

Importing Libraries

```
In [1]: !pip install matplotlib  
!pip install tensorflow  
!pip install pandas  
!pip install seaborn
```

Requirement already satisfied: matplotlib in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (3.8.3)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(1.2.0)
Requirement already satisfied: cycler>=0.10 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(4.50.0)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(1.4.5)
Requirement already satisfied: numpy<2,>=1.21 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(1.26.4)
Requirement already satisfied: packaging>=20.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(24.0)
Requirement already satisfied: pillow>=8 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(10.2.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(3.1.2)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib)(2.9.0.post0)
Requirement already satisfied: six>=1.5 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from python-dateutil>=2.7>matplotlib)(1.16.0)
Requirement already satisfied: tensorflow in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (2.16.1)
Requirement already satisfied: tensorflow-intel==2.16.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow)(2.16.1)
Requirement already satisfied: absl-py>=1.0.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(2.1.0)
Requirement already satisfied: astunparse>=1.6.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(1.6.3)
Requirement already satisfied: flatbuffers>=23.5.26 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-low-intel==2.16.1->tensorflow)(24.3.25)
Requirement already satisfied: gast!=0.5.0,!0.5.1,!0.5.2,>=0.2.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(0.5.4)
Requirement already satisfied: google-pasta>=0.1.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-low-intel==2.16.1->tensorflow)(0.2.0)
Requirement already satisfied: h5py>=3.10.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(3.10.0)
Requirement already satisfied: libclang>=13.0.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(18.1.1)
Requirement already satisfied: ml-dtypes~0.3.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(0.3.2)
Requirement already satisfied: opt-einsum>=2.3.2 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(3.3.0)
Requirement already satisfied: packaging in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(24.0)
Requirement already satisfied: protobuf!=4.21.0,!4.21.1,!4.21.2,!4.21.3,!4.21.4,!4.21.5,<5.0.0dev,>=3.20.3 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(4.25.3)
Requirement already satisfied: requests<3,>=2.21.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-low-intel==2.16.1->tensorflow)(2.31.0)
Requirement already satisfied: setuptools in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(63.2.0)
Requirement already satisfied: six>=1.12.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(1.16.0)
Requirement already satisfied: termcolor>=1.1.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(2.4.0)
Requirement already satisfied: typing-extensions>=3.6.6 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(4.10.0)
Requirement already satisfied: wrapt>=1.11.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(1.16.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-low-intel==2.16.1->tensorflow)(1.62.1)
Requirement already satisfied: tensorboard<2.17,>=2.16 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(2.16.2)
Requirement already satisfied: keras>=3.0.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(3.1.1)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel==2.16.1->tensorflow)(0.31.0)
Requirement already satisfied: numpy<2.0.0,>=1.23.5 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-low-intel==2.16.1->tensorflow)(1.26.4)
Requirement already satisfied: wheel<1.0,>=0.23.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from astunparse>=1.6.0->tensorflow-intel==2.16.1->tensorflow)(0.43.0)
Requirement already satisfied: rich in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from keras>=3.0.0->tensorflow-intel==2.16.1->tensorflow)(13.7.1)
Requirement already satisfied: namex in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from keras>=3.0.0->tensorflow-intel==2.16.1->tensorflow)(0.0.7)
Requirement already satisfied: optree in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from keras>=3.0.0->tensorflow-low-intel==2.16.1->tensorflow)(0.11.0)
Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from requests<3,>=2.21.0->tensorflow-intel==2.16.1->tensorflow)(3.3.2)
Requirement already satisfied: idna<4,>=2.5 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from requests<3,>=2.21.0->tensorflow-intel==2.16.1->tensorflow)(3.6)
Requirement already satisfied: urllib3<3,>=1.21.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from requests<3,>=2.21.0->tensorflow-intel==2.16.1->tensorflow)(2.2.1)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from requests<3,>=2.21.0->tensorflow-intel==2.16.1->tensorflow)(2024.2.2)
Requirement already satisfied: markdown>=2.6.8 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow<2.17,>=2.16->tensorflow-intel==2.16.1->tensorflow)(3.6)
Requirement already satisfied: tensorflow-data-server<0.8.0,>=0.7.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow-intel<2.17,>=2.16->tensorflow-intel==2.16.1->tensorflow)(0.7.2)
Requirement already satisfied: werkzeug>=1.0.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from tensorflow<2.17,>=2.16->tensorflow-intel==2.16.1->tensorflow)(3.0.1)
Requirement already satisfied: MarkupSafe>=2.1.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from werkzeug>=1.0.1->tensorflow<2.17,>=2.16->tensorflow-intel==2.16.1->tensorflow)(2.1.5)
Requirement already satisfied: markdown-it-py>=2.2.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from rich->

```
keras>=3.0.0->tensorflow-intel==2.16.1->tensorflow (3.0.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from rich->keras>=3.0.0->tensorflow-intel==2.16.1->tensorflow) (2.17.2)
Requirement already satisfied: mdurl~=0.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from markdown-it-py>=2.2.0->rich->keras>=3.0.0->tensorflow-intel==2.16.1->tensorflow) (0.1.2)
Requirement already satisfied: pandas in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (2.2.1)
Requirement already satisfied: numpy<2,>=1.22.4 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from pandas) (1.26.4)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from pandas) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from pandas) (2024.1)
Requirement already satisfied: six>=1.5 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from python-dateutil>=2.8.2->pandas) (1.16.0)
Requirement already satisfied: seaborn in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (0.13.2)
Requirement already satisfied: numpy!=1.24.0,>=1.20 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from seaborn) (1.26.4)
Requirement already satisfied: pandas>=1.2 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from seaborn) (2.2.1)
Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from seaborn) (3.8.3)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.2.0)
Requirement already satisfied: cycler>=0.10 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (4.50.0)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.4.5)
Requirement already satisfied: packaging>=20.0 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (24.0)
Requirement already satisfied: pillow>=8 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (10.2.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (3.1.2)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from pandas>=1.2->seaborn) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from pandas>=1.2->seaborn) (2024.1)
Requirement already satisfied: six>=1.5 in c:\users\gupta\desktop\plant disease detection\tfenv\lib\site-packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.4->seaborn) (1.16.0)
```

```
In [2]:  
import tensorflow as tf  
import matplotlib.pyplot as plt  
import pandas as pd  
import seaborn as sns
```

Data Preprocessing

Training Image Preprocessing

```
In [3]: training_set = tf.keras.utils.image_dataset_from_directory(  
    'train',  
    labels="inferred",  
    label_mode="categorical",  
    class_names=None,  
    color_mode="rgb",  
    batch_size=32,  
    image_size=(128, 128),  
    shuffle=True,  
    seed=None,  
    validation_split=None,  
    subset=None,  
    interpolation="bilinear",  
    follow_links=False,  
    crop_to_aspect_ratio=False,  
    #      pad_to_aspect_ratio=False,  
    #      data_format=None,  
    #      verbose=True,  
)
```

Found 70295 files belonging to 38 classes.

Validation Image Preprocessing

```
In [4]: validation_set = tf.keras.utils.image_dataset_from_directory(  
    'valid',  
    labels="inferred",  
    label_mode="categorical",  
    class_names=None,  
    color_mode="rgb",  
    batch_size=32,  
    image_size=(128, 128),  
    shuffle=True,  
    seed=None,  
    validation_split=None,  
    subset=None,
```

```
interpolation="bilinear",
follow_links=False,
crop_to_aspect_ratio=False,
#     pad_to_aspect_ratio=False,
#     data_format=None,
#     verbose=True,
)
```

Found 17572 files belonging to 38 classes.

In [5]: `training_set`

Out[5]: <`_PrefetchDataset` element_spec=(`TensorSpec(shape=(None, 128, 128, 3), dtype=tf.float32, name=None)`, `TensorSpec(shape=(None, 38), dtype=tf.float32, name=None)`)>

In [6]: `for x,y in training_set:`
 `print(x)`
 `print(y)`
 `break`

```
tf.Tensor(  
[[[[ 79.75  68.25  76.25]  
[146.25 131.   137.75]  
[147.   128.75 132.   ]  
...  
[174.5  162.5  162.5 ]  
[172.   160.   160.   ]  
[176.   164.   164.   ]]  
  
[[ 72.25  60.75  66.25]  
[136.   121.25 124.   ]  
[150.75 134.5  134.   ]  
...  
[174.5  162.5  162.5 ]  
[160.   148.   148.   ]  
[179.5  167.5  167.5 ]]  
  
[[ 76.25  67.   67.25]  
[152.75 140.75 139.   ]  
[138.   123.75 116.   ]  
...  
[183.5  171.5  171.5 ]  
[175.   163.   163.   ]  
[167.   155.   155.   ]]  
  
...  
  
[[ 70.5   60.5   63.5 ]  
[140.75 129.75 133.75]  
[152.   141.   145.   ]  
...  
[163.75 152.75 156.75]  
[170.5   159.5   163.5 ]  
[166.   155.   159.   ]]  
  
[[ 81.75  71.75  74.75]  
[140.25 129.25 133.25]  
[147.   136.   140.   ]  
...  
[158.25 147.25 151.25]  
[174.   163.   167.   ]  
[172.75 161.75 165.75]]  
  
[[ 74.   64.   67.   ]  
[140.   129.   133.   ]  
[142.25 131.25 135.25]  
...  
[170.75 159.75 163.75]  
[163.   152.   156.   ]  
[163.5  152.5  156.5 ]]]  
  
[[[[156.   151.5  162.   ]  
[153.5  151.   165.   ]  
[146.   146.   164.   ]  
...  
[ 90.25  83.25  99.25]  
[ 92.75  85.75 101.75]  
[ 94.25  87.25 103.25]]]  
  
[[[158.75 154.25 165.75]  
[152.   149.25 163.5 ]  
[150.5  150.5  168.5 ]  
...  
[ 86.75  79.75  95.75]  
[ 87.25  80.25  96.25]  
[ 87.75  80.75  96.75]]]  
  
[[[162.   157.5  169.   ]  
[155.5  153.   167.25]  
[154.75 153.75 171.25]  
...  
[ 88.5   81.5   97.5 ]  
[ 89.5   82.5   98.5 ]  
[ 89.75  82.75  98.75]]]  
  
...  
  
[[[168.75 161.75 168.75]  
[169.   162.   169.   ]  
[170.   163.   170.   ]  
...  
[153.25 148.25 154.25]  
[154.25 149.25 155.25]  
[157.   152.   158.   ]]]  
  
[[[167.5  160.5  167.5 ]  
[169.75 162.75 169.75]  
[168.25 161.25 168.25]  
...  
[154.5  149.5  155.5 ]  
[155.   150.   156.   ]  
[155.75 150.75 156.75]]]
```

```
[[166. 159. 166. ]
 [169.5 162.5 169.5 ]
 [166. 159. 166. ]
 ...
 [156.25 151.25 157.25]
 [157.5 152.5 158.5 ]
 [156.75 151.75 157.75]]]
```

```
[[[245.25 237.25 234.25]
 [244.25 236.25 233.25]
 [250.75 242.75 239.75]
 ...
 [224.75 213.75 209.75]
 [232.5 221.5 217.5 ]
 [235.75 224.75 220.75]]]
```

```
[[250. 242. 239. ]
 [244. 236. 233. ]
 [245.25 237.25 234.25]
 ...
 [234.25 223.25 219.25]
 [234.5 223.5 219.5 ]
 [232.25 221.25 217.25]]
```

```
[[247.25 239.25 236.25]
 [248.75 240.75 237.75]
 [247.25 239.25 236.25]
 ...
 [231. 220. 216. ]
 [229.5 218.5 214.5 ]
 [233. 222. 218. ]]
```

...

```
[[242. 233. 228. ]
 [245.75 237.25 232.25]
 [237.25 228.25 223.25]
 ...
 [200.5 182.5 178.5 ]
 [213. 195. 191. ]
 [236. 218. 214. ]]
```

```
[[247.5 239.75 234.75]
 [236.5 227.5 222.5 ]
 [247.75 238.75 233.75]
 ...
 [207.25 189.25 185.25]
 [210. 192. 188. ]
 [220.75 202.75 198.75]]
```

```
[[239.5 230.5 225.5 ]
 [244.25 235.25 230.25]
 [238.5 229.5 224.5 ]
 ...
 [215.75 197.75 193.75]
 [215.75 197.75 193.75]
 [234.75 217. 213. ]]]
```

...

```
[[[153. 143. 141. ]
 [158.5 148.5 146.5 ]
 [157.5 147.5 145.5 ]
 ...
 [133.5 122.5 120.5 ]
 [128. 117. 115. ]
 [142.75 131.75 129.75]]]
```

```
[[159.25 149.25 147.25]
 [158.25 148.25 146.25]
 [165.75 155.75 153.75]
 ...
 [148. 137. 135. ]
 [147.25 136.25 134.25]
 [146. 135. 133. ]]]
```

```
[[162.25 152.25 150.25]
 [167. 157. 155. ]
 [154.25 144.25 142.25]
 ...
 [152.25 141.25 139.25]
 [138. 127. 125. ]
 [143. 132. 130. ]]]
```

...

```
[[188.75 184.75 185.75]
 [194. 190. 191. ]
 [193.25 189.25 190.25]]
```

...
[171.5 167.5 168.5]
[181.5 177.5 178.5]
[181. 177. 178.]]

[[187. 183. 184.]
[187. 183. 184.]
[183. 179. 180.]
...
[168.5 164.5 165.5]
[170.75 166.75 167.75]
[172.75 168.75 169.75]]

[[187. 183. 184.]
[193.5 189.5 190.5]
[191.25 187.25 188.25]
...
[175.5 171.5 172.5]
[188.75 184.75 185.75]
[177. 173. 174.]]]

[[[147. 136. 144.]
[150. 139. 147.]
[149.75 138.75 146.75]
...
[132.75 112.75 121.75]
[133.75 113.75 122.75]
[145.25 125.25 134.25]]

[[155.75 144.75 152.75]
[145.75 134.75 142.75]
[149. 138. 146.]
...
[134.75 114.75 123.75]
[129.75 109.75 118.75]
[125.5 105.5 114.5]]

[[143.25 132.25 140.25]
[148.5 137.5 145.5]
[139.25 128.25 136.25]
...
[135.5 118.5 126.5]
[133.5 116.5 124.5]
[123.75 106.75 114.75]]

...
[[153.75 141.75 153.75]
[152.75 140.75 152.75]
[152.25 140.25 152.25]
...
[111.25 99.25 109.25]
[111. 99. 109.]
[115.5 103.5 113.5]]

[[149.5 137.5 149.5]
[148.5 136.5 148.5]
[149.25 137.25 149.25]
...
[117. 105. 115.]
[120.75 108.75 118.75]
[116.5 104.5 114.5]]

[[148. 136. 148.]
[145.5 133.5 145.5]
[146.25 134.25 146.25]
...
[112.75 100.75 110.75]
[117. 105. 115.]
[113. 101. 111.]]]

[[[185.5 182.5 177.5]
[186.75 183.75 178.75]
[187.5 184.5 179.5]
...
[150. 147. 140.]
[150.75 147.75 140.75]
[147.5 144.5 137.5]]

[[185.75 182.75 177.75]
[187.75 184.75 179.75]
[188.75 185.75 180.75]
...
[147.25 144.25 137.25]
[145.25 142.25 135.25]
[148. 145. 138.]]

[[186.5 183.5 178.5]
[188.75 185.75 180.75]
[191. 188. 183.]
...]

```
[146.75 143.75 136.75]
[145.75 142.75 135.75]
[138.    135.    128.  ]]
...
[[123.25 119.25 110.25]
 [117.25 113.25 104.25]
 [111.75 107.75  98.75]
 ...
[108.25 102.25  90.25]
[108.25 102.25  90.25]
[117.    111.    99.  ]]
...
[[108.    104.    95.  ]
 [113.5   109.5   100.5 ]
 [115.5   111.5   102.5 ]
 ...
[104.25  98.25  86.25]
[108.75  102.75  90.75]
[111.5   105.5   93.5 ]]
...
[[112.25 108.25  99.25]
 [109.25 105.25  96.25]
 [107.5   103.5   94.5 ]
 ...
[103.    97.    85.  ]
[105.75  99.75  87.75]
[104.5   98.5   86.5 ]]], shape=(32, 128, 128, 3), dtype=float32)
tf.Tensor(
[[0. 1. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 ...
[0. 0. 0. ... 0. 0. 0.]
[0. 0. 0. ... 0. 0. 0.]
[0. 0. 0. ... 0. 0. 0.]], shape=(32, 38), dtype=float32)
```

```
In [7]: for x,y in training_set:
    print(x,x.shape)
    print(y,y.shape)
    break
```

```
tf.Tensor(  
[[[[[225.5 211.5 210.5 ]  
[225.75 211.75 210.75]  
[226. 212. 211. ]  
...  
[228.25 214.25 213.25]  
[225.25 211.25 210.25]  
[221.25 207.25 206.25]]  
  
[[235.75 221.75 220.75]  
[228.75 214.75 213.75]  
[225.75 211.75 210.75]  
...  
[236.25 222.25 221.25]  
[231. 217. 216. ]  
[228. 214.25 213.25]]  
  
[[234. 220. 219. ]  
[241. 227. 226. ]  
[229.25 215.25 214.25]  
...  
[225.75 211.75 210.75]  
[227.5 213.5 212.5 ]  
[227.25 213.25 212.25]]  
  
...  
  
[[219. 205. 202. ]  
[218.5 204.5 201.5 ]  
[225.75 211.75 208.75]  
...  
[234.25 220.25 217.25]  
[227.25 213.25 210.25]  
[234.25 220.25 217.25]]  
  
[[233.25 220.75 217.75]  
[225.75 211.75 208.75]  
[215.25 201.25 198.25]  
...  
[232.25 220.75 217.75]  
[236.25 222.25 219.25]  
[219.75 205.75 202.75]]  
  
[[[231. 217. 214. ]  
[218.5 204.5 201.5 ]  
[229. 215. 212. ]  
...  
[230.5 216.5 213.5 ]  
[227.25 213.25 210.25]  
[229. 215. 212. ]]]  
  
[[[[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
...  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]]  
  
[[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
...  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]]  
  
[[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
...  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]]  
  
...  
  
[[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
...  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]]  
  
[[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
...  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]  
[ 0. 0. 0. ]]]
```

```
[[ 0.    0.    0.  ]
 [ 0.    0.    0.  ]
 [ 0.    0.    0.  ]
 ...
 [ 0.    0.    0.  ]
 [ 0.    0.    0.  ]
 [ 0.    0.    0.  ]]]
```

```
[[[150.5 142.5 153.5 ]
 [131.75 123.75 134.75]
 [127.75 119.75 130.75]
 ...
 [162.25 154.25 165.25]
 [166.    158.    169.  ]
 [158.75 150.75 161.75]]]
```

```
[[140.25 132.25 143.25]
 [125.25 117.25 128.25]
 [137.5 129.5 140.5 ]
 ...
 [160.    152.    163.  ]
 [153.5 145.5 156.5 ]
 [159.25 151.25 162.25]]
```

```
[[145.25 137.25 148.25]
 [160.75 152.75 163.75]
 [142.    134.    145.  ]
 ...
 [150.75 142.75 153.75]
 [156.25 148.25 159.25]
 [157.    149.    160.  ]]
```

