

Data Collection

In [1]:

```
from pydrive.auth import GoogleAuth
from pydrive.drive import GoogleDrive
from google.colab import auth
from oauth2client.client import GoogleCredentials

auth.authenticate_user()
gauth = GoogleAuth()
gauth.credentials = GoogleCredentials.get_application_default()
drive = GoogleDrive(gauth)

file_id = '1WkCy13kFta2GwV2ZJes0ZNNFNbfB11LE'

# YD_dataset 1WkCy13kFta2GwV2ZJes0ZNNFNbfB11LE
# Yawn_dataset 1LE3BHEpRuIDe-yedGriGg1Ix6_ZliEsp
downloaded = drive.CreateFile({'id': file_id})
downloaded.GetContentFile('yawn_dataset.zip')
```

In [2]:

```
!unzip yawn_dataset.zip
```

```
Archive:  yawn_dataset.zip
  creating: YD_dataset/test/
  creating: YD_dataset/test/no_yawn/
 inflating: YD_dataset/test/no_yawn/1004.jpg
 inflating: YD_dataset/test/no_yawn/1007.jpg
 inflating: YD_dataset/test/no_yawn/1010.jpg
 inflating: YD_dataset/test/no_yawn/1033.jpg
 inflating: YD_dataset/test/no_yawn/1044.jpg
 inflating: YD_dataset/test/no_yawn/1050.jpg
 inflating: YD_dataset/test/no_yawn/1063.jpg
 inflating: YD_dataset/test/no_yawn/1067.jpg
 inflating: YD_dataset/test/no_yawn/1096.jpg
 inflating: YD_dataset/test/no_yawn/1114.jpg
 inflating: YD_dataset/test/no_yawn/1118.jpg
 inflating: YD_dataset/test/no_yawn/1129.jpg
 inflating: YD_dataset/test/no_yawn/113.jpg
 inflating: YD_dataset/test/no_yawn/1134.jpg
 inflating: YD_dataset/test/no_yawn/115.jpg
 inflating: YD_dataset/test/no_yawn/1213.jpg
 inflating: YD_dataset/test/no_yawn/1267.jpg
```

Data Augmentation

In [3]:

```
import numpy as np
import pandas as pd

train = pd.read_csv('csv_dataset.csv')
train
```

Out[3]:

	image_names	yawn_or_not
0	YD_dataset/train/yawn/1.jpg	1
1	YD_dataset/train/yawn/10.jpg	1
2	YD_dataset/train/yawn/101.jpg	1
3	YD_dataset/train/yawn/103.jpg	1
4	YD_dataset/train/yawn/104.jpg	1
...
1443	YD_dataset/train/no_yawn/992.jpg	0
1444	YD_dataset/train/no_yawn/993.jpg	0
1445	YD_dataset/train/no_yawn/994.jpg	0
1446	YD_dataset/train/no_yawn/997.jpg	0
1447	YD_dataset/train/no_yawn/998.jpg	0

1448 rows × 2 columns

In [4]:

```
from skimage.io import imread
from skimage.transform import resize
import matplotlib.pyplot as plt
%matplotlib inline

train_img = []
for img_name in train['image_names']:
    # defining the image path
    image_path = '/content/' + img_name
    # reading the image
    img = imread(image_path)
    # normalizing the pixel values
    img = img/255
    # resizing the image to (224,224,3)
    img = resize(img, output_shape=(224,224,3), mode='constant', anti_aliasing=True)
    # converting the type of pixel to float 32
    img = img.astype('float32')
    # appending the image into the list
    train_img.append(img)

images = np.array(train_img)
images.shape
```

Out[4]:

(1448, 224, 224, 3)

In [5]:

```
labels = train['yawn_or_not'].values
labels.shape
```

Out[5]:

(1448,)

In [6]:

```
labels
```

Out[6]:

array([1, 1, 1, ..., 0, 0, 0])

In [7]:

```
from sklearn.model_selection import train_test_split

Xtrain,Xtest,Ytrain,Ytest = train_test_split(images,labels, test_size = 0.148, random_state
(Xtrain.shape, Ytrain.shape), (Xtest.shape, Ytest.shape))
```

Out[7]:

((1233, 224, 224, 3), (1233,)), ((215, 224, 224, 3), (215,)))

In [8]:

```
Ytest_bin = Ytest
Ytest_bin
```

Out[8]:

```
array([0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0,
        1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1,
        1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0,
        0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1,
        1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0,
        1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0,
        1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0,
        1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1,
        0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1,
        1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1])
```

In [9]:

```
from keras.utils import to_categorical
```

```
Ytrain = to_categorical(Ytrain)
Ytrain
```

Out[9]:

```
array([[0., 1.],
        [0., 1.],
        [0., 1.],
        ...,
        [0., 1.],
        [0., 1.],
        [0., 1.]], dtype=float32)
```

In [10]:

```
Ytest = to_categorical(Ytest)
Ytest
```

Out[10]:

```
array([[1., 0.],
        [0., 1.],
        [0., 1.],
        [0., 1.],
        [0., 1.],
        [1., 0.],
        [1., 0.],
        [0., 1.],
        [1., 0.],
        [0., 1.],
        [1., 0.],
        [0., 1.],
        [0., 1.],
        [0., 1.],
        [1., 0.],
        [1., 0.],
        [1., 0.],
        [0., 1.]])
```

In [11]:

```
print(Ytest.shape,Ytest_bin.shape)
```

```
(215, 2) (215,)
```

Building model architecture

In [14]:

```

import tensorflow as tf
from keras.models import Sequential
from keras import layers
from keras.layers import Conv2D, MaxPooling2D, Dropout, Flatten, Dense, Activation, GlobalMaxPooling2D
from keras import applications
from keras.applications import VGG16
from keras.models import Model
from keras import optimizers

base_model = VGG16(input_shape = (224, 224, 3),
include_top = False,
weights = 'imagenet')

for layer in base_model.layers:
    layer.trainable = False

x = layers.Flatten()(base_model.output)
x = layers.Dense(512, activation='relu')(x)

x = layers.Dropout(0.5)(x)
x = layers.Dense(2, activation='softmax')(x)

model = tf.keras.models.Model(base_model.input, x)

model.compile(optimizer = tf.keras.optimizers.RMSprop(lr=0.0001), loss = 'binary_crossentropy')
model.summary()

```

Model: "model"

Layer (type)	Output Shape	Param #
input_3 (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
flatten_2 (Flatten)	(None, 25088)	0
dense (Dense)	(None, 512)	12845568
dropout (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 2)	1026
=====		
Total params: 27,561,282		
Trainable params: 12,846,594		
Non-trainable params: 14,714,688		

Model Training

In [15]:

```
from keras.callbacks import ModelCheckpoint

checkpoint = ModelCheckpoint("model_weights.h5", monitor='val_acc', verbose=1, save_best_only=True,
                             save_weights_only=True, mode='max', period=1)
callbacks_list = [checkpoint]
```

In [16]:

```
training = model.fit(Xtrain,Ytrain,callbacks=callbacks_list,validation_data=(Xtest,Ytest),epochs=50)
```

Epoch 1/50

39/39 [=====] - 54s 392ms/step - loss: 1.0809 - acc: 0.5795 - val_loss: 0.5304 - val_acc: 0.6372

Epoch 00001: val_acc improved from -inf to 0.63721, saving model to model_weights.h5

Epoch 2/50

39/39 [=====] - 7s 168ms/step - loss: 0.5094 - acc: 0.7302 - val_loss: 0.5431 - val_acc: 0.6744

Epoch 00002: val_acc improved from 0.63721 to 0.67442, saving model to model_weights.h5

Epoch 3/50

39/39 [=====] - 7s 169ms/step - loss: 0.3921 - acc: 0.8209 - val_loss: 0.3999 - val_acc: 0.7860

Epoch 00003: val_acc improved from 0.67442 to 0.78605, saving model to model_weights.h5

Epoch 4/50

39/39 [=====] - 7s 168ms/step - loss: 0.3515 - acc: 0.8515 - val_loss: 0.3515 - val_acc: 0.8515

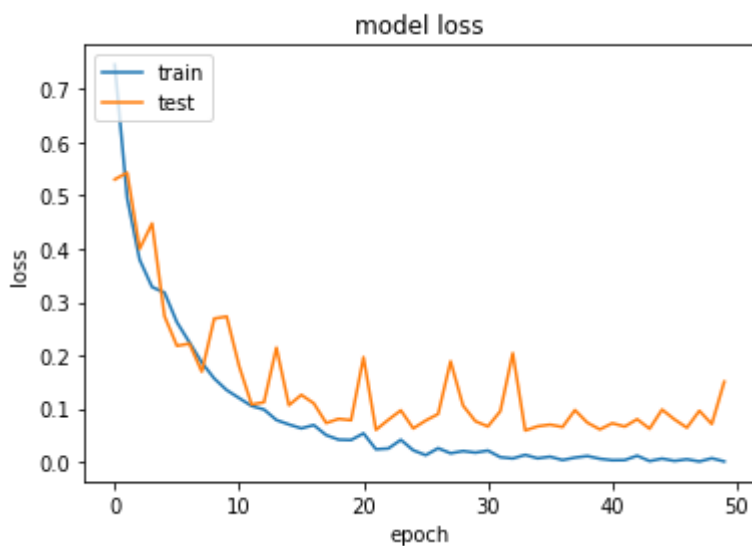
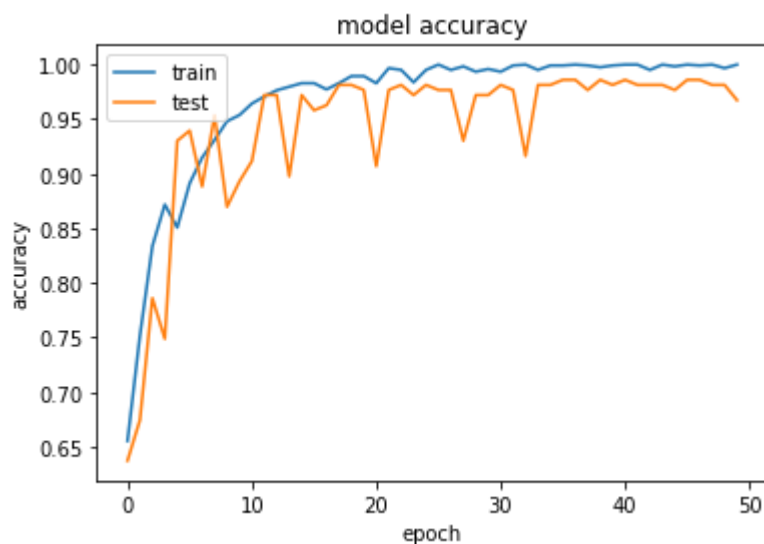
Model Evaluation

In [18]:

```
import matplotlib.pyplot as plt

# summarize training for accuracy
plt.plot(training.history['acc'])
plt.plot(training.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()

# summarize training for loss
plt.plot(training.history['loss'])
plt.plot(training.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```



In [17]:

```
result = model.evaluate(Xtest,Ytest,verbose=1)

print("test loss : ",result[0])
print("test acc. :",result[1])
```

```
7/7 [=====] - 1s 130ms/step - loss: 0.1505 - acc:
0.9674
test loss : 0.1505407989025116
test acc. : 0.9674418568611145
```

In [19]:

```
predictions= model.predict(Xtest,verbose=1)
predicted_classes = np.argmax(predictions,axis=1)
```

```
7/7 [=====] - 1s 155ms/step
```

In [20]:

```
from sklearn.metrics import confusion_matrix,classification_report,accuracy_score

print(accuracy_score(Ytest_bin,predicted_classes))
```

```
0.9674418604651163
```

In [21]:

```
print(confusion_matrix(Ytest_bin,predicted_classes))
```

```
[[102  0]
 [ 7 106]]
```

In [22]:

```
print(classification_report(Ytest_bin,predicted_classes,target_names=["yawn","no_yawn"]))
```

	precision	recall	f1-score	support
yawn	0.94	1.00	0.97	102
no_yawn	1.00	0.94	0.97	113
accuracy			0.97	215
macro avg	0.97	0.97	0.97	215
weighted avg	0.97	0.97	0.97	215

In [23]:

```
training.history.keys()
```

Out[23]:

```
dict_keys(['loss', 'acc', 'val_loss', 'val_acc'])
```

In [24]:

```
from sklearn.metrics import roc_curve, roc_auc_score  
  
roc_auc_score(Ytest_bin, predicted_classes)
```

Out[24]:

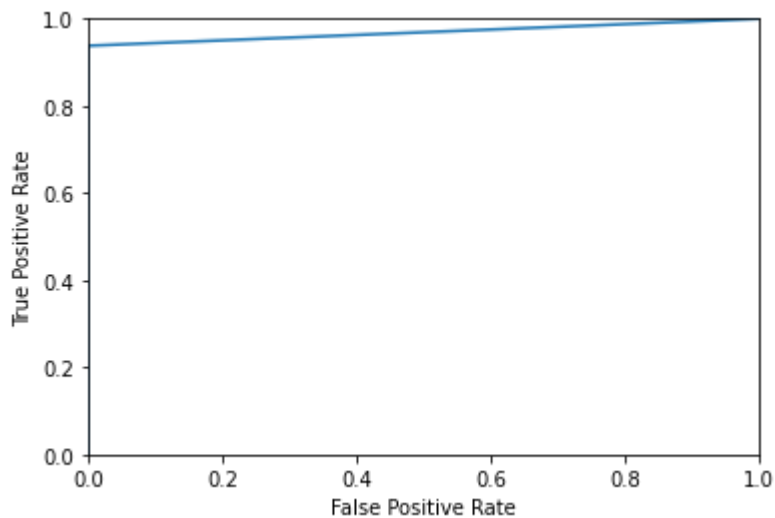
0.9690265486725664

In [25]:

```
fpr , tpr , thresholds = roc_curve(Ytest_bin, predicted_classes)
```

In [26]:

```
import matplotlib.pyplot as plt  
  
def plot_roc_curve(fpr, tpr):  
    plt.plot(fpr, tpr)  
    plt.axis([0,1,0,1])  
    plt.xlabel('False Positive Rate')  
    plt.ylabel('True Positive Rate')  
    plt.show()  
  
plot_roc_curve(fpr, tpr)
```



Saving model

In []:

```
model_json = model.to_json()  
with open("model.json", "w") as json_file:  
    json_file.write(model_json)
```

In []:

```
model.save("Yawn_classifier.model")
```

In []:

```
!zip -r /content/model.zip /content/Yawn_classifier.model
```

In []:

```
"""
!wget http://skulddata.cs.umass.edu/traces/mmsys/2014/user06.tar
!tar -xvf /content/user06.tar
!pip install patool
!pip install unrar
import patoolib
patoolib.extract_archive("/content/user06/YawDD dataset.rar", outdir="/content/")
"""
```