

(A)VGG16 Model to train and test dataset of car-views

1.Import Libraries

```
In [61]: import tensorflow as tf
import os
import numpy as np
from tensorflow.keras.layers import Input, Flatten, Dense
from tensorflow.keras.models import Model
from tensorflow.keras.applications.vgg16 import VGG16
from tensorflow.keras.models import Sequential
import matplotlib.pyplot as plt
```

2.Image data generator

```
In [62]: IMAGE_SIZE=224
BATCH_SIZE=64

train_datagen=tf.keras.preprocessing.image.ImageDataGenerator(
    rescale=1./255,
    zoom_range=0.2,
    horizontal_flip=True,
    validation_split=0.1)

validation_datagen=tf.keras.preprocessing.image.ImageDataGenerator(
    rescale=1./255,
    validation_split=0.1
)
```

```
In [64]: train="training"
train_generator=train_datagen.flow_from_directory(
    train,
    target_size=(IMAGE_SIZE, IMAGE_SIZE),
    batch_size=BATCH_SIZE
)

test="validation"
validation_generator=validation_datagen.flow_from_directory(
    test,
    target_size=(IMAGE_SIZE, IMAGE_SIZE),
    batch_size=BATCH_SIZE
)

Found 960 images belonging to 3 classes.
Found 171 images belonging to 3 classes.
```

```
In [74]: print("Integer values of classes:")
train_generator.class_indices

Integer values of classes:
Out[74]: {'front': 0, 'rear': 1, 'side': 2}
```

3.VGG16 model

```
In [65]: IMAGE_SIZE=[224, 224]
vgg=VGG16(input_shape=IMAGE_SIZE+[3],weights='imagenet',include_top=False)
vgg.output
```

Out[65]: <KerasTensor: shape=(None, 7, 7, 512) dtype=float32 (created by layer 'block5_pool')>

```
In [66]: for layer in vgg.layers:
layer.trainable=False
```

```
In [67]: x=Flatten()(vgg.output)
prediction=Dense(3,activation='softmax')(x)
model=Model(inputs=vgg.input,outputs=prediction)
model.summary()
```

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_2 (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		

4.Train the model

```
In [75]: model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

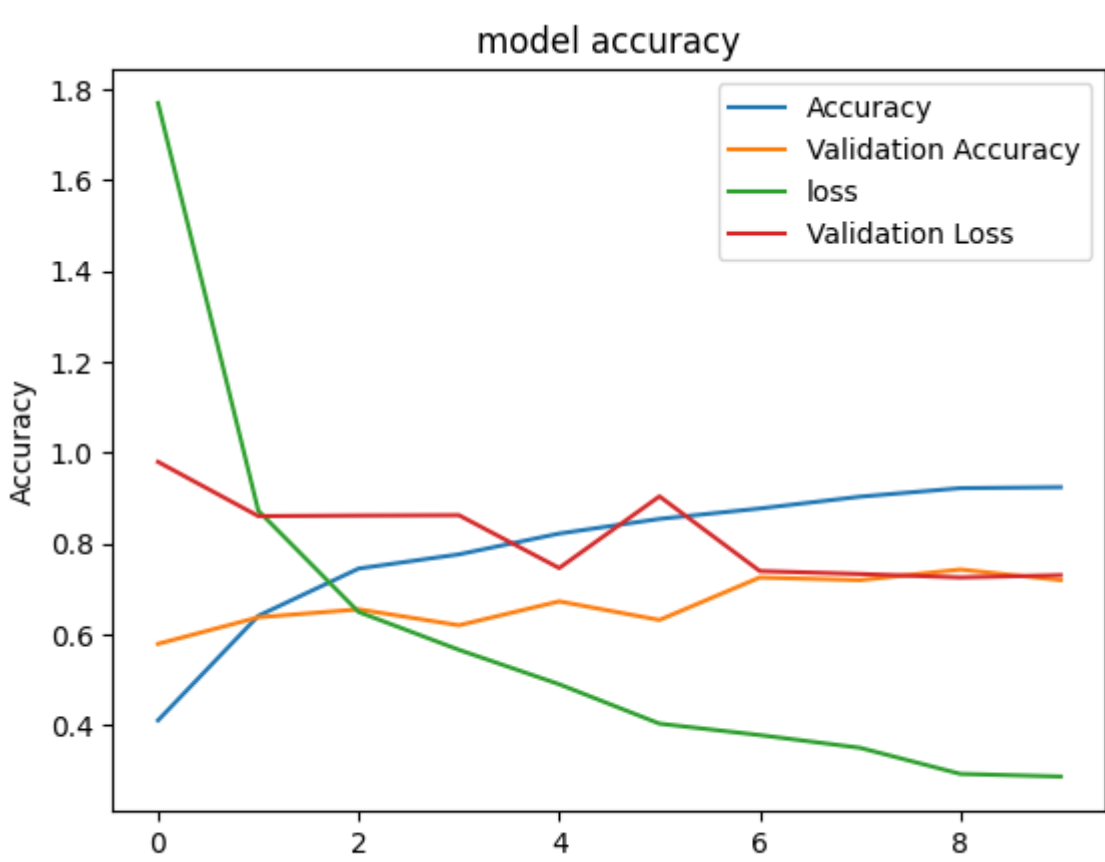
```
In [69]: epoch=10
```

```
history=model.fit(train_generator,
    steps_per_epoch=len(train_generator),
    epochs=epoch,
    validation_data=validation_generator,
    validation_steps=len(validation_generator)
)
```

Epoch 1/10
15/15 [=====] - 223s 15s/step - loss: 1.7701 - accuracy: 0.4104 - val_loss: 0.9798 - val_accuracy: 0.5789
Epoch 2/10
15/15 [=====] - 251s 17s/step - loss: 0.8721 - accuracy: 0.6406 - val_loss: 0.8603 - val_accuracy: 0.6374
Epoch 3/10
15/15 [=====] - 239s 16s/step - loss: 0.6492 - accuracy: 0.7448 - val_loss: 0.8616 - val_accuracy: 0.6550
Epoch 4/10
15/15 [=====] - 243s 16s/step - loss: 0.5658 - accuracy: 0.7760 - val_loss: 0.8623 - val_accuracy: 0.6199
Epoch 5/10
15/15 [=====] - 302s 21s/step - loss: 0.4900 - accuracy: 0.8219 - val_loss: 0.7459 - val_accuracy: 0.6725
Epoch 6/10
15/15 [=====] - 357s 24s/step - loss: 0.4032 - accuracy: 0.8542 - val_loss: 0.9038 - val_accuracy: 0.6316
Epoch 7/10
15/15 [=====] - 337s 22s/step - loss: 0.3780 - accuracy: 0.8771 - val_loss: 0.7393 - val_accuracy: 0.7251
Epoch 8/10
15/15 [=====] - 236s 16s/step - loss: 0.3499 - accuracy: 0.9031 - val_loss: 0.7328 - val_accuracy: 0.7193
Epoch 9/10
15/15 [=====] - 243s 16s/step - loss: 0.2925 - accuracy: 0.9219 - val_loss: 0.7252 - val_accuracy: 0.7427
Epoch 10/10
15/15 [=====] - 249s 17s/step - loss: 0.2867 - accuracy: 0.9240 - val_loss: 0.7302 - val_accuracy: 0.7193

Model accuracy

```
In [70]: plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title("model accuracy")
plt.ylabel("Accuracy")
plt.xlabel("Epoch")
plt.legend(['Accuracy', "Validation Accuracy", "loss", "Validation Loss"])
plt.show()
```



5.Test the model

Test_Image1

```
In [71]: from tensorflow.keras.utils import array_to_img
from tensorflow.keras.utils import load_img
from tensorflow.keras.utils import img_to_array
import numpy as np
img_pred=load_img("test/side.jpg",target_size=(224,224))
plt.imshow(img_pred, cmap=plt.get_cmap('gray'))

img_pred=img_to_array(img_pred)

img_pred=np.expand_dims(img_pred, axis=0)

rslt= model.predict(img_pred)

print(rslt)
if rslt[0][0]>rslt[0][1]:
    if rslt[0][2]>rslt[0][0]:
        prediction="rear_side"
    else:
        prediction="front"
else:
    prediction="side"
print(prediction)
```

1/1 [=====] - 1s 727ms/step

[0. 0. 1.]

side



Test_Image2

```
In [72]: from tensorflow.keras.utils import load_img
from tensorflow.keras.utils import img_to_array
import numpy as np
img_pred=load_img("test/frontside.jpg",target_size=(224,224))
plt.imshow(img_pred, cmap=plt.get_cmap('gray'))

img_pred=img_to_array(img_pred)

img_pred=np.expand_dims(img_pred, axis=0)

rslt= model.predict(img_pred)

print(rslt)
if rslt[0][0]>rslt[0][1]:
    if rslt[0][2]>rslt[0][0]:
        prediction="rear_side"
    else:
        prediction="front"
else:
    prediction="side"
print(prediction)
```

1/1 [=====] - 0s 303ms/step

[1. 0. 0.]

front

