

## Day 01

```

1 from google.colab import drive
2 drive.mount('/content/drive')
3
4 data_path = '/content/drive/MyDrive/F1CarsDataset/Formula One Cars'
5

```



Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

```

1 from tensorflow.keras.preprocessing.image import ImageDataGenerator
2
3 IMG_SIZE = (224, 224)
4 BATCH_SIZE = 32
5
6 datagen = ImageDataGenerator(
7     rescale=1./255,    # Normalize pixel values
8     validation_split=0.2 # 20% validation split
9 )
10
11 train_generator = datagen.flow_from_directory(
12     data_path,
13     target_size=IMG_SIZE,
14     batch_size=BATCH_SIZE,
15     class_mode='categorical',
16     subset='training',
17     color_mode='rgb' # Add this to ensure images are converted to RGB
18 )
19
20 val_generator = datagen.flow_from_directory(
21     data_path,
22     target_size=IMG_SIZE,
23     batch_size=BATCH_SIZE,
24     class_mode='categorical',
25     subset='validation',
26     color_mode='rgb' # Add this to ensure images are converted to RGB
27 )

```



Found 1928 images belonging to 8 classes.  
Found 479 images belonging to 8 classes.

```

1 import matplotlib.pyplot as plt
2 import numpy as np
3
4 class_labels = list(train_generator.class_indices.keys())
5
6 images, labels = next(train_generator) # get a batch
7 plt.figure(figsize=(10,10))
8 for i in range(9):

```

```

9 plt.subplot(3,3,i+1)
10 plt.imshow(images[i])
11 plt.title(class_labels[np.argmax(labels[i])])
12 plt.axis('off')
13 plt.show()
14

```



Ferrari F1 car



Williams F1 car



Williams F1 car



Red Bull Racing F1 car



Williams F1 car



Williams F1 car



McLaren F1 car



Williams F1 car



Ferrari F1 car



```
1 print(f"Classes found: {train_generator.class_indices}")
2 print(f"Number of training samples: {train_generator.samples}")
3 print(f"Number of validation samples: {val_generator.samples}")
4
```



Classes found: {'AlphaTauri F1 car': 0, 'Ferrari F1 car': 1, 'McLaren F1 car': 2, 'Mercedes F1 car': 3, 'Racing Point F1 car': 4, 'Red Bull Racing F1 car': 5, 'Scuderia Toro Rosso F1 car': 6, 'Williams F1 car': 7}  
Number of training samples: 1928  
Number of validation samples: 479

## Day 02

```
1 import tensorflow as tf
2 from tensorflow.keras.models import Sequential
3 from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
4 import matplotlib.pyplot as plt
5
```

```
1 from tensorflow.keras.preprocessing.image import ImageDataGenerator
2
3 data_path = '/content/drive/MyDrive/F1CarsDataset/Formula One Cars'
4
5 IMG_SIZE = (224, 224)
6 BATCH_SIZE = 32
7
8 datagen = ImageDataGenerator(
9     rescale=1./255,
10    validation_split=0.2
11 )
12
13 train_generator = datagen.flow_from_directory(
14     data_path,
15     target_size=IMG_SIZE,
16     batch_size=BATCH_SIZE,
17     class_mode='categorical',
18     subset='training'
19 )
20
21 val_generator = datagen.flow_from_directory(
22     data_path,
23     target_size=IMG_SIZE,
24     batch_size=BATCH_SIZE,
25     class_mode='categorical',
26     subset='validation'
27 )
28
```



Found 1928 images belonging to 8 classes.  
Found 479 images belonging to 8 classes.

```
1 import tensorflow as tf
2 from tensorflow.keras.models import Sequential
3 from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
4
5 num_classes = train_generator.num_classes
6 input_shape = (224, 224, 3)
7
8 model = Sequential([
9     Conv2D(32, (3,3), activation='relu', input_shape=input_shape),
10    MaxPooling2D(2, 2),
11
12    Conv2D(64, (3,3), activation='relu'),
13    MaxPooling2D(2, 2),
14
15    Conv2D(128, (3,3), activation='relu'),
16    MaxPooling2D(2, 2),
17
18    Flatten(),
19    Dense(256, activation='relu'),
20    Dropout(0.5),
21    Dense(num_classes, activation='softmax')
22 ])
23
24 model.compile(
25     optimizer='adam',
26     loss='categorical_crossentropy',
27     metrics=['accuracy']
28 )
29
30 model.summary()
31
```

```

/usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv
super().__init__(activity_regularizer=activity_regularizer, **kwargs)

