```
import os
os.environ['TF_CPP_MIN_LOG_LEVEL'] = '2'
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Batc
from tensorflow.keras.optimizers import Adam, Adamax
from tensorflow.keras.metrics import binary_crossentropy
from tensorflow.keras import regularizers
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Model, Sequential
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from PIL import Image
from IPython.core.display import display, HTML
import logging
logging.getLogger("tensorflow").setLevel(logging.ERROR)
from pathlib import PosixPath
import random
import matplotlib.pyplot as plt
%matplotlib inline
from keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau
import seaborn as sns
from sklearn.metrics import f1_score, confusion_matrix
                                ____ication_report
                                > onMatrixDisplay
 Başarıyla kaydedildi.
seed = random.randint(0,100)
random.seed(seed)
seed = random.randint(0,1250)
print(f"Current seed : {seed}")
     Current seed: 384
root_path = '/content/drive/MyDrive/archive/train'
data_dir = PosixPath(root_path)
image count = len(list(data dir.glob("*/*")))
print(f"Image count: {image_count}")
     Image count: 340
class_names = os.listdir(root_path)
class_names
```

```
['pepsi', 'cocacola']

class_distribution = [len(os.listdir(root_path + '/' + name)) for name in class_names]
class_distribution

   [170, 170]

img_path=r'/content/drive/MyDrive/archive/train/cocacola/100.jpg'
img=plt.imread(img_path)
print (img.shape)
plt.axis('off')
plt.imshow(img)
plt.show()

   (270, 230, 3)
```



def preprocess (sdir, trsplit, random_seed):

```
filepaths=[]
  labels=[]
Başarıyla kaydedildi.
   for klass in classlist:
       classpath=os.path.join(sdir,klass)
       flist=os.listdir(classpath)
       for f in flist:
           fpath=os.path.join(classpath,f)
           filepaths.append(fpath)
           labels.append(klass)
   Fseries=pd.Series(filepaths, name='filepaths')
   Lseries=pd.Series(labels, name='labels')
  df=pd.concat([Fseries, Lseries], axis=1)
   # split df into train_df and test_df
  dsplit=(1-trsplit)
   strat=df['labels']
   train_df, valid_df=train_test_split(df, train_size=trsplit, shuffle=True, random_state
```

```
print('train_df length: ', len(train_df), ' valid_df length: ', len(valid_df))
    return train df, valid df
sdir=r'/content/drive/MyDrive/archive/train'
train_df, valid_df = preprocess(sdir, .8, seed)
    train_df length: 272 valid_df length: 68
filepaths=[]
labels=[]
classlist = os.listdir('/content/drive/MyDrive/archive/test')
for klass in classlist:
   classpath=os.path.join(sdir,klass)
   flist=os.listdir(classpath)
   for f in flist:
       fpath=os.path.join(classpath,f)
       filepaths.append(fpath)
       labels.append(klass)
Fseries=pd.Series(filepaths, name='filepaths')
Lseries=pd.Series(labels, name='labels')
test_df=pd.concat([Fseries, Lseries], axis=1)
print('test_df length: ', len(test_df))
    test_df length: 340
 Başarıyla kaydedildi.
img_shape=(img_size[0], img_size[1], channels)
length=len(test_df)
test_batch_size=sorted([int(length/n) for n in range(1,length+1) if length % n ==0 and ler
test_steps=int(length/test_batch_size)
print ( 'test batch size: ' ,test_batch_size, ' test steps: ', test_steps)
     test batch size: 68
                          test steps: 5
def scalar(img):
    return img # EfficientNet expects pixels in range 0 to 255 so no scaling is required
trgen=ImageDataGenerator(preprocessing function=scalar, horizontal flip=True)
tvgen=ImageDataGenerator(preprocessing_function=scalar)
train_gen=trgen.flow_from_dataframe( train_df, x_col='filepaths', y_col='labels', target_s
                                    color mode='rgb', shuffle=True, batch size=batch size)
test_gen=tvgen.flow_from_dataframe( test_df, x_col='filepaths', y_col='labels', target_siz
                                    color_mode='rgb', shuffle=False, batch_size=test_batch
```

```
valid_gen=tvgen.flow_from_dataframe( valid_df, x_col='filepaths', y_col='labels', target_s
                                    color_mode='rgb', shuffle=True, batch_size=batch_size)
classes=list(train_gen.class_indices.keys())
     Found 272 validated image filenames belonging to 2 classes.
     Found 340 validated image filenames belonging to 2 classes.
     Found 68 validated image filenames belonging to 2 classes.
class count=len(classes)
print(class_count)
     2
best_model = ModelCheckpoint('/content/drive/MyDrive/bestmodel.hdf5', monitor='accuracy',
best_val_acc = ModelCheckpoint('/content/drive/MyDrive/best_val_acc.hdf5', monitor='val_ac
model = Sequential([
Conv2D(32, (3, 3), activation='relu', input_shape=(100, 100, channels)),
BatchNormalization(),
MaxPooling2D(pool_size=(2, 2)),
Dropout(0.25),
Conv2D(64, (3, 3), activation='relu'),
BatchNormalization(),
MaxPooling2D(pool_size=(2, 2)),
Dropout(0.25),
Conv2D(128, (3, 3), activation='relu'),
BatchNormalization().
 Başarıyla kaydedildi.
Flatten(),
Dense(512, activation='relu'),
BatchNormalization(),
Dropout(0.5),
Dense(2, activation='softmax'),
])
opt = tf.keras.optimizers.Adam(learning_rate=0.001, decay=1e-6)
model.compile(loss='binary_crossentropy', optimizer= opt, metrics=['accuracy'])
history=model.fit(x=train_gen, epochs=5, verbose=2,validation_data=valid_gen, callbacks=[
    best_model, best_val_acc],
    shuffle=True, initial epoch=0)
     Epoch 1/5
     9/9 - 12s - loss: 0.6005 - accuracy: 0.7684 - val_loss: 2.9342 - val_accuracy: 0.750
     Epoch 2/5
     9/9 - 8s - loss: 0.1660 - accuracy: 0.9375 - val_loss: 0.2019 - val_accuracy: 0.9559
```

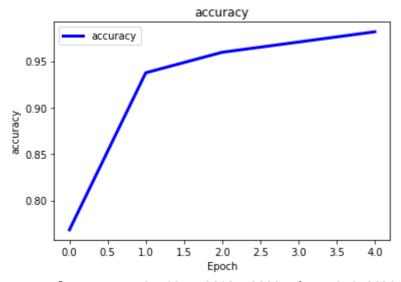
```
Epoch 3/5
9/9 - 8s - loss: 0.1372 - accuracy: 0.9596 - val_loss: 0.7882 - val_accuracy: 0.8529
Epoch 4/5
9/9 - 8s - loss: 0.1003 - accuracy: 0.9706 - val_loss: 0.5809 - val_accuracy: 0.8676
Epoch 5/5
9/9 - 8s - loss: 0.0488 - accuracy: 0.9816 - val_loss: 0.3212 - val_accuracy: 0.8971
```

```
def visualization(name,h,color):
    t = h.history[name]
    my_max = max(t)
    my_min = min(t)
    print(f'Name : {name} max : {my_max} min : {my_min}')
    plt.plot(t,color=color,linewidth=3.0)
    plt.title(name)
    plt.ylabel(name)
    plt.xlabel('Epoch')
    plt.legend([name],loc='upper left')
    plt.show()

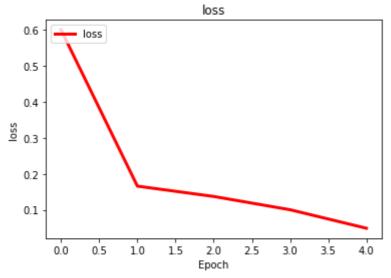
visualization('accuracy',history,'Blue')
visualization('loss',history,'Red')
visualization('val_accuracy',history,'Green')
visualization('val_loss',history,'Black')
```

Başarıyla kaydedildi.

Name : accuracy max : 0.9816176295280457 min : 0.7683823704719543



Name : loss max : 0.6005160808563232 min : 0.048830702900886536



```
Başarıyla kaydedildi. X racy
```

model.load_weights('/content/drive/MyDrive/bestmodel.hdf5')
accuracy=model.evaluate (test_gen, steps=test_steps)[1]
print(f'accuracy on test set = {accuracy:6.2f}')

```
• 5.55
```

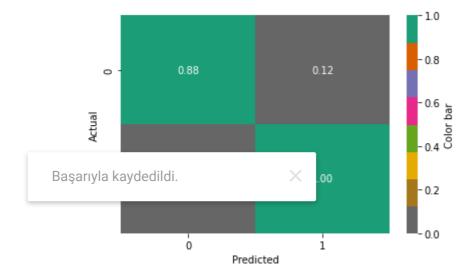
```
predictions = model.predict_generator(test_gen, steps=test_steps)
test_steps_per_epoch = np.math.ceil(test_gen.samples / test_batch_size)
predicted_classes = np.argmax(predictions, axis=1)
true_classes = test_gen.classes
class_labels = list(test_gen.class_indices.keys())
report = classification_report(true_classes, predicted_classes, target_names=class_labels)
print(report)
```

cm=confusion_matrix(true_classes,predicted_classes)
print(cm)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: UserWarning: `Model. """Entry point for launching an IPython kernel.

```
recall f1-score
              precision
                                                support
                    1.00
                              0.88
                                         0.93
    cocacola
                                                    170
                    0.89
                              1.00
                                         0.94
                                                     170
       pepsi
                                         0.94
    accuracy
                                                    340
   macro avg
                    0.95
                              0.94
                                         0.94
                                                     340
                                         0.94
                    0.95
                              0.94
                                                    340
weighted avg
```

```
[[149 21]
[ 0 170]]
```



```
model.load_weights('/content/drive/MyDrive/best_val_acc.hdf5')
accuracy=model.evaluate (test_gen, steps=test_steps)[1]
print(f'accuracy on test set = {accuracy:6.2f}')

predictions = model.predict_generator(test_gen, steps=test_steps)
test_steps_per_epoch = np.math.ceil(test_gen.samples / test_batch_size)
predicted_classes = np.argmax(predictions, axis=1)
true_classes = test_gen.classes
class_labels = list(test_gen.classe_indices.keys())
report = classification_report(true_classes, predicted_classes, target_names=class_labels)
print(report)

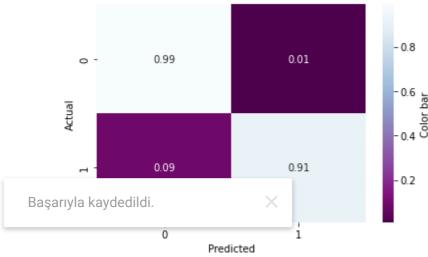
cm=confusion_matrix(true_classes, predicted_classes)
print(cm)
```

1/ / ± a\ L / a a\ a\

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:5: UserWarning: `Model. """

	precision	recall	f1-score	support
cocacola	0.92	0.99	0.95	170
pepsi	0.99	0.91	0.95	170
accuracy			0.95	340
macro avg	0.96	0.95	0.95	340
weighted avg	0.96	0.95	0.95	340





model.summary()

Model: "sequential_4"

Layer (type)	Output Shape	Param #
conv2d_12 (Conv2D)	(None, 98, 98, 32)	896
<pre>batch_normalization_16 (Bat chNormalization)</pre>	(None, 98, 98, 32)	128
<pre>max_pooling2d_12 (MaxPoolin g2D)</pre>	(None, 49, 49, 32)	0
dropout_16 (Dropout)	(None, 49, 49, 32)	0
conv2d_13 (Conv2D)	(None, 47, 47, 64)	18496

<pre>batch_normalization_17 (Bat chNormalization)</pre>	(None, 47, 47, 64)	256
<pre>max_pooling2d_13 (MaxPoolin g2D)</pre>	(None, 23, 23, 64)	0
dropout_17 (Dropout)	(None, 23, 23, 64)	0
conv2d_14 (Conv2D)	(None, 21, 21, 128)	73856
<pre>batch_normalization_18 (Bat chNormalization)</pre>	(None, 21, 21, 128)	512
<pre>max_pooling2d_14 (MaxPoolin g2D)</pre>	(None, 10, 10, 128)	0
dropout_18 (Dropout)	(None, 10, 10, 128)	0
flatten_4 (Flatten)	(None, 12800)	0
dense_8 (Dense)	(None, 512)	6554112
<pre>batch_normalization_19 (Bat chNormalization)</pre>	(None, 512)	2048
dropout_19 (Dropout)	(None, 512)	0
dense_9 (Dense)	(None, 2)	1026

Total params: 6,651,330 Trainable params: 6,649,858 Non-trainable params: 1,472

Başarıyla kaydedildi.

v_shapes=True)

