Assignment-Ⅲ

Fertilizer recommendation system for disease prediction

from tensorflow.keras.preprocessing.image import ImageDataGenerator

train\_datagen = ImageDataGenerator(rescale= 1./255,horizontal\_flip = True,vertical\_flip = True,zoom\_range = 0.2)

test\_datagen = ImageDataGenerator(rescale= 1./255) x\_train =

train\_datagen.flow\_from\_directory(r"C:\Users\maris\_q3mm6nk\Desktop\data\_for\_ibm\Flowers-

Dataset\flowers",target\_size = (64,64),

class\_mode = "categorical",batch\_size = 24)

Found 4317 images belonging to 5 classes.

x\_test = test\_datagen.flow\_from\_directory(r"C:\Users\maris\_q3mm6nk\Desktop\data\_for\_ibm\Flowers- Dataset\flowers",target\_size = (64,64),

class\_mode = "categorical",batch\_size = 24) Found 4317 images belonging to 5 classes. x\_train.class\_indices

{'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}

from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense

from tensorflow.keras.layers import 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

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Total params: 896

Trainable params: 896

Non-trainable params: 0

model.add(Dense(300,activation='relu')) model.add(Dense(150,activation='relu'))

model.add(Dense(5,activation='softmax')) len(x\_train)

180

model.compile(loss='categorical\_crossentropy',optimizer='adam',metrics=['accuracy'])

model.fit(x\_train,steps\_per\_epoch=len(x\_train),validation\_data=x\_test,validation\_steps=le n(x\_test),epochs=10)

Epoch 1/10

180/180 [==============================] - 33s 183ms/step - loss: 1.3003 - accuracy:

0.4691 - val\_loss: 1.1679 - val\_accuracy: 0.5342 Epoch 2/10

180/180 [==============================] - 28s 157ms/step - loss: 1.0616 - accuracy:

0.5812 - val\_loss: 1.0829 - val\_accuracy: 0.5800 Epoch 3/10

180/180 [==============================] - 28s 157ms/step - loss: 0.9799 - accuracy:

0.6185 - val\_loss: 1.1128 - val\_accuracy: 0.5821 Epoch 4/10

180/180 [==============================] - 29s 161ms/step - loss: 0.9217 - accuracy:

0.6366 - val\_loss: 0.9303 - val\_accuracy: 0.6386 Epoch 5/10

180/180 [==============================] - 28s 158ms/step - loss: 0.8893 - accuracy:

0.6583 - val\_loss: 0.8627 - val\_accuracy: 0.6650 Epoch 6/10

180/180 [==============================] - 29s 162ms/step - loss: 0.8509 - accuracy:

0.6755 - val\_loss: 0.8262 - val\_accuracy: 0.6880 Epoch 7/10

180/180 [==============================] - 30s 169ms/step - loss: 0.8274 - accuracy:

0.6755 - val\_loss: 0.8372 - val\_accuracy: 0.6796 Epoch 8/10

180/180 [==============================] - 30s 166ms/step - loss: 0.7923 - accuracy:

0.6965 - val\_loss: 0.8437 - val\_accuracy: 0.6734 Epoch 9/10

180/180 [==============================] - 28s 157ms/step - loss: 0.7745 - accuracy:

0.7072 - val\_loss: 0.6995 - val\_accuracy: 0.7306 Epoch 10/10

180/180 [==============================] - 28s 158ms/step - loss: 0.7363 - accuracy:

0.7192 - val\_loss: 0.7278 - val\_accuracy: 0.7278

<keras.callbacks.History at 0x16061cf68f0> model.save('IBM\_flowers.h5')

pwd 'C:\\Users\\maris\_q3mm6nk\\Desktop\\data\_for\_ibm'

import numpy as np

from tensorflow.keras.models import load\_model from tensorflow.keras.preprocessing import image

model=load\_model('IBM\_flowers.h5')

img=image.load\_img(r'C:\Users\maris\_q3mm6nk\Desktop\data\_for\_ibm\Flowers- Dataset\flowers\rose/394990940\_7af082cf8d\_n.jpg')

img



img=image.load\_img(r'C:\Users\maris\_q3mm6nk\Desktop\data\_for\_ibm\Flowers- Dataset\flowers\rose/394990940\_7af082cf8d\_n.jpg',target\_size=(64,64)) img



x=image.img\_to\_array(img) x

array([[[ 4., 14., 3.],

[ 4., 15., 0.],

|  |  |  |
| --- | --- | --- |
| [ 7.,  ..., | 10., | 3.], |
| [ 1., | 1., | 1.], |
| [ 1., | 1., | 1.], |
| [ 3., | 3., | 3.]], |
| [[21., | 37., | 8.], |
| [ 7., | 18., | 1.], |
| [ 5., | 11., | 1.], |
| ...,  [ 1., | 1., | 3.], |
| [ 1., | 1., | 1.], |
| [ 2., | 2., | 2.]], |
| [[15., | 34., | 4.], |
| [ 5., | 18., | 0.], |
| [ 6., | 14., | 3.], |
| ..., |  |  |
| [ 1., | 2., | 4.], |
| [ 0., | 0., | 0.], |
| [ 1., | 1., | 1.]], |

...,

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| [[ 7., | | 11., | 10.], | |
| [ 7., | | 16., | 15.], | |
| [17.,  ..., | | 23., | 21.], | |
| [ 1., | | 1., | 1.], | |
| [ 2., | | 2., | 2.], | |
| [ 0., | | 0., | 0.]], | |
| [[ | 9., | 18., | 15.], | |
| [ | 2., | 7., | 3.], | |
| [ | 5., | 11., | 7.], | |
| ..., | | | | |
| [ | 0., | 0., | 0.], | |
| [ | 1., | 1., | 1.], | |
| [ | 1., | 1., | 1.]], | |
| [[18., | | 26., | 28.], |  |
| [ 0., | | 10., | 2.], |  |
| [ 8., | | 14., | 10.], |  |
| ..., | |  |  |  |
| [ 2., | | 6., | 9.], |  |
| [ 1., | | 1., | 1.], |  |
| [ 1., | | 1., | 1.]]], | dtype=float32) |

x=np.expand\_dims(x,axis=0) x

|  |  |  |  |
| --- | --- | --- | --- |
| array([[[[ | 4., | 14., | 3.], |
| [ | 4., | 15., | 0.], |
| [ | 7., | 10., | 3.], |

...,

|  |  |  |  |
| --- | --- | --- | --- |
| [ | 1., | 1., | 1.], |
| [ | 1., | 1., | 1.], |
| [ | 3., | 3., | 3.]], |
| [[21., | | 37., | 8.], |
| [ 7., | | 18., | 1.], |
| [ 5.,  ..., | | 11., | 1.], |
| [ 1., | | 1., | 3.], |
| [ 1., | | 1., | 1.], |
| [ 2., | | 2., | 2.]], |
| [[15., | | 34., | 4.], |
| [ 5., | | 18., | 0.], |
| [ 6., | | 14., | 3.], |
| ...,  [ 1., | | 2., | 4.], |
| [ 0., | | 0., | 0.], |
| [ 1., | | 1., | 1.]], |

...,

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| [[ 7., | | 11., | 10.], | |
| [ 7., | | 16., | 15.], | |
| [17., | | 23., | 21.], | |
| ..., | |  |  | |
| [ 1., | | 1., | 1.], | |
| [ 2., | | 2., | 2.], | |
| [ 0., | | 0., | 0.]], | |
| [[ | 9., | 18., | 15.], | |
| [ | 2., | 7., | 3.], | |
| [ | 5., | 11., | 7.], | |
| ..., | | | | |
| [ | 0., | 0., | 0.], | |
| [ | 1., | 1., | 1.], | |
| [ | 1., | 1., | 1.]], | |
| [[18., | | 26., | 28.], |  |
| [ 0., | | 10., | 2.], |  |
| [ 8., | | 14., | 10.], |  |
| ...,  [ 2., | | 6., | 9.], |  |
| [ 1., | | 1., | 1.], |  |
| [ 1., | | 1., | 1.]]]], | dtype=float32) |

y=np.argmax(model.predict(x),axis=1) y

1/1 [==============================] - 0s 74ms/step

array([2], dtype=int64) x\_train.class\_indices

{'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}

index=['daisy','dandelion','rose','sunflower','tulip'] index[y[0]]

'rose'