

# Assignment 3

## 1. Download And unzip dataset

# run this to download the dataset directly to the kernal

```
!gdown 1xkynpL15pt6KT3YS1Dimu4A5iRU9qYck
```

Downloading...

From: <https://drive.google.com/uc?id=1xkynpL15pt6KT3YS1Dimu4A5iRU9qYck>

To: /content/Flowers-Dataset.zip

100% 236M/236M [00:00<00:00, 286MB/s]

# Unzip

```
!unzip '/content/Flowers-Dataset.zip'
```

```

      g      / y/      _      jpg
C inflating: flowers/daisy/13826249325_f61cb15f86_n.jpg      inflating:
flowers/daisy/13901930939_a7733c03f0_n.jpg      inflating:
flowers/daisy/1392131677_116ec04751.jpg      inflating:
flowers/daisy/1392946544_115acbb2d9.jpg      inflating:
flowers/daisy/13953307149_f8de6a768c_m.jpg      inflating:
flowers/daisy/1396526833_fb867165be_n.jpg      inflating:
flowers/daisy/13977181862_f8237b6b52.jpg      inflating:
flowers/daisy/14021430525_e06baf93a9.jpg      inflating:
flowers/daisy/14073784469_ffb12f3387_n.jpg      inflating:
flowers/daisy/14087947408_9779257411_n.jpg      inflating:
flowers/daisy/14088053307_1a13a0bf91_n.jpg      inflating:
flowers/daisy/14114116486_0bb6649bc1_m.jpg      inflating:
flowers/daisy/14147016029_8d3cf2414e.jpg      inflating:
flowers/daisy/14163875973_467224aaf5_m.jpg      inflating:
flowers/daisy/14167534527_781ceb1b7a_n.jpg      inflating:
flowers/daisy/14167543177_cd36b54ac6_n.jpg      inflating:
flowers/daisy/14219214466_3ca6104eae_m.jpg      inflating:
flowers/daisy/14221836990_90374e6b34.jpg      inflating:
flowers/daisy/14221848160_7f0a37c395.jpg      inflating:
flowers/daisy/14245834619_153624f836.jpg      inflating:
flowers/daisy/14264136211_9531fbc144.jpg      inflating:
flowers/daisy/14272874304_47c0a46f5a.jpg      inflating:
flowers/daisy/14307766919_fac3c37a6b_m.jpg      inflating:
flowers/daisy/14330343061_99478302d4_m.jpg      inflating:
flowers/daisy/14332947164_9b13513c71_m.jpg      inflating:
flowers/daisy/14333681205_a07c9f1752_m.jpg      inflating:
flowers/daisy/14350958832_29bdd3a254.jpg      inflating:
flowers/daisy/14354051035_1037b30421_n.jpg      inflating:
flowers/daisy/14372713423_61e2daae88.jpg      inflating:
flowers/daisy/14399435971_ea5868c792.jpg      inflating:
flowers/daisy/14402451388_56545a374a_n.jpg      inflating:
flowers/daisy/144076848_57e1d662e3_m.jpg      inflating:
flowers/daisy/144099102_bf63a41e4f_n.jpg      inflating:
flowers/daisy/1441939151_b271408c8d_n.jpg      inflating:
flowers/daisy/14421389519_d5fd353eb4.jpg

```

```

a g      o e s/da sy/      3895 9_d5 d353eb jpg
inflating: flowers/daisy/144603918_b9de002f60_m.jpg
inflating: flowers/daisy/14471433500_cdaa22e3ea_m.jpg
inflating: flowers/daisy/14485782498_fb342ec301.jpg      inflating:
flowers/daisy/14507818175_05219b051c_m.jpg      inflating:
flowers/daisy/14523675369_97c31d0b5b.jpg      inflating:
flowers/daisy/14551098743_2842e7a004_n.jpg      inflating:
flowers/daisy/14554906452_35f066ffe9_n.jpg      inflating:
flowers/daisy/14564545365_1f1d267bf1_n.jpg      inflating:
flowers/daisy/14569895116_32f0dcb0f9.jpg      inflating:
flowers/daisy/14591326135_930703dbed_m.jpg      inflating:
flowers/daisy/14600779226_7bbc288d40_m.jpg      inflating:
flowers/daisy/14613443462_d4ed356201.jpg      inflating:
flowers/daisy/14621687774_ec52811acd_n.jpg      inflating:
flowers/daisy/14674743211_f68b13f6d9.jpg      inflating:
flowers/daisy/14698531521_0c2f0c6539.jpg      inflating:
flowers/daisy/147068564_32bb4350cc.jpg      inflating:
flowers/daisy/14707111433_cce08ee007.jpg      inflating:
flowers/daisy/14716799982_ed6d626a66.jpg      inflating:
flowers/daisy/14816364517_2423021484_m.jpg      inflating:
flowers/daisy/14866200659_6462c723cb_m.jpg      inflating:
flowers/daisy/14907815010_bff495449f.jpg      inflating:
flowers/daisy/14921511479_7b0a647795.jpg      inflating:
flowers/daisy/15029936576_8d6f96c72c_n.jpg

```



## Importing Necessary Libs

```

from tensorflow.keras.preprocessing.image import ImageDataGenerator from
tensorflow.keras.models import Sequential from tensorflow.keras.layers import
Convolution2D, MaxPooling2D, Flatten, Dense from tensorflow.keras.preprocessing
import image import numpy as np
import matplotlib.pyplot as plt

```



## 2. Data Augmentation

```

# For training

train_datagen = ImageDataGenerator(rescale=1./255,
horizontal_flip=True,
zoom_range=0.2)

# for testing test_datagen =

ImageDataGenerator(rescale=1./255)

# To split the dataset into Train and test

!pip install split_folders import
splitfolders input_folder =
"/content/flowers" output =

```

```
"/content/Dataset"
splitfolders.ratio(input_folder,
output=output, seed=42,
ratio=(0.7,0.3))
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-w
Requirement already satisfied: split_folders in /usr/local/lib/python3.7/dist
Copying files: 4317 files [00:01, 3937.44 files/s]
```

```
# data generation
```

```
xtrain = train_datagen.flow_from_directory('/content/Dataset/train',
target_size=(64,64),
class_mode='categorical',
batch_size=100)
```

```
xtest = test_datagen.flow_from_directory('/content/Dataset/val',
target_size=(64,64),
class_mode='categorical',
batch_size=100)
```

```
Found 3019 images belonging to 5 classes.
Found 1298 images belonging to 5 classes.
```

### 3. Build Model



#### Adding layers



```
# Build a CNN block
```

```
model = Sequential() # Initializing sequential model
model.add(Convolution2D(32, (3,3), activation='relu', input_shape=(64,64,3))) # convol
model.add(MaxPooling2D(pool_size=(2, 2))) # Max pooling layer
model.add(Flatten()) # Flatten layer
model.add(Dense(300, activation='relu')) # Hidden layer 1
model.add(Dense(150, activation='relu')) # Hidden layer 2
model.add(Dense(5, activation='softmax')) # Output layer
```

