

FOR BODY DAMAGE

▼ IMAGE PRE PROCESSING

1. Import The ImageDataGenerator Library

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

2. Configure ImageDataGenerator Class

```
train_datagen = ImageDataGenerator(rescale = 1./255,  
                                   shear_range = 0.1,  
                                   zoom_range = 0.1,  
                                   horizontal_flip = True)
```

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

3. Apply ImageDataGenerator Functionality To Trainset And Testset

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

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount('/content/drive') again

```
training_set = train_datagen.flow_from_directory('/content/drive/MyDrive/Car_damage/body/train',  
                                                target_size=(224, 224),  
                                                batch_size=10,  
                                                class_mode='categorical')  
test_set = test_datagen.flow_from_directory('/content/drive/MyDrive/Car_damage/body/validation',  
                                           target_size=(224, 224),  
                                           batch_size=10,  
                                           class_mode='categorical')
```

```
Found 979 images belonging to 3 classes.  
Found 171 images belonging to 3 classes.
```

▼ MODEL BUILDING

1. 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
```

2. Loading The Model

```
IMAGE_SIZE = [224, 224]
```

```
train_path = '/content/drive/MyDrive/Car_damage/body/training'
valid_path = '/content/drive/MyDrive/Car_damage/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\_data\_format.h5
58889256/58889256 [=====] - 0s 0us/step
```



3. Adding Flatten Layer

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

```
folders = glob('/content/drive/MyDrive/Car_damage/body/training/*')
```

```
folders
```

```
['/content/drive/MyDrive/Car_damage/body/training/02-side',
 '/content/drive/MyDrive/Car_damage/body/training/01-rear',
 '/content/drive/MyDrive/Car_damage/body/training/00-front']
```

```
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 |
| 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

6. Configure The Learning Process

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

7. Train The Model

```
r = model.fit_generator(
    training_set,
    validation_data=test_set,
    epochs=25,
    steps_per_epoch=len(training_set),
    validation_steps=len(test_set)
)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: UserWarning: `Model.fit`

```
Epoch 1/25
98/98 [=====] - 498s 5s/step - loss: 1.3552 - accuracy: 0.53
Epoch 2/25
98/98 [=====] - 421s 4s/step - loss: 0.6244 - accuracy: 0.76
Epoch 3/25
98/98 [=====] - 422s 4s/step - loss: 0.4846 - accuracy: 0.81
Epoch 4/25
98/98 [=====] - 420s 4s/step - loss: 0.3649 - accuracy: 0.86
Epoch 5/25
98/98 [=====] - 417s 4s/step - loss: 0.2899 - accuracy: 0.89
Epoch 6/25
98/98 [=====] - 418s 4s/step - loss: 0.3088 - accuracy: 0.88
Epoch 7/25
98/98 [=====] - 420s 4s/step - loss: 0.1704 - accuracy: 0.93
Epoch 8/25
98/98 [=====] - 423s 4s/step - loss: 0.1787 - accuracy: 0.93
Epoch 9/25
98/98 [=====] - 421s 4s/step - loss: 0.1622 - accuracy: 0.94
Epoch 10/25
98/98 [=====] - 417s 4s/step - loss: 0.0985 - accuracy: 0.97
Epoch 11/25
98/98 [=====] - 423s 4s/step - loss: 0.0962 - accuracy: 0.96
Epoch 12/25
98/98 [=====] - 426s 4s/step - loss: 0.0909 - accuracy: 0.98
Epoch 13/25
98/98 [=====] - 422s 4s/step - loss: 0.0796 - accuracy: 0.98
Epoch 14/25
98/98 [=====] - 421s 4s/step - loss: 0.0625 - accuracy: 0.98
Epoch 15/25
98/98 [=====] - 421s 4s/step - loss: 0.0995 - accuracy: 0.97
Epoch 16/25
98/98 [=====] - 422s 4s/step - loss: 0.1147 - accuracy: 0.96
Epoch 17/25
```

```

98/98 [=====] - 423s 4s/step - loss: 0.1107 - accuracy: 0.97
Epoch 18/25
98/98 [=====] - 423s 4s/step - loss: 0.0688 - accuracy: 0.98
Epoch 19/25
98/98 [=====] - 424s 4s/step - loss: 0.0563 - accuracy: 0.98
Epoch 20/25
98/98 [=====] - 426s 4s/step - loss: 0.0415 - accuracy: 0.99
Epoch 21/25
98/98 [=====] - 425s 4s/step - loss: 0.0386 - accuracy: 0.99
Epoch 22/25
98/98 [=====] - 425s 4s/step - loss: 0.0600 - accuracy: 0.98
Epoch 23/25
98/98 [=====] - 425s 4s/step - loss: 0.0519 - accuracy: 0.98
Epoch 24/25
98/98 [=====] - 432s 4s/step - loss: 0.1167 - accuracy: 0.96
Epoch 25/25
98/98 [=====] - 426s 4s/step - loss: 0.0424 - accuracy: 0.98

```

8. Save The Model

```

from tensorflow.keras.models import load_model

model.save('/content/drive/MyDrive/IBM/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/drive/MyDrive/IBM/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/drive/MyDrive/Car_damage/level/training/01-minor/0008.JPEG"
image = cv2.imread(data)
print(detect(image))

```

```
1/1 [=====] - 1s 536ms/step  
side
```

FOR LEVEL DAMAGE

▼ IMAGE PRE PROCESSING

1. Import The ImageDataGenerator Library

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

2. Configure ImageDataGenerator Class

```
train_datagen = ImageDataGenerator(rescale = 1./255,  
                                   shear_range = 0.1,  
                                   zoom_range = 0.1,  
                                   horizontal_flip = True)
```

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

3. Apply ImageDataGenerator Functionality To Trainset And Testset

```
training_set = train_datagen.flow_from_directory('/content/drive/MyDrive/Car_damage/level/  
                                                target_size = (224, 224),  
                                                batch_size = 10,  
                                                class_mode = 'categorical')  
test_set = test_datagen.flow_from_directory('/content/drive/MyDrive/Car_damage/level/train  
                                                target_size = (224, 224),  
                                                batch_size = 10,  
                                                class_mode = 'categorical')
```

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

▼ MODEL BUILDING

1. 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

```

2. Loading The Model

```

IMAGE_SIZE = [224, 224]

train_path = '/content/drive/MyDrive/Car_damage/level/training'
valid_path = '/content/drive/MyDrive/Car_damage/level/training'

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/drive/MyDrive/Car_damage/level/training/*')

folders

['/content/drive/MyDrive/Car_damage/body/training/02-side',
 '/content/drive/MyDrive/Car_damage/body/training/01-rear',
 '/content/drive/MyDrive/Car_damage/body/training/00-front']

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.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 |
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| 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(
    training_set,
    validation_data=test_set,
    epochs=25,
    steps_per_epoch=len(training_set),
    validation_steps=len(test_set)
)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: UserWarning: `Model.1

```
Epoch 1/25
98/98 [=====] - 615s 6s/step - loss: 1.2465 - accuracy: 0.55
Epoch 2/25
98/98 [=====] - 604s 6s/step - loss: 0.6654 - accuracy: 0.75
Epoch 3/25
98/98 [=====] - 604s 6s/step - loss: 0.5950 - accuracy: 0.76
Epoch 4/25
98/98 [=====] - 601s 6s/step - loss: 0.4964 - accuracy: 0.86
Epoch 5/25
98/98 [=====] - 603s 6s/step - loss: 0.3559 - accuracy: 0.86
Epoch 6/25
98/98 [=====] - 604s 6s/step - loss: 0.2425 - accuracy: 0.91
Epoch 7/25
98/98 [=====] - 604s 6s/step - loss: 0.1964 - accuracy: 0.93
Epoch 8/25
98/98 [=====] - 598s 6s/step - loss: 0.2119 - accuracy: 0.92
Epoch 9/25
98/98 [=====] - 597s 6s/step - loss: 0.1111 - accuracy: 0.96
Epoch 10/25
98/98 [=====] - 595s 6s/step - loss: 0.1394 - accuracy: 0.94
Epoch 11/25
98/98 [=====] - 598s 6s/step - loss: 0.1167 - accuracy: 0.96
Epoch 12/25
98/98 [=====] - 598s 6s/step - loss: 0.0823 - accuracy: 0.97
Epoch 13/25
98/98 [=====] - 602s 6s/step - loss: 0.1062 - accuracy: 0.96
Epoch 14/25
98/98 [=====] - 599s 6s/step - loss: 0.0717 - accuracy: 0.97
Epoch 15/25
98/98 [=====] - 598s 6s/step - loss: 0.0692 - accuracy: 0.98
Epoch 16/25
98/98 [=====] - 595s 6s/step - loss: 0.0449 - accuracy: 0.98
Epoch 17/25
98/98 [=====] - 609s 6s/step - loss: 0.0522 - accuracy: 0.98
Epoch 18/25
98/98 [=====] - 607s 6s/step - loss: 0.0386 - accuracy: 0.99
Epoch 19/25
98/98 [=====] - 595s 6s/step - loss: 0.0381 - accuracy: 0.99
Epoch 20/25
98/98 [=====] - 596s 6s/step - loss: 0.0196 - accuracy: 1.00
```

```

Epoch 21/25
98/98 [=====] - 597s 6s/step - loss: 0.0394 - accuracy: 0.99
Epoch 22/25
98/98 [=====] - 595s 6s/step - loss: 0.0377 - accuracy: 0.99
Epoch 23/25
98/98 [=====] - 595s 6s/step - loss: 0.0387 - accuracy: 0.99
Epoch 24/25
98/98 [=====] - 596s 6s/step - loss: 0.0279 - accuracy: 0.99
Epoch 25/25
98/98 [=====] - 603s 6s/step - loss: 0.0271 - accuracy: 0.99

```

8. Save The Model

```

from tensorflow.keras.models import load_model

model.save('/content/drive/MyDrive/IBM/Model/level.h5')

```

9. Test The Model

```

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

model = load_model('/content/drive/MyDrive/IBM/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/drive/MyDrive/Car_damage/level/validation/01-minor/0002.JPEG"
image = cv2.imread(data)
print(detect(image))

1/1 [=====] - 1s 510ms/step
severe

```

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