```
import tensorflow as tf
import os
import numpy as np
import pandas as pd
from keras.layers import Input, Lambda, Dense, Flatten
from keras.models import Model
from keras.applications.vgg16 import VGG16
from keras.applications.vgg16 import preprocess input
from keras.preprocessing import image
from keras.layers import Dense, Conv2D, MaxPool2D , Flatten
from keras.callbacks import Callback
from keras.callbacks import TensorBoard
import datetime
from tensorflow.keras.utils import plot model
!gdown --id 1Z4TyI7FcFVEx8qdl4j09qxvxaqLSqoEu
/usr/local/lib/python3.8/dist-packages/gdown/cli.py:127:
FutureWarning: Option `--id` was deprecated in version 4.3.1 and will
be removed in 5.0. You don't need to pass it anymore to use a file ID.
  warnings.warn(
Access denied with the following error:
     Cannot retrieve the public link of the file. You may need to
change
     the permission to 'Anyone with the link', or have had many
accesses.
You may still be able to access the file from the browser:
      https://drive.google.com/uc?id=1Z4TyI7FcFVEx8gdl4j09gxvxaqLSgoEu
```

```
!wget --header="Host: doc-10-58-docs.googleusercontent.com" --
header="User-Agent: Mozilla/5.0 (Windows NT 10.0; Win64; x64)
AppleWebKit/537.36 (KHTML, like Gecko) Chrome/109.0.0.0 Safari/537.36"
--header="Accept:
text/html,application/xhtml+xml,application/xml;q=0.9,image/avif,image
/webp,image/apng,*/*;q=0.8,application/signed-exchange;v=b3;q=0.9" --
header="Accept-Language: en-US,en;q=0.9" --header="Referer:
https://drive.google.com/" --header="Cookie:
AUTH_csgqhf4p22bquld8f6s95mu6pd817erv_nonce=q22o5hnasoge2" --
header="Connection: keep-alive" "https://doc-10-58-
docs.googleusercontent.com/docs/securesc/ka55uvoaagpve0kkmfos6gnjgr7s6
tea/lu2lpn9jj7baqgopd776arvajqn3jj9j/
1675136025000/00484516897554883881/00389752900367068778/1Z4TyI7FcFVEx8
qd14j09qxvxaqLSqoEu?e=download&ax=AB85Z1Ch6Ny_pkChc2waWK-3Q9-Q0clF-
Se1JmTx6lVAME1A6Bv6JHnvowFy19BGI -lp-Z1nsV04u-
```

```
vT5CSJsucdXKk1MWiansc7De4qX10WzZAmNvd0jq0hLyau9U6wWKr070yiWs9FqThw7aq7
L-BIVhdfYFFxhP4tfDXURCJ9UI95vBGeE4HPTNjYZ-
js1hxeSzcLigKiROjDMgmJv4eue dGlzxxtz2cKHNpseRruCp3bA-
BaX2VLvFGFJaM0kzSm99izHH3ieOAvAhmuvnFZecBe30bpvIh2G9TmnpDcVI-
edoAEng j-SA5oj8dbiLMbt6i7wujHPqvGEV3q1pQyfqiwDcf 13VpOvNMMG-
sntYh TitFYKy7hdtWTnCJfmn9Ud VqU1Kz4p6E9nM2B8sApbK2nD9lW47qnWHQRPIouwA
B8AlaARZAiOsPvm8C3vtxPu5la-
5ilfliwrBLnzdiZz5kymgVSWdjd48g 9GveTCddZFicK64rktETIRAgfDGiCjuEkU81dr
6oGeworByXAWGvqzaAzcL85i1D1YP69RbVdcEzQ2 kZioPhiTnnVGBllbnYCqJcl j-
U6nS0jg5lxUarhH2P3v1SV8mH40tekwFxd2RqMqc0hUoiNDhFZi2cKI9IosTvwc-
F9YKatRRs8eCCNmfNyBTbBmK0phQpGk48CuHM0RkiRq95guFUSyXCZ2UhM9ckygT3o8SXq
cZ-doyX4SXsT-iZgAcvM0iQuxw-IbjFWKJqvFTNy-6ZXKTIWqVoWq5K4yP-
zn64R4Mv96BY64-d9l-vFY-
r7zUIvoLPBm3tS 2ktWxVjS MWW03ktSKYLSVyMM4zIED95G0m0x5GlLtqZ3nKxyfJq7fi
8XH-2TrbFr4xw&uuid=d1144c7b-aca1-4755-a75b-
80324461cd83&authuser=0&nonce=q22o5hnasoge2&user=00389752900367068778&
hash=glqhcu6ltvnv2qeqk8k4l8sild5u3bvk" -c -0 'rvl-cdip.rar'
--2023-01-31 03:34:43-- https://doc-10-58-
docs.googleusercontent.com/docs/securesc/ka55uvoaagpve0kkmfos6gnjgr7s6
tea/lu2lpn9jj7baggopd776arvajgn3jj9j/
1675136025000/00484516897554883881/00389752900367068778/1Z4TyI7FcFVEx8
qdl4j09qxvxaqLSqoEu?e=download&ax=AB85Z1Ch6Ny pkChc2waWK-3Q9-Q0clF-
Se1JmTx6lVAME1A6Bv6JHnvowFyI9BGI -lp-Z1nsV04u-
vT5CSJsucdXKk1MWiansc7De4gX10WzZAmNvd0jq0hLyau9U6wWKr070yiWs9FqThw7aq7
L-BIVhdfYFFxhP4tfDXURCJ9UI95vBGeE4HPTNjYZ-
is1hxeSzcLigKiROjDMgmJv4eue dGlzxxtz2cKHNpseRruCp3bA-
BqX2VLvFGFJgM0kzSm99jzHH3ie0AvAhmuvnFZecBe30bpvIh2G9TmnpDcVI-
edoAEng j-SA5oj8dbiLMbt6i7wujHPgvGEV3g1pQyfqiwDcf 13VpOvNMMG-
sntYh TitFYKy7hdtWTnCJfmn9Ud VqU1Kz4p6E9nM2B8sApbK2nD9lW47qnWHQRPIouwA
B8AlgARZAjQsPym8C3ytxPu5la-
5ilf1iwrBLnzdiZz5kymqVSWdjd48q 9GveTCddZFicK64rktETIRAqfDGiCjuEkU81dr
6oGeworByXAWGvgzaAzcL85i1D1YP69RbVdcEzQ2 kZioPhiTnnVGBllbnYCgJcl j-
U6nS0jg5lxUarhH2P3v1SV8mH40tekwFxd2RqMqc0hUoiNDhFZi2cKI9IosTvwc-
F9YKatRRs8eCCNmfNyBTbBmK0phQpGk48CuHM0RkiRq95quFUSyXCZ2UhM9ckyqT3o8SXq
cZ-doyX4SXsT-iZqAcvM0iQuxw-IbjFWKJqvFTNy-6ZXKTIWqVoWq5K4yP-
zn64R4Mv96BY64-d9l-vFY-
