## Configure Image Data Generator

ImageDataGenerator class is instantiated and the configuration for the types of data augmentation

There are five main types of data augmentation techniques for image data; specifically:

Image shifts via the width shift range and height shift range arguments.

The image flips via the horizontal flip and vertical flip arguments.

Image rotations via the rotation range argument

Image brightness via the brightness range argument.

Image zoom via the zoom\_range argument.

An instance of the ImageDataGenerator class can be constructed for train and test.

## **CODING**

```
In [10]:
train_df, test_df = train_test_split(data, test_size=0.2)
In [11]:
train datagen = ImageDataGenerator(
    preprocessing_function=preprocess_input,
    validation_split=0.2
test datagen = ImageDataGenerator(
    preprocessing_function=preprocess_input
)
train_gen = train_datagen.flow_from_dataframe(
    dataframe=train df,
    x_col='image',
    y_col='classes',
    target_size=(224, 224),
    color_mode='rgb',
    batch_size=32,
    shuffle=True,
    seed=0
val_gen = train_datagen.flow_from_dataframe(
    dataframe=train df,
    x_col='image',
    y_col='classes',
```

```
target_size=(224, 224),
   batch size=32,
   shuffle=True,
   seed=0
)
test_gen = test_datagen.flow_from_dataframe(
   dataframe=test_df,
   x_col='image',
   y col='classes'
   target size=(224, 224),
   color_mode='rgb',
   class mode='categorical',
   batch_size=32,
   shuffle=False
)
Found 1275 validated image filenames belonging to 8 classes.
Found 1275 validated image filenames belonging to 8 classes.
Found 319 validated image filenames belonging to 8 classes.
In [13]:
pretrained model = MobileNetV2(
   input_shape=(224, 224, 3),
   include_top=False,
   weights='imagenet',
   pooling='avg'
)
pretrained model.trainable = False
2021-09-29 13:56:21.323371: I tensorflow/compiler/jit/xla cpu device.cc:41] Not c
reating XLA devices, tf_xla_enable_xla_devices not set
2021-09-29 13:56:21.326316: W tensorflow/stream executor/platform/default/dso loa
der.cc:60] Could not load dynamic library 'libcuda.so.1'; dlerror: libcuda.so.1:
cannot open shared object file: No such file or directory; LD LIBRARY PATH: /opt/
conda/lib
2021-09-29 13:56:21.326351: W tensorflow/stream executor/cuda/cuda driver.cc:326]
failed call to cuInit: UNKNOWN ERROR (303)
2021-09-29 13:56:21.326382: I tensorflow/stream executor/cuda/cuda diagnostics.cc
:156] kernel driver does not appear to be running on this host (38ac2691f9a8): /p
roc/driver/nvidia/version does not exist
2021-09-29 13:56:21.326737: I tensorflow/core/platform/cpu feature guard.cc:142]
This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneD
NN) to use the following CPU instructions in performance-critical operations: AV
X2 FMA
To enable them in other operations, rebuild TensorFlow with the appropriate compi
ler flags.
2021-09-29 13:56:21.327090: I tensorflow/compiler/jit/xla gpu device.cc:99] Not c
reating XLA devices, tf_xla_enable_xla_devices not set
Downloading data from https://storage.googleapis.com/tensorflow/keras-application
s/mobilenet v2/mobilenet v2 weights tf dim ordering tf kernels 1.0 224 no top.h5
9412608/9406464 [=========== ] - Os Ous/step
In [14]:
inputs = pretrained model.input
```

```
x = Dense(120, activation='relu')(pretrained_model.output)
x = Dense(120, activation='relu')(x)
outputs = Dense(8, activation='softmax')(x)
model = Model(inputs=inputs, outputs=outputs)
In [15]:
model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy']
In [16]:
my callbacks = [EarlyStopping(monitor='val accuracy',
                       min delta=0,
                       patience=2,
                       mode='auto')]
In [17]:
history = model.fit(train gen, validation data=val gen, epochs=50, callbacks=my callb
acks)
2021-09-29 13:56:23.423199: I tensorflow/compiler/mlir_graph_optimization_pa
ss.cc:116] None of the MLIR optimization passes are enabled (registered 2)
2021-09-29 13:56:23.429278: I tensorflow/core/platform/profile utils/cpu utils.cc
:112] CPU Frequency: 2199995000 Hz
Epoch 1/50
.4445 - val loss: 0.6562 - val accuracy: 0.8016
Epoch 2/50
40/40 [============== ] - 67s 2s/step - loss: 0.6024 - accuracy: 0
.8231 - val loss: 0.3426 - val accuracy: 0.9106
Epoch 3/50
.8751 - val_loss: 0.2268 - val_accuracy: 0.9420
Epoch 4/50
.9464 - val loss: 0.2268 - val accuracy: 0.9255
40/40 [=============== ] - 67s 2s/step - loss: 0.1770 - accuracy: 0
.9496 - val_loss: 0.0854 - val_accuracy: 0.9851
Epoch 6/50
40/40 [============== ] - 67s 2s/step - loss: 0.0672 - accuracy: 0
.9925 - val loss: 0.0541 - val accuracy: 0.9969
Epoch 7/50
40/40 [=============== ] - 88s 2s/step - loss: 0.0717 - accuracy: 0
.9869 - val loss: 0.0275 - val accuracy: 0.9992
Epoch 8/50
40/40 [============== ] - 69s 2s/step - loss: 0.0300 - accuracy: 0
.9967 - val loss: 0.0147 - val_accuracy: 1.0000
.0000 - val_loss: 0.0094 - val_accuracy: 1.0000
Epoch 10/50
40/40 [============== ] - 69s 2s/step - loss: 0.0085 - accuracy: 1
.0000 - val loss: 0.0076 - val accuracy: 1.0000
In [18]:
```

```
# Plotting Accuracy and val_accuracy
pd.DataFrame(history.history)[['accuracy','val_accuracy']].plot()
plt.title("Accuracy")
plt.show()
# Plotting loss and val loss
pd.DataFrame(history.history)[['loss','val_loss']].plot()
plt.title("Loss")
plt.show()
In [19]:
# Calculating Test Accuracy and Loss
results = model.evaluate(test_gen, verbose=0)
print("
          Test Loss: {:.5f}".format(results[0]))
print("Test Accuracy: {:.2f}%".format(results[1] * 100))
    Test Loss: 0.79887
Test Accuracy: 78.06%
In [20]:
pred = model.predict(test_gen )
pred = np.argmax(pred,axis=1)
# Map the Label
labels = (train_gen.class_indices)
labels = dict((v,k) for k,v in labels.items())
pred = [labels[k] for k in pred]
In [21]:
# Classification report
y_test = list(test_df.classes)
print(classification_report(y_test, pred))
                             recall f1-score
                precision
                                                  support
   bumper dent
                     0.55
                                0.63
                                          0.59
                                                       27
bumper_scratch
                      0.91
                                1.00
                                           0.95
                                                       39
     door_dent
                     0.71
                                0.61
                                           0.66
                                                       41
                                                       25
  door scratch
                     0.65
                                0.80
                                          0.71
 glass_shatter
                     0.84
                                0.84
                                          0.84
                                                       25
     head_lamp
                     0.54
                                0.71
                                          0.61
                                                       21
                                                       37
     tail_lamp
                     0.86
                                0.68
                                          0.76
       unknown
                     0.90
                                0.84
                                          0.87
                                                      104
                                          0.78
                                                      319
      accuracy
                     0.74
                                0.76
                                          0.75
                                                      319
     macro avg
  weighted avg
                     0.79
                                0.78
                                          0.78
                                                      319
In [22]:
linkcode
fig, axes = plt.subplots(nrows=2, ncols=4, figsize=(15, 7),
                        subplot_kw={'xticks': [], 'yticks': []})
```

```
for i, ax in enumerate(axes.flat):
    ax.imshow(plt.imread(test_df.image.iloc[i]))
    ax.set_title(f"True: {test_df.classes.iloc[i]}\nPredicted: {pred[i]}")
plt.tight_layout()
plt.show()
```

## **Image Data Augmentation**

```
#image data augmentation to the testing data.
val_datagen = ImageDataGenerator(rescale = 1./255)
```