...

```
[[186.25 181.25 188.25]
 [172.5 167.5 174.5 ]
 [174.25 169.25 176.25]
 ...
 [182.    176.    186.  ]
 [177.75 171.75 181.75]
 [174.5 168.5 178.5 ]]]
```

```
[[170.25 165.25 172.25]
 [171.75 166.75 173.75]
 [191.5 186.5 193.5 ]
 ...
 [173.    167.    177.  ]
 [172.25 166.25 176.25]
 [180.5 174.5 184.5 ]]]
```

```
[[167.    162.    169.  ]
 [174.25 169.25 176.25]
 [179.75 174.75 181.75]
 ...
 [175.25 169.25 179.25]
 [173.25 167.25 177.25]
 [175.75 169.75 179.75]]]]
```

...

```
[[[144.75 139.75 135.75]
 [141.25 136.25 132.25]
 [153.75 148.75 144.75]
 ...
 [173.25 171.25 172.25]
 [175.75 173.75 174.75]
 [176.25 174.25 175.25]]]
```

```
[[153.75 148.75 144.75]
 [147.5 142.5 138.5 ]
 [149.5 144.5 140.5 ]
 ...
 [167.75 165.75 166.75]
 [167.75 165.75 166.75]
 [170.5 168.5 169.5 ]]]
```

```
[[150.25 145.25 141.25]
 [137.25 132.25 128.25]
 [131.75 126.75 122.75]
 ...
 [171.75 169.75 170.75]
 [175.25 173.25 174.25]
 [174.75 172.75 173.75]]]
```

...

```
[[109.75 100.75 95.75]
 [111.25 102.25 97.25]
 [112.5 103.5 98.5 ]]
```

...
[148.5 143.5 139.5]
[155. 150. 146.]
[159. 154. 150.]]

[[109.25 100.25 95.25]
[110.25 101.25 96.25]
[110.5 101.5 96.5]
...
[151.75 146.75 142.75]
[150.25 145.25 141.25]
[152.75 147.75 143.75]]

[[118. 109. 104.]
[112.75 103.75 98.75]
[109. 100. 95.]
...
[150.5 145.5 141.5]
[147. 142. 138.]
[147.5 142.5 138.5]]]

[[[124.75 116.75 114.75]
[124.75 116.75 114.75]
[121.75 113.75 111.75]
...
[118.75 108.75 107.75]
[119. 109. 108.]
[120.5 110.5 109.5]]

[[[132.5 124.5 122.5]
[120.75 112.75 110.75]
[126. 118. 116.]
...
[115.75 105.75 104.75]
[114. 104. 103.]
[116.25 106.25 105.25]]

[[[116.25 108.25 106.25]
[123.25 115.25 113.25]
[124.5 116.5 114.5]
...
[124. 114. 113.]
[119.5 109.5 108.5]
[116.25 106.25 105.25]]

...
[[[158.5 154.5 155.5]
[155.25 151.25 152.25]
[155. 151. 152.]
...
[154.5 149.5 146.5]
[154.5 149.5 146.5]
[155.25 150.25 147.25]]

[[[160. 156. 157.]
[156. 152. 153.]
[154.75 150.75 151.75]
...
[154.75 149.75 146.75]
[156.25 151.25 148.25]
[157.75 152.75 149.75]]

[[[156. 152. 153.]
[156.25 152.25 153.25]
[156.5 152.5 153.5]
...
[154.75 149.75 146.75]
[152.75 147.75 144.75]
[148.5 143.5 140.5]]]

[[[155. 147. 158.]
[153.75 145.75 156.75]
[151.75 143.75 154.75]
...
[76. 76.75 25.5]
[92.5 90.25 32.]
[108. 102. 39.75]]

[[[153.75 145.75 156.75]
[152.75 144.75 155.75]
[154. 146. 157.]
...
[89.25 96. 47.25]
[91. 96. 41.25]
[91.5 96. 37.5]]]

[[[153.5 145.5 156.5]
[150. 142. 153.]
[149. 141. 152.]
...]

```

[ 73.5  90.25 44.75]
[ 73.    89.75 39.25]
[ 80.5  96.5  44.5 ]]

...
[[105.25 97.25 108.25]
 [103.25 95.25 106.25]
 [107.25 99.25 110.25]
 ...
 [ 49.25 66.75 29. ]
 [ 51.    68.75 31. ]
 [ 58.5   76.25 38.5 ]]

[[103.25 95.25 106.25]
 [106.5   98.5   109.5 ]
 [102.25 94.25 105.25]
 ...
 [ 36.    50.    13.75]
 [ 50.25 63.25 27.25]
 [ 54.    66.5   30.75]]

[[ 98.75 90.75 101.75]
 [ 94.5   86.5   97.5 ]
 [ 98.75 90.75 101.75]
 ...
 [ 56.25 68.75 32.75]
 [ 48.75 57.25 22.75]
 [ 58.25 66.    31.75]]]], shape=(32, 128, 128, 3), dtype=float32) (32, 128, 128, 3)
tf.Tensor(
[[0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 ...
 [0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]], shape=(32, 38), dtype=float32) (32, 38)

```

To avoid overshooting

1. Choose small learning rate default 0.001 , we are taking 0.0001
2. There may be chance of underfitting, so increase number of neuron
3. Add more Convolution layer to extract more important feature from images, there may be possibility that model unable to capture relevant feature or model is confusing due to lack of feature so feed with more feature

Building Model

```
In [8]: from tensorflow.keras.layers import Dense,Conv2D,MaxPool2D,Flatten, Dropout
from tensorflow.keras.models import Sequential
```

```
In [9]: model=Sequential()
```

Building Convolutional Layer

```
In [10]: model.add(Conv2D(filters=32, kernel_size=3, padding='same', activation='relu', input_shape=[128, 128, 3]))
model.add(Conv2D(filters=32,kernel_size=3,activation='relu'))
model.add(MaxPool2D(pool_size=2,strides=2))
```

C:\Users\gupta\Desktop\plant disease detection\tfenv\lib\site-packages\keras\src\layers\convolutional\base_conv.py:99: 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__()

```
In [11]: model.add(Conv2D(filters=64,kernel_size=3,padding='same',activation='relu'))
model.add(Conv2D(filters=64,kernel_size=3,activation='relu'))
model.add(MaxPool2D(pool_size=2,strides=2))
```

```
In [12]: model.add(Conv2D(filters=128,kernel_size=3,padding='same',activation='relu'))
model.add(Conv2D(filters=128,kernel_size=3,activation='relu'))
model.add(MaxPool2D(pool_size=2,strides=2))
```

```
In [13]: model.add(Conv2D(filters=256,kernel_size=3,padding='same',activation='relu'))
model.add(Conv2D(filters=256,kernel_size=3,activation='relu'))
model.add(MaxPool2D(pool_size=2,strides=2))
```

```
In [14]: model.add(Conv2D(filters=512,kernel_size=3,padding='same',activation='relu'))
model.add(Conv2D(filters=512,kernel_size=3,activation='relu'))
model.add(MaxPool2D(pool_size=2,strides=2))
```

```
In [15]: model.add(Dropout(0.25)) #To avoid overfitting
```

```
In [16]: model.add(Flatten())
```

```
In [17]: model.add(Dense(units=1500,activation='relu'))
```

```
In [18]: model.add(Dropout(0.4))
```

```
In [19]: ## Output Layer  
model.add(Dense(units=38,activation='softmax'))
```

Compiling Model

```
In [20]: model.compile(optimizer=tf.keras.optimizers.Adam(  
    learning_rate=0.0001),loss='categorical_crossentropy',metrics=['accuracy'])
```

```
In [21]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 128, 128, 32)	896
conv2d_1 (Conv2D)	(None, 126, 126, 32)	9,248
max_pooling2d (MaxPooling2D)	(None, 63, 63, 32)	0
conv2d_2 (Conv2D)	(None, 63, 63, 64)	18,496
conv2d_3 (Conv2D)	(None, 61, 61, 64)	36,928
max_pooling2d_1 (MaxPooling2D)	(None, 30, 30, 64)	0
conv2d_4 (Conv2D)	(None, 30, 30, 128)	73,856
conv2d_5 (Conv2D)	(None, 28, 28, 128)	147,584
max_pooling2d_2 (MaxPooling2D)	(None, 14, 14, 128)	0
conv2d_6 (Conv2D)	(None, 14, 14, 256)	295,168
conv2d_7 (Conv2D)	(None, 12, 12, 256)	590,080
max_pooling2d_3 (MaxPooling2D)	(None, 6, 6, 256)	0
conv2d_8 (Conv2D)	(None, 6, 6, 512)	1,180,160
conv2d_9 (Conv2D)	(None, 4, 4, 512)	2,359,808
max_pooling2d_4 (MaxPooling2D)	(None, 2, 2, 512)	0
dropout (Dropout)	(None, 2, 2, 512)	0
flatten (Flatten)	(None, 2048)	0
dense (Dense)	(None, 1500)	3,073,500
dropout_1 (Dropout)	(None, 1500)	0
dense_1 (Dense)	(None, 38)	57,038

Total params: 7,842,762 (29.92 MB)

Trainable params: 7,842,762 (29.92 MB)

Non-trainable params: 0 (0.00 B)

Model Training

```
In [22]: training_history=model.fit(x=training_set,validation_data=validation_set,epochs=10)
```

```
Epoch 1/10  
2197/2197 2537s 1s/step - accuracy: 0.4113 - loss: 2.0918 - val_accuracy: 0.8525 - val_loss: 0.4730  
Epoch 2/10  
2197/2197 4921s 2s/step - accuracy: 0.8420 - loss: 0.5009 - val_accuracy: 0.9230 - val_loss: 0.2429  
Epoch 3/10  
2197/2197 2447s 1s/step - accuracy: 0.9105 - loss: 0.2775 - val_accuracy: 0.9383 - val_loss: 0.1906  
Epoch 4/10  
2197/2197 3893s 2s/step - accuracy: 0.9350 - loss: 0.1960 - val_accuracy: 0.9451 - val_loss: 0.1671  
Epoch 5/10  
2197/2197 2301s 1s/step - accuracy: 0.9528 - loss: 0.1431 - val_accuracy: 0.9573 - val_loss: 0.1397  
Epoch 6/10  
2197/2197 2292s 1s/step - accuracy: 0.9627 - loss: 0.1119 - val_accuracy: 0.9642 - val_loss: 0.1125  
Epoch 7/10  
2197/2197 2272s 1s/step - accuracy: 0.9711 - loss: 0.0880 - val_accuracy: 0.9568 - val_loss: 0.1393  
Epoch 8/10  
2197/2197 2343s 1s/step - accuracy: 0.9748 - loss: 0.0779 - val_accuracy: 0.9653 - val_loss: 0.1217  
Epoch 9/10  
2197/2197 2237s 1s/step - accuracy: 0.9771 - loss: 0.0690 - val_accuracy: 0.9706 - val_loss: 0.0947  
Epoch 10/10  
2197/2197 2353s 1s/step - accuracy: 0.9816 - loss: 0.0581 - val_accuracy: 0.9525 - val_loss: 0.1637
```

Model Evaluation

```
In [23]: #Model Evaluation on Training set  
train_loss, train_acc = model.evaluate(training_set)  
#Training set Accuracy  
print('Training accuracy:', train_acc)  
  
2197/2197 ————— 719s 327ms/step - accuracy: 0.9815 - loss: 0.0584  
Training accuracy: 0.9813784956932068
```

```
In [24]: print(train_loss,train_acc)  
  
0.05601634085178375 0.9813784956932068
```

```
In [33]: #Validation set Accuracy  
val_loss, val_acc = model.evaluate(validation_set)  
print('Validation accuracy:', val_acc)  
  
550/550 ————— 176s 319ms/step - accuracy: 0.9524 - loss: 0.1655  
Validation accuracy: 0.952481210231781
```

Saving Model

```
In [38]: model.save("trained_model.h5")  
model.save('trained_plant_disease_model.keras') #it is used to compress file size
```

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

```
In [39]: training_history.history #Return Dictionary of history
```

```
Out[39]: {'accuracy': [0.6125613451004028,  
 0.8652962446212769,  
 0.9173910021781921,  
 0.9422718286514282,  
 0.9566398859024048,  
 0.9648481607437134,  
 0.9728999137878418,  
 0.9761576056480408,  
 0.9793157577514648,  
 0.9830998182296753],  
 'loss': [1.31917142868042,  
 0.4199684262275696,  
 0.25422152876853943,  
 0.1763271987438202,  
 0.132105752825737,  
 0.10680293291807175,  
 0.08436506986618042,  
 0.07233617454767227,  
 0.06455418467521667,  
 0.054157961159944534],  
 'val_accuracy': [0.8525494933128357,  
 0.9230024814605713,  
 0.9383109211921692,  
 0.9450830817222595,  
 0.9573184847831726,  
 0.9642044305801392,  
 0.9568063020706177,  
 0.9652856588363647,  
 0.9706351161003113,  
 0.952481210231781],  
 'val_loss': [0.47296813130378723,  
 0.24292133748531342,  
 0.19062699377536774,  
 0.16713154315948486,  
 0.1396910697221756,  
 0.11251150071620941,  
 0.1393156349658966,  
 0.12166524678468704,  
 0.09469074755907059,  
 0.16374102234840393]}
```

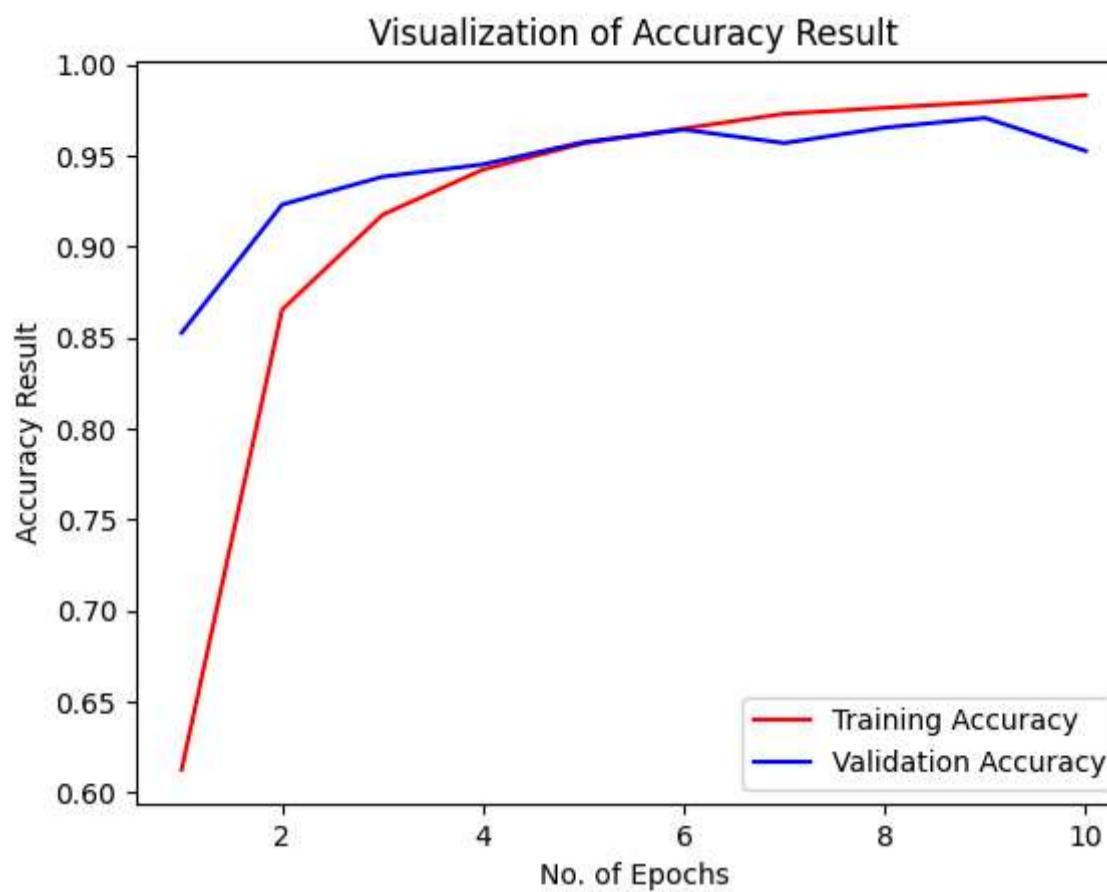
```
In [40]: #Recording History in json  
import json  
with open('training_hist.json', 'w') as f:  
    json.dump(training_history.history,f)
```

```
In [41]: print(training_history.history.keys())  
  
dict_keys(['accuracy', 'loss', 'val_accuracy', 'val_loss'])
```

Accuracy Visualization

```
In [43]: epochs = [i for i in range(1,11)]  
plt.plot(epochs,training_history.history['accuracy'],color='red',label='Training Accuracy')  
plt.plot(epochs,training_history.history['val_accuracy'],color='blue',label='Validation Accuracy')  
plt.xlabel('No. of Epochs')  
plt.ylabel('Accuracy Result')  
plt.title('Visualization of Accuracy Result')
```

```
plt.legend()  
plt.show()
```



Some other metrics for model evaluation

```
In [44]: class_name = validation_set.class_names  
class_name
```

```
Out[44]: ['Apple__Apple_scab',  
          'Apple__Black_rot',  
          'Apple__Cedar_apple_rust',  
          'Apple__healthy',  
          'Blueberry__healthy',  
          'Cherry_(including_sour)__Powdery_mildew',  
          'Cherry_(including_sour)__healthy',  
          'Corn_(maize)__Cercospora_leaf_spot_Gray_leaf_spot',  
          'Corn_(maize)__Common_rust',  
          'Corn_(maize)__Northern_Leaf_Blight',  
          'Corn_(maize)__healthy',  
          'Grape__Black_rot',  
          'Grape__Esca_(Black_Measles)',  
          'Grape__Leaf_blight_(Isariopsis_Leaf_Spot)',  
          'Grape__healthy',  
          'Orange__Haunglongbing_(Citrus_greening)',  
          'Peach__Bacterial_spot',  
          'Peach__healthy',  
          'Pepper,_bell__Bacterial_spot',  
          'Pepper,_bell__healthy',  
          'Potato__Early_blight',  
          'Potato__Late_blight',  
          'Potato__healthy',  
          'Raspberry__healthy',  
          'Soybean__healthy',  
          'Squash__Powdery_mildew',  
          'Strawberry__Leaf_scorch',  
          'Strawberry__healthy',  
          'Tomato__Bacterial_spot',  
          'Tomato__Early_blight',  
          'Tomato__Late_blight',  
          'Tomato__Leaf_Mold',  
          'Tomato__Septoria_leaf_spot',  
          'Tomato__Spider_mites_Two-spotted_spider_mite',  
          'Tomato__Target_Spot',  
          'Tomato__Tomato_Yellow_Leaf_Curl_Virus',  
          'Tomato__Tomato_mosaic_virus',  
          'Tomato__healthy']
```

```
In [45]: test_set = tf.keras.utils.image_dataset_from_directory(  
           'valid', #not test directory  
           labels="inferred",  
           label_mode="categorical",  
           class_names=None,  
           color_mode="rgb",  
           batch_size=1,  
           image_size=(128, 128),  
           shuffle=False, #not true coz it will be passed sequentially  
           seed=None,  
           validation_split=None,  
           subset=None,  
           interpolation="bilinear",  
           follow_links=False,  
           crop_to_aspect_ratio=False  
)
```

Found 17572 files belonging to 38 classes.

```
In [46]: y_pred = model.predict(test_set)
y_pred, y_pred.shape
```

17572/17572 ————— 521s 30ms/step

```
Out[46]: (array([[9.9999762e-01, 2.34691072e-07, 3.19311439e-12, ...,
   2.67704432e-17, 2.38235979e-15, 3.23502847e-15],
  [9.9999762e-01, 2.06792038e-07, 2.36440988e-12, ...,
   1.38692712e-16, 3.18356385e-14, 4.53157951e-14],
  [1.00000000e+00, 1.19024290e-09, 1.27916408e-14, ...,
   1.14301859e-19, 4.85212699e-17, 1.36849647e-16],
  ...,
  [7.57148833e-09, 4.39102639e-13, 5.03816711e-10, ...,
   1.47297632e-10, 2.62246058e-10, 9.99999166e-01],
  [2.69355804e-09, 1.51434204e-13, 3.35975164e-10, ...,
   2.37843412e-10, 9.09668937e-11, 9.99999881e-01],
  [3.99693619e-16, 8.81078990e-20, 3.40523762e-16, ...,
   1.01956678e-17, 1.06512765e-16, 1.00000000e+00]], dtype=float32),
 (17572, 38))
```

```
In [49]: predicted_categories = tf.argmax(y_pred, axis=1) #argmax() extract maximum value from above prob arr and return its index
#axis=1 means scan vertically
```

```
In [50]: predicted_categories
```

```
Out[50]: <tf.Tensor: shape=(17572,), dtype=int64, numpy=array([ 0,  0,  0, ..., 37, 37, 37], dtype=int64)>
```

```
In [51]: true_categories = tf.concat([y for x, y in test_set], axis=0)
```

```
In [52]: true_categories
```

```
Out[52]: <tf.Tensor: shape=(17572, 38), dtype=float32, numpy=
array([[1., 0., 0., ..., 0., 0., 0.],
       [1., 0., 0., ..., 0., 0., 0.],
       [1., 0., 0., ..., 0., 0., 0.],
       ...,
       [0., 0., 0., ..., 0., 0., 1.],
       [0., 0., 0., ..., 0., 0., 1.],
       [0., 0., 0., ..., 0., 0., 1.]], dtype=float32)>
```

```
In [53]: Y_true = tf.argmax(true_categories, axis=1)
```

```
In [54]: Y_true
```

```
Out[54]: <tf.Tensor: shape=(17572,), dtype=int64, numpy=array([ 0,  0,  0, ..., 37, 37, 37], dtype=int64)>
```

```
In [56]: !pip install scikit-learn
from sklearn.metrics import confusion_matrix,classification_report
```

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Downloading threadpoolctl-3.4.0-py3-none-any.whl (17 kB)

Installing collected packages: threadpoolctl, scipy, joblib, scikit-learn

Successfully installed joblib-1.4.0 scikit-learn-1.4.2 scipy-1.13.0 threadpoolctl-3.4.0

```
In [57]: # Precision Recall Fscore  
print(classification_report(Y_true,predicted_categories,target_names=class_name))
```

		precision	recall	f1-score	support
	Apple__Apple_scab	0.89	0.97	0.93	504
	Apple__Black_rot	0.98	0.98	0.98	497
	Apple__Cedar_apple_rust	0.96	0.98	0.97	440
	Apple__healthy	0.97	0.90	0.93	502
	Blueberry__healthy	0.95	0.99	0.97	454
	Cherry_(including_sour)__Powdery_mildew	0.99	0.97	0.98	421
	Cherry_(including_sour)__healthy	0.99	0.98	0.98	456
Corn_(maize)	Cercospora_leaf_spot_Gray_leaf_spot	0.96	0.89	0.92	410
	Corn_(maize)__Common_rust_	0.99	0.99	0.99	477
	Corn_(maize)__Northern_Leaf_Blight	0.93	0.97	0.95	477
	Corn_(maize)__healthy	0.99	1.00	0.99	465
	Grape__Black_rot	0.95	1.00	0.97	472
	Grape__Esca_(Black_Measles)	1.00	0.95	0.97	480
Grape	Leaf_blight_(Isariopsis_Leaf_Spot)	1.00	1.00	1.00	430
	Grape__healthy	0.98	1.00	0.99	423
Orange	Haunglongbing_(Citrus_greening)	0.99	0.98	0.99	503
	Peach__Bacterial_spot	0.96	0.95	0.96	459
	Peach__healthy	0.97	1.00	0.99	432
Pepper,_bell	Bacterial_spot	0.91	0.99	0.95	478
	Pepper,_bell__healthy	0.88	0.96	0.92	497
	Potato__Early_blight	0.98	0.94	0.96	485
	Potato__Late_blight	0.95	0.94	0.95	485
	Potato__healthy	0.98	0.92	0.95	456
	Raspberry__healthy	1.00	0.88	0.94	445
	Soybean__healthy	0.99	0.94	0.97	505
	Squash__Powdery_mildew	0.99	0.97	0.98	434
	Strawberry__Leaf_scorch	1.00	0.95	0.97	444
	Strawberry__healthy	0.97	0.98	0.98	456
	Tomato__Bacterial_spot	0.99	0.95	0.97	425
	Tomato__Early_blight	0.96	0.84	0.89	480
	Tomato__Late_blight	0.89	0.92	0.91	463
	Tomato__Leaf_Mold	0.93	0.98	0.96	470
	Tomato__Septoria_leaf_spot	0.79	0.93	0.85	436
Tomato	Spider_mites_Two-spotted_spider_mite	0.95	0.92	0.93	435
	Tomato__Target_Spot	0.93	0.72	0.81	457
	Tomato__Tomato_Yellow_Leaf_Curl_Virus	0.98	0.99	0.98	490
	Tomato__Tomato_mosaic_virus	0.97	0.96	0.96	448
	Tomato__healthy	0.82	1.00	0.90	481
	accuracy			0.95	17572
	macro avg	0.96	0.95	0.95	17572
	weighted avg	0.95	0.95	0.95	17572

```
In [60]: cm = confusion_matrix(Y_true,predicted_categories)
#cm.shape
cm
```

```
Out[60]: array([[489,    5,    0, ...,    0,    0,    0],
   [ 1, 489,    0, ...,    0,    0,    0],
   [ 0,    0, 432, ...,    0,    0,    2],
   ...,
   [ 0,    0,    2, ..., 483,    0,    0],
   [ 0,    0,    0, ...,    0, 428,    3],
   [ 0,    0,    0, ...,    0,    0, 481]], dtype=int64)
```

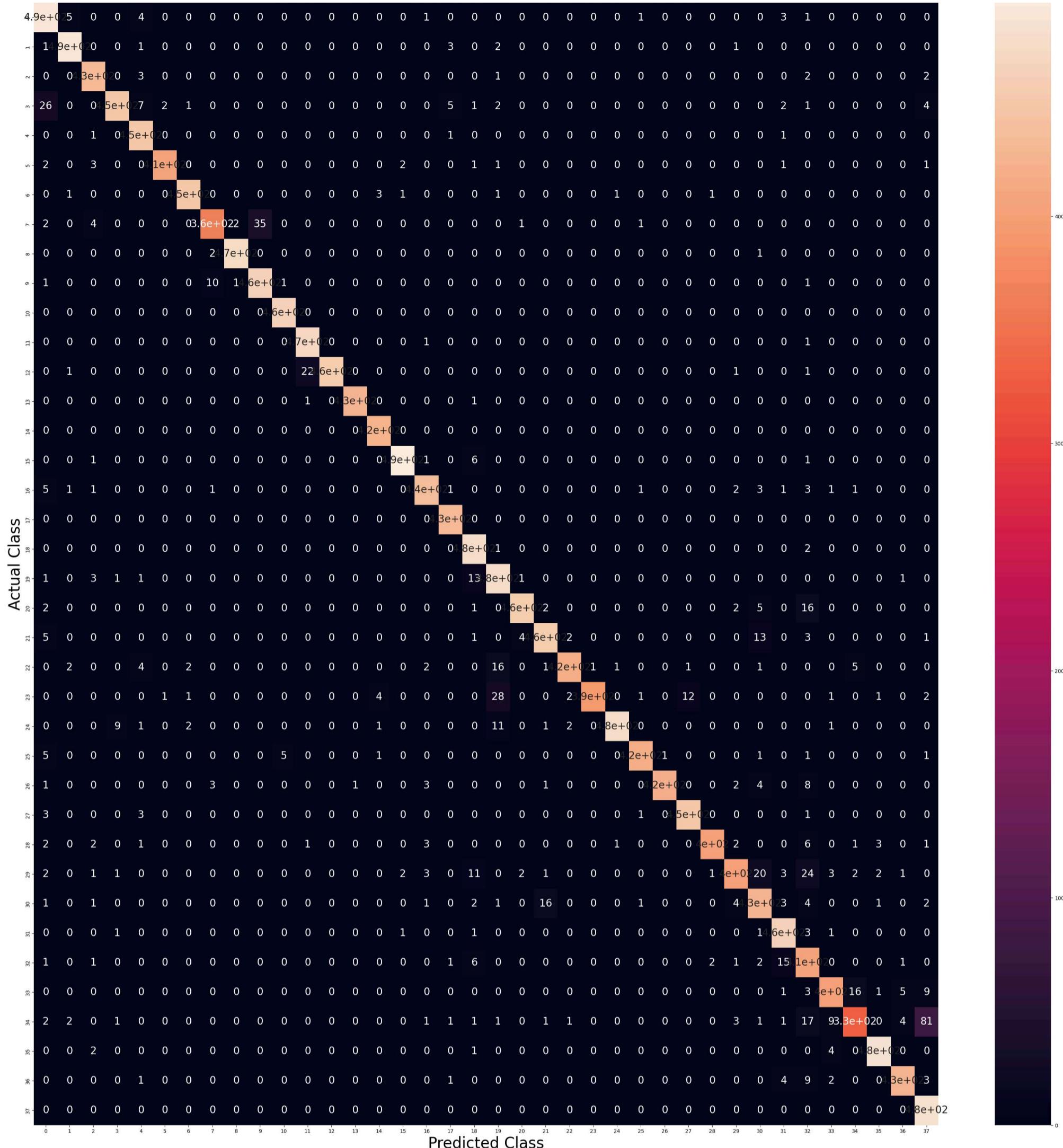
Confusion Matrix Visualization

```
In [62]: #sns.heatmap(cm)
```

```
In [66]: plt.figure(figsize=(40, 40))
sns.heatmap(cm,annot=True,annot_kws={"size": 20})

plt.xlabel('Predicted Class',fontsize = 30)
plt.ylabel('Actual Class',fontsize = 30)
plt.title('Plant Disease Prediction Confusion Matrix',fontsize = 40)
plt.show()
```

Plant Disease Prediction Confusion Matrix



In []:

In []: