

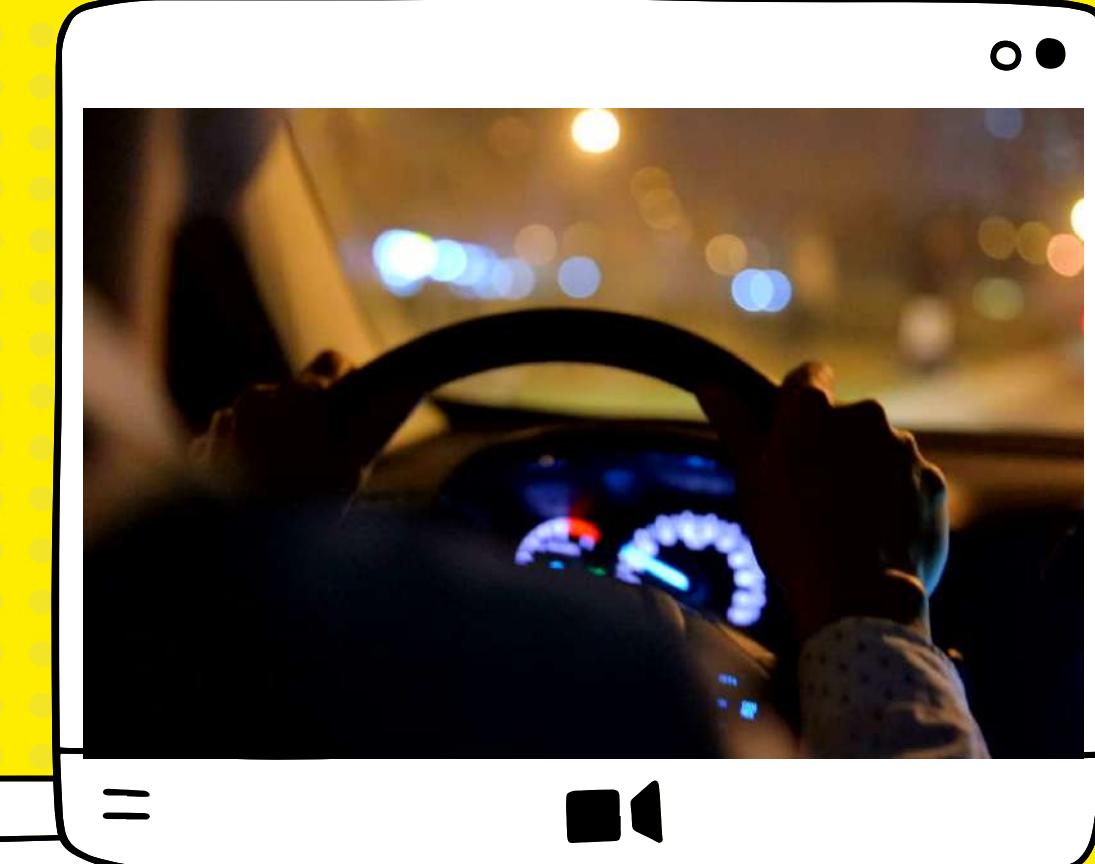
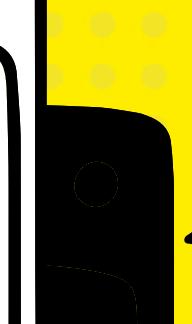


DRIVER DROWSINESS DETECTION SYSTEM



A countless number of people drive on the highway day and night. Taxi drivers, bus drivers, truck drivers and people traveling long-distance suffer from lack of sleep. Due to which it becomes very dangerous to drive when feeling sleepy.

The majority of accidents happen due to the drowsiness of the driver. So, to prevent these accidents we will build a drowsiness detection system using Python, OpenCV, and Keras which will alert the driver when he feels sleepy.



Problem Statement

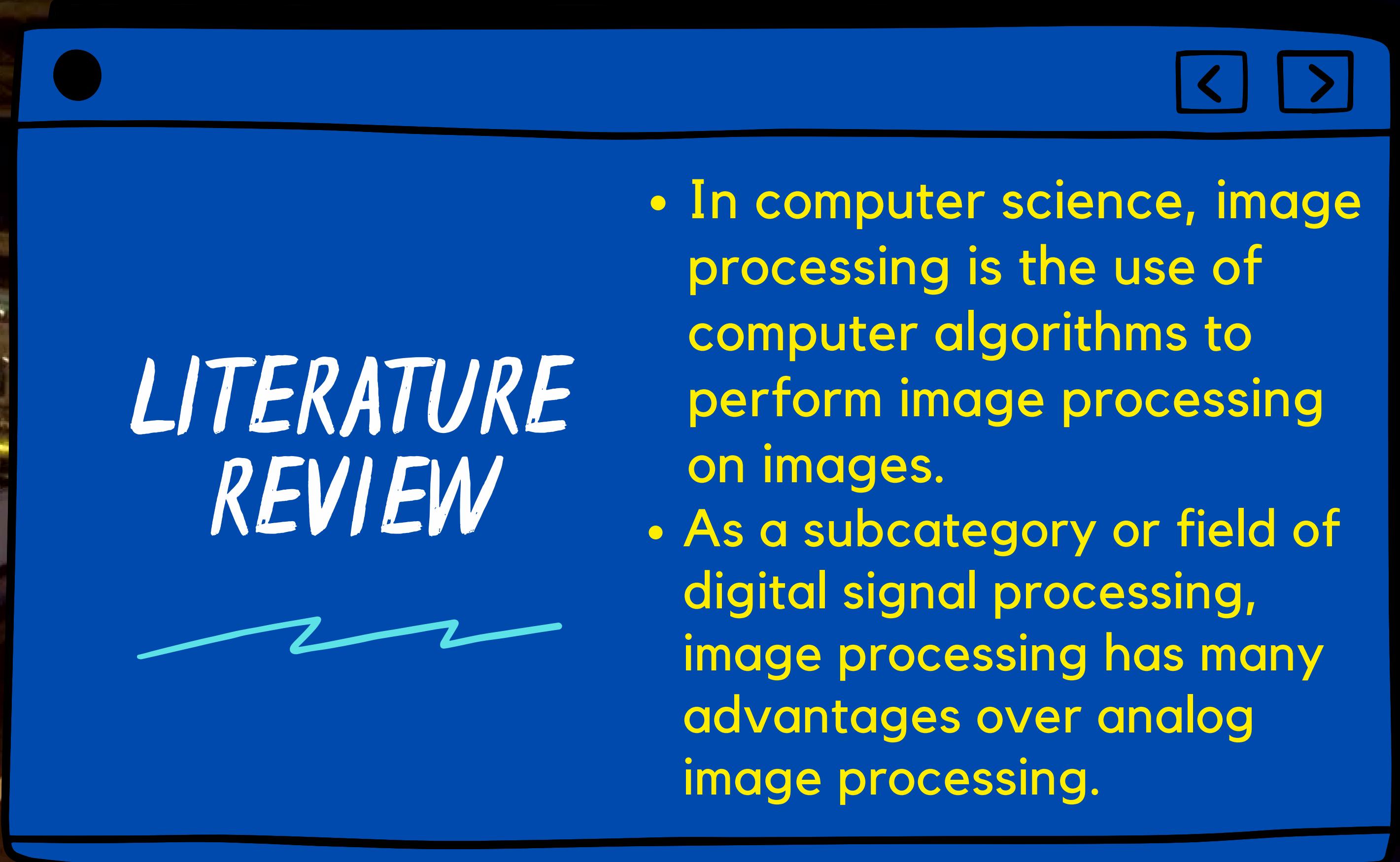
- *In this system the driver assistance System is presented in order to reduce the number of accidents caused by driver fatigue and thus improve road safety.*
- *This system treats the automatic detection of driver drowsiness based on visual information and transfer learning .We locate ,track and analyze the driver face and eyes and check whether the driver is drowsy .*



REQUIREMENTS

- Python - jupyter notebook.
- OpenCV - Face and eye detection.
- Keras and transfer learning- To build our classification model.
- TensorFlow - Keras uses TensorFlow as backend.
- Pygame - To play alarm sound.

LITERATURE REVIEW

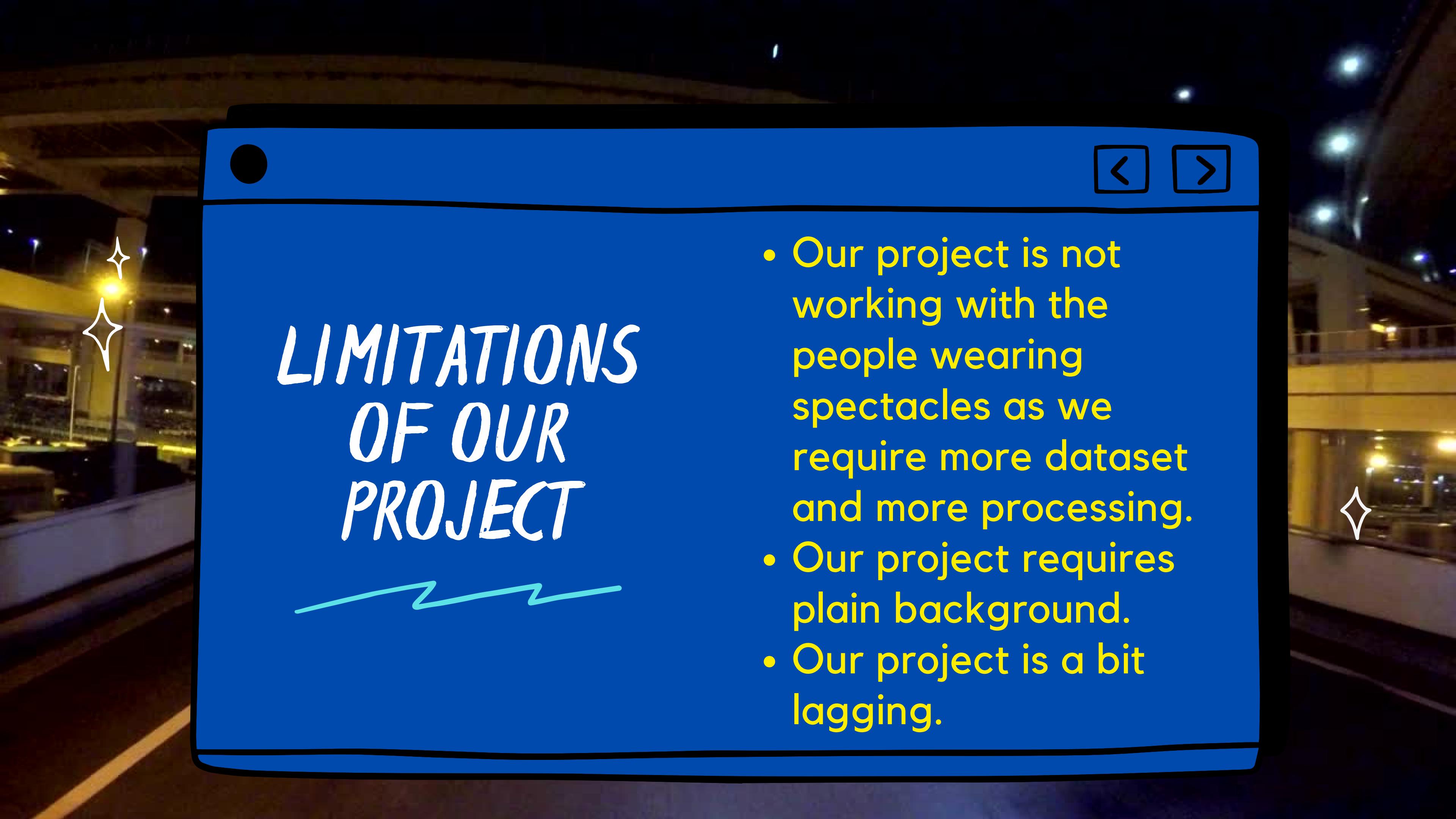


- In computer science, image processing is the use of computer algorithms to perform image processing on images.
- As a subcategory or field of digital signal processing, image processing has many advantages over analog image processing.

NOVELTY OR UNIQUENESS

- We have used the transfer learning model than CNN nueral network as it is more accurate than CNN





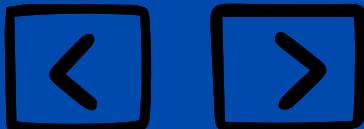
LIMITATIONS OF OUR PROJECT

- Our project is not working with the people wearing spectacles as we require more dataset and more processing.
- Our project requires plain background.
- Our project is a bit lagging.

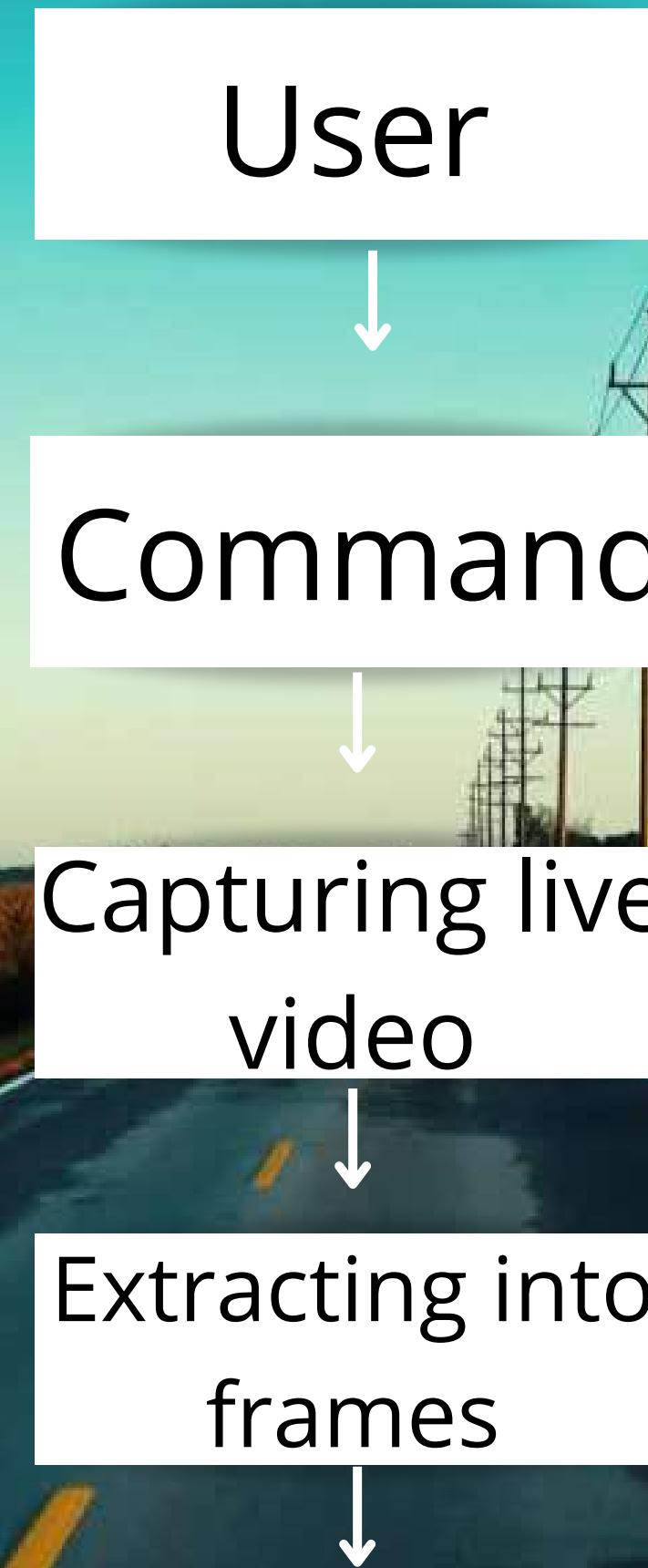
THE MODEL ARCHITECTURE

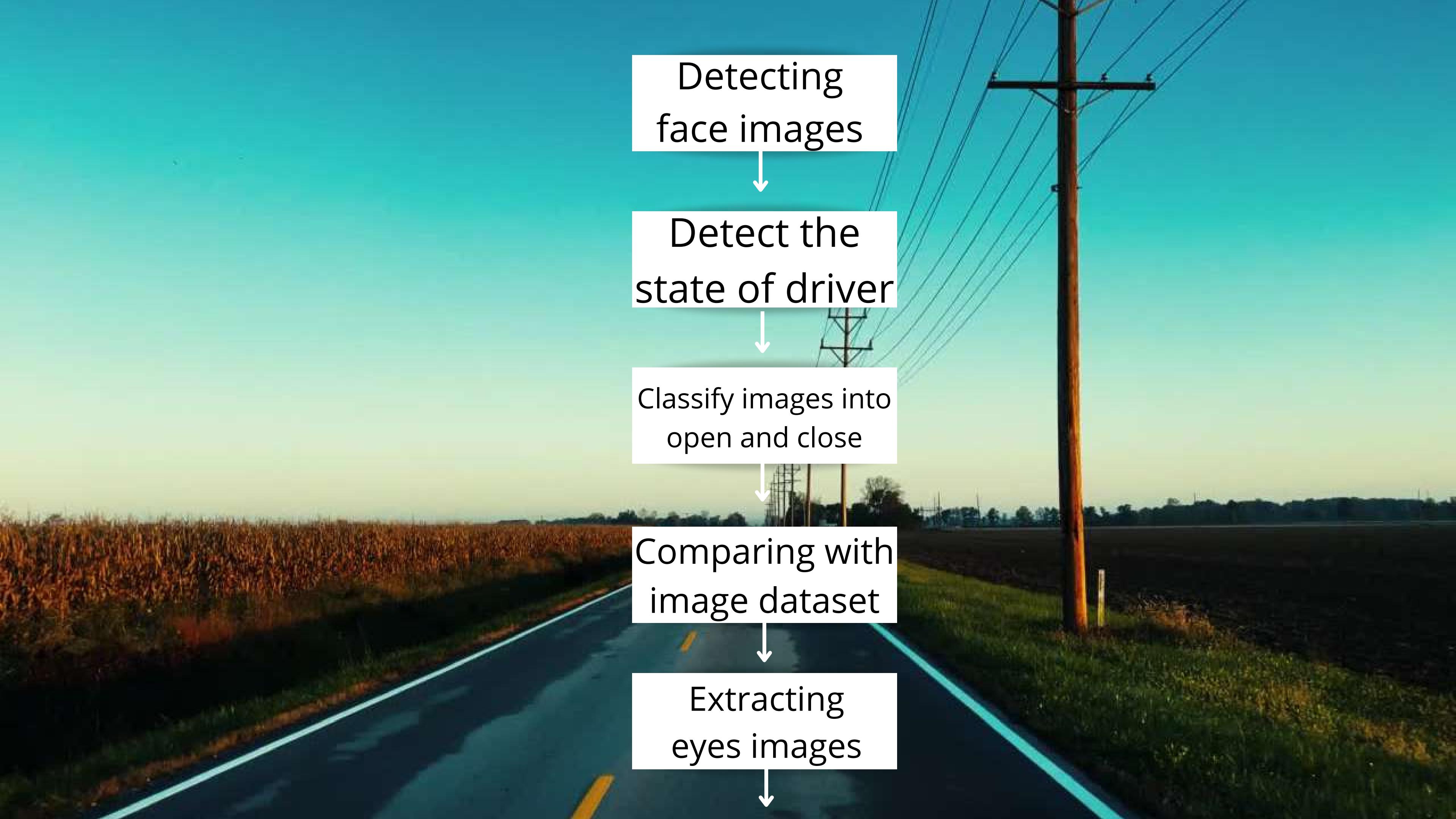


- *The model we used is built with Keras, tensorflow and transfer learning. Transfer learning (TL) is a research problem in machine learning (ML) that focuses on storing knowledge gained while solving one problem and applying it to a different but related problem.*



