Project Development Phase Model Performance Test

Date	17 November 2022	
Team ID	PNT2022TMID50383	
Project Name	Fertilizer Recommendation System For	
	Disease Prediction	
Maximum Marks	10 Marks	

Model Performance Testing:

Project team shall fill the following information in model performance testing template.

S.No.	Parameter	Values	Screenshot
1.	Model Summary	Total params: 36,160,485 Trainable params: 1,024,005 Non-trainable params: 20,861,480	Attached below
2.	Accuracy	Training Accuracy - 72%	Attached below
		Validation Accuracy - 59%	
3.	Confidence Score (Only	Class Detected - NILL	NILL
	Yolo Projects)	Confidence Score - NILL	

SCREENSHOTS:

Veg:

```
In [1]: from keras.preprocessing.image import ImageDataGenerator
                             train\_datagen=ImageDataGenerator(rescale=1./255, shear\_range=0.2, zoom\_range=0.2, horizontal\_flip=True)
                            test_datagen=ImageDataGenerator(rescale=1)
In [2]: x_train=train_datagen.flow_from_directory(r'C:\Users\princ\OneDrive\Desktop\Dataset Plant Disease\Veg-dataset\Veg-dataset\train_set',target_size=(128, x_test=test_datagen.flow_from_directory(r'C:\Users\princ\OneDrive\Desktop\Dataset Plant Disease\Veg-dataset\Veg-dataset\test_set',target_size=(128,128, x_test=test_datagen.flow_from_directory(r'C:\Users\princ\OneDrive\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Desktop\Deskt
                          Found 11386 images belonging to 9 classes.
                          Found 3416 images belonging to 9 classes.
In [3]: from keras.models import Sequential
                             from keras.layers import Dense
                             from keras.layers import Convolution2D
                             from keras.layers import MaxPooling2D
                            from keras.layers import Flatten
In [4]: model=Sequential()
In [5]: model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))
In [6]: model.add(MaxPooling2D(pool_size=(2,2)))
In [7]: model.add(Flatten())
In [8]: model.add(Dense(units=300,kernel_initializer='uniform',activation='relu'))
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In [8]:
           model.add(Dense(units=300.kernel initializer='uniform',activation='relu'))
 In [9]:
           model.add(Dense(units=150,kernel_initializer='uniform',activation='relu'))
 In [10]:
           model.add(Dense(units=75,kernel_initializer='uniform',activation='relu'))
           model.add(Dense(units=9,kernel initializer='uniform',activation='softmax'))
           model.compile(loss='categorical_crossentropy',optimizer="adam",metrics=["accuracy"])
 In [13]:
          model.fit(x_train,steps_per_epoch=89,epochs=20,validation_data=x_test,validation_steps=27)
          Epoch 1/20
          89/89 [====
                         Epoch 2/20
                                  :========] - 34s 386ms/step - loss: 1.3260 - accuracy: 0.5105 - val_loss: 144.9129 - val_accuracy: 0.5104
          89/89 [====
          Epoch 3/20
          89/89 [===
                                      =======] - 31s 348ms/step - loss: 1.0517 - accuracy: 0.6204 - val_loss: 452.8666 - val_accuracy: 0.2465
          Epoch 4/20
                                     =======] - 30s 340ms/step - loss: 0.9506 - accuracy: 0.6570 - val loss: 1062.1256 - val accuracy: 0.2801
          89/89 [===
          Epoch 5/20
          89/89 [===
                                                - 33s 368ms/step - loss: 0.7732 - accuracy: 0.7268 - val_loss: 713.5864 - val_accuracy: 0.3264
          Epoch 6/20
                                   ========] - 33s 372ms/step - loss: 0.6780 - accuracy: 0.7574 - val_loss: 1175.1545 - val_accuracy: 0.2801
          89/89 [====
          Epoch 7/20
          89/89 [===
                                   =========] - 31s 352ms/step - loss: 0.6116 - accuracy: 0.7745 - val_loss: 1305.8286 - val_accuracy: 0.2743
          Epoch 8/20
          89/89 [====
                                 ========] - 29s 323ms/step - loss: 0.5763 - accuracy: 0.7903 - val_loss: 1302.3727 - val_accuracy: 0.2269
          Epoch 9/20
          89/89 [====
                                =========] - 30s 342ms/step - loss: 0.5767 - accuracy: 0.7946 - val_loss: 1277.9509 - val_accuracy: 0.3056
        Epoch 10/20
        89/89 [====
                               ========] - 31s 348ms/step - loss: 0.5556 - accuracy: 0.8013 - val_loss: 1235.4139 - val_accuracy: 0.3113
        Epoch 11/20
        89/89 [=:
                                        ==] - 32s 355ms/step - loss: 0.4403 - accuracy: 0.8392 - val_loss: 1530.7563 - val_accuracy: 0.3461
        Epoch 12/20
        89/89 [=====
                                        ===] - 31s 344ms/step - loss: 0.4323 - accuracy: 0.8522 - val_loss: 1574.8284 - val_accuracy: 0.3125
        Epoch 13/20
                                ========] - 29s 320ms/step - loss: 0.4533 - accuracy: 0.8343 - val loss: 1657.8656 - val accuracy: 0.2627
        89/89 [=====
        Epoch 14/20
        89/89 [=====
                                ========] - 29s 324ms/step - loss: 0.4202 - accuracy: 0.8498 - val_loss: 1603.1924 - val_accuracy: 0.3021
        Epoch 15/20
        89/89 [====
                                     ======] - 26s 289ms/step - loss: 0.3590 - accuracy: 0.8792 - val_loss: 1485.8334 - val_accuracy: 0.3623
        Epoch 16/20
        89/89 [==
                                     =====] - 25s 281ms/step - loss: 0.3680 - accuracy: 0.8694 - val_loss: 2185.7332 - val_accuracy: 0.2708
        Epoch 17/20
        89/89 [====
                                  =======] - 24s 270ms/step - loss: 0.3488 - accuracy: 0.8792 - val_loss: 2317.3721 - val_accuracy: 0.2870
        Epoch
              18/20
        89/89 [=====
                                ========] - 24s 264ms/step - loss: 0.3258 - accuracy: 0.8902 - val loss: 1627.2305 - val accuracy: 0.3009
        Epoch 19/20
        89/89 [====
                              ========] - 26s 293ms/step - loss: 0.3268 - accuracy: 0.8869 - val loss: 1958.1587 - val accuracy: 0.2546
        Epoch
        89/89 [======
                          ===========] - 28s 313ms/step - loss: 0.2834 - accuracy: 0.9031 - val_loss: 2081.9343 - val_accuracy: 0.3090
In [14]:
         model.save('vegetabledata.h5')
```

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89/89 [====
                     Epoch 20/20
       89/89 [=============] - 28s 313ms/step - loss: 0.2834 - accuracy: 0.9031 - val loss: 2081.9343 - val accuracy: 0.3090
        model.save('vegetabledata.h5')
In [15]:
       model.summary()
       Model: "sequential"
       Layer (type)
                               Output Shape
                                                    Param #
       conv2d (Conv2D)
                               (None, 126, 126, 32)
       max_pooling2d (MaxPooling2D) (None, 63, 63, 32)
                                                    0
       flatten (Flatten)
                               (None, 127008)
       dense (Dense)
                                                    38102700
                               (None, 300)
       dense_1 (Dense)
                               (None, 150)
                                                    45150
       dense_2 (Dense)
                                                    11325
                               (None, 75)
       dense_3 (Dense)
                               (None, 9)
                                                    684
       Total params: 38,160,755
       Trainable params: 38,160,755
       Non-trainable params: 0
```

Fruit:

```
In [1]:
             from keras.preprocessing.image import ImageDataGenerator
train_datagen=ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal_flip=True)
             test_datagen=ImageDataGenerator(rescale=1)
In [2]: x_train=train_datagen.flow_from_directory(r'C:\Users\princ\OneDrive\Desktop\Dataset Plant Disease\fruit-dataset\fruit-dataset\train',target_size=(128, x_test=test_datagen.flow_from_directory(r'C:\Users\princ\OneDrive\Desktop\Dataset Plant Disease\fruit-dataset\fruit-dataset\train',target_size=(128, 128)
             Found 5384 images belonging to 6 classes.
            Found 1686 images belonging to 6 classes.
In [3]: from keras.models import Sequential
             from keras.layers import Dense
from keras.layers import Convolution2D
             from keras.layers import MaxPooling2D
from keras.layers import Flatten
In [4]: model=Sequential()
In [5]: model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))
In [6]: model.add(MaxPooling2D(pool_size=(2,2)))
In [7]:
             model.add(Flatten())
In [8]:
             model.add(Dense(units=40,kernel_initializer='uniform',activation='relu'))
model.add(Dense(units=70,kernel_initializer='random_uniform',activation='relu'))
model.add(Dense(units=6,kernel_initializer='random_uniform',activation='softmax'))
             model.compile(loss='categorical_crossentropy',optimizer="adam",metrics=["accuracy"])
```

Output	Shape	Param #
(None,	126, 126, 32)	896
(None,	63, 63, 32)	0
(None,	127008)	0
(None,	40)	5080360
(None,	70)	2870
(None,	6)	426
	(None, (None, (None, (None,	Output Shape (None, 126, 126, 32) (None, 63, 63, 32) (None, 127008) (None, 40) (None, 70) (None, 6)

Total params: 5,084,552 Trainable params: 5,084,552 Non-trainable params: 0