



# Plant Seedling Classification:

## CNN, PreTrained ResNet34, VGG and Ensemble Approach



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## INTRODUCTION

## DATA LOADER

Exploratory  
Data  
Analysis  
(EDA)

DATA PRE-  
PROCESSING  
AND  
MODELLING

## EVALUATION

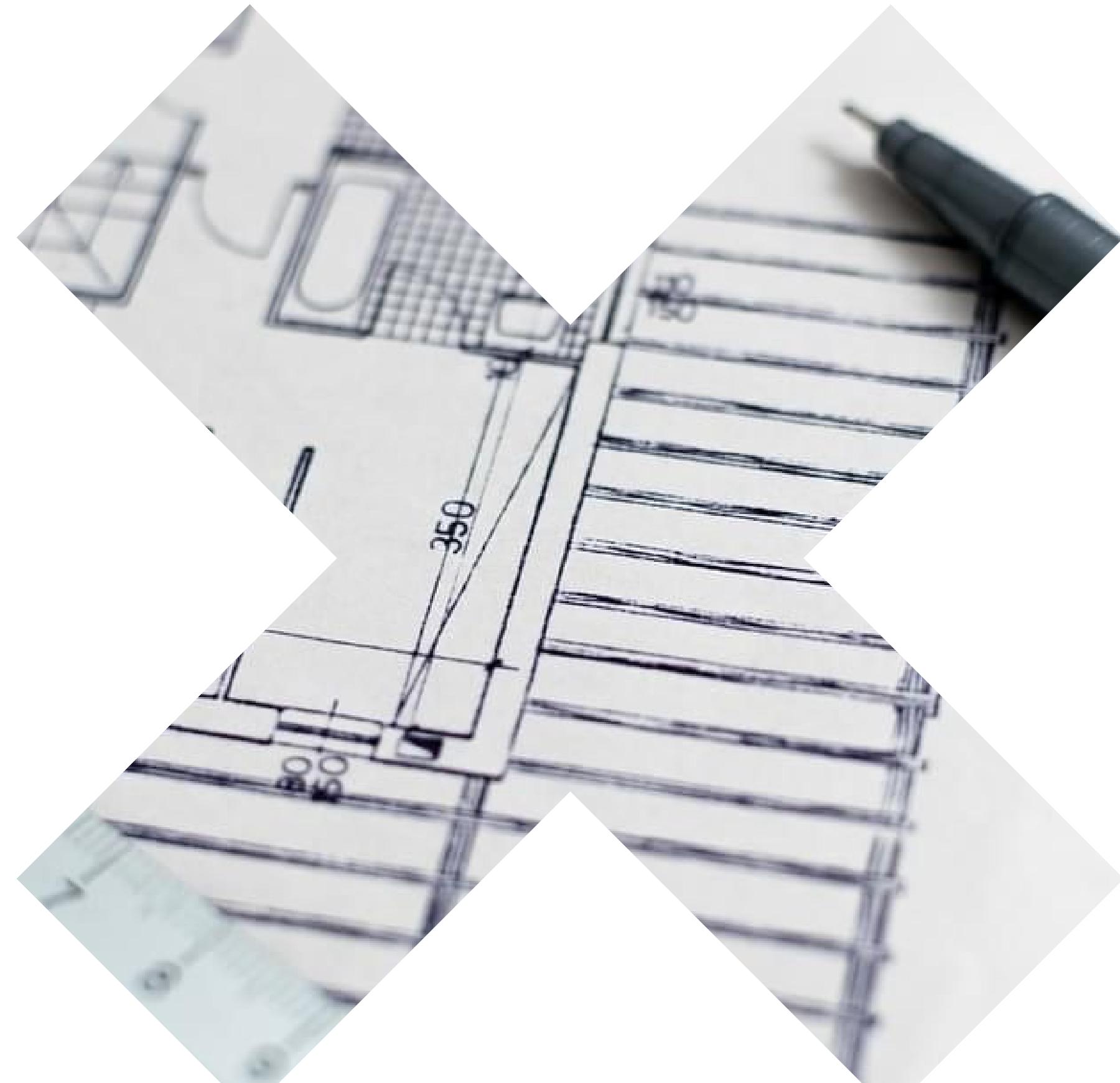
# INTRODUCTION



**Aims to solve Kaggle Problem: Classify  
between seeds and weeds**

Different categories of seedling and their images

Aarhus University Department of Engineering  
Signal Processing Group for hosting the  
[original data](#).



”

# Why this is important?

## Global Biofertilizer Market Trends, Forecast and Competitive Analysis to 2024: Promising Opportunities in Seed Treatment and Soil Treatment Applications



Research and Markets

21 September 2020 · 4-min read



Dublin, Sept. 21, 2020 (GLOBE NEWSWIRE) -- The "Biofertilizer Market Report: Trends, Forecast and Competitive Analysis" report has been added to [ResearchAndMarkets.com's](#) offering.

The global biofertilizer market is expected to grow with a CAGR of 14% from 2019 to 2024

### TRENDING

1. Sen. Lindsey Graham says Black people are free to 'go anywhere' in South Carolina so long as they're conservative
2. William Barr discovers that he is not immune to Trump's wrath
3. Former NJ Gov. Chris Christie released from hospital a week after testing positive for COVID-19



Reduce labour cost



Increase Efficiency

INTRODUCTION

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# Data Loaders:

## Train dataset:

```
In [5]: #transforming and loading the dataset
transform = transforms.Compose(
    [transforms.Resize([224,224])
])
train_data = datasets.ImageFolder(train_dir, transform=transform)
```