```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 222, 222, 32)	896
max_pooling2d (MaxPooling2D)	(None, 111, 111, 32)	0
conv2d_1 (Conv2D)	(None, 109, 109, 64)	18,496
max_pooling2d_1 (MaxPooling2D)	(None, 54, 54, 64)	0
conv2d_2 (Conv2D)	(None, 52, 52, 128)	73,856
max_pooling2d_2 (MaxPooling2D)	(None, 26, 26, 128)	0
flatten (Flatten)	(None, 86528)	0
dense (Dense)	(None, 256)	22,151,424
dropout (Dropout)	(None, 256)	0
dense_1 (Dense)	(None, 8)	2,056

Total params: 22,246,728 (84.86 MB)

Trainable params: 22,246,728 (84.86 MB)

Non-trainable params: 0 (0.00 B)

```

1 from PIL import Image
2 import os
3
4 data_path = '/content/drive/MyDrive/F1CarsDataset/Formula One Cars'
5
6 for root, dirs, files in os.walk(data_path):
7     for fname in files:
8         fpath = os.path.join(root, fname)
9         try:
10             img = Image.open(fpath)
11             img.verify() # verify if image file is broken
12         except (IOError, SyntaxError) as e:
13             print(f"Bad file detected: {fpath} - {e}")
14

```

```

1 import os
2 from PIL import Image
3
4 # Path to your dataset folder
5 data_path = '/content/drive/MyDrive/F1CarsDataset/Formula One Cars'
6
7 # Allowed image extensions
8 valid_extensions = {'.jpg', '.jpeg', '.png', '.bmp', '.gif'}
9

```

```

10 deleted_files = []
11
12 for root, dirs, files in os.walk(data_path):
13     for fname in files:
14         fpath = os.path.join(root, fname)
15         ext = os.path.splitext(fname)[1].lower()
16
17         # Remove files with invalid extensions
18         if ext not in valid_extensions:
19             os.remove(fpath)
20             deleted_files.append(f"Removed non-image file: {fpath}")
21             continue
22
23         # Check if image is corrupted
24         try:
25             img = Image.open(fpath)
26             img.verify() # Verify image integrity
27         except Exception as e:
28             os.remove(fpath)
29             deleted_files.append(f"Removed corrupted image: {fpath}")
30
31 print(f"Total files removed: {len(deleted_files)}")
32 for msg in deleted_files:
33     print(msg)
34

```

 Total files removed: 0

```

1 epochs = 10 # start with 10, adjust as needed
2
3 history = model.fit(
4     train_generator,
5     validation_data=val_generator,
6     epochs=epochs
7 )
8


```

 /usr/local/lib/python3.11/dist-packages/keras/src/trainers/data\_adapters/py\_dataset\_adapter.py:121: UserWarning: Your `self._warn_if_super_not_called()`

/usr/local/lib/python3.11/dist-packages/PIL/Image.py:1047: UserWarning: Palette images with Transparency expressed in RGB should warn of banded colors

warnings.warn(


Epoch 1/10

**61/61**  **309s** 5s/step - accuracy: 0.1765 - loss: 3.0811 - val\_accuracy: 0.3549 - val\_loss: 1.60

Epoch 2/10

**61/61**  **261s** 4s/step - accuracy: 0.3839 - loss: 1.6410 - val\_accuracy: 0.5052 - val\_loss: 1.53


Epoch 3/10

**61/61**  **263s** 4s/step - accuracy: 0.5592 - loss: 1.2820 - val\_accuracy: 0.6159 - val\_loss: 1.11

Epoch 4/10

**61/61**  **261s** 4s/step - accuracy: 0.7010 - loss: 0.9136 - val\_accuracy: 0.6347 - val\_loss: 1.00

Epoch 5/10

**61/61**  **262s** 4s/step - accuracy: 0.8039 - loss: 0.6301 - val\_accuracy: 0.6305 - val\_loss: 1.00

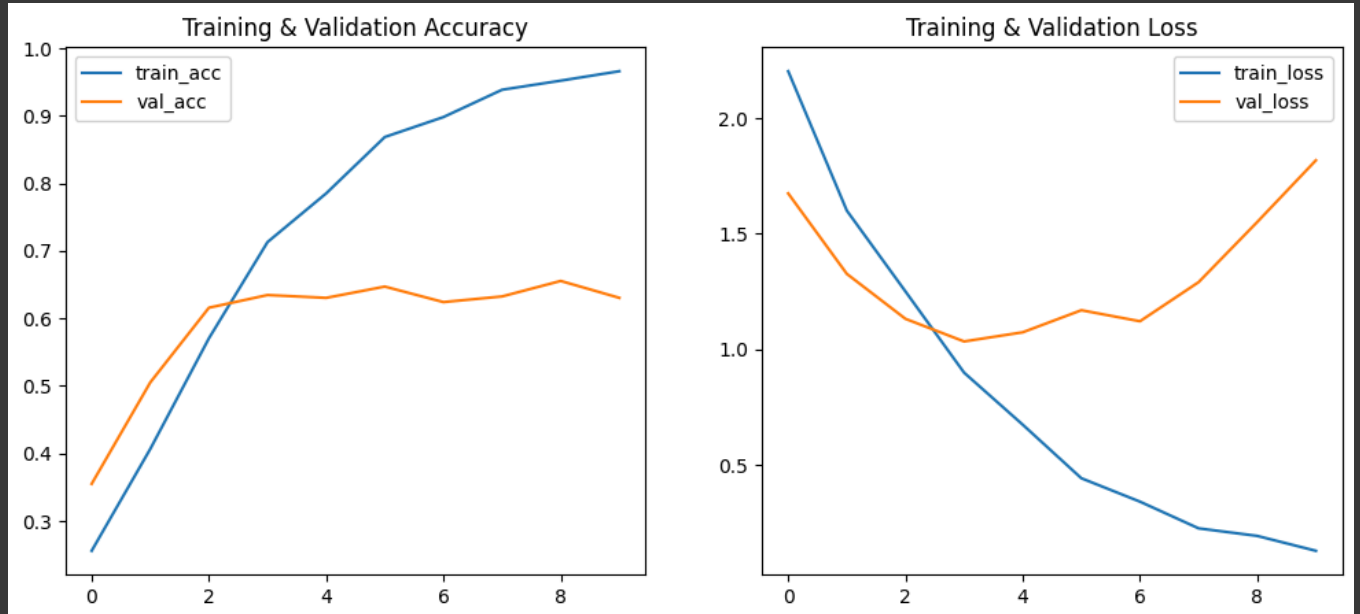
Epoch 6/10

**61/61**  **262s** 4s/step - accuracy: 0.8715 - loss: 0.4415 - val\_accuracy: 0.6472 - val\_loss: 1.11

Epoch 7/10

61/61 ————— 262s 4s/step - accuracy: 0.9035 - loss: 0.3329 - val\_accuracy: 0.6242 - val\_loss: 1.1  
Epoch 8/10  
61/61 ————— 262s 4s/step - accuracy: 0.9334 - loss: 0.2337 - val\_accuracy: 0.6326 - val\_loss: 1.2  
Epoch 9/10  
61/61 ————— 262s 4s/step - accuracy: 0.9487 - loss: 0.1947 - val\_accuracy: 0.6555 - val\_loss: 1.5  
Epoch 10/10  
61/61 ————— 279s 5s/step - accuracy: 0.9689 - loss: 0.1177 - val\_accuracy: 0.6305 - val\_loss: 1.8

```
1 plt.figure(figsize=(12,5))
2
3 plt.subplot(1,2,1)
4 plt.plot(history.history['accuracy'], label='train_acc')
5 plt.plot(history.history['val_accuracy'], label='val_acc')
6 plt.title("Training & Validation Accuracy")
7 plt.legend()
8
9 plt.subplot(1,2,2)
10 plt.plot(history.history['loss'], label='train_loss')
11 plt.plot(history.history['val_loss'], label='val_loss')
12 plt.title("Training & Validation Loss")
13 plt.legend()
14
15 plt.show()
16
```




```
1 model.save('/content/drive/MyDrive/F1CarsDataset/f1cars_cnn_model.h5')
2
```

 WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save\_model(model)`. T

## Day 03

```
1 from tensorflow.keras.preprocessing.image import ImageDataGenerator
2
3 data_path = '/content/drive/MyDrive/F1CarsDataset/Formula One Cars'
4 IMG_SIZE = (224, 224)
5 BATCH_SIZE = 32
6
7 train_datagen = ImageDataGenerator(
8     rescale=1./255,
9     rotation_range=20,
10    width_shift_range=0.2,
11    height_shift_range=0.2,
12    shear_range=0.15,
13    zoom_range=0.15,
14    horizontal_flip=True,
15    fill_mode='nearest',
16    validation_split=0.2
17 )
18
19 val_datagen = ImageDataGenerator(rescale=1./255, validation_split=0.2)
20
21 train_generator = train_datagen.flow_from_directory(
22     data_path,
23     target_size=IMG_SIZE,
24     batch_size=BATCH_SIZE,
25     class_mode='categorical',
26     subset='training',
27     shuffle=True
28 )
29
30 val_generator = val_datagen.flow_from_directory(
31     data_path,
32     target_size=IMG_SIZE,
33     batch_size=BATCH_SIZE,
34     class_mode='categorical',
35     subset='validation',
36     shuffle=False
37 )
38
```

 Found 1928 images belonging to 8 classes.  
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```
1 import tensorflow as tf
2
```



```
3 base_model = tf.keras.applications.MobileNetV2(
4     input_shape=(224, 224, 3),
5     include_top=False,
6     weights='imagenet'
7 )
8
9 base_model.trainable = False # freeze base layers initially
10
11 model = tf.keras.Sequential([
12     base_model,
13     tf.keras.layers.GlobalAveragePooling2D(),
14     tf.keras.layers.Dropout(0.3),
15     tf.keras.layers.Dense(train_generator.num_classes, activation='softmax')
16 ])
17
18 model.compile(
19     optimizer='adam',
20     loss='categorical_crossentropy',
21     metrics=['accuracy']
22 )
23
24 model.summary()
25
```

Downloading data from <https://storage.googleapis.com/tensorflow/keras-applications/9406464/9406464> 0s 0us/step  
Model: "sequential\_1"

Layer (type)	Output Shape	Param #
mobilenetv2_1.00_224 (Functional)	(None, 7, 7, 1280)	2,257,984
global_average_pooling2d (GlobalAveragePooling2D)	(None, 1280)	0
dropout_1 (Dropout)	(None, 1280)	0
dense_2 (Dense)	(None, 8)	10,248

Total params: 2,268,232 (8.65 MB)  
Trainable params: 10,248 (40.03 KB)  
Non-trainable params: 2,257,984 (8.61 MB)

```
1 epochs = 10
2
3 history = model.fit(
4     train_generator,
5     validation_data=val_generator,
6     epochs=epochs
7 )
8
```

Epoch 1/10  
61/61 178s 3s/step - accuracy: 0.2357 - loss: 2.1375 - val\_accuracy: 0.5470 - val\_loss: 1.5

```

Epoch 2/10
61/61 ————— 157s 3s/step - accuracy: 0.5218 - loss: 1.3816 - val_accuracy: 0.6534 - val_loss: 1.0
Epoch 3/10
61/61 ————— 169s 3s/step - accuracy: 0.6169 - loss: 1.1148 - val_accuracy: 0.6868 - val_loss: 0.9
Epoch 4/10
61/61 ————— 170s 3s/step - accuracy: 0.6483 - loss: 1.0352 - val_accuracy: 0.7182 - val_loss: 0.9
Epoch 5/10
61/61 ————— 168s 3s/step - accuracy: 0.6942 - loss: 0.9081 - val_accuracy: 0.7077 - val_loss: 0.8
Epoch 6/10
61/61 ————— 152s 2s/step - accuracy: 0.7059 - loss: 0.9053 - val_accuracy: 0.7349 - val_loss: 0.8
Epoch 7/10
61/61 ————— 154s 3s/step - accuracy: 0.7114 - loss: 0.8527 - val_accuracy: 0.7557 - val_loss: 0.7
Epoch 8/10
61/61 ————— 154s 3s/step - accuracy: 0.7465 - loss: 0.8075 - val_accuracy: 0.7182 - val_loss: 0.8
Epoch 9/10
61/61 ————— 153s 2s/step - accuracy: 0.7372 - loss: 0.7930 - val_accuracy: 0.7745 - val_loss: 0.7
Epoch 10/10
61/61 ————— 170s 3s/step - accuracy: 0.7309 - loss: 0.7612 - val_accuracy: 0.7808 - val_loss: 0.7

```

```
1 model.save('/content/drive/MyDrive/F1CarsDataset/f1cars_mobilenetv2_augmented.h5')
```

```
2
```

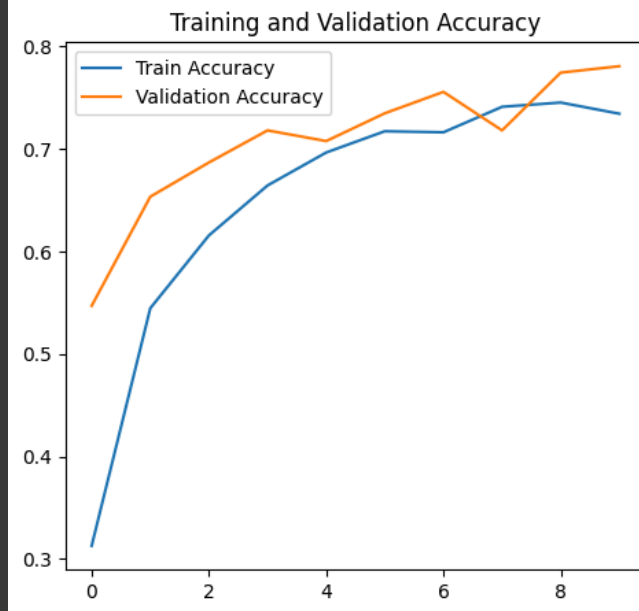


WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save\_model(model)`. T

```

1 import matplotlib.pyplot as plt
2
3 acc = history.history['accuracy'] + history_fine.history['accuracy'] if 'history_fine' in globals() else history.history['accuracy']
4 val_acc = history.history['val_accuracy'] + history_fine.history['val_accuracy'] if 'history_fine' in globals() else history.history['val_
5 loss = history.history['loss'] + history_fine.history['loss'] if 'history_fine' in globals() else history.history['loss']
6 val_loss = history.history['val_loss'] + history_fine.history['val_loss'] if 'history_fine' in globals() else history.history['val_loss']
7
8 plt.figure(figsize=(12,5))
9 plt.subplot(1,2,1)
10 plt.plot(acc, label='Train Accuracy')
11 plt.plot(val_acc, label='Validation Accuracy')
12 plt.title('Training and Validation Accuracy')
13 plt.legend()
14
15 plt.subplot(1,2,2)
16 plt.plot(loss, label='Train Loss')
17 plt.plot(val_loss, label='Validation Loss')
18 plt.title('Training and Validation Loss')
19 plt.legend()
20
21 plt.show()
22

```



## Day 04

```

1 import tensorflow as tf
2
3 model_path = '/content/drive/MyDrive/F1CarsDataset/f1cars_mobilenetv2_augmented.h5' # Adjust path/filename
4 model = tf.keras.models.load_model(model_path)
5

```



WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile\_metrics` wi

```

1 import numpy as np
2
3 # Reset the validation generator for consistent batch order
4 val_generator.reset()
5
6 # Predict probabilities for all validation images
7 pred_probs = model.predict(val_generator)
8
9 # Predicted class labels
10 pred_labels = np.argmax(pred_probs, axis=1)
11
12 # True class labels
13 true_labels = val_generator.classes
14

```

```

15 # Class label names
16 class_names = list(val_generator.class_indices.keys())
17

```

15/15 ————— 29s 2s/step

