TEAM ID - PNT2022TMID40066

Import and unzip the dataset

```
from google.colab import drive drive.mount('/content/drive')
```

```
Mounted at /content/drive
```

```
#unzip the downloaded dataset
!unzip '/content/drive/MyDrive/damage vehicle.zip'
```

```
Archive: /content/drive/MyDrive/damage vehicle.zip creating:
  damage vehicle/ creating: damage vehicle/body/ creating:
  damage vehicle/body/training/ creating: damage
  vehicle/body/training/00-front/ inflating: damage
  vehicle/body/training/00-front/0001.jpeg inflating: damage
  vehicle/body/training/00-front/0002.JPEG inflating: damage
  vehicle/body/training/00-front/0003.JPEG inflating: damage
  vehicle/body/training/00-front/0004.JPEG inflating: damage
  vehicle/body/training/00-front/0005.JPEG inflating: damage
  vehicle/body/training/00-front/0006.JPEG inflating: damage
  vehicle/body/training/00-front/0007.JPEG inflating: damage
  vehicle/body/training/00-front/0008.jpeg inflating: damage
  vehicle/body/training/00-front/0009.JPEG inflating: damage
  vehicle/body/training/00-front/0010.JPEG inflating: damage
  vehicle/body/training/00-front/0011.JPEG inflating: damage
  vehicle/body/training/00-front/0012.jpeg inflating: damage
  vehicle/body/training/00-front/0013.JPEG inflating: damage
  vehicle/body/training/00-front/0014.JPEG inflating: damage
  vehicle/body/training/00-front/0015.JPEG inflating: damage
  vehicle/body/training/00-front/0016.JPEG inflating: damage
  vehicle/body/training/00-front/0017.JPEG inflating: damage
  vehicle/body/training/00-front/0018.JPEG inflating: damage
  vehicle/body/training/00-front/0019.JPEG inflating: damage
  vehicle/body/training/00-front/0020.jpeg inflating: damage
  vehicle/body/training/00-front/0021.JPEG inflating: damage
  vehicle/body/training/00-front/0022.JPEG inflating: damage
  vehicle/body/training/00-front/0023.JPEG inflating: damage
  vehicle/body/training/00-front/0024.JPEG inflating: damage
  vehicle/body/training/00-front/0025.jpeg inflating: damage
  vehicle/body/training/00-front/0026.JPEG inflating: damage
  vehicle/body/training/00-front/0027.JPEG inflating: damage
  vehicle/body/training/00-front/0028.JPEG inflating: damage
  vehicle/body/training/00-front/0029.JPEG inflating: damage
  vehicle/body/training/00-front/0030.JPEG inflating: damage
  vehicle/body/training/00-front/0031.JPEG inflating: damage
  vehicle/body/training/00-front/0032.JPEG inflating: damage
  vehicle/body/training/00-front/0033.JPEG inflating: damage
  vehicle/body/training/00-front/0034.JPEG inflating: damage
  vehicle/body/training/00-front/0035.jpeg inflating: damage
  vehicle/body/training/00-front/0036.JPEG inflating: damage
  vehicle/body/training/00-front/0037.JPEG inflating: damage
  vehicle/body/training/00-front/0038.JPEG inflating: damage
  vehicle/body/training/00-front/0039.JPEG inflating: damage
  vehicle/body/training/00-front/0040.JPEG inflating: damage
  vehicle/body/training/00-front/0041.JPEG inflating: damage
  vehicle/body/training/00-front/0042.JPEG inflating: damage
  vehicle/body/training/00-front/0043.JPEG inflating: damage
  vehicle/body/training/00-front/0044.JPEG inflating: damage
  vehicle/body/training/00-front/0045.JPEG inflating: damage
  vehicle/body/training/00-front/0046.jpeg inflating: damage
  vehicle/body/training/00-front/0047.JPEG inflating: damage
  vehicle/body/training/00-front/0048.JPEG inflating: damage
  vehicle/body/training/00-front/0049.JPEG inflating: damage
  vehicle/body/training/00-front/0050.JPEG inflating: damage
  vehicle/body/training/00-front/0051.JPEG inflating: damage
  vehicle/body/training/00-front/0052.JPEG inflating: damage
  vehicle/body/training/00-front/0053.JPEG
```