## Test dataset:

```
#for Test Dataset
class SeedlingsTestDataset(Dataset):
    def __init__(self, data_path, transform=None):
        self.data_path = data_path
        self.transform = transform

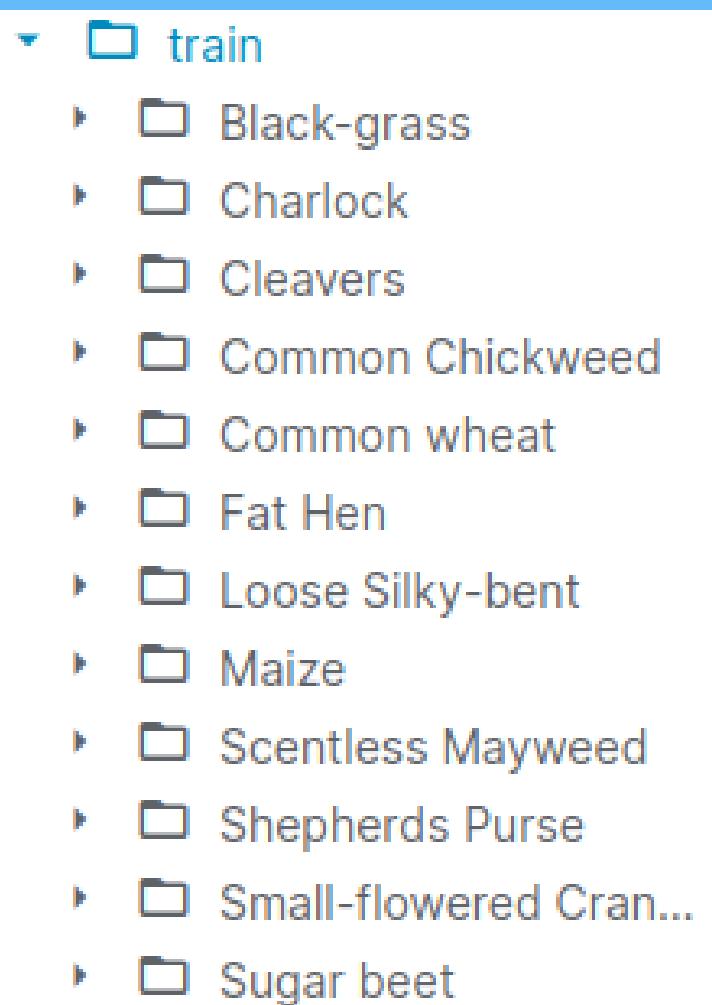
        self.images = [os.path.join(self.data_path, dI) for dI in os.listdir(self.data_path)]

    def __len__(self):
        return len(self.images)

    def __getitem__(self, index):
        img_path = self.images[index]
        img = Image.open(img_path)

        if self.transform is not None:
            img = self.transform(img)

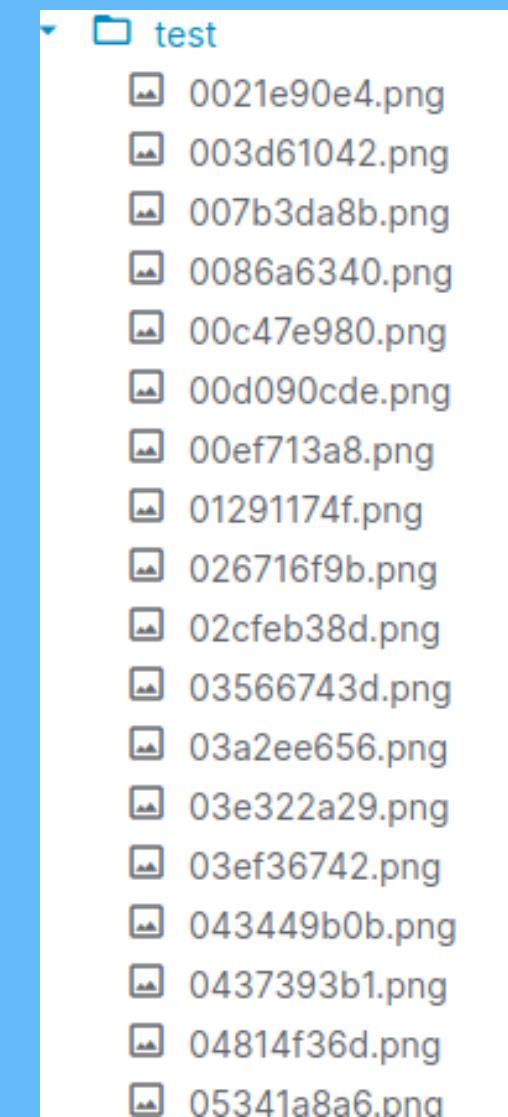
        return img
```



```
{0: 'Black-grass',
1: 'Charlock',
2: 'Cleavers',
3: 'Common Chickweed',
4: 'Common wheat',
5: 'Fat Hen',
6: 'Loose Silky-bent',
7: 'Maize',
8: 'Scentless Mayweed',
9: 'Shepherds Purse',
10: 'Small-flowered Cranesbill',
11: 'Sugar beet'}
```

```
#test transform and load
test_transforms = transforms.Compose([
    transforms.Resize([224, 224]),
    transforms.ToTensor(),
    transforms.ToPILImage()
])

test_dataset = SeedlingsTestDataset(
    test_dir,
    transform=test_transforms)
```



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MODELLING**

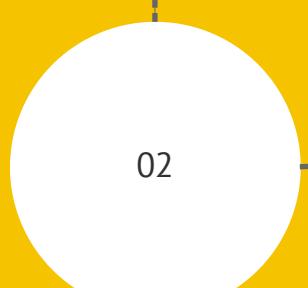
**EVALUATION**

”

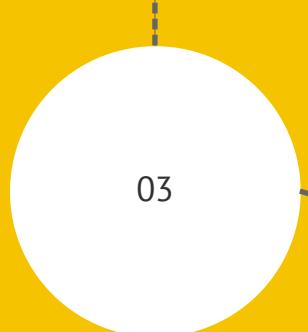
## Exploratory Data Analysis (EDA)



Print some dataset details



**Different Plots**



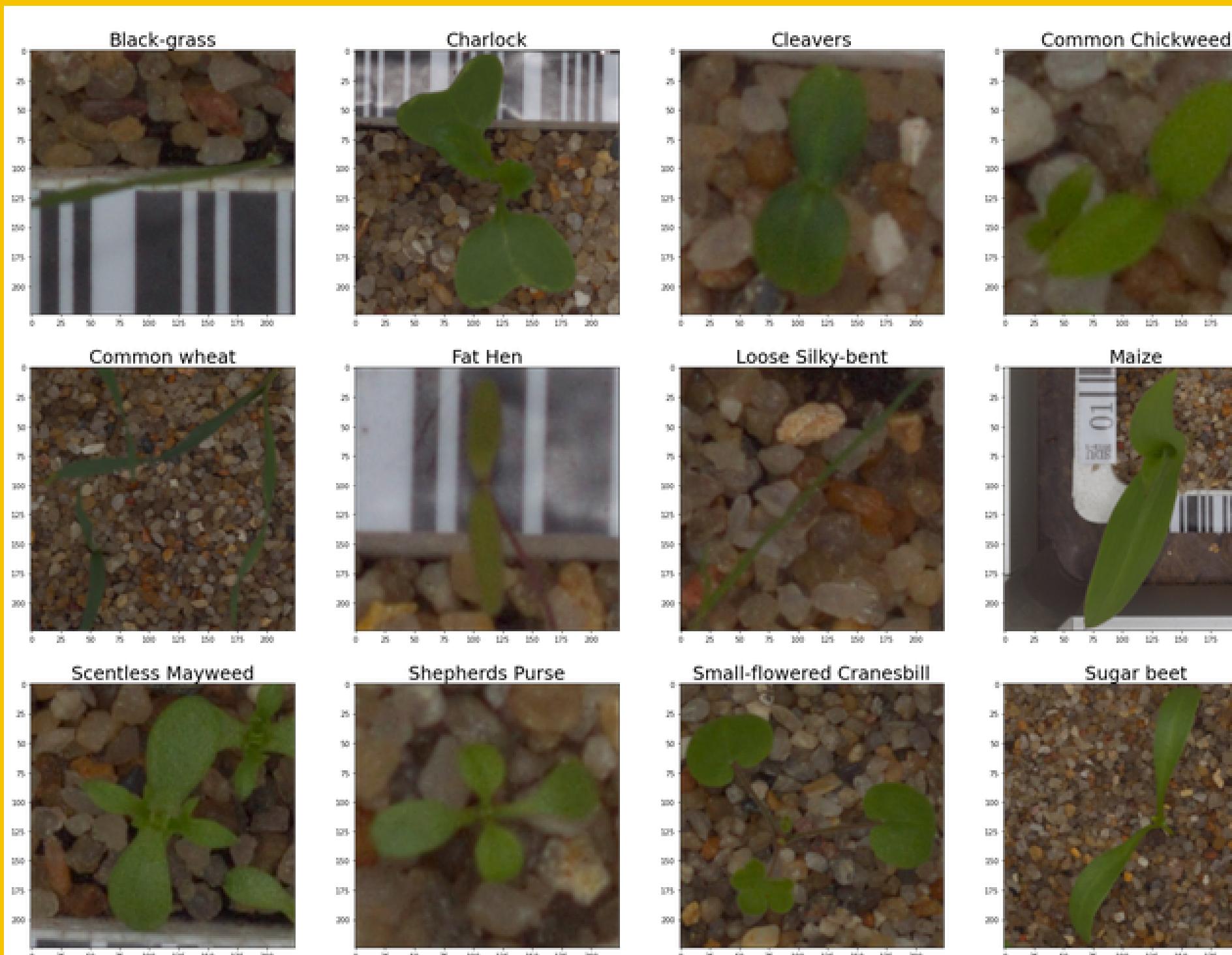
**Categories with different IDs**

”

