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    "from keras.preprocessing.image import ImageDataGenerator\n",
"train datagen=ImageDataGenerator(rescale=1./255, shear range=0.2, zoom range
=0.2, horizontal flip=True) \n",
    "test datagen=ImageDataGenerator(rescale=1)"
  ]
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      "Found 11386 images belonging to 9 classes.\n",
      "Found 3416 images belonging to 9 classes.\n"
     ]
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   "source": [
"x_train=train_datagen.flow_from_directory(r'C:\\Users\\sridevi\\project\\D
ataset Plant Disease\\Veg-dataset\\Veg-
dataset\\train set',target size=(128,128),batch size=2,class mode='categori
cal') \n",
```

```
"x_test=test_datagen.flow_from_directory(r'C:\\Users\\sridevi\\project\\Dat
aset Plant Disease\\Veg-dataset\\Veg-
dataset\\test set',target size=(128,128),batch size=2,class mode='categoric
al')"
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    "from keras.models import Sequential\n",
    "from keras.layers import Dense\n",
    "from keras.layers import Convolution2D\n",
    "from keras.layers import MaxPooling2D\n",
    "from keras.layers import Flatten"
  ]
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    "model=Sequential()"
  1
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```

```
"outputs": [],
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"model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'
) ) "
   ]
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    "model.add(MaxPooling2D(pool size=(2,2)))"
   ]
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    "model.add(Flatten())"
   ]
  },
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```

```
"model.add(Dense(units=300, kernel_initializer='uniform', activation='relu'))
  ]
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"model.add(Dense(units=150,kernel initializer='uniform',activation='relu'))
  ]
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"model.add(Dense(units=75,kernel initializer='uniform',activation='relu'))"
  ]
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   "source": [
```

```
"model.add(Dense(units=9, kernel initializer='uniform', activation='softmax')
  ]
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"model.compile(loss='categorical crossentropy',optimizer=\"adam\",metrics=[
\"accuracy\"])"
  1
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     "Epoch 1/20\n",
     "89/89 [=========== ] - 95s 1s/step - loss: 2.1765
- accuracy: 0.1404 - val_loss: 107.0669 - val_accuracy: 0.2407\n",
      "Epoch 2/20\n",
     "89/89 [============= ] - 61s 679ms/step - loss:
2.1010 - accuracy: 0.2303 - val loss: 73.7251 - val accuracy: 0.0741\n",
     "Epoch 3/20\n",
     "89/89 [=========== ] - 67s 755ms/step - loss:
2.1514 - accuracy: 0.1348 - val loss: 56.0996 - val accuracy: 0.1111\n",
     "Epoch 4/20\n",
```

```
"89/89 [====================] - 64s 717ms/step - loss:
2.0868 - accuracy: 0.1573 - val loss: 23.7097 - val accuracy: 0.3148\n",
     "Epoch 5/20\n",
     "89/89 [============= ] - 60s 671ms/step - loss:
2.0239 - accuracy: 0.3090 - val loss: 99.1493 - val accuracy: 0.2222\n",
     "Epoch 6/20\n",
     1.9236 - accuracy: 0.2753 - val loss: 172.7210 - val accuracy: 0.1296\n",
     "Epoch 7/20\n",
     "89/89 [============== ] - 86s 971ms/step - loss:
1.9143 - accuracy: 0.2753 - val loss: 107.2718 - val accuracy: 0.2778\n",
     "Epoch 8/20\n",
     "89/89 [============== ] - 58s 646ms/step - loss:
1.7796 - accuracy: 0.3034 - val loss: 64.2221 - val accuracy: 0.3148\n",
     "Epoch 9/20\n",
     "89/89 [============== ] - 60s 676ms/step - loss:
1.7756 - accuracy: 0.3427 - val loss: 182.9076 - val accuracy: 0.3519\n",
     "Epoch 10/20\n",
     "89/89 [============ ] - 59s 665ms/step - loss:
1.8444 - accuracy: 0.2978 - val loss: 138.7072 - val accuracy: 0.2407\n",
     "Epoch 11/20\n",
     "89/89 [============ ] - 53s 598ms/step - loss:
1.7811 - accuracy: 0.2640 - val loss: 111.3470 - val accuracy: 0.3333\n",
     "Epoch 12/20\n",
     "89/89 [============= ] - 57s 637ms/step - loss:
1.8700 - accuracy: 0.2809 - val_loss: 104.8549 - val_accuracy: 0.2778\n",
     "Epoch 13/20\n",
     "89/89 [============ ] - 53s 593ms/step - loss:
1.8179 - accuracy: 0.3371 - val loss: 88.9790 - val accuracy: 0.3519\n",
     "Epoch 14/20\n",
     "89/89 [====================] - 53s 590ms/step - loss:
1.7108 - accuracy: 0.2921 - val loss: 79.7810 - val accuracy: 0.4074\n",
     "Epoch 15/20\n",
     "89/89 [============= ] - 57s 644ms/step - loss:
1.8212 - accuracy: 0.2416 - val loss: 187.6725 - val accuracy: 0.2222\n",
     "Epoch 16/20\n",
     "89/89 [===================] - 53s 589ms/step - loss:
1.7251 - accuracy: 0.3483 - val loss: 148.7835 - val accuracy: 0.4259\n",
     "Epoch 17/20\n",
```

```
"89/89 [============== ] - 56s 624ms/step - loss:
1.6795 - accuracy: 0.3146 - val_loss: 109.7393 - val_accuracy: 0.2593\n",
     "Epoch 18/20\n",
     "89/89 [======== ] - 54s 608ms/step - loss:
1.8275 - accuracy: 0.3202 - val loss: 90.4495 - val accuracy: 0.3148\n",
     "Epoch 19/20\n",
     "89/89 [============ ] - 53s 591ms/step - loss:
1.7737 - accuracy: 0.3427 - val loss: 141.6376 - val accuracy: 0.2593\n",
     "Epoch 20/20\n",
     "89/89 [============= ] - 53s 592ms/step - loss:
1.7060 - accuracy: 0.4213 - val loss: 96.3340 - val accuracy: 0.3889\n"
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"model.fit(x train, steps per epoch=89, epochs=20, validation data=x test, vali
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  ]
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```

```
"model.save(r'C:\\Users\\ishu\\project\\flask\\uploads\\vegetable.h5')
  ]
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  "id": "abce82ea",
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                                             ____\n",
                    Output Shape
    " Layer (type)
                                                Param #
\n",
"-----\n",
    " conv2d (Conv2D)
                           (None, 126, 126, 32) 896
\n",
\n",
    " max pooling2d (MaxPooling2D (None, 63, 63, 32) 0
\n",
    " )
\n",
\n",
    " flatten (Flatten) (None, 127008) 0
\n",
\n",
    " dense (Dense) (None, 300) 38102700
\n",
\n",
```

```
" dense 1 (Dense) (None, 150)
                                                     45150
\n",
\n",
     " dense 2 (Dense) (None, 75)
                                                    11325
\n",
\n",
   " dense_3 (Dense) (None, 9)
                                                       684
\n",
\n",
     "Total params: 38,160,755\n",
     "Trainable params: 38,160,755\n",
     "Non-trainable params: 0\n",
                                                               \n"
   ]
  }
  ],
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  "model.summary()"
  ]
 }
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  "version": 3
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