r7zUIvoLPBm3tS 2ktWxVjS MWWQ3ktSKYLSVyMM4zIED95G0m0x5GlLtqZ3nKxyfJq7fi
8XH-2TrbFr4xw&uuid=d1144c7b-aca1-4755-a75b-
80324461cd83&authuser=0&nonce=q22o5hnasoge2&user=00389752900367068778&
hash=g1qhcu6ltvnv2qeqk8k4l8sild5u3bvk
Resolving doc-10-58-docs.googleusercontent.com (doc-10-58-
docs.googleusercontent.com)... 142.251.111.132, 2607:f8b0:4004:c19::84
Connecting to doc-10-58-docs.googleusercontent.com (doc-10-58-
docs.googleusercontent.com) | 142.251.111.132 | :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 4660541790 (4.3G) [application/rar]
Saving to: 'rvl-cdip.rar'
```

```
2023-01-31 03:35:07 (190 MB/s) - 'rvl-cdip.rar' saved
[4660541790/4660541790]
get ipvthon().system raw("unrar x rvl-cdip.rar")
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
df = pd.read csv('labels final.csv',dtype=str)
df.head()
                                         path label
   imagesv/v/o/h/voh71d00/509132755+-2755.tif
         imagesl/l/x/t/lxt19d00/502213303.tif
                                                   3
1
        imagesx/x/e/d/xed05a00/2075325674.tif
                                                   2
3
  imageso/o/j/b/ojb60d00/517511301+-1301.tif
                                                   3
        imagesq/q/z/k/qzk17e00/2031320195.tif
from keras preprocessing.image import ImageDataGenerator
datagen = \overline{I}mageDataGenerator(rescale=1/255., validation split=0.2)
#Help in Image augmentation
dir p = "/content/data final"
from keras preprocessing.image import ImageDataGenerator
# Help in Image augmentation
datagen=tf.keras.preprocessing.image.ImageDataGenerator(rescale=1/255,
validation split=0.2)
# flow from dataframe used to loads the image dataset in memory and
generates batches of augmented data.
train_generator = datagen.flow_from_dataframe(dataframe= df,
directory=dir p, x col='path',y col='label',
target size=(128,128),class mode='categorical',batch size=128,subset='
training',seed=7)
Found 38400 validated image filenames belonging to 16 classes.
valid generator = datagen.flow from dataframe(dataframe = df,
directory = dir p,x col='path',y col='label',
target_size=(128,128),class_mode='categorical',batch_size=128,subset='
validation',seed=7)
Found 9600 validated image filenames belonging to 16 classes.
%reload ext tensorboard
```

```
logdir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S")
# tensorboard
tensorboard callback = TensorBoard(log dir=logdir, histogram freg=1)
Model 1
from tensorflow.keras.applications.vgg16 import VGG16
from tensorflow.keras.layers import
Dense, Input, Conv2D, MaxPooling2D, Flatten, Dropout
from tensorflow.keras.models import Model
# initilizing all the weights with Imagenet trained weights
# adding VGG-16 network without FC layers
vgg16 model = VGG16(weights='imagenet', include top=False,
input shape=(*(128, 128), 3))
# adding a new Conv block ( 1 Conv layer and 1 Maxpooling )
for layer in vgg16 model.layers:
  layer.trainable = False #parameters from trainable goes into
non trainable parameters
conv layer=
Conv2D(512,kernel size=(3,3),activation="relu",name="conv")
(vgg16 model.output)
max pool layer = MaxPooling2D(name='Max Pool')(conv layer)
flatten layer = Flatten(name="Flatten layer")(max pool layer)
# 2 FC layers
fully connected1=
Dense(1054,activation="relu",name="fully connected1")(flatten layer)
fully connected2= Dense(256,activation="relu",name="fully connected2")
(fully connected1)
# output layer to classify 16 classes
op= Dense(16,activation="softmax")(fully connected2)
#https://www.tensorflow.org/api docs/python/tf/keras/Model
model1= Model(inputs=vgg16 model.input,outputs= op,name="model 1")
model1.compile(optimizer="Adam",
loss="categorical_crossentropy",metrics=["accuracy"])
model1.summary()
Model: "model 1"
Layer (type)
                             Output Shape
                                                        Param #
 input 2 (InputLayer)
                             [(None, 128, 128, 3)]
 block1 conv1 (Conv2D)
                             (None, 128, 128, 64)
                                                        1792
```