## Exploratory Data Analysis (EDA)

01

# Print some dataset details



`#length of the training data  
data_length`

4750

`#shape of the image  
image_shape`

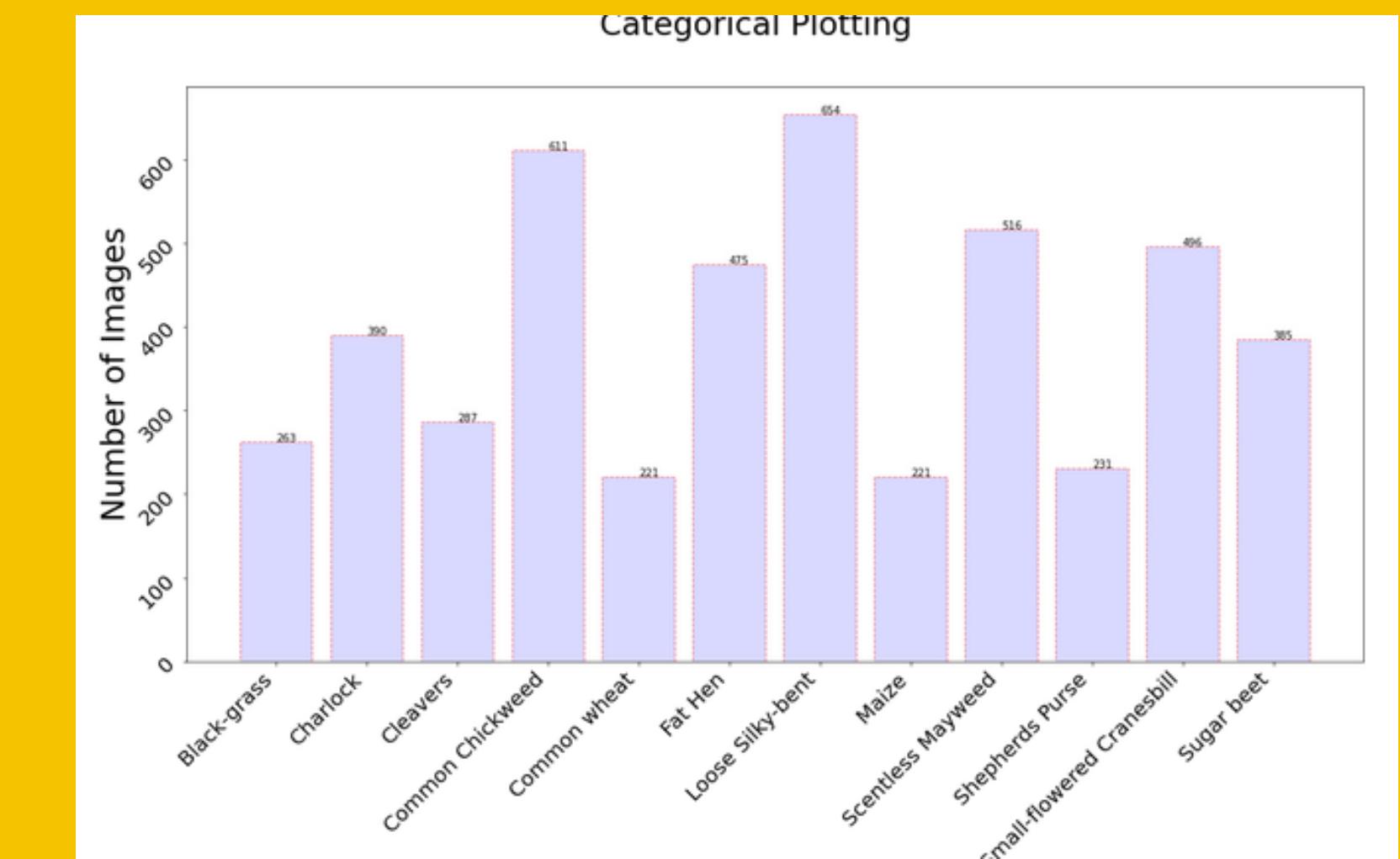
(224, 224, 3)

”

## Exploratory Data Analysis (EDA)

02

# Different Plots



02

# Categories with different IDs

```
{0: 'Black-grass',  
 1: 'Charlock',  
 2: 'Cleavers',  
 3: 'Common Chickweed',  
 4: 'Common wheat',  
 5: 'Fat Hen',  
 6: 'Loose Silky-bent',  
 7: 'Maize',  
 8: 'Scentless Mayweed',  
 9: 'Shepherds Purse',  
 10: 'Small-flowered Cranesbill',  
 11: 'Sugar beet'}
```

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AND  
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**EVALUATION**



# DATA PRE-PROCESSING

## ” DATA AND LABELS

```
#Making empty labels and data
data_shape = list(image_shape)
data_shape.insert(0, data_length)
data = np.zeros(data_shape, dtype=np.uint8)
labels = np.zeros([data_length], dtype=np.int64)
```

```
#dividing labels and data
i = 0
for image, label in tqdm(train_data, desc="Reading Images"):
    data[i] = np.array(image)
    labels[i] = label
    i += 1
```

```
#Creatig dictionary where train_dict contains data ,labels and shape, train_infor contains ids and image paths
train_dict = {"data": data, "labels": labels, 'data_shape': image_shape}
train_info = {"label_names": idx_to_class, "file_paths": image_paths}
```

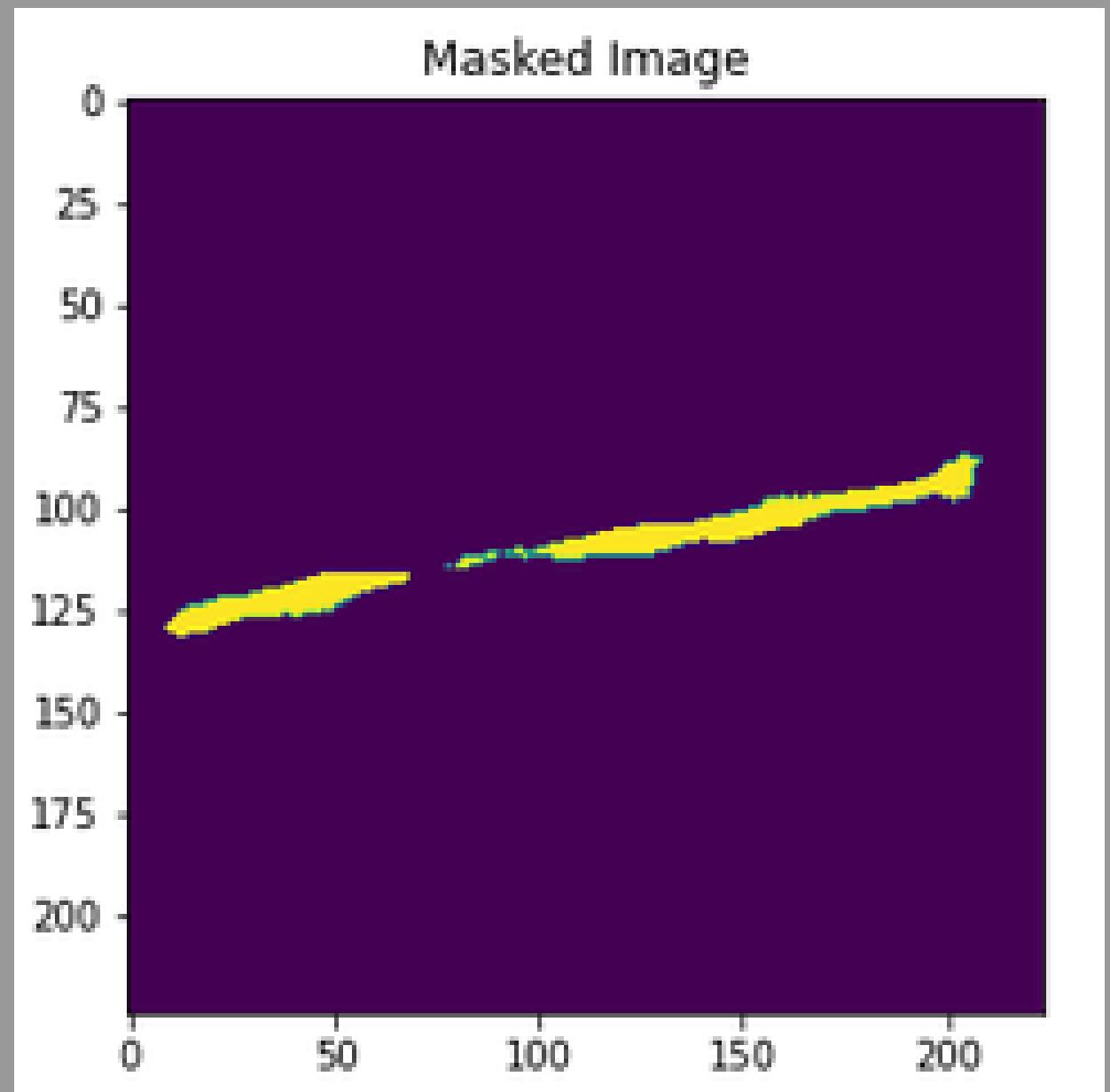


# DATA PRE-PROCESSING

## ” DIGITAL IMAGE PROCESSING

### 1. MASKING

```
def create_mask_for_plant(image):
    image_hsv = cv2.cvtColor(image, cv2.COLOR_BGR2HSV)
    sensitivity = 35
    lower_hsv = np.array([60 - sensitivity, 100, 50])
    upper_hsv = np.array([60 + sensitivity, 255, 255])
    mask = cv2.inRange(image_hsv, lower_hsv, upper_hsv)
    kernel = cv2.getStructuringElement(cv2.MORPH_ELLIPSE, (11,11))
    mask = cv2.morphologyEx(mask, cv2.MORPH_CLOSE, kernel)
    return mask
```



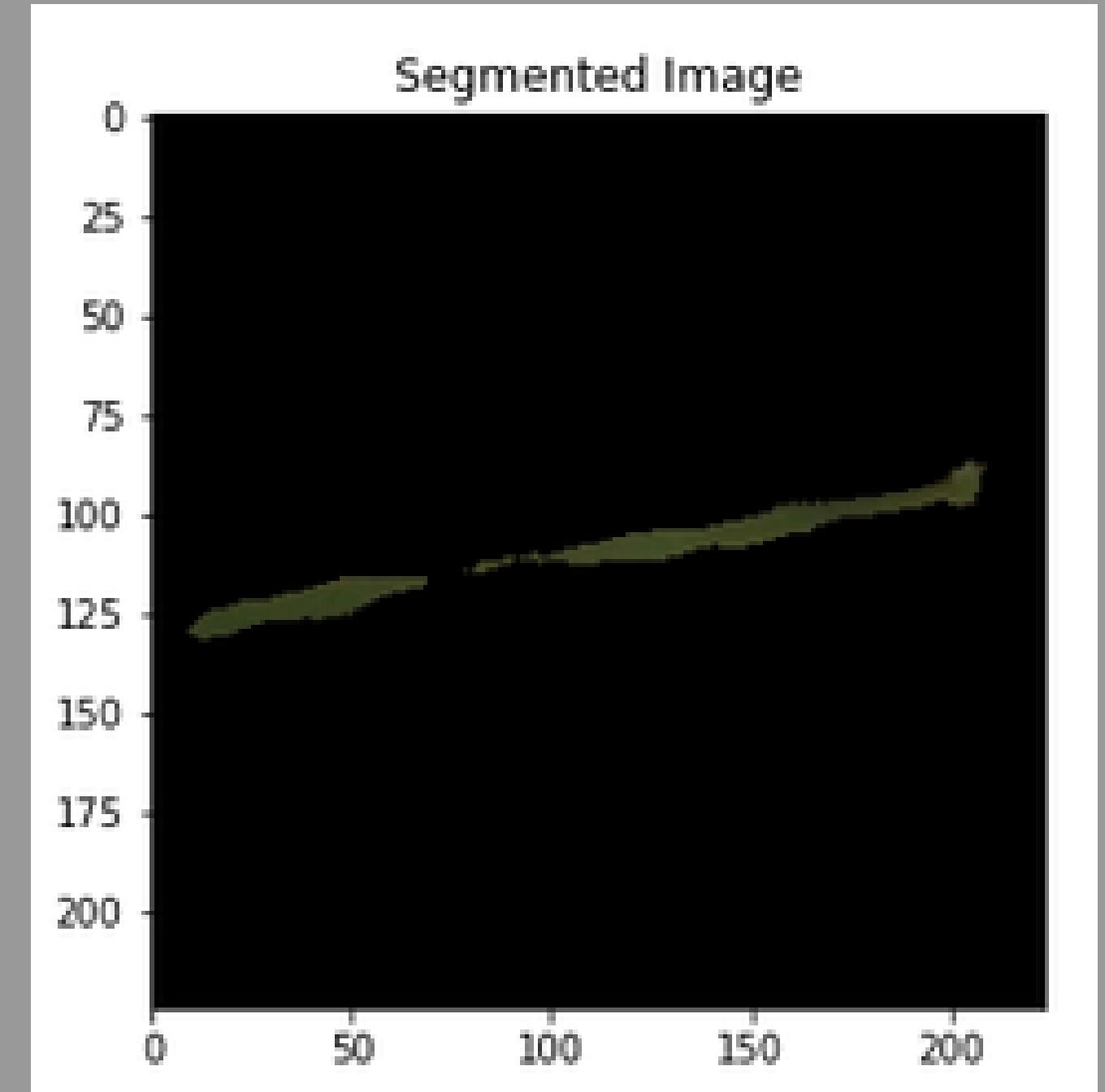


# DATA PRE-PROCESSING

## ” DIGITAL IMAGE PROCESSING

### 2. SEGMENTATION

```
#segmentation
def segment_plant(image):
    mask = create_mask_for_plant(image)
    output = cv2.bitwise_and(image, image, mask = mask)
    return output
```



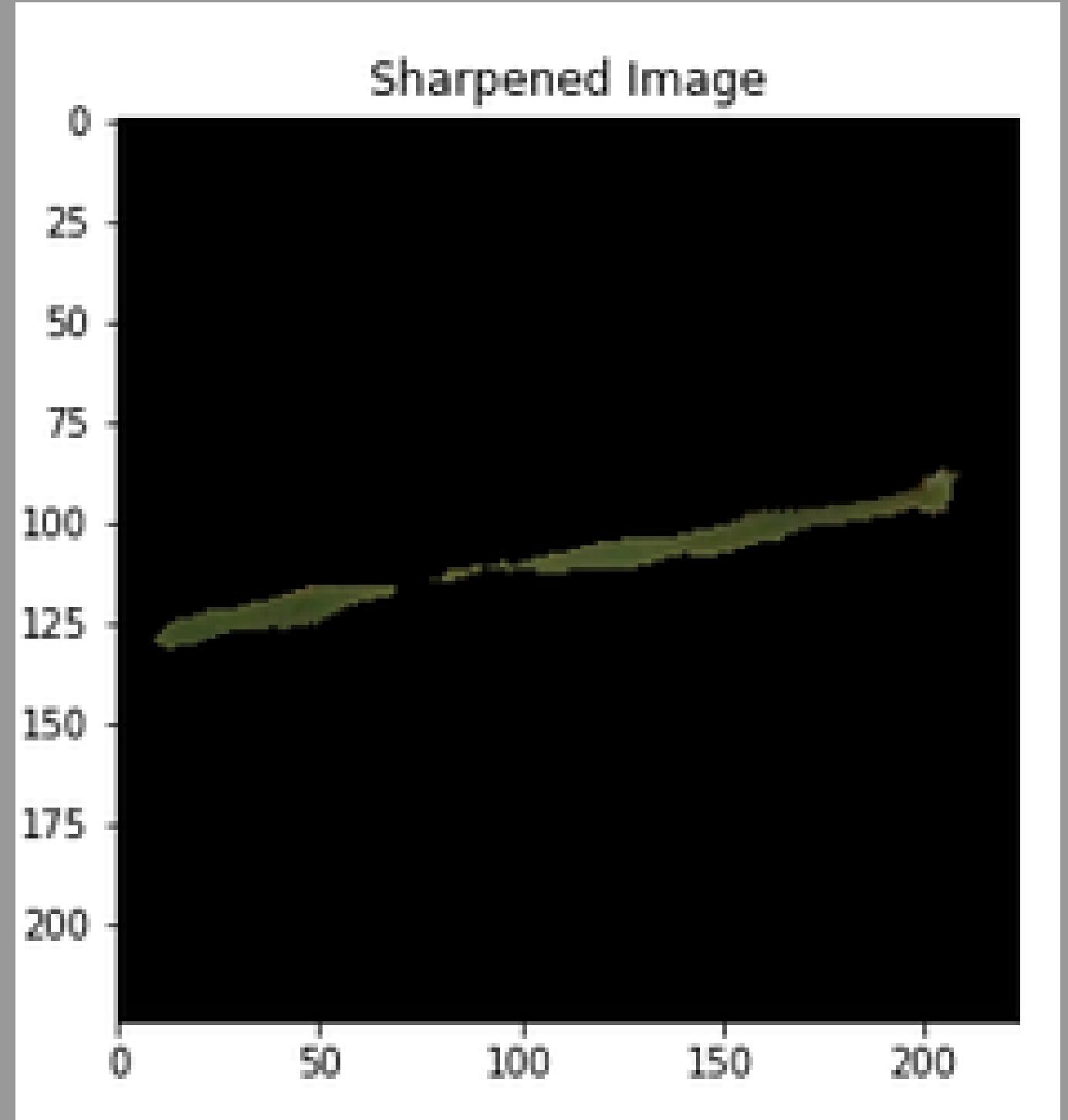


# DATA PRE-PROCESSING

## ” DIGITAL IMAGE PROCESSING

### 3. SHARPENING

```
#sharpening
def sharpen_image(image):
    image_blurred = cv2.GaussianBlur(image, (0, 0), 3)
    image_sharp = cv2.addWeighted(image, 1.5, image_blurred, -0.5, 0)
    return image_sharp
```





# DATA PRE-PROCESSING

## ” CONVERTING IMAGES TO TENSOR

```
image_dataset = ImageDataset(train_dict["data"], train_dict["labels"])
#Converting images to Tensor
print(image_dataset[0][0].shape)
print(image_dataset[4610])

#Image Dataset class returns item and length
class ImageDataset(torch.utils.data.Dataset):
    def __init__(self, x_data, y_data):
        self.x_data = x_data
        self.y_data = y_data
        self.to_tensor = torchvision.transforms.ToTensor()
    def __getitem__(self, index):
        return self.to_tensor(self.x_data[index]), self.y_data[index]
    def __len__(self):
        return len(self.x_data)
```

```
torch.Size([3, 224, 224])
(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.]],

          [[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.]]], 11)
```



# DATA PRE-PROCESSING

## ” DIVIDING VALIDATION AND TRAINING

```
#Dividing the dataset to training set and validation set
def array_random_pick(array, pick_num):
    index = np.arange(len(array))
    pick = np.random.choice(len(array), pick_num, replace=False)
    unpick = np.equal(np.in1d(index, pick), False)
    return array[unpick], array[pick]
#10% to validation and remaining to training data
train_mask, valid_mask = array_random_pick(np.arange(len(image_dataset)), 475)
train_set = torch.utils.data.Subset(image_dataset, train_mask)
valid_set = torch.utils.data.Subset(image_dataset, valid_mask)
```

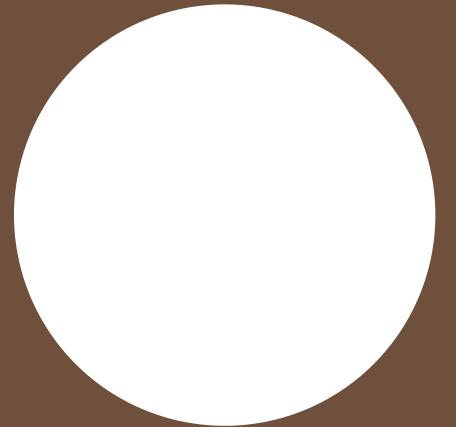
```
#length of training data
len(train_loader)
```

4275

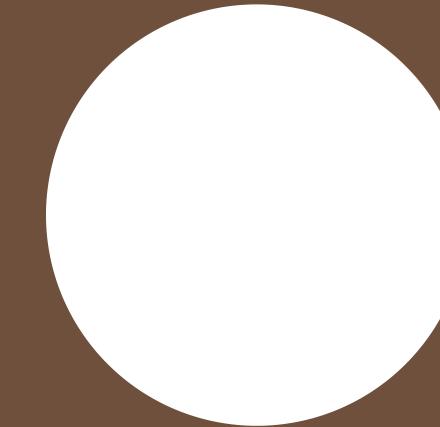
```
#length of validation data
len(valid_loader)
```

475

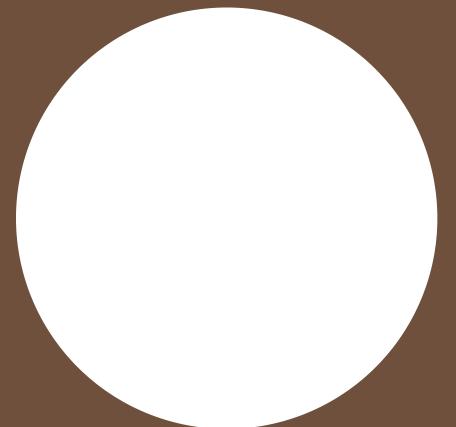
# DATA MODELLING



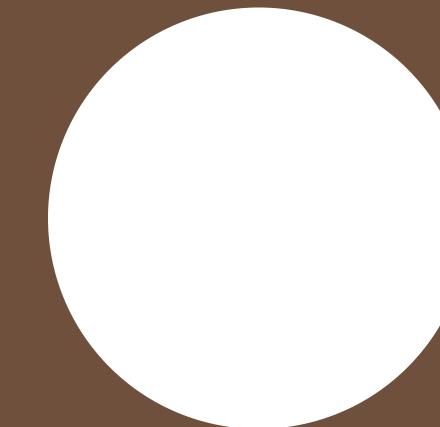
CONVOLUTIONAL  
NEURAL NETWORK



RESNET 34  
(PYTORCH PRE-  
TRAINED WEIGHTS)



