## Eye detection

November 20, 2023

```
[35]: import numpy as np
      import os
      import matplotlib.pylab as plt
      import tensorflow as tf
      from tensorflow import keras
      from tensorflow.keras.preprocessing import image
      from keras.preprocessing.image import ImageDataGenerator
      from tensorflow.keras.models import Sequential, load model
      from tensorflow.keras.layers import Dense, Conv2D, MaxPooling2D, Flatten
      from keras.callbacks import ModelCheckpoint
      from sklearn.metrics import f1_score
      from sklearn.metrics import confusion_matrix, classification_report
      import seaborn as sns
[23]: train_path="data_eyes_closure/train"
      test_path="data_eyes_closure/test"
      train_datagen = ImageDataGenerator (rescale= 1./255, shear_range= 0.2, __
       →zoom_range= 0.2, horizontal_flip=True)
      test datagen = ImageDataGenerator (rescale= 1./255)
      x_train=train_datagen.flow_from_directory(train_path, target_size =__
       →(256,256),batch_size = 128, color_mode = 'grayscale',class_mode = _
      x_test=train_datagen.flow_from_directory(test_path,target_size =_u
       →(256,256), batch size = 128, color mode = 'grayscale', class mode = 'grayscale'
       x_train.class_indices
     Found 1234 images belonging to 2 classes.
     Found 218 images belonging to 2 classes.
[23]: {'Closed': 0, 'Open': 1}
[24]: classes=2
      model = Sequential()
```

Model: "sequential\_1"

· · · · ·	Output Shape	Param #
conv2d_3 (Conv2D)		
<pre>max_pooling2d_3 (MaxPoolin g2D)</pre>	(None, 128, 128, 32)	0
conv2d_4 (Conv2D)	(None, 128, 128, 64)	18496
<pre>max_pooling2d_4 (MaxPoolin g2D)</pre>	(None, 64, 64, 64)	0
conv2d_5 (Conv2D)	(None, 64, 64, 128)	73856
<pre>max_pooling2d_5 (MaxPoolin g2D)</pre>	(None, 32, 32, 128)	0
flatten_1 (Flatten)	(None, 131072)	0
dense_2 (Dense)	(None, 64)	8388672
dense_3 (Dense)	(None, 2)	130
Total params: 8481474 (32 35	MD)	

Total params: 8481474 (32.35 MB)
Trainable params: 8481474 (32.35 MB)
Non-trainable params: 0 (0.00 Byte)

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None

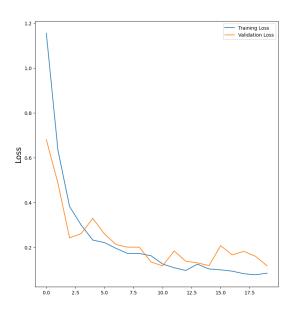
```
[25]: model.compile(loss = 'categorical_crossentropy',optimizer = 'adam', metrics = ___
     [26]: model_path="eye_detection.h5"
    checkpoint = ModelCheckpoint(model_path, monitor='val_accuracy', verbose=1,
                         save best only=True, mode='max')
    callbacks_list = [checkpoint]
[27]: num\_epochs = 20
    training_steps=x_train.n//x_train.batch_size
    validation_steps =x_test.n//x_test.batch_size
    history = model.fit_generator(x_train, epochs=num_epochs,__
     ⇒steps_per_epoch=training_steps, validation_data=x_test,
                  validation_steps=validation_steps, callbacks =_
     ⇔callbacks list)
   C:\Users\ajars\AppData\Local\Temp\ipykernel_12804\87117149.py:5: UserWarning:
   `Model.fit_generator` is deprecated and will be removed in a future version.
   Please use `Model.fit`, which supports generators.
     history = model.fit_generator(x_train, epochs=num_epochs,
   steps_per_epoch=training_steps, validation_data=x_test,
   Epoch 1/20
   Epoch 1: val_accuracy improved from -inf to 0.51562, saving model to
   eye detection.h5
   0.5741 - val_loss: 0.6810 - val_accuracy: 0.5156
   Epoch 2/20
   Epoch 2: val_accuracy improved from 0.51562 to 0.89844, saving model to
   eye_detection.h5
   0.7098 - val_loss: 0.4861 - val_accuracy: 0.8984
   Epoch 3/20
   Epoch 3: val_accuracy improved from 0.89844 to 0.92969, saving model to
   eve detection.h5
   0.8644 - val_loss: 0.2423 - val_accuracy: 0.9297
   Epoch 4/20
   Epoch 4: val_accuracy did not improve from 0.92969
   0.8924 - val_loss: 0.2611 - val_accuracy: 0.9141
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Epoch 5/20
Epoch 5: val_accuracy did not improve from 0.92969
0.9114 - val_loss: 0.3293 - val_accuracy: 0.8516
Epoch 6/20
Epoch 6: val_accuracy did not improve from 0.92969
0.9168 - val_loss: 0.2600 - val_accuracy: 0.8828
Epoch 7/20
Epoch 7: val_accuracy improved from 0.92969 to 0.93750, saving model to
eve detection.h5
9/9 [============== ] - 44s 5s/step - loss: 0.1954 - accuracy:
0.9231 - val_loss: 0.2124 - val_accuracy: 0.9375
Epoch 8/20
Epoch 8: val_accuracy improved from 0.93750 to 0.95312, saving model to
eye detection.h5
0.9295 - val_loss: 0.2007 - val_accuracy: 0.9531
Epoch 9/20
Epoch 9: val_accuracy did not improve from 0.95312
0.9331 - val_loss: 0.2010 - val_accuracy: 0.9141
Epoch 10/20
9/9 [============= - ETA: Os - loss: 0.1627 - accuracy: 0.9394
Epoch 10: val_accuracy did not improve from 0.95312
0.9394 - val_loss: 0.1350 - val_accuracy: 0.9531
Epoch 11/20
Epoch 11: val_accuracy did not improve from 0.95312
0.9548 - val_loss: 0.1172 - val_accuracy: 0.9531
Epoch 12/20
Epoch 12: val_accuracy did not improve from 0.95312
0.9602 - val_loss: 0.1837 - val_accuracy: 0.9453
Epoch 13/20
Epoch 13: val_accuracy did not improve from 0.95312
0.9647 - val_loss: 0.1381 - val_accuracy: 0.9531
Epoch 14/20
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Epoch 14: val_accuracy did not improve from 0.95312
   0.9494 - val_loss: 0.1308 - val_accuracy: 0.9531
   Epoch 15/20
   Epoch 15: val accuracy did not improve from 0.95312
   0.9627 - val_loss: 0.1184 - val_accuracy: 0.9375
   Epoch 16/20
   Epoch 16: val_accuracy did not improve from 0.95312
   0.9593 - val_loss: 0.2074 - val_accuracy: 0.9219
   Epoch 17/20
   Epoch 17: val_accuracy did not improve from 0.95312
   0.9602 - val_loss: 0.1663 - val_accuracy: 0.9453
   Epoch 18/20
   Epoch 18: val_accuracy did not improve from 0.95312
   9/9 [========== - - 48s 5s/step - loss: 0.0822 - accuracy:
   0.9711 - val_loss: 0.1820 - val_accuracy: 0.9375
   Epoch 19/20
   Epoch 19: val_accuracy did not improve from 0.95312
   0.9756 - val_loss: 0.1598 - val_accuracy: 0.9453
   Epoch 20/20
   Epoch 20: val_accuracy improved from 0.95312 to 0.97656, saving model to
   eye_detection.h5
   0.9620 - val_loss: 0.1174 - val_accuracy: 0.9766
[28]: plt.figure(figsize=(20,10))
   plt.subplot(1, 2, 1)
   plt.suptitle('Optimizer : Adam', fontsize=10)
   plt.ylabel('Loss', fontsize=16)
   plt.plot(history.history['loss'], label='Training Loss')
   plt.plot(history.history['val_loss'], label='Validation Loss')
   plt.legend(loc='upper right')
   plt.subplot(1, 2, 2)
   plt.ylabel('Accuracy', fontsize=16)
   plt.plot(history.history['accuracy'], label='Training Accuracy')
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plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend(loc='lower right')
plt.show()
```

Optimizer : Adam



[]:

