

Baseline Model: Implementation and Rationale

1. Goal of the Baseline

The aim of our baseline model was to quickly establish a from-scratch convolutional network to confirm our dataset loading, shape consistency, and minimal data augmentation pipeline. By building a small "two-block CNN," we gain an early reference point (both in accuracy and training behavior) before applying more sophisticated or pretrained approaches.

2. Architectural Decisions

- **Input Shape:** (180,180,3) to align with `image_dataset_from_directory(..., image_size=(180,180))`
- **Two Convolution Blocks:**
 - Each block has `Conv2D(32,3)` or `Conv2D(64,3)` repeated twice, then a `MaxPooling2D`
 - This standard pattern follows a typical VGG-like design, albeit in a smaller scale
- **Flatten → Dense:**
 - We flatten the pooled feature maps, then apply a `Dense(128, relu)` with a `Dropout(0.5)` for partial regularization
 - Lastly, a `Dense(num_classes, softmax)` for multi-class classification

3. Data Augmentation

Before the first convolution block, we incorporate Keras's built-in augmentation layers:

- `RandomFlip` horizontally
- `RandomRotation` ~10%
- `RandomZoom` ~10%

These transformations help the model learn invariance to flips, minor rotation, and scale changes, presumably beneficial given our limited dataset sizes (~100–172 images per class).

4. Key Observations from Training

- **Fluctuating Loss:** The training loss sometimes dips, then spikes. This can be symptomatic of a somewhat high learning rate or an architecture that is quickly overfitting in certain epochs
- **Reasonable Test Accuracy:** (~77–80%) by the final epoch, demonstrating that even a modest CNN can differentiate the classes with moderate reliability
- **Overfitting Tendency:** Our training accuracy occasionally outstripped validation, but data augmentation plus dropout helps mitigate it, showing a measure of stability

5. Future Enhancements

- **Hyperparameter Tuning:** Investigate a smaller or variable learning rate, or experiment with an additional conv block for deeper feature extraction
- **Transfer Learning:** As recommended in our roadmap, consider using a pretrained network (e.g., VGG16 or MobileNet). This often boosts accuracy for small or moderate datasets
- **Offline Augmentation:** Potentially expand the dataset with the separate offline augmentation script—generating more samples to reduce overfitting

6. Conclusion and Next Steps

The baseline model proves that a simple two-block CNN, plus minimal on-the-fly augmentation, yields a workable solution. This foundation is well-suited for iterative refinement—either by adjusting hyperparameters or transitioning to a more advanced Transfer Learning pipeline—as guided by our overall project roadmap.

```
In [1]: %pip install tensorflow
        %pip install matplotlib
        %pip install numpy
        %pip install scikit-learn
```

Requirement already satisfied: tensorflow in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (2.18.0)

Requirement already satisfied: absl-py>=1.0.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (2.1.0)

Requirement already satisfied: astunparse>=1.6.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (1.6.3)

Requirement already satisfied: flatbuffers>=24.3.25 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (24.12.23)

Requirement already satisfied: gast!=0.5.0,!0.5.1,!0.5.2,>=0.2.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (0.6.0)

Requirement already satisfied: google-pasta>=0.1.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (0.2.0)

Requirement already satisfied: libclang>=13.0.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (18.1.1)

Requirement already satisfied: opt-einsum>=2.3.2 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (3.4.0)

Requirement already satisfied: packaging in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (24.2)

Requirement already satisfied: protobuf!=4.21.0,!4.21.1,!4.21.2,!4.21.3,!4.21.4,!4.21.5,<6.0.0dev,>=3.20.3 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (5.29.3)

Requirement already satisfied: requests<3,>=2.21.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (2.32.3)

Requirement already satisfied: setuptools in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (75.1.0)

Requirement already satisfied: six>=1.12.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (1.16.0)

Requirement already satisfied: termcolor>=1.1.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (2.5.0)

Requirement already satisfied: typing-extensions>=3.6.6 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (4.12.2)

Requirement already satisfied: wrapt>=1.11.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (1.17.2)

