

# Exploratory Data Analysis

```
# Required imports
import os
import json
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import cv2
from PIL import Image
import hashlib
import os
import cv2

# Set base directory for dataset
BASE_DIR = "/kaggle/input/cassava-leaf-disease-classification"

with
open("/kaggle/input/cassava-leaf-disease-classification/label_num_to_d
isease_map.json") as file:
    print("yes")

yes

# Step 1: Load and inspect label map (mapping from numerical labels to
disease names)
with open(os.path.join(BASE_DIR, "label_num_to_disease_map.json")) as
file:
    map_classes = json.loads(file.read())
    map_classes = {int(k): v for k, v in map_classes.items()}

# Display the mapping
print("Class Mapping: ")
print(json.dumps(map_classes, indent=4))

Class Mapping:
{
    "0": "Cassava Bacterial Blight (CBB)",
    "1": "Cassava Brown Streak Disease (CBDSD)",
    "2": "Cassava Green Mottle (CGM)",
    "3": "Cassava Mosaic Disease (CMD)",
    "4": "Healthy"
}
os.listdir(os.path.join(BASE_DIR, "train_images"))
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'2118008819.jpg',
'2557274516.jpg',
'3887762313.jpg',
'1419604978.jpg',
'3567504840.jpg',
'1358792990.jpg',
'2611538307.jpg',
'3500950357.jpg',
'2139984567.jpg',
'1658205752.jpg',
'2000281372.jpg',
'2002346677.jpg',
'1218794118.jpg',
'3647482405.jpg',
'4274749483.jpg',
'4040861068.jpg',
'2042895796.jpg',
'3419507688.jpg',
'2318645335.jpg',
'91285032.jpg',
'385391871.jpg',
'2448870532.jpg',
'3276292509.jpg',
'785251696.jpg',
'3740700061.jpg',
'2689770530.jpg',
'2086436188.jpg',
'846824837.jpg',
'2623245968.jpg',
'2587436758.jpg',
'2606643559.jpg',
'1948267522.jpg',
'1436718296.jpg',
'3042292188.jpg',
'127776052.jpg',
```

```
'48153077.jpg',
'3964029612.jpg',
'2209303439.jpg',
'3373304739.jpg',
'2045580929.jpg',
'3440246067.jpg',
'1214627201.jpg',
'3979826725.jpg',
'639060341.jpg',
'849182498.jpg',
'4233882902.jpg',
'2047500718.jpg',
'161212223.jpg',
'3237441683.jpg',
'2405316066.jpg',
'1796222011.jpg',
'1909574970.jpg',
'598746218.jpg',
'3797289739.jpg',
'394451018.jpg',
'4284813323.jpg',
'1691647530.jpg',
'15982075.jpg',
'3197563521.jpg',
'1960012199.jpg',
'3870634420.jpg',
'2800300240.jpg',
'1561998426.jpg',
'2226045745.jpg',
'920623788.jpg',
'731084892.jpg',
'2137780185.jpg',
'4130557422.jpg',
'1639780135.jpg',
'4256968855.jpg',
'4170665280.jpg',
'1067379852.jpg',
'4038568741.jpg',
'3808794161.jpg',
'208826492.jpg',
'1474434656.jpg',
'1951683874.jpg',
'2597470760.jpg',
'3077845683.jpg',
'4116414929.jpg',
'2601706130.jpg',
'1826527887.jpg',
'4234605337.jpg',
'936775758.jpg',
```

```
'2941780886.jpg',
'2871575859.jpg',
'3058839740.jpg',
'3298030379.jpg',
'3567421807.jpg',
'701148586.jpg',
'2164873412.jpg',
'4261671268.jpg',
'4056070889.jpg',
'3121142461.jpg',
'1117199954.jpg',
'3602830290.jpg',
'1296168681.jpg',
'3044190781.jpg',
'910617288.jpg',
'1048581072.jpg',
'2476543961.jpg',
'2733802395.jpg',
'780779910.jpg',
'2583614987.jpg',
'2930575492.jpg',
'3951384519.jpg',
'1149596528.jpg',
'2978135052.jpg',
'133303828.jpg',
'3114522519.jpg',
'2213611476.jpg',
'2859411048.jpg',
'1805115397.jpg',
'2606136202.jpg',
'1829519358.jpg',
'1677822348.jpg',
'3977934674.jpg',
'2596954655.jpg',
'4522938.jpg',
'3464607854.jpg',
'3267434230.jpg',
'2146353282.jpg',
'1736448195.jpg',
'1723407805.jpg',
'2915309072.jpg',
'3436413534.jpg',
'472489554.jpg',
'4254213032.jpg',
'611507457.jpg',
'503224990.jpg',
'451261982.jpg',
'3766633636.jpg',
'639068838.jpg',
```

'557774617.jpg',  
'1047894047.jpg',  
'1546329958.jpg',  
'3568729258.jpg',  
'3633505917.jpg',  
'2072537637.jpg',  
'3667619405.jpg',  
'3011700717.jpg',  
'2818289247.jpg',  
'424999624.jpg',  
'3451069987.jpg',  
'1476112995.jpg',  
'3676837791.jpg',  
'1472183727.jpg',  
'2836107083.jpg',  
'370989494.jpg',  
'2633910453.jpg',  
'1891182946.jpg',  
'52672633.jpg',  
'635279232.jpg',  
'2740240477.jpg',  
'3849865547.jpg',  
'4215655540.jpg',  
'3471618012.jpg',  
'895910836.jpg',  
'2240297228.jpg',  
'2209356814.jpg',  
'1365404548.jpg',  
'4279248558.jpg',  
'4018307313.jpg',  
'2504965655.jpg',  
'1809122626.jpg',  
'2430296671.jpg',  
'3964194408.jpg',  
'2441927798.jpg',  
'2097619367.jpg',  
'1445369057.jpg',  
'1515163743.jpg',  
'2907186124.jpg',  
'3680354956.jpg',  
'2435254407.jpg',  
'2543879211.jpg',  
'2286124427.jpg',  
'2059469005.jpg',  
'2526623317.jpg',  
'3811396790.jpg',  
'1239825198.jpg',  
'2027324099.jpg',  
'335796643.jpg',

```
'2150392038.jpg',
'2752466458.jpg',
'673868311.jpg',
'854627770.jpg',
'3511001525.jpg',
'4205544766.jpg',
'2057007338.jpg',
'4135070493.jpg',
'1079224858.jpg',
'2587457959.jpg',
'2477858047.jpg',
'551875095.jpg',
'2872638305.jpg',
'2932008073.jpg',
'4228467711.jpg',
'2276509518.jpg',
'813060428.jpg',
'4096337072.jpg',
'1391911669.jpg',
'315046129.jpg',
'2321993342.jpg',
'2445348448.jpg',
'114081332.jpg',
'3684621874.jpg',
'248804719.jpg',
'2997295640.jpg',
'5511383.jpg',
'4114035268.jpg',
'2023813792.jpg',
'4121046251.jpg',
'4059993044.jpg',
'2338213285.jpg',
'2289989107.jpg',
'4120621808.jpg',
'3964080612.jpg',
'1735461481.jpg',
'2854111497.jpg',
'4020138210.jpg',
'4263725317.jpg',
'3638122648.jpg',
'745723934.jpg',
'2035397744.jpg',
'157149350.jpg',
'396528848.jpg',
'1399448037.jpg',
'3058038323.jpg',
'2949246528.jpg',
'3746522482.jpg',
'3020460837.jpg',
```

```

['3345615928.jpg',
'232481441.jpg',
'1824113054.jpg',
'1927144610.jpg',
'3158054107.jpg',
'2717129685.jpg',
'3145516632.jpg',
'1220365722.jpg',
'2085011211.jpg',
'1776880375.jpg',
'3427476782.jpg',
'80482467.jpg',
'568226171.jpg',
'1160075077.jpg',
'3277409188.jpg',
'4025247063.jpg',
...]

# Step 2: Load training image filenames and display the count
input_files = os.listdir(os.path.join(BASE_DIR, "train_images"))
print(f"Number of train images: {len(input_files)})")

Number of train images: 21397

# Step 3: Load train.csv and add a human-readable class name based on the mapping
df_train = pd.read_csv(os.path.join(BASE_DIR, "train.csv"))
df_train.head()

      image_id  label
0  1000015157.jpg      0
1  1000201771.jpg      3
2  100042118.jpg       1
3  1000723321.jpg       1
4  1000812911.jpg      3

df_train["class_name"] = df_train["label"].map(map_classes)
df_train

      image_id  label      class_name
0  1000015157.jpg      0  Cassava Bacterial Blight (CBB)
1  1000201771.jpg      3  Cassava Mosaic Disease (CMD)
2  100042118.jpg       1  Cassava Brown Streak Disease (CBSD)
3  1000723321.jpg       1  Cassava Brown Streak Disease (CBSD)
4  1000812911.jpg      3  Cassava Mosaic Disease (CMD)
...   ...
21392  999068805.jpg      3  Cassava Mosaic Disease (CMD)
21393  999329392.jpg      3  Cassava Mosaic Disease (CMD)
21394  999474432.jpg      1  Cassava Brown Streak Disease (CBSD)
21395  999616605.jpg      4  Healthy
21396  999998473.jpg      4  Healthy

```

```
[21397 rows x 3 columns]

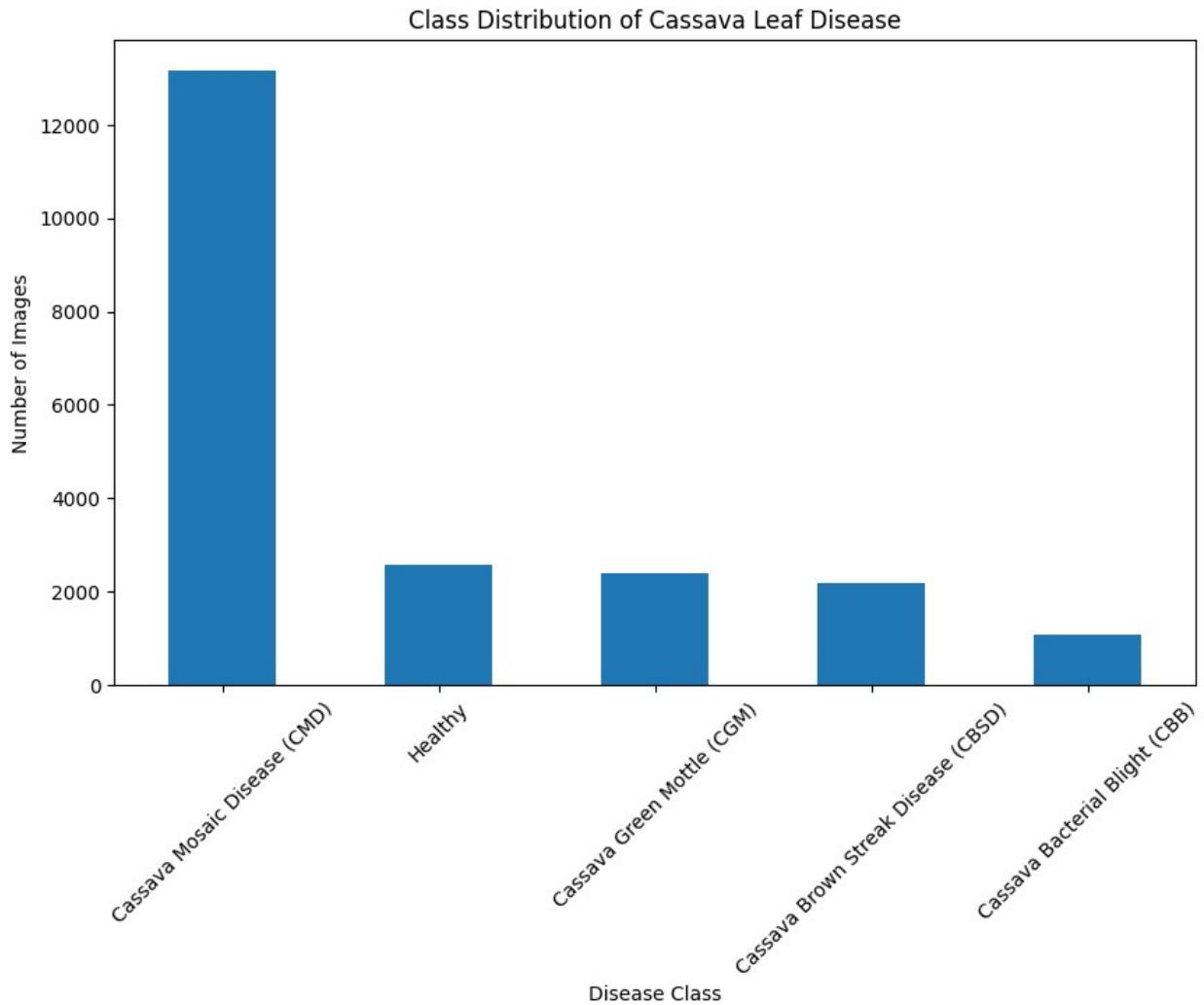
df_train['class_name'].value_counts()

class_name
Cassava Mosaic Disease (CMD)      13158
Healthy                           2577
Cassava Green Mottle (CGM)        2386
Cassava Brown Streak Disease (CBSD) 2189
Cassava Bacterial Blight (CBB)    1087
Name: count, dtype: int64
```

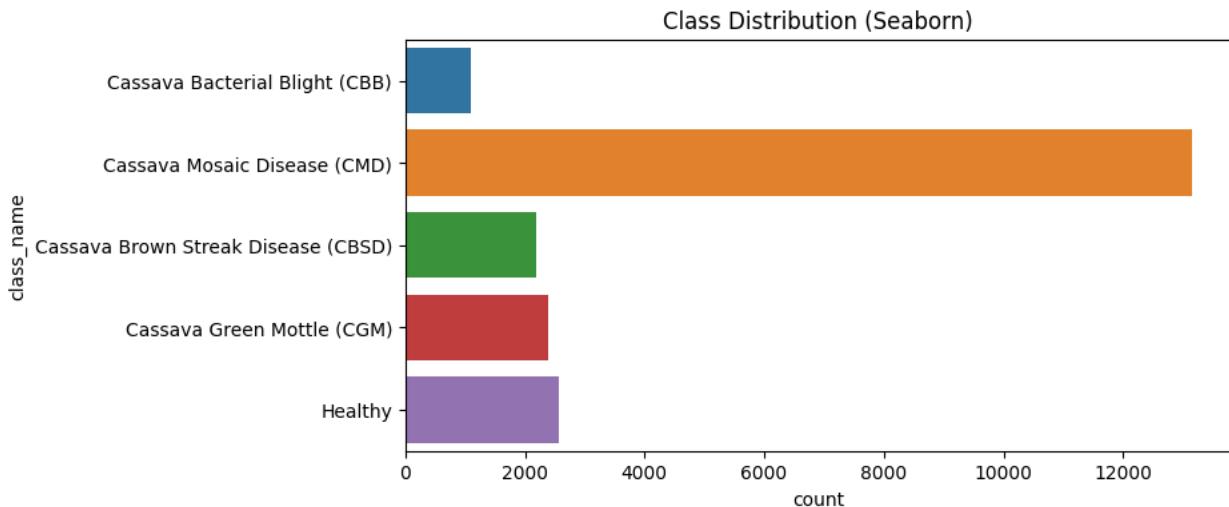
Techniques like transfer learning, you don't need to balance the data. Internally model will handle. (Pre-trained model)

Scratch - then you need to do something about imbalance..

```
# Step 4: Check class distribution
class_distribution = df_train['class_name'].value_counts()
# Plot the class distribution
plt.figure(figsize=(10, 6))
class_distribution.plot(kind='bar')
plt.title('Class Distribution of Cassava Leaf Disease')
plt.ylabel('Number of Images')
plt.xlabel('Disease Class')
plt.xticks(rotation=45)
plt.show()
```



```
# Alternatively, use seaborn for a countplot visualization
plt.figure(figsize=(8, 4))
sns.countplot(y="class_name", data=df_train)
plt.title('Class Distribution (Seaborn)')
plt.show()
```



```
# Step 5: Basic dataset exploration
# Show data info and summary statistics
print("Dataset Info:")
print(df_train.info())
```

```
Dataset Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 21397 entries, 0 to 21396
Data columns (total 3 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   image_id    21397 non-null   object 
 1   label       21397 non-null   int64  
 2   class_name  21397 non-null   object 
dtypes: int64(1), object(2)
memory usage: 501.6+ KB
None
```

```
print("\nDataset Summary Statistics:")
print(df_train.describe())
```

```
Dataset Summary Statistics:
label
count  21397.000000
mean    2.651914
std     0.988565
min    0.000000
25%   2.000000
50%   3.000000
75%   3.000000
max    4.000000
```

```

# Step 6: Check for missing values and duplicates
print(f"\nMissing values in each column:{df_train.isnull().sum()}")
print(f"Number of duplicate rows: {df_train.duplicated().sum()}")


Missing values in each column:
image_id      0
label         0
class_name    0
dtype: int64
Number of duplicate rows: 0

```

Images - let's check the shape of the images.