(None, 128, 128, 64)

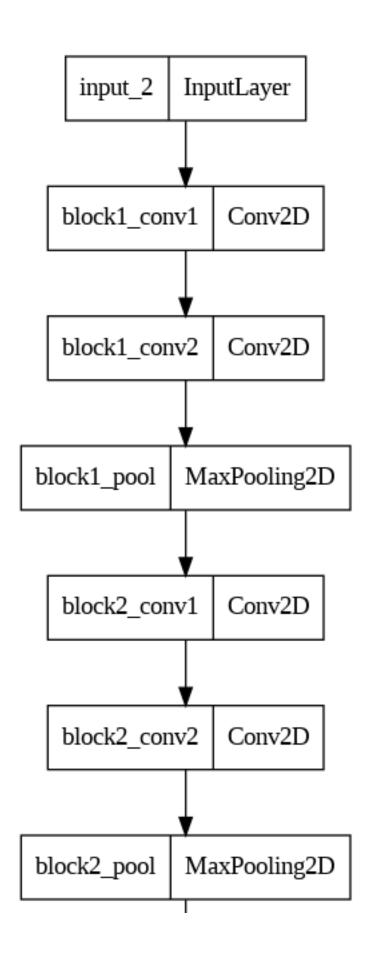
36928

block1 conv2 (Conv2D)

