

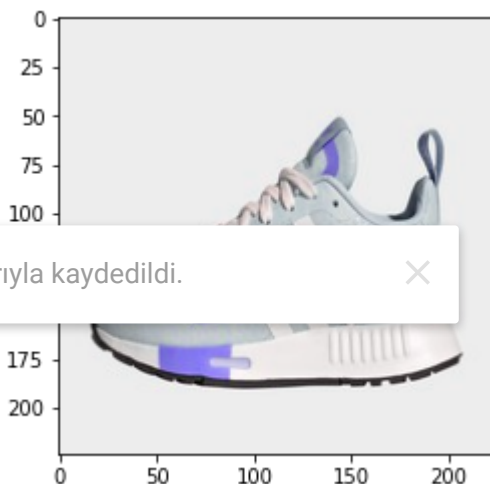
```

import time
start = time.time()

import os
import cv2
import random
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
import tensorflow as tf
import keras
import keras.metrics
from keras.models import Sequential
from keras.layers import Flatten, MaxPooling2D, Conv2D, Dense
from sklearn.metrics import f1_score, confusion_matrix, ConfusionMatrixDisplay
from sklearn.pipeline import Pipeline, make_pipeline
from keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau

img_array = cv2.imread('/content/drive/MyDrive/archive (1)/test/adidas/Adidas (18).jpg')
# img_array = cv2.cvtColor(img_array, cv2.COLOR_BGR2RGB)
plt.imshow(img_array)
plt.show()

```



```

data = '/content/drive/MyDrive/archive (1)/train'
categories = ['adidas', 'nike']
# For test our create_data function will be work
for ct in categories:
    path = os.path.join(data, ct)
    for img in os.listdir(path):
        img_array = cv2.imread(os.path.join(path, img))
        # img_array = cv2.cvtColor(img_array, cv2.COLOR_BGR2RGB)
        plt.imshow(img_array)
        plt.show()
        break
    break

```



```
print(img_array.shape)
```

```
(224, 224, 3)
```

```
IMG_Size = 150
new_array = cv2.resize(img_array, (IMG_Size,IMG_Size))
plt.imshow(new_array)
plt.show()
```



```
training_data = []
test_data = []
val_data = []

def create_data(my_data_path,my_data):
    for ct in categories:
        path = os.path.join(my_data_path,ct)
        class_num = categories.index(ct)
        for img in os.listdir(path):
            datagen = tf.keras.preprocessing.image.ImageDataGenerator(rescale=1./225)
            generator = datagen.flow_from_directory(my_data_path, shuffle=True, batch_size=32)
            try:
                img_array = cv2.imread(os.path.join(path,img))
                # img_array = cv2.cvtColor(img_array,cv2.COLOR_GRAY2RGB)
                new_array = cv2.resize(img_array, (IMG_Size,IMG_Size))
                my_data.append([new_array,class_num])
```

```
except Exception as e:
    pass

test_data_path = '/content/drive/MyDrive/archive (1)/test'
train_data_path = '/content/drive/MyDrive/archive (1)/test'
val_data_path = '/content/drive/MyDrive/archive (1)/validation'
create_data(train_data_path, training_data)
create_data(test_data_path, test_data)
create_data(val_data_path, val_data)

random.shuffle(training_data)
random.shuffle(test_data)

for sample in training_data[:10]: # For test shuffle
    print(sample[1])

1
0
1
0
0
1
1
0
1
0

x = []
y = []

for features, label in training_data:
    x.append(features)

x_test = []
y_test = []

for features, label in test_data:
    x_test.append(features)
    y_test.append(label)

x_val = []
y_val = []

for features, label in val_data:
    x_val.append(features)
    y_val.append(label)

x = np.array(x)
x_test = np.array(x_test)
x_val = np.array(x_val)
y = np.array(y)
y_test = np.array(y_test)
y_val = np.array(y_val)
```

