DRIVER DROWSINESS DETECTION

```
import numpy as np
import pandas as pd
from glob import glob
import os
import PIL
import tensorflow as tf
import pathlib
from tensorflow import keras
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
```

C:\Users\user\anaconda3\lib\site-packages\scipy__init__.py:155: UserWarning: A NumPy
 warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"
WARNING:tensorflow:From C:\Users\user\anaconda3\lib\site-packages\keras\src\losses.py</pre>

```
data_path = r"C:\Users\user\Downloads\drowsiness 1\eyes open closed"
all_faces = glob(r"C:\Users\user\Downloads\drowsiness 1\eyes open closed/*/*")
print(len(all_faces))
```

→ 1452

```
batch_size = 64
num_classes = 2
img_height = 180
img_width = 180
```

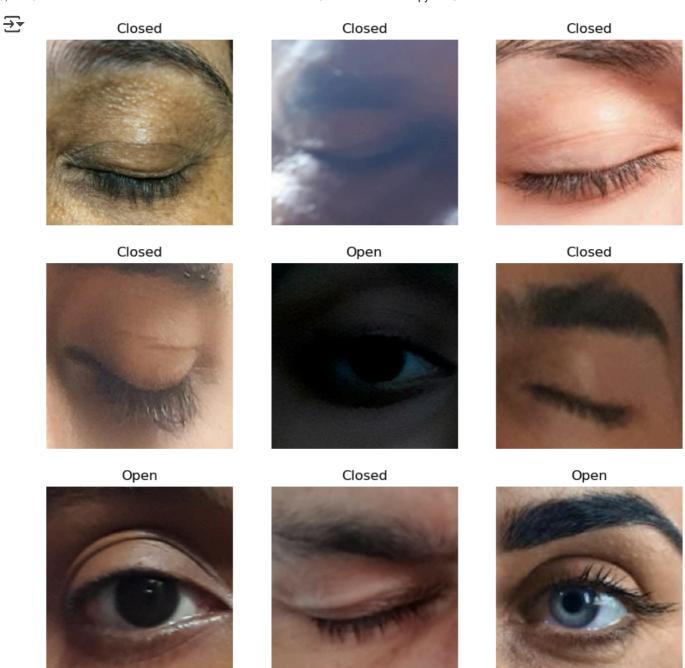
```
Found 1452 files belonging to 2 classes.
Using 1089 files for training.
Found 1452 files belonging to 2 classes.
Using 363 files for validation.
```

```
class_names = train_ds.class_names
print(class_names)
```

```
→ ['Closed', 'Open']
```

```
import matplotlib.pyplot as plt

plt.figure(figsize=(10, 10))
for images, labels in train_ds.take(1):
    for i in range(9):
        ax = plt.subplot(3, 3, i + 1)
        ax.imshow(images[i].numpy().astype("uint8"))
        plt.title(class_names[labels[i]])
        plt.axis("off");
```



First CNN

```
layers.Conv2D(64, 3, padding='same', activation='relu'),
layers.MaxPooling2D(),
layers.Dropout(0.2),
layers.Flatten(),
layers.Dense(128, activation='relu'),
layers.Dense(num_classes, activation = 'sigmoid')
])
```

WARNING:tensorflow:From C:\Users\user\anaconda3\lib\site-packages\keras\src\backend.p

WARNING:tensorflow:From C:\Users\user\anaconda3\lib\site-packages\keras\src\layers\pc

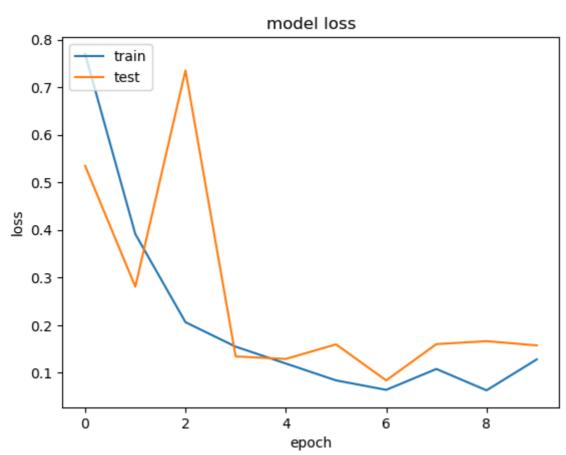
→ WARNING:tensorflow:From C:\Users\user\anaconda3\lib\site-packages\keras\src\optimizer

 \rightarrow

```
model1_acc = model.evaluate(val_ds)[1]
model1_acc
```

```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')

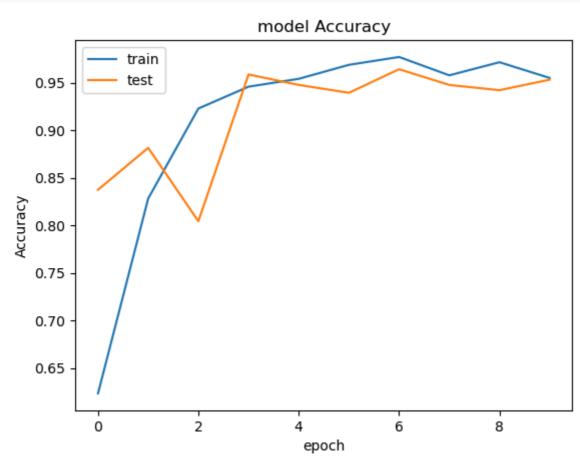
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```



 $\overline{\Rightarrow}$

```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model Accuracy')
plt.ylabel('Accuracy')

plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```



Second CNN

```
7/22/25, 12:29 PM
                        CNN-drowsinesssss.ipynb - Colab
  model.compile(optimizer='adam',
         loss=tf.keras.losses.SparseCategoricalCrossentropy(),
         metrics=['accuracy'])
  early_stopping = tf.keras.callbacks.EarlyStopping(monitor='val_loss',
                             patience=3,
                             mode='min',
                             restore_best_weights=True
                             )
  filepath = 'model_2.h5'
  model_checkpoint = tf.keras.callbacks.ModelCheckpoint(filepath,
                               monitor="val loss",
                               save_best_only=True,
                               save_weights_only=False,
                               mode="min",
                               save_freq="epoch",
                              )
  epochs = 15
  history = model.fit(train_ds, validation_data = val_ds, epochs = epochs,
             verbose = 1, callbacks=[early_stopping, model_checkpoint])
  → Epoch 1/15
    Epoch 2/15
    Epoch 3/15
    Epoch 4/15
    Epoch 5/15
    18/18 [============== ] - 200s 11s/step - loss: 0.1055 - accuracy: 0.9
    Epoch 6/15
    Epoch 7/15
    Epoch 8/15
    Epoch 9/15
    model2_acc = model.evaluate(val_ds)[1]
  model2 acc
  0.9862259030342102
  plt.plot(history.history['loss'])
  plt.plot(history.history['val_loss'])
```

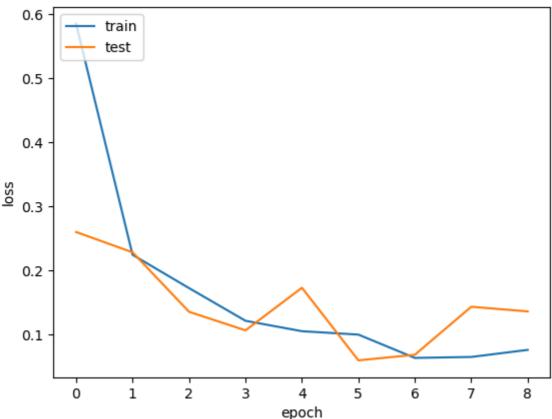
plt.title('model loss')

plt.ylabel('loss')

```
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```

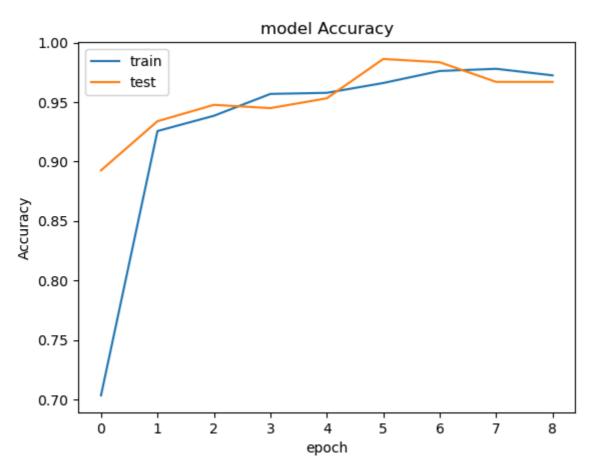


model loss



```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```





Third CNN

```
model = Sequential([
                  layers.experimental.preprocessing.Rescaling(1./255, input_shape=(img_he
                  layers.Conv2D(32, 3, padding='same', activation='relu'),
                  layers.MaxPooling2D(),
                  layers.Conv2D(64, 3, padding='same', activation='relu'),
                  layers.MaxPooling2D(),
                  layers.Conv2D(128, 3, padding='same', activation='relu'),
                  layers.MaxPooling2D(),
                  layers.Dropout(0.2),
                  layers.Conv2D(256, 3, padding='same', activation='relu'),
                  layers.MaxPooling2D(),
                  layers.Dropout(0.2),
                  layers.Conv2D(512, 3, padding='same', activation='relu'),
                  layers.MaxPooling2D(),
                  layers.Dropout(0.2),
                  layers.Flatten(),
                  layers.Dense(256, activation='relu'),
                  layers.Dropout(0.2),
                  layers.Dense(128, activation = 'relu'),
                  layers.Dense(num_classes, activation = 'sigmoid')
])
```

```
metrics=['accuracy'])
```

```
\rightarrow Epoch 1/15
Epoch 2/15
Epoch 3/15
Epoch 4/15
Epoch 5/15
Epoch 6/15
Epoch 7/15
Epoch 8/15
Epoch 9/15
Epoch 10/15
Epoch 11/15
Epoch 12/15
Epoch 13/15
Epoch 14/15
Epoch 15/15
```

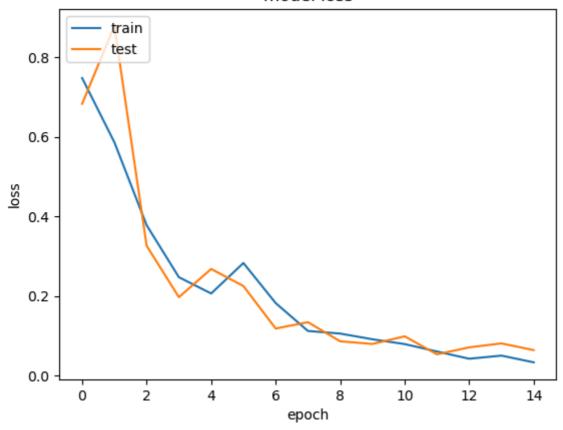
```
model3_acc = model.evaluate(val_ds)[1]
model3_acc
```

```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')

plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```

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model loss



```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model Accuracy')
plt.ylabel('Accuracy')

plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```

