

```

import tensorflow as tf
from tensorflow import keras
import matplotlib.pyplot as plt
import numpy as np
from os import listdir
from os.path import join
import pandas
import cv2
import os
import random

```

In [6]:

```

data_lead = 'D:/IBM Project/Flowers-Dataset/flowers'
folders_lead = os.listdir(data_lead)

```

```

print(folders_lead)

['daisy', 'dandelion', 'rose', 'sunflower', 'tulip']

```

In [10]:

```

image_names = []
train_labels = []
train_images = []

size = 64, 64

for folder in folders_lead:
    for file in os.listdir(os.path.join(data_lead, folder)):
        if file.endswith(".jpg"):
            image_names.append(os.path.join(data_lead, folder, file))
            train_labels.append(folder)
            img = cv2.imread(os.path.join(data_lead, folder, file))
            im = cv2.resize(img, size)
            train_images.append(im)
        else:
            continue

```

In [11]:

```

train = np.array(train_images)

```

```

train.shape

```

Out[11]:

```

(4317, 64, 64, 3)

```

In [12]:

```

train = train.astype('float32') / 255.0

```

In [13]:

```

label_dummies = pandas.get_dummies(train_labels)

```

```

labels = label_dummies.values.argmax(1)

```

In [14]:

```

pandas.unique(train_labels)

```

Out[14]:

```
array(['daisy', 'dandelion', 'rose', 'sunflower', 'tulip'], dtype=object)
```

In [15]:

```
union_list = list(zip(train, labels))
random.shuffle(union_list)
train, labels = zip(*union_list)
```

```
# Convert the shuffled list to numpy array type
```

```
train = np.array(train)
labels = np.array(labels)
```

In [17]:

```
model = keras.Sequential([
    keras.layers.Flatten(input_shape=(64, 64, 3)),
    keras.layers.Dense(256, activation=tf.nn.relu),
    keras.layers.Dense(128, activation=tf.nn.relu),
    keras.layers.Dense(6, activation=tf.nn.softmax)
])
```

In [18]:

```
model.summary()
```

```
Model: "sequential_1"
```

Layer (type)	Output Shape	Param #
flatten_1 (Flatten)	(None, 12288)	0
dense_3 (Dense)	(None, 256)	3145984
dense_4 (Dense)	(None, 128)	32896
dense_5 (Dense)	(None, 6)	774

Total params: 3,179,654
Trainable params: 3,179,654
Non-trainable params: 0

In [19]:

```
model.compile(optimizer= tf.optimizers.Adam(),
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
```

In [21]:

```
model.fit(train, labels, epochs=15)
```

```
Epoch 1/15
```

```
135/135 [=====] - 4s 33ms/step - loss: 0.9817 - accuracy: 0.6046
```

```
Epoch 2/15
```

```
135/135 [=====] - 4s 33ms/step - loss: 0.9438 - accuracy: 0.6278
```

```
Epoch 3/15
```

```
135/135 [=====] - 4s 32ms/step - loss: 0.9001 -  
accuracy: 0.6416  
Epoch 4/15  
135/135 [=====] - 4s 32ms/step - loss: 0.8361 -  
accuracy: 0.6854  
Epoch 5/15  
135/135 [=====] - 4s 33ms/step - loss: 0.8594 -  
accuracy: 0.6674  
Epoch 6/15  
135/135 [=====] - 4s 32ms/step - loss: 0.8078 -  
accuracy: 0.6787  
Epoch 7/15  
135/135 [=====] - 4s 32ms/step - loss: 0.7158 -  
accuracy: 0.7239  
Epoch 8/15  
135/135 [=====] - 4s 32ms/step - loss: 0.7496 -  
accuracy: 0.7130  
Epoch 9/15  
135/135 [=====] - 4s 33ms/step - loss: 0.7025 -  
accuracy: 0.7308  
Epoch 10/15  
135/135 [=====] - 4s 33ms/step - loss: 0.6381 -  
accuracy: 0.7640  
Epoch 11/15  
135/135 [=====] - 4s 33ms/step - loss: 0.5484 -  
accuracy: 0.8024  
Epoch 12/15  
135/135 [=====] - 4s 33ms/step - loss: 0.5464 -  
accuracy: 0.8024  
Epoch 13/15  
135/135 [=====] - 4s 33ms/step - loss: 0.5362 -  
accuracy: 0.8110  
Epoch 14/15  
135/135 [=====] - 4s 33ms/step - loss: 0.5055 -  
accuracy: 0.8114  
Epoch 15/15  
135/135 [=====] - 4s 32ms/step - loss: 0.4804 -  
accuracy: 0.8279
```

```
<keras.callbacks.History at 0x1ab1493c670>
```

```
model.save("D:/IBM Project/Flowers-Dataset/flowers.h5")
```

Out[21]:

In [22]:

In []: