

1. Load the Corn 3-Classes image dataset.
2. Preprocess the images: \
  - a. Resize images to a fixed size (e.g., 224×224) \
  - b. Normalize pixel values.
3. Split the dataset into training and testing sets.
4. Create a CNN model using: \
  - a. Convolution layer \
  - b. Max Pooling layer \
  - c. Flatten layer \
  - d. Dense layer
5. Train the CNN model for multi-class classification.
6. Test the model on unseen images.
7. Plot graphs: \
  - a. Training vs Validation Accuracy \
  - b. Training vs Validation Loss (Error)

```
In [1]: import numpy as np
import pandas as pd
import tensorflow as tf
```

```
2026-01-06 12:39:09.095358: I tensorflow/core/util/port.cc:153] oneDNN custom
operations are on. You may see slightly different numerical results due to
floating-point round-off errors from different computation orders. To turn
them off, set the environment variable `TF_ENABLE_ONEDNN_OPTS=0`.
2026-01-06 12:39:09.102589: E external/local_xla/xla/stream_executor/cuda/
cuda_fft.cc:467] Unable to register cuFFT factory: Attempting to register
factory for plugin cuFFT when one has already been registered
WARNING: All log messages before absl::InitializeLog() is called are written
to STDERR
E0000 00:00:1767683349.111243 227125 cuda_dnn.cc:8579] Unable to register
cuDNN factory: Attempting to register factory for plugin cuDNN when one has
already been registered
E0000 00:00:1767683349.113840 227125 cuda_blas.cc:1407] Unable to register
cuBLAS factory: Attempting to register factory for plugin cuBLAS when one has
already been registered
W0000 00:00:1767683349.120493 227125 computation_placer.cc:177] computation
placer already registered. Please check linkage and avoid linking the same
target more than once.
W0000 00:00:1767683349.120504 227125 computation_placer.cc:177] computation
placer already registered. Please check linkage and avoid linking the same
target more than once.
W0000 00:00:1767683349.120505 227125 computation_placer.cc:177] computation
placer already registered. Please check linkage and avoid linking the same
target more than once.
W0000 00:00:1767683349.120506 227125 computation_placer.cc:177] computation
placer already registered. Please check linkage and avoid linking the same
target more than once.
2026-01-06 12:39:09.122645: I tensorflow/core/platform/
cpu_feature_guard.cc:210] This TensorFlow binary is optimized to use
available CPU instructions in performance-critical operations.
To enable the following instructions: AVX2 AVX_VNNI FMA, in other operations,
rebuild TensorFlow with the appropriate compiler flags.
```

```
In [2]: from tensorflow.keras.preprocessing import image_dataset_from_directory
```

```
In [ ]: train_dataset = image_dataset_from_directory(
        'Data/Corn_3_Classes_Image_Dataset',
        image_size=(224, 224),
        batch_size=32,
        label_mode='int',
        validation_split=0.2,
        subset='training',
        seed=123
    )

    val_dataset = image_dataset_from_directory(
        'Data/Corn_3_Classes_Image_Dataset',
        image_size=(224, 224),
        batch_size=32,
        label_mode='int',
        validation_split=0.2,
        subset='validation',
        seed=123
    )

    classes = train_dataset.class_names
    print("Classes:", classes)
```

Found 1050 files belonging to 3 classes.  
Using 840 files for training.

I0000 00:00:1767683350.500896 227125 gpu\_device.cc:2019] Created device /  
job:localhost/replica:0/task:0/device:GPU:0 with 9286 MB memory: -> device:  
0, name: NVIDIA GeForce RTX 4070 SUPER, pci bus id: 0000:01:00.0, compute  
capability: 8.9

Found 1050 files belonging to 3 classes.  
Using 210 files for validation.  
Classes: ['Zea\_mays\_Chulpi\_Cancha', 'Zea\_mays\_Indurata', 'Zea\_mays\_Rugosa']

```
In [4]: from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense, Conv2D, MaxPooling2D, Flatten ,
        BatchNormalization
        from tensorflow.keras.optimizers import Adam
```

```
In [5]: model = Sequential([
        tf.keras.layers.Input(shape=(224, 224, 3)),
        tf.keras.layers.Rescaling(1./255),
        Conv2D(32, (3, 3), activation='relu'),
        MaxPooling2D((2, 2)),
        Conv2D(64, (3, 3), activation='relu'),
        MaxPooling2D((2, 2)),
        Flatten(),
        Dense(128, activation='relu'),
        Dense(len(classes), activation='softmax')
    ])
```

```
In [6]: model.compile(optimizer=Adam(), loss='sparse_categorical_crossentropy',
        metrics=['accuracy'])
```

```
In [7]: history = model.fit(train_dataset, validation_data=val_dataset, epochs=10)
```

Epoch 1/10

```
WARNING: All log messages before absl::InitializeLog() is called are written
to STDERR
I0000 00:00:1767683351.776831 227337 service.cc:152] XLA service
0x7f1c840b0bae0 initialized for platform CUDA (this does not guarantee that
XLA will be used). Devices:
I0000 00:00:1767683351.776852 227337 service.cc:160]   StreamExecutor device
(0): NVIDIA GeForce RTX 4070 SUPER, Compute Capability 8.9
2026-01-06 12:39:11.791202: I tensorflow/compiler/mlir/tensorflow/utils/
dump_mlir_util.cc:269] disabling MLIR crash reproducer, set env var
`MLIR_CRASH_REPRODUCER_DIRECTORY` to enable.
I0000 00:00:1767683351.865194 227337 cuda_dnn.cc:529] Loaded cuDNN version
91701
```

```
[1m 5/27[0m [32m[0m[37m[0m [1m0s[0m 25ms/step - accuracy:
0.5364 - loss: 0.9860
```

```
I0000 00:00:1767683354.483140 227337 device_compiler.h:188] Compiled cluster
using XLA! This line is logged at most once for the lifetime of the process.
```

```
[1m27/27[0m [32m[0m[37m[0m [1m0s[0m 71ms/step - accuracy:
0.8236 - loss: 0.4718
```

```
2026-01-06 12:39:17.292341: I external/local_xla/xla/stream_executor/cuda/
subprocess_compilation.cc:346] ptxas warning : Registers are spilled to local
memory in function 'gemm_fusion_dot_72', 4 bytes spill stores, 4 bytes spill
loads
```

```
[1m27/27[0m [32m[0m[37m[0m [1m6s[0m 127ms/step - accuracy:
0.8277 - loss: 0.4622 - val_accuracy: 0.9857 - val_loss: 0.0202
```

Epoch 2/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 25ms/step - accuracy:
0.9961 - loss: 0.0114 - val_accuracy: 0.9952 - val_loss: 0.0046
```

Epoch 3/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 25ms/step - accuracy:
0.9960 - loss: 0.0198 - val_accuracy: 0.9952 - val_loss: 0.0112
```

Epoch 4/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 25ms/step - accuracy:
1.0000 - loss: 0.0011 - val_accuracy: 1.0000 - val_loss: 8.0215e-04
```

Epoch 5/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 25ms/step - accuracy:
1.0000 - loss: 4.6605e-04 - val_accuracy: 1.0000 - val_loss: 9.8933e-04
```

Epoch 6/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 26ms/step - accuracy:
1.0000 - loss: 1.0943e-04 - val_accuracy: 1.0000 - val_loss: 4.7261e-04
```

Epoch 7/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 25ms/step - accuracy:
1.0000 - loss: 4.5000e-05 - val_accuracy: 1.0000 - val_loss: 4.1222e-04
```

Epoch 8/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 26ms/step - accuracy:
1.0000 - loss: 3.7742e-05 - val_accuracy: 1.0000 - val_loss: 3.5011e-04
```

Epoch 9/10

```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 25ms/step - accuracy:
1.0000 - loss: 2.1122e-05 - val_accuracy: 1.0000 - val_loss: 3.0246e-04
```

Epoch 10/10

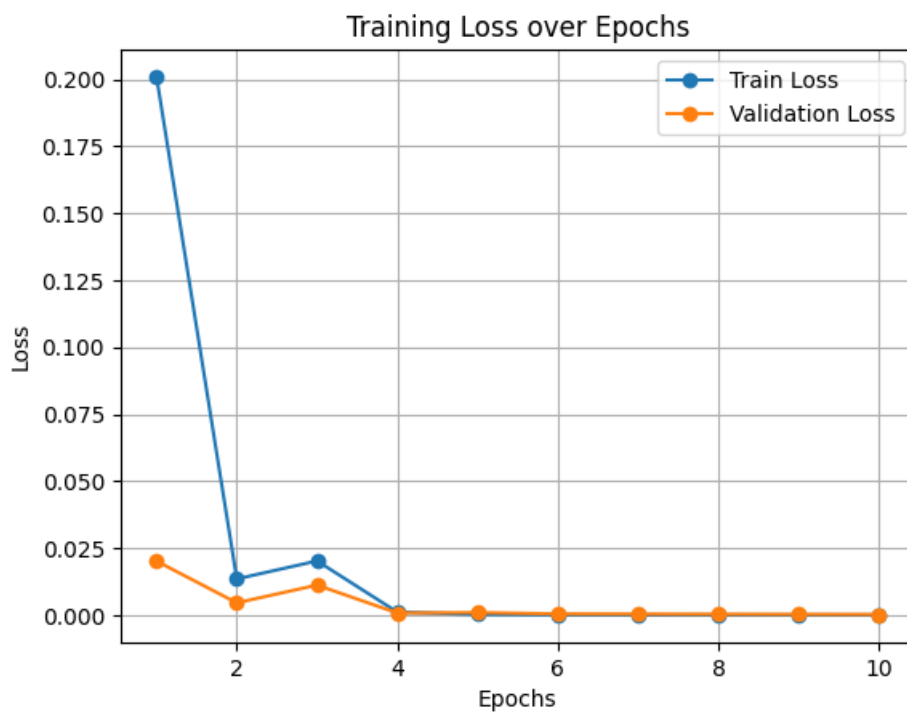
```
[1m27/27[0m [32m[0m[37m[0m [1m1s[0m 26ms/step - accuracy:
1.0000 - loss: 2.3673e-05 - val_accuracy: 1.0000 - val_loss: 2.6323e-04
```

```
In [8]: train_losses = history.history['loss']
        train_accuracies = history.history['accuracy']

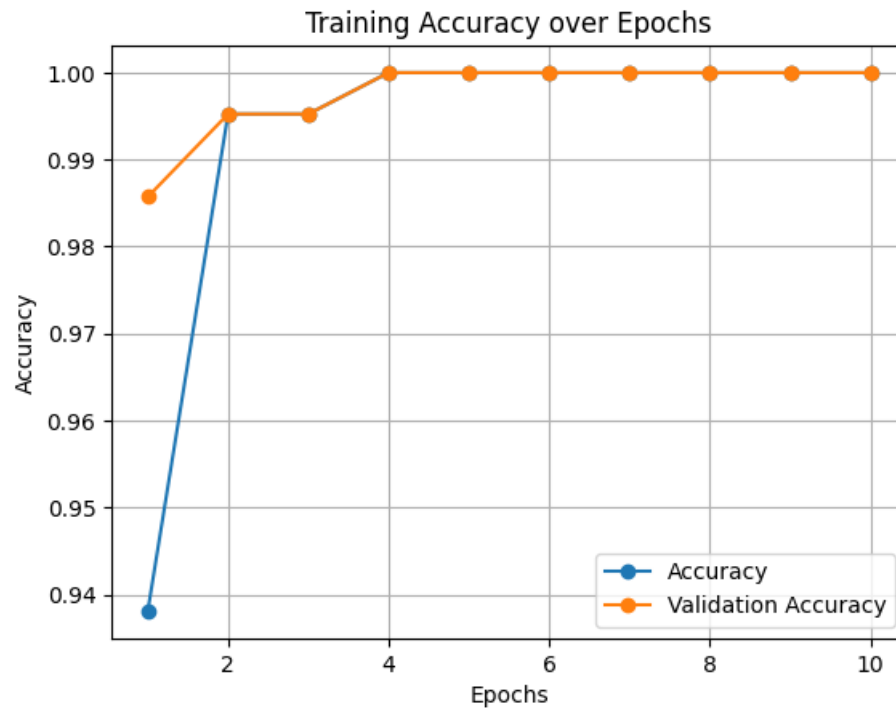
        val_losses = history.history['val_loss']
        val_accuracies = history.history['val_accuracy']
```

```
In      x = np.arange(1, len(train_losses) + 1)
[10]:
```

```
In      import matplotlib.pyplot as plt
[11]: plt.plot(x, train_losses, label='Train Loss', marker='o')
        plt.plot(x, val_losses, label='Validation Loss', marker='o')
        plt.xlabel('Epochs')
        plt.ylabel('Loss')
        plt.title('Training Loss over Epochs')
        plt.legend()
        plt.grid()
        plt.show()
```



```
In [12]: plt.plot(x, train_accuracies, label='Accuracy', marker='o')
plt.plot(x, val_accuracies, label='Validation Accuracy', marker='o')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.title('Training Accuracy over Epochs')
plt.legend()
plt.grid()
plt.show()
```



In [ ]: