```
In [ ]:
# references:
# https://github.com/Kaggle/kaggle-api
# https://towardsdatascience.com/downloading-datasets-into-google-drive-via-google-colab-bcb1b30b0166
# https://www.pyimagesearch.com/2018/12/24/how-to-use-keras-fit-and-fit generator-a-hands-on-tutorial/
In [ ]:
from google.colab import drive
drive.mount('/content/gdrive')
Mounted at /content/gdrive
In [ ]:
# importing necessay libraries
import tensorflow as tf
import datetime
import os
from tensorflow.keras.layers import Dense, Activation, Conv2D, Flatten, MaxPooling2D, Dropout
from tensorflow.keras import regularizers, optimizers, initializers
from tensorflow.keras.models import Model
from keras_preprocessing.image import ImageDataGenerator
import pandas as pd
import numpy as np
In [ ]:
!pip install -q kaggle
from google.colab import files
files.upload()
!mkdir ~/.kaggle
!cp kaggle.json ~/.kaggle/
! chmod 600 ~/.kaggle/kaggle.json
In [ ]:
!kaggle datasets download -d brahma0545/aaic-assignment-tl
Downloading aaic-assignment-tl.zip to /content
100% 4.32G/4.34G [01:27<00:00, 78.2MB/s]
100% 4.34G/4.34G [01:28<00:00, 52.9MB/s]
In [ ]:
!unzip "/content/aaic-assignment-tl.zip" -d "/content/TL"
In [ ]:
# label data file
dir path = "/content/TL/labels final.csv"
In [ ]:
# fetch labels final.csv
train df = pd.read csv(dir path)
In [ ]:
train df.head()
```

### Out[]:

# path label 0 imagesv/v/o/h/voh71d00/509132755+-2755.tif 3 1 imagesl/l/x/t/lx19d00/502213303.tif 3 2 imagesx/x/e/d/xed05a00/2075325674.tif 2 3 imageso/o/j/b/ojb60d00/517511301+-1301.tif 3 4 imagesq/q/z/k/qzk17e00/2031320195.tif 7

### In [ ]:

```
# replacing labels
train df = train df.replace({'label':
                            {0:"letter",
                            1:"form",
                            2:"email",
                            3: "handwritten",
                            4:"advertisement",
                            5:"scientifit report",
                             6:"scientific publication",
                            7:"specification",
                            8:"file folder",
                            9:"news article",
                            10:"budget",
                            11:"invoice",
                             12:"presentation",
                            13: "questionnaire",
                            14:"resume",
                            15:"memo"}})
```

# In [ ]:

```
# how much data for each category:
train_df['label'].value_counts()
```

# Out[]:

letter	3016
questionnaire	3007
resume	3006
presentation	3006
handwritten	3005
file folder	3003
news article	3002
budget	3002
specification	3000
scientifit report	2999
memo	2996
advertisement	2994
form	2994
email	2993
invoice	2992
scientific publication	2985
Name: label, dtype: int64	

### Observations:

1. data is balanced

# In [ ]:

```
datagen = ImageDataGenerator(rotation_range=90,width_shift_range=0.25,height_shift_range=0.25,horizonta
1_flip=0.25,vertical_flip=0.25,
    rescale=1./255,validation_split=0.30)
```

### In [ ]:

```
# train data generator
train_generator = datagen.flow_from_dataframe(
    dataframe=train_df,
    directory="/content/TL/data_final/",
    x_col = "path",
    y_col = "label",
    subset = "training",
    batch_size = 96,
    seed = 39,
    shuffle = True,
    class_mode="categorical",
    target_size = (224,224)
)
```

Found 33600 validated image filenames belonging to 16 classes.

### In [ ]:

```
# validation data generator
valid_generator = datagen.flow_from_dataframe(
    dataframe=train_df,
    directory="/content/TL/data_final/",
    x_col = "path",
    y_col = "label",
    subset = "validation",
    batch_size = 32,
    seed = 42,
    shuffle = True,
    class_mode="categorical",
    target_size = (224,224)
}
```

Found 14400 validated image filenames belonging to 16 classes.

