

# PlantVillage Image Classification

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Aygaz Introduction to Artificial Intelligence Bootcamp

# INTRODUCTION

Today's agricultural sector needs new technologies in the fight against plant diseases and pests.

Artificial intelligence and deep learning have significant potential to increase efficiency and sustainability in agriculture.

- In this project, we developed an artificial intelligence model that can detect healthy and unhealthy conditions of pepper plants.
- Our model aims to provide farmers with an important diagnostic tool by accelerating disease detection processes in agriculture.



# METHODOLOGY

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## Librarys

- Numpy
- Pandas
- Seaborn
- Matplotlib
- Sklearn
- Tensorflow

## Data Collection

- Kaggle - PlantVillage Dataset

## Algorithms

Deep Learning algorithms

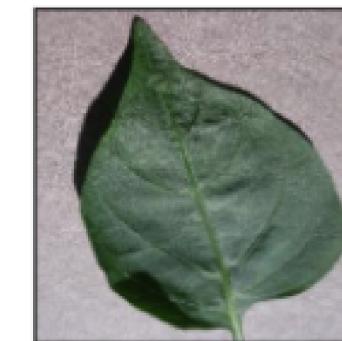
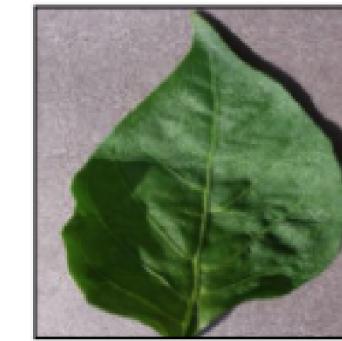
- Convolutional Neural Network

Machine learning algorithms

- K-Nearest Neighbor(KNN)
- Random Forest
- Support vector machine
- Decision Tree

# EXPLORE THE DATA

Dataset includes 880 healthy pepper leaves, and 748 unhealthy pepper leaves



Our dataset consists of two categories: visual and label. The label is divided into healthy and unhealthy.

```
DatasetDict({  
    train: Dataset({  
        features: ['image', 'label'],  
        num_rows: 1628  
    })  
})
```

# MODEL TRAINING

After separating our images into train and test, we normalized them and trained them with different algorithms to check which one fits our model.

## K-Nearest Neighbor(KNN)

Our model trained with KNN has an accuracy of 0.63.

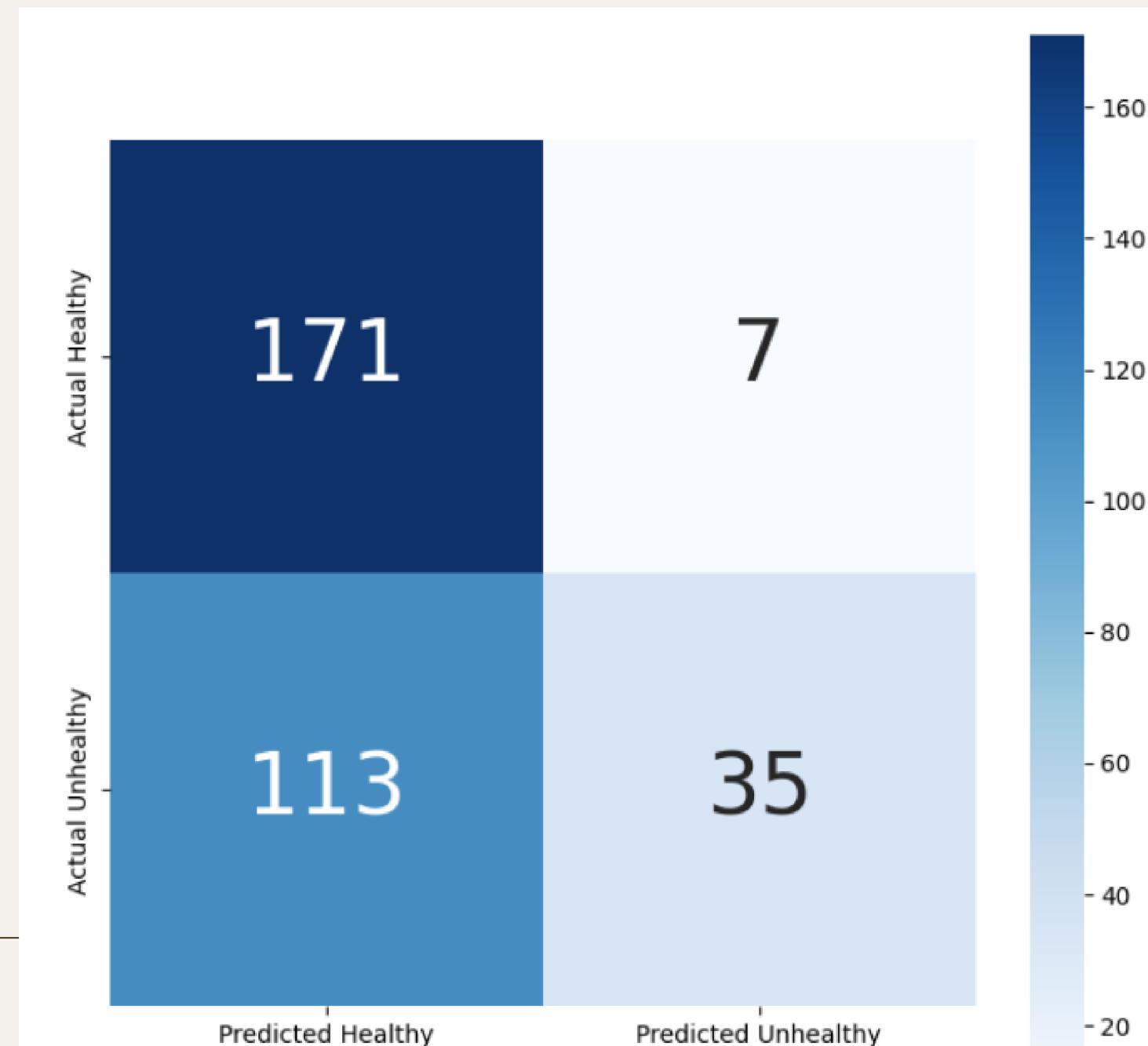
As we see in the confusion matrix, although it knows the healthy ones well, it predicts the unhealthy ones very poorly.

**Accuracy: 0.6319**

**F1 Score: 0.5714**

**Recall: 0.6319**

**Precision: 0.7071**



# Convolutional Neural Network(CNN)

Epoch 1/5 loss: 1.1098 - accuracy: 0.7642 - val\_loss: 0.2528 - val\_accuracy: 0.8926

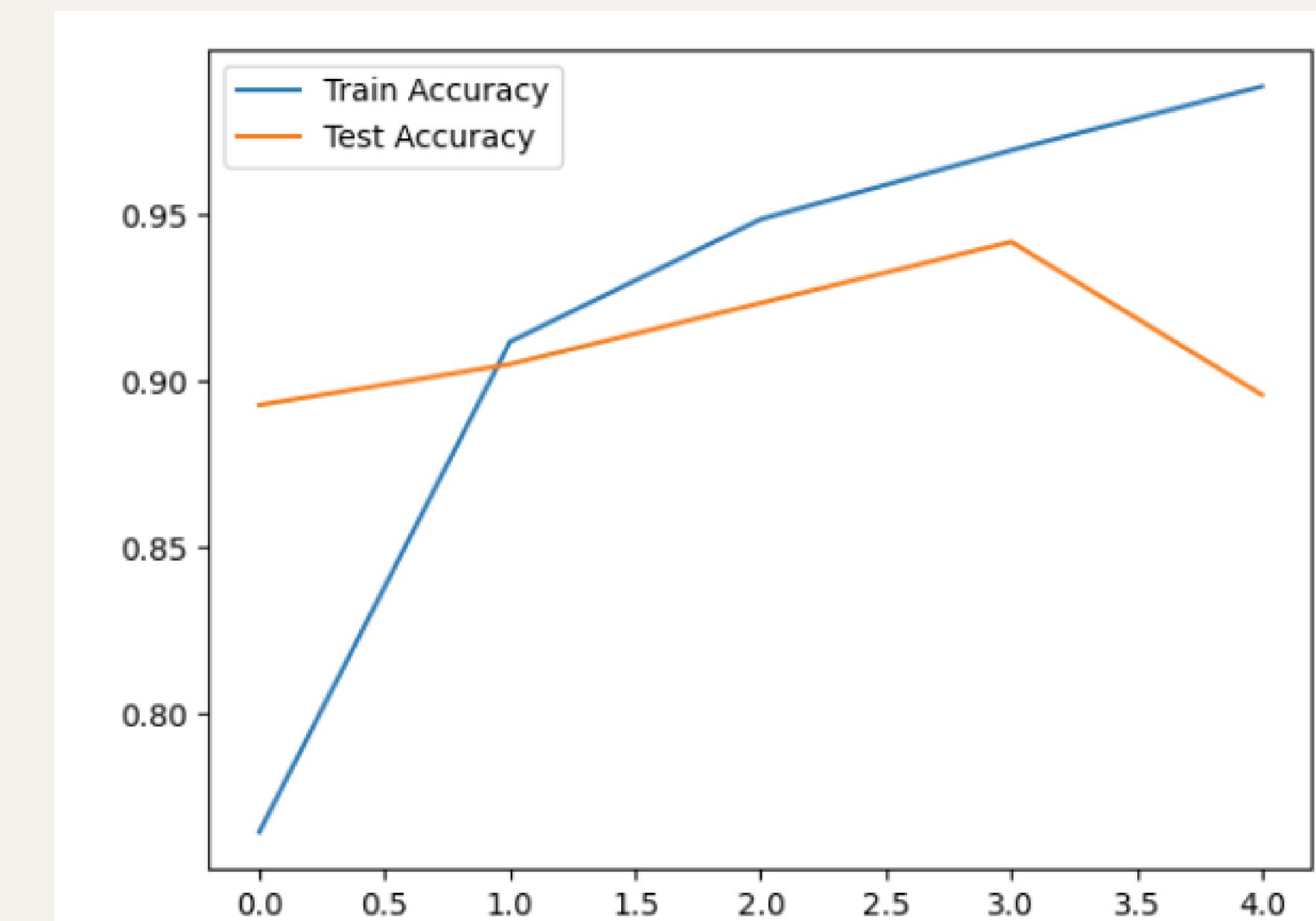
Epoch 2/5 loss: 0.2232 - accuracy: 0.9117 - val\_loss: 0.2020 - val\_accuracy: 0.9049

Epoch 3/5 5s/step - loss: 0.1398 - accuracy: 0.9485 - val\_loss: 0.1604 - val\_accuracy: 0.9233

**Epoch 4/5 4s/step - loss: 0.0938 - accuracy: 0.9693 - val\_loss: 0.1285 - val\_accuracy: 0.9417**

Epoch 5/5 loss: 0.0501 - accuracy: 0.9885 - val\_loss: 0.2510 - val\_accuracy: 0.8957

Increasing the number of Epochs too much may cause the model to overfit the training data. 4 epochs seems ideal for this data.