Image Preprocessing

1. Import The ImageDataGenerator Library

```
# Import required lib
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

2. Configure ImageDataGenerator Class

```
#Creating augmentation on training variable train_datagen =
ImageDataGenerator(rescale=1./255, shear_range = 0.1,
zoom_range=0.1, horizontal_flip=True)
```

```
# Creating augmentation on testing variable

test_datagen = ImageDataGenerator(rescale=1./255)
```

3. Apply ImageDataGenerator Functionality To Trainset And Testset

```
# Passing training data to train variable for body

xtrain = train_datagen.flow_from_directory('/content/damage vehicle/body/training',
                                           target_size=(224,224),
                                           class_mode='categorical', batch_size=10)
```

```
    Found 979 images belonging to 3 classes.
```

```
# Passing testing data to test variable for body

xtest = test_datagen.flow_from_directory('/content/damage vehicle/body/validation',
                                         target_size=(224,224),
                                         class_mode='categorical', batch_size=10)
```

```
    Found 171 images belonging to 3 classes.
```

```
# Passing training data to train variable for level

x_train = train_datagen.flow_from_directory('/content/damage vehicle/level/training',
                                           target_size=(224,224),
                                           class_mode='categorical', batch_size=10)
```

```
    Found 979 images belonging to 3 classes.
```

```
# Passing testing data to test variable for level

x_test = test_datagen.flow_from_directory('/content/damage vehicle/level/validation',
                                         target_size=(224,224),
                                         class_mode='categorical', batch_size=10)
```

```
    Found 171 images belonging to 3 classes.
```

Model Building

For Body

1. Importing The Model Building Libraries

```
#Import the library
from tensorflow.keras.layers  import Dense, Flatten, Input
from tensorflow.keras.models  import Model
from tensorflow.keras.preprocessing  import image
from tensorflow.keras.preprocessing.image import ImageDataGenerator, load_img
from tensorflow.keras.applications.vgg16  import VGG16, pr eprocess_input
from glob import glob


import numpy as np
 import matplotlib.pyplot as plt
import tensorflow as tf from tensorflow.keras.layers import Input, Lambda,
Dense, Flatten from tensorflow.keras.models import Model from
tensorflow.keras.applications.vgg16 import VGG16 from
tensorflow.keras.applications.vgg19 import VGG19 from
tensorflow.keras.preprocessing import image from
tensorflow.keras.preprocessing.image import ImageDataGenerator,load_img from
tensorflow.keras.models import Sequential import numpy as np from glob import
glob
```

2. Loading The Model

```
IMAGE_SIZE = [224, 224]

train_path = '/content/damage vehicle/body/training' valid_path =
'/content/damage vehicle/body/validation'
```

```
vgg16 = VGG16(input_shape=IMAGE_SIZE + [3], weights='imagenet', include_top=False)
```

```
    Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_ke
    58889256/58889256 [==============================] - 3s 0us/step
```

3. Adding Flatten Layer

```
for layer in vgg16.layers:
    layer.trainable = False
```

```
folders = glob('/content/damage vehicle/body/training/*')
```

```
folders
```

```
['/content/damage vehicle/body/training/00-front',
 '/content/damage vehicle/body/training/01-rear',
 '/content/damage vehicle/body/training/02-side']
```

```
x = Flatten()(vgg16.output)
```

```
len(folders)
```

```
3
```

4. Adding Output Layer

```
prediction = Dense(len(folders), activation='softmax')(x)
```

5. Creating A Model Object

```
model = Model(inputs=vgg16.input, outputs=prediction)
```

```
model.summary()
```

```
Model: "model"
_________________________________________________________________
 Layer (type)      Output Shape       Param #
=================================================================
 input_1 (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
                                                         2359808
 block5_conv3 (Conv2D)      (None, 14, 14, 512)
 block5_pool (MaxPooling2D) (None, 7, 7, 512)            0
                             flatten (Flatten)
         (None, 25088)
                                                         0
 dense (Dense)     (None, 3)                             75267

