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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/
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/
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)))
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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
35/35 [============== ] - 540s 15s/step - loss: 1.3555
- accuracy: 0.4087 - val loss: 1.2931 - val accuracy: 0.4884
Epoch 2/10
- accuracy: 0.5733 - val loss: 1.2284 - val accuracy: 0.5081
Epoch 3/10
- accuracy: 0.6069 - val loss: 1.1106 - val accuracy: 0.5698
Epoch 4/10
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- accuracy: 0.6312 - val loss: 1.0596 - val accuracy: 0.6070
Epoch 5/10
35/35 [============== ] - 17s 491ms/step - loss: 0.8886
- accuracy: 0.6587 - val loss: 1.1208 - val accuracy: 0.5826
Epoch 6/10
- accuracy: 0.6691 - val loss: 1.1426 - val accuracy: 0.5860
Epoch 7/10
- accuracy: 0.6789 - val loss: 1.0249 - val accuracy: 0.6291
Epoch 8/10
- accuracy: 0.7102 - val loss: 1.1002 - val accuracy: 0.5965
Epoch 9/10
- accuracy: 0.7292 - val loss: 1.0417 - val accuracy: 0.6372
Epoch 10/10
- accuracy: 0.7495 - val loss: 1.0116 - val accuracy: 0.6488
<keras.callbacks.History at 0x7f2af156d610>
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/Flowers-Dataset/flowers/dandeli
on/10200780773 c6051a7d71 n.jpg' ,target size=(64,64))
img
x=image.img_to_array(img)
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array([[[ 73., 115., 3.],
     [ 75., 120., 3.],
      [ 74., 119., 0.1,
      . . . ,
```

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[118., 142.,
                         44.],
         [135., 158.,
                         54.],
         [136., 162.,
                         53.]],
        [[ 75., 117.,
                          5.],
                          5.],
         [ 78., 121.,
         [ 82., 123.,
                          3.],
         [118., 141.,
                         35.],
         [132., 156.,
                         44.],
                         44.]],
         [134., 161.,
        [[ 80., 119.,
                         10.],
         [ 86., 124.,
                         13.],
         [ 88., 127.,
                         12.],
         . . . ,
         [119., 145.,
                         35.],
         [130., 154.,
                         40.],
         [133., 163.,
                         41.]],
        . . . ,
        [[117., 120.,
                         33.],
         [113., 134.,
                         31.],
         [114., 142.,
                         31.],
         . . . ,
         [114., 134.,
                         45.],
         [109., 135.,
                         38.],
         [ 98., 130.,
                         21.]],
        [[118., 125.,
                         32.],
         [113., 134.,
                         29.],
         [113., 142.,
                         32.],
         . . . ,
                         48.],
         [113., 136.,
         [106., 134.,
                         34.],
         [ 98., 131.,
                         18.]],
        [[117., 135.,
                         35.],
         [109., 136.,
                         23.],
                         40.],
         [118., 145.,
         . . . ,
         [110., 134.,
                         38.],
         [104., 132.,
                         29.],
         [100., 135.,
                         19.]]], dtype=float32)
x=np.expand dims(x,axis=0)
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array([[[ 73., 115.,
                           3.],
          [ 75., 120.,
                           3.],
          [ 74., 119.,
                           0.],
          [118., 142.,
                          44.],
                          54.],
          [135., 158.,
          [136., 162.,
                          53.]],
         [[ 75., 117.,
                           5.],
          [ 78., 121.,
                           5.],
          [ 82., 123.,
                           3.],
          . . . ,
          [118., 141.,
                          35.],
          [132., 156.,
                          44.],
          [134., 161.,
                          44.]],
         [[ 80., 119.,
                          10.],
          [ 86., 124.,
                          13.],
          [ 88., 127.,
                          12.],
          [119., 145.,
                          35.],
          [130., 154.,
                          40.],
          [133., 163.,
                          41.]],
         . . . ,
         [[117., 120.,
                          33.],
                          31.],
          [113., 134.,
          [114., 142.,
                          31.],
          . . . ,
          [114., 134.,
                          45.],
          [109., 135.,
                          38.],
          [ 98., 130.,
                          21.]],
         [[118., 125.,
                          32.],
          [113., 134.,
                          29.],
          [113., 142.,
                          32.],
          . . . ,
          [113., 136.,
                          48.],
                          34.],
          [106., 134.,
          [ 98., 131.,
                          18.]],
         [[117., 135.,
                          35.],
          [109., 136.,
                          23.],
          [118., 145.,
                          40.],
          [110., 134.,
                          38.],
          [104., 132.,
                          29.],
          [100., 135.,
                          19.]]]], dtype=float32)
```

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model.predict(x)
array([[0.0000000e+00, 0.0000000e+00, 0.0000000e+00, 1.9213412e-21,
        1.0000000e+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/Flowers-Dataset/flowers/tulip/
10094729603_eeca3f2cb6.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"}
```