## In [ ]:

```
%load_ext tensorboard
import tensorflow as tf
import datetime, os
```

# VGG16 Pretrained model as base model

# Model-1

### In [ ]:

```
tf.keras.backend.clear_session()
# loading vgg16 from keras
from keras.applications import VGG16
# load model
base model = VGG16(include_top=False,input_shape = (224, 224, 3),weights='imagenet')
# no need to train the VGG-16 network
for layer in base_model.layers:
    layer.trainable = False

# Conv layer
conv1 = Conv2D(256,kernel_size=(3,3),padding='valid',strides=1,activation='relu') (base_model.output)
# max pool layer
max = MaxPooling2D(pool_size=(2,2),strides=(2,2),padding='valid') (conv1)
# Flatten
flat = Flatten() (max)
# FC layer-1
```

# In [ ]:

final\_model1.summary()

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 224, 224, 3)]	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2359808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2359808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0
conv2d (Conv2D)	(None, 5, 5, 256)	1179904
max_pooling2d (MaxPooling2D)	(None, 2, 2, 256)	0
flatten (Flatten)	(None, 1024)	0
dense (Dense)	(None, 1024)	1049600
dense_1 (Dense)	(None, 256)	262400
dense_2 (Dense)	(None, 16)	4112

Total params: 17,210,704 Trainable params: 2,496,016 Non-trainable params: 14,714,688 In [ ]: chech path = "/content/gdrive/MyDrive/checkpoint/" In [ ]: log path = "/content/gdrive/MyDrive/logs/" In [ ]: # checkpoint callback: cp callback = tf.keras.callbacks.ModelCheckpoint(chech path,monitor='accuracy',verbose=1,save weights o nly=True, save freq='epoch') In [ ]: !rm -rf /content/gdrive/MyDrive/logs/ logdir = os.path.join(log\_path,datetime.datetime.now().strftime("%Y%m%d-%H%M%S")) tensorboard callback = tf.keras.callbacks.TensorBoard(logdir, histogram freq=1) vgghist = final model1.fit(train generator, validation data=valid generator, callbacks=[tensorboard callback,cp callback], steps\_per\_epoch = 350, epochs = 70Epoch 1/70 350/350 [== ======] - 915s 3s/step - loss: 2.1734 - accuracy: 0.3032 - val loss: 1 .7295 - val accuracy: 0.4519 Epoch 00001: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 2/70 350/350 [== .6136 - val accuracy: 0.4872 Epoch 00002: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 3/70 ======] - 842s 2s/step - loss: 1.6037 - accuracy: 0.4878 - val\_loss: 1 350/350 [== .5800 - val accuracy: 0.5013 Epoch 00003: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 4/70 350/350 [== =======] - 814s 2s/step - loss: 1.5538 - accuracy: 0.5039 - val loss: 1 .5648 - val accuracy: 0.5067 Epoch 00004: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 5/70 .5284 - val accuracy: 0.5182 Epoch 00005: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 6/70 350/350 [== .4938 - val accuracy: 0.5253 Epoch 00006: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 7/70 350/350 [=== .4615 - val accuracy: 0.5392 Epoch 00007: saving model to /content/gdrive/MyDrive/checkpoint/ Epoch 8/70 ======] - 796s 2s/step - loss: 1.4354 - accuracy: 0.5416 - val loss: 1 350/350 [== .4660 - val accuracy: 0.5385

Though 00000: accion model to /content/advice/McDvice/absolution/

```
Epoch UUUUU: saving model to /content/garive/MyDrive/checkpoint/
Epoch 9/70
.4604 - val accuracy: 0.5412
Epoch 00009: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 10/70
350/350 [==
          .4159 - val accuracy: 0.5517
Epoch 00010: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 11/70
               350/350 [===
.4310 - val accuracy: 0.5540
Epoch 00011: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 12/70
.4358 - val accuracy: 0.5561
Epoch 00012: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 13/70
.3903 - val accuracy: 0.5651
Epoch 00013: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 14/70
350/350 [===========] - 751s 2s/step - loss: 1.3503 - accuracy: 0.5734 - val loss: 1
.3868 - val accuracy: 0.5667
Epoch 00014: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 15/70
350/350 [===
                     .3890 - val_accuracy: 0.5671
Epoch 00015: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 16/70
350/350 [========
                      =======] - 753s 2s/step - loss: 1.3496 - accuracy: 0.5706 - val loss: 1
.3716 - val_accuracy: 0.5697
Epoch 00016: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 17/70
350/350 [====
                      ======] - 760s 2s/step - loss: 1.3388 - accuracy: 0.5762 - val loss: 1
.3882 - val accuracy: 0.5644
Epoch 00017: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 18/70
                     350/350 [===
.3684 - val accuracy: 0.5705
Epoch 00018: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 19/70
          350/350 [====
.3647 - val_accuracy: 0.5728
Epoch 00019: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 20/70
.3892 - val accuracy: 0.5652
Epoch 00020: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 21/70
350/350 [===
                     ======] - 712s 2s/step - loss: 1.3088 - accuracy: 0.5891 - val loss: 1
.3780 - val_accuracy: 0.5699
Epoch 00021: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 22/70
350/350 [===
                      =======] - 724s 2s/step - loss: 1.2951 - accuracy: 0.5963 - val loss: 1
.3691 - val accuracy: 0.5776
Epoch 00022: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 23/70
350/350 [===========
                      =======] - 717s 2s/step - loss: 1.2962 - accuracy: 0.5904 - val loss: 1
.3816 - val accuracy: 0.5708
Epoch 00023: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 24/70
                              710- 0-/---- 1---- 1 0700 ------- 0 5016 ---1 1---- 1
250/250
```

```
.3463 - val accuracy: 0.5869
Epoch 00024: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 25/70
.3454 - val accuracy: 0.5793
Epoch 00025: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 26/70
                   350/350 [===
.3782 - val accuracy: 0.5710
Epoch 00026: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 27/70
350/350 [=====
                   .3563 - val accuracy: 0.5779
Epoch 00027: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 28/70
.3554 - val accuracy: 0.5797
Epoch 00028: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 29/70
.3964 - val accuracy: 0.5684
Epoch 00029: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 30/70
350/350 [===========] - 723s 2s/step - loss: 1.2658 - accuracy: 0.5977 - val loss: 1
.3586 - val accuracy: 0.5753
Epoch 00030: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 31/70
350/350 [===========] - 718s 2s/step - loss: 1.2622 - accuracy: 0.6019 - val loss: 1
.3275 - val accuracy: 0.5916
Epoch 00031: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 32/70
350/350 [===========] - 719s 2s/step - loss: 1.2348 - accuracy: 0.6114 - val loss: 1
.3432 - val accuracy: 0.5852
Epoch 00032: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 33/70
                     ======] - 720s 2s/step - loss: 1.2425 - accuracy: 0.6053 - val loss: 1
350/350 [==
.3336 - val accuracy: 0.5827
Epoch 00033: saving model to /content/gdrive/MyDrive/checkpoint/
.3143 - val accuracy: 0.5983
Epoch 00034: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 35/70
.3398 - val accuracy: 0.5835
Epoch 00035: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 36/70
350/350 [======
                   =======] - 737s 2s/step - loss: 1.2285 - accuracy: 0.6121 - val loss: 1
.3523 - val accuracy: 0.5806
Epoch 00036: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 37/70
350/350 [===========] - 731s 2s/step - loss: 1.2358 - accuracy: 0.6061 - val loss: 1
.3202 - val accuracy: 0.5956
Epoch 00037: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 38/70
.3288 - val accuracy: 0.5933
Epoch 00038: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 39/70
350/350 [============] - 724s 2s/step - loss: 1.2187 - accuracy: 0.6158 - val loss: 1
```

.3121 - val accuracy: 0.5908

```
Epoch 00039: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 40/70
         350/350 [===
.3183 - val_accuracy: 0.5913
Epoch 00040: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 41/70
                    350/350 [==
.3309 - val accuracy: 0.5928
Epoch 00041: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 42/70
350/350 [==
                  In [ ]:
final model1.load weights (chech path)
Out[]:
<tensorflow.python.training.tracking.util.CheckpointLoadStatus at 0x7f8b901a89e8>
In [ ]:
logdir = os.path.join(log path,datetime.datetime.now().strftime("%Y%m%d-%H%M%S"))
tensorboard callback = tf.keras.callbacks.TensorBoard(logdir, histogram freq=1)
vgghist = final model1.fit(train generator,
                 validation data=valid generator,
                 callbacks=[tensorboard callback,cp callback],
                 steps per epoch = 350,
                 epochs = 30
Epoch 1/30
350/350 [==
                   =======] - 730s 2s/step - loss: 1.2161 - accuracy: 0.6130 - val loss: 1
.3257 - val accuracy: 0.5862
Epoch 00001: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 2/30
350/350 [==
                   .2997 - val accuracy: 0.6022
Epoch 00002: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 3/30
.3140 - val accuracy: 0.5972
Epoch 00003: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 4/30
.3197 - val_accuracy: 0.5997
Epoch 00004: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 5/30
350/350 [==
        .3112 - val accuracy: 0.5978
Epoch 00005: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 6/30
       350/350 [==
.3247 - val accuracy: 0.6001
Epoch 00006: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 7/30
        350/350 [==
.3014 - val accuracy: 0.6022
Epoch 00007: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 8/30
350/350 [==
        .3258 - val accuracy: 0.5946
```

```
Epoch 00008: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 9/30
350/350 [==
        .3099 - val accuracy: 0.6037
Epoch 00009: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 10/30
350/350 [===========] - 726s 2s/step - loss: 1.1924 - accuracy: 0.6194 - val loss: 1
.2875 - val accuracy: 0.6036
Epoch 00010: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 11/30
.2921 - val_accuracy: 0.6044
Epoch 00011: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 12/30
350/350 [=======
                  .3261 - val accuracy: 0.5965
Epoch 00012: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 13/30
.2885 - val_accuracy: 0.6028
Epoch 00013: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 14/30
        350/350 [===
.3016 - val_accuracy: 0.6069
Epoch 00014: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 15/30
350/350 [===
                 .3641 - val_accuracy: 0.5867
Epoch 00015: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 16/30
.2984 - val accuracy: 0.6006
Epoch 00016: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 17/30
350/350 [===
                   .3072 - val accuracy: 0.6019
Epoch 00017: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 18/30
                   ======] - 758s 2s/step - loss: 1.1687 - accuracy: 0.6279 - val loss: 1
350/350 [===
.3305 - val accuracy: 0.5994
Epoch 00018: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 19/30
.2936 - val accuracy: 0.6016
Epoch 00019: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 20/30
.3154 - val accuracy: 0.6049
Epoch 00020: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 21/30
                   350/350 [===
.3148 - val accuracy: 0.6047
Epoch 00021: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 22/30
350/350 [=======
                 ========] - 725s 2s/step - loss: 1.1667 - accuracy: 0.6303 - val loss: 1
.3030 - val accuracy: 0.6045
Epoch 00022: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 23/30
350/350 [========
                  .3192 - val accuracy: 0.5978
Epoch 00023: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 24/30
```

```
.2934 - val_accuracy: 0.6082
Epoch 00024: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 25/30
350/350 [==
                       ======] - 743s 2s/step - loss: 1.1588 - accuracy: 0.6310 - val loss: 1
.2896 - val accuracy: 0.6073
Epoch 00025: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 26/30
                       =======] - 709s 2s/step - loss: 1.1529 - accuracy: 0.6327 - val loss: 1
350/350 [==
.2894 - val accuracy: 0.6070
Epoch 00026: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 27/30
350/350 [===
                      .3281 - val accuracy: 0.5948
Epoch 00027: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 28/30
.3081 - val accuracy: 0.6065
Epoch 00028: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 29/30
.3076 - val accuracy: 0.6017
Epoch 00029: saving model to /content/gdrive/MyDrive/checkpoint/
Epoch 30/30
350/350 [==
                      ========] - 714s 2s/step - loss: 1.1450 - accuracy: 0.6364 - val loss: 1
.3097 - val accuracy: 0.6069
Epoch 00030: saving model to /content/gdrive/MyDrive/checkpoint/
In [ ]:
#!tensorboard dev upload --logdir /content/gdrive/MyDrive/logs/20210202-154031 \
# --name "Transfer Learning Model:1 (initial 42 epochs)" \
   --description " from TF 1.ipynb " \
  --one shot
In [ ]:
# !tensorboard dev upload --logdir /content/gdrive/MyDrive/logs/20210203-061009 \
  --name "Transfer Learning Model:1 (next 30 epochs)" \
  --description " from TF_1.ipynb " \
#
  --one shot
In [ ]:
```