1. Download all the data in this folder https://drive.google.com/open?id=1Z4TyI7FcFVEx8qdl-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 spl:
- 3. Try not to load all the images into memory, use the gernarators that we have given the or you can use this method also

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
- 5. Use tensorboard for every model and analyse your gradients. (you need to upload the scre

Note: fit genarator() method will have problems with the tensorboard histograms, try to del

6. You can check about Transfer Learning in this link - <a href="https://blog.keras.io/building-pow">https://blog.keras.io/building-pow</a>

#!wget --header="Host: doc-08-6g-docs.googleusercontent.com" --header="User-Agent: Mozilla/5.

```
get ipython().system raw("unrar x rvl-cdip.rar")
%load ext tensorboard
from keras_preprocessing.image import ImageDataGenerator
from tensorflow import keras
from sklearn.model selection import train test split
import pandas as pd
df = pd.read csv('/content/labels final.csv')
train_data, valid_data = train_test_split(df, test_size=0.3, random_state=42)
#data augmentation
datagen = ImageDataGenerator(rescale=1./255)
#get images from different directories
train generator = datagen.flow from dataframe(dataframe = train data,
                  directory="data final", x col="path", y col="label",
                  class mode="raw", target size=(256,256), batch size=32)
valid_generator = datagen.flow_from_dataframe(dataframe = valid_data,
                  directory="data final", x col="path", y col="label",
                  class_mode="raw", target_size=(256,256), batch_size=32)
 Found 33600 validated image filenames.
     Found 14400 validated image filenames.
```

#### Model-1

- 1. Use  $\underline{\text{VGG-16}}$  pretrained network without Fully Connected layers and initilize all the weight
- 2. After VGG-16 network without FC layers, add a new Conv block ( 1 Conv layer and 1 Maxpor
- 3. Final architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 10 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> May 12 architecture will be INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> Conv
- 4. Train only new Conv block, FC layers, output layer. Don't train the VGG-16 network.

```
input_shape=(256, 256, 3))
```

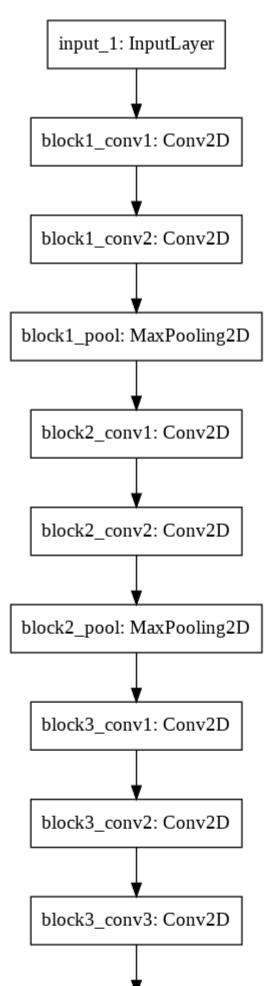
```
x = vgg model.output
x = keras.layers.Conv2D(32, (3, 3),padding = 'same')(x)
x = keras.layers.MaxPooling2D(pool_size=(2, 2))(x)
x = keras.layers.Flatten()(x)
x = keras.layers.Dense(128, activation='relu')(x)
x = keras.layers.Dropout(.5)(x)
x = keras.layers.Dense(64, activation='relu')(x)
x = keras.layers.Dropout(.5)(x)
preds = keras.layers.Dense(16, activation='softmax')(x)
model1 = Model(inputs = vgg model.input,outputs=preds)
    Downloading data from <a href="https://storage.googleapis.com/tensorflow/keras-applications/vgg1">https://storage.googleapis.com/tensorflow/keras-applications/vgg1</a>
     58892288/58889256 [============ ] - Os Ous/step
for layer in model1.layers[:19]:
    layer.trainable=False
for layer in model1.layers[19:]:
    layer.trainable=True
model1.summary()
```