```

1 from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
2
3 print("Validation Accuracy:", accuracy_score(true_labels, pred_labels))
4 print("\nClassification Report:\n")
5 print(classification_report(true_labels, pred_labels, target_names=class_names))
6

```

Validation Accuracy: 0.7807933194154488

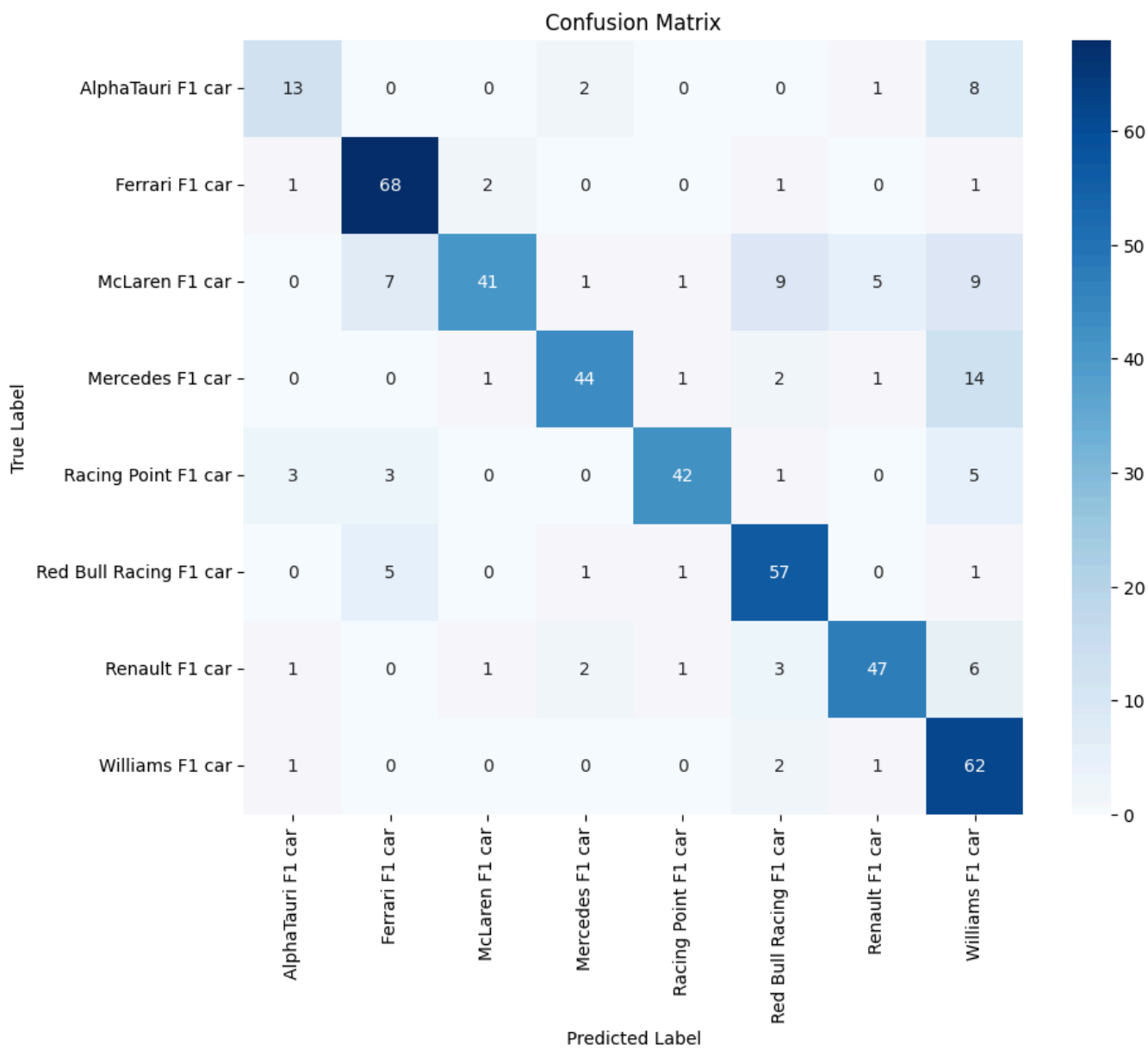
Classification Report:

	precision	recall	f1-score	support
AlphaTauri F1 car	0.68	0.54	0.60	24
Ferrari F1 car	0.82	0.93	0.87	73
McLaren F1 car	0.91	0.56	0.69	73
Mercedes F1 car	0.88	0.70	0.78	63
Racing Point F1 car	0.91	0.78	0.84	54
Red Bull Racing F1 car	0.76	0.88	0.81	65
Renault F1 car	0.85	0.77	0.81	61
Williams F1 car	0.58	0.94	0.72	66
accuracy		0.78		479
macro avg	0.80	0.76	0.77	479
weighted avg	0.81	0.78	0.78	479

```

1 import matplotlib.pyplot as plt
2 import seaborn as sns
3 from sklearn.metrics import ConfusionMatrixDisplay
4
5 cm = confusion_matrix(true_labels, pred_labels)
6 plt.figure(figsize=(10,8))
7 sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=class_names, yticklabels=class_names)
8 plt.ylabel('True Label')
9 plt.xlabel('Predicted Label')
10 plt.title('Confusion Matrix')
11 plt.show()
12

```



```

1 with open('/content/drive/MyDrive/F1CarsDataset/classification_report.txt', 'w') as f:
2     f.write(classification_report(true_labels, pred_labels, target_names=class_names))
3

```

## Day 05

```

1 import tensorflow as tf
2 import tensorflow_model_optimization as tfmot
3 import numpy as np
4

```

```

1 from tensorflow.keras.preprocessing.image import ImageDataGenerator
2
3 data_path = '/content/drive/MyDrive/F1CarsDataset/Formula One Cars'
4 IMG_SIZE = (224, 224)
5 BATCH_SIZE = 32
6
7 datagen = ImageDataGenerator(rescale=1./255, validation_split=0.2)
8
9 train_generator = datagen.flow_from_directory(
10     data_path,
11     target_size=IMG_SIZE,
12     batch_size=BATCH_SIZE,
13     class_mode='categorical',
14     subset='training'
15 )
16
17 val_generator = datagen.flow_from_directory(
18     data_path,
19     target_size=IMG_SIZE,
20     batch_size=BATCH_SIZE,
21     class_mode='categorical',
22     subset='validation'
23 )
24

```



Found 1928 images belonging to 8 classes.  
Found 479 images belonging to 8 classes.

```

1 num_classes = train_generator.num_classes
2 input_shape = (224, 224, 3)
3
4 model = tf.keras.Sequential([
5     tf.keras.layers.Conv2D(32, (3,3), activation='relu', input_shape=input_shape),
6     tf.keras.layers.MaxPooling2D(2, 2),
7     tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
8     tf.keras.layers.MaxPooling2D(2, 2),
9     tf.keras.layers.Flatten(),
10    tf.keras.layers.Dense(256, activation='relu'),
11    tf.keras.layers.Dropout(0.5),
12    tf.keras.layers.Dense(num_classes, activation='softmax'),
13 ])
14
15 model.compile(optimizer='adam',
16               loss='categorical_crossentropy',

```

```
17 metrics=['accuracy'])
```

```
18
```



```
/usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:113: UserWarning: Do not pass an `
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
1 epochs = 10
2 batch_size = train_generator.batch_size
3 num_training_samples = train_generator.samples
4
5 steps_per_epoch = np.ceil(num_training_samples / batch_size).astype(np.int32)
6 end_step = steps_per_epoch * epochs
7
```

```
1 import numpy as np
2
3 # Assuming you have your train_generator and epochs set
4 batch_size = train_generator.batch_size      # e.g., 32
5 epochs = 10                                # number of epochs to train
6 num_training_samples = train_generator.samples  # total training images
7
8 steps_per_epoch = np.ceil(num_training_samples / batch_size).astype(np.int32)
9 end_step = steps_per_epoch * epochs
10
```

```
1 import tensorflow_model_optimization as tfmot
2
3 pruning_params = {
4     'pruning_schedule': tfmot.sparsity.keras.PolynomialDecay(
5         initial_sparsity=0.0,
6         final_sparsity=0.5,
7         begin_step=0,
8         end_step=end_step
9     )
10 }
11
```