Images - 5050 (*very small pixels*) it doesn't make any sense to upscale to 224224 ( mess up informaiton )

images 600 \* 600 resize to may be 224 224 500 \* 500

images size distributin you need to tune your resize image size.

```

path =
"/kaggle/input/cassava-leaf-disease-classification/train_images/100001
5157.jpg"
path2 = "1000015157.jpg"

cv2.imread(path2)

[ WARN:0@638.540] global loadsave.cpp:241 findDecoder
imread_('1000015157.jpg'): can't open/read file: check file
path/integrity

cv2.imread(path2)

[ WARN:0@706.511] global loadsave.cpp:241 findDecoder
imread_('1000015157.jpg'): can't open/read file: check file
path/integrity

# Step 7: Analyze image shapes (size dimensions) for a sample of 300
# images
# Dictionary to store image shapes and their counts
img_shapes = {}
for image_name in os.listdir(os.path.join(BASE_DIR, "train_images"))[:1000]:
    image = cv2.imread(os.path.join(BASE_DIR, "train_images",
image_name))
    img_shapes[image.shape] = img_shapes.get(image.shape, 0) + 1

# Display image shapes
print("\nSample Image Shapes and their Frequencies (from 1000

```

```
images):")
print(img_shapes)
```

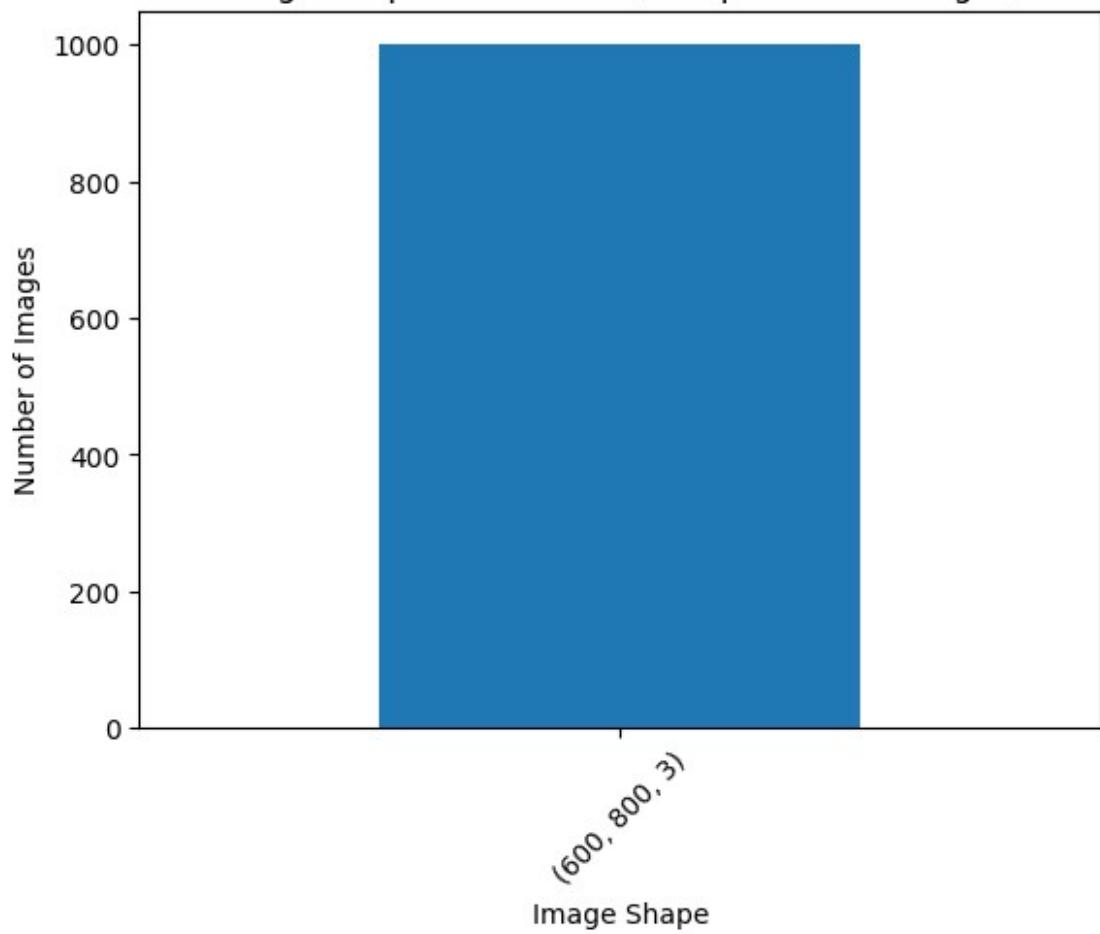
```
Sample Image Shapes and their Frequencies (from 1000 images):
{(600, 800, 3): 1000}
```

I have my images with the size 600 \* 800 \* 3 (rgb channels) color full image.

```
# Plot the image size distribution
img_shapes_df = pd.DataFrame(list(img_shapes.items()), columns=['Image
Shape', 'Count'])
plt.figure(figsize=(10, 6))
img_shapes_df.sort_values(by='Count',
ascending=False).plot(kind='bar', x='Image Shape', y='Count',
legend=False)
plt.title('Image Shape Distribution (Sample of 300 Images)')
plt.xlabel('Image Shape')
plt.ylabel('Number of Images')
plt.xticks(rotation=45)
plt.show()
```

```
<Figure size 1000x600 with 0 Axes>
```

Image Shape Distribution (Sample of 300 Images)



# Loading data

shapes

info

null vlaues

describe

image shapes

```
df_train.head()

   image_id  label      class_name
0  1000015157.jpg     0  Cassava Bacterial Blight (CBB)
1  1000201771.jpg     3  Cassava Mosaic Disease (CMD)
2  100042118.jpg      1  Cassava Brown Streak Disease (CBS)
3  1000723321.jpg      1  Cassava Brown Streak Disease (CBS)
4  1000812911.jpg     3  Cassava Mosaic Disease (CMD)

# Step 8: Function to plot sample images from a specific class
def plot_images_from_class(class_id, num_images=9):
    """
    Plot sample images from a specific class in a 3x3 grid.

    Parameters:
        class_id (int): The class label to filter images.
        num_images (int): The number of images to plot.
    """
    # Filter images for the specified class
    class_images = df_train[df_train['label'] == class_id]
    num_images = min(len(class_images), num_images) # Adjust if fewer
    images than requested

    plt.figure(figsize=(15, 15)) # Set figure size for better
    visualization
    images = class_images.sample(num_images) # Randomly sample images

    # Plot images in a 3x3 grid
```

```
for i, (_, row) in enumerate(images.iterrows()):
    img_path = os.path.join(BASE_DIR, "train_images",
row['image_id']))
    img = Image.open(img_path)
    plt.subplot(3, 3, i + 1)
    plt.imshow(img)
    plt.title(map_classes[class_id]) # Use class name for the
title
    plt.axis('off') # Hide axis for better visualization

plt.tight_layout() # Adjust layout to prevent overlap
plt.show()
```

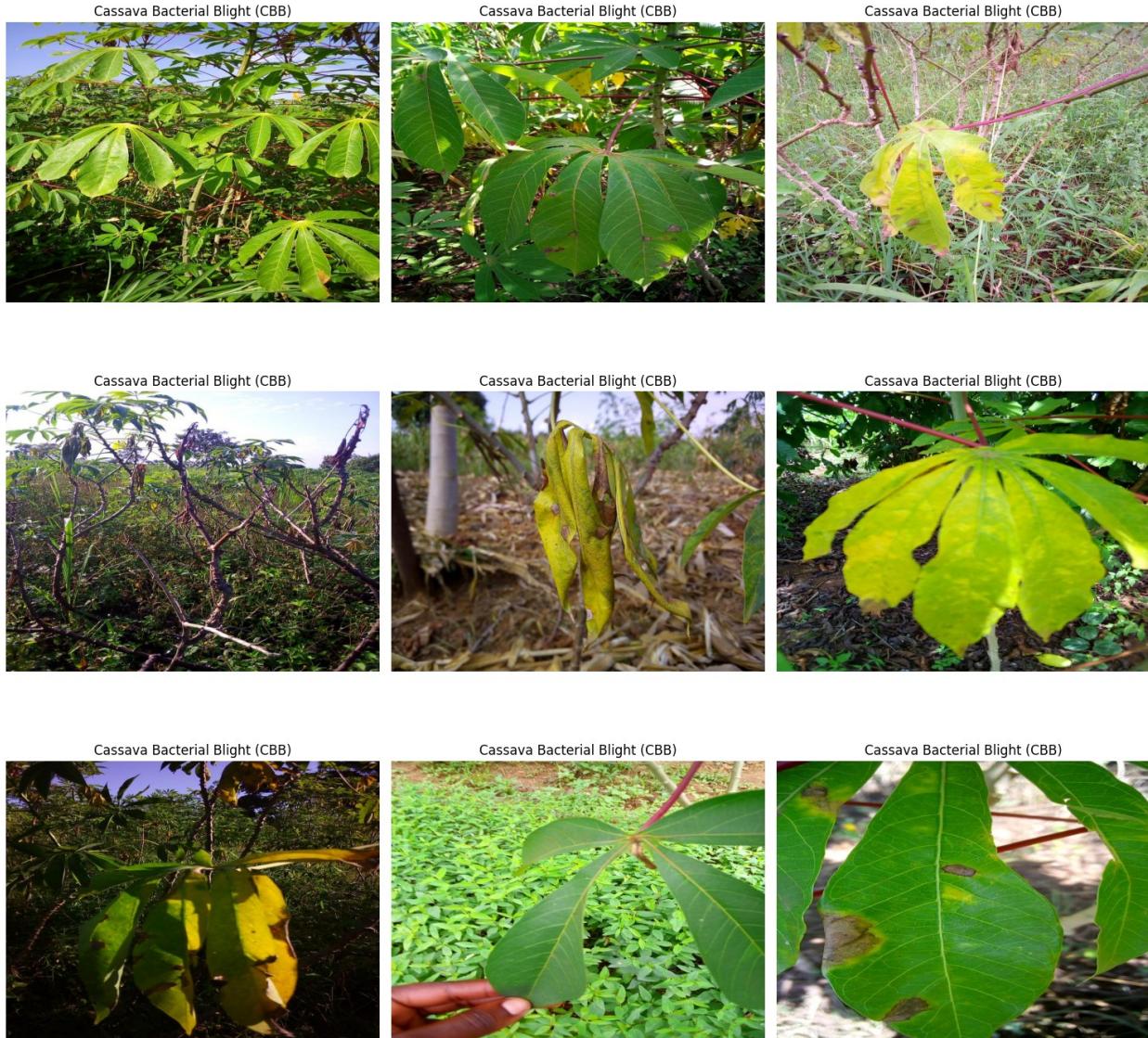
0 1 2 3 4 5

```
plot_images_from_class(0)
```



```
# Step 9: Visualize sample images for each class (0 to 4)
for i in range(5):
    print(f"Displaying sample images for class: {map_classes[i]}")
    plot_images_from_class(i)
```

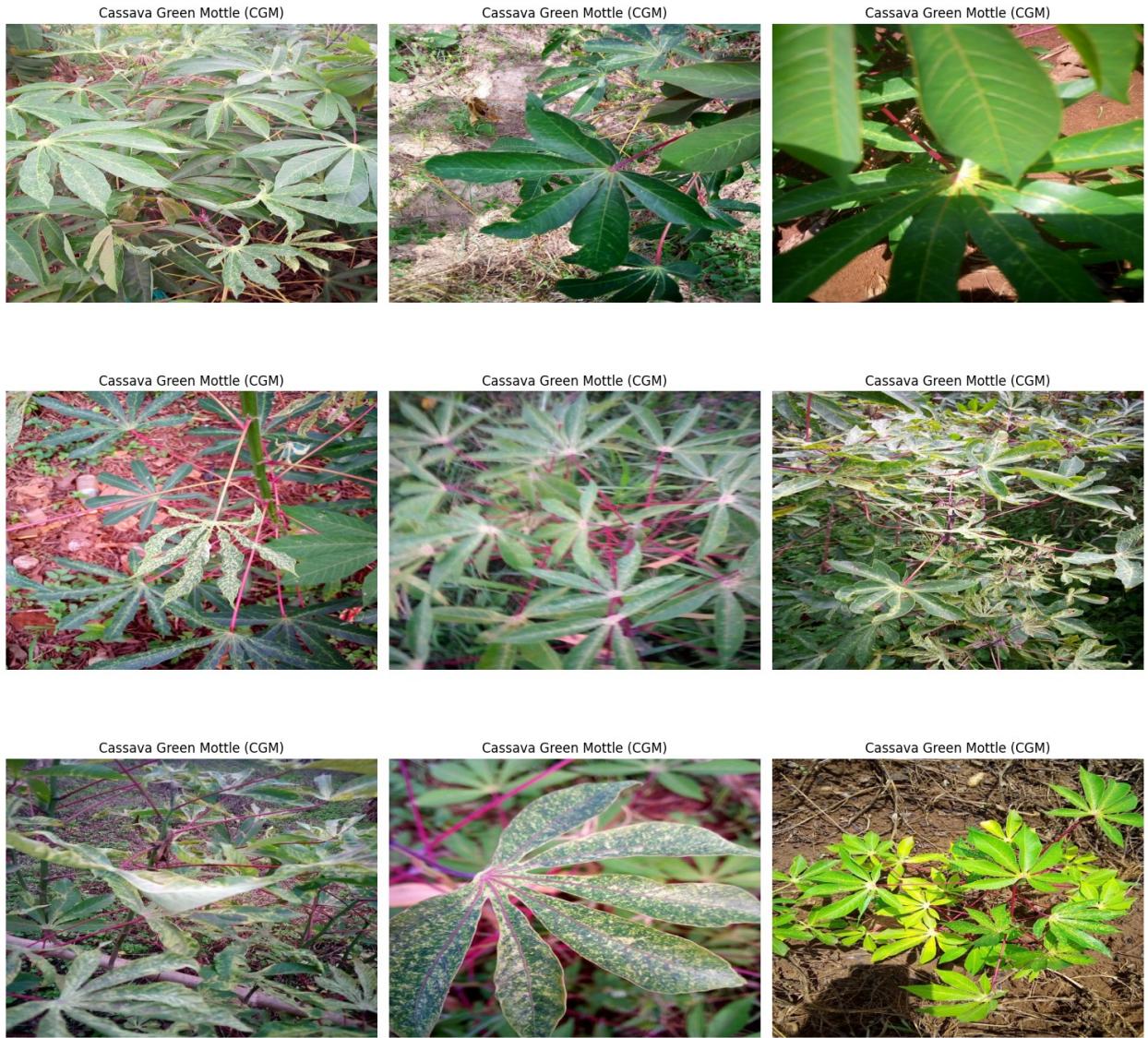
```
Displaying sample images for class: Cassava Bacterial Blight (CBB)
```



Displaying sample images for class: Cassava Brown Streak Disease (CBSD)



Displaying sample images for class: Cassava Green Mottle (CGM)



Displaying sample images for class: Cassava Mosaic Disease (CMD)



Displaying sample images for class: Healthy



we are leaf doctors

```
# Step 10: Check image shapes for the entire dataset
df_train['image_shape'] = df_train['image_id'].apply(lambda x:
cv2.imread(os.path.join(BASE_DIR, "train_images", x)).shape)

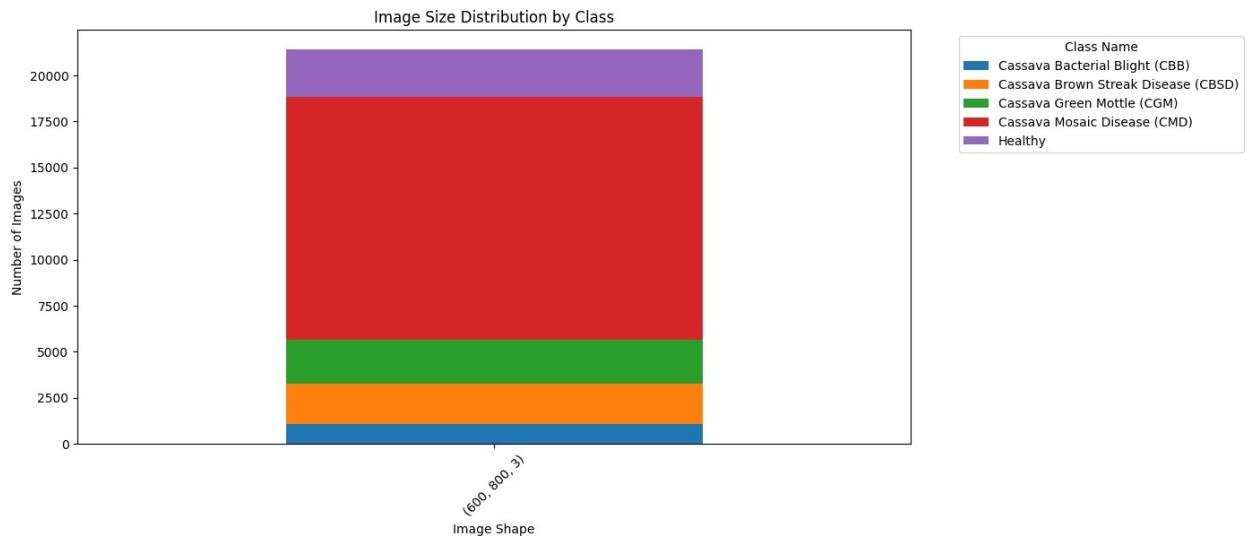
# Group by class and image shape to see if there's a pattern in image
# size by disease type
shape_class_dist = df_train.groupby(['class_name',
'image_shape']).size().unstack(fill_value=0)

# Plot the distribution of image sizes for each class
shape_class_dist.T.plot(kind='bar', stacked=True, figsize=(12, 6))
plt.title('Image Size Distribution by Class')
plt.xlabel('Image Shape')
plt.ylabel('Number of Images')
```

```

plt.legend(title='Class Name', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.xticks(rotation=45)
plt.show()

```



Step 1 : Analyzing the Patient¶ What would be a first step that a Leaf doctor should do before anything considering the fact that his client can't speak ? The answer is simple right , analyze what's wrong by looking at the patient

But how does a doctor understand if something is wrong by just looking at it? For this as a doctor , he should know what a normal Patient/Leaf looks like and observe deviations (in pattern ,color, texture,etc) from the normal behavior to separate the healthy patients from infected ones . Now to further classify the infected ones into specific class of diseases doctor should also know how the patient/leaf condition looks like in different diseases

With these pointers in mind let's start with basic familiarity

This can help you understand if certain classes are more commonly associated with specific image sizes, which might indicate that certain images were taken under different conditions or with different devices.

```

df_train.head()

      image_id  label           class_name
image_shape
0  1000015157.jpg    0  Cassava Bacterial Blight (CBB)  (600,
800, 3)
1  1000201771.jpg    3  Cassava Mosaic Disease (CMD)  (600,
800, 3)
2  100042118.jpg     1  Cassava Brown Streak Disease (CBSD)  (600,
800, 3)
3  1000723321.jpg     1  Cassava Brown Streak Disease (CBSD)  (600,
800, 3)

```

```
4 1000812911.jpg      3          Cassava Mosaic Disease (CMD) (600,  
800, 3)  
df_train['label_names'] = df_train['label'].map(map_classes)
```

## Duplicate Images

```
# Finding Duplicate Images  
# Identifying Exact Duplicate Images  
  
def get_image_hash(image_path):  
    """Generate an MD5 hash for the image."""  
    with open(image_path, "rb") as f:  
        file_hash = hashlib.md5(f.read()).hexdigest()  
    return file_hash  
  
# Dictionary to store image hashes and their file names  
image_hashes = {}  
  
# Check for duplicate images  
duplicate_images = []  
  
for image_name in os.listdir(os.path.join(BASE_DIR, "train_images")):  
    image_path = os.path.join(BASE_DIR, "train_images", image_name)  
    image_hash = get_image_hash(image_path)  
  
    if image_hash in image_hashes:  
        duplicate_images.append((image_name,  
image_hashes[image_hash])) # Image is a duplicate  
    else:  
        image_hashes[image_hash] = image_name # Store the hash  
  
print(f"Found {len(duplicate_images)} exact duplicate images.")  
for dup in duplicate_images:  
    print(f"Duplicate pair: {dup[0]} and {dup[1]}")  
  
Found 0 exact duplicate images.  
  
len(image_hashes)  
21397
```

# Model -- 1: Let's train Base Model

```
import tensorflow as tf
from tensorflow.keras import layers, models

# Define the CNN model architecture
model = models.Sequential()

# 1st Convolutional block
model.add(layers.Conv2D(32, (3, 3), activation='relu',
input_shape=(224, 224, 3)))
model.add(layers.BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))

# 2nd Convolutional block
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))

# 3rd Convolutional block
model.add(layers.Conv2D(128, (3, 3), activation='relu'))
model.add(layers.BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))

# 4th Convolutional block
model.add(layers.Conv2D(256, (3, 3), activation='relu'))
model.add(layers.BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))

# Flatten the output to feed it into fully connected layers
model.add(layers.Flatten())

# Dense layer with 512 units
model.add(layers.Dense(512, activation='relu'))
model.add(layers.BatchNormalization())
model.add(layers.Dropout(0.5))

# Output layer for classification (assuming 5 classes)
model.add(layers.Dense(5, activation='softmax'))

# Compile the model
model.compile(optimizer='adam',
              loss='categorical_crossentropy', # Assuming sparse
              labels (integers)
              metrics=['accuracy'])

# Model summary
model.summary()
```

```
/opt/conda/lib/python3.10/site-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 #	
conv2d (Conv2D) 896	(None, 222, 222, 32)
batch_normalization 128 (BatchNormalization)	(None, 222, 222, 32)
max_pooling2d (MaxPooling2D) 0	(None, 111, 111, 32)
conv2d_1 (Conv2D) 18,496	(None, 109, 109, 64)
batch_normalization_1 256 (BatchNormalization)	(None, 109, 109, 64)
max_pooling2d_1 (MaxPooling2D) 0	(None, 54, 54, 64)
conv2d_2 (Conv2D) 73,856	(None, 52, 52, 128)
batch_normalization_2 512	(None, 52, 52, 128)

	(BatchNormalization)		
0	max_pooling2d_2 (MaxPooling2D)	(None, 26, 26, 128)	
295,168	conv2d_3 (Conv2D)	(None, 24, 24, 256)	
1,024	batch_normalization_3 (BatchNormalization)	(None, 24, 24, 256)	
0	max_pooling2d_3 (MaxPooling2D)	(None, 12, 12, 256)	
0	flatten (Flatten)	(None, 36864)	
18,874,880	dense (Dense)	(None, 512)	
2,048	batch_normalization_4 (BatchNormalization)	(None, 512)	
0	dropout (Dropout)	(None, 512)	
2,565	dense_1 (Dense)	(None, 5)	

Total params: 19,269,829 (73.51 MB)

Trainable params: 19,267,845 (73.50 MB)

Non-trainable params: 1,984 (7.75 KB)

```

df_train.head()

      image_id  label          class_name
image_shape \
0  1000015157.jpg     0  Cassava Bacterial Blight (CBB) (600,
800, 3)
1  1000201771.jpg     3  Cassava Mosaic Disease (CMD) (600,
800, 3)
2  100042118.jpg      1  Cassava Brown Streak Disease (CBSD) (600,
800, 3)
3  1000723321.jpg      1  Cassava Brown Streak Disease (CBSD) (600,
800, 3)
4  1000812911.jpg      3  Cassava Mosaic Disease (CMD) (600,
800, 3)

      label_names
0  Cassava Bacterial Blight (CBB)
1  Cassava Mosaic Disease (CMD)
2  Cassava Brown Streak Disease (CBSD)
3  Cassava Brown Streak Disease (CBSD)
4  Cassava Mosaic Disease (CMD)

```

## Deprecated Techniques.

```

images_dir =
'/kaggle/input/cassava-leaf-disease-classification/train_images'

from tensorflow.keras.preprocessing.image import ImageDataGenerator
# Create an ImageDataGenerator to preprocess images
datagen = ImageDataGenerator(rescale=1./255,
                             validation_split=0.2) # Normalize pixel
values

images_dir =
'/kaggle/input/cassava-leaf-disease-classification/train_images'

# Generate training and validation datasets
train_dataset = datagen.flow_from_dataframe(
    dataframe=df_train,
    directory=images_dir,
    x_col="image_id", # Assuming the image path column is named
"image_path"
    y_col="class_name",
    target_size=(224, 224), # Adjust image size as needed
    batch_size=32,
    class_mode="categorical",
    subset="training",
    shuffle=True

```

```

)
val_dataset = datagen.flow_from_dataframe(
    dataframe=df_train,
    directory=images_dir,
    x_col="image_id",
    y_col="class_name",
    target_size=(224, 224),
    batch_size=32,
    class_mode="categorical",
    subset="validation",
    shuffle=True
)

Found 17118 validated image filenames belonging to 5 classes.
Found 4279 validated image filenames belonging to 5 classes.

# Train the model for 10 epochs
history = model.fit(
    train_dataset, # Assumed preprocessed training dataset
    validation_data=val_dataset, # Assumed preprocessed validation
    dataset
    epochs=1
)

/opt/conda/lib/python3.10/site-packages/keras/src/trainers/
data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset`'
class should call `super().__init__(**kwargs)` in its constructor.
`**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will
be ignored.
    self._warn_if_super_not_called()
WARNING: All log messages before absl::InitializeLog() is called are
written to STDERR
I0000 00:00:1726408133.649689    184 service.cc:145] XLA service
0x7bbbfc0165b0 initialized for platform CUDA (this does not guarantee
that XLA will be used). Devices:
I0000 00:00:1726408133.649758    184 service.cc:153] StreamExecutor
device (0): Tesla T4, Compute Capability 7.5
I0000 00:00:1726408133.649763    184 service.cc:153] StreamExecutor
device (1): Tesla T4, Compute Capability 7.5

2/535 ━━━━━━━━━━ 43s 81ms/step - accuracy: 0.2734 - loss:
3.4142

I0000 00:00:1726408142.126726    184 device_compiler.h:188] Compiled
cluster using XLA! This line is logged at most once for the lifetime
of the process.

535/535 ━━━━━━━━━━ 122s 204ms/step - accuracy: 0.4949 -
loss: 1.6661 - val_accuracy: 0.6244 - val_loss: 1.0734

```

```

# Train the model for 10 epochs
history = model.fit(
    train_dataset, # Assumed preprocessed training dataset
    validation_data=val_dataset, # Assumed preprocessed validation
    dataset
    epochs=10
)

Epoch 1/10

/opt/conda/lib/python3.10/site-packages/keras/src/trainers/
data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class
should call `super().__init__(**kwargs)` in its constructor.
`**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will
be ignored.
    self._warn_if_super_not_called()
WARNING: All log messages before absl::InitializeLog() is called are
written to STDERR
I0000 00:00:1726310729.097077    172 service.cc:145] XLA service
0x7f9b812d7e30 initialized for platform CUDA (this does not guarantee
that XLA will be used). Devices:
I0000 00:00:1726310729.097144    172 service.cc:153] StreamExecutor
device (0): Tesla T4, Compute Capability 7.5
I0000 00:00:1726310729.097150    172 service.cc:153] StreamExecutor
device (1): Tesla T4, Compute Capability 7.5

2/535 ━━━━━━━━━━ 42s 80ms/step - accuracy: 0.2578 - loss:
2.8330

I0000 00:00:1726310738.154399    172 device_compiler.h:188] Compiled
cluster using XLA! This line is logged at most once for the lifetime
of the process.

535/535 ━━━━━━━━━━ 122s 203ms/step - accuracy: 0.4931 -
loss: 1.6367 - val_accuracy: 0.1358 - val_loss: 3.2566
Epoch 2/10
535/535 ━━━━━━━━━━ 98s 180ms/step - accuracy: 0.6455 - loss:
0.9746 - val_accuracy: 0.4288 - val_loss: 1.3975
Epoch 3/10
535/535 ━━━━━━━━━━ 98s 182ms/step - accuracy: 0.6913 - loss:
0.8239 - val_accuracy: 0.6413 - val_loss: 1.0349
Epoch 4/10
535/535 ━━━━━━━━━━ 97s 180ms/step - accuracy: 0.7269 - loss:
0.7433 - val_accuracy: 0.7097 - val_loss: 0.7598
Epoch 5/10
535/535 ━━━━━━━━━━ 139s 174ms/step - accuracy: 0.7536 -
loss: 0.6656 - val_accuracy: 0.6586 - val_loss: 1.1666
Epoch 6/10
535/535 ━━━━━━━━━━ 98s 180ms/step - accuracy: 0.7515 - loss:

```

```
0.6782 - val_accuracy: 0.7062 - val_loss: 0.8929
Epoch 7/10
535/535 ━━━━━━━━ 97s 179ms/step - accuracy: 0.8047 - loss:
0.5223 - val_accuracy: 0.7184 - val_loss: 0.7765
Epoch 8/10
535/535 ━━━━━━ 96s 178ms/step - accuracy: 0.7986 - loss:
0.5452 - val_accuracy: 0.6284 - val_loss: 1.2401
Epoch 9/10
535/535 ━━━━━━ 96s 177ms/step - accuracy: 0.8219 - loss:
0.4828 - val_accuracy: 0.7214 - val_loss: 0.8700
Epoch 10/10
535/535 ━━━━━━ 95s 175ms/step - accuracy: 0.9100 - loss:
0.2516 - val_accuracy: 0.5901 - val_loss: 1.3391
```

## CONCULTIOSN - Model overfitting

Complete the training for more epochs 20

Do model plots

Confusiton matrix, multi class

Model Evaluation.

## Method 2 -- Using Tensorlfow Dataset

```
df_train.head()

          image_id  label \
0 /kaggle/input/cassava-leaf-disease-classificat...      0
1 /kaggle/input/cassava-leaf-disease-classificat...      3
2 /kaggle/input/cassava-leaf-disease-classificat...      1
3 /kaggle/input/cassava-leaf-disease-classificat...      1
4 /kaggle/input/cassava-leaf-disease-classificat...      3

          class_name    image_shape \
0   Cassava Bacterial Blight (CBB)  (600, 800, 3)
1   Cassava Mosaic Disease (CMD)  (600, 800, 3)
```

```

2 Cassava Brown Streak Disease (CBSD) (600, 800, 3)
3 Cassava Brown Streak Disease (CBSD) (600, 800, 3)
4 Cassava Mosaic Disease (CMD) (600, 800, 3)

label_names
0 Cassava Bacterial Blight (CBB)
1 Cassava Mosaic Disease (CMD)
2 Cassava Brown Streak Disease (CBSD)
3 Cassava Brown Streak Disease (CBSD)
4 Cassava Mosaic Disease (CMD)

# Option 1: Using pandas' vectorized string operations
df_train['image_id'] = "/kaggle/input/cassava-leaf-disease-
classification/train_images/" + df_train['image_id']
df_train.head()

image_id    label \
0 /kaggle/input/cassava-leaf-disease-classificat... 0
1 /kaggle/input/cassava-leaf-disease-classificat... 3
2 /kaggle/input/cassava-leaf-disease-classificat... 1
3 /kaggle/input/cassava-leaf-disease-classificat... 1
4 /kaggle/input/cassava-leaf-disease-classificat... 3

class_name    image_shape \
0 Cassava Bacterial Blight (CBB) (600, 800, 3)
1 Cassava Mosaic Disease (CMD) (600, 800, 3)
2 Cassava Brown Streak Disease (CBSD) (600, 800, 3)
3 Cassava Brown Streak Disease (CBSD) (600, 800, 3)
4 Cassava Mosaic Disease (CMD) (600, 800, 3)

label_names
0 Cassava Bacterial Blight (CBB)
1 Cassava Mosaic Disease (CMD)
2 Cassava Brown Streak Disease (CBSD)
3 Cassava Brown Streak Disease (CBSD)
4 Cassava Mosaic Disease (CMD)

from sklearn.model_selection import train_test_split

# Define the validation split ratio
VALIDATION_SPLIT = 0.2 # 20% for validation

# Perform stratified split to maintain class distribution
train_df, val_df = train_test_split(
    df_train,
    test_size=VALIDATION_SPLIT,
    stratify=df_train['label'],
    random_state=42
)

```

```
print(f"Training samples: {len(train_df)}")
print(f"Validation samples: {len(val_df)}")

Training samples: 17117
Validation samples: 4280

print("Training class distribution:")
print(train_df['label'].value_counts())

print("\nValidation class distribution:")
print(val_df['label'].value_counts())

Training class distribution:
label
3    10526
4     2061
2     1909
1     1751
0      870
Name: count, dtype: int64

Validation class distribution:
label
3    2632
4      516
2      477
1      438
0      217
Name: count, dtype: int64

# Image dimensions
IMG_HEIGHT = 224
IMG_WIDTH = 224
CHANNELS = 3

# Batch size
BATCH_SIZE = 32

# Buffer size for shuffling
BUFFER_SIZE = 1000

# AUTOTUNE for performance optimization
AUTOTUNE = tf.data.AUTOTUNE

def process_image(file_path, label):
    # Read the image from disk
    image = tf.io.read_file(file_path)

    # Decode the image (assuming JPEG format)
    image = tf.image.decode_jpeg(image, channels=CHANNELS)
```

```

# Resize the image
image = tf.image.resize(image, [IMG_HEIGHT, IMG_WIDTH])

# Normalize pixel values to [0,1]
image = image / 255.0

return image, label

df_train.head()

          image_id  label \
0 /kaggle/input/cassava-leaf-disease-classificat...    0
1 /kaggle/input/cassava-leaf-disease-classificat...    3
2 /kaggle/input/cassava-leaf-disease-classificat...    1
3 /kaggle/input/cassava-leaf-disease-classificat...    1
4 /kaggle/input/cassava-leaf-disease-classificat...    3

          class_name   image_shape \
0  Cassava Bacterial Blight (CBB)  (600, 800, 3)
1  Cassava Mosaic Disease (CMD)  (600, 800, 3)
2  Cassava Brown Streak Disease (CBSD)  (600, 800, 3)
3  Cassava Brown Streak Disease (CBSD)  (600, 800, 3)
4  Cassava Mosaic Disease (CMD)  (600, 800, 3)

          label_names
0  Cassava Bacterial Blight (CBB)
1  Cassava Mosaic Disease (CMD)
2  Cassava Brown Streak Disease (CBSD)
3  Cassava Brown Streak Disease (CBSD)
4  Cassava Mosaic Disease (CMD)

# Create TensorFlow Dataset from training DataFrame
train_ds =
tf.data.Dataset.from_tensor_slices((train_df['image_id'].values,
train_df['label'].values))

# Map the processing function to each (image, label) pair
train_ds = train_ds.map(process_image, num_parallel_calls=AUTOTUNE)

# Data Augmentation (optional but recommended)
data_augmentation = tf.keras.Sequential([
    layers.RandomFlip('horizontal'),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.18),
    layers.RandomContrast(0.2),
])

def augment(image, label):
    image = data_augmentation(image)
    return image, label

```

```

train_ds = train_ds.map(augment, num_parallel_calls=AUTOTUNE)

# Shuffle, batch, and prefetch
train_ds =
train_ds.shuffle(BUFFER_SIZE).batch(BATCH_SIZE).prefetch(AUTOTUNE)

# Create TensorFlow Dataset from validation DataFrame
val_ds =
tf.data.Dataset.from_tensor_slices((val_df['image_id'].values,
val_df['label'].values))

# Map the processing function
val_ds = val_ds.map(process_image, num_parallel_calls=AUTOTUNE)

# Batch and prefetch
val_ds = val_ds.batch(BATCH_SIZE).prefetch(AUTOTUNE)

from tensorflow.keras import regularizers

num_classes = 5
# Define the CNN model architecture
def create_cnn_model(input_shape=(IMG_HEIGHT, IMG_WIDTH, CHANNELS),
num_classes=5):
    model = models.Sequential([
        # 1st Convolutional block
        layers.Input(shape=input_shape),
        layers.Conv2D(32, (3, 3), activation='relu', padding='same'),
        layers.BatchNormalization(),
        layers.MaxPooling2D((2, 2)),

        # 2nd Convolutional block
        layers.Conv2D(64, (3, 3), activation='relu', padding='same'),
        layers.BatchNormalization(),
        layers.MaxPooling2D((2, 2)),

        # 3rd Convolutional block
        layers.Conv2D(128, (3, 3), activation='relu', padding='same'),
        layers.BatchNormalization(),
        layers.MaxPooling2D((2, 2)),

        # 4th Convolutional block
        layers.Conv2D(256, (3, 3), activation='relu', padding='same'),
        layers.BatchNormalization(),
        layers.MaxPooling2D((2, 2)),

        # Flatten and Dense layers
        layers.Flatten(),
        layers.Dense(512, activation='relu'),
        layers.BatchNormalization(),
        layers.Dropout(0.5),
    ])

```

```

        # Output layer
        layers.Dense(5, activation='softmax')
    ])

return model

# Instantiate the model
model = create_cnn_model(input_shape=(IMG_HEIGHT, IMG_WIDTH,
CHANNELS), num_classes=num_classes)

# Display the model architecture
model.summary()

Model: "sequential_3"

```

Layer (type)	Output Shape
Param #	
conv2d_8 (Conv2D) 896	(None, 224, 224, 32)
batch_normalization_10 128 (BatchNormalization)	(None, 224, 224, 32)
max_pooling2d_8 (MaxPooling2D) 0	(None, 112, 112, 32)
conv2d_9 (Conv2D) 18,496	(None, 112, 112, 64)
batch_normalization_11 256 (BatchNormalization)	(None, 112, 112, 64)
max_pooling2d_9 (MaxPooling2D) 0	(None, 56, 56, 64)

	conv2d_10 (Conv2D)	(None, 56, 56, 128)	
73,856			
	batch_normalization_12 (BatchNormalization)	(None, 56, 56, 128)	
512			
	max_pooling2d_10 (MaxPooling2D)	(None, 28, 28, 128)	
0			
	conv2d_11 (Conv2D)	(None, 28, 28, 256)	
295,168			
	batch_normalization_13 (BatchNormalization)	(None, 28, 28, 256)	
1,024			
	max_pooling2d_11 (MaxPooling2D)	(None, 14, 14, 256)	
0			
	flatten_2 (Flatten)	(None, 50176)	
0			
	dense_4 (Dense)	(None, 512)	
25,690,624			
	batch_normalization_14 (BatchNormalization)	(None, 512)	
2,048			
	dropout_2 (Dropout)	(None, 512)	
0			
	dense_5 (Dense)	(None, 5)	
2,565			

```
Total params: 26,085,573 (99.51 MB)
Trainable params: 26,083,589 (99.50 MB)
Non-trainable params: 1,984 (7.75 KB)

# Compile the model
model.compile(
    optimizer='adam', # You can experiment with different optimizers
    loss='categorical_crossentropy', # Suitable for integer-encoded
    labels
    metrics=['accuracy']
)

from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint,
ReduceLROnPlateau

# Early stopping to prevent overfitting
early_stop = EarlyStopping(
    monitor='val_loss',
    patience=3,
    restore_best_weights=True,
    verbose=1
)

# Model checkpoint to save the best model
checkpoint = ModelCheckpoint(
    'best_cnn_model.keras',
    monitor='val_accuracy',
    save_best_only=True,
    verbose=1
)

# Reduce learning rate when a metric has stopped improving
reduce_lr = ReduceLROnPlateau(
    monitor='val_loss',
    factor=0.2,
    patience=3,
    verbose=1,
    min_lr=1e-6
)

# Combine callbacks
callbacks = [early_stop, checkpoint, reduce_lr]

val_dataset
```

```
<keras.src.legacy.preprocessing.image.DataFrameIterator at
0x7bbd305ec250>

# Train the model for 10 epochs
history = model.fit(
    train_dataset, # Assumed preprocessed training dataset
    validation_data=val_dataset, # Assumed preprocessed validation
    dataset
    epochs=1
)

535/535 ━━━━━━━━━━ 148s 253ms/step - accuracy: 0.4954 -
loss: 1.6679 - val_accuracy: 0.6392 - val_loss: 1.0716

# Define the number of epochs
EPOCHS = 10 # Adjust based on your requirements

# Train the model
history = model.fit(
    train_ds,
    validation_data=val_ds,
    epochs=EPOCHS,
    callbacks=callbacks
)

Epoch 1/10
535/535 ━━━━━━━━━━ 0s 258ms/step - accuracy: 0.4856 - loss:
1.6838
Epoch 1: val_accuracy improved from -inf to 0.38131, saving model to
best_cnn_model.keras
535/535 ━━━━━━━━━━ 165s 277ms/step - accuracy: 0.4857 -
loss: 1.6832 - val_accuracy: 0.3813 - val_loss: 1.5359 -
learning_rate: 0.0010
Epoch 2/10
535/535 ━━━━━━━━━━ 0s 233ms/step - accuracy: 0.6354 - loss:
1.0151
Epoch 2: val_accuracy improved from 0.38131 to 0.61355, saving model
to best_cnn_model.keras
535/535 ━━━━━━━━━━ 142s 255ms/step - accuracy: 0.6354 -
loss: 1.0150 - val_accuracy: 0.6136 - val_loss: 1.0416 -
learning_rate: 0.0010
Epoch 3/10
535/535 ━━━━━━━━━━ 0s 232ms/step - accuracy: 0.6684 - loss:
0.8863
Epoch 3: val_accuracy improved from 0.61355 to 0.65678, saving model
to best_cnn_model.keras
535/535 ━━━━━━━━━━ 137s 246ms/step - accuracy: 0.6684 -
loss: 0.8863 - val_accuracy: 0.6568 - val_loss: 0.9554 -
learning_rate: 0.0010
Epoch 4/10
```

```
535/535 ━━━━━━━━ 0s 233ms/step - accuracy: 0.6813 - loss:  
0.8348  
Epoch 4: val_accuracy did not improve from 0.65678  
535/535 ━━━━━━━━ 136s 244ms/step - accuracy: 0.6813 -  
loss: 0.8348 - val_accuracy: 0.5591 - val_loss: 1.2806 -  
learning_rate: 0.0010  
Epoch 5/10  
535/535 ━━━━━━━━ 0s 232ms/step - accuracy: 0.6964 - loss:  
0.7986  
Epoch 5: val_accuracy improved from 0.65678 to 0.70374, saving model  
to best_cnn_model.keras  
535/535 ━━━━━━━━ 137s 246ms/step - accuracy: 0.6964 -  
loss: 0.7986 - val_accuracy: 0.7037 - val_loss: 0.7824 -  
learning_rate: 0.0010  
Epoch 6/10  
535/535 ━━━━━━━━ 0s 232ms/step - accuracy: 0.7239 - loss:  
0.7384  
Epoch 6: val_accuracy did not improve from 0.70374  
535/535 ━━━━━━━━ 138s 246ms/step - accuracy: 0.7239 -  
loss: 0.7384 - val_accuracy: 0.5614 - val_loss: 1.1939 -  
learning_rate: 0.0010  
Epoch 7/10  
535/535 ━━━━━━━━ 0s 232ms/step - accuracy: 0.7412 - loss:  
0.7004  
Epoch 7: val_accuracy did not improve from 0.70374  
535/535 ━━━━━━━━ 135s 243ms/step - accuracy: 0.7412 -  
loss: 0.7004 - val_accuracy: 0.6881 - val_loss: 0.8473 -  
learning_rate: 0.0010  
Epoch 8/10  
535/535 ━━━━━━━━ 0s 232ms/step - accuracy: 0.7599 - loss:  
0.6557  
Epoch 8: val_accuracy improved from 0.70374 to 0.73084, saving model  
to best_cnn_model.keras  
535/535 ━━━━━━━━ 137s 246ms/step - accuracy: 0.7599 -  
loss: 0.6557 - val_accuracy: 0.7308 - val_loss: 0.7404 -  
learning_rate: 0.0010  
Epoch 9/10  
535/535 ━━━━━━━━ 0s 232ms/step - accuracy: 0.7609 - loss:  
0.6630  
Epoch 9: val_accuracy did not improve from 0.73084  
535/535 ━━━━━━━━ 140s 251ms/step - accuracy: 0.7609 -  
loss: 0.6630 - val_accuracy: 0.2318 - val_loss: 1.9677 -  
learning_rate: 0.0010  
Epoch 10/10  
535/535 ━━━━━━━━ 0s 230ms/step - accuracy: 0.7354 - loss:  
0.7071  
Epoch 10: val_accuracy did not improve from 0.73084  
535/535 ━━━━━━━━ 135s 241ms/step - accuracy: 0.7354 -  
loss: 0.7071 - val_accuracy: 0.6563 - val_loss: 0.9383 -
```

```
learning_rate: 0.0010
Restoring model weights from the end of the best epoch: 8.

# Step 14: Evaluate the Model
val_loss, val_accuracy = model.evaluate(val_ds)
print(f"\nValidation Loss: {val_loss:.4f}")
print(f"Validation Accuracy: {val_accuracy:.4f}")

134/134 ━━━━━━━━━━━━ 6s 46ms/step - accuracy: 0.7293 - loss:
0.7448

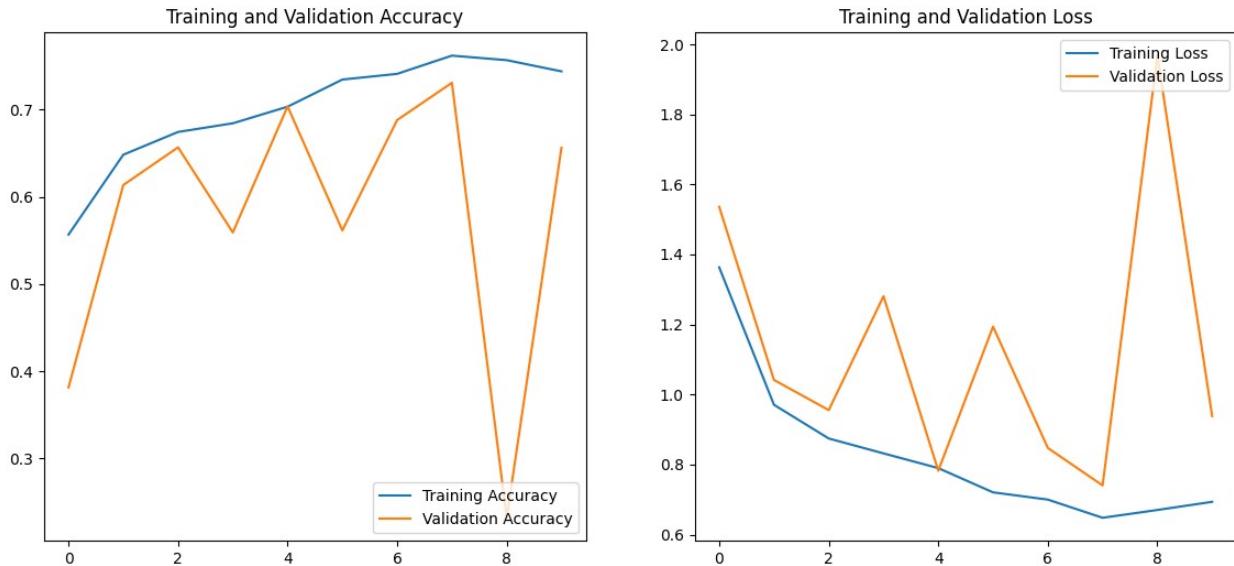
Validation Loss: 0.7404
Validation Accuracy: 0.7308

acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs_range = range(len(acc))

# Plot Accuracy
plt.figure(figsize=(14, 6))
plt.subplot(1, 2, 1)
plt.plot(epochs_range, acc, label='Training Accuracy')
plt.plot(epochs_range, val_acc, label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')

# Plot Loss
plt.subplot(1, 2, 2)
plt.plot(epochs_range, loss, label='Training Loss')
plt.plot(epochs_range, val_loss, label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')

plt.show()
```



## Visualising the augmentations

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator

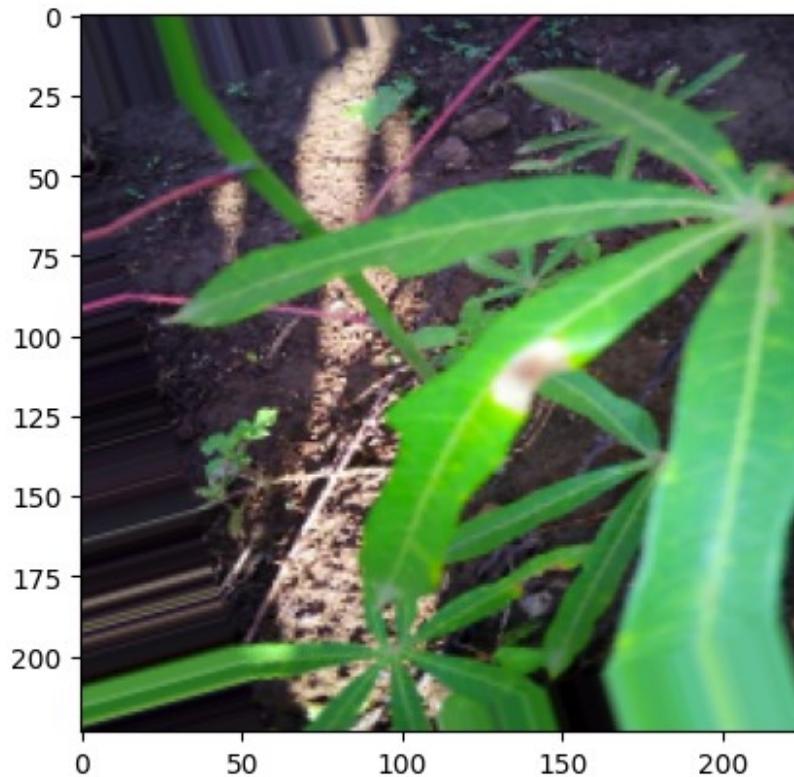
# Define your data augmentation generator
datagen = ImageDataGenerator(
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom_range=0.2,
    horizontal_flip=True,
    fill_mode='nearest'
)

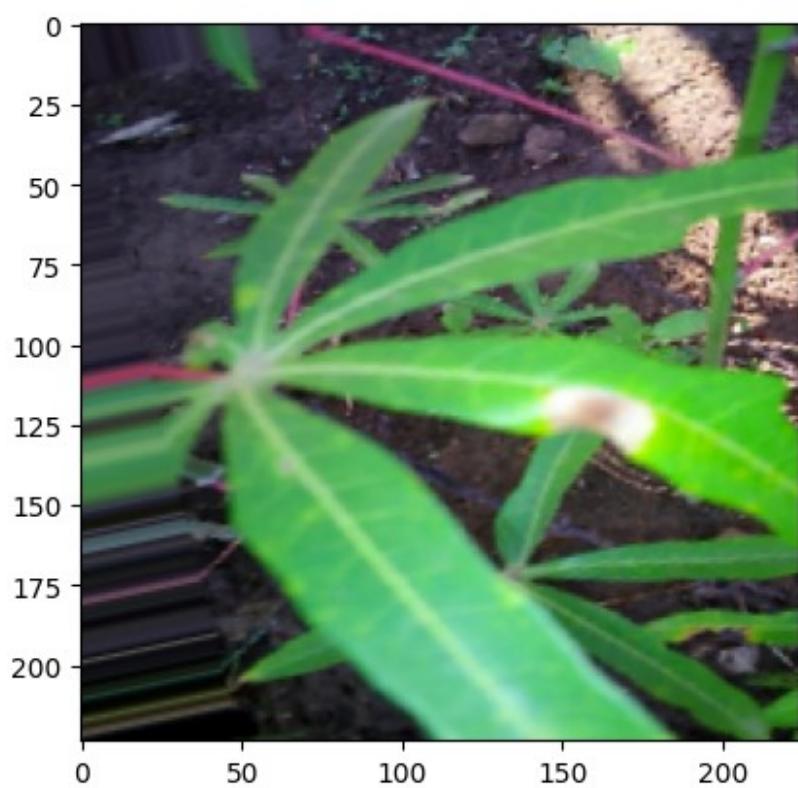
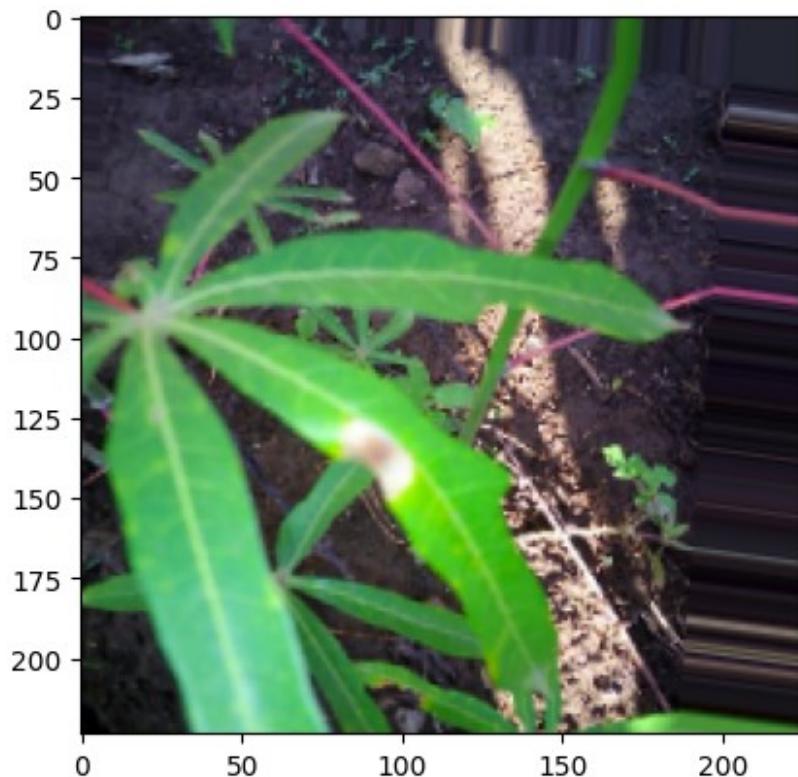
# Example usage on one image
from tensorflow.keras.preprocessing import image

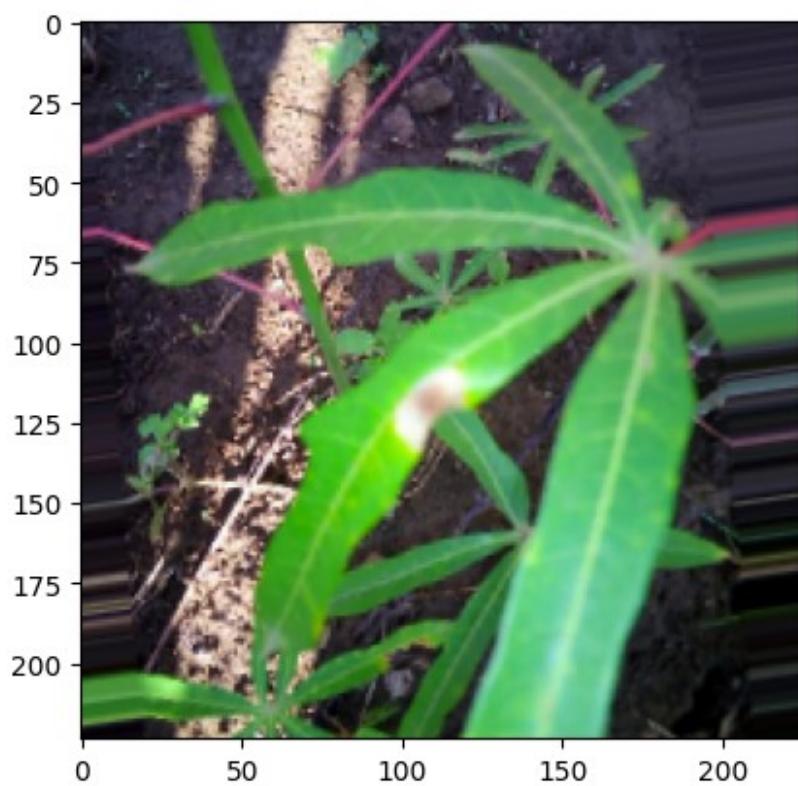
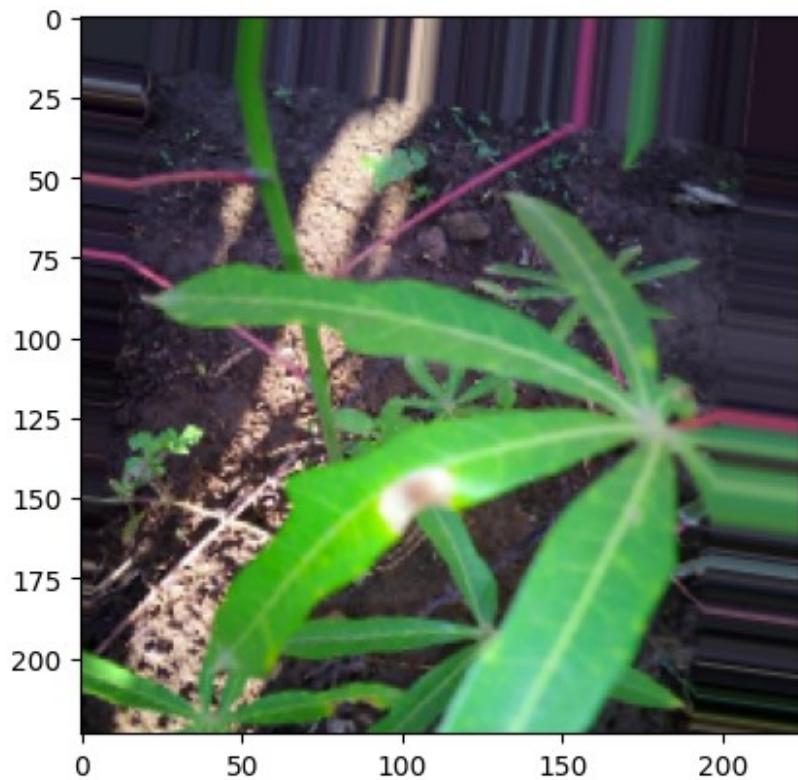
img = image.load_img('/kaggle/input/cassava-leaf-disease-
classification/train_images/1001320321.jpg',
                     target_size=(224, 224))
x = image.img_to_array(img)
x = x.reshape((1,) + x.shape)

# Generate 5 augmented images
i = 0
for batch in datagen.flow(x, batch_size=1):
    plt.figure(i)
    imgplot = plt.imshow(image.array_to_img(batch[0]))
    i += 1
    if i % 5 == 0:
```

```
    break  
plt.show()
```







# Understanding class weights

```
# Compute class weights
from sklearn.utils import class_weight
import numpy as np

class_weights = class_weight.compute_class_weight(
    'balanced',
    classes=np.unique(df_train['label']),
    y=df_train['label']
)
class_weights = dict(enumerate(class_weights))
class_weights

{0: 3.9368905243790246,
 1: 1.954956601187757,
 2: 1.7935456831517183,
 3: 0.3252317981456148,
 4: 1.6606131160263873}

model.fit(class_weights = class_weights )
```

# Implementing Focal Loss

```
focal_loss = tfa.losses.SparseCategoricalFocalCrossentropy(
    from_logits=False, # Since the model outputs probabilities with softmax
    alpha=class_weights, # Class weights computed earlier
    gamma=2.0 # Focusing parameter
)

-----
NameError                               Traceback (most recent call
last)
Cell In[72], line 1
----> 1 focal_loss = tfa.losses.SparseCategoricalFocalCrossentropy(
      2     from_logits=False, # Since the model outputs
probabilities with softmax
      3     alpha=class_weights, # Class weights computed earlier
      4     gamma=2.0 # Focusing parameter
      5 )

NameError: name 'tfa' is not defined

model.compile(
    optimizer='adam',
    loss=focal_loss,
```

```

        metrics=['accuracy']
    )

history = model.fit(
    train_ds,
    validation_data=val_ds,
    epochs=10,
    callbacks=callbacks
)

# Step 14: Evaluate the Model
val_loss, val_accuracy = model.evaluate(val_ds)
print(f"\nValidation Loss: {val_loss:.4f}")
print(f"Validation Accuracy: {val_accuracy:.4f}")

# Step 15: Save the Model
model.save('cnn_cassava_leaf_disease_focal_loss.keras')
print("\nModel saved successfully!")

acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs_range = range(len(acc))

# Plot Accuracy
plt.figure(figsize=(14, 6))
plt.subplot(1, 2, 1)
plt.plot(epochs_range, acc, label='Training Accuracy')
plt.plot(epochs_range, val_acc, label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')

# Plot Loss
plt.subplot(1, 2, 2)
plt.plot(epochs_range, loss, label='Training Loss')
plt.plot(epochs_range, val_loss, label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')

plt.show()

```

# Implementing Transfer Learning RESNET

Home work fine tuning.

Model training, plotting, evalution, confusion matirix.

```
from tensorflow.keras.applications import ResNet50

# Efficient net

# Load the ResNet50 model without the top classification layers
base_model = ResNet50(
    weights='imagenet',
    include_top=False,
    input_shape=(224, 224, 3)
)

# Freeze the base model
base_model.trainable = False

# Create a new model on top
model = models.Sequential([
    base_model,
    layers.GlobalAveragePooling2D(),
    layers.Dense(256, activation='relu'),
    layers.BatchNormalization(),
    layers.Dense(512, activation='relu'),
    layers.BatchNormalization(),
    layers.Dropout(0.5),
    layers.Dense(5, activation='softmax')
])

# Compile with Focal Loss
model.compile(
    optimizer='adam',
    loss='categorical_crossentropy',
    metrics=['accuracy']
)

# Model summary
model.summary()
```

# Transfer Learning Efficient Net

```
!pip uninstall tensorflow_addons

Found existing installation: tensorflow-addons 0.23.0
Uninstalling tensorflow-addons-0.23.0:
Would remove:
  /opt/conda/lib/python3.10/site-packages/_foo.cpython-310-x86_64-
linux-gnu.so
  /opt/conda/lib/python3.10/site-packages/tensorflow_addons-
0.23.0.dist-info/*
  /opt/conda/lib/python3.10/site-packages/tensorflow_addons/*
Proceed (Y/n)?
```

*# Step 1: Import Necessary Libraries*

```
import os
import json
from collections import Counter

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras import layers, models, regularizers
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint,
ReduceLROnPlateau
# import tensorflow_addons as tfa

from sklearn.model_selection import train_test_split
from sklearn.utils import class_weight


import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import GlobalAveragePooling2D, Flatten,
Dense, Dropout, BatchNormalization
from tensorflow.keras.optimizers import RMSprop, Adam
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint,
ReduceLROnPlateau
from tensorflow.keras.applications import EfficientNetB3
```

*# Step 2: Define Paths*

```
work_dir = '../input/cassava-leaf-disease-classification/' # Update
if necessary
train_path = os.path.join(work_dir, 'train_images') # Path to
training images
train_csv_path = os.path.join(work_dir, 'train.csv') # Path to
```

```

train.csv
label_map_path = os.path.join(work_dir,
'label_num_to_disease_map.json') # Path to label mapping JSON

# a. Load Train Data
data = pd.read_csv(train_csv_path)
print("Initial DataFrame:")
print(data.head())

# Check label frequencies
print("\nLabel Frequencies:")
print(Counter(data['label']))

Initial DataFrame:
   image_id  label
0  1000015157.jpg      0
1  1000201771.jpg      3
2  100042118.jpg       1
3  1000723321.jpg       1
4  1000812911.jpg      3

Label Frequencies:
Counter({3: 13158, 4: 2577, 2: 2386, 1: 2189, 0: 1087})

# b. Load Label Mapping
with open(label_map_path, 'r') as f:
    real_labels = json.load(f)

# Convert string keys to integers
real_labels = {int(k): v for k, v in real_labels.items()}

# Map label numbers to class names
data['class_name'] = data['label'].map(real_labels)
print("\nDataFrame with 'class_name':")
print(data.head())


DataFrame with 'class_name':
   image_id  label          class_name
0  1000015157.jpg      0  Cassava Bacterial Blight (CBB)
1  1000201771.jpg      3  Cassava Mosaic Disease (CMD)
2  100042118.jpg       1  Cassava Brown Streak Disease (CBS)
3  1000723321.jpg       1  Cassava Brown Streak Disease (CBS)
4  1000812911.jpg      3  Cassava Mosaic Disease (CMD)

# Step 4: Compute Class Weights
labels = data['label'].values # Integer-encoded labels

# Compute class weights
class_weights_array = class_weight.compute_class_weight(

```

```

        class_weight='balanced',
        classes=np.unique(labels),
        y=labels
    )

# Create a dictionary mapping class indices to weights
class_weights_dict = {i: weight for i, weight in
enumerate(class_weights_array)}
print("\nClass Weights:")
print(class_weights_dict)

Class Weights:
{0: 3.9368905243790246, 1: 1.954956601187757, 2: 1.7935456831517183,
3: 0.3252317981456148, 4: 1.6606131160263873}

# Step 5: Define ImageDataGenerators
IMG_SIZE = 224
TARGET_SIZE = (IMG_SIZE, IMG_SIZE)
NUM_CLASSES = 5 # Update based on actual number of classes
BATCH_SIZE = 15

datagen_train = ImageDataGenerator(
    preprocessing_function=tf.keras.applications.efficientnet.preprocess_input, # EfficientNet preprocessing
    rotation_range=40,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom_range=0.2,
    horizontal_flip=True,
    vertical_flip=True,
    fill_mode='nearest'
)

# ImageDataGenerator for validation (only normalization)
datagen_val = ImageDataGenerator(
    preprocessing_function=tf.keras.applications.efficientnet.preprocess_input # EfficientNet preprocessing
)

```

Effecient net uses upscaling metod. 512 \* 512

```

train_df, val_df = train_test_split(
    data,
    test_size=0.05, # 5% for validation
    random_state=42,
)

```

```
    stratify=data['class_name']
)
print(f"\nNumber of training samples: {len(train_df)}")
print(f"Number of validation samples: {len(val_df)}")
```

```
Number of training samples: 20327
Number of validation samples: 1070
```

```
train_set = datagen_train.flow_from_dataframe(train_df,
                                              directory = train_path,
                                              seed=42,
                                              x_col = 'image_id',
                                              y_col = 'class_name',
                                              target_size = TARGET_SIZE,
                                              #color_mode="rgb",
                                              class_mode = 'categorical',
                                              interpolation = 'nearest',
                                              shuffle = True,
                                              batch_size = BATCH_SIZE)
```

```
val_set = datagen_val.flow_from_dataframe(val_df,
                                             directory = train_path,
                                             seed=42,
                                             x_col = 'image_id',
                                             y_col = 'class_name',
                                             target_size = TARGET_SIZE,
                                             #color_mode="rgb",
                                             class_mode = 'categorical',
                                             interpolation = 'nearest',
                                             shuffle = True,
                                             batch_size = BATCH_SIZE)
```

```
Found 20327 validated image filenames belonging to 5 classes.
Found 1070 validated image filenames belonging to 5 classes.
```

```
base_model = EfficientNetB3(input_shape=(224, 224, 3),
                           include_top=False,
                           weights='imagenet',
                           drop_connect_rate=0.6)
```

```
base_model.summary()
```

```
Model: "efficientnetb3"
```

Layer (type)	Output Shape	Param #	Connected to

input_layer_10	(None, 224, 224,	0	-
(InputLayer)	3)		
rescaling_18	(None, 224, 224,	0	
input_layer_10[0...]			
(Rescaling)	3)		
normalization_9	(None, 224, 224,	7	
rescaling_18[0][...]			
(Normalization)	3)		
rescaling_19	(None, 224, 224,	0	
normalization_9[...]			
(Rescaling)	3)		
stem_conv_pad	(None, 225, 225,	0	
rescaling_19[0][...]			
(ZeroPadding2D)	3)		
stem_conv (Conv2D)	(None, 112, 112,	1,080	
stem_conv_pad[0]...	40)		
stem_bn [0]	(None, 112, 112,	160	stem_conv[0]
(BatchNormalizatio...	40)		
stem_activation	(None, 112, 112,	0	stem_bn[0][0]
(Activation)	40)		
block1a_dwconv	(None, 112, 112,	360	

stem_activation[...]			
(DepthwiseConv2D)	40)		
block1a_bn	(None, 112, 112,	160	
block1a_dwconv[0...]	(BatchNormalizatio...	40)	
block1a_activation	(None, 112, 112,	0	block1a_bn[0]
[0]	(Activation)	40)	
block1a_se_squeeze	(None, 40)	0	
block1a_activati...	(GlobalAveragePool...		
block1a_se_reshape	(None, 1, 1, 40)	0	
block1a_se_squeee...	(Reshape)		
block1a_se_reduce	(None, 1, 1, 10)	410	
block1a_se_resha...	(Conv2D)		
block1a_se_expand	(None, 1, 1, 40)	440	
block1a_se_reduc...	(Conv2D)		
block1a_se_excite	(None, 112, 112,	0	
block1a_activati...	(Multiply)	40)	
block1a_se_expan...			
block1a_project_co...	(None, 112, 112,	960	
block1a_se_excit...			