VGG-16



ENSEMBLE  
APPROACH

# DATA MODELLING



## CONVOLUTIONAL NEURAL NETWORK

Layer (type)	Output Shape	Param #
Conv2d-1	[ -1, 6, 220, 220]	456
BatchNorm2d-2	[ -1, 6, 220, 220]	12
MaxPool2d-3	[ -1, 6, 110, 110]	0
Dropout-4	[ -1, 6, 110, 110]	0
Conv2d-5	[ -1, 16, 106, 106]	2,416
BatchNorm2d-6	[ -1, 16, 106, 106]	32
MaxPool2d-7	[ -1, 16, 53, 53]	0
Dropout-8	[ -1, 16, 53, 53]	0
Conv2d-9	[ -1, 32, 49, 49]	12,832
Linear-10	[ -1, 512]	39,338,496
ReLU-11	[ -1, 512]	0
Linear-12	[ -1, 100]	51,300
ReLU-13	[ -1, 100]	0
Linear-14	[ -1, 12]	1,212
<hr/>		
Total params: 39,406,756		
Trainable params: 39,406,756		
Non-trainable params: 0		

[submission\\_cnn.csv](#)

a few seconds ago by [Arpit Garg](#)

[add submission details](#)

0.96347

0.96347

# DATA MODELLING

## RESNET34

Layer Name	Shape	Params
BatchNorm2d-110	[ 1, 512, 7, 7]	1,024
ReLU-111	[ -1, 512, 7, 7]	0
Conv2d-112	[ -1, 512, 7, 7]	2,359,296
BatchNorm2d-113	[ -1, 512, 7, 7]	1,024
ReLU-114	[ -1, 512, 7, 7]	0
BasicBlock-115	[ -1, 512, 7, 7]	0
Conv2d-116	[ -1, 512, 7, 7]	2,359,296
BatchNorm2d-117	[ -1, 512, 7, 7]	1,024
ReLU-118	[ -1, 512, 7, 7]	0
Conv2d-119	[ -1, 512, 7, 7]	2,359,296
BatchNorm2d-120	[ -1, 512, 7, 7]	1,024
ReLU-121	[ -1, 512, 7, 7]	0
BasicBlock-122	[ -1, 512, 7, 7]	0
AdaptiveAvgPool2d-123	[ -1, 512, 1, 1]	0
Linear-124	[ -1, 12]	6,156
ResNet-125	[ -1, 12]	0
<hr/>		
Total params:	21,290,828	
Trainable params:	21,059,340	
Non-trainable params:	231,488	

[submission\\_resnet.csv](#)

3 minutes ago by Arpit Garg

0.97732

0.97732

# DATA MODELLING

## VGG-16

<a href="#">submission.csv</a>	0.96725	0.96725
3 minutes ago by Arpit Garg		

## ENSEMBLE

<a href="#">submission_ensemble.csv</a>	0.69269	0.69269
a minute ago by Arpit Garg		

Layer (type)	Output Shape	Param #
Conv2d-1	[ -1, 64, 224, 224]	1,792
ReLU-2	[ -1, 64, 224, 224]	0
Conv2d-3	[ -1, 64, 224, 224]	36,928
ReLU-4	[ -1, 64, 224, 224]	0
MaxPool2d-5	[ -1, 64, 112, 112]	0
Conv2d-6	[ -1, 128, 112, 112]	73,856
ReLU-7	[ -1, 128, 112, 112]	0
Conv2d-8	[ -1, 128, 112, 112]	147,584
ReLU-9	[ -1, 128, 112, 112]	0
MaxPool2d-10	[ -1, 128, 56, 56]	0
Conv2d-11	[ -1, 256, 56, 56]	295,168
ReLU-12	[ -1, 256, 56, 56]	0
Conv2d-13	[ -1, 256, 56, 56]	590,080
ReLU-14	[ -1, 256, 56, 56]	0
Conv2d-15	[ -1, 256, 56, 56]	590,080
ReLU-16	[ -1, 256, 56, 56]	0
MaxPool2d-17	[ -1, 256, 28, 28]	0
Conv2d-18	[ -1, 512, 28, 28]	1,180,160
ReLU-19	[ -1, 512, 28, 28]	0
Conv2d-20	[ -1, 512, 28, 28]	2,359,808
ReLU-21	[ -1, 512, 28, 28]	0
Conv2d-22	[ -1, 512, 28, 28]	2,359,808
ReLU-23	[ -1, 512, 28, 28]	0
MaxPool2d-24	[ -1, 512, 14, 14]	0
Conv2d-25	[ -1, 512, 14, 14]	2,359,808
ReLU-26	[ -1, 512, 14, 14]	0
Conv2d-27	[ -1, 512, 14, 14]	2,359,808
ReLU-28	[ -1, 512, 14, 14]	0
Conv2d-29	[ -1, 512, 14, 14]	2,359,808
ReLU-30	[ -1, 512, 14, 14]	0
MaxPool2d-31	[ -1, 512, 7, 7]	0
AdaptiveAvgPool2d-32	[ -1, 512, 7, 7]	0
Linear-33	[ -1, 4096]	102,764,544
ReLU-34	[ -1, 4096]	0
Dropout-35	[ -1, 4096]	0
Linear-36	[ -1, 4096]	16,781,312
ReLU-37	[ -1, 4096]	0
Dropout-38	[ -1, 4096]	0
Linear-39	[ -1, 1000]	4,097,000

Total params: 138,357,544  
Trainable params: 138,357,544  
Non-trainable params: 0