Flowchart





Detecting
face images

↓

Detect the
state of driver

↓

Classify images into
open and close

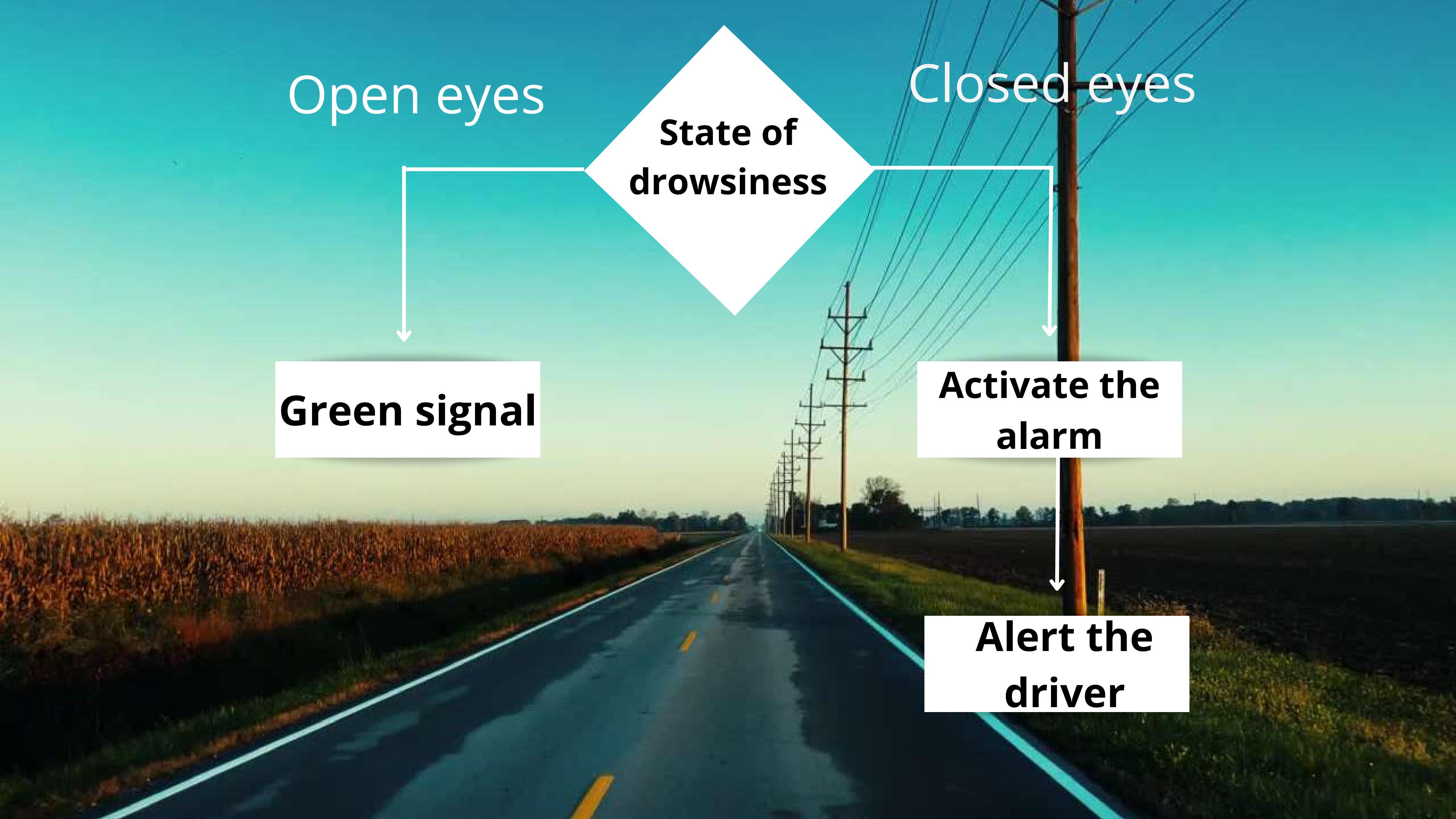
↓

Comparing with
image dataset

↓

Extracting
eyes images