Requirement already satisfied: grpcio<2.0,>=1.24.3 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (1.69.0)

Requirement already satisfied: tensorboard<2.19,>=2.18 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (2.18.0)

Requirement already satisfied: keras>=3.5.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (3.8.0)

Requirement already satisfied: numpy<2.1.0,>=1.26.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (2.0.2)

Requirement already satisfied: h5py>=3.11.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (3.11.0)

forge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (3.12.1)

Requirement already satisfied: ml-dtypes<0.5.0,>=0.4.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorflow) (0.4.1)

Requirement already satisfied: wheel<1.0,>=0.23.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from astunparse>=1.6.0->tensorflow) (0.44.0)

Requirement already satisfied: rich in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from keras>=3.5.0->tensorflow) (13.9.4)

Requirement already satisfied: namex in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from keras>=3.5.0->tensorflow) (0.0.8)

Requirement already satisfied: optree in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from keras>=3.5.0->tensorflow) (0.14.0)

Requirement already satisfied: charset-normalizer<4,>=2 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from requests<3,>=2.21.0->tensorflow) (3.4.1)

Requirement already satisfied: idna<4,>=2.5 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from requests<3,>=2.21.0->tensorflow) (3.10)

Requirement already satisfied: urllib3<3,>=1.21.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from requests<3,>=2.21.0->tensorflow) (2.3.0)

Requirement already satisfied: certifi>=2017.4.17 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from requests<3,>=2.21.0->tensorflow) (2024.12.14)

Requirement already satisfied: markdown>=2.6.8 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorboard<2.19,>=2.18->tensorflow) (3.7)

Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorboard<2.19,>=2.18->tensorflow) (0.7.2)

Requirement already satisfied: werkzeug>=1.0.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from tensorboard<2.19,>=2.18->tensorflow) (3.1.3)

Requirement already satisfied: MarkupSafe>=2.1.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from werkzeug>=1.0.1->tensorboard<2.19,>=2.18->tensorflow) (3.0.2)

Requirement already satisfied: markdown-it-py>=2.2.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from rich->keras>=3.5.0->tensorflow) (3.0.0)

Requirement already satisfied: pygments<3.0.0,>=2.13.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from rich->keras>=3.5.0->tensorflow) (2.19.1)

Requirement already satisfied: mdurl~=0.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from markdown-it-py>=2.2.0->rich->keras>=3.5.0->tensorflow) (0.1.2)

Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: matplotlib in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (3.10.0)

Requirement already satisfied: contourpy>=1.0.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (1.3.1)

Requirement already satisfied: cycler>=0.10 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (0.12.1)

Requirement already satisfied: fonttools>=4.22.0 in /opt/homebrew/Caskroom

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Requirement already satisfied: kiwisolver<=1.3.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (1.4.8)
Requirement already satisfied: numpy<=1.23 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (2.0.2)
Requirement already satisfied: packaging<=20.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (24.2)
Requirement already satisfied: pillow<=8 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (11.1.0)
Requirement already satisfied: pyparsing<=2.3.1 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (3.2.1)
Requirement already satisfied: python-dateutil<=2.7 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from matplotlib) (2.9.0.post0)
Requirement already satisfied: six<=1.5 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from python-dateutil<=2.7->matplotlib) (1.16.0)
Note: you may need to restart the kernel to use updated packages.
Requirement already satisfied: numpy in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (2.0.2)
Note: you may need to restart the kernel to use updated packages.
Requirement already satisfied: scikit-learn in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (1.6.1)
Requirement already satisfied: numpy<=1.19.5 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from scikit-learn) (2.0.2)
Requirement already satisfied: scipy<=1.6.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from scikit-learn) (1.15.1)
Requirement already satisfied: joblib<=1.2.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from scikit-learn) (1.4.2)
Requirement already satisfied: threadpoolctl<=3.1.0 in /opt/homebrew/Caskroom/miniforge/base/envs/mlx-env/lib/python3.12/site-packages (from scikit-learn) (3.5.0)
Note: you may need to restart the kernel to use updated packages.
```

```
In [2]: #!/usr/bin/env python
# baseline_implementation_no_dip.py

import os
import numpy as np
import matplotlib.pyplot as plt
import cv2

import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers

#####
# Data Augmentation Pipeline
#####
myDataAug = keras.Sequential([
    layers.RandomFlip("horizontal"),
```

```

layers.RandomRotation(0.1),
layers.RandomZoom(0.1),
# You can add layers.RandomContrast(0.1) if desired.
], name="MyDataAug")

#####
# Baseline CNN Model: repeated 3x3 conv + maxpool
#####
def build_baseline_cnn(num_classes=3, input_shape=(180,180,3)):
    """
    Basic CNN: repeated 3x3 conv -> maxpool -> flatten -> dense.
    """
    # The input shape is (180,180,3) to match your dataset resizing
    inputs = keras.Input(shape=input_shape, name="input_image")

    # (Optional) Data augmentation first
    x = myDataAug(inputs)

    # 1st conv block
    x = layers.Conv2D(32, kernel_size=3, padding='same', activation='relu')(x)
    x = layers.Conv2D(32, kernel_size=3, padding='same', activation='relu')(x)
    x = layers.MaxPooling2D()(x)

    # 2nd conv block
    x = layers.Conv2D(64, kernel_size=3, padding='same', activation='relu')(x)
    x = layers.Conv2D(64, kernel_size=3, padding='same', activation='relu')(x)
    x = layers.MaxPooling2D()(x)

    # (Optional) 3rd conv block
    # x = layers.Conv2D(128, kernel_size=3, padding='same', activation='relu')(x)
    # x = layers.Conv2D(128, kernel_size=3, padding='same', activation='relu')(x)
    # x = layers.MaxPooling2D()(x)

    # Flatten and dense
    x = layers.Flatten()(x)
    x = layers.Dense(128, activation='relu')(x)
    x = layers.Dropout(0.5)(x)

    outputs = layers.Dense(num_classes, activation='softmax')(x)
    model = keras.Model(inputs, outputs, name="BaselineCNN")
    return model

#####
# Main
#####
def main():
    # 1) Paths
    train_dir = "/Users/ryangichuru/Documents/SSD-K/Uni/2nd year/Intro to AI/CNN/assignment-2-ryantigi254-main/train"
    val_dir = "/Users/ryangichuru/Documents/SSD-K/Uni/2nd year/Intro to AI/CNN/assignment-2-ryantigi254-main/val"
    test_dir = "/Users/ryangichuru/Documents/SSD-K/Uni/2nd year/Intro to AI/CNN/assignment-2-ryantigi254-main/test"

    # 2) Load Datasets
    batch_size = 32
    img_size = (180, 180)

    train_ds = tf.keras.preprocessing.image_dataset_from_directory(
        train_dir,
        image_size=img_size,
        batch_size=batch_size,
        label_mode='categorical'
    )

```

```
)
val_ds = tf.keras.preprocessing.image_dataset_from_directory(
    val_dir,
    image_size=img_size,
    batch_size=batch_size,
    label_mode='categorical'
)
test_ds = tf.keras.preprocessing.image_dataset_from_directory(
    test_dir,
    image_size=img_size,
    batch_size=batch_size,
    label_mode='categorical'
)

# 3) Build model
num_classes = 3 # e.g. 3 people or classes
baseline_model = build_baseline_cnn(num_classes=num_classes,
                                     input_shape=(180,180,3))
baseline_model.summary()

# 4) Compile
baseline_model.compile(optimizer='rmsprop',
                       loss='categorical_crossentropy',
                       metrics=['accuracy'])

# 5) Train
callbacks_list = [
    keras.callbacks.ModelCheckpoint("baseline_cnn_best.h5",
                                    save_best_only=True,
                                    monitor="val_loss")
]
epochs = 10
history = baseline_model.fit(
    train_ds,
    validation_data=val_ds,
    epochs=epochs,
    callbacks=callbacks_list
)

# 6) Evaluate on test set
print("\nEvaluating on test set ...")
test_loss, test_acc = baseline_model.evaluate(test_ds)
print(f"Test loss: {test_loss:.4f}")
print(f"Test accuracy: {test_acc:.4f}")

# 7) Plot training vs. validation
acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs_range = range(1, len(acc)+1)

plt.figure(figsize=(12,5))
plt.subplot(1,2,1)
plt.plot(epochs_range, acc, 'bo-', label='Training Acc')
plt.plot(epochs_range, val_acc, 'ro-', label='Validation Acc')
plt.title('Training & Validation Accuracy')
plt.legend()

plt.subplot(1,2,2)
```

```

plt.plot(epochs_range, loss, 'bo-', label='Training Loss')
plt.plot(epochs_range, val_loss, 'ro-', label='Validation Loss')
plt.title('Training & Validation Loss')
plt.legend()
plt.show()

# 8) Confusion Matrix
print("\nGenerating confusion matrix ...")
all_labels = []
all_preds = []
for images, labels in test_ds:
    preds = baseline_model.predict(images)
    all_preds.extend(tf.argmax(preds, axis=1).numpy())
    all_labels.extend(tf.argmax(labels, axis=1).numpy())

from sklearn.metrics import confusion_matrix, classification_report
cm = confusion_matrix(all_labels, all_preds)
print("Confusion Matrix:\n", cm)
print("Classification Report:\n",
      classification_report(all_labels, all_preds))

if __name__ == "__main__":
    main()

```

Found 457 files belonging to 3 classes.

Found 144 files belonging to 3 classes.

Found 154 files belonging to 3 classes.

Model: "BaselineCNN"

Layer (type)	Output Shape	
input_image (InputLayer)	(None, 180, 180, 3)	
MyDataAug (Sequential)	(None, 180, 180, 3)	
conv2d (Conv2D)	(None, 180, 180, 32)	
conv2d_1 (Conv2D)	(None, 180, 180, 32)	
max_pooling2d (MaxPooling2D)	(None, 90, 90, 32)	
conv2d_2 (Conv2D)	(None, 90, 90, 64)	
conv2d_3 (Conv2D)	(None, 90, 90, 64)	
max_pooling2d_1 (MaxPooling2D)	(None, 45, 45, 64)	
flatten (Flatten)	(None, 129600)	
dense (Dense)	(None, 128)	16,5
dropout (Dropout)	(None, 128)	
dense_1 (Dense)	(None, 3)	

Total params: 16,654,883 (63.53 MB)

Trainable params: 16,654,883 (63.53 MB)

Non-trainable params: 0 (0.00 B)

Epoch 1/10

15/15 ————— 0s 513ms/step - accuracy: 0.3692 - loss: 278.9919

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

15/15 ————— 9s 575ms/step - accuracy: 0.3711 - loss: 269.9249 - val_accuracy: 0.5069 - val_loss: 1.2182

Epoch 2/10

15/15 ————— 0s 673ms/step - accuracy: 0.5656 - loss: 0.9462

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

15/15 ————— 11s 749ms/step - accuracy: 0.5656 - loss: 0.9446 - val_accuracy: 0.7014 - val_loss: 0.7083

Epoch 3/10

15/15 ————— 0s 816ms/step - accuracy: 0.6459 - loss: 0.7838

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

15/15 ————— 14s 895ms/step - accuracy: 0.6465 - loss: 0.7859 - val_accuracy: 0.6806 - val_loss: 0.6920

Epoch 4/10

15/15 ————— 13s 831ms/step - accuracy: 0.6652 - loss: 0.7544 - val_accuracy: 0.6875 - val_loss: 0.7130

Epoch 5/10

15/15 ————— 0s 812ms/step - accuracy: 0.5838 - loss: 0.9427

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

15/15 ————— 13s 882ms/step - accuracy: 0.5858 - loss: 0.9381 - val_accuracy: 0.7778 - val_loss: 0.5346

Epoch 6/10

15/15 ————— 13s 870ms/step - accuracy: 0.5381 - loss: 7.5380 - val_accuracy: 0.7292 - val_loss: 0.6102

Epoch 7/10

15/15 ————— 14s 895ms/step - accuracy: 0.5873 - loss: 2.1275 - val_accuracy: 0.7083 - val_loss: 0.8159

Epoch 8/10

15/15 ————— 0s 763ms/step - accuracy: 0.6870 - loss: 0.7045

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

15/15 ————— 13s 842ms/step - accuracy: 0.6886 - loss: 0.7034 - val_accuracy: 0.8333 - val_loss: 0.4530

Epoch 9/10

15/15 ————— 0s 804ms/step - accuracy: 0.7388 - loss: 0.6577

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

15/15 ————— 14s 884ms/step – accuracy: 0.7381 – loss: 0.656
 7 – val_accuracy: 0.8264 – val_loss: 0.4324

Epoch 10/10

15/15 ————— 0s 777ms/step – accuracy: 0.6912 – loss: 1.0076

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

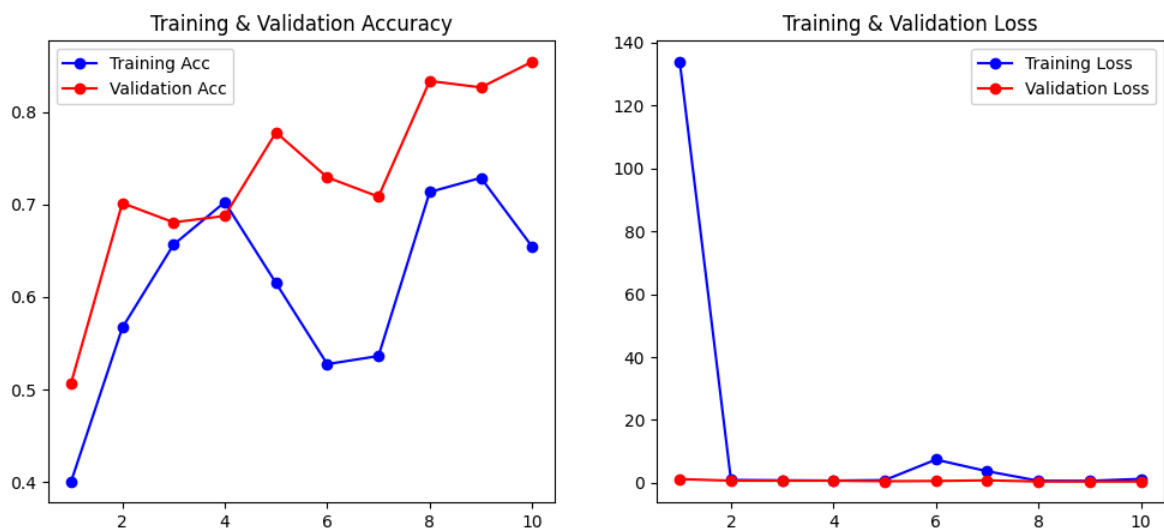
15/15 ————— 13s 850ms/step – accuracy: 0.6889 – loss: 1.024
 8 – val_accuracy: 0.8542 – val_loss: 0.4314

Evaluating on test set ...

5/5 ————— 1s 185ms/step – accuracy: 0.7931 – loss: 0.5251

Test loss: 0.4888

Test accuracy: 0.8312



Generating confusion matrix ...

1/1 ————— 0s 288ms/step

1/1 ————— 0s 164ms/step

1/1 ————— 0s 193ms/step

1/1 ————— 0s 182ms/step

1/1 ————— 0s 194ms/step

Confusion Matrix:

[[26 2 4]

[6 60 2]

[2 10 42]]

Classification Report:

	precision	recall	f1-score	support
0	0.76	0.81	0.79	32
1	0.83	0.88	0.86	68
2	0.88	0.78	0.82	54
accuracy			0.83	154
macro avg	0.82	0.82	0.82	154
weighted avg	0.83	0.83	0.83	154

2025-01-25 13:06:56.865287: I tensorflow/core/framework/local_rendezvous.cc:405] Local rendezvous is aborting with status: OUT_OF_RANGE: End of sequence

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