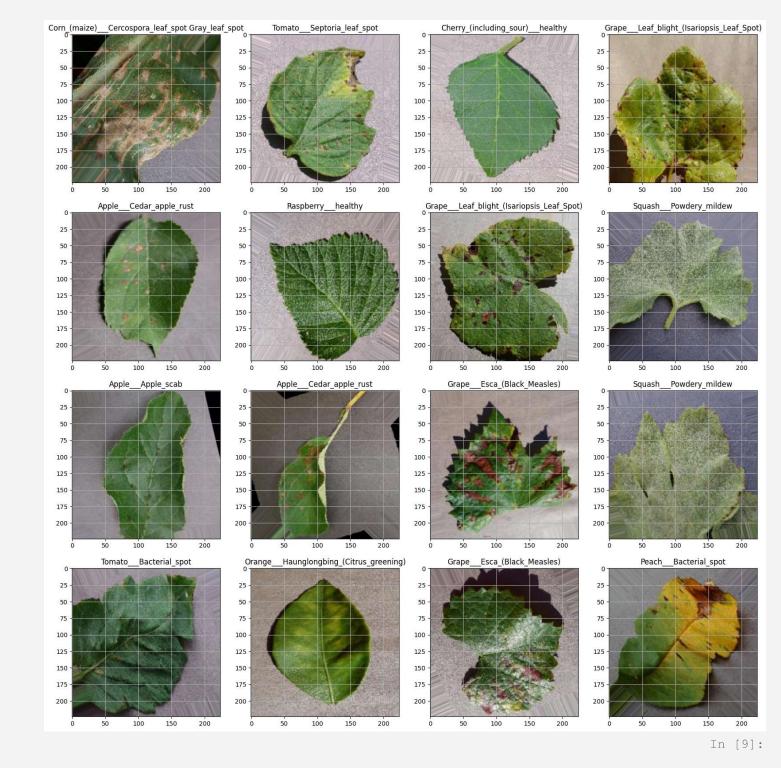
```
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
                                                                                       In [2]:
import zipfile
import os
# Define the path to the zip file and the extraction directory
zip_file_path = "/content/drive/MyDrive/archive (2).zip"
extract_dir = "/content"
# Check if the zip file exists
if os.path.exists(zip_file_path):
    try:
        # Create a ZipFile object
        with zipfile.ZipFile(zip file path, 'r') as zip ref:
            # Extract all contents to the specified directory
            zip ref.extractall(extract dir)
        print(f"Successfully extracted '{zip file path}' to '{extract dir}'")
    except zipfile.BadZipFile:
        print(f"Error: '{zip file path}' is not a valid zip file.")
    except Exception as e:
        print(f"An error occurred: {e}")
else:
    print(f"Error: '{zip file path}' does not exist.")
Successfully extracted '/content/drive/MyDrive/archive (2).zip' to '/content'
                                                                                       In [3]:
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import pathlib
import os
import glob as gb
import glob
import cv2
import tensorflow as tf
                                                                                       In [4]:
train='/content/New Plant Diseases Dataset(Augmented)/New Plant Diseases
Dataset (Augmented) / train'
```

Mount the drive here

```
In [5]:
size=224
                                                                                      In [6]:
train generator=tf.keras.preprocessing.image.ImageDataGenerator(
    rotation range=90,
   width shift range=0.0,
   height shift range=0.0,
   shear_range=0.0,
   zoom range=0.0,
   horizontal flip=False,
   vertical flip=False,
   rescale=1/255.0,
   preprocessing function=None,
   validation split=0.1,
).flow_from_directory(train,
                     batch size=164,
                     target size=(size, size),
                     subset="training",
                     color_mode='rgb', #"rgb", "rgba", or "grayscale"
                      class mode='categorical', # Use 'binary', 'sparse', 'categorical' or
None as needed
                     shuffle=True)
Found 63282 images belonging to 38 classes.
                                                                                      In [7]:
train generator.class indices
                                                                                      Out[7]:
{'Apple__Apple_scab': 0,
 'Apple Black_rot': 1,
 'Apple___Cedar_apple_rust': 2,
 'Apple healthy': 3,
 'Blueberry healthy': 4,
 'Cherry_(including_sour)___Powdery_mildew': 5,
 'Cherry_(including_sour)___healthy': 6,
 'Corn (maize) Cercospora leaf spot Gray leaf spot': 7,
 'Corn (maize) Common rust ': 8,
 'Corn_(maize)___Northern_Leaf_Blight': 9,
 'Corn (maize) healthy': 10,
 'Grape Black rot': 11,
 'Grape___Esca_(Black_Measles)': 12,
 'Grape___Leaf_blight_(Isariopsis_Leaf_Spot)': 13,
 'Grape healthy': 14,
 'Orange Haunglongbing (Citrus greening)': 15,
 'Peach Bacterial spot': 16,
 'Peach healthy': 17,
 'Pepper, bell Bacterial spot': 18,
```

```
'Pepper,_bell__healthy': 19,
 'Potato Early blight': 20,
 'Potato___Late_blight': 21,
 'Potato healthy': 22,
 'Raspberry healthy': 23,
 'Soybean healthy': 24,
 'Squash__Powdery_mildew': 25,
 'Strawberry___Leaf_scorch': 26,
 'Strawberry healthy': 27,
 'Tomato Bacterial spot': 28,
 'Tomato___Early_blight': 29,
 'Tomato Late blight': 30,
 'Tomato Leaf Mold': 31,
 'Tomato __Septoria_leaf_spot': 32,
 'Tomato___Spider_mites Two-spotted_spider_mite': 33,
 'Tomato Target Spot': 34,
 'Tomato Tomato Yellow Leaf Curl Virus': 35,
 'Tomato Tomato mosaic virus': 36,
 'Tomato healthy': 37}
                                                                                     In [8]:
#let's display some images from training set
classes=list(train generator.class indices.keys())
plt.figure(figsize=(20,20))
for X_batch, y_batch in train_generator:
   # create a grid of 3x3 images
   for i in range (0,16):
       plt.subplot(4,4,i+1)
       plt.imshow(X batch[i])
       plt.title(classes[np.where(y batch[i]==1)[0][0]]) # when y is categorical
       # plt.title(classes[int(y_batch[i])]) # when y is binary or sparse
       plt.grid(None)
    # show the plot
   plt.show()
   break
```



valid='/content/New Plant Diseases Dataset(Augmented)/New Plant Diseases
Dataset(Augmented)/valid'

In [10]:

```
valid_generator=tf.keras.preprocessing.image.ImageDataGenerator(
    rescale=1/255.0,
    preprocessing_function=None,
    validation_split=0.1,
).flow_from_directory(valid,
```

```
batch size=164,
                      target size=(224,224),
                      subset='validation',
                      color mode='rgb', #"rgb", "rgba", or "grayscale"
                      class mode='categorical', # Use 'binary', 'sparse', 'categorical' or
None as needed
                     shuffle=False)
Found 1742 images belonging to 38 classes.
                                                                                     In [11]:
test='/content/New Plant Diseases Dataset(Augmented)/New Plant Diseases
Dataset (Augmented) / valid'
                                                                                     In [12]:
test generator=tf.keras.preprocessing.image.ImageDataGenerator(
   rescale=1/255.0,
   preprocessing function=None,
).flow from directory(test,
                     batch size=164,
                     target size=(224, 224),
                     color mode='rgb', #"rgb", "rgba", or "grayscale"
                     class mode='categorical', # Use 'binary', 'sparse', 'categorical' or
None as needed
                     shuffle=False)
Found 17572 images belonging to 38 classes.
                                                                                     In [13]:
test generator.class indices
                                                                                     Out[13]:
{'Apple Apple scab': 0,
 'Apple Black rot': 1,
 'Apple___Cedar_apple_rust': 2,
 'Apple___healthy': 3,
 'Blueberry healthy': 4,
 'Cherry (including sour) Powdery mildew': 5,
 'Cherry (including sour) healthy': 6,
 'Corn (maize) Cercospora leaf spot Gray leaf spot': 7,
 'Corn (maize) Common rust ': 8,
 'Corn_(maize)___Northern_Leaf_Blight': 9,
 'Corn_(maize)__healthy': 10,
 'Grape Black rot': 11,
 'Grape Esca (Black Measles)': 12,
 'Grape Leaf blight_(Isariopsis_Leaf_Spot)': 13,
 'Grape___healthy': 14,
 'Orange Haunglongbing (Citrus greening)': 15,
 'Peach___Bacterial_spot': 16,
 'Peach healthy': 17,
 'Pepper,_bell___Bacterial_spot': 18,
```

```
'Pepper,_bell___healthy': 19,
 'Potato___Early_blight': 20,
 'Potato___Late_blight': 21,
 'Potato healthy': 22,
 'Raspberry healthy': 23,
 'Soybean healthy': 24,
 'Squash Powdery mildew': 25,
 'Strawberry___Leaf_scorch': 26,
 'Strawberry healthy': 27,
 'Tomato Bacterial spot': 28,
 'Tomato____Early_blight': 29,
 'Tomato___Late_blight': 30,
 'Tomato Leaf Mold': 31,
 'Tomato Septoria leaf spot': 32,
 'Tomato___Spider_mites Two-spotted spider mite': 33,
 'Tomato Target Spot': 34,
 'Tomato Tomato Yellow Leaf Curl Virus': 35,
 'Tomato Tomato mosaic virus': 36,
 'Tomato healthy': 37}
                                                                                    In [14]:
from tensorflow import keras
model = keras.models.Sequential() # To build NN
model.add(keras.layers.Conv2D(filters=32, kernel size=7, strides=1,
                padding="same", activation="relu", name="Conv1", input shape= (224,224,3)))
model.add(keras.layers.MaxPool2D(pool size=2, name="Pool1"))
model.add(keras.layers.Conv2D(filters=64, kernel size=5, strides=1,
                             padding="same", activation="relu", name="Conv2"))
model.add(keras.layers.MaxPool2D(pool size=2, name="Pool2"))
model.add(keras.layers.Conv2D(filters=128, kernel size=3, strides=1,
                             padding="same", activation="relu", name="Conv3"))
model.add(keras.layers.Conv2D(filters=256, kernel size=3, strides=1,
                              padding="same", activation="relu", name="Conv4"))
model.add(keras.layers.MaxPool2D(pool_size=2, name="Pool3"))
model.add(keras.layers.Flatten(name="Flatten1")) #flatten layer - to convert into 1d
vector
model.add(keras.layers.Dense(128, activation="relu", name="Dense1")) #hidden layer
tf.keras.layers.Dropout(0.5)
model.add(keras.layers.Dense(64, activation="relu", name="Dense2")) #hidden layer
tf.keras.layers.Dropout(0.5)
model.add(keras.layers.Dense(38, activation="softmax", name="Output")) #output layer
# The model's summary() method displays all the model's layers
print(model.summary())
```

/usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

super().__init__(activity_regularizer=activity_regularizer, **kwargs)

Model: "sequential"

Layer (type)	Output Shape	Param #
L L		Tatam #
Conv1 (Conv2D)	(None, 224, 224, 32)	4,736
Pool1 (MaxPooling2D)	(None, 112, 112, 32)	0
Conv2 (Conv2D)	(None, 112, 112, 64)	51,264
Pool2 (MaxPooling2D)	(None, 56, 56, 64)	0
Conv3 (Conv2D)	(None, 56, 56, 128)	73,856
Conv4 (Conv2D)	(None, 56, 56, 256)	295,168
Pool3 (MaxPooling2D)	(None, 28, 28, 256)	0
Flatten1 (Flatten)	(None, 200704)	0
Densel (Dense)	(None, 128)	25,690,240
Dense2 (Dense)	(None, 64)	8,256
Output (Dense)	(None, 38)	2,470

Total params: 26,125,990 (99.66 MB)

Trainable params: 26,125,990 (99.66 MB)

Non-trainable params: 0 (0.00 B)

None