Assignment -3

Build CNN Model for Classification Of Flowers

Assignment Date	09 October 2022
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Maximum Marks	2 Marks

Task 1:

1. Download the Dataset: Dataset

Solution:

from google.colab import drive drive.mount('/content/drive')

Build CNN model for Classification of Flowers

- 1. Download the Dataset dataset

```
from google.colab import drive drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
```

unzip the file

!unzip "/content/drive/MyDrive/Colab Notebooks/muthamizhan/Flowers-Dataset.zip"

```
lunzip "/content/drive/MyDrive/Colab Notebooks/muthamizhan/Flowers-Dataset.zip"

Archive: /content/drive/MyDrive/Colab Notebooks/muthamizhan/Flowers-Dataset.zip
inflating: flowers/daisy/100808076_fsze8ce07e_n.jpg
inflating: flowers/daisy/10172379554_b296050f82_n.jpg
inflating: flowers/daisy/10172379554_b296050f82_n.jpg
inflating: flowers/daisy/10172567486_274802638b.jpg
inflating: flowers/daisy/1017265073_lbededa7s_n.jpg
inflating: flowers/daisy/102841525_bd6628a26.jpg
inflating: flowers/daisy/1031799732_e7f40080.03.jpg
inflating: flowers/daisy/1031799732_e7f40080.03.jpg
inflating: flowers/daisy/104177954174_22ec996b77_n.jpg
inflating: flowers/daisy/10437772966_bbb677bdd3_m.jpg
inflating: flowers/daisy/10437772966_bbb677bdd3_m.jpg
inflating: flowers/daisy/10437929963_bc13eebe0.jpg
inflating: flowers/daisy/10450528166_d.jpg
inflating: flowers/daisy/10555816204_d211569b0_jpg
inflating: flowers/daisy/1055806016_d1211669b0_jpg
inflating: flowers/daisy/1055806016_d21166d.jpg
inflating: flowers/daisy/1055806016_d21166d.jpg
inflating: flowers/daisy/1055806016_d12116e0.jpg
inflating: flowers/daisy/1055806016_d12116e0.jpg
inflating: flowers/daisy/1055806016_d12116e0.jpg
inflating: flowers/daisy/1055806016_d12116e0.jpg
inflating: flowers/daisy/1055806016_d12116e0.jpg
inflating: flowers/daisy/1055806016_d2116e0.jpg
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inflating: flowers/daisy/1055806016_d2116e0.jpg
inflating: flowers/daisy/1055806016_d2116e0.jpg
inflating: flowers/daisy/10593079.aga-00f678_m.jpg
inflating: flowers/daisy/10903710036_203222201_jpg
inflating: flowers/daisy/10903710036_203222201_jpg
inflating: flowers/daisy/10903710036_203222201_jpg
inflating: flowers/daisy/10903710036_203222201_jpg
inflating: flowers/daisy/10903710036_203322201_jpg
inflating: flowers/daisy/10903710036_203322201_jpg
inflating: flowers/daisy/10903710036_203322201_jpg
inflating: flowers/daisy/10903710036_203322201_jpg
inflating: flowers/daisy/10903710036_203322201_jpg
inflating: flowers/daisy/10903710036_203322201_jpg
```

Task 2:

2. Image Augmentation

Solution:

from tensorflow.keras.preprocessing.image import ImageDataGenerator

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

x_train = train_datagen.flow_from_directory(r"/content/flowers", target_size = (64,64), class_mode = "categorical", batch_size = 100)
"

→ 2. Image Augmentation

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
    train_datagen = ImageDataGenerator(rescale = 1./255, horizontal_flip = True, vertical_flip = True, zoom_range = 0.2)
    x_train = train_datagen.flow_from_directory(r"/content/flowers", target_size = (64,64) , class_mode = "categorical", batch_size = 100)
    Found 4317 images belonging to 5 classes.
```

Task 3:

3. Create Model

Solution:

from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense

model = Sequential()

- 3. Create Model

```
[23] from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense

model = Sequential()
```

Task 4:

Add Layers (Convolution, MaxPooling, Flatten, Dense-(Hidden Layers), Output)

Solution:

```
model.add(Convolution2D(32, (3,3), activation = "relu", input_shape = (64,64,3) ))
model.add(MaxPooling2D(pool_size = (2,2)))
model.add(Flatten())
model.add(Dense(300, activation = "relu"))
```

model.add(Dense(150, activation = "relu")) #multiple dense layers model.add(Dense(5, activation = "softmax")) #output layer

- 4. Add the layers (Convolution, MaxPooling, Flatten, Dense-(HiddenLayers), Output)

```
[24] model.add(Convolution2D(32, (3,3), activation = "relu", input_shape = (64,64,3) ))

model.add(MaxPooling2D(pool_size = (2,2)))

model.add(Flatten())

model.add(Dense(300, activation = "relu"))

model.add(Dense(150, activation = "relu")) #multiple dense layers

model.add(Dense(5, activation = "softmax")) #output layer
```

Task 5:

5. Compile The Model

Solution:

```
model.compile(loss = "categorical_crossentropy", metrics = ["accuracy"], optimizer = "adam") len(x train)
```

→ 5. Compile The Model

```
    [25] model.compile(loss = "categorical_crossentropy", metrics = ["accuracy"], optimizer = "adam")
    len(x_train)

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```

Task 6:

6. Fit The Model

Solution:

model.fit(x train, epochs = 15, steps per epoch = len(x train))

- 6. Fit The Model

```
[26] model.fit(x_train, epochs = 15, steps_per_epoch = len(x_train))
     Epoch 1/15
     44/44 [====
Epoch 2/15
                 ========] - 14s 299ms/step - loss: 1.4965 - accuracy: 0.3864
                 Epoch 3/15
44/44 [====
                    =======] - 13s 294ms/step - loss: 1.0495 - accuracy: 0.5879
     Epoch 4/15
     44/44 [====
Epoch 5/15
44/44 [=====
                   ======== ] - 13s 293ms/step - loss: 0.9727 - accuracy: 0.6196
              Epoch 6/15
     Epoch 7/15
     44/44 [====
Epoch 8/15
             ======== 0.8627 - 135 291ms/step - loss: 0.8661 - accuracy: 0.6627
     44/44 [============] - 13s 299ms/step - loss: 0.8560 - accuracy: 0.6648
Epoch 9/15
             -----] - 13s 292ms/step - loss: 0.8039 - accuracy: 0.6875
     44/44 [====
     Epoch 10/15
     44/44 [=====
Epoch 11/15
                 44/44 [====
Epoch 12/15
                    -----] - 13s 292ms/step - loss: 0.7879 - accuracy: 0.6931
     44/44 [====
Epoch 13/15
                   44/44 [====
                   Epoch 14/15
     44/44 [====
Epoch 15/15
                   -----] - 13s 290ms/step - loss: 0.7279 - accuracy: 0.7181
     <keras.callbacks.History at 0x7f72c927a6d0>
```

model.summary()

/ [27] model.summary()

Model: "sequential_2"

Layer (type)	Output Shape	Param #
conv2d_3 (Conv2D)	(None, 62, 62, 32)	896
<pre>max_pooling2d_2 (MaxPooling 2D)</pre>	(None, 31, 31, 32)	0
flatten_2 (Flatten)	(None, 30752)	0
dense_6 (Dense)	(None, 300)	9225900
dense_7 (Dense)	(None, 150)	45150
dense_8 (Dense)	(None, 5)	755
Total params: 9,272,701 Trainable params: 9,272,701 Non-trainable params: 0		

Task 7:

7. Save The Model

Solution:

model.save("flowers.h5")

→ 7. Save The Model

```
/ [28] model.save("flowers.h5")
```

Task 8:

9. Test The Model

Solution:

```
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
import numpy as np

model = load_model("/content/flowers.h5")
img = image.load_img("/content/flower.jpe", target_size = (64,64))
x = image.img_to_array(img)
x = np.expand_dims(x,axis = 0)
pred = model.predict(x)

labels = ['daisy','dandelion','roses','sunflowers','tulips']
print("Input image is")
img

print("Classification of Flower is:",labels[np.argmax(pred)])
```


Classification of Flower is: sunflowers