

Transfer Learning Inception V3

In [2]: *# importing the libraries*

```
from tensorflow.keras.layers import Dense, Flatten
from tensorflow.keras.models import Model
from tensorflow.keras.applications.inception_v3 import InceptionV3

from tensorflow.keras.preprocessing.image import ImageDataGenerator, load_img
import numpy as np
from glob import glob
```

In [3]: *# re-size all the images to this*

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

```
train_path = 'Datasets/train'
test_path = 'Datasets/test'
valid_path = 'Datasets/val'
```

In [5]: *# Import the InceptionV3 library as shown below and add preprocessing layer to the model*
Here we will be using imagenet weights

```
inception = InceptionV3(input_shape=IMAGE_SIZE + [3], weights='imagenet', include_top=False)
```

In [4]: *# don't train existing weights*

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

In [5]: *# useful for getting number of output classes*

```
folders = glob('Datasets/train/*')
```

In [6]: *# output layers - you can add more if you want*

```
x = Flatten()(inception.output)
prediction = Dense(len(folders), activation='softmax')(x)
```

In [7]:

```
# create a model object
model = Model(inputs=inception.input, outputs=prediction)
```

In [8]:

```
# view the structure of the model
model.summary()
```

```

batch_normalization_48 (BatchNo (None, 12, 12, 192) 576      conv2d_48[0]
[0])

batch_normalization_49 (BatchNo (None, 12, 12, 192) 576      conv2d_49[0]
[0])

activation_40 (Activation)      (None, 12, 12, 192) 0      batch_normal
ization_40[0][0])

activation_43 (Activation)      (None, 12, 12, 192) 0      batch_normal
ization_43[0][0])

activation_48 (Activation)      (None, 12, 12, 192) 0      batch_normal
ization 48[0][0])

```

In [9]:

```
# tell the model what cost and optimization method to use
model.compile(
    loss='categorical_crossentropy',
    optimizer='adam',
    metrics=['accuracy']
)
```

In [10]:

```
# Use the Image Data Generator to import the images from the dataset
from tensorflow.keras.preprocessing.image import ImageDataGenerator

train_datagen = ImageDataGenerator(rescale = 1./255,
                                   shear_range = 0.2,
                                   zoom_range = 0.2,
                                   horizontal_flip = True)

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

In [13]:

```
# Make sure you provide the same target size as initialied for the image size
training_set = train_datagen.flow_from_directory('Datasets/train',
                                                target_size = (224, 224),
                                                batch_size = 32,
                                                class_mode = 'categorical')
```

Found 744 images belonging to 4 classes.

```
In [14]: validation_set = test_datagen.flow_from_directory('Datasets/val',
                                                         target_size = (224, 224),
                                                         batch_size = 32,
                                                         class_mode = 'categorical')
```

Found 101 images belonging to 4 classes.

fitting the model