(Conv2D)	24)		
block1a_project_bn	(None, 112, 112,	96	
block1a_project_...	(BatchNormalizatio...	24)	
block1b_dwconv	(None, 112, 112,	216	
block1a_project_...	(DepthwiseConv2D)	24)	
block1b_bn	(None, 112, 112,	96	
block1b_dwconv[0...]	(BatchNormalizatio...	24)	
block1b_activation	(None, 112, 112,	0	block1b_bn[0]
[0]	(Activation)	24)	
block1b_se_squeeze	(None, 24)	0	
block1b_activati...	(GlobalAveragePool...		
block1b_se_reshape	(None, 1, 1, 24)	0	
block1b_se_squee...	(Reshape)		
block1b_se_reduce	(None, 1, 1, 6)	150	
block1b_se_resha...	(Conv2D)		
block1b_se_expand	(None, 1, 1, 24)	168	
block1b_se_reduc...	(Conv2D)		

block1b_se_excite	(None, 112, 112,	0	
block1b_activati...			
(Multiply)	24)		
block1b_se_expan...			
block1b_project_co...	(None, 112, 112,	576	
block1b_se_excit...			
(Conv2D)	24)		
block1b_project_bn	(None, 112, 112,	96	
block1b_project_...			
(BatchNormalizatio...	24)		
block1b_drop	(None, 112, 112,	0	
block1b_project_...			
(Dropout)	24)		
block1b_add (Add)	(None, 112, 112,	0	
block1b_drop[0][...]			
24)			
block1a_project_...			
block2a_expand_conv	(None, 112, 112,	3,456	
block1b_add[0][0]			
(Conv2D)	144)		
block2a_expand_bn	(None, 112, 112,	576	
block2a_expand_c...			
(BatchNormalizatio...	144)		
block2a_expand_act...	(None, 112, 112,	0	
block2a_expand_b...			
(Activation)	144)		

	block2a_dwconv_pad   (None, 113, 113,   0		
	block2a_expand_a...   (ZeroPadding2D)   144)		
	block2a_dwconv   (None, 56, 56,   1,296		
	block2a_dwconv_p...   (DepthwiseConv2D)   144)		
	block2a_bn   (None, 56, 56,   576		
	block2a_dwconv[0...   (BatchNormalizatio...   144)		
	block2a_activation   (None, 56, 56,   0   block2a_bn[0]		
	[0]   (Activation)   144)		
	block2a_se_squeeze   (None, 144)   0		
	block2a_activati...   (GlobalAveragePool...)		
	block2a_se_reshape   (None, 1, 1, 144)   0		
	block2a_se_squeee...   (Reshape)		
	block2a_se_reduce   (None, 1, 1, 6)   870		
	block2a_se_resha...   (Conv2D)		
	block2a_se_expand   (None, 1, 1, 144)   1,008		
	block2a_se_reduc...   (Conv2D)		

block2a_se_excite	(None, 56, 56,	0
block2a_activati...	(Multiply)	144)
block2a_se_expan...		
block2a_project_co...	(Conv2D)	(None, 56, 56,
block2a_se_excit...		4,608
block2a_project_bn	(BatchNormalizatio...	(None, 56, 56,
block2a_project_...		128
block2b_expand_conv	(Conv2D)	(None, 56, 56,
block2a_project_...		6,144
block2b_expand_bn	(BatchNormalizatio...	(None, 56, 56,
block2b_expand_c...		768
block2b_expand_act...	(Activation)	(None, 56, 56,
block2b_expand_b...		0
block2b_dwconv	(DepthwiseConv2D)	(None, 56, 56,
block2b_expand_a...		1,728
block2b_bn	(BatchNormalizatio...	(None, 56, 56,
block2b_dwconv[0...		768

block2b_activation	(None, 56, 56,	0	block2b_bn[0]
[0] (Activation)	192)		
block2b_se_squeeze	(None, 192)	0	
block2b_activati... (GlobalAveragePool...)			
block2b_se_reshape	(None, 1, 1, 192)	0	
block2b_se_squee... (Reshape)			
block2b_se_reduce	(None, 1, 1, 8)	1,544	
block2b_se_resha... (Conv2D)			
block2b_se_expand	(None, 1, 1, 192)	1,728	
block2b_se_reduc... (Conv2D)			
block2b_se_excite	(None, 56, 56,	0	
block2b_activati... (Multiply)	192)		
block2b_se_expan... (Conv2D)			
block2b_project_co... (Conv2D)	(None, 56, 56,   32)	6,144	
block2b_project_bn	(None, 56, 56,	128	
block2b_project_... (BatchNormalizatio...   32)			
block2b_drop	(None, 56, 56,	0	

block2b_project_...			
(Dropout)	32)		
block2b_add (Add)	(None, 56, 56,	0	
block2b_drop[0][...]	32)		
block2a_project_...			
block2c_expand_conv	(None, 56, 56,	6,144	
block2b_add[0][0]	(Conv2D)	192)	
block2c_expand_bn	(None, 56, 56,	768	
block2c_expand_c...	(BatchNormalizatio...	192)	
block2c_expand_act...	(None, 56, 56,	0	
block2c_expand_b...	(Activation)	192)	
block2c_dwconv	(None, 56, 56,	1,728	
block2c_expand_a...	(DepthwiseConv2D)	192)	
block2c_bn	(None, 56, 56,	768	
block2c_dwconv[0...]	(BatchNormalizatio...	192)	
block2c_activation	(None, 56, 56,	0   block2c_bn[0]	
[0]	(Activation)	192)	
block2c_se_squeeze	(None, 192)	0	
block2c_activati...			

(GlobalAveragePool...			
block2c_se_reshape   (None, 1, 1, 192)   0			
block2c_se_squee...   (Reshape)			
block2c_se_reduce   (None, 1, 1, 8)   1,544			
block2c_se_resha...   (Conv2D)			
block2c_se_expand   (None, 1, 1, 192)   1,728			
block2c_se_reduc...   (Conv2D)			
block2c_se_excite   (None, 56, 56,   0			
block2c_activati...   (Multiply)	192		
block2c_se_expan...			
block2c_project_co...   (None, 56, 56,   6,144			
block2c_se_excit...   (Conv2D)	32		
block2c_project_bn   (None, 56, 56,   128			
block2c_project_...   (BatchNormalizatio...	32		
block2c_drop   (None, 56, 56,   0			
block2c_project_...   (Dropout)	32		
block2c_add (Add)   (None, 56, 56,   0			
block2c_drop[0][...   (32)			

block2b_add[0][0]			
block3a_expand_conv	(None, 56, 56,	6,144	
block2c_add[0][0]	(Conv2D)	192	
block3a_expand_bn	(None, 56, 56,	768	
block3a_expand_c...	(BatchNormalizatio...	192	
block3a_expand_act...	(None, 56, 56,	0	
block3a_expand_b...	(Activation)	192	
block3a_dwconv_pad	(None, 59, 59,	0	
block3a_expand_a...	(ZeroPadding2D)	192	
block3a_dwconv	(None, 28, 28,	4,800	
block3a_dwconv_p...	(DepthwiseConv2D)	192	
block3a_bn	(None, 28, 28,	768	
block3a_dwconv[0...]	(BatchNormalizatio...	192	
block3a_activation	(None, 28, 28,	0	block3a_bn[0]
[0]	(Activation)	192	
block3a_se_squeeze	(None, 192)	0	
block3a_activati...	(GlobalAveragePool...		

block3a_se_reshape	(None, 1, 1, 192)	0
block3a_se_squeeze (Reshape)		
block3a_se_reduce	(None, 1, 1, 8)	1,544
block3a_se_resha... (Conv2D)		
block3a_se_expand	(None, 1, 1, 192)	1,728
block3a_se_reduc... (Conv2D)		
block3a_se_excite	(None, 28, 28,	0
block3a_activati... (Multiply)	192)	
block3a_se_expan...		
block3a_project_co...	(None, 28, 28,	9,216
block3a_se_excit... (Conv2D)	48)	
block3a_project_bn	(None, 28, 28,	192
block3a_project_... (BatchNormalizatio...)	48)	
block3b_expand_conv	(None, 28, 28,	13,824
block3a_project_... (Conv2D)	288)	
block3b_expand_bn	(None, 28, 28,	1,152
block3b_expand_c... (BatchNormalizatio...)	288)	

block3b_expand_act...	(None, 28, 28,	0
block3b_expand_b... (Activation)	288)	
block3b_dwconv block3b_expand_a... (DepthwiseConv2D)	(None, 28, 28,   288)	7,200
block3b_bn block3b_dwconv[0... (BatchNormalizatio...   288)	(None, 28, 28,   288)	1,152
block3b_activation [0] (Activation)	(None, 28, 28,   288)	0   block3b_bn[0]
block3b_se_squeeze block3b_activati... (GlobalAveragePool... 	(None, 288)	0
block3b_se_reshape block3b_se_squee... (Reshape)	(None, 1, 1, 288)	0
block3b_se_reduce block3b_se_resha... (Conv2D)	(None, 1, 1, 12)	3,468
block3b_se_expand block3b_se_reduc... (Conv2D)	(None, 1, 1, 288)	3,744
block3b_se_excite	(None, 28, 28,	0

block3b_activation		
(Multiply)   288)		
block3b_se_expand		
block3b_project_co...   (None, 28, 28,	13,824	
block3b_se_excit...   (Conv2D)   48)		
block3b_project_bn   (None, 28, 28,	192	
block3b_project_...   (BatchNormalizatio...   48)		
block3b_drop   (None, 28, 28,	0	
block3b_project_...   (Dropout)   48)		
block3b_add (Add)   (None, 28, 28,	0	
block3b_drop[0][...   48)		
block3a_project_...		
block3c_expand_conv   (None, 28, 28,	13,824	
block3b_add[0][0]   (Conv2D)   288)		
block3c_expand_bn   (None, 28, 28,	1,152	
block3c_expand_c...   (BatchNormalizatio...   288)		
block3c_expand_act...   (None, 28, 28,	0	
block3c_expand_b...   (Activation)   288)		
block3c_dwconv   (None, 28, 28,	7,200	
block3c_expand_a...		

(DepthwiseConv2D)	288)		
block3c_bn	(None, 28, 28,	1,152	
block3c_dwconv[0...]	(BatchNormalizatio...	288)	
block3c_activation [0]	(Activation)   288)	0	block3c_bn[0]
block3c_se_squeeze	(None, 288)	0	
block3c_activati...	(GlobalAveragePool...		
block3c_se_reshape	(None, 1, 1, 288)	0	
block3c_se_squeee...	(Reshape)		
block3c_se_reduce	(None, 1, 1, 12)	3,468	
block3c_se_resha...	(Conv2D)		
block3c_se_expand	(None, 1, 1, 288)	3,744	
block3c_se_reduc...	(Conv2D)		
block3c_se_excite	(None, 28, 28,	0	
block3c_activati...	(Multiply)   288)		
block3c_se_expan...			
block3c_project_co...	(None, 28, 28,	13,824	
block3c_se_excit...	(Conv2D)   48)		

block3c_project_bn block3c_project_... (BatchNormalizatio...)	(None, 28, 28, 48)		192
block3c_drop block3c_project_... (Dropout)	(None, 28, 28, 48)		0
block3c_add (Add) block3c_drop[0][...] block3b_add[0][0]	(None, 28, 28, 48)		0
block4a_expand_conv block3c_add[0][0] (Conv2D)	(None, 28, 28, 288)		13,824
block4a_expand_bn block4a_expand_c... (BatchNormalizatio...)	(None, 28, 28, 288)		1,152
block4a_expand_act... block4a_expand_b... (Activation)	(None, 28, 28, 288)		0
block4a_dwconv_pad block4a_expand_a... (ZeroPadding2D)	(None, 29, 29, 288)		0
block4a_dwconv block4a_dwconv_p... (DepthwiseConv2D)	(None, 14, 14, 288)		2,592

block4a_bn	(None, 14, 14,	1,152	
block4a_dwconv[0...	(BatchNormalizatio...	288)	
block4a_activation	(None, 14, 14,	0	block4a_bn[0]
[0]	(Activation)	288)	
block4a_se_squeeze	(None, 288)	0	
block4a_activati...	(GlobalAveragePool...		
block4a_se_reshape	(None, 1, 1, 288)	0	
block4a_se_squee...	(Reshape)		
block4a_se_reduce	(None, 1, 1, 12)	3,468	
block4a_se_resha...	(Conv2D)		
block4a_se_expand	(None, 1, 1, 288)	3,744	
block4a_se_reduc...	(Conv2D)		
block4a_se_excite	(None, 14, 14,	0	
block4a_activati...	(Multiply)	288)	
block4a_se_expan...			
block4a_project_co...	(None, 14, 14,	27,648	
block4a_se_excit...	(Conv2D)	96)	

block4a_project_bn	(None, 14, 14, 384)	
block4a_project_... (BatchNormalizatio...)	96)	
block4b_expand_conv	(None, 14, 14, 55,296)	
block4a_project_... (Conv2D)	576)	
block4b_expand_bn	(None, 14, 14, 2,304)	
block4b_expand_c... (BatchNormalizatio...)	576)	
block4b_expand_act... block4b_expand_b... (Activation)	(None, 14, 14, 0)	
block4b_dwconv	(None, 14, 14, 5,184)	
block4b_expand_a... (DepthwiseConv2D)	576)	
block4b_bn	(None, 14, 14, 2,304)	
block4b_dwconv[0... (BatchNormalizatio...)	576)	
block4b_activation [0] (Activation)	(None, 14, 14, 0)	block4b_bn[0]
block4b_se_squeeze	(None, 576)	0
block4b_activati... (GlobalAveragePool...)		

block4b_se_reshape	(None, 1, 1, 576)	0
block4b_se_squeeze (Reshape)		
block4b_se_reduce (Conv2D)	(None, 1, 1, 24)	13,848
block4b_se_reduce (Conv2D)	(None, 1, 1, 576)	14,400
block4b_se_excite (Multiply)	(None, 14, 14, 576)	0
block4b_se_expand (Conv2D)	(None, 14, 14, 96)	55,296
block4b_project_bn (BatchNormalization)	(None, 14, 14, 96)	384
block4b_drop (Dropout)	(None, 14, 14, 96)	0
block4b_add (Add) block4b_drop[0][...] block4a_project [...]	(None, 14, 14, 96)	0
block4c_expand_conv	(None, 14, 14, 55,296)	55,296

block4b_add[0][0]			
(Conv2D)   576)			
block4c_expand_bn   (None, 14, 14,   2,304			
block4c_expand_c...			
(BatchNormalizatio...   576)			
block4c_expand_act...   (None, 14, 14,   0			
block4c_expand_b...			
(Activation)   576)			
block4c_dwconv   (None, 14, 14,   5,184			
block4c_expand_a...			
(DepthwiseConv2D)   576)			
block4c_bn   (None, 14, 14,   2,304			
block4c_dwconv[0...			
(BatchNormalizatio...   576)			
block4c_activation   (None, 14, 14,   0			block4c_bn[0]
[0]			
(Activation)   576)			
block4c_se_squeeze   (None, 576)   0			
block4c_activati...			
(GlobalAveragePool...			
block4c_se_reshape   (None, 1, 1, 576)   0			
block4c_se_squeee...			
(Reshape)			
block4c_se_reduce   (None, 1, 1, 24)   13,848			
block4c_se_resha...			

(Conv2D)			
block4c_se_expand	(None, 1, 1, 576)	14,400	
block4c_se_reduce	(Conv2D)		
block4c_se_excite	(None, 14, 14,	0	
block4c_activation	576)		
block4c_se_expand			
block4c_project_combine	(None, 14, 14,	55,296	
block4c_se_excite	96)		
block4c_project_bn	(None, 14, 14,	384	
block4c_project	96)		
block4c_drop	(None, 14, 14,	0	
block4c_project	96)		
block4c_add (Add)	(None, 14, 14,	0	
block4c_drop[0][0]	96)		
block4b_add[0][0]			
block4d_expand_conv	(None, 14, 14,	55,296	
block4c_add[0][0]	576)		
block4d_expand_bn	(None, 14, 14,	2,304	
block4d_expand_c	576)		
block4d_expand	(BatchNormalizatio...		

block4d_expand_act...	(None, 14, 14,	0	
block4d_expand_b... (Activation)	576)		
block4d_dwconv block4d_expand_a... (DepthwiseConv2D)	(None, 14, 14, 576)	5,184	
block4d_bn block4d_dwconv[0... (BatchNormalizatio...]	(None, 14, 14, 576)	2,304	
block4d_activation [0] (Activation)	(None, 14, 14, 576)	0	block4d_bn[0]
block4d_se_squeeze block4d_activati... (GlobalAveragePool...)	(None, 576)	0	
block4d_se_reshape block4d_se_squee... (Reshape)	(None, 1, 1, 576)	0	
block4d_se_reduce block4d_se_resha... (Conv2D)	(None, 1, 1, 24)	13,848	
block4d_se_expand block4d_se_reduc... (Conv2D)	(None, 1, 1, 576)	14,400	

block4d_se_excite	(None, 14, 14,	0
block4d_activati...	(Multiply)	576)
block4d_se_expan...		
block4d_project_co...	(None, 14, 14,	55,296
block4d_se_excit...	(Conv2D)	96)
block4d_project_bn	(None, 14, 14,	384
block4d_project_...	(BatchNormalizatio...	96)
block4d_drop	(None, 14, 14,	0
block4d_project_...	(Dropout)	96)
block4d_add (Add)	(None, 14, 14,	0
block4d_drop[0][...]	96)	
block4c_add[0][0]		
block4e_expand_conv	(None, 14, 14,	55,296
block4d_add[0][0]	(Conv2D)	576)
block4e_expand_bn	(None, 14, 14,	2,304
block4e_expand_c...	(BatchNormalizatio...	576)
block4e_expand_act...	(None, 14, 14,	0
block4e_expand_b...	(Activation)	576)

block4e_dwconv	(None, 14, 14,	5,184
block4e_expand_a... (DepthwiseConv2D)	576)	
block4e_bn	(None, 14, 14,	2,304
block4e_dwconv[0... (BatchNormalizatio...)	576)	
block4e_activation [0] (Activation)	(None, 14, 14,   576)	0   block4e_bn[0]
block4e_se_squeeze block4e_activati... (GlobalAveragePool...)	(None, 576)	0
block4e_se_reshape block4e_se_squee... (Reshape)	(None, 1, 1, 576)	0
block4e_se_reduce block4e_se_resha... (Conv2D)	(None, 1, 1, 24)	13,848
block4e_se_expand block4e_se_reduc... (Conv2D)	(None, 1, 1, 576)	14,400
block4e_se_excite block4e_activati... (Multiply)	(None, 14, 14,   576)	0
block4e_se_expan...		
block4e_project_co...	(None, 14, 14,	55,296

block4e_se_excit...		
(Conv2D)	96)	
block4e_project_bn	(None, 14, 14,	384
block4e_project_...	(BatchNormalizatio...	
96)		
block4e_drop	(None, 14, 14,	0
block4e_project_...	(Dropout)	
96)		
block4e_add (Add)	(None, 14, 14,	0
block4e_drop[0][...]	96)	
block4d_add[0][0]		
block5a_expand_conv	(None, 14, 14,	55,296
block4e_add[0][0]	(Conv2D)	
576)		
block5a_expand_bn	(None, 14, 14,	2,304
block5a_expand_c...	(BatchNormalizatio...	
576)		
block5a_expand_act...	(None, 14, 14,	0
block5a_expand_b...	(Activation)	
576)		
block5a_dwconv	(None, 14, 14,	14,400
block5a_expand_a...	(DepthwiseConv2D)	
576)		
block5a_bn	(None, 14, 14,	2,304
block5a_dwconv[0...]		

(BatchNormalizatio...	576)			
block5a_activation [0]	(None, 14, 14,		0	block5a_bn[0]
(Activation)	576)			
block5a_se_squeeze	(None, 576)		0	
block5a_activati...				
(GlobalAveragePool...				
block5a_se_reshape	(None, 1, 1, 576)		0	
block5a_se_squee...				
(Reshape)				
block5a_se_reduce	(None, 1, 1, 24)		13,848	
block5a_se_resha...				
(Conv2D)				
block5a_se_expand	(None, 1, 1, 576)		14,400	
block5a_se_reduc...				
(Conv2D)				
block5a_se_excite	(None, 14, 14,		0	
block5a_activati...				
(Multiply)	576)			
block5a_se_expan...				
block5a_project_co...	(None, 14, 14,		78,336	
block5a_se_excit...				
(Conv2D)	136)			
block5a_project_bn	(None, 14, 14,		544	
block5a_project_...				
(BatchNormalizatio...	136)			

block5b_expand_conv	(None, 14, 14, 816)	110,976	
block5a_project_...			
(Conv2D)			
block5b_expand_bn	(None, 14, 14, 816)	3,264	
block5b_expand_c...			
(BatchNormalizatio...			
block5b_expand_act...	(None, 14, 14, 816)	0	
block5b_expand_b...			
(Activation)			
block5b_dwconv	(None, 14, 14, 816)	20,400	
block5b_expand_a...			
(DepthwiseConv2D)			
block5b_bn	(None, 14, 14, 816)	3,264	
block5b_dwconv[0...]			
(BatchNormalizatio...			
block5b_activation	(None, 14, 14, 816)	0	block5b_bn[0]
[0]			
(Activation)			
block5b_se_squeeze	(None, 816)	0	
block5b_activati...			
(GlobalAveragePool...			
block5b_se_reshape	(None, 1, 1, 816)	0	
block5b_se_squee...			
(Reshape)			

block5b_se_reduce	(None, 1, 1, 34)	27,778	
block5b_se_resha...			
(Conv2D)			
block5b_se_expand	(None, 1, 1, 816)	28,560	
block5b_se_reduc...			
(Conv2D)			
block5b_se_excite	(None, 14, 14,	0	
block5b_activati...			
(Multiply)	816)		
block5b_se_expan...			
block5b_project_co...	(None, 14, 14,	110,976	
block5b_se_excit...			
(Conv2D)	136)		
block5b_project_bn	(None, 14, 14,	544	
block5b_project_...			
(BatchNormalizatio...	136)		
block5b_drop	(None, 14, 14,	0	
block5b_project_...			
(Dropout)	136)		
block5b_add (Add)	(None, 14, 14,	0	
block5b_drop[0][...]			
136)			
block5a_project_...			
block5c_expand_conv	(None, 14, 14,	110,976	
block5b_add[0][0]			
(Conv2D)	816)		

block5c_expand_bn	(None, 14, 14, 816)	3,264
block5c_expand_c...		
(BatchNormalizatio...	816)	
block5c_expand_act...	(None, 14, 14, 816)	0
block5c_expand_b...		
(Activation)		
block5c_dwconv	(None, 14, 14, 816)	20,400
block5c_expand_a...		
(DepthwiseConv2D)		
block5c_bn	(None, 14, 14, 816)	3,264
block5c_dwconv[0...		
(BatchNormalizatio...	816)	
block5c_activation	(None, 14, 14, 816)	0
[0]		block5c_bn[0]
(Activation)		
block5c_se_squeeze	(None, 816)	0
block5c_activati...		
(GlobalAveragePool...		
block5c_se_reshape	(None, 1, 1, 816)	0
block5c_se_squeee...		
(Reshape)		
block5c_se_reduce	(None, 1, 1, 34)	27,778
block5c_se_resha...		
(Conv2D)		

block5c_se_expand	(None, 1, 1, 816)	28,560
block5c_se_reduce (Conv2D)		
block5c_se_excite	(None, 14, 14,	0
block5c_activation (Multiply)	816)	
block5c_se_expand		
block5c_project_com... block5c_se_excite (Conv2D)	(None, 14, 14,   136)	110,976
block5c_project_bn	(None, 14, 14,	544
block5c_project_... (BatchNormalization)	136)	
block5c_drop	(None, 14, 14,	0
block5c_project_... (Dropout)	136)	
block5c_add (Add)	(None, 14, 14,	0
block5c_drop[0][...]	136)	
block5b_add[0][0]		
block5d_expand_conv block5c_add[0][0]	(None, 14, 14,   816)	110,976
block5d_expand_bn	(None, 14, 14,	3,264
block5d_expand_c... (BatchNormalizatio...)	816)	
block5d_expand_act...	(None, 14, 14,	0

block5d_expand_b...	(Activation)	816)			
block5d_dwconv		(None, 14, 14,	20,400		
block5d_expand_a...	(DepthwiseConv2D)	816)			
block5d_bn		(None, 14, 14,	3,264		
block5d_dwconv[0...	(BatchNormalizatio...	816)			
block5d_activation		(None, 14, 14,	0	block5d_bn[0]	
[0]	(Activation)	816)			
block5d_se_squeeze		(None, 816)	0		
block5d_activati...	(GlobalAveragePool...				
block5d_se_reshape		(None, 1, 1, 816)	0		
block5d_se_squee...	(Reshape)				
block5d_se_reduce		(None, 1, 1, 34)	27,778		
block5d_se_resha...	(Conv2D)				
block5d_se_expand		(None, 1, 1, 816)	28,560		
block5d_se_reduc...	(Conv2D)				
block5d_se_excite		(None, 14, 14,	0		
block5d_activati...					

(Multiply)	816)		
block5d_se_expan...			
block5d_project_co...	(None, 14, 14,	110,976	
block5d_se_excit...	(Conv2D)	136)	
block5d_project_bn	(None, 14, 14,	544	
block5d_project_...	(BatchNormalizatio...	136)	
block5d_drop	(None, 14, 14,	0	
block5d_project_...	(Dropout)	136)	
block5d_add (Add)	(None, 14, 14,	0	
block5d_drop[0] [...	136)		
block5c_add[0][0]			
block5e_expand_conv	(None, 14, 14,	110,976	
block5d_add[0][0]	(Conv2D)	816)	
block5e_expand_bn	(None, 14, 14,	3,264	
block5e_expand_c...	(BatchNormalizatio...	816)	
block5e_expand_act...	(None, 14, 14,	0	
block5e_expand_b...	(Activation)	816)	
block5e_dwconv	(None, 14, 14,	20,400	
block5e_expand_a...	(DepthwiseConv2D)	816)	

block5e_bn	(None, 14, 14, 3,264)		
block5e_dwconv[0...]	(BatchNormalizatio...   816)		
block5e_activation [0]	(Activation)   816	0	block5e_bn[0]
block5e_se_squeeze	(None, 816)	0	
block5e_activati...	(GlobalAveragePool...)		
block5e_se_reshape	(None, 1, 1, 816)	0	
block5e_se_squeee...	(Reshape)		
block5e_se_reduce	(None, 1, 1, 34)	27,778	
block5e_se_resha...	(Conv2D)		
block5e_se_expand	(None, 1, 1, 816)	28,560	
block5e_se_reduc...	(Conv2D)		
block5e_se_excite	(None, 14, 14, 0)		
block5e_activati...	(Multiply)   816)		
block5e_se_expan...			
block5e_project_co...	(None, 14, 14, 110,976)		
block5e_se_excit...	(Conv2D)   136)		

block5e_project_bn	(None, 14, 14, 136)	BatchNormalizatio...	544
block5e_project_...			
block5e_drop	(None, 14, 14, 136)	Dropout	0
block5e_project_...			
block5e_add (Add)	(None, 14, 14, 136)	block5e_drop[0][...]	0
block5d_add[0][0]			
block6a_expand_conv	(None, 14, 14, 816)	block5e_add[0][0]   Conv2D	110,976
block6a_expand_c...			
block6a_expand_bn	(None, 14, 14, 816)	BatchNormalizatio...	3,264
block6a_expand_c...			
block6a_expand_act...	(None, 14, 14, 816)	block6a_expand_b...   Activation	0
block6a_expand_b...			
block6a_dwconv_pad	(None, 17, 17, 816)	block6a_expand_a...   ZeroPadding2D	0
block6a_expand_a...			
block6a_dwconv	(None, 7, 7, 816)	block6a_dwconv_p...   DepthwiseConv2D	20,400
block6a_dwconv_p...			

block6a_bn	(None, 7, 7, 816)	3,264
block6a_dwconv[0...   (BatchNormalizatio...		
block6a_activation [0] (Activation)	(None, 7, 7, 816)	0   block6a_bn[0]
block6a_se_squeeze block6a_activati...   (GlobalAveragePool...	(None, 816)	0
block6a_se_reshape block6a_se_squee...   (Reshape)	(None, 1, 1, 816)	0
block6a_se_reduce block6a_se_resha...   (Conv2D)	(None, 1, 1, 34)	27,778
block6a_se_expand block6a_se_reduc...   (Conv2D)	(None, 1, 1, 816)	28,560
block6a_se_excite block6a_activati...   (Multiply)	(None, 7, 7, 816)	0
block6a_se_expan...		
block6a_project_co... block6a_se_excit...   (Conv2D)	(None, 7, 7, 232)	189,312
block6a_project_bn	(None, 7, 7, 232)	928

block6a_project_...	(BatchNormalizatio...		
block6b_expand_conv	(None, 7, 7,	322,944	
block6a_project_...	(Conv2D)	1392	
block6b_expand_bn	(None, 7, 7,	5,568	
block6b_expand_c...	(BatchNormalizatio...	1392	
block6b_expand_act...	(None, 7, 7,	0	
block6b_expand_b...	(Activation)	1392	
block6b_dwconv	(None, 7, 7,	34,800	
block6b_expand_a...	(DepthwiseConv2D)	1392	
block6b_bn	(None, 7, 7,	5,568	
block6b_dwconv[0...	(BatchNormalizatio...	1392	
block6b_activation	(None, 7, 7,	0	block6b_bn[0]
[0]	(Activation)	1392	
block6b_se_squeeze	(None, 1392)	0	
block6b_activati...	(GlobalAveragePool...		
block6b_se_reshape	(None, 1, 1,	0	
block6b_se_squeez...			

(Reshape)	1392)		
block6b_se_reduce	(None, 1, 1, 58)	80,794	
block6b_se_resha... (Conv2D)			
block6b_se_expand	(None, 1, 1,	82,128	
block6b_se_reduc... (Conv2D)	1392)		
block6b_se_excite	(None, 7, 7,	0	
block6b_activati... (Multiply)	1392)		
block6b_se_expan...			
block6b_project_co... block6b_se_excit... (Conv2D)	(None, 7, 7, 232)	322,944	
block6b_project_bn	(None, 7, 7, 232)	928	
block6b_project_... (BatchNormalizatio...)			
block6b_drop	(None, 7, 7, 232)	0	
block6b_project_... (Dropout)			
block6b_add (Add)	(None, 7, 7, 232)	0	
block6b_drop[0][...]			
block6a_project_...			
block6c_expand_conv block6b_add[0][0] (Conv2D)	(None, 7, 7,	322,944	
	1392)		

block6c_expand_bn	(None, 7, 7, 5,568)		
block6c_expand_c... (BatchNormalizatio...)	1392		
block6c_expand_act... block6c_expand_b... (Activation)	(None, 7, 7, 0)		
block6c_dwconv block6c_expand_a... (DepthwiseConv2D)	(None, 7, 7, 34,800)		
block6c_bn block6c_dwconv[0... (BatchNormalizatio...)	(None, 7, 7, 5,568)		
block6c_activation [0] (Activation)	(None, 7, 7, 0)	block6c_bn[0]	
block6c_se_squeeze block6c_activati... (GlobalAveragePool...)	(None, 1392, 0)		
block6c_se_reshape block6c_se_squee... (Reshape)	(None, 1, 1, 0)		
block6c_se_reduce block6c_se_resha... (Conv2D)	(None, 1, 1, 58, 80,794)		

block6c_se_expand   (None, 1, 1,   82,128			
block6c_se_reduc...			
(Conv2D)   1392)			
block6c_se_excite   (None, 7, 7,   0			
block6c_activati...			
(Multiply)   1392)			
block6c_se_expan...			
block6c_project_co...   (None, 7, 7, 232)   322,944			
block6c_se_excit...			
(Conv2D)			
block6c_project_bn   (None, 7, 7, 232)   928			
block6c_project_...			
(BatchNormalizatio...			
block6c_drop   (None, 7, 7, 232)   0			
block6c_project_...			
(Dropout)			
block6c_add (Add)   (None, 7, 7, 232)   0			
block6c_drop[0][...			
block6b_add[0][0]			
block6d_expand_conv   (None, 7, 7,   322,944			
block6c_add[0][0]			
(Conv2D)   1392)			
block6d_expand_bn   (None, 7, 7,   5,568			
block6d_expand_c...			
(BatchNormalizatio...   1392)			

block6d_expand_act...	(None, 7, 7,	0
block6d_expand_b...	(Activation)	1392
block6d_dwconv	(None, 7, 7,	34,800
block6d_expand_a...	(DepthwiseConv2D)	1392
block6d_bn	(None, 7, 7,	5,568
block6d_dwconv[0...]	(BatchNormalizatio...	1392
block6d_activation	(None, 7, 7,	0
[0]	(Activation)	1392
block6d_se_squeeze	(None, 1392)	0
block6d_activati...	(GlobalAveragePool...	
block6d_se_reshape	(None, 1, 1,	0
block6d_se_squeee...	(Reshape)	1392
block6d_se_reduce	(None, 1, 1, 58)	80,794
block6d_se_resha...	(Conv2D)	
block6d_se_expand	(None, 1, 1,	82,128
block6d_se_reduc...	(Conv2D)	1392

block6d_se_excite	(None, 7, 7,	0
block6d_activati...	(Multiply)	1392
block6d_se_expan...		
block6d_project_co...	(None, 7, 7, 232)	322,944
block6d_se_excit...	(Conv2D)	
block6d_project_bn	(None, 7, 7, 232)	928
block6d_project_...	(BatchNormalizatio...	
block6d_drop	(None, 7, 7, 232)	0
block6d_project_...	(Dropout)	
block6d_add (Add)	(None, 7, 7, 232)	0
block6d_drop[0][...]		
block6c_add[0][0]		
block6e_expand_conv	(None, 7, 7,	322,944
block6d_add[0][0]	(Conv2D)	1392
block6e_expand_bn	(None, 7, 7,	5,568
block6e_expand_c...	(BatchNormalizatio...	
block6e_expand_act...	(None, 7, 7,	0
block6e_expand_b...	(Activation)	1392
block6e_dwconv	(None, 7, 7,	34,800

block6e_expand_a...			
(DepthwiseConv2D)	1392)		
block6e_bn	(None, 7, 7,	5,568	
block6e_dwconv[0...	(BatchNormalizatio...	1392)	
block6e_activation	(None, 7, 7,	0	block6e_bn[0]
[0]	(Activation)	1392)	
block6e_se_squeeze	(None, 1392)	0	
block6e_activati...	(GlobalAveragePool...		
block6e_se_reshape	(None, 1, 1,	0	
block6e_se_squeee...	(Reshape)	1392)	
block6e_se_reduce	(None, 1, 1, 58)	80,794	
block6e_se_resha...	(Conv2D)		
block6e_se_expand	(None, 1, 1,	82,128	
block6e_se_reduc...	(Conv2D)	1392)	
block6e_se_excite	(None, 7, 7,	0	
block6e_activati...	(Multiply)	1392)	
block6e_se_expan...			
block6e_project_co...	(None, 7, 7, 232)	322,944	
block6e_se_excit...			

(Conv2D)			
block6e_project_bn	(None, 7, 7, 232)	928	
block6e_project_...	(BatchNormalizatio...		
block6e_drop	(None, 7, 7, 232)	0	
block6e_project_...	(Dropout)		
block6e_add (Add)	(None, 7, 7, 232)	0	
block6e_drop[0][...]			
block6d_add[0][0]			
block6f_expand_conv	(None, 7, 7,	322,944	
block6e_add[0][0]	(Conv2D)	1392	
block6f_expand_bn	(None, 7, 7,	5,568	
block6f_expand_c...	(BatchNormalizatio...		
block6f_expand_act...	(None, 7, 7,	0	
block6f_expand_b...	(Activation)	1392	
block6f_dwconv	(None, 7, 7,	34,800	
block6f_expand_a...	(DepthwiseConv2D)	1392	
block6f_bn	(None, 7, 7,	5,568	
block6f_dwconv[0...	(BatchNormalizatio...		
	1392)		

block6f_activation [0] (Activation)	(None, 7, 7, 1392)		0	block6f_bn[0]
block6f_se_squeeze block6f_activati... (GlobalAveragePool...)	(None, 1392)		0	
block6f_se_reshape block6f_se_squee... (Reshape)	(None, 1, 1, 1392)		0	
block6f_se_reduce block6f_se_resha... (Conv2D)	(None, 1, 1, 58)		80,794	
block6f_se_expand block6f_se_reduc... (Conv2D)	(None, 1, 1, 1392)		82,128	
block6f_se_excite block6f_activati... (Multiply)	(None, 7, 7, 1392)		0	
block6f_project_co... block6f_se_excit... (Conv2D)	(None, 7, 7, 232)		322,944	
block6f_project_bn block6f_project_... (BatchNormalizatio...)	(None, 7, 7, 232)		928	

block6f_drop	(None, 7, 7, 232)	0
block6f_project_... (Dropout)		
block6f_add (Add) block6f_drop[0][...]	(None, 7, 7, 232)	0
block6e_add[0][0]		
block7a_expand_conv block6f_add[0][0] (Conv2D)	(None, 7, 7, 1392)	322,944
block7a_expand_bn block7a_expand_c... (BatchNormalizatio...)	(None, 7, 7, 1392)	5,568
block7a_expand_act... block7a_expand_b... (Activation)	(None, 7, 7, 1392)	0
block7a_dwconv block7a_expand_a... (DepthwiseConv2D)	(None, 7, 7, 1392)	12,528
block7a_bn block7a_dwconv[0... (BatchNormalizatio...)	(None, 7, 7, 1392)	5,568
block7a_activation [0] (Activation)	(None, 7, 7, 1392)	0   block7a_bn[0]

block7a_se_squeeze	(None, 1392)	0
block7a_activation	(GlobalAveragePool...)	
block7a_se_reshape	(None, 1, 1, 1392)	0
block7a_se_reduce	(None, 1, 1, 58)	80,794
block7a_se_resha...	(Conv2D)	
block7a_se_expand	(None, 1, 1, 1392)	82,128
block7a_se_reduc...	(Conv2D)	
block7a_se_excite	(None, 7, 7, 1392)	0
block7a_activation	(Multiply)	
block7a_se_expan...	(Conv2D)	
block7a_project_co...	(None, 7, 7, 384)	534,528
block7a_se_excit...	(Conv2D)	
block7a_project_bn	(None, 7, 7, 384)	1,536
block7a_project_...	(BatchNormalizatio...	
block7b_expand_conv	(None, 7, 7, 2304)	884,736
block7a_project_...	(Conv2D)	
block7b_expand_bn	(None, 7, 7, 9,216)	9,216

block7b_expand_c...	(BatchNormalizatio...	2304		
block7b_expand_act...	(None, 7, 7,		0	
block7b_expand_b...	(Activation)	2304		
block7b_dwconv	(None, 7, 7,	20,736		
block7b_expand_a...	(DepthwiseConv2D)	2304		
block7b_bn	(None, 7, 7,	9,216		
block7b_dwconv[0...]	(BatchNormalizatio...	2304		
block7b_activation	(None, 7, 7,	0	block7b_bn[0]	
[0]	(Activation)	2304		
block7b_se_squeeze	(None, 2304)	0		
block7b_activati...	(GlobalAveragePool...			
block7b_se_reshape	(None, 1, 1,	0		
block7b_se_squee...	(Reshape)	2304		
block7b_se_reduce	(None, 1, 1, 96)	221,280		
block7b_se_resha...	(Conv2D)			
block7b_se_expand	(None, 1, 1,	223,488		
block7b_se_reduc...				

(Conv2D)	2304)		
block7b_se_excite	(None, 7, 7,	0	
block7b_activati...			
(Multiply)	2304)		
block7b_se_expan...			
block7b_project_co...	(None, 7, 7, 384)	884,736	
block7b_se_excit...			
(Conv2D)			
block7b_project_bn	(None, 7, 7, 384)	1,536	
block7b_project_...			
(BatchNormalizatio...			
block7b_drop	(None, 7, 7, 384)	0	
block7b_project_...			
(Dropout)			
block7b_add (Add)	(None, 7, 7, 384)	0	
block7b_drop[0][...]			
block7a_project_...			
top_conv (Conv2D)	(None, 7, 7,	589,824	
block7b_add[0][0]	1536)		
top_bn [0]	(None, 7, 7,	6,144	top_conv[0]
(BatchNormalizatio...	1536)		
top_activation	(None, 7, 7,	0	top_bn[0][0]
(Activation)	1536)		

```

Total params: 10,783,535 (41.14 MB)
Trainable params: 10,696,232 (40.80 MB)
Non-trainable params: 87,303 (341.03 KB)

import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.applications import EfficientNetB3

# Set up the model
model = Sequential()

# Add an input layer
model.add(tf.keras.Input(shape=(224, 224, 3)))

# Add EfficientNetB3 base model
base_model = EfficientNetB3(include_top=False, weights='imagenet',
drop_connect_rate=0.6)
model.add(base_model)

# Now the model is built, you can call summary
model.summary()

```

Downloading data from https://storage.googleapis.com/keras-applications/efficientnetb3\_notop.h5  
43941136/43941136 ————— 0s 0us/step

Model: "sequential\_4"

Layer (type)	Output Shape
Param #	
efficientnetb3 (Functional)	(None, 7, 7, 1536)
10,783,535	

```

Total params: 10,783,535 (41.14 MB)
Trainable params: 10,696,232 (40.80 MB)
Non-trainable params: 87,303 (341.03 KB)

import tensorflow as tf
from tensorflow.keras.models import Sequential

```

```

from tensorflow.keras.layers import GlobalAveragePooling2D, Dense,
Dropout
from tensorflow.keras.applications import EfficientNetB3

def create_model():

    model = Sequential()
    # Add an input layer
    model.add(tf.keras.Input(shape=(224, 224, 3)))

    # Initialize EfficientNetB3 with input_shape explicitly defined
    base_model = EfficientNetB3(input_shape=(224, 224, 3),
include_top=False,
                           weights='imagenet',
drop_connect_rate=0.6)

    # Freeze layers (except last 40 layers)
    for layer in base_model.layers[:-40]:
        layer.trainable = False

    # Add the base model
    model.add(base_model)

    # Add global average pooling layer
    model.add(GlobalAveragePooling2D())

    # Add Dense layer with regularization
    model.add(Dense(256, activation='relu',
bias_regularizer=tf.keras.regularizers.L1L2(l1=0.01, l2=0.001)))

    # Add Dropout layer
    model.add(Dropout(0.5))

    # Add output Dense layer for classification (5 classes)
    model.add(Dense(5, activation='softmax'))

    return model

# Create and summarize the model
leaf_model = create_model()
leaf_model.summary()

Model: "sequential_5"

```

Layer (type)	Output Shape
Param #	

	efficientnetb3 (Functional)	(None, 7, 7, 1536)	
10,783,535			
0	global_average_pooling2d (GlobalAveragePooling2D)	(None, 1536)	
393,472	dense_6 (Dense)	(None, 256)	
0	dropout_3 (Dropout)	(None, 256)	
1,285	dense_7 (Dense)	(None, 5)	

Total params: 11,178,292 (42.64 MB)

Trainable params: 4,758,201 (18.15 MB)

Non-trainable params: 6,420,091 (24.49 MB)

```
EPOCHS = 50
STEP_SIZE_TRAIN = train_set.n//train_set.batch_size
STEP_SIZE_VALID = val_set.n//val_set.batch_size

def Model_fit():

    #leaf_model = None

    leaf_model = create_model()

    '''Compiling the model'''

    loss = tf.keras.losses.CategoricalCrossentropy(from_logits =
False,
label_smoothing=0.0001,
name='categorical_crossentropy' )

    leaf_model.compile(optimizer = Adam(learning_rate = 1e-3),
loss = loss, #'categorical_crossentropy'
```

```

        metrics = ['categorical_accuracy']) #'acc'

    # Stop training when the val_loss has stopped decreasing for 3
    epochs.
    es = EarlyStopping(monitor='val_loss', mode='min', patience=3,
                        restore_best_weights=True, verbose=1)

    # Save the model with the minimum validation loss
    checkpoint_cb = ModelCheckpoint("Cassava_best_model.keras",
                                    save_best_only=True,
                                    monitor = 'val_loss',
                                    mode='min')

    # reduce learning rate
    reduce_lr = ReduceLROnPlateau(monitor = 'val_loss',
                                  factor = 0.2,
                                  patience = 2,
                                  min_lr = 1e-6,
                                  mode = 'min',
                                  verbose = 1)

    history = leaf_model.fit(train_set,
                            validation_data = val_set,
                            epochs= EP0CHS,
                            batch_size = BATCH_SIZE,
                            # class_weight = d_class_weights,
                            steps_per_epoch = STEP_SIZE_TRAIN,
                            validation_steps = STEP_SIZE_VALID,
                            callbacks=[es, checkpoint_cb, reduce_lr])

    leaf_model.save('Cassava_model'+'.keras')

    return history

results = Model_fit()

Epoch 1/50

/opt/conda/lib/python3.10/site-packages/keras/src/trainers/
data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset`'
class should call `super().__init__(**kwargs)` in its constructor.
`**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will
be ignored.
    self._warn_if_super_not_called()
WARNING: All log messages before absl::InitializeLog() is called are
written to STDERR
I0000 00:00:1726315338.916660      167 service.cc:145] XLA service
0x7e91c001d690 initialized for platform CUDA (this does not guarantee
that XLA will be used). Devices:
I0000 00:00:1726315338.916741      167 service.cc:153]     StreamExecutor

```

```
device (0): Tesla T4, Compute Capability 7.5
I0000 00:00:1726315338.916745      167 service.cc:153] StreamExecutor
device (1): Tesla T4, Compute Capability 7.5

    2/1355 ━━━━━━━━━━ 1:32 68ms/step - categorical_accuracy:
0.1667 - loss: 1.7674

I0000 00:00:1726315384.574417      167 device_compiler.h:188] Compiled
cluster using XLA! This line is logged at most once for the lifetime
of the process.

1355/1355 ━━━━━━━━━━ 624s 410ms/step - categorical_accuracy:
0.6583 - loss: 0.9676 - val_categorical_accuracy: 0.7690 - val_loss:
0.6937 - learning_rate: 0.0010
Epoch 2/50
    1/1355 ━━━━━━━━━━ 1:28 65ms/step - categorical_accuracy:
0.8000 - loss: 0.7827

/opt/conda/lib/python3.10/contextlib.py:153: UserWarning: Your input
ran out of data; interrupting training. Make sure that your dataset or
generator can generate at least `steps_per_epoch * epochs` batches.
You may need to use the `repeat()` function when building your
dataset.
    self.gen.throw(typ, value, traceback)

1355/1355 ━━━━━━━━━━ 7s 5ms/step - categorical_accuracy:
0.8000 - loss: 0.7827 - val_categorical_accuracy: 1.0000 - val_loss:
0.0762 - learning_rate: 0.0010
Epoch 3/50
1355/1355 ━━━━━━━━━━ 351s 257ms/step - categorical_accuracy:
0.7274 - loss: 0.7747 - val_categorical_accuracy: 0.7606 - val_loss:
0.6555 - learning_rate: 0.0010
Epoch 4/50
    1/1355 ━━━━━━━━━━ 57s 43ms/step - categorical_accuracy:
0.5333 - loss: 1.0227
Epoch 4: ReduceLROnPlateau reducing learning rate to
0.00020000000949949026.
1355/1355 ━━━━━━━━━━ 0s 31us/step - categorical_accuracy:
0.5333 - loss: 1.0227 - val_categorical_accuracy: 0.6000 - val_loss:
0.8312 - learning_rate: 0.0010
Epoch 5/50
1355/1355 ━━━━━━━━━━ 345s 253ms/step - categorical_accuracy:
0.7538 - loss: 0.6843 - val_categorical_accuracy: 0.7972 - val_loss:
0.5733 - learning_rate: 2.0000e-04
Epoch 5: early stopping
Restoring model weights from the end of the best epoch: 2.

print('Train_Cat-Acc: ', max(results.history['categorical_accuracy']))
print('Val_Cat-Acc: ',
max(results.history['val_categorical_accuracy']))
```

```
Train_Cat-Acc:  0.800000011920929
Val_Cat-Acc:  1.0

def Train_Val_Plot(acc,val_acc,loss,val_loss):

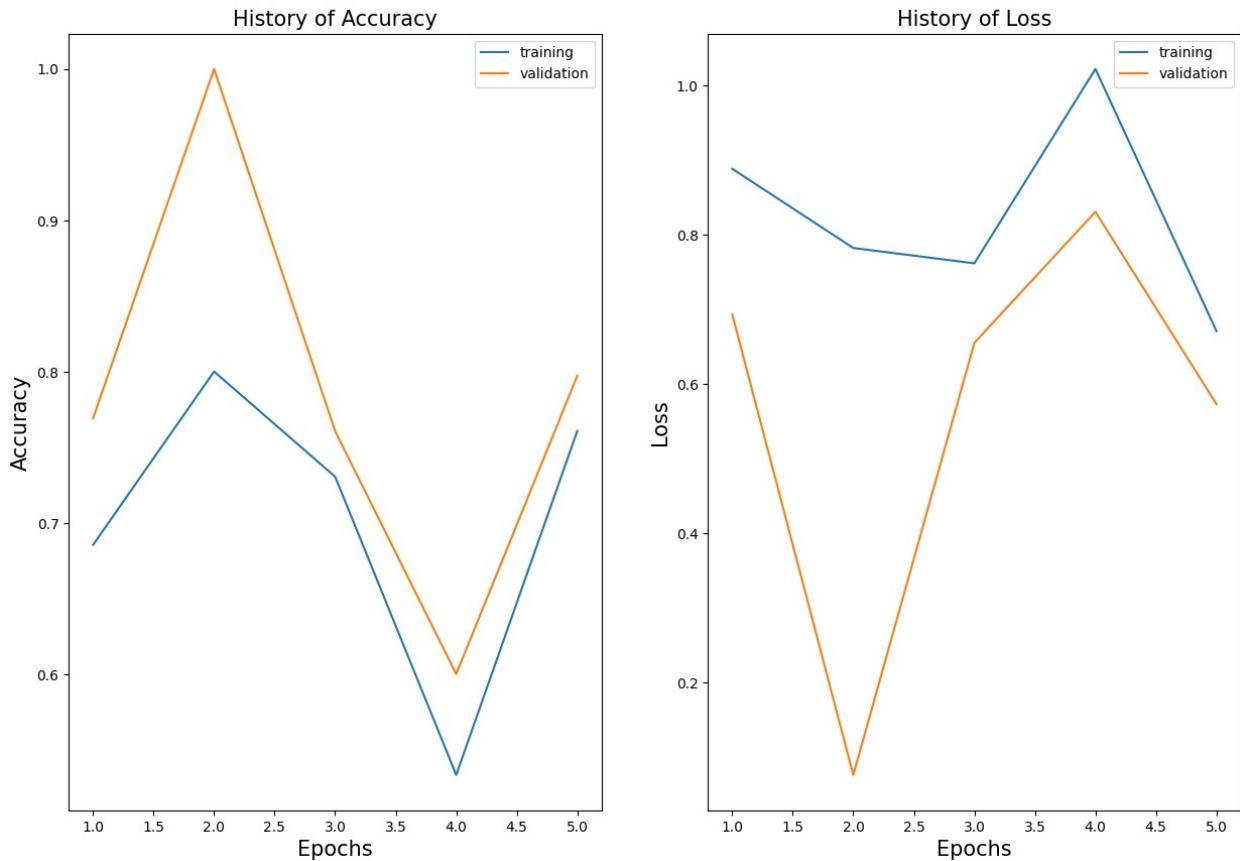
    fig, (ax1, ax2) = plt.subplots(1,2, figsize= (15,10))
    fig.suptitle(" MODEL'S METRICS VISUALIZATION ", fontsize=20)

    ax1.plot(range(1, len(acc) + 1), acc)
    ax1.plot(range(1, len(val_acc) + 1), val_acc)
    ax1.set_title('History of Accuracy', fontsize=15)
    ax1.set_xlabel('Epochs', fontsize=15)
    ax1.set_ylabel('Accuracy', fontsize=15)
    ax1.legend(['training', 'validation'])

    ax2.plot(range(1, len(loss) + 1), loss)
    ax2.plot(range(1, len(val_loss) + 1), val_loss)
    ax2.set_title('History of Loss', fontsize=15)
    ax2.set_xlabel('Epochs', fontsize=15)
    ax2.set_ylabel('Loss', fontsize=15)
    ax2.legend(['training', 'validation'])
    plt.show()

Train_Val_Plot(results.history['categorical_accuracy'],results.history
['val_categorical_accuracy'],
               results.history['loss'],results.history['val_loss'])
```

## MODEL'S METRICS VISUALIZATION



```
# EVALUATING THE MODEL
```

```
import keras
final_model = keras.models.load_model('Cassava_best_model.keras')
```

## Test Time Augmentaiton

```
TEST_DIR = '../input/cassava-leaf-disease-classification/test_images/'
test_images = os.listdir(TEST_DIR)
datagen = ImageDataGenerator(horizontal_flip=True)

def pred(images):
    for image in test_images:
        img = Image.open(TEST_DIR + image)
        img = img.resize(size)
        samples = np.expand_dims(img, axis=0)
        it = datagen.flow(samples, batch_size=10)
```

```
    yhats = final_model.predict_generator(it, steps=10, verbose=0)
    summed = np.sum(yhats, axis=0)
    return np.argmax(summed)

predictions = pred(test_images)
```