Başarıyla kaydedildi.



```
x.shape,y.shape,x_test.shape,y_test.shape,x_val.shape,y_val.shape
```

```
((60, 150, 150, 3), (60,), (60, 150, 150, 3), (60,), (55, 150, 150, 3), (55,))
```

```
x = np.array(x).reshape(-1, IMG_Size, IMG_Size, 3)
```

```
x_test = np.array(x_test).reshape(-1, IMG_Size, IMG_Size, 3)
```

```
x_val = np.array(x_val).reshape(-1, IMG_Size, IMG_Size, 3)
```

```
x.shape,y.shape,x_test.shape,y_test.shape,x_val.shape,y_val.shape
```

```
((60, 150, 150, 3), (60,), (60, 150, 150, 3), (60,), (55, 150, 150, 3), (55,))
```

```
y = y.reshape(-1,1)
```

```
y_test = y_test.reshape(-1,1)
```

```
y_val = y_val.reshape(-1,1)
```

```
from sklearn.preprocessing import OneHotEncoder
```

```
enc = OneHotEncoder(handle_unknown='ignore')
```

```
enc.fit(y)
```

```
enc.fit(y_test)
```

```
enc.fit(y_val)
```

```
x.shape,y.shape,x_test.shape,y_test.shape,x_val.shape,y_val.shape
```

```
((60, 150, 150, 3),
```

```
(60, 1),
```

```
(60, 150, 150, 3),
```

Başarıyla kaydedildi.

```
(55, 1))
```

```
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()
```

```
model.add(Conv2D(128, (3,3), input_shape=(150, 150, 3) , activation='relu'))
```

```
model.add(MaxPooling2D(2,2))
```

```
model.add(Conv2D(256, (3,3), activation='relu'))
```

```
model.add(MaxPooling2D(2,2))
```

```
model.add(Conv2D(256, (3,3), activation='relu'))
```

```
model.add(Flatten())
```

```
model.add(Dense(256, activation='relu'))
```

```
model.add(Dense(128, activation='relu'))
```

```
model.add(tf.keras.layers.Dropout(.2))
```

```
model.add(Dense(6, activation='softmax'))
```

```
opt = tf.keras.optimizers.Adam(learning_rate=0.001, decay=1e-6)
```

```
# Compile model
model.compile(
    loss='sparse_categorical_crossentropy', # binary_crossentropy, sparse_categorical_crossentropy
    optimizer=opt,
    run_eagerly=True,
    metrics=['accuracy']
)
```

```
hist = model.fit(x, y, validation_data=(x_val,y_val), verbose=2, epochs=20, shuffle=True,
```

```
Epoch 1/20
2/2 - 35s - loss: 575.5352 - accuracy: 0.2500 - val_loss: 1714.5830 - val_accuracy:
Epoch 2/20
2/2 - 32s - loss: 954.5818 - accuracy: 0.5000 - val_loss: 45.8089 - val_accuracy: 0.
Epoch 3/20
2/2 - 35s - loss: 33.4230 - accuracy: 0.5167 - val_loss: 6.2965 - val_accuracy: 0.50
Epoch 4/20
2/2 - 34s - loss: 3.0799 - accuracy: 0.6667 - val_loss: 0.8517 - val_accuracy: 0.454
Epoch 5/20
2/2 - 46s - loss: 1.1580 - accuracy: 0.7333 - val_loss: 1.1766 - val_accuracy: 0.672
Epoch 6/20
2/2 - 37s - loss: 0.7841 - accuracy: 0.8167 - val_loss: 3.7669 - val_accuracy: 0.545
Epoch 7/20
2/2 - 28s - loss: 1.1789 - accuracy: 0.7667 - val_loss: 3.3433 - val_accuracy: 0.490
Epoch 8/20
2/2 - 32s - loss: 2.0127 - accuracy: 0.8000 - val_loss: 2.1312 - val_accuracy: 0.618
Epoch 9/20
2/2 - 34s - loss: 0.3079 - accuracy: 0.9167 - val_loss: 2.9074 - val_accuracy: 0.490
Epoch 10/20
2/2 - 37s - loss: 0.2324 - accuracy: 0.9500 - val_loss: 1.9650 - val_accuracy: 0.563
Epoch 11/20
2/2 - 35s - loss: 0.1417 - accuracy: 0.9667 - val_loss: 2.0513 - val_accuracy: 0.600
Epoch 12/20
2/2 - 35s - loss: 0.0991 - accuracy: 0.9167 - val_loss: 4.1698 - val_accuracy: 0.545
Epoch 13/20
2/2 - 35s - loss: 0.0991 - accuracy: 0.9167 - val_loss: 1.9645 - val_accuracy: 0.581
Epoch 14/20
2/2 - 31s - loss: 0.2179 - accuracy: 0.9833 - val_loss: 2.4311 - val_accuracy: 0.527
Epoch 15/20
2/2 - 29s - loss: 0.3749 - accuracy: 0.9167 - val_loss: 1.9064 - val_accuracy: 0.563
Epoch 16/20
2/2 - 24s - loss: 0.2024 - accuracy: 0.9667 - val_loss: 1.9749 - val_accuracy: 0.563
Epoch 17/20
2/2 - 30s - loss: 0.0680 - accuracy: 1.0000 - val_loss: 2.3094 - val_accuracy: 0.563
Epoch 18/20
2/2 - 27s - loss: 0.0367 - accuracy: 1.0000 - val_loss: 3.1910 - val_accuracy: 0.600
Epoch 19/20
2/2 - 25s - loss: 0.0444 - accuracy: 0.9833 - val_loss: 5.2440 - val_accuracy: 0.545
Epoch 20/20
2/2 - 25s - loss: 0.5225 - accuracy: 0.9167 - val_loss: 2.6957 - val_accuracy: 0.618
```

