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|----------------------|---------------------|
| <b>Name</b>          | <b>Krishna T</b>    |
| <b>Roll No</b>       | <b>210419104092</b> |
| <b>Assignment No</b> | <b>3</b>            |

```

1. import numpy as np
2. import tensorflow as tf
3. from tensorflow.keras import layers
4. from tensorflow.keras.models import Sequential
5. import matplotlib.pyplot as plt

6. batch_size = 16

```

## IMAGE AUGMENTATION

```

1. data_augmentation = Sequential(
2. [
3. layers.RandomFlip("horizontal",input_shape=(180, 180, 3)),
4. layers.RandomRotation(0.1),
5. layers.RandomZoom(0.1),
6. ]
7. )

```

### Splitting dataset into training and test

```

1. train_data_set = tf.keras.utils.image_dataset_from_directory(
2. "flowers",
3. validation_split=0.25,
4. subset="training",
5. seed=132,
6. image_size=(180, 180),
7. batch_size=batch_size)

```

Found 4317 files belonging to 5 classes.  
Using 3238 files for training.

```

1. val_data_set = tf.keras.utils.image_dataset_from_directory(
2. "flowers",
3. validation_split=0.25,
4. subset="validation",
5. seed=132,
6. image_size=(180, 180),
7. batch_size=batch_size)

```

Found 4317 files belonging to 5 classes.  
Using 1079 files for validation.

```

1. class_names = train_data_set.class_names

2. plt.figure(figsize=(15, 15))
3. for images, labels in train_data_set.take(1):

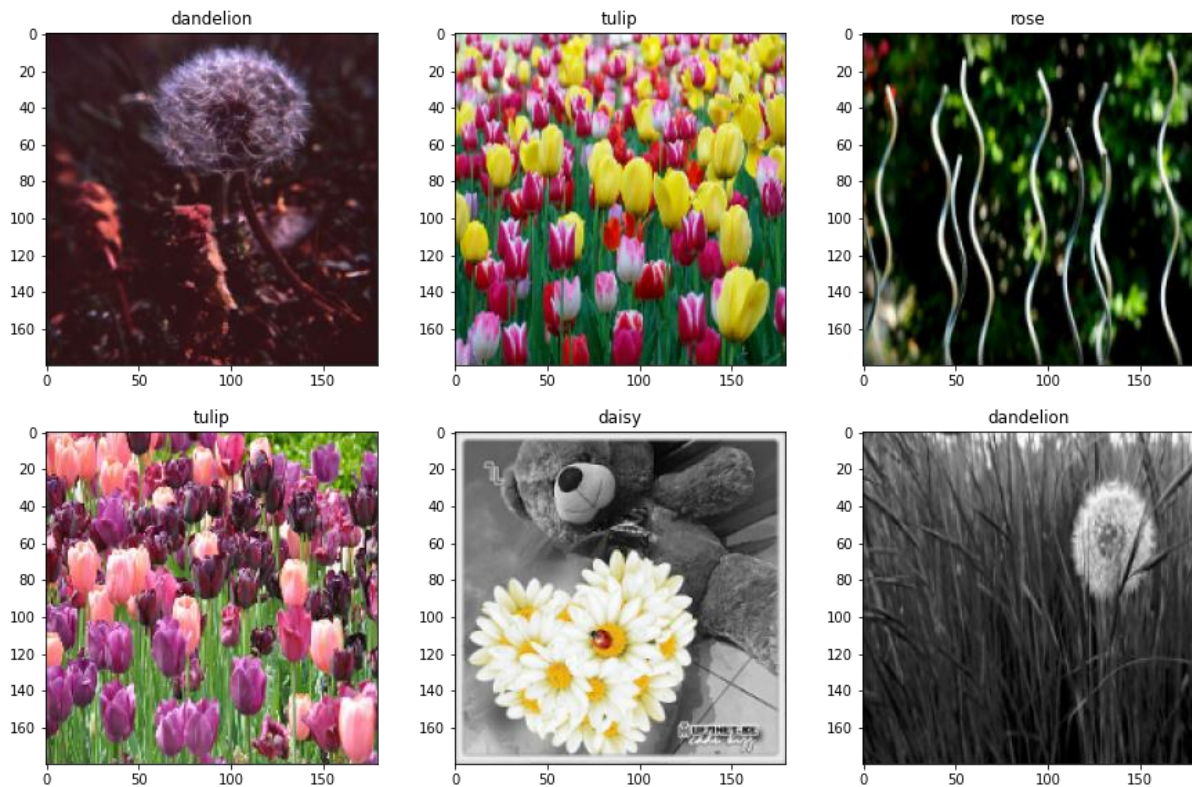
```

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```

4. for i in range(6):
5.     ax = plt.subplot(3, 3, i + 1)
6.     plt.imshow(images[i].numpy().astype("uint8"))
7.     plt.title(class_names[labels[i]])

```



Normalizing pixel value from 0 - 255 to 0 - 1

```

1. normalization_layer = layers.Rescaling(1./255)

2. dataset_normalized = train_data_set.map(lambda x, y:
    (normalization_layer(x), y))
3. image_batch, labels_batch = next(iter(dataset_normalized))
4. first_image = image_batch[0]
5. print(np.min(first_image), np.max(first_image))

```

```
0.0 1.0
```

## MODEL CREATION AND ADDITION OF LAYERS

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```

1. num_classes = len(class_names)

2. model = Sequential([
3. data_augmentation,
4. layers.Rescaling(1./255, input_shape=(180, 180, 3)),
5. # adding convolutional layer
6. layers.Conv2D(16, 3, padding='same', activation='relu'),
7. # adding maxpooling layer
8. layers.MaxPooling2D(),
9. layers.Conv2D(32, 3, padding='same', activation='relu'),
10. layers.MaxPooling2D(),
11. layers.Conv2D(64, 3, padding='same', activation='relu'),
12. layers.MaxPooling2D(),
13. # adding flatten
14. layers.Flatten(),
15. # adding dense hidden layer
16. layers.Dense(128, activation='relu'),
17. # adding dense output layer
18. layers.Dense(num_classes)
19. ])

```

## COMPIATION OF MODEL

```

1. # compiling model with categorical cross entropy and adam optimizer
2. model.compile(optimizer='adam',
3. loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
4. metrics=['accuracy'])

```

## FITTING THE MODEL

```

epochs=15
history =
model.fit(train_data_set, validation_data=val_data_set, epochs=epochs)

Epoch 1/15
203/203 [=====] - 113s 554ms/step - loss: 1.2347 -
accuracy: 0.4700 - val_loss: 1.0847 - val_accuracy: 0.5672
Epoch 2/15
203/203 [=====] - 109s 537ms/step - loss: 0.9885 -
accuracy: 0.6124 - val_loss: 1.0040 - val_accuracy: 0.5996
Epoch 3/15
203/203 [=====] - 88s 432ms/step - loss: 0.8784 -
accuracy: 0.6563 - val_loss: 1.0471 - val_accuracy: 0.6154
Epoch 4/15
203/203 [=====] - 97s 479ms/step - loss: 0.8347 -
accuracy: 0.6776 - val_loss: 0.8886 - val_accuracy: 0.6793

```

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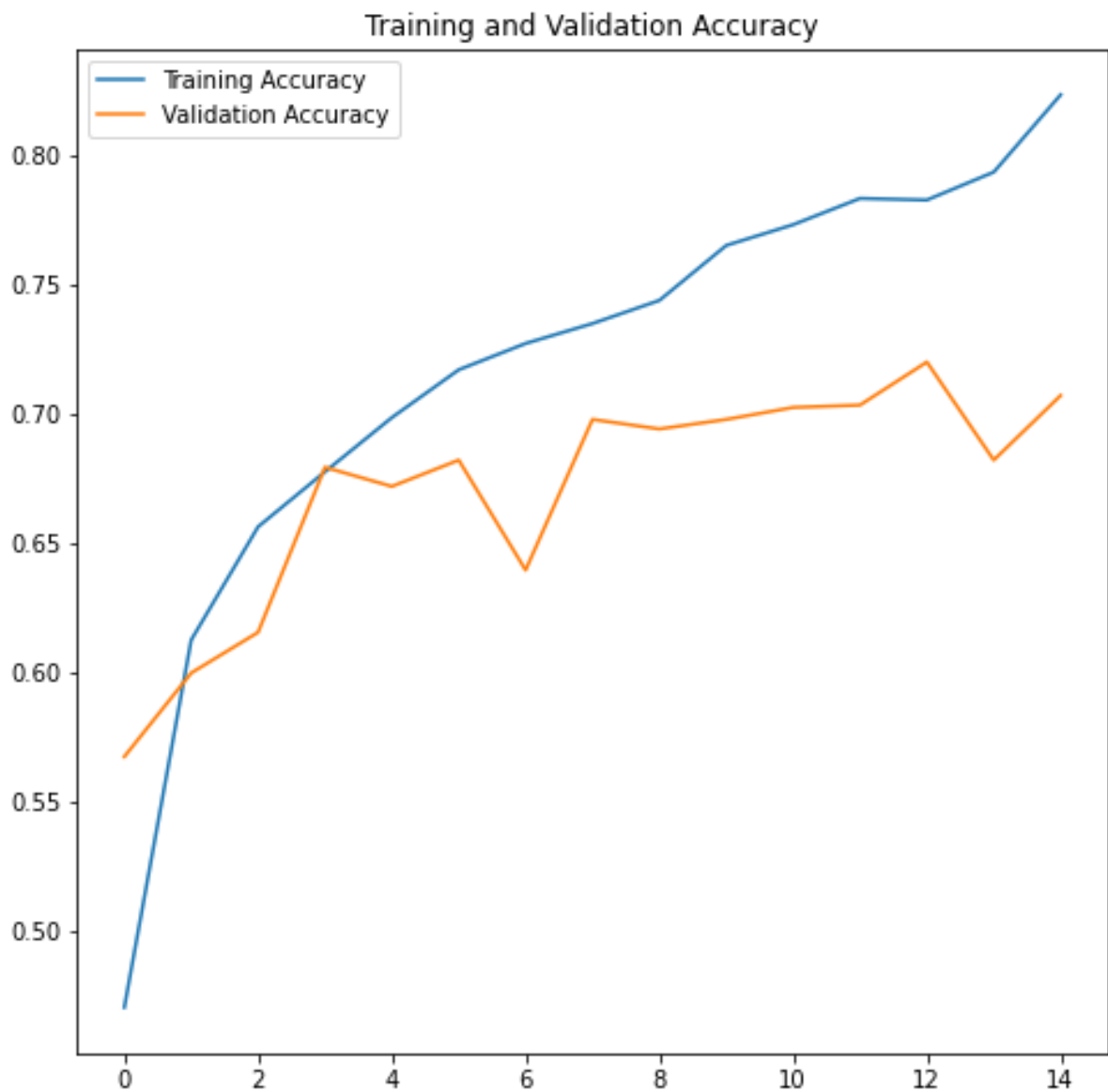
```

Epoch 5/15
203/203 [=====] - 98s 482ms/step - loss: 0.7949 -
accuracy: 0.6986 - val_loss: 0.9212 - val_accuracy: 0.6719
Epoch 6/15
203/203 [=====] - 92s 454ms/step - loss: 0.7518 -
accuracy: 0.7171 - val_loss: 0.8591 - val_accuracy: 0.6821
Epoch 7/15
203/203 [=====] - 82s 402ms/step - loss: 0.7336 -
accuracy: 0.7273 - val_loss: 1.0130 - val_accuracy: 0.6395
Epoch 8/15
203/203 [=====] - 95s 465ms/step - loss: 0.7143 -
accuracy: 0.7350 - val_loss: 0.8428 - val_accuracy: 0.6979
Epoch 9/15
203/203 [=====] - 97s 478ms/step - loss: 0.6683 -
accuracy: 0.7440 - val_loss: 0.8703 - val_accuracy: 0.6942
Epoch 10/15
203/203 [=====] - 91s 448ms/step - loss: 0.6359 -
accuracy: 0.7653 - val_loss: 0.8908 - val_accuracy: 0.6979
Epoch 11/15
203/203 [=====] - 90s 444ms/step - loss: 0.6027 -
accuracy: 0.7733 - val_loss: 0.8223 - val_accuracy: 0.7025
Epoch 12/15
203/203 [=====] - 78s 383ms/step - loss: 0.5753 -
accuracy: 0.7835 - val_loss: 0.9151 - val_accuracy: 0.7034
Epoch 13/15
203/203 [=====] - 76s 373ms/step - loss: 0.5684 -
accuracy: 0.7829 - val_loss: 0.8659 - val_accuracy: 0.7201
Epoch 14/15
203/203 [=====] - 76s 373ms/step - loss: 0.5285 -
accuracy: 0.7937 - val_loss: 1.0205 - val_accuracy: 0.6821
Epoch 15/15
203/203 [=====] - 75s 371ms/step - loss: 0.4781 -
accuracy: 0.8237 - val_loss: 0.9877 - val_accuracy: 0.7071

```

1. `epochs_range = range(epochs)`
2. `plt.figure(figsize=(8, 8))`
3. `plt.plot(epochs_range, history.history['accuracy'], label='Training Accuracy')`
4. `plt.plot(epochs_range, history.history['val_accuracy'], label='Validation Accuracy')`
5. `plt.legend()`
6. `plt.title('Training and Validation Accuracy')`
7. `plt.show()`

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In [35]:

```

1. plt.figure(figsize=(8, 8))
2. plt.plot(epochs_range, history.history['loss'], label='Training
   Loss')
3. plt.plot(epochs_range, history.history['val_loss'], label='Validation
   Loss')
4. plt.legend()
5. plt.title('Training and Validation Loss')
6. plt.show()

```

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## SAVING THE MODEL

1. `model.save("CNN Model for Classification Of Flowers.h5")`
2. `model.load_weights('CNN Model for Classification Of Flowers.h5')`

## TESTING THE MODEL

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```

1. sunflower_url =
   "https://storage.googleapis.com/download.tensorflow.org/example_images/592px-Red_sunflower.jpg"
2. sunflower_path = tf.keras.utils.get_file('Red_sunflower',
   origin=sunflower_url)

3. img = tf.keras.utils.load_img(
4. sunflower_path, target_size=(180, 180)
5. )
6. img_array = tf.keras.utils.img_to_array(img)
7. img_array = tf.expand_dims(img_array, 0) # Create a batch

8. predictions = model.predict(img_array)
9. score = tf.nn.softmax(predictions[0])

10. print(class_names[np.argmax(score)], 100 * np.max(score))

```

sunflower 98.68776202201843