#### Compiling Model

```
# Compiling the model
```



```
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
```

#### Fit Model

```
# Train model
```

```
model.fit(xtrain,
          steps_per_epoch=len(xtrain),
          epochs=50,
          validation_data=xtest,
          validation_steps=len(xtest))
```

```
Epoch 1/50
31/31 [=====] - 23s 394ms/step - loss: 2.2765 - accu
Epoch 2/50
31/31 [=====] - 12s 393ms/step - loss: 1.2581 - accu
Epoch 3/50
31/31 [=====] - 16s 508ms/step - loss: 1.1536 - accu
Epoch 4/50
31/31 [=====] - 12s 393ms/step - loss: 1.0853 - accu
Epoch 5/50
31/31 [=====] - 13s 414ms/step - loss: 1.0715 - accu
Epoch 6/50
31/31 [=====] - 12s 389ms/step - loss: 1.0135 - accu
Epoch 7/50
31/31 [=====] - 13s 415ms/step - loss: 0.9690 - accu
Epoch 8/50
31/31 [=====] - 14s 456ms/step - loss: 0.9308 - accu
Epoch 9/50
31/31 [=====] - 14s 446ms/step - loss: 0.8920 - accu
Epoch 10/50
31/31 [=====] - 14s 451ms/step - loss: 0.8516 - accu
Epoch 11/50
31/31 [=====] - 13s 432ms/step - loss: 0.8382 - accu
Epoch 12/50
31/31 [=====] - 13s 418ms/step - loss: 0.8205 - accu
Epoch 13/50
31/31 [=====] - 13s 431ms/step - loss: 0.7762 - accu
Epoch 14/50
31/31 [=====] - 12s 392ms/step - loss: 0.7396 - accu
Epoch 15/50
31/31 [=====] - 12s 392ms/step - loss: 0.7333 - accu
Epoch 16/50
31/31 [=====] - 12s 412ms/step - loss: 0.7193 - accu
Epoch 17/50
31/31 [=====] - 14s 447ms/step - loss: 0.6993 - accu
Epoch 18/50
31/31 [=====] - 13s 424ms/step - loss: 0.6708 - accu
Epoch 19/50
31/31 [=====] - 13s 407ms/step - loss: 0.6495 - accu
Epoch 20/50
31/31 [=====] - 13s 405ms/step - loss: 0.6217 - accu
Epoch 21/50
31/31 [=====] - 13s 411ms/step - loss: 0.5822 - accu
Epoch 22/50
31/31 [=====] - 12s 391ms/step - loss: 0.6147 - accu
Epoch 23/50
31/31 [=====] - 12s 391ms/step - loss: 0.5687 - accu
Epoch 24/50
31/31 [=====] - 12s 389ms/step - loss: 0.5583 - accu
Epoch 25/50
31/31 [=====] - 12s 407ms/step - loss: 0.5564 - accu
Epoch 26/50
31/31 [=====] - 12s 391ms/step - loss: 0.4828 - accu
Epoch 27/50
```

```

31/31 [=====] - 12s 389ms/step - loss: 0.4835 - accu
Epoch 28/50
31/31 [=====] - 13s 408ms/step - loss: 0.4777 - accu
Epoch 29/50
31/31 [=====] - 12s 401ms/step - loss: 0.4616 - accu

```

## 4. Save Model

```
model.save('Flowers.h5')
```

5.

## Testing The Model

```

def predict_flower(img_path):
    img = image.load_img(img_path, target_size=(64,64)) # Reading image    x =
    image.img_to_array(img) # Converting image into array    x =
    np.expand_dims(x,axis=0) # expanding Dimensions    pred =
    np.argmax(model.predict(x)) # Predicting the higher probablity index    op =
    ['Daisy','Dandelion','Rose','SunFlower','Tulip'] # Creating list
    print(op[pred]) # List indexing with output    plt.imshow(img) # Printing
    the image

```

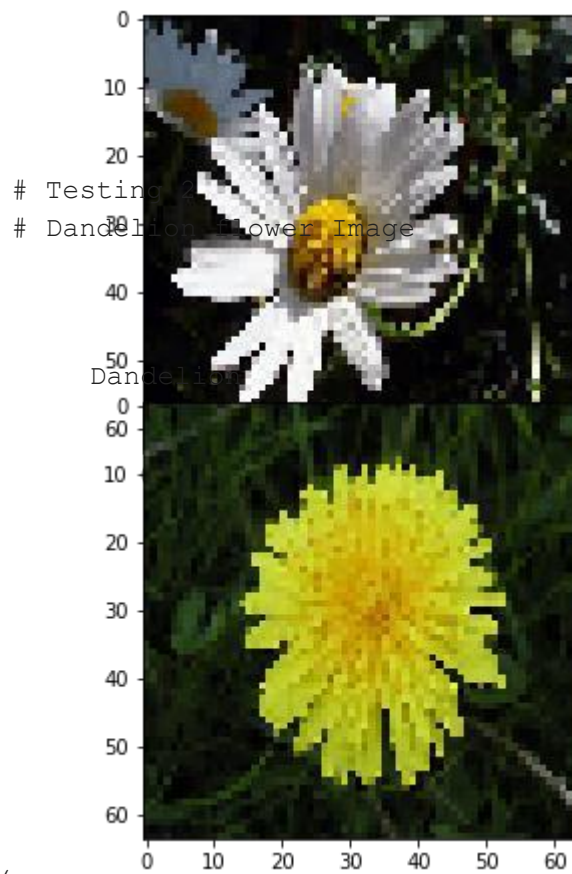
## With Test Data Images

```

# Testing 1 # Daisy flower Image
predict_flower('/content/Dataset/val/daisy/1150395827_6f94a5c6e4_n.jpg') #
Predicti

Daisy

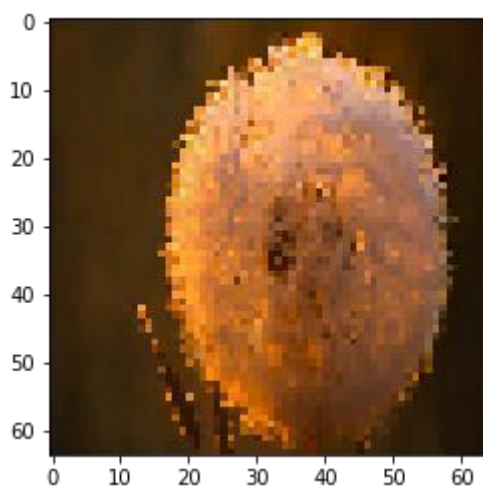
```



```
predict_flower(
  '/content/Dataset/val/dandelion/1128626197_3f52424215_n.jpg')
```

```
# Testing 3
# Dandelion flower Image
```

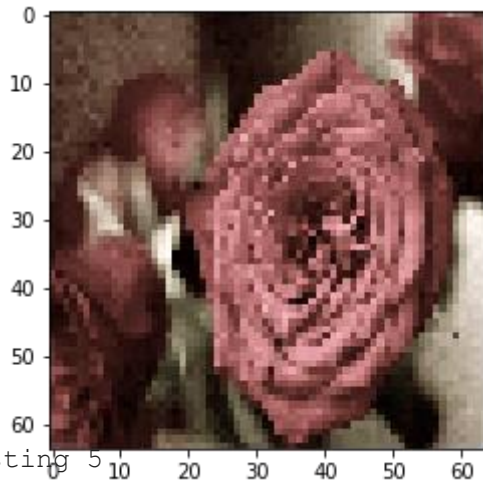
```
predict_flower('/content/Dataset/val/dandelion/14199664556_188b37e51e.jpg')  Rose
```



```
# Testing 4 # Rose flower Image
```

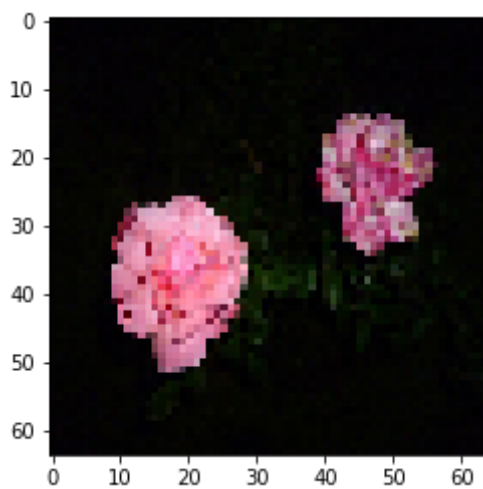
```
predict_flower('/content/Dataset/val/rose/12202373204_34fb07205b.jpg')
```

```
Rose
```



```
# Testing 5  
# Rose flower Image
```