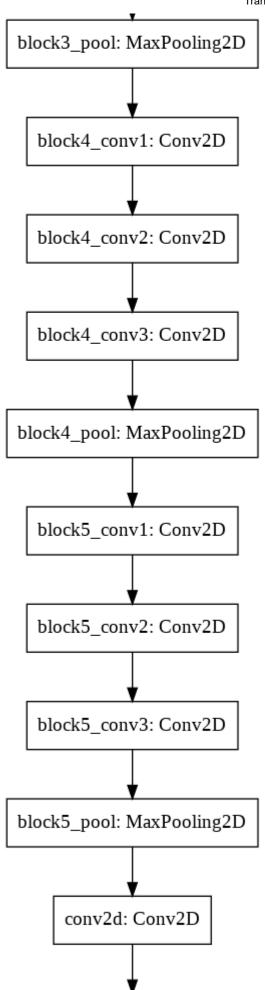
Model: "functional 1"

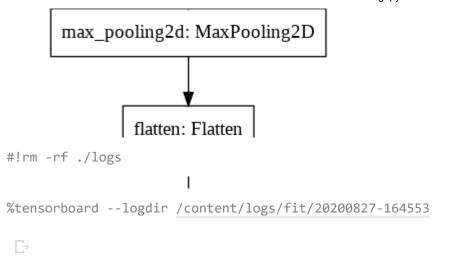
Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 256, 256, 3)]	0
block1_conv1 (Conv2D)	(None, 256, 256, 64)	1792
block1_conv2 (Conv2D)	(None, 256, 256, 64)	36928
block1_pool (MaxPooling2D)	(None, 128, 128, 64)	0
block2_conv1 (Conv2D)	(None, 128, 128, 128)	73856
block2_conv2 (Conv2D)	(None, 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

```
impor
model1.compile(optimizer= 'adam',loss='sparse_categorical_crossentropy',metrics=['accuracy'])
#tensorflow callback
log_dir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S")
tensorboard callback = keras.callbacks.TensorBoard(log dir=log dir,histogram freq=1, write gr
#early stopping callback
early stopping = keras.callbacks.EarlyStopping(monitor='val accuracy', patience=2)
STEP SIZE TRAIN=train generator.n//train generator.batch size
STEP SIZE VALID=valid generator.n//valid generator.batch size
with tf.device('/device:GPU:0'):
  model1.fit_generator(generator=train_generator,
                    steps_per_epoch=STEP_SIZE_TRAIN,
                    validation data=valid generator,
                    validation steps=STEP SIZE VALID,
                    callbacks = [tensorboard_callback, early_stopping],
                    epochs=5)
```

```
WARNING:tensorflow:`write grads` will be ignored in TensorFlow 2.0 for the `TensorBoard
    WARNING:tensorflow:From <ipython-input-10-f5fb2a2567c6>:21: Model.fit_generator (from t
    Instructions for updating:
    Please use Model.fit, which supports generators.
    Epoch 1/5
       1/1050 [.....] - ETA: 0s - loss: 3.3809 - accuracy: 0.0312W
    Instructions for updating:
    use `tf.profiler.experimental.stop` instead.
    1050/1050 [================ ] - 541s 516ms/step - loss: 2.0371 - accuracy:
    Epoch 2/5
    1050/1050 [=============== ] - 541s 515ms/step - loss: 1.5413 - accuracy:
    Epoch 3/5
    1050/1050 [=============== ] - 540s 515ms/step - loss: 1.3827 - accuracy:
    Epoch 4/5
    1050/1050 [================ ] - 540s 515ms/step - loss: 1.2765 - accuracy:
    Epoch 5/5
    1050/1050 [=============== ] - 541s 515ms/step - loss: 1.1900 - accuracy:
keras.utils.plot model(
   model1, to file='model1.png', show shapes=False, show layer names=True,
   rankdir='TB', expand nested=False, dpi=96
)
```







```
for layer in model_2.layers[:13]:
    layer.trainable=False
for layer in model_2.layers[13:]:
    layer.trainable=True

model_2.summary()
```

• The dataset difference between the pretrained model and our models are also high regarding sizes, variance of categories.

TensorBoard

**SCALARS** 

**GRAPHS** 

**INACTIVE** 

### Observations:

- The validation accuracy for Mode1 1 = 0.69
- The imagenet dataset contains 14 million images and our modle trained on 33600 images, a huge differences in dataset
- · Maybe the imagenet dataset has less category of file type data



#### Model-2



- 1. Use  $\underline{\text{VGG-16}}$  pretrained network without Fully Connected layers and initilize all the weight
- 2. After VGG-16 network without FC layers, don't use FC layers, use conv layers only as Fu.
- 3. Final architecture will be VGG-16 without FC layers(without top), 2 Conv layers identicated
- 3. Train only last 2 Conv layers identical to FC layers, 1 output layer. Don't train the VC

Model: "functional\_3"

Output Shape	Param #
[(None, 256, 256, 3)]	0
(None, 256, 256, 64)	1792
(None, 256, 256, 64)	36928
(None, 128, 128, 64)	0
(None, 128, 128, 128)	73856
(None, 128, 128, 128)	147584
(None, 64, 64, 128)	0
(None, 64, 64, 256)	295168
(None, 64, 64, 256)	590080
(None, 64, 64, 256)	590080
(None, 32, 32, 256)	0
(None, 32, 32, 512)	1180160
(None, 32, 32, 512)	2359808
(None, 32, 32, 512)	2359808
(None, 16, 16, 512)	0
(None, 16, 16, 512)	2359808
(None, 16, 16, 512)	2359808
(None, 16, 16, 512)	2359808
(None, 8, 8, 512)	0
(None, 1, 1, 4096)	134221824
(None, 1, 1, 4096)	16781312
(None, 4096)	0
	[(None, 256, 256, 3)] (None, 256, 256, 64) (None, 256, 256, 64) (None, 128, 128, 64) (None, 128, 128, 128) (None, 64, 64, 128) (None, 64, 64, 256) (None, 64, 64, 256) (None, 32, 32, 256) (None, 32, 32, 512) (None, 32, 32, 512) (None, 16, 16, 512) (None, 16, 16, 512) (None, 16, 16, 512) (None, 8, 8, 512) (None, 1, 1, 4096) (None, 1, 1, 4096)

model\_2.compile(optimizer= 'adam',loss='sparse\_categorical\_crossentropy',metrics=['accuracy']

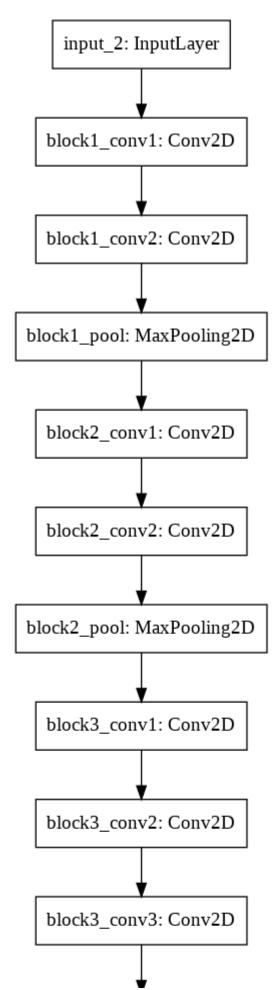
```
#tensorflow callback
```

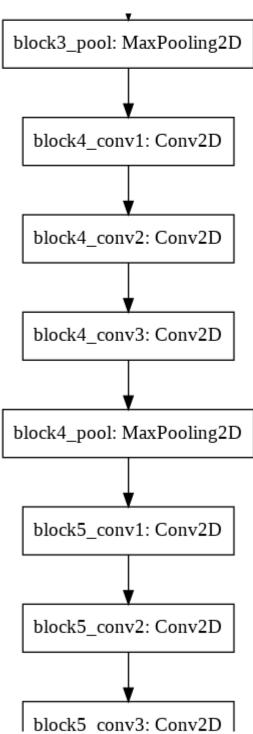
log\_dir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S")
tensorboard\_callback = keras.callbacks.TensorBoard(log\_dir=log\_dir,histogram\_freq=1, write\_gr