↓



Open eyes

State of drowsiness

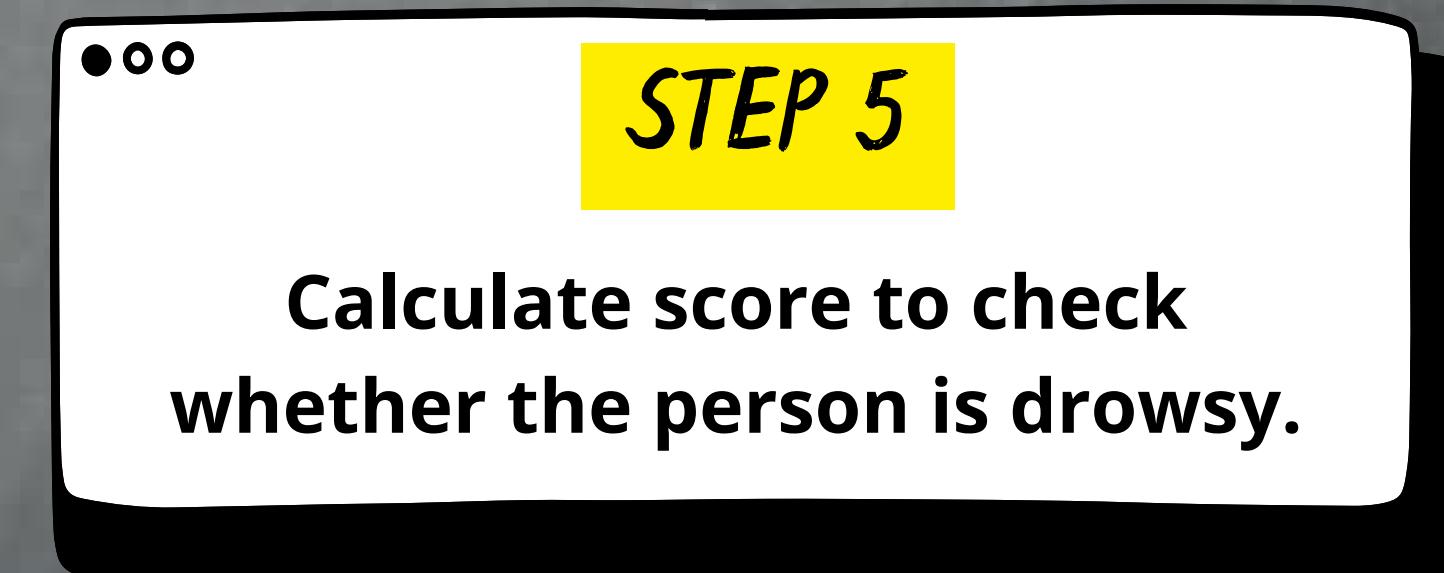
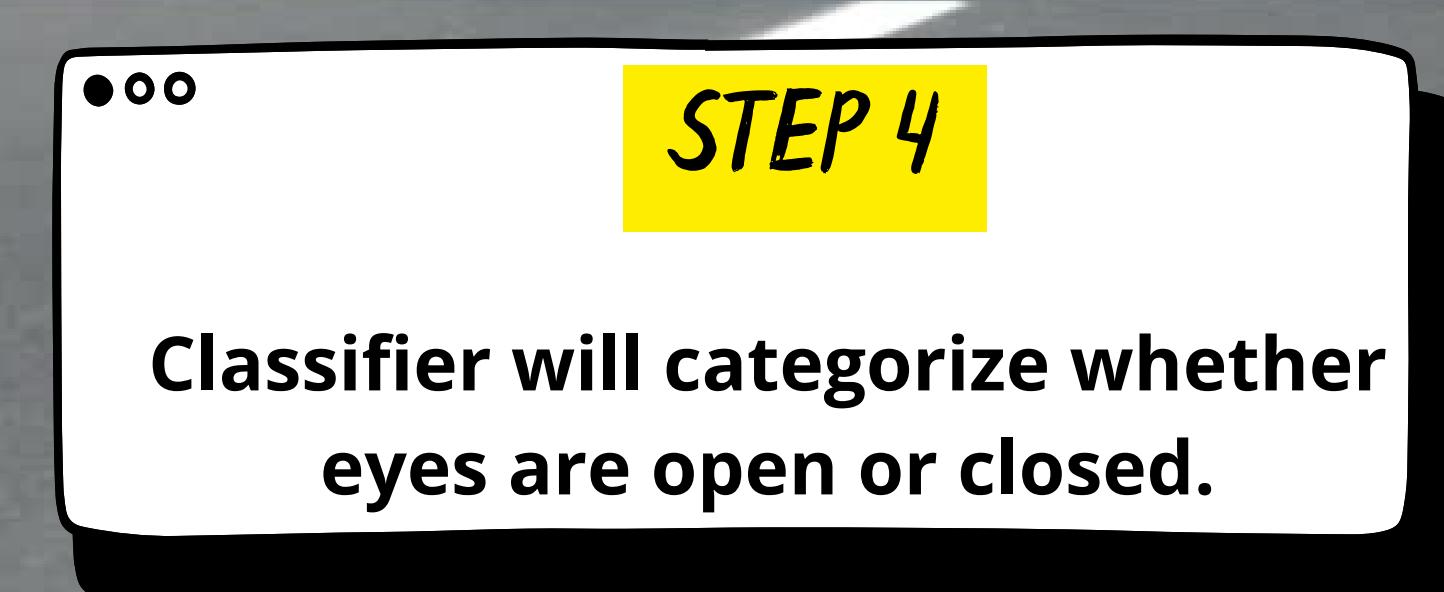
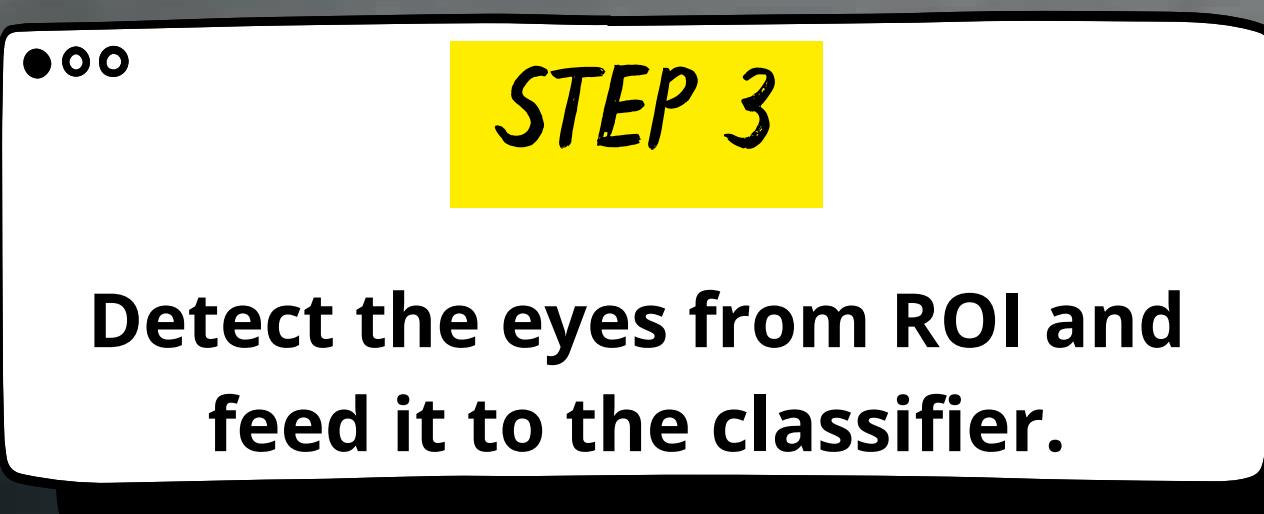
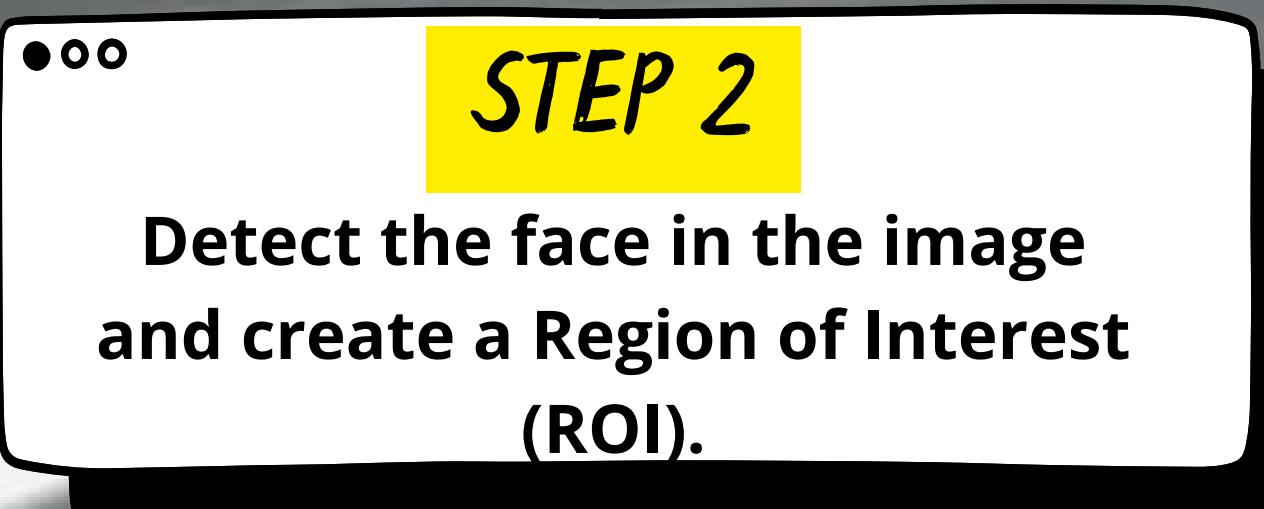
Closed eyes

Green signal

Activate the alarm

Alert the driver

In this Python project, we will be using OpenCV for gathering the images from webcam and feed them into a Deep Learning model which will classify whether the person's eyes are 'Open' or 'Closed'. The approach we will be using for this Python project is as follows :



STEP 1

Take Image as Input
from a Camera

- With a **webcam**, we will take images as input.
- So to access the **webcam**, we made an infinite loop that will capture each frame.
- We will use the method provided by OpenCV.

STEP 2

Detect Face in the
Image and Create a
Region of Interest
(ROI)

- To detect the face in the image, we need to first convert the image into grayscale as the OpenCV algorithm for object detection takes gray images in the input.

STEP 3

Detect the eyes from
ROI and feed it to the
classifier

- First, we set the cascade classifier for eyes in eye and reye respectively then detect the eyes.
- Now we need to extract only the eyes data from the full image.
- This can be achieved by extracting the eye's boundary box and then pulling out the eye image from the frame with this code.

STEP 4

Classifier will
Categorize whether
Eyes are Open or
Closed

- We are using transfer learning model to classify our images into open and closed eyes .

STEP 5

Calculate Score to
Check whether
Person is Drowsy

- The score is a value we will use to determine how long the person has closed his eyes.
- So if both eyes are closed, we will keep on increasing the score and when eyes are open, we decrease the score.
- And if the score is more than 15 then it will beep the alarm to alert the driver of the vehicle.

**SOURCE
CODE**

Dataset

We have taken a MRLEYES dataset from google which contains face images of 36 different persons in 36 folders , each folder containing about 1200-2000 face images. So we have created a code to classify these 75000 images into two subfolders i.e. closed and open eyes. Then we have created a test folder containing 2500 images of both subfolders and a train folder containing both the subfolder.

In [1]:

```
import os
import shutil
import glob
from tqdm import tqdm
```

-We have imported some libraries.

In [2]:

```
Raw_DIR= r'C:\Users\ARIHANT\Desktop\driver drowsiness\mr1Eyes_2018_01\mr1Eyes_2018_01'
for dirname, dirpath, filenames in os.walk(Raw_DIR):
    for i in tqdm([f for f in filenames if f.endswith('.png')]):
        if i.split('_')[4]=='0':
            shutil.copy(src=dirpath+'/'+i, dst=r'C:\Users\ARIHANT\Desktop\driver drowsiness\open eyes')
        elif i.split('_')[4]=='1':
            shutil.copy(src=dirpath+'/'+i, dst=r'C:\Users\ARIHANT\Desktop\driver drowsiness\closed eyes')
```

We have divided our database into two categories:

1)open eyes

2)closed eyes

OUTPUT

0it [00:00, ?it/s]	
100%	1114/1114 [00:13<00:00, 84.39it/s]
100%	679/679 [00:08<00:00, 76.08it/s]
100%	1069/1069 [00:15<00:00, 67.53it/s]
100%	736/736 [00:10<00:00, 67.85it/s]
100%	1012/1012 [00:14<00:00, 71.57it/s]
100%	624/624 [00:08<00:00, 69.85it/s]
100%	832/832 [00:12<00:00, 64.38it/s]
100%	387/387 [00:05<00:00, 70.85it/s]
100%	399/399 [00:05<00:00, 78.57it/s]
100%	1648/1648 [00:18<00:00, 87.86it/s]
100%	8728/8728 [01:36<00:00, 90.23it/s]
100%	3605/3605 [00:37<00:00, 95.17it/s]
100%	8884/8884 [01:36<00:00, 92.43it/s]
100%	1132/1132 [00:12<00:00, 93.45it/s]
100%	1889/1889 [00:22<00:00, 82.87it/s]
100%	1648/1648 [00:22<00:00, 74.53it/s]
100%	4410/4410 [00:54<00:00, 80.61it/s]
100%	6175/6175 [01:18<00:00, 78.59it/s]
100%	704/704 [00:09<00:00, 73.68it/s]
100%	987/987 [00:11<00:00, 88.23it/s]

**SOURCE CODE
(MODEL)**

In [1]:

```
import tensorflow as tf
from tensorflow.keras.applications import InceptionV3
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Dropout, Input, Flatten, Dense, MaxPooling2D
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.preprocessing.image import ImageDataGenerator # Data
```

**We have imported some
libraries.**

```
In [2]: tf.test.is_gpu_available()
```

```
WARNING:tensorflow:From C:\Users\ARIHANT\AppData\Local\Temp\ipykernel_20024\337460  
670.py:1: is_gpu_available (from tensorflow.python.framework.test_util) is depreca-  
ted and will be removed in a future version.
```

```
Instructions for updating:
```

```
Use `tf.config.list_physical_devices('GPU')` instead.
```

```
False
```

```
Out[2]:
```

We have checked if gpu is available or not.

```
In [3]: batchsize=8
```

```
In [4]: train_datagen= ImageDataGenerator(rescale=1./255, rotation_range=0.2,shear_range=0
zoom_range=0.2,width_shift_range=0.2,
height_shift_range=0.2, validation_split=0.2)

train_data= train_datagen.flow_from_directory(r'C:\Users\ARIHANT\Desktop\driver\driv
target_size=(80,80),batch_size=batchsize,class_mode='categorical')

validation_data= train_datagen.flow_from_directory(r'C:\Users\ARIHANT\Desktop\driver\driv
target_size=(80,80),batch_size=batchsize,class_mode='categorical')
```

Found 61178 images belonging to 2 classes.

Found 15294 images belonging to 2 classes.

We have set batchsize to 8,we are classifying images into 2 classes .

```
In [5]: test_datagen = ImageDataGenerator(rescale=1./255)

test_data = test_datagen.flow_from_directory(r'C:\Users\ARIHANT\Desktop\driver drop',
                                             target_size=(80,80),batch_size=batchsize,class_mode='categorical')

Found 5184 images belonging to 2 classes.
```

Rescale

```
In [6]: bmodel = InceptionV3(include_top=False, weights='imagenet', input_tensor=Input(shape=(299, 299, 3)))
hmodel = bmodel.output
hmodel = Flatten()(hmodel)
hmodel = Dense(64, activation='relu')(hmodel)
hmodel = Dropout(0.5)(hmodel)
hmodel = Dense(2,activation= 'softmax')(hmodel)

model = Model(inputs=bmodel.input, outputs= hmodel)
for layer in bmodel.layers:
    layer.trainable = False
```

We have used trasnfer learning to import the model using inception V3 which converted grey scale images into RGB.

model

```
In [7]: model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
<hr/>			
<hr/>			
input_1 (InputLayer)	[None, 80, 80, 3]	0	[]
conv2d (Conv2D)	(None, 39, 39, 32)	864	['input_1[0][0]']
batch_normalization (BatchNorm alization)	(None, 39, 39, 32)	96	['conv2d[0][0]']
activation (Activation)	(None, 39, 39, 32)	0	['batch_normaliza tion[0][0]']
conv2d_1 (Conv2D)	(None, 37, 37, 32)	9216	['activation[0] [0]']
batch_normalization_1 (BatchNo rmalization)	(None, 37, 37, 32)	96	['conv2d_1[0] [0]']
activation_1 (Activation)	(None, 37, 37, 32)	0	['batch_normaliza tion_1[0][0]']
conv2d_2 (Conv2D)	(None, 37, 37, 64)	18432	['activation_1[0] [0]']
batch_normalization_2 (BatchNo rmalization)	(None, 37, 37, 64)	192	['conv2d_2[0] [0]']

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model

In [8]: `from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPl:`

In [9]: `checkpoint = ModelCheckpoint(r'C:\Users\ARIHANT\Desktop\driver drowsiness\models\mo
monitor='val_loss', save_best_only=True, verbose=3)`

`earlystop = EarlyStopping(monitor = 'val_loss', patience=7, verbose= 3, restore_be:`

`learning_rate = ReduceLROnPlateau(monitor= 'val_loss', patience=3, verbose= 3,)`

`callbacks=[checkpoint,earlystop,learning_rate]`

Now we are saving our model.

```
In [ ]:
```

```
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['accuracy'])

model.fit_generator(train_data, steps_per_epoch=train_data.samples//batchsize,
                    validation_data=validation_data,
                    validation_steps=validation_data.samples//batchsize,
                    callbacks=callbacks,
                    epochs=20)
```

```
In [ ]:
```

```
acc_tr, loss_tr = model.evaluate_generator(train_data)
print(acc_tr)
print(loss_tr)
```

```
In [ ]:
```

```
acc_vr, loss_vr = model.evaluate_generator(validation_data)
print(acc_vr)
print(loss_vr)
```

```
In [ ]:
```

```
acc_test, loss_test = model.evaluate_generator(test_data)
print(acc_tr)
print(loss_tr)
```

```
In [ ]:
```

Now we are training our model:

- 1)Training train data**
- 2)Training validation data**
- 3)Training test data**

SOURCE CODE

(MAIN)

In [1]:

```
import cv2
import tensorflow as tf
from tensorflow.keras.models import load_model
import numpy as np
from pygame import mixer
```

pygame 2.1.2 (SDL 2.0.18, Python 3.9.12)

Hello from the pygame community. <https://www.pygame.org/contribute.html> (<https://www.pygame.org/contribute.html>)

-We have imported some libraries.

In [2]:

```
face_cascade = cv2.CascadeClassifier(cv2.data.haarcascades + 'haarcascade_frontalface_default.xml')
eye_cascade = cv2.CascadeClassifier(cv2.data.haarcascades + 'haarcascade_eye.xml')
model = load_model(r'C:\Users\ARIHANT\Desktop\driver drowsiness\models\model.h5')
```

-We have use haarcascade to detect the face quickly, and then loaded our model.

In [3]:

```
mixer.init()
sound= mixer.Sound(r'C:\Users\ARIHANT\Desktop\driver drowsiness\alarm.wav')
cap = cv2.VideoCapture(0)
Score = 0
while True:
    ret, frame = cap.read()
    height,width = frame.shape[0:2]
    gray = cv2.cvtColor(frame, cv2.COLOR_BGR2GRAY)
    faces= face_cascade.detectMultiScale(gray, scaleFactor= 1.2, minNeighbors=3)
    eyes= eye_cascade.detectMultiScale(gray, scaleFactor= 1.1, minNeighbors=1)

    cv2.rectangle(frame, (0,height-50),(200,height),(0,0,0),thickness=cv2.FILLED)
```

-Taking the file of alarm from our directory and then capturing our video.

```
for (x,y,w,h) in faces:  
    cv2.rectangle(frame,pt1=(x,y),pt2=(x+w,y+h), color= (255,0,0), thickness=3 )  
  
for (ex,ey,ew,eh) in eyes:  
    #cv2.rectangle(frame,pt1=(ex,ey),pt2=(ex+ew,ey+eh), color= (255,0,0), thickness=3 )  
  
# preprocessing steps  
eye= frame[ey:ey+eh,ex:ex+w]  
eye= cv2.resize(eye,(80,80))  
eye= eye/255  
eye= eye.reshape(80,80,3)  
eye= np.expand_dims(eye, axis=0)  
# preprocessing is done now model prediction  
prediction = model.predict(eye)
```

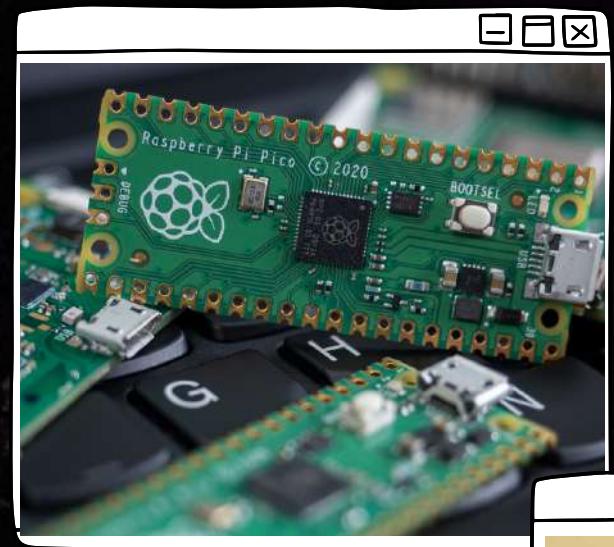
**-Processing the steps to make region
of interest i.e. in this case eyes.**

```
# if eyes are closed
if prediction[0][0]>0.30:
    cv2.putText(frame,'closed',(10,height-20),fontFace=cv2.FONT_HERSHEY_COMPLEX_SMALL
               thickness=1,lineType=cv2.LINE_AA)
    cv2.putText(frame,'Score'+str(Score),(100,height-20),fontFace=cv2.FONT_HERSHEY_
               thickness=1,lineType=cv2.LINE_AA)
    Score=Score+1
    if(Score>15):
        try:
            sound.play()
        except:
            pass

# if eyes are open
elif prediction[0][1]>0.90:
    cv2.putText(frame,'open',(10,height-20),fontFace=cv2.FONT_HERSHEY_COMPLEX_SMALL
               thickness=1,lineType=cv2.LINE_AA)
    cv2.putText(frame,'Score'+str(Score),(100,height-20),fontFace=cv2.FONT_HERSHEY_
               thickness=1,lineType=cv2.LINE_AA)
    Score = Score-1
    if (Score<0):
        Score=0

cv2.imshow('frame',frame)
if cv2.waitKey(33) & 0xFF==ord('q'):
    break

cap.release()
cv2.destroyAllWindows()
```



REAL WORLD APPLICATION:

- This System can be used in any automobiles with ease.
- This can be done by using Raspberry pi and with the help of a Web camera and an alarm for the alarm sound .



-
- A screenshot of a mobile application interface is shown, featuring a search bar at the top and a road scene in the background. A callout bubble contains the following list:
- The Driver eyes are constantly monitored by Webcam.
 - This Data is sent to the image classifier model which is running on Raspberry Pi.
 - The classifier checks every frame and classifies the frame as open or close.
 - If the eyes are being closed for a few seconds straight then the alarm is Triggered.
 - The alarm goes off when the driver is active and his eyes are open without any drowsiness.



★ | ⚓

- This system can be used to reduce the amount of roads accidents that happens to great extent.
- This can save a lot of lives, which is a main motive of this system.
- This system does not need any complex system to work effectively.
- Taking the facts into consideration driver drowsiness detection system is the future of road safety.

CONCLUSION



THANK YOU !!