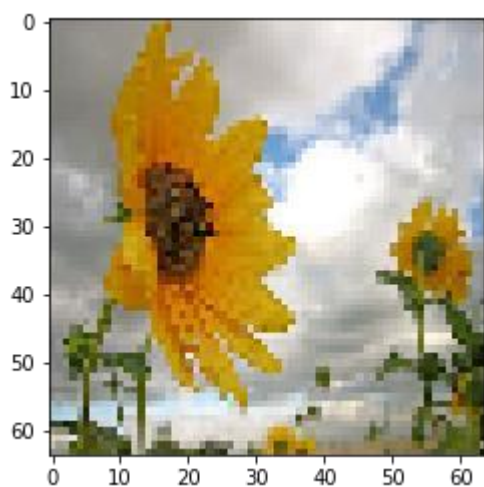
```
predict_flower('/content/Dataset/val/rose/15820572326_be2ea4a55c_n.jpg')  Rose
```



```
# Testing 6 # Sunflower Image
```

```
predict_flower('/content/Dataset/val/sunflower/1596293240_2d5b53495a_m.jpg')
```

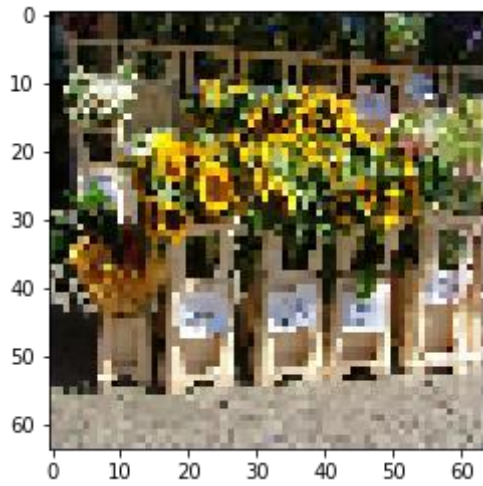
```
SunFlower
```



```
# Testing 7 # Sunflower Image
```

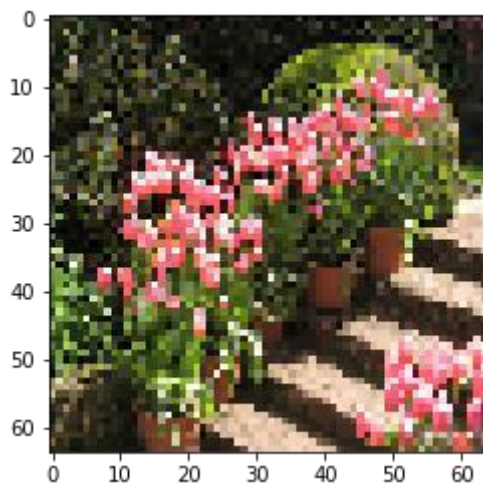
```
predict_flower('/content/Dataset/val/sunflower/210076535_80951bc5d5.jpg')
```

```
SunFlower
```



```
# Testing 8 # Tulip Flower Image
```

```
predict_flower('/content/Dataset/val/tulip/13530690445_9f1f5cf43a_n.jpg') Rose
```

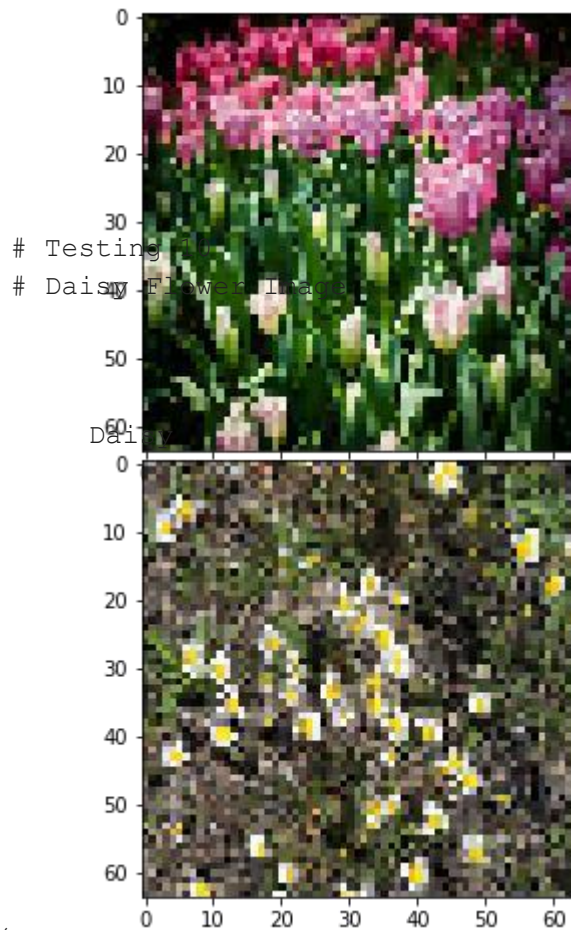


```
# Testing 9 # Tulip Flower Image
```

```
predict_flower('/content/Dataset/val/tulip/16680927427_07ca6e4552_n.jpg')
```

```
Tulip
```





```
predict_flower(
  '/content/Dataset/val/daisy/34542837641_10492bf600_n.jpg')
```

## With Google Images

```
# Run To download test images
```

```
!gdown 1Q-QTRIfXjV0BbLcIvopbiYfbAD3hJfmw
```

```
Downloading...
```

```
From: https://drive.google.com/uc?id=1Q-QTRIfXjV0BbLcIvopbiYfbAD3hJfmw
```

```
To: /content/IBM Flower_Test dataset.zip 100%
```

```
1.01M/1.01M [00:00<00:00, 163MB/s]
```

```
# unzip
```

```
!unzip '/content/IBM Flower_Test dataset.zip'
```

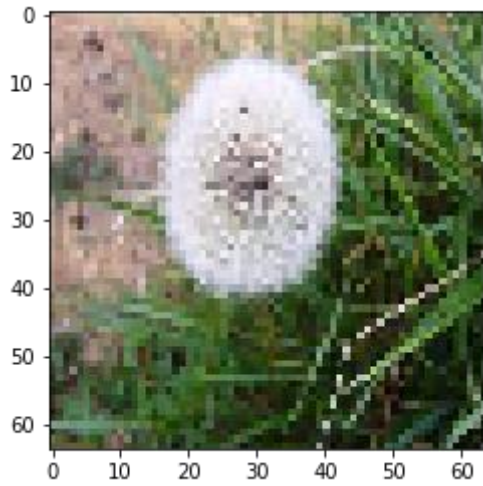
```
Archive: /content/IBM Flower_Test dataset.zip
```

```
replace IBM Flower_Test dataset/tulip_2.jpg? [y]es, [n]o, [A]ll, [N]one, [r]e
```

```
# Test 1
```

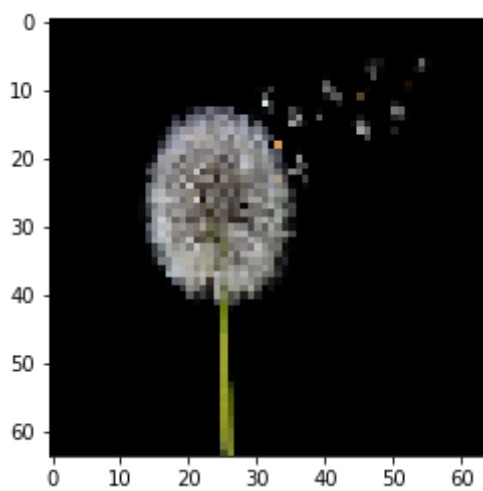
```
# Dandelion Flower predict_flower('/content/IBM Flower_Test
```

```
dataset/Dandelion.jpeg') Tulip
```



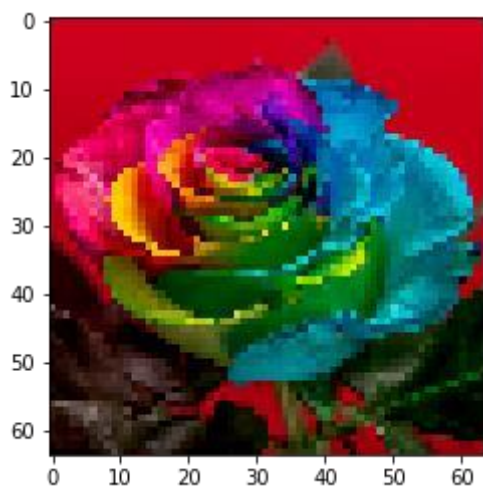
```
# Test 2 # Dandelion Flower predict_flower('/content/IBM  
Flower_Test dataset/Dandelion_2.jpeg')
```

Daisy



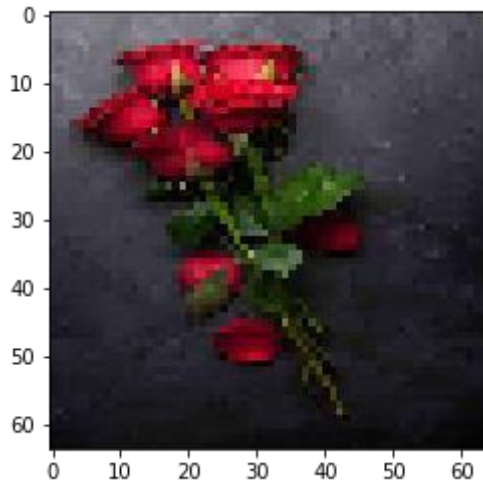
```
# Test 3 # Rose Flower predict_flower('/content/IBM  
Flower_Test dataset/Rose.jpeg')
```

Tulip

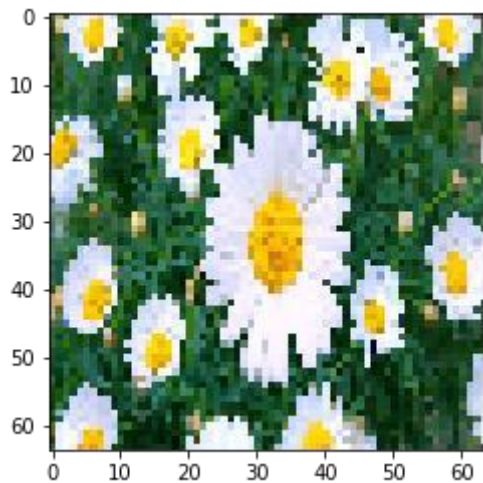


```
# Test 4 # Rose Flower predict_flower('/content/IBM  
Flower_Test dataset/Rose_2.jpeg')
```

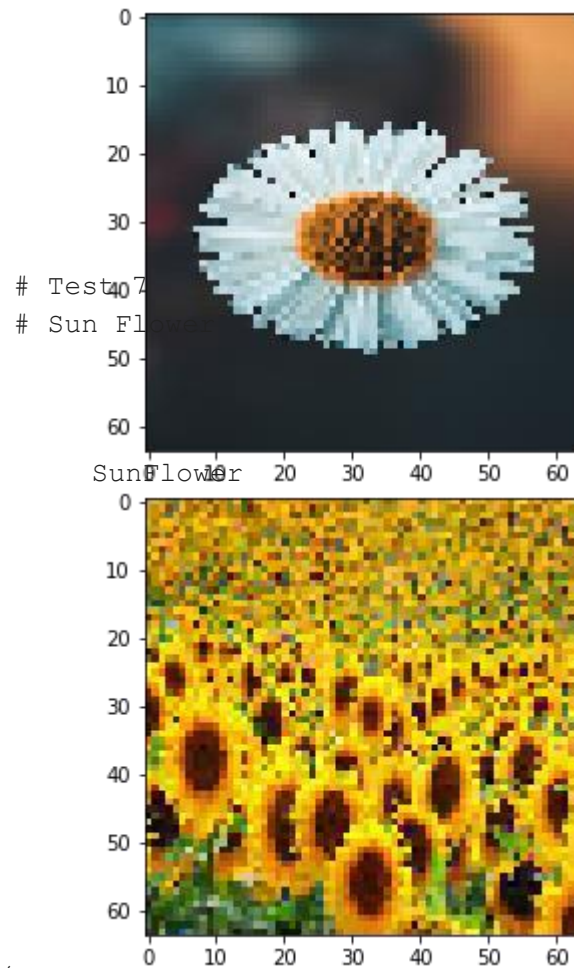
Rose



```
# Test 5 # Daisy Flower predict_flower('/content/IBM Flower_Test dataset/daisy-  
flower-1532449822.jpg') Daisy
```



