

```
# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES
# TO THE CORRECT LOCATION (/kaggle/input) IN YOUR NOTEBOOK,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.

import os
import sys
from tempfile import NamedTemporaryFile
from urllib.request import urlopen
from urllib.parse import unquote, urlparse
from urllib.error import HTTPError
from zipfile import ZipFile
import tarfile
import shutil

CHUNK_SIZE = 40960
DATA_SOURCE_MAPPING = 'new-plant-diseases-dataset:https%3A%2F%2Fstorage.googleapis.com%2Fkaggle-data-sets%2F78313%2F182633%2Fbundle%2Farchive.zip%3FX-Goog-Alt'

KAGGLE_INPUT_PATH='/kaggle/input'
KAGGLE_WORKING_PATH='/kaggle/working'
KAGGLE_SYMLINK='kaggle'

!umount /kaggle/input/ 2> /dev/null
shutil.rmtree('/kaggle/input', ignore_errors=True)
os.makedirs(KAGGLE_INPUT_PATH, 0o777, exist_ok=True)
os.makedirs(KAGGLE_WORKING_PATH, 0o777, exist_ok=True)

try:
    os.symlink(KAGGLE_INPUT_PATH, os.path.join(".", 'input'), target_is_directory=True)
except FileExistsError:
    pass
try:
    os.symlink(KAGGLE_WORKING_PATH, os.path.join(".", 'working'), target_is_directory=True)
except FileExistsError:
    pass

for data_source_mapping in DATA_SOURCE_MAPPING.split(','):
    directory, download_url_encoded = data_source_mapping.split(':')
    download_url = unquote(download_url_encoded)
    filename = urlparse(download_url).path
    destination_path = os.path.join(KAGGLE_INPUT_PATH, directory)
    try:
        with urlopen(download_url) as fileres, NamedTemporaryFile() as tfile:
            total_length = fileres.headers['content-length']
            print(f'Downloading {directory}, {total_length} bytes compressed')
            dl = 0
            data = fileres.read(CHUNK_SIZE)
            while len(data) > 0:
                dl += len(data)
                tfile.write(data)
                done = int(50 * dl / int(total_length))
                sys.stdout.write(f"\r[{'=' * done}{' ' * (50-done)}] {dl} bytes downloaded")
                sys.stdout.flush()
                data = fileres.read(CHUNK_SIZE)
            if filename.endswith('.zip'):
                with ZipFile(tfile) as zfile:
                    zfile.extractall(destination_path)
            else:
                with tarfile.open(tfile.name) as tarfile:
                    tarfile.extractall(destination_path)
            print(f'\nDownloaded and uncompressed: {directory}')
    except HTTPError as e:
        print(f'Failed to load (likely expired) {download_url} to path {destination_path}')
        continue
    except OSError as e:
        print(f'Failed to load {download_url} to path {destination_path}')
        continue

print('Data source import complete.')
```

 Downloading new-plant-diseases-dataset, 2897709187 bytes compressed
 [=====] 2897709187 bytes downloaded
 Downloaded and uncompressed: new-plant-diseases-dataset
 Data source import complete.

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load

import numpy as np # linear algebra
```

```
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
```

```
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
```

```
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

```
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session
```



Streaming output truncated to the last 5000 lines.

```
/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/12913612-
/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/c1721601-
/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/d824b284-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/d2729b47-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/e5a94b14-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/c1721601-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/d42d1a5-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/287d7355-
/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/3a5abb6b-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/359365bc-
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/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Grape___Black_rot/632803d3-
```

```
import tensorflow as tf
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

```
validation_set = tf.keras.utils.image_dataset_from_directory(
    '/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/valid',
    labels="inferred",
    label_mode="categorical",
    class_names=None,
    color_mode="rgb",
    batch_size=32,
    image_size=(128, 128),
    shuffle=True,
    seed=None,
    validation_split=None,
```

```
subset=None,
interpolation="bilinear",
follow_links=False,
crop_to_aspect_ratio=False
)
```

```
class_name = validation_set.class_names
print(class_name)
```

Found 17572 files belonging to 38 classes.
 ['Apple__Apple_scab', 'Apple__Black_rot', 'Apple__Cedar_apple_rust', 'Apple__healthy', 'Blueberry__healthy', 'Cherry_(including_sour)__Powdery_mil']

```
training_set = tf.keras.utils.image_dataset_from_directory(
    '/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train',
    labels="inferred",
    label_mode="categorical",
    class_names=None,
    color_mode="rgb",
    batch_size=32,
    image_size=(128, 128),
    shuffle=True,
    seed=None,
    validation_split=None,
    subset=None,
    interpolation="bilinear",
    follow_links=False,
    crop_to_aspect_ratio=False
)
```

Found 70295 files belonging to 38 classes.

Building Model

```
cnn = tf.keras.models.Sequential()
```

Building Convolution Layer

```
cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, padding='same', activation='relu', input_shape=[128, 128, 3]))
cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu'))
cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
```

```
cnn.add(tf.keras.layers.Conv2D(filters=64, kernel_size=3, padding='same', activation='relu'))
cnn.add(tf.keras.layers.Conv2D(filters=64, kernel_size=3, activation='relu'))
cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
```

```
cnn.add(tf.keras.layers.Conv2D(filters=128, kernel_size=3, padding='same', activation='relu'))
cnn.add(tf.keras.layers.Conv2D(filters=128, kernel_size=3, activation='relu'))
cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
```

```
cnn.add(tf.keras.layers.Conv2D(filters=256, kernel_size=3, padding='same', activation='relu'))
cnn.add(tf.keras.layers.Conv2D(filters=256, kernel_size=3, activation='relu'))
cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
```

```
cnn.add(tf.keras.layers.Conv2D(filters=512, kernel_size=3, padding='same', activation='relu'))
cnn.add(tf.keras.layers.Conv2D(filters=512, kernel_size=3, activation='relu'))
cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
```

```
cnn.add(tf.keras.layers.Dropout(0.25))
```

```
cnn.add(tf.keras.layers.Flatten())
```

```
cnn.add(tf.keras.layers.Dense(units=1500, activation='relu'))
```

```
cnn.add(tf.keras.layers.Dropout(0.4))
```

```
cnn.add(tf.keras.layers.Dense(units=38, activation='softmax'))
```

Compiling and Training Phase

```
cnn.compile(optimizer=tf.keras.optimizerslegacy.Adam(
    learning_rate=0.0001), loss='categorical_crossentropy', metrics=['accuracy'])
```

cnn.summary()



Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 128, 128, 32)	896
conv2d_1 (Conv2D)	(None, 126, 126, 32)	9248
max_pooling2d (MaxPooling2D)	(None, 63, 63, 32)	0
conv2d_2 (Conv2D)	(None, 63, 63, 64)	18496
conv2d_3 (Conv2D)	(None, 61, 61, 64)	36928
max_pooling2d_1 (MaxPooling2D)	(None, 30, 30, 64)	0
conv2d_4 (Conv2D)	(None, 30, 30, 128)	73856
conv2d_5 (Conv2D)	(None, 28, 28, 128)	147584
max_pooling2d_2 (MaxPooling2D)	(None, 14, 14, 128)	0
conv2d_6 (Conv2D)	(None, 14, 14, 256)	295168
conv2d_7 (Conv2D)	(None, 12, 12, 256)	590080
max_pooling2d_3 (MaxPooling2D)	(None, 6, 6, 256)	0
conv2d_8 (Conv2D)	(None, 6, 6, 512)	1180160
conv2d_9 (Conv2D)	(None, 4, 4, 512)	2359808
max_pooling2d_4 (MaxPooling2D)	(None, 2, 2, 512)	0
dropout (Dropout)	(None, 2, 2, 512)	0
flatten (Flatten)	(None, 2048)	0
dense (Dense)	(None, 1500)	3073500
dropout_1 (Dropout)	(None, 1500)	0
dense_1 (Dense)	(None, 38)	57038
Total params: 7842762 (29.92 MB)		
Trainable params: 7842762 (29.92 MB)		
Non-trainable params: 0 (0.00 Byte)		

```
training_history = cnn.fit(x=training_set,validation_data=validation_set,epochs=10)
```



```
Epoch 1/10
2197/2197 [=====] - 162s 70ms/step - loss: 1.4444 - accuracy: 0.5755 - val_loss: 0.5067 - val_accuracy: 0.8424
Epoch 2/10
2197/2197 [=====] - 150s 68ms/step - loss: 0.4673 - accuracy: 0.8511 - val_loss: 0.2903 - val_accuracy: 0.9102
Epoch 3/10
2197/2197 [=====] - 152s 69ms/step - loss: 0.2817 - accuracy: 0.9083 - val_loss: 0.2028 - val_accuracy: 0.9356
Epoch 4/10
2197/2197 [=====] - 152s 69ms/step - loss: 0.1949 - accuracy: 0.9362 - val_loss: 0.1634 - val_accuracy: 0.9483
Epoch 5/10
2197/2197 [=====] - 152s 69ms/step - loss: 0.1401 - accuracy: 0.9537 - val_loss: 0.2065 - val_accuracy: 0.9384
Epoch 6/10
2197/2197 [=====] - 161s 73ms/step - loss: 0.1091 - accuracy: 0.9641 - val_loss: 0.1267 - val_accuracy: 0.9594
Epoch 7/10
2197/2197 [=====] - 151s 69ms/step - loss: 0.0939 - accuracy: 0.9692 - val_loss: 0.1455 - val_accuracy: 0.9533
Epoch 8/10
2197/2197 [=====] - 151s 69ms/step - loss: 0.0816 - accuracy: 0.9727 - val_loss: 0.1161 - val_accuracy: 0.9637
Epoch 9/10
2197/2197 [=====] - 161s 73ms/step - loss: 0.0648 - accuracy: 0.9789 - val_loss: 0.1507 - val_accuracy: 0.9582
Epoch 10/10
2197/2197 [=====] - 151s 69ms/step - loss: 0.0587 - accuracy: 0.9809 - val_loss: 0.1763 - val_accuracy: 0.9537
```

```
#Training set Accuracy
train_loss, train_acc = cnn.evaluate(training_set)
print('Training accuracy:', train_acc)
```



```
2197/2197 [=====] - 46s 21ms/step - loss: 0.0477 - accuracy: 0.9843
Training accuracy: 0.9842520952224731
```

```
#Validation set Accuracy
val_loss, val_acc = cnn.evaluate(validation_set)
print('Validation accuracy:', val_acc)
```

```
↗ 550/550 [=====] - 12s 21ms/step - loss: 0.1763 - accuracy: 0.9537
Validation accuracy: 0.9537332057952881
```

```
cnn.save('trained_plant_disease_model.keras')
```

```
training_history.history
```

```
↗ {'loss': [1.4443633556365967,
0.4673091471195221,
0.28167495131492615,
0.19489221274852753,
0.14010187983512878,
0.10909982025623322,
0.09389577805995941,
0.08163385838270187,
0.06479869037866592,
0.05872979760169983],
'accuracy': [0.5755032300949097,
0.8511131405830383,
0.9082865118980408,
0.9361547827720642,
0.9536666870117188,
0.9641368389129639,
0.9692296981811523,
0.9726865291595459,
0.9788604974746704,
0.9808521270751953],
'val_loss': [0.5066662430763245,
0.2903105914592743,
0.20278093218803406,
0.16342882812023163,
0.20645461976528168,
0.1266852766275406,
0.145468071103096,
0.11605484783649445,
0.15067780017852783,
0.17630502581596375],
'val_accuracy': [0.8423628211021423,
0.9101980328559875,
0.9356362223625183,
0.9483268857002258,
0.9383678436279297,
0.9593671560287476,
0.9533348679542542,
0.9637491703033447,
0.958172082901001,
0.9537332057952881]}
```

```
import numpy as np
import tensorflow as tf
from keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt
```

```
cnn = tf.keras.models.load_model('trained_plant_disease_model.keras')
```

```
import cv2
image_path = '/kaggle/input/new-plant-diseases-dataset/New Plant Diseases Dataset(Augmented)/New Plant Diseases Dataset(Augmented)/train/Strawberry___health
# Reading an image in default mode
img = cv2.imread(image_path)
img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
# Displaying the image
plt.imshow(img)
plt.title('Test Image')
plt.xticks([])
plt.yticks([])
plt.show()
```



Test Image

**Loading Model**

```
image = tf.keras.preprocessing.image.load_img(image_path,target_size=(128,128))
input_arr = tf.keras.preprocessing.image.img_to_array(image)
input_arr = np.array([input_arr]) # Convert single image to a batch.
predictions = cnn.predict(input_arr)
```



```
1/1 [=====] - 1s 727ms/step
```



```
print(predictions)
```



```
[[[5.010552e-13 1.3067867e-12 1.0732161e-13 5.3974633e-12 6.2073080e-10
1.8932536e-12 1.4026540e-16 5.0029766e-17 6.8041122e-19 1.1565596e-15
1.4242793e-18 7.2466303e-15 8.9017025e-18 6.0567762e-14 1.4675618e-09
2.7214923e-14 4.3491018e-16 1.8718665e-17 1.4204996e-15 1.7397306e-13
2.8894090e-14 1.4291915e-10 8.3225504e-10 1.7983723e-07 7.3056934e-12
7.8461702e-11 5.9628729e-07 9.9999917e-01 1.5113362e-14 1.2415093e-13
3.3360532e-13 5.0125474e-17 7.7981692e-12 1.3234882e-11 1.1702539e-10
2.4612673e-14 1.1255341e-13 2.3143443e-10]]]
```

```
result_index = np.argmax(predictions) #Return index of max element
print(result_index)
```



```
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```

```
# Displaying the disease prediction
model_prediction = class_name[result_index]
plt.imshow(img)
plt.title(f"Disease Name: {model_prediction}")
plt.xticks([])
plt.yticks([])
plt.show()
```



Disease Name: Strawberry__healthy



Start coding or [generate](#) with AI.