```
In [33]: r = model.fit(
    training_set,
    validation_data=validation_set,
    epochs=10,
    steps_per_epoch=len(training_set),
    validation_steps=len(validation_set)
)
```

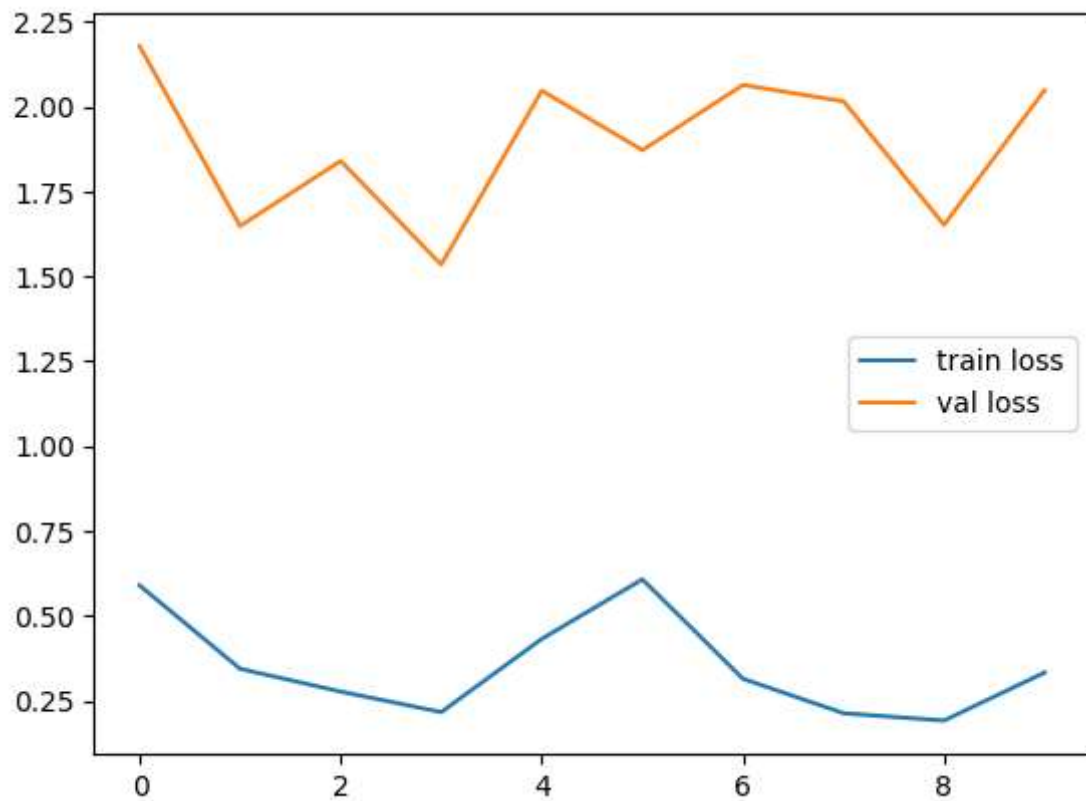
```
Epoch 1/10
24/24 [=====] - 64s 3s/step - loss: 0.5898 - accuracy: 0.8495 - val_loss: 2.1789 - val_accuracy: 0.6832
Epoch 2/10
24/24 [=====] - 60s 2s/step - loss: 0.3438 - accuracy: 0.9073 - val_loss: 1.6488 - val_accuracy: 0.7030
Epoch 3/10
24/24 [=====] - 63s 3s/step - loss: 0.2767 - accuracy: 0.9234 - val_loss: 1.8407 - val_accuracy: 0.6931
Epoch 4/10
24/24 [=====] - 59s 2s/step - loss: 0.2161 - accuracy: 0.9368 - val_loss: 1.5350 - val_accuracy: 0.7129
Epoch 5/10
24/24 [=====] - 58s 2s/step - loss: 0.4320 - accuracy: 0.8978 - val_loss: 2.0477 - val_accuracy: 0.6832
Epoch 6/10
24/24 [=====] - 58s 2s/step - loss: 0.6074 - accuracy: 0.8737 - val_loss: 1.8724 - val_accuracy: 0.7228
Epoch 7/10
24/24 [=====] - 50s 2s/step - loss: 0.3145 - accuracy: 0.9145 - val_loss: 1.5350 - val_accuracy: 0.7129
```

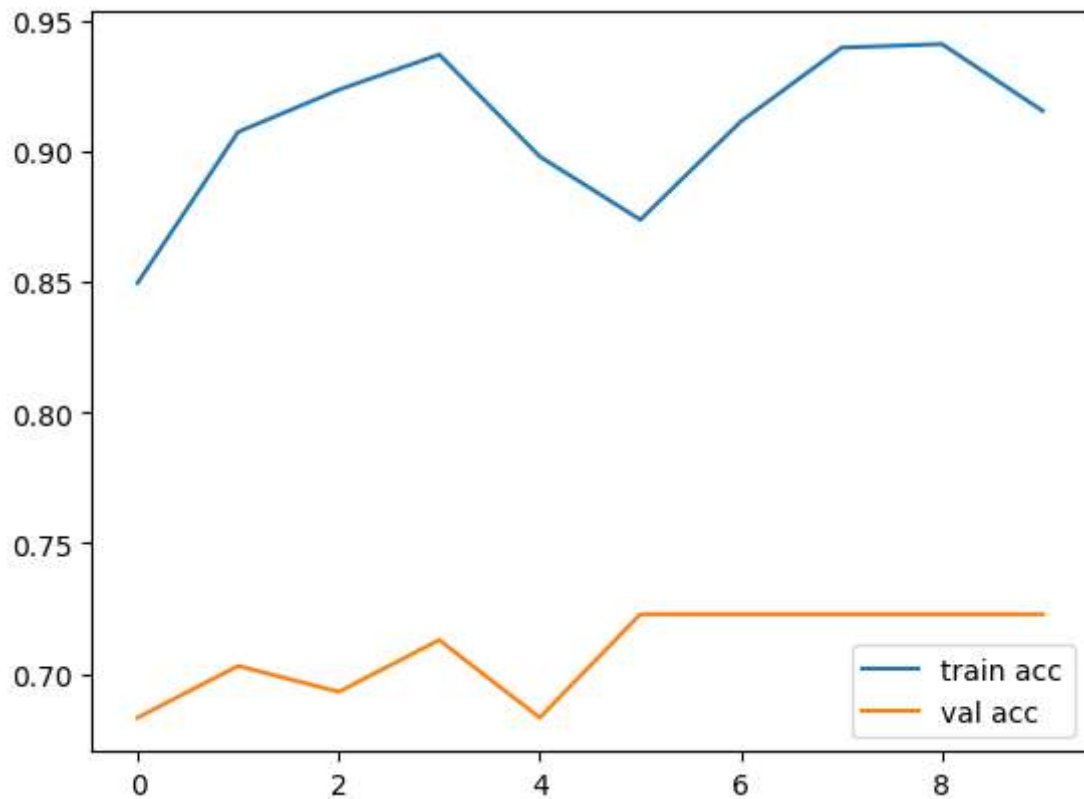
Plotting loss and accuracy

```
In [16]: import matplotlib.pyplot as plt
```

```
In [34]: # plot the loss
plt.plot(r.history['loss'], label='train loss')
plt.plot(r.history['val_loss'], label='val loss')
plt.legend()
plt.show()
plt.savefig('LossVal_loss')

# plot the accuracy
plt.plot(r.history['accuracy'], label='train acc')
plt.plot(r.history['val_accuracy'], label='val acc')
plt.legend()
plt.show()
plt.savefig('AccVal_acc')
```





Saving the model to reuse

```
In [35]: model.save('model_inception_car_brand.h5')
```

predicting using validation set

```
In [36]: y_pred = model.predict(validation_set)
```

```
In [37]: import numpy as np
y_pred = np.argmax(y_pred, axis=1)#getting index having max value
```

```
In [38]: y_pred
```

```
Out[38]: array([1, 1, 1, 0, 1, 0, 2, 2, 1, 1, 0, 3, 3, 0, 2, 3, 0, 3, 0, 0, 3, 0,
                0, 1, 3, 2, 3, 1, 3, 1, 0, 2, 0, 0, 3, 3, 0, 0, 2, 2, 1, 0, 0, 2,
                3, 1, 0, 1, 2, 1, 3, 3, 3, 1, 2, 1, 2, 0, 0, 3, 1, 0, 3, 0, 2, 1,
                1, 0, 1, 1, 3, 1, 0, 2, 2, 0, 0, 1, 1, 1, 0, 1, 2, 1, 2, 1, 0, 3,
                3, 1, 3, 3, 0, 1, 2, 1, 0, 1, 0, 2, 0], dtype=int64)
```

Predicting Multiple images in test set

```
In [39]: from tensorflow.keras.models import load_model  
         from tensorflow.keras.preprocessing import image
```

```
In [40]: model=load_model('model_inception_car_brand.h5')
```

```
In [43]: import glob  
         cv_img=[]  
         col_dir='Datasets/test/Jaguar/*.jpg'  
         for img in glob.glob(col_dir):  
             img=image.load_img(img,target_size=(224,224))  
             x=image.img_to_array(img)  
             x=x/255  
             x=np.expand_dims(x,axis=0)  
             preds = model.predict(x)  
             preds=np.argmax(preds, axis=1)  
             if preds==0:  
                 preds="The car is BMW"  
             elif preds==1:  
                 preds="The car is Jaguar"  
             elif preds==2:  
                 preds="The car is ROLls Royce"  
             else:  
                 preds="The car is BEnz"  
             print(preds)
```

```
The car is Jaguar  
The car is Jaguar  
The car is BMW  
The car is BMW  
The car is Jaguar  
The car is Jaguar  
The car is Jaguar  
The car is Jaguar  
The car is Jaguar  
The car is Jaguar
```

```
In [ ]:
```