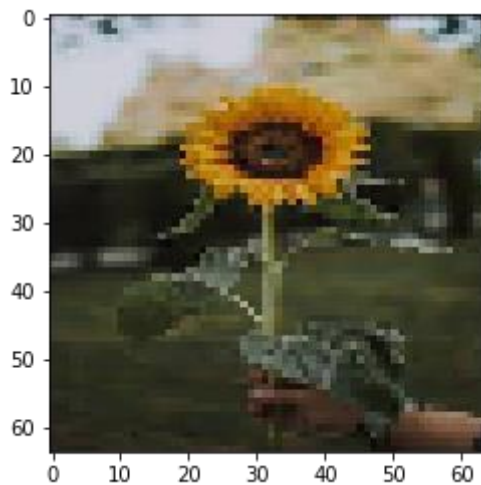
```
# Test 6 # Daisy Flower predict_flower('/content/IBM Flower_Test dataset/photo-  
1606041008023-472dfb5e530f.j Rose
```



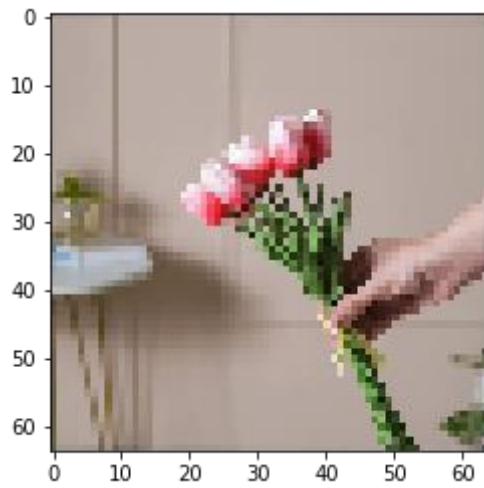
```
predict_flower(
dataset/sunflower.jpeg') '/content/IBM Flower_Test
```

```
# Test 8 # Sun Flower predict_flower('/content/IBM Flower_Test
dataset/sunflower_2.jpeg')
```

SunFlower

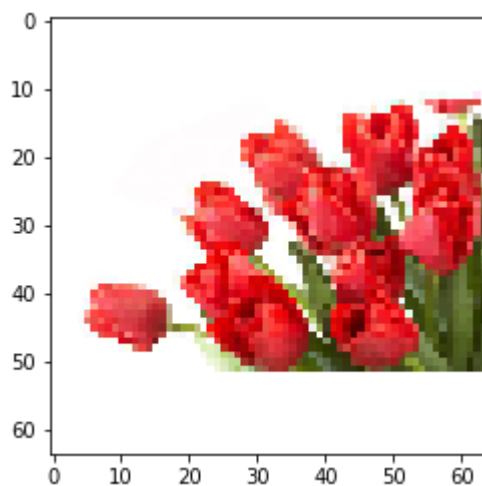


```
# Test 9 # Tulip Flower predict_flower('/content/IBM
Flower_Test dataset/tulip.webp') Tulip
```



```
# Test 10 # Tulip Flower predict_flower('/content/IBM  
Flower_Test dataset/tulip_2.jpg')
```

Tulip



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