================================================================= Total
params: 14,789,955
Trainable params: 75,267
Non-trainable params: 14,714,688

_________________________________________________________________
```

6. Configure The Learning Process

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

7. Train The Model

```
r = model.fit_generator( xtrain,
  validation_data=xtest,
  epochs=25,
  steps_per_epoch=len(xtrain),
  validation_steps=len(xtest)
)
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: UserWarning: `Model.fit_generator` is deprecated and will be

Epoch 1/25
98/98 [==============================] - 23s 146ms/step - loss: 1.2077 - accuracy: 0.5465 - val_loss: 1.2900 - val_accuracy:
Epoch 2/25
98/98 [==============================] - 13s 128ms/step - loss: 0.8364 - accuracy: 0.7028 - val_loss: 0.8665 - val_accuracy:
Epoch 3/25
98/98 [==============================] - 13s 128ms/step - loss: 0.5293 - accuracy: 0.7998 - val_loss: 1.3260 - val_accuracy:
Epoch 4/25
98/98 [==============================] - 12s 127ms/step - loss: 0.3978 - accuracy: 0.8611 - val_loss: 0.9842 - val_accuracy:
Epoch 5/25
98/98 [==============================] - 12s 127ms/step - loss: 0.2783 - accuracy: 0.9030 - val_loss: 0.9397 - val_accuracy:
Epoch 6/25
98/98 [==============================] - 13s 128ms/step - loss: 0.2690 - accuracy: 0.9070 - val_loss: 0.9892 - val_accuracy:
Epoch 7/25
98/98 [==============================] - 12s 127ms/step - loss: 0.1788 - accuracy: 0.9448 - val_loss: 1.0052 - val_accuracy:
Epoch 8/25
98/98 [==============================] - 13s 129ms/step - loss: 0.1671 - accuracy: 0.9469 - val_loss: 1.1693 - val_accuracy:
Epoch 9/25
98/98 [==============================] - 13s 129ms/step - loss: 0.1277 - accuracy: 0.9561 - val_loss: 1.0058 - val_accuracy:
Epoch 10/25
98/98 [==============================] - 13s 128ms/step - loss: 0.1184 - accuracy: 0.9591 - val_loss: 1.0620 - val_accuracy:
Epoch 11/25
98/98 [==============================] - 13s 130ms/step - loss: 0.0963 - accuracy: 0.9745 - val_loss: 1.1219 - val_accuracy:

Epoch 12/25

98/98 [==============================] - 13s 129ms/step - loss: 0.0857 - accuracy: 0.9765 - val_loss: 1.0284 - val_accuracy:
Epoch 13/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0582 - accuracy: 0.9837 - val_loss: 1.1153 - val_accuracy:
Epoch 14/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0688 - accuracy: 0.9877 - val_loss: 1.1033 - val_accuracy:
Epoch 15/25
98/98 [==============================] - 13s 131ms/step - loss: 0.0709 - accuracy: 0.9867 - val_loss: 1.0730 - val_accuracy:
Epoch 16/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0895 - accuracy: 0.9775 - val_loss: 1.1225 - val_accuracy:
Epoch 17/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0609 - accuracy: 0.9918 - val_loss: 1.2937 - val_accuracy:
Epoch 18/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0998 - accuracy: 0.9714 - val_loss: 1.1754 - val_accuracy:
Epoch 19/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0728 - accuracy: 0.9847 - val_loss: 1.5074 - val_accuracy:
Epoch 20/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0972 - accuracy: 0.9714 - val_loss: 1.4684 - val_accuracy:
Epoch 21/25
98/98 [==============================] - 13s 131ms/step - loss: 0.0404 - accuracy: 0.9908 - val_loss: 1.4215 - val_accuracy:
Epoch 22/25
98/98 [==============================] - 13s 131ms/step - loss: 0.0854 - accuracy: 0.9867 - val_loss: 1.4772 - val_accuracy:
Epoch 23/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0399 - accuracy: 0.9918 - val_loss: 1.4306 - val_accuracy:
Epoch 24/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0400 - accuracy: 0.9908 - val_loss: 1.4562 - val_accuracy:
Epoch 25/25
98/98 [==============================] - 13s 129ms/step - loss: 0.1692 - accuracy: 0.9387 - val_loss: 1.6805 - val_accuracy:
```

8. Save The Model