#early stopping callback

earlv stonning = keras.callhacks.FarlvStonning(monitor='val accuracv'. natience=2)

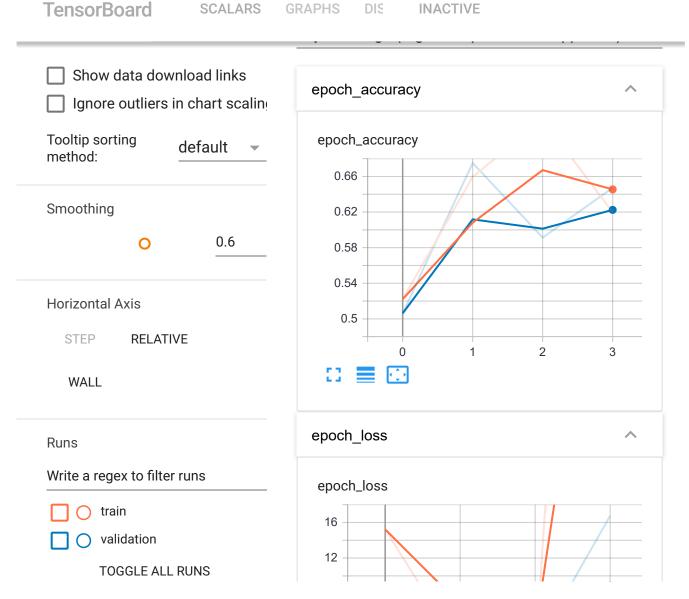
```
var_accaracy , pacacine = /
model 2.fit generator(generator=train generator,
               steps per epoch=STEP SIZE TRAIN,
               validation data=valid generator,
               validation_steps=STEP_SIZE_VALID,
               callbacks = [tensorboard callback, early stopping],
               epochs=5)
flow:`write_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.
   ========= - - 1091s 1s/step - loss: 1.7481 - accuracy: 0.6596 - val los
   ========= - - 1093s 1s/step - loss: 0.9951 - accuracy: 0.7239 - val los
   thon.keras.callbacks.History at 0x7f17ceb93588>
keras.utils.plot model(
  model_2, to_file='model2.png', show_shapes=False, show_layer_names=True,
   rankdir='TB', expand nested=False, dpi=96)
```





%tensorboard --logdir logs/fit/20200827-173114

L?



### Observations:

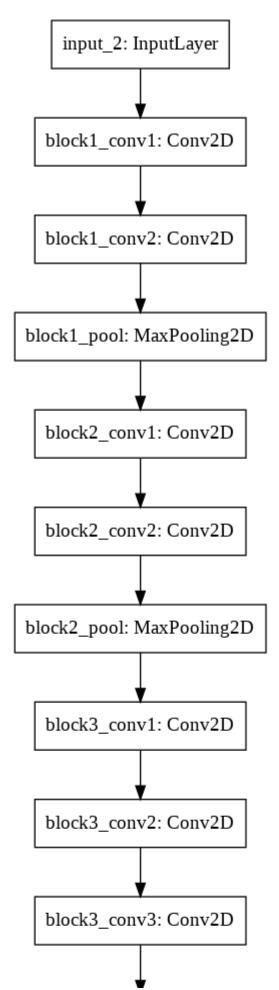
- In Model 2 the validation accuracy = 0.64
- Even though the only difference in vgg\_16 and this model was last two layers in vgg-16 were dense and model 2 had convoluted data
- · Maybe imagenet data doesn't consist variety of file type images

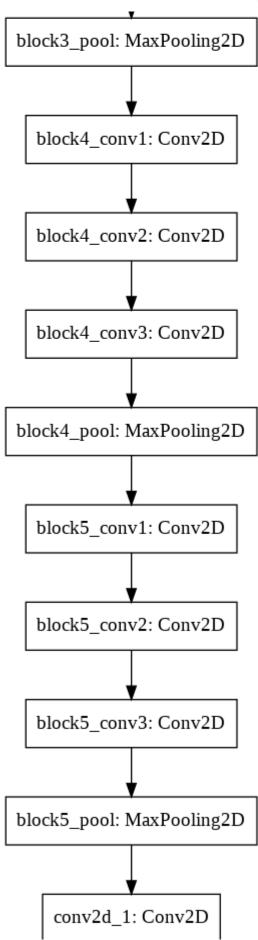
#### Model-3

1. Use same network as Model-2 'INPUT --> VGG-16 without Top layers(FC) --> 2 Conv Layers :

```
Model: "functional 3"
```

```
Layer (type)
                         Output Shape
                                             Param #
    ______
   input 2 (InputLayer)
                         [(None, 256, 256, 3)]
model 2.compile(optimizer= 'adam',loss='sparse categorical crossentropy',metrics=['accuracy']
#tensorflow callback
log dir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S")
tensorboard callback = keras.callbacks.TensorBoard(log dir=log dir, histogram freq=1, write gr
#early stopping callback
early stopping = keras.callbacks.EarlyStopping(monitor='val accuracy', patience=2)
model 2.fit generator(generator=train generator,
               steps per epoch=STEP SIZE TRAIN,
               validation data=valid generator,
               validation steps=STEP SIZE VALID,
               callbacks = [tensorboard callback, early stopping],
               epochs=5)
□ sorflow:`write grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.
   i.python.keras.callbacks.History at 0x7f17cc1229b0>
keras.utils.plot model(
   model 2, to file='model 2.png', show shapes=False, show layer names=True,
   rankdir='TB', expand nested=False, dpi=96)
```

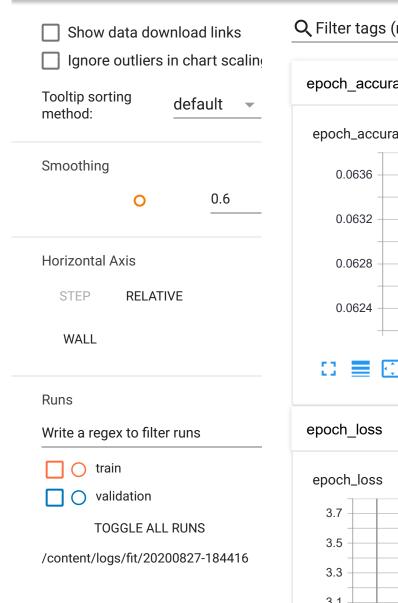




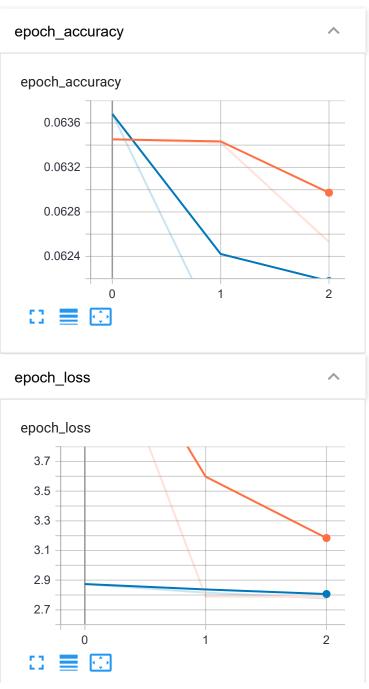
%tensorboard --logdir /content/logs/fit/20200827-184416

 $\Box$ 

TensorBoard SCALARS GRAPHS DIS INACTIVE



# Q Filter tags (regular expressions supported)



# **Observations:**

- The validation accuracy for model 3 is 0.0619
- In these model we have also trained the pre trained layers of vgg-16 models
- The accuracy is so low than other two models maybe the weights generated by pretrained models were not useful for the images we have.