In [1]:
```
import matplotlib.pyplot as plt
import numpy as np
import os
import PIL          #Python Image Library-provides support for opening, manipulating, and saving many different image file format
import tensorflow as tf
from tensorflow import keras
from tensorflow. keras import layers
from tensorflow.python.keras.layers import Dense, Flatten
from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import Adam
```

In [2]:
```
from tensorflow.keras import layers
```

In [3]:
```
flowers_url = "https://storage.googleapis.com/download.tensorflow.org/example_images/flower_photos.tgz"
```

The code will download the dataset in a compressed format. You then have to extract the dataset and save it into a data directory in Google Colab:

In [4]:
```
import pathlib
flowers_data = tf.keras.utils.get_file('flower_photos', origin=flowers_url, untar=True)
flowers_data = pathlib.Path(flowers_data)
```

```
Downloading data from https://storage.googleapis.com/download.tensorflow.org/example_images/flower_photos.tgz
228813984/228813984 [==============================] - 62s 0us/step
```

In [5]:
```
print(flowers_data)
```

```
C:\Users\lenovo\.keras\datasets\flower_photos
```

In [6]:
```
all_sunflowers = list(flowers_data.glob('sunflowers/*'))
```

In [7]:
```
import PIL
print(all_sunflowers[1])
PIL.Image.open(str(all_sunflowers[1]))
```

```
C:\Users\lenovo\.keras\datasets\flower_photos\sunflowers\1022552002_2b93faf9e7_n.jpg
```

Out[7]:

In [8]:
```
all_roses = list(flowers_data.glob('roses/*'))
```

In [9]:
```
import PIL
print(all_roses[1])
PIL.Image.open(str(all_roses[8]))
```

```
C:\Users\lenovo\.keras\datasets\flower_photos\roses\102501987_3cdb8e5394_n.jpg
```

Out[9]:

In [10]:
```
height,width=180,180
```

In [11]:
```
training_batch_size=32
```

In [12]:
```
train_set = tf.keras.preprocessing.image_dataset_from_directory(
flowers_data,
validation_split=0.2,
subset="training",
seed=123,
image_size=(height,width),
batch_size=training_batch_size)
```

```
Found 3670 files belonging to 5 classes.
Using 2936 files for training.
```

In [13]:
```
image_cat = train_set.class_names
print(image_cat)
```

```
['daisy', 'dandelion', 'roses', 'sunflowers', 'tulips']
```

In [14]:
```
validation_set = tf.keras.preprocessing.image_dataset_from_directory(
flowers_data,
validation_split=0.2,
subset="validation",
seed=123,
image_size=(height, width),
batch_size=training_batch_size)
```

```
Found 3670 files belonging to 5 classes.
Using 734 files for validation.
```

In [15]:
```
dnn_model = Sequential()
```

In [16]:
```
imported_model= tf.keras.applications.ResNet50(include_top=False,
  input_shape=(180,180,3),
  pooling='avg',classes=5,
  weights='imagenet')


for layer in imported_model.layers:
  layer.trainable=False
```

```
Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50_weights_tf_dim_ordering_tf_kernels_notop.h5
94765736/94765736 [==============================] - 23s 0us/step
```

In [17]:
```
dnn_model.add(imported_model)
dnn_model.add(Flatten())
dnn_model.add(Dense(512, activation='relu'))
dnn_model.add(Dense(5, activation='softmax'))
```

In [18]:
```
dnn_model.summary()
```

```
Model: "sequential"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 resnet50 (Functional)       (None, 2048)              23587712

 module_wrapper (ModuleWrap  (None, 2048)              0
 per)

 module_wrapper_1 (ModuleWr  (None, 512)               1049088
 apper)

 module_wrapper_2 (ModuleWr  (None, 5)                 2565
 apper)

=================================================================
```

In [24]:
```
from tensorflow.keras.optimizers import Adam
dnn_model.compile(optimizer=Adam(learning_rate=0.001),loss='sparse_categorical_crossentropy',metrics=['accuracy'])
```

In [23]:
```
history = dnn_model.fit(
train_set,
validation_data=validation_set,
epochs=10
)
```

```
Epoch 1/10
92/92 [==============================] - 184s 2s/step - loss: 2.5682e-04 - accuracy: 1.0000 - val_loss: 0.5139 - val_accuracy:
0.8869
Epoch 2/10
92/92 [==============================] - 224s 2s/step - loss: 2.3840e-04 - accuracy: 1.0000 - val_loss: 0.5156 - val_accuracy:
0.8842
Epoch 3/10
92/92 [==============================] - 242s 3s/step - loss: 2.2109e-04 - accuracy: 1.0000 - val_loss: 0.5163 - val_accuracy:
0.8869
Epoch 4/10
92/92 [==============================] - 229s 2s/step - loss: 2.0410e-04 - accuracy: 1.0000 - val_loss: 0.5210 - val_accuracy:
0.8856
Epoch 5/10
92/92 [==============================] - 202s 2s/step - loss: 1.8787e-04 - accuracy: 1.0000 - val_loss: 0.5237 - val_accuracy:
0.8842
Epoch 6/10
92/92 [==============================] - 193s 2s/step - loss: 1.7597e-04 - accuracy: 1.0000 - val_loss: 0.5270 - val_accuracy:
0.8856
Epoch 7/10
92/92 [==============================] - 195s 2s/step - loss: 1.6396e-04 - accuracy: 1.0000 - val_loss: 0.5273 - val_accuracy:
0.8828
Epoch 8/10
92/92 [==============================] - 194s 2s/step - loss: 1.5179e-04 - accuracy: 1.0000 - val_loss: 0.5323 - val_accuracy:
0.8842
Epoch 9/10
92/92 [==============================] - 193s 2s/step - loss: 1.4124e-04 - accuracy: 1.0000 - val_loss: 0.5336 - val_accuracy:
0.8856
Epoch 10/10
```