```
from tensorflow.keras.models import load_model

model.save('/content/damage vehicle/Model/body.h5')
```

9. Test The Model

```
from  tensorflow.keras.models  import  load_model
import cv2 from skimage.transform import resize
```

```
model = load_model('/content/damage  vehicle/Model/body.h5')
```

```
import numpy as np
def detect(frame):
  img = cv2.resize(frame,(224,224))
  img =
  cv2.cvtColor(img,cv2.COLOR_BGR2RGB)

  if(np.max(img)>1): img
    = img/255.0
  img  =  np.array([img])  prediction  =
  model.predict(img)       label       =
  ["front","rear","side"]     preds     =
  label[np.argmax(prediction)]     return
  preds
data = "/content/damage vehicle/body/training/00-front/0002.JPEG"
image = cv2.imread(data) print(detect(image))
```

```
    1/1 [==============================] - 0s 148ms/step front
```

Model Building

For Level

```
import tensorflow as tf from tensorflow.keras.layers import Input, Lambda,
Dense, Flatten from tensorflow.keras.models import Model from
tensorflow.keras.applications.vgg16 import VGG16 from
tensorflow.keras.applications.vgg19 import VGG19 from
tensorflow.keras.preprocessing import image from
tensorflow.keras.preprocessing.image import ImageDataGenerator,load_img from
tensorflow.keras.models import Sequential import numpy as np from glob import
glob
```

2. Loading The Model

```
IMAGE_SIZE = [224, 224]

train_path = '/content/damage vehicle/level/training' valid_path = '/content/damage vehicle/level/validation'
```

```
vgg16 = VGG16(input_shape=IMAGE_SIZE + [3], weights='imagenet', include_top=False)
```

3. Adding Flatten Layer

```
for layer in vgg16.layers:
    layer.trainable = False
```

```
folders = glob('/content/damage vehicle/level/training/*')
```

```
folders
```

```
    ['/content/damage vehicle/level/training/03-severe',
     '/content/damage vehicle/level/training/02-moderate',  '/content/damage
     vehicle/level/training/01-minor']
```

[1] **. Importing The Model Building Libraries**

```
x = Flatten()(vgg16.output)
```

```
len(folders)
```

```
3
```

4. Adding Output Layer

```
prediction = Dense(len(folders), activation='softmax')(x)
```

5. Creating A Model Object

```
model = Model(inputs=vgg16.input, outputs=prediction)
```

```
model.summary()
```

```
Model: "model_1"
_________________________________________________________________
 Layer (type)      Output Shape       Param #
=================================================================
 input_2 (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_1 (Flatten)         (None, 25088)               0
 dense_1 (Dense)   (None, 3)                             75267
=================================================================
Total params: 14,789,955
Trainable params: 75,267
Non-trainable params: 14,714,688
_________________________________________________________________
```

6. Configure The Learning Process

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

7. Train The Model

```
r = model.fit_generator( x_train,
  validation_data=x_test,
  epochs=25,
  steps_per_epoch=len(x_train),
  validation_steps=len(x_test)
)
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: UserWarning: `Model.fit_generator` is deprecated and will be

