```
PROBLEM STATEMENT: Build CNN Model for Classification Of Flowers
Mounting drive
from google.colab import drive
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
DATA AUGMENTATION
from tensorflow.keras.preprocessing.image import ImageDataGenerator
train datagen =
ImageDataGenerator(rescale=1./255,zoom range=0.2,horizontal flip=True,
vertical flip=False, validation split=0.2)
test datagen = ImageDataGenerator(rescale=1./255, validation split=0.2)
x train=train datagen.flow from directory(r"/content/drive/MyDrive/
IBM/Flowers-Dataset/
flowers", target_size=(64,64), class_mode='categorical', batch_size=100,s
ubset = 'training')
Found 3457 images belonging to 5 classes.
x test=test datagen.flow from directory(r"/content/drive/MyDrive/IBM/
Flowers-Dataset/
flowers", target size=(64,64), class mode='categorical', batch size=100, s
ubset = 'validation')
Found 860 images belonging to 5 classes.
x train.class indices
{'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}
CNN MODEL( Adding Layers :Convolution, MaxPooling, Flatten, Dense-(Hidden
Layers),Output)
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Dense, Convolution2D, MaxPooling2D, Flatten
model=Sequential()
model.add(Convolution2D(32,
(3,3),input shape=(64,64,3),activation='relu'))
model.add(MaxPooling2D(pool size=(2,2)))
```

```
model.add(Flatten())
model.summary()
Model: "sequential"
Layer (type)
                        Output Shape
                                               Param #
conv2d (Conv2D)
                         (None, 62, 62, 32)
                                               896
max pooling2d (MaxPooling2D (None, 31, 31, 32)
                                               0
flatten (Flatten)
                         (None, 30752)
                                               0
Total params: 896
Trainable params: 896
Non-trainable params: 0
#hidden layers
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))
model.add(Dense(75,activation='relu'))
model.add(Dense(5,activation='softmax'))#op layer
model.compile(loss='categorical crossentropy',optimizer='adam',metrics
=['accuracy'])
FIT THE MODEL
model.fit generator(x train, steps per epoch=len(x train), validation da
ta=x test, validation steps=len(x test), epochs=10)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:1:
UserWarning: `Model.fit generator` is deprecated and will be removed
in a future version. Please use `Model.fit`, which supports
generators.
 """Entry point for launching an IPython kernel.
Epoch 1/10
- accuracy: 0.6312 - val loss: 1.0603 - val accuracy: 0.5953
Epoch 2/10
35/35 [============== ] - 17s 501ms/step - loss: 0.8961
- accuracy: 0.6503 - val loss: 1.0812 - val accuracy: 0.5826
Epoch 3/10
- accuracy: 0.6743 - val loss: 1.0225 - val accuracy: 0.6244
Epoch 4/10
```

```
- accuracy: 0.6963 - val loss: 1.0625 - val accuracy: 0.6070
Epoch 5/10
35/35 [============== ] - 17s 502ms/step - loss: 0.7726
- accuracy: 0.7081 - val loss: 1.0203 - val accuracy: 0.6105
Epoch 6/10
- accuracy: 0.7249 - val loss: 1.0809 - val accuracy: 0.6326
Epoch 7/10
- accuracy: 0.7495 - val loss: 1.1533 - val accuracy: 0.6081
Epoch 8/10
- accuracy: 0.7524 - val loss: 1.0465 - val accuracy: 0.6349
Epoch 9/10
- accuracy: 0.7712 - val loss: 0.9939 - val accuracy: 0.6581
Epoch 10/10
- accuracy: 0.7926 - val loss: 1.0939 - val accuracy: 0.6291
<keras.callbacks.History at 0x7f40c213dbd0>
SAVING THE MODEL
model.save('flowers.h5')
TESTING THE MODEL
import numpy as np
from tensorflow.keras.preprocessing import image
imq =
image.load img('/content/drive/MyDrive/IBM/Flowers-Dataset/flowers/
daisy/10172636503 21bededa75 n.jpg' ,target size=(64,64))
img
x=image.img to array(img)
Χ
array([[[ 6., 6., 4.],
     [21., 22., 16.],
     [12., 13., 8.],
      . . . ,
```

```
[ 1.,
                      0.],
                1.,
                2.,
         [ 2.,
                      0.],
         [ 2.,
                2.,
                      0.]],
        [[20., 21., 16.],
        [ 7., 7., 5.],
