s/step - loss: 1.4058 - accuracy: 0.6052 - val_loss: 1.5813 - val_accuracy: 0.6809
 Epoch 2/5
 1200/1200 [=====] - 1409s 1s/step - loss: 0.9071 - accuracy: 0.7223 - val_loss: 1.2431 - val_accuracy: 0.7105
 Epoch 3/5
 1200/1200 [=====] - 1410s 1s/step - loss: 0.7603 - accuracy: 0.7663 - val_loss: 0.6085 - val_accuracy: 0.7309
 Epoch 4/5
 1200/1200 [=====] - 1409s 1s/step - loss: 0.6576 - accuracy: 0.7952 - val_loss: 0.8318 - val_accuracy: 0.7357
 Epoch 5/5
 1200/1200 [=====] - 1409s 1s/step - loss: 0.5836 - accuracy: 0.8188 - val_loss: 0.7888 - val_accuracy: 0.7315

Out[32]:

<keras.callbacks.callbacks.History at 0x7f6a08762668>

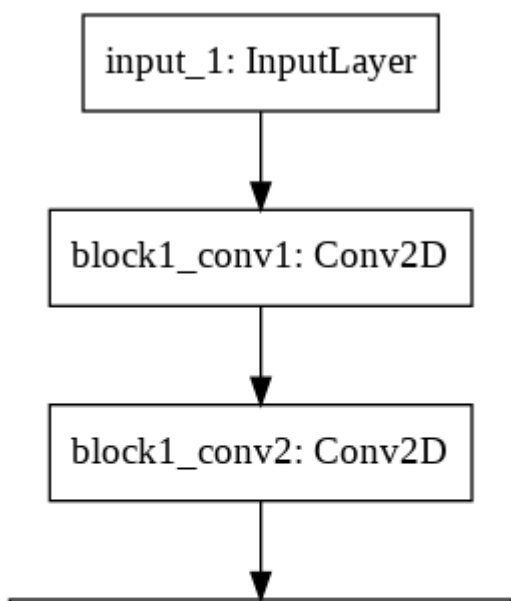
In []:

```

1 # model graphs
2 tf.keras.utils.plot_model(
3     model_2, to_file='model_2.png', show_shapes=False, show_layer_names=True,
4     rankdir='TB', expand_nested=False, dpi=96
5 )

```

Out[33]:



In []:

```
1 %tensorboard --logdir logs
```

<IPython.core.display.Javascript object>

observations

- number of trainable parameter large as compared to model_1 so time taken for each epoch is increases
- as number of trainable parameters increases and we use epoch number as previous model we got lesser accuracy and higher loss value(we got only .73 accuracy)
- in the conv2d_6 layer change of weights after every epoch is very small because that layer consist of large number of parameter.

In []:

```
1 !rm -rf ./logs/
```

Model-3

In []:

```
1 for layer in model.layers[-6:]: # training last 6 layers of vgg16
2     layer.trainable = True
3     print("Layer '%s' is trainable" % layer.name)
```

```
Layer 'block4_conv3' is trainable
Layer 'block4_pool' is trainable
Layer 'block5_conv1' is trainable
Layer 'block5_conv2' is trainable
Layer 'block5_conv3' is trainable
Layer 'block5_pool' is trainable
```

In []:

```
1 #model_3
2 #Adding custom Layers
3 x = model.output
4 x = Conv2D(filters=4096,kernel_size=8 ,strides=1,activation="relu")(x)
5 x = Conv2D(filters=4096,kernel_size=1 ,strides=1,activation="relu")(x)
6 x = Flatten()(x)
7 # creating the final model
8 output = Dense(16, activation="softmax")(x)
9 model_3 = Model(inputs = model.input, outputs = output)
10 # compile the model
11 model_3.compile(loss="categorical_crossentropy",optimizer = 'Adam',metrics=['accuracy'])
```

In []:

1 model_3.summary()

Model: "model_5"

Layer (type)	Output Shape	Param #
=====		
input_1 (InputLayer)	(None, 256, 256, 3)	0
block1_conv1 (Conv2D)	(None, 256, 256, 64)	1792
block1_conv2 (Conv2D)	(None, 256, 256, 64)	36928
block1_pool (MaxPooling2D)	(None, 128, 128, 64)	0
block2_conv1 (Conv2D)	(None, 128, 128, 128)	73856
block2_conv2 (Conv2D)	(None, 128, 128, 128)	147584
block2_pool (MaxPooling2D)	(None, 64, 64, 128)	0
block3_conv1 (Conv2D)	(None, 64, 64, 256)	295168
block3_conv2 (Conv2D)	(None, 64, 64, 256)	590080
block3_conv3 (Conv2D)	(None, 64, 64, 256)	590080
block3_pool (MaxPooling2D)	(None, 32, 32, 256)	0
block4_conv1 (Conv2D)	(None, 32, 32, 512)	1180160
block4_conv2 (Conv2D)	(None, 32, 32, 512)	2359808
block4_conv3 (Conv2D)	(None, 32, 32, 512)	2359808
block4_pool (MaxPooling2D)	(None, 16, 16, 512)	0
block5_conv1 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block5_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
block5_pool (MaxPooling2D)	(None, 8, 8, 512)	0
conv2d_8 (Conv2D)	(None, 1, 1, 4096)	134221824
conv2d_9 (Conv2D)	(None, 1, 1, 4096)	16781312
flatten_5 (Flatten)	(None, 4096)	0
dense_7 (Dense)	(None, 16)	65552
=====		
Total params: 165,783,376		
Trainable params: 160,507,920		
Non-trainable params: 5,275,456		

In [39]:

```

1 #fitting model_3
2 model_3.fit_generator(train_generator, steps_per_epoch=train_steps, epochs=5,
3                       validation_data=validation_generator, validation_steps=va
4

```

Epoch 1/5

1200/1200 [=====] - 2155s 2s/step - loss: 2.8802 - accuracy: 0.0624 - val_loss: 2.7756 - val_accuracy: 0.0601

Epoch 2/5

1200/1200 [=====] - 2181s 2s/step - loss: 2.7728 - accuracy: 0.0608 - val_loss: 2.7724 - val_accuracy: 0.0617

Epoch 3/5

1200/1200 [=====] - 2155s 2s/step - loss: 2.7728 - accuracy: 0.0615 - val_loss: 2.7754 - val_accuracy: 0.0584

Epoch 4/5

1200/1200 [=====] - 2160s 2s/step - loss: 2.7728 - accuracy: 0.0602 - val_loss: 2.7708 - val_accuracy: 0.0601

Epoch 5/5

1200/1200 [=====] - 2162s 2s/step - loss: 2.7728 - accuracy: 0.0597 - val_loss: 2.7729 - val_accuracy: 0.0601

Out[39]:

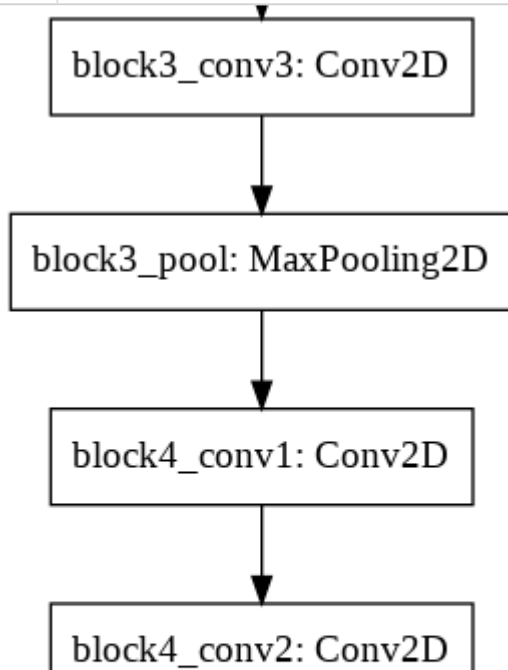
<keras.callbacks.callbacks.History at 0x7f6a0c669eb8>

In [40]:

```

1 # model graphs
2 tf.keras.utils.plot_model(
3     model_3, to_file='model_3.png', show_shapes=False, show_layer_names=True,
4     rankdir='TB', expand_nested=False, dpi=96
5 )

```



In [42]:

```
1 %tensorboard --logdir logs
```

Reusing TensorBoard on port 6006 (pid 5579), started 3:18:59 ago. (Use '!kill 5579' to kill it.)

<IPython.core.display.Javascript object>

observations

- we get only .0601 accuracy on model_3.
- from the histograms of last layers we understand that layers are not completely learned .
- low accuracy is because of model_3 consist of very large number of trainable parameters so that it need more epoch to attain more accuracy.
- as trainable parameters increases more epoch is needed to get good accuracy loss convergence.
- if we run the model for 25-30 epochs we get accuracy above 90

In [3]:

```
1 from prettytable import PrettyTable
2 from prettytable import ALL as ALL
3 table=PrettyTable(hrules=ALL)
4 table.field_names = [ "S1.N0", "Model", "Number of epochs", " val_accuracy"] # # http://
5 table.add_row([1, "model_1", "5", 0.7444])
6 table.add_row([2, "model_2", "5", 0.7315])
7 table.add_row([3, "model_3", "5", 0.0601])
8 print(table)
```

S1.N0	Model	Number of epochs	val_accuracy
1	model_1	5	0.7444
2	model_2	5	0.7315
3	model_3	5	0.0601

In []:

```
1
```