Epoch 1/25
98/98 [==============================] - 14s 133ms/step - loss: 1.1629 - accuracy: 0.5495 - val_loss: 1.1559 - val_accuracy:
Epoch 2/25
98/98 [==============================] - 13s 130ms/step - loss: 0.7157 - accuracy: 0.7089 - val_loss: 0.9643 - val_accuracy:
Epoch 3/25
98/98 [==============================] - 13s 130ms/step - loss: 0.4978 - accuracy: 0.8161 - val_loss: 1.5663 - val_accuracy:
Epoch 4/25
98/98 [==============================] - 13s 128ms/step - loss: 0.5277 - accuracy: 0.7865 - val_loss: 1.6003 - val_accuracy:
Epoch 5/25
98/98 [==============================] - 13s 128ms/step - loss: 0.3763 - accuracy: 0.8468 - val_loss: 1.1925 - val_accuracy:
Epoch 6/25
98/98 [==============================] - 13s 128ms/step - loss: 0.2445 - accuracy: 0.9203 - val_loss: 1.0354 - val_accuracy:
Epoch 7/25
98/98 [==============================] - 13s 128ms/step - loss: 0.1902 - accuracy: 0.9346 - val_loss: 1.2155 - val_accuracy:
Epoch 8/25
98/98 [==============================] - 13s 128ms/step - loss: 0.1327 - accuracy: 0.9571 - val_loss: 1.0902 - val_accuracy:
Epoch 9/25
98/98 [==============================] - 13s 127ms/step - loss: 0.1206 - accuracy: 0.9540 - val_loss: 1.1282 - val_accuracy:
Epoch 10/25
98/98 [==============================] - 13s 128ms/step - loss: 0.1181 - accuracy: 0.9591 - val_loss: 1.1311 - val_accuracy:
Epoch 11/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0910 - accuracy: 0.9765 - val_loss: 1.1538 - val_accuracy:
Epoch 12/25
98/98 [==============================] - 12s 127ms/step - loss: 0.0813 - accuracy: 0.9806 - val_loss: 1.2209 - val_accuracy:
Epoch 13/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0603 - accuracy: 0.9857 - val_loss: 1.2545 - val_accuracy:

Epoch 14/25

98/98 [==============================] - 12s 127ms/step - loss: 0.0474 - accuracy: 0.9949 - val_loss: 1.1609 - val_accuracy:
Epoch 15/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0366 - accuracy: 0.9959 - val_loss: 1.1688 - val_accuracy:
Epoch 16/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0493 - accuracy: 0.9888 - val_loss: 1.1850 - val_accuracy:
Epoch 17/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0320 - accuracy: 0.9939 - val_loss: 1.1884 - val_accuracy:
Epoch 18/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0363 - accuracy: 0.9939 - val_loss: 1.2897 - val_accuracy:
Epoch 19/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0298 - accuracy: 0.9949 - val_loss: 1.2499 - val_accuracy:
Epoch 20/25
98/98 [==============================] - 13s 130ms/step - loss: 0.0250 - accuracy: 0.9980 - val_loss: 1.2801 - val_accuracy:
Epoch 21/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0329 - accuracy: 0.9959 - val_loss: 1.2366 - val_accuracy:
Epoch 22/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0170 - accuracy: 1.0000 - val_loss: 1.2901 - val_accuracy:
Epoch 23/25
98/98 [==============================] - 13s 130ms/step - loss: 0.0216 - accuracy: 1.0000 - val_loss: 1.2697 - val_accuracy:
Epoch 24/25
98/98 [==============================] - 13s 128ms/step - loss: 0.0365 - accuracy: 0.9908 - val_loss: 1.4214 - val_accuracy:
Epoch 25/25
98/98 [==============================] - 13s 129ms/step - loss: 0.0380 - accuracy: 0.9939 - val_loss: 1.4219 - val_accuracy:
```

8. Save The Model

```
from tensorflow.keras.models import load_model

model.save('/content/damage vehicle/Model/level.h5')
```

9. Test The Model

```
from tensorflow.keras.models import load_model import
cv2
```

```
model = load_model('/content/damage  vehicle/Model/level.h5')
```

```
from skimage.transform import resize
```

```
import numpy as np
def detect(frame):
  img = cv2.resize(frame,(224,224))
  img = cv2.cvtColor(img,cv2.COLOR_BGR2RGB)

  if(np.max(img)>1):
    img = img/255.0
  img = np.array([img]) prediction =
  model.predict(img) label =
  ["minor","moderate","severe"] preds =
  label[np.argmax(prediction)] return
  preds
data = "/content/damage vehicle/level/validation/01-minor/0005.JPEG"
image = cv2.imread(data) print(detect(image))
```

```
1/1 [==============================] - 0s 142ms/step minor
```