        [22., 23., 18.],
         . . . ,
         [ 0.,
                0.,
                      0.],
         [ 2.,
                2.,
                      0.],
                1.,
         [ 1.,
                      0.]],
        [[16., 17., 12.],
        [20., 21., 16.],
         [11., 12., 7.],
         . . . ,
         [ 1.,
                1.,
                      0.],
                1.,
                      0.],
         [ 1.,
         [ 1.,
                1.,
                      0.]],
        . . . ,
                      9.],
        [[13., 14.,
                9.,
        [ 9.,
                      9.],
        [ 6.,
                6.,
                      4.],
         . . . ,
         [ 1.,
                1.,
                      0.],
                2.,
         [ 2.,
                      0.],
         [ 2.,
                2.,
                      0.]],
        [[ 5.,
                5.,
                      3.],
        [ 7., 7.,
                      5.],
        [14., 15., 10.],
         . . . ,
         [23., 23., 21.],
         [ 1., 1.,
                      0.],
                1.,
                      0.]],
         [ 1.,
                2.,
        [[ 2.,
                      0.],
                7.,
                      5.],
        [ 7.,
                3.,
         [ 3.,
                      1.],
         [22., 22., 20.],
                2.,
         [ 2.,
                      0.],
         [ 2.,
                2.,
                     0.]]], dtype=float32)
x=np.expand dims(x,axis=0)
```

Χ

```
array([[[ 6., 6., 4.],
          [21., 22., 16.],
          [12., 13., 8.],
          . . . ,
                 1.,
          [ 1.,
                       0.],
          [ 2.,
                 2.,
                       0.],
          [ 2.,
                 2.,
                       0.]],
         [[20., 21., 16.],
         [7., 7., 5.],
          [22., 23., 18.],
          . . . ,
          [ 0.,
                 0.,
                       0.],
          [ 2.,
                 2.,
                       0.],
                 1.,
          [ 1.,
                       0.]],
         [[16., 17., 12.],
          [20., 21., 16.],
          [11., 12., 7.],
          . . . ,
          [ 1.,
                 1.,
                       0.],
          [ 1.,
                 1.,
                       0.],
          [ 1.,
                 1.,
                       0.]],
         . . . ,
         [[13., 14.,
                       9.],
         [ 9., 9.,
                       9.],
          [ 6.,
                 6.,
                       4.],
          . . . ,
                 1.,
          [ 1.,
                       0.],
          [ 2.,
                 2.,
                       0.],
                 2.,
          [ 2.,
                       0.]],
         [[ 5.,
                 5.,
                       3.],
          [7., 7., 5.],
          [14., 15., 10.],
          . . . ,
          [23., 23., 21.],
          [ 1.,
                 1.,
                       0.],
                 1.,
          [ 1.,
                       0.]],
                 2.,
         [[ 2.,
                       0.],
          [ 7.,
                 7.,
                       5.],
                       1.],
          [ 3.,
                 3.,
          [22., 22., 20.],
          [ 2., 2.,
                       0.],
                 2.,
                       0.]]]], dtype=float32)
          [ 2.,
```

```
model.predict(x)
array([[0., 0., 1., 0., 0.]], dtype=float32)
x_train.class_indices
{'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}
op = ['daisy','dandelion','rose','sunflower','tulip']
pred = np.argmax(model.predict(x))
op[pred]
{"type":"string"}
img =
image.load img('/content/drive/MyDrive/IBM/Flowers-Dataset/flowers/
dandelion/10828951106 c3cd47983f.jpg',target size=(64,64))
x = image.img_to_array(img)
x = np.expand dims(x,axis=0)
pred = np.argmax(model.predict(x))
op[pred]
{"type":"string"}
```