## INTRODUCTION

## DATA LOADER

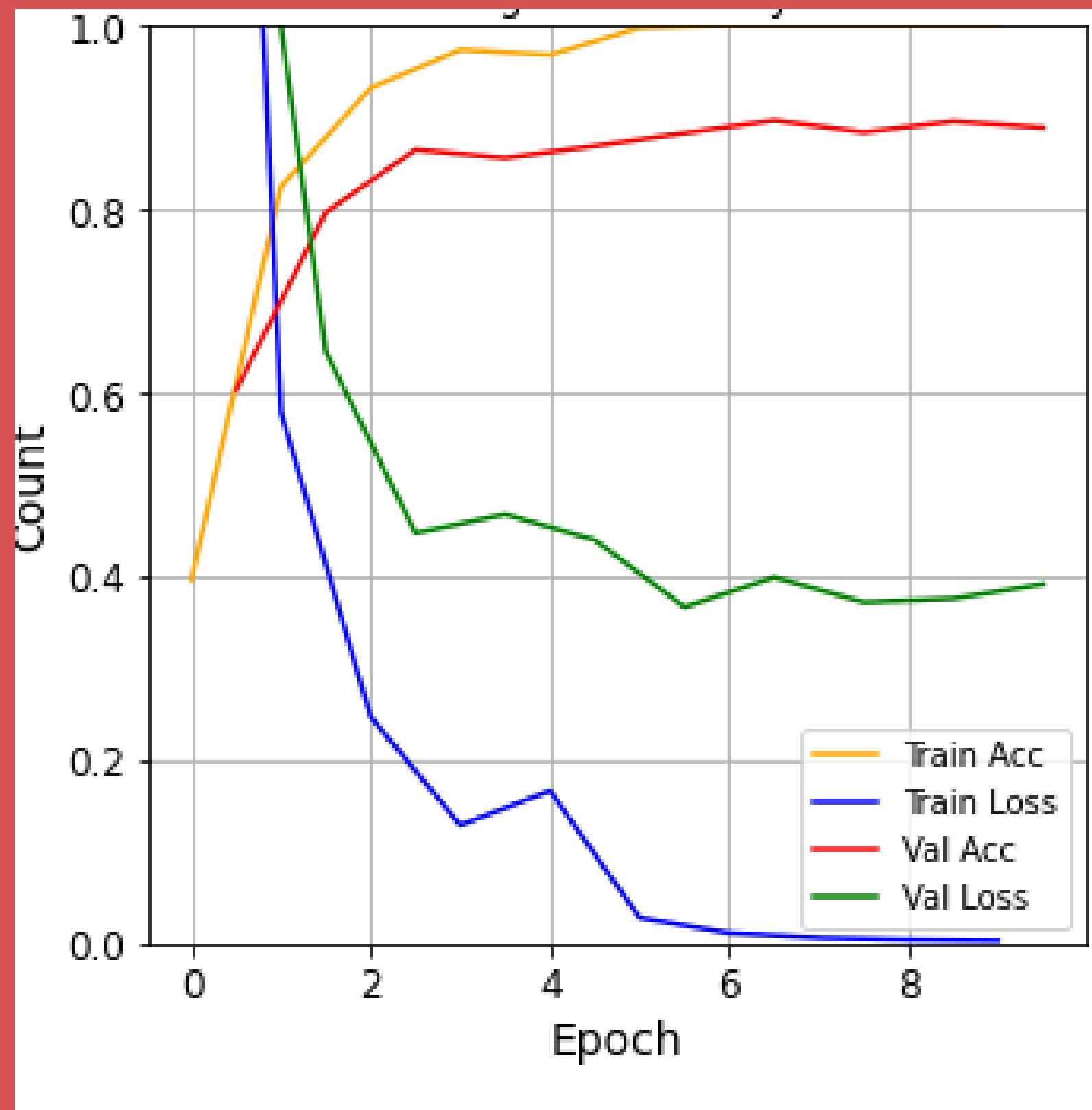
Exploratory  
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DATA PRE-  
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AND  
MODELLING

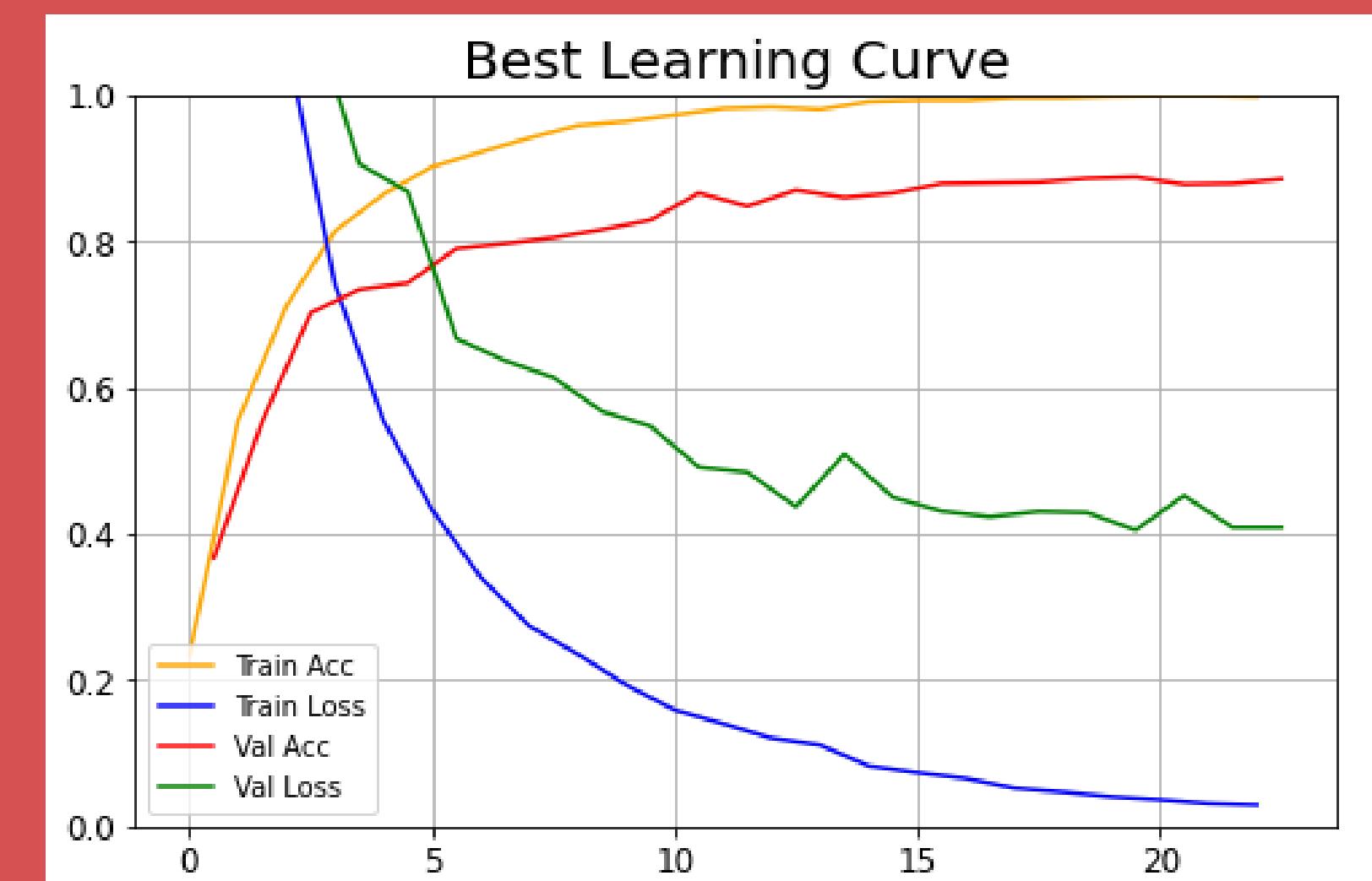
## EVALUATION

# EVALUATION

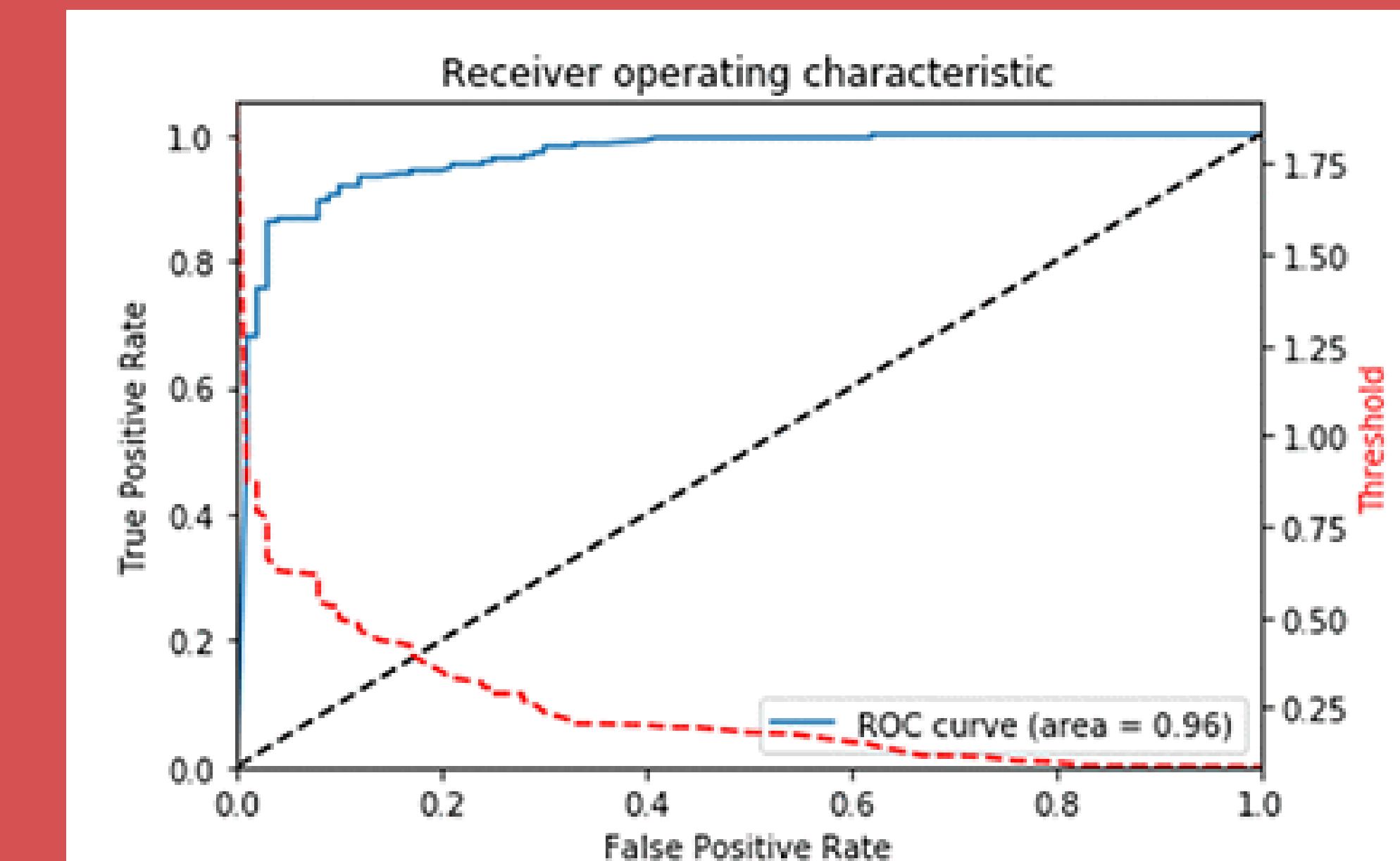
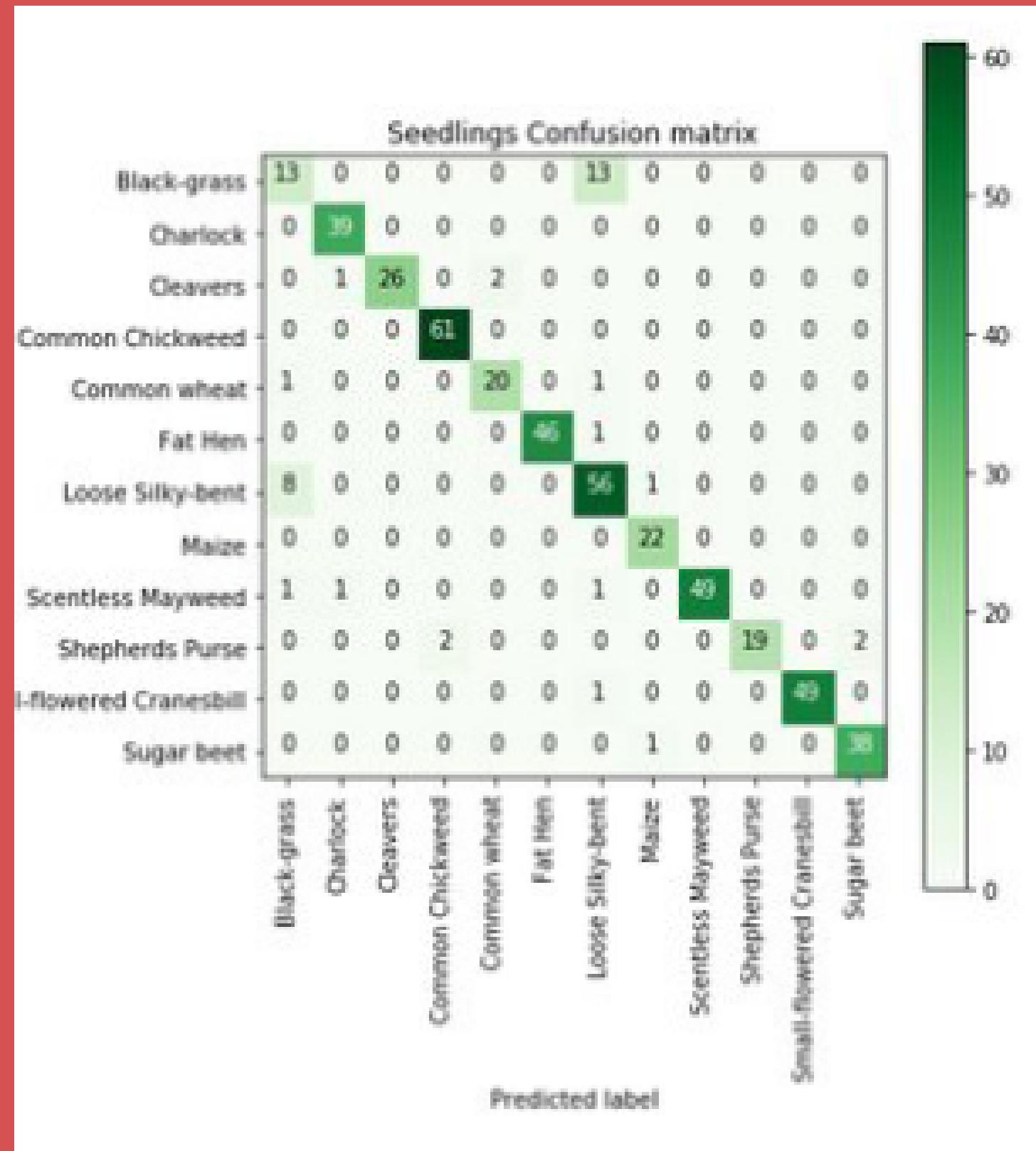
CNN LEARNING CURVE



RESNET34 LEARNING CURVE



# EVALUATION



# EVALUATION

Table 1: Evaluation Metrics

S.No.	Model	F Score
1	Convolutional Neural Network	0.96347
2	VGG-16	0.96725
3	ResNet34	0.97732
4	Ensemble	0.69269

[submission\\_cnn.csv](#) 0.96347 0.96347

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[submission.csv](#) 0.96725 0.96725

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[submission\\_resnet.csv](#) 0.97732 0.97732

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[submission\\_ensemble.csv](#) 0.69269 0.69269

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DANKE !

THANK YOU !

MERCI !

GRAZIE !

GRACIAS !

DANK JE WEL !