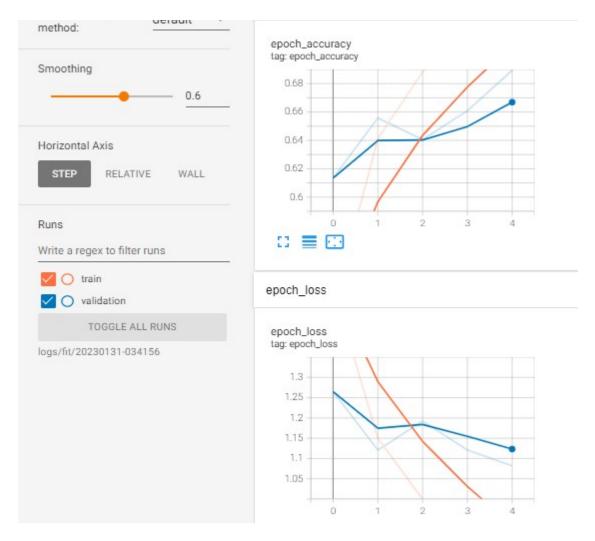
<pre>block1_pool (MaxPooling2D)</pre>	(None, 64, 64, 64)	0
block2_conv1 (Conv2D)	(None, 64, 64, 128)	73856
block2_conv2 (Conv2D)	(None, 64, 64, 128)	147584
<pre>block2_pool (MaxPooling2D)</pre>	(None, 32, 32, 128)	0
block3_conv1 (Conv2D)	(None, 32, 32, 256)	295168
block3_conv2 (Conv2D)	(None, 32, 32, 256)	590080
block3_conv3 (Conv2D)	(None, 32, 32, 256)	590080
<pre>block3_pool (MaxPooling2D)</pre>	(None, 16, 16, 256)	0
block4_conv1 (Conv2D)	(None, 16, 16, 512)	1180160
block4_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block4_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
<pre>block4_pool (MaxPooling2D)</pre>	(None, 8, 8, 512)	Θ
block5_conv1 (Conv2D)	(None, 8, 8, 512)	2359808
block5_conv2 (Conv2D)	(None, 8, 8, 512)	2359808
block5_conv3 (Conv2D)	(None, 8, 8, 512)	2359808
<pre>block5_pool (MaxPooling2D)</pre>	(None, 4, 4, 512)	0
conv (Conv2D)	(None, 2, 2, 512)	2359808
<pre>Max_Pool (MaxPooling2D)</pre>	(None, 1, 1, 512)	0
Flatten_layer (Flatten)	(None, 512)	Θ
<pre>fully_connected1 (Dense)</pre>	(None, 1054)	540702
fully_connected2 (Dense)	(None, 256)	270080
	(None, 250)	
dense (Dense)	(None, 16)	4112

Total params: 17,889,390 Trainable params: 3,174,702 Non-trainable params: 14,714,688

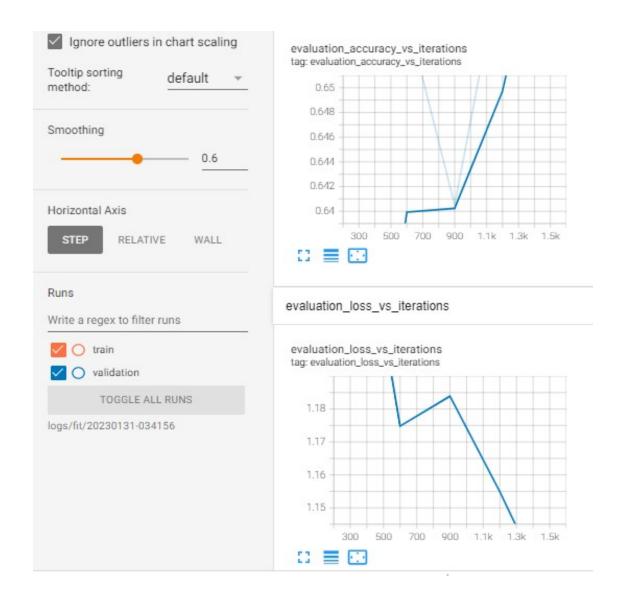
plot_model(model1)



```
steps = len(train_generator.labels) // train generator.batch size
early stoping = tf.keras.callbacks.EarlyStopping(patience=2)
log dir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M
%S")
tensorboard=
tf.keras.callbacks.TensorBoard(log dir=log dir,histogram freg=1,
write graph=True)
#callbacks list = [early stoping, tensorboard]
history = model1.fit(train generator,
validation_data=valid_generator,epochs=5,
steps per epoch=steps,callbacks=tensorboard)
Epoch 1/5
300/300 [============= ] - 224s 706ms/step - loss:
1.5222 - accuracy: 0.5223 - val loss: 1.2649 - val accuracy: 0.6134
Epoch 2/5
1.1499 - accuracy: 0.6414 - val loss: 1.1208 - val accuracy: 0.6558
Epoch 3/5
300/300 [============= ] - 194s 647ms/step - loss:
0.9998 - accuracy: 0.6883 - val loss: 1.1926 - val accuracy: 0.6405
Epoch 4/5
300/300 [============ ] - 191s 638ms/step - loss:
0.9013 - accuracy: 0.7177 - val loss: 1.1208 - val accuracy: 0.6608
Epoch 5/5
300/300 [============= ] - 188s 628ms/step - loss:
0.8103 - accuracy: 0.7458 - val loss: 1.0818 - val accuracy: 0.6894
%reload ext tensorboard
%tensorboard --logdir {log dir}
<IPython.core.display.Javascript object>
import IPython.display as display
from PIL import Image
display.display(Image.open('t1.png'))
```



display.display(Image.open('t2.png'))



Model 2

```
vgg16_model2 = VGG16(weights='imagenet', include_top=False,
input_shape=(*(128,128), 3))

for layer in vgg16_model2.layers:
    layer.trainable = False #from book 14-61.17e ,as we see we freeze
    the layers

# using conv layers only as Fully connected layer
# This conversion will reduce the No of Trainable parameters in FC
layers
conv_layer1=
Conv2D(512,kernel_size=(3,3),activation="relu",padding="same",name="conv1")(vgg16_model2.output)
conv_layer2=
Conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_size=(3,3),activation="relu",padding="same",name="conv2D(128,kernel_si
```

```
nv2")(conv_layer1)
flatten= Flatten(name="Flatten_layer")(conv_layer2)
# output layer for 16 class classification
op= Dense(16,activation="softmax")(flatten)
#https://www.tensorflow.org/api_docs/python/tf/keras/Model
model2= Model(inputs=vgg16_model2.input,outputs=op,name="model_2")
model2.compile(optimizer="Adam",
loss="categorical_crossentropy",metrics=["accuracy"])
```

model2.summary()

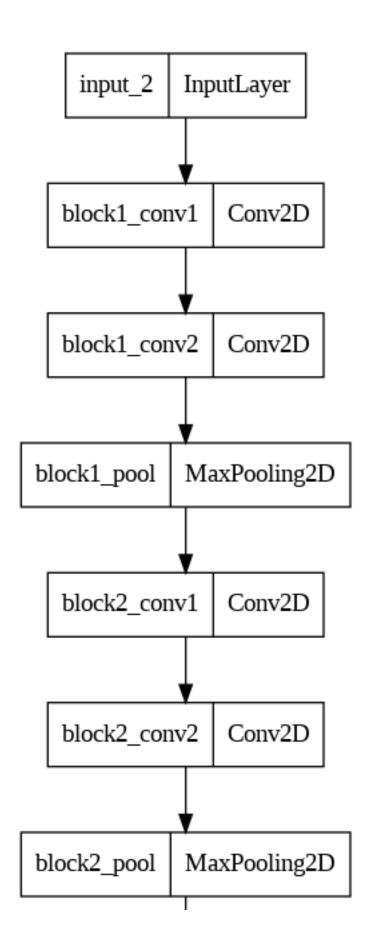
Model: "model_2"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 128, 128, 3)]	0
block1_conv1 (Conv2D)	(None, 128, 128, 64)	1792
block1_conv2 (Conv2D)	(None, 128, 128, 64)	36928
<pre>block1_pool (MaxPooling2D)</pre>	(None, 64, 64, 64)	0
block2_conv1 (Conv2D)	(None, 64, 64, 128)	73856
block2_conv2 (Conv2D)	(None, 64, 64, 128)	147584
<pre>block2_pool (MaxPooling2D)</pre>	(None, 32, 32, 128)	0
block3_conv1 (Conv2D)	(None, 32, 32, 256)	295168
block3_conv2 (Conv2D)	(None, 32, 32, 256)	590080
block3_conv3 (Conv2D)	(None, 32, 32, 256)	590080
<pre>block3_pool (MaxPooling2D)</pre>	(None, 16, 16, 256)	0
block4_conv1 (Conv2D)	(None, 16, 16, 512)	1180160
block4_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block4_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
<pre>block4_pool (MaxPooling2D)</pre>	(None, 8, 8, 512)	0
block5_conv1 (Conv2D)	(None, 8, 8, 512)	2359808
block5_conv2 (Conv2D)	(None, 8, 8, 512)	2359808
block5_conv3 (Conv2D)	(None, 8, 8, 512)	2359808

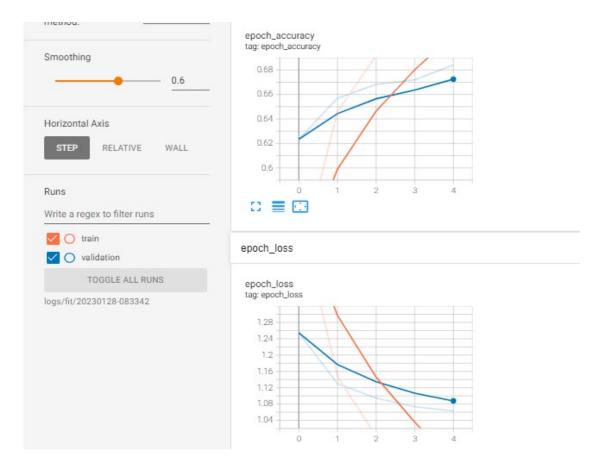
block5_pool (MaxPooling2D)	(None, 4, 4, 512)	0
conv1 (Conv2D)	(None, 4, 4, 512)	2359808
conv2 (Conv2D)	(None, 4, 4, 128)	589952
Flatten_layer (Flatten)	(None, 2048)	0
dense_1 (Dense)	(None, 16)	32784

Total params: 17,697,232 Trainable params: 2,982,544 Non-trainable params: 14,714,688

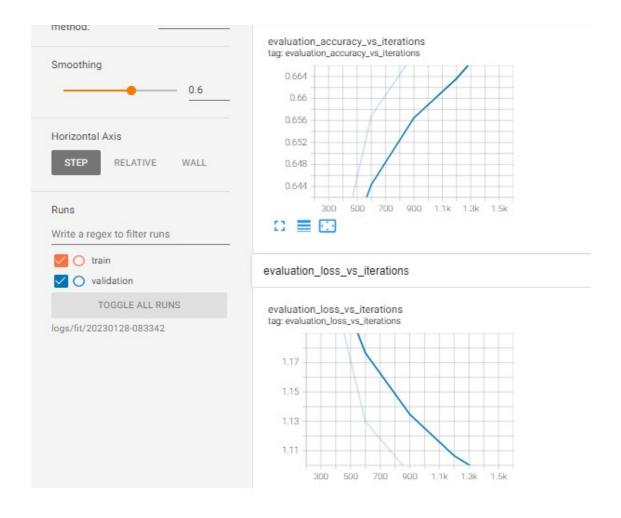
plot_model(model2)



```
steps = len(train generator.labels) // train generator.batch size
log dir="logs/fit/" + datetime.datetime.now().strftime("%Y%m%d-%H%M
%S")
tensorboard=
tf.keras.callbacks.TensorBoard(log dir=log dir,histogram freq=1,
write graph=True)
history = model2.fit(train generator,
validation data=valid generator,epochs=5,
steps per epoch=steps,callbacks=tensorboard)
Epoch 1/5
300/300 [============= ] - 218s 724ms/step - loss:
1.5175 - accuracy: 0.5300 - val loss: 1.3092 - val accuracy: 0.5885
Epoch 2/5
300/300 [============= ] - 206s 688ms/step - loss:
1.1280 - accuracy: 0.6524 - val loss: 1.1621 - val accuracy: 0.6427
Epoch 3/5
300/300 [============= ] - 205s 683ms/step - loss:
0.9902 - accuracy: 0.6936 - val loss: 1.0685 - val accuracy: 0.6791
Epoch 4/5
300/300 [============ ] - 205s 682ms/step - loss:
0.8818 - accuracy: 0.7253 - val loss: 1.1086 - val accuracy: 0.6673
Epoch 5/5
300/300 [============ ] - 211s 705ms/step - loss:
0.7889 - accuracy: 0.7537 - val loss: 1.0636 - val accuracy: 0.6970
%reload ext tensorboard
%tensorboard --logdir {log dir}
Reusing TensorBoard on port 6006 (pid 8578), started 0:01:23 ago. (Use
'!kill 8578' to kill it.)
<IPython.core.display.Javascript object>
import IPython.display as display
from PIL import Image
display.display(Image.open('ts.png'))
```



display.display(Image.open('ts2.png'))



Model 3

```
vgg16base model3 = VGG16(weights='imagenet',
include top=False,input shape=(*(128,128), 3))
for layer in vgg16base model3.layers[0:-6]: #training only Last 6
Layers of VGG-16 network
 layer.trainable = False
conv layer1=
Conv2D(512, kernel size=(3,3), activation="relu", padding="same", name="co
nv1")(vgg16base model3.output)
conv layer2=
Conv2D(128, kernel size=(3,3), activation="relu", padding="same", name="co
nv2")(conv layer1)
flatten= Flatten(name="Flatten_layer")(conv_layer2)
otput= Dense(16,activation="softmax")(flatten)
#optm = tf.keras.optimizers.SGD(learning rate=0.001, momentum=0.8)
#https://www.tensorflow.org/api docs/python/tf/keras/Model
model3=
```

Model(inputs=vgg16base_model3.input,outputs=otput,name="model_3")
model3.compile(optimizer="Adam",
loss="categorical_crossentropy",metrics=["accuracy"])

model3.summary()

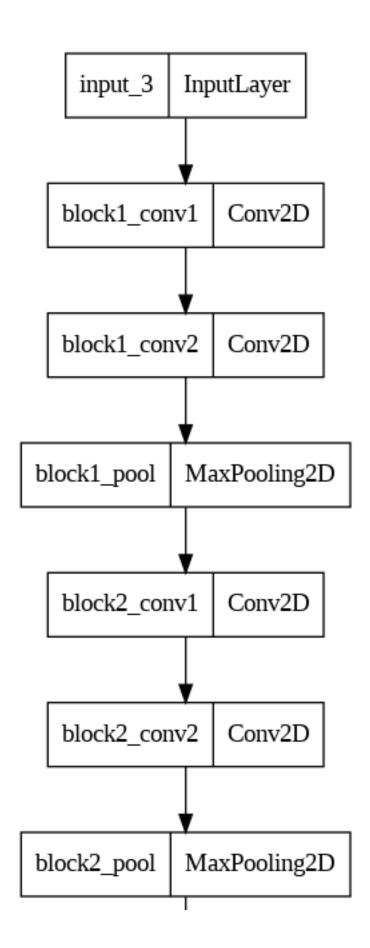
Model: "model_3"

Layer (type)	Output Shape	Param #
input_6 (InputLayer)	[(None, 128, 128, 3)]	0
block1_conv1 (Conv2D)	(None, 128, 128, 64)	1792
block1_conv2 (Conv2D)	(None, 128, 128, 64)	36928
<pre>block1_pool (MaxPooling2D)</pre>	(None, 64, 64, 64)	0
block2_conv1 (Conv2D)	(None, 64, 64, 128)	73856
block2_conv2 (Conv2D)	(None, 64, 64, 128)	147584
<pre>block2_pool (MaxPooling2D)</pre>	(None, 32, 32, 128)	0
block3_conv1 (Conv2D)	(None, 32, 32, 256)	295168
block3_conv2 (Conv2D)	(None, 32, 32, 256)	590080
block3_conv3 (Conv2D)	(None, 32, 32, 256)	590080
<pre>block3_pool (MaxPooling2D)</pre>	(None, 16, 16, 256)	0
block4_conv1 (Conv2D)	(None, 16, 16, 512)	1180160
block4_conv2 (Conv2D)	(None, 16, 16, 512)	2359808
block4_conv3 (Conv2D)	(None, 16, 16, 512)	2359808
<pre>block4_pool (MaxPooling2D)</pre>	(None, 8, 8, 512)	0
block5_conv1 (Conv2D)	(None, 8, 8, 512)	2359808
block5_conv2 (Conv2D)	(None, 8, 8, 512)	2359808
block5_conv3 (Conv2D)	(None, 8, 8, 512)	2359808
<pre>block5_pool (MaxPooling2D)</pre>	(None, 4, 4, 512)	0
conv1 (Conv2D)	(None, 4, 4, 512)	2359808
conv2 (Conv2D)	(None, 4, 4, 128)	589952

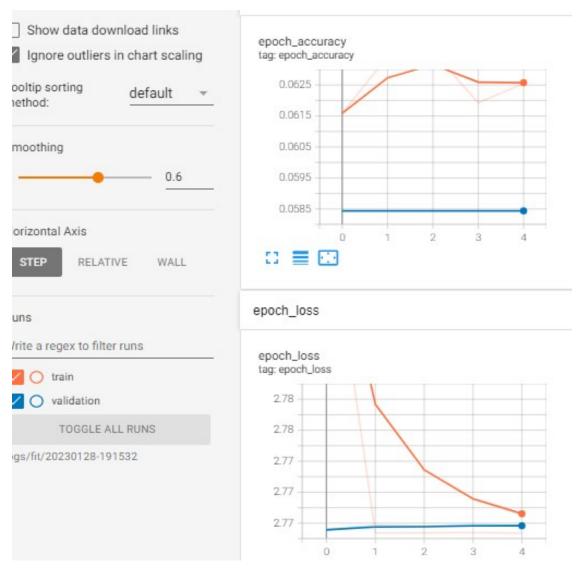
Flatten_layer (Flatten) (None, 2048) 0
dense_6 (Dense) (None, 16) 32784

Total params: 17,697,232 Trainable params: 12,421,776 Non-trainable params: 5,275,456

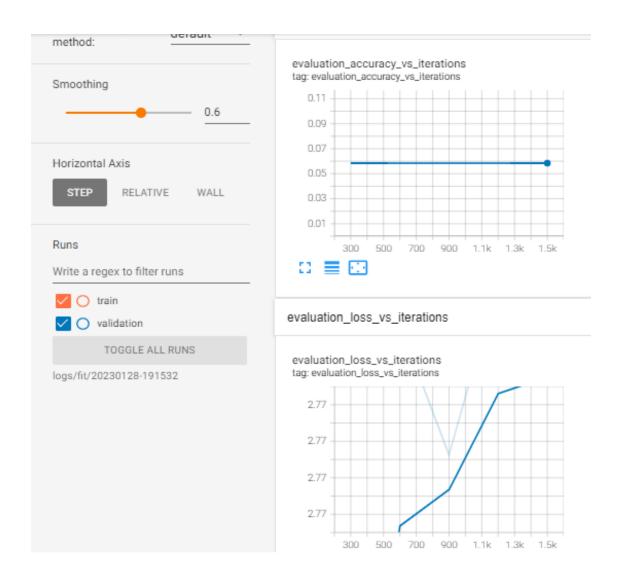
plot_model(model3)



```
history = model3.fit(train generator,
validation data=valid generator,epochs=5,
steps_per_epoch=steps,callbacks=tensorboard)
Epoch 1/5
2.7928 - accuracy: 0.0582 - val loss: 2.7727 - val accuracy: 0.0584
2.7727 - accuracy: 0.0626 - val loss: 2.7728 - val accuracy: 0.0584
Epoch 3/5
2.7727 - accuracy: 0.0626 - val loss: 2.7729 - val accuracy: 0.0584
Epoch 4/5
300/300 [============= ] - 209s 697ms/step - loss:
2.7727 - accuracy: 0.0614 - val_loss: 2.7729 - val_accuracy: 0.0584
Epoch 5/5
300/300 [============ ] - 210s 701ms/step - loss:
2.7727 - accuracy: 0.0620 - val loss: 2.7730 - val accuracy: 0.0584
%tensorboard --logdir {log dir}
<IPython.core.display.Javascript object>
display.display(Image.open('ts5.png'))
```



display.display(Image.open('ts6.png'))



Summary

```
from prettytable import PrettyTable
myTable = PrettyTable(['Model
','Max_Train_Accuracy','Max_Validation_Accuracy'])
myTable.add_row(['Model-1','0.7458','0.6894'])
myTable.add_row(['Model-2','0.7537','0.6970'])
myTable.add_row(['Model-3','0.0620','0.0584'])
print(myTable)
```

Model		Max_Validation_Accuracy
Model-1	0.7458	0.6894
Model-2	0.7537	0.6970
Model-3	0.0620	0.0584