**TEAM ID-PNT2022TMID50946**

**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**

* **Import The ImageDataGenerator Library**
* **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)

* **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**

* **Importing The Model Building Libraries**

import matplotlib.pyplot as plt

* **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](https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels_notop.h5) 58889256/58889256 [==============================] - 3s 0us/step

* **Adding Flatten Layer**

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

3

* **Adding Output Layer**

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

* **Creating A Model Object**

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

model.summary()

Model: "model"

|  |  |  |
| --- | --- | --- |
| Layer (type)  ============================  input\_1 (InputLayer) | Output Shape  ========================= [(None, 224, 224, 3)] | Param #  ============ 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 (Flatten) | (None, | 25088) | |  | 0 |
| dense (Dense) | (None, | 3) |  |  | 75267 |

=================================================================

Total params: 14,789,955

Trainable params: 75,267

Non-trainable params: 14,714,688

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

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

* **Save The Model**

from tensorflow.keras.models import load\_model

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

* **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') |
| 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 |
| import numpy as np |
| 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**

* **Importing The Model Building Libraries**

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

* **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)

* **Adding Flatten Layer**

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

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

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

3

* **Adding Output Layer**

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

* **Creating A Model Object**

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

model.summary()

Model: "model\_1"

|  |  |  |
| --- | --- | --- |
| Layer (type)  ============================  input\_2 (InputLayer) | Output Shape  ========================= [(None, 224, 224, 3)] | Param #  ============ 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 |

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Total params: 14,789,955

Trainable params: 75,267

Non-trainable params: 14,714,688

* **Configure The Learning Process**

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

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

* **Save The Model**

from tensorflow.keras.models import load\_model

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

* **Test The Model**

|  |
| --- |
| from tensorflow.keras.models import load\_model import cv2  from skimage.transform import resize |
|  |
| model = load\_model('/content/damage vehicle/Model/level.h5') |
|  |
| 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 |
|  |
| import numpy as np |
|  |
| data = "/content/damage vehicle/level/validation/01 -minor/0005.JPEG" image = cv2.imread(data)  print(detect(image)) |

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

minor