Başarıyla kaydedildi.

```
hist.history??
```

```
def visualization(name,h,color):
    t = h.history[name]
```

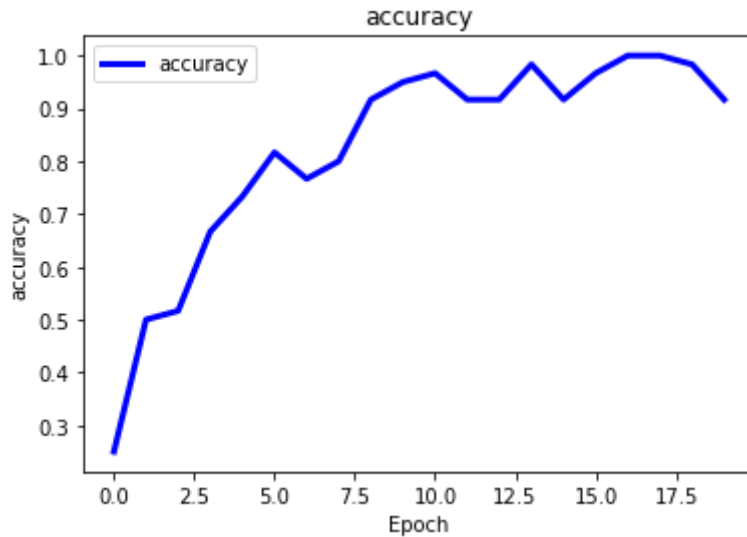
```
my_max = max(t)
my_min = min(t)
print(f'Name : {name:10} max : {my_max:10} min : {my_min:10}')
plt.plot(t,color=color,linewidth=3.0)
plt.title(name)
plt.ylabel(name)
plt.xlabel('Epoch')
plt.legend([name],loc='upper left')
plt.show()

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

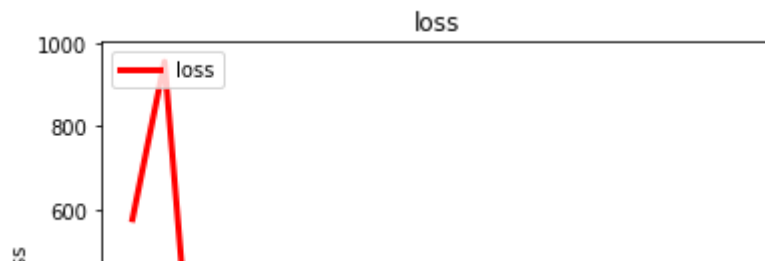
Başarıyla kaydedildi.



Name : accuracy max : 1.0 min : 0.25



Name : loss max : 954.581787109375 min : 0.03665284439921379



```
model.load_weights('/content/drive/MyDrive/bestmodel.hdf5')
res = model.evaluate(x_test, y_test)
print("test loss, test acc:", res)
```

```
2/2 [=====] - 5s 2s/step - loss: 0.0461 - accuracy: 1.0000
test loss, test acc: [0.04606626182794571, 1.0]
```

Epoch

Başarıyla kaydedildi.



MyDrive/best_val_acc.hdf5')

)

```
2/2 [=====] - 5s 2s/step - loss: 0.6011 - accuracy: 0.9500
test loss, test acc: [0.6010614633560181, 0.949999988079071]
```

```
def my_predict(my_model, my_x_test):
```

```
    y_pred = my_model.predict(my_x_test)
    return y_pred
```

```
def my_f1_score(my_y_test, my_y_pred):
    f1 = f1_score(my_y_test, my_y_pred, average="micro")
    return f1
```

```
def my_conf_matrix(my_y_test, my_y_pred):
    cm = confusion_matrix(my_y_test, my_y_pred)
    cm_norm = np.round(cm/np.sum(cm, axis=1).reshape(-1,1), 2)
    sns.heatmap(cm_norm, cmap='Greens', annot=True,
                cbar_kws={'orientation' : 'vertical', 'label' : 'Color bar'},
                fmt='.2f')
    )
```

```
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
# cm_display = ConfusionMatrixDisplay(cm)
# cm_display.plot()
# plt.show()
```

```
y_pred_res = my_predict(model,x_test)
y_pred_res = np.argmax(y_pred_res, axis=-1)
```

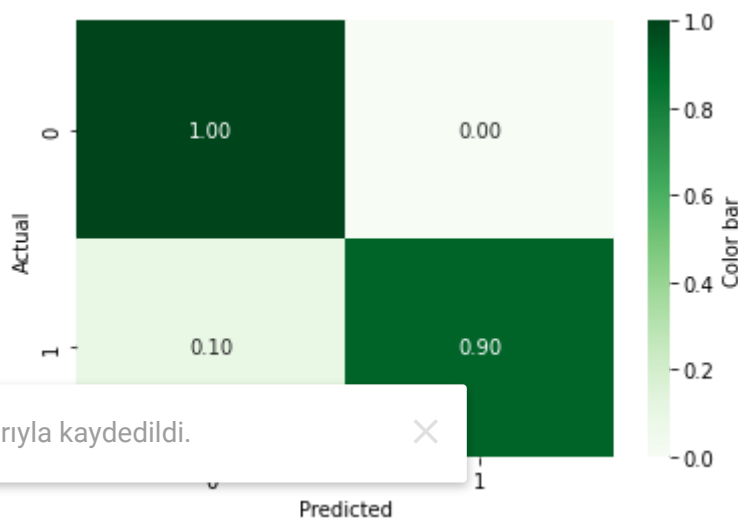
```
y_pred_res.shape,y_test.shape
```

```
((60,), (60, 1))
```

```
print(my_f1_score(y_test,y_pred_res))
```

```
0.9500000000000001
```

```
my_conf_matrix(y_test,y_pred_res)
```



```
newpath = r'/content/drive/MyDrive/Model'
if not os.path.exists(newpath):
    os.makedirs(newpath)
```

```
import pickle
# To save model || You can use tensorflow model.save
pickle_out = open('/content/drive/MyDrive/Model/model.pickle','wb')
pickle.dump(model,pickle_out)
pickle_out.close()
```

```
model.summary()
```

```
Model: "sequential_9"
```

Layer (type)	Output Shape	Param #
=====		

conv2d_27 (Conv2D)	(None, 148, 148, 128)	3584
max_pooling2d_18 (MaxPooling2D)	(None, 74, 74, 128)	0
conv2d_28 (Conv2D)	(None, 72, 72, 256)	295168
max_pooling2d_19 (MaxPooling2D)	(None, 36, 36, 256)	0
conv2d_29 (Conv2D)	(None, 34, 34, 256)	590080
flatten_9 (Flatten)	(None, 295936)	0
dense_23 (Dense)	(None, 256)	75759872
dense_24 (Dense)	(None, 128)	32896
dropout_9 (Dropout)	(None, 128)	0
dense_25 (Dense)	(None, 6)	774

```

=====
Total params: 76,682,374
Trainable params: 76,682,374
Non-trainable params: 0

```

```

end = time.time()
print((end - start)/60)

```

```
30.865265921751657
```

Başarıyla kaydedildi.



Drive/Model/model.pickle', 'rb')

```
model = pickle.load(pickle_in)
```