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PROBLEM STATEMENT: Build CNN Model for Classification Of Flowers
Mounting 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
assignment
3/flowers", target size=(64,64), class mode='categorical', batch size=100
,subset = 'training')
Found 3503 images belonging to 5 classes.
x test=test datagen.flow from directory(r"/content/drive/MyDrive/IBM
assignment
3/flowers", target size=(64,64), class mode='categorical', batch size=100
,subset = 'validation')
Found 872 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)))
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model.add(Flatten())

model.summary()

Model: "sequential"

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Output Shape
Layer (type)
                                             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
 """Entry point for launching an IPython kernel.
Epoch 1/10
- accuracy: 0.3908 - val loss: 1.2657 - val_accuracy: 0.4782
Epoch 2/10
- accuracy: 0.5344 - val loss: 1.1854 - val accuracy: 0.5482
Epoch 3/10
- accuracy: 0.5712 - val loss: 1.2353 - val accuracy: 0.5126
Epoch 4/10
36/36 [============== ] - 18s 482ms/step - loss: 1.0093
- accuracy: 0.5958 - val loss: 1.0953 - val accuracy: 0.5654
Epoch 5/10
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- accuracy: 0.6280 - val loss: 1.1978 - val accuracy: 0.5585
Epoch 6/10
- accuracy: 0.6474 - val loss: 1.0435 - val accuracy: 0.6009
Epoch 7/10
- accuracy: 0.6566 - val loss: 1.1237 - val accuracy: 0.5573
Epoch 8/10
36/36 [============== ] - 17s 486ms/step - loss: 0.8625
- accuracy: 0.6634 - val loss: 1.0356 - val accuracy: 0.6204
Epoch 9/10
- accuracy: 0.6871 - val loss: 1.0691 - val accuracy: 0.6181
Epoch 10/10
36/36 [============== ] - 19s 526ms/step - loss: 0.7800
- accuracy: 0.6951 - val loss: 1.0292 - val accuracy: 0.6089
<keras.callbacks.History at 0x7fe6ccac6d50>
SAVING THE MODEL
model.save('flowers.h5')
TESTING THE MODEL
import numpy as np
from tensorflow.keras.preprocessing import image
img = image.load img('/content/drive/MyDrive/IBM assignment
3/flowers/daisy/10140303196 b88d3d6cec.jpg' ,target size=(64,64))
imq
x=image.img to array(img)
array([[[221., 224., 231.],
      [208., 211., 218.],
      [214., 219., 225.],
      [ 15., 24., 19.],
            16., 17.],
      [ 12..
      [ 14.,
            13., 18.]],
```

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[[228., 231., 240.],
        [220., 221., 226.],
        [222., 216., 226.],
         . . . ,
                 36.,
         [ 27.,
                        33.],
                        20.],
         [ 19.,
                 17.,
           4.,
                  2.,
        [
                        7.]],
       [[223., 224., 229.],
        [234., 239., 245.],
        [228., 233., 239.],
         . . . ,
         [ 34.,
                 32.,
                        37.],
         [ 10.,
                 18.,
                        20.],
        [ 11.,
                 10.,
                        16.]],
        . . . ,
       [[182., 177., 183.],
        [179., 178., 184.],
        [185., 186., 190.],
         [152., 143., 136.],
         [143., 131., 119.],
        [176., 167., 158.]],
        [[174., 178., 181.],
        [169., 168., 173.],
        [199., 196., 203.],
         . . . ,
         [149., 140., 131.],
         [136., 131., 109.],
        [182., 171., 165.]],
        [[225., 230., 236.],
        [229., 232., 239.],
        [229., 232., 239.],
         [149., 141., 128.],
         [144., 140., 111.],
         [182., 174., 161.]]], dtype=float32)
x=np.expand_dims(x,axis=0)
array([[[[221., 224., 231.],
          [208., 211., 218.],
          [214., 219., 225.],
```

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[ 15.,
                  24.,
                         19.],
          [ 12.,
                  16.,
                         17.],
          [ 14.,
                  13.,
                         18.]],
         [[228., 231., 240.],
          [220., 221., 226.],
          [222., 216., 226.],
          [ 27.,
                  36.,
                         33.],
          [ 19.,
                  17.,
                         20.],
          [ 4.,
                   2.,
                          7.]],
         [[223., 224., 229.],
          [234., 239., 245.],
          [228., 233., 239.],
          . . . ,
          [ 34.,
                  32.,
                         37.],
          [ 10.,
                  18.,
                         20.],
          [ 11.,
                  10.,
                         16.]],
         . . . ,
         [[182., 177., 183.],
          [179., 178., 184.],
          [185., 186., 190.],
          [152., 143., 136.],
          [143., 131., 119.],
          [176., 167., 158.]],
         [[174., 178., 181.],
          [169., 168., 173.],
          [199., 196., 203.],
          . . . ,
          [149., 140., 131.],
          [136., 131., 109.],
          [182., 171., 165.]],
         [[225., 230., 236.],
          [229., 232., 239.],
          [229., 232., 239.],
          . . . ,
          [149., 141., 128.],
          [144., 140., 111.],
          [182., 174., 161.]]]], dtype=float32)
model.predict(x)
array([[7.816458e-17, 0.000000e+00, 0.000000e+00, 1.000000e+00,
        0.000000e+00]], 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 assignment
3/flowers/dandelion/10294487385_92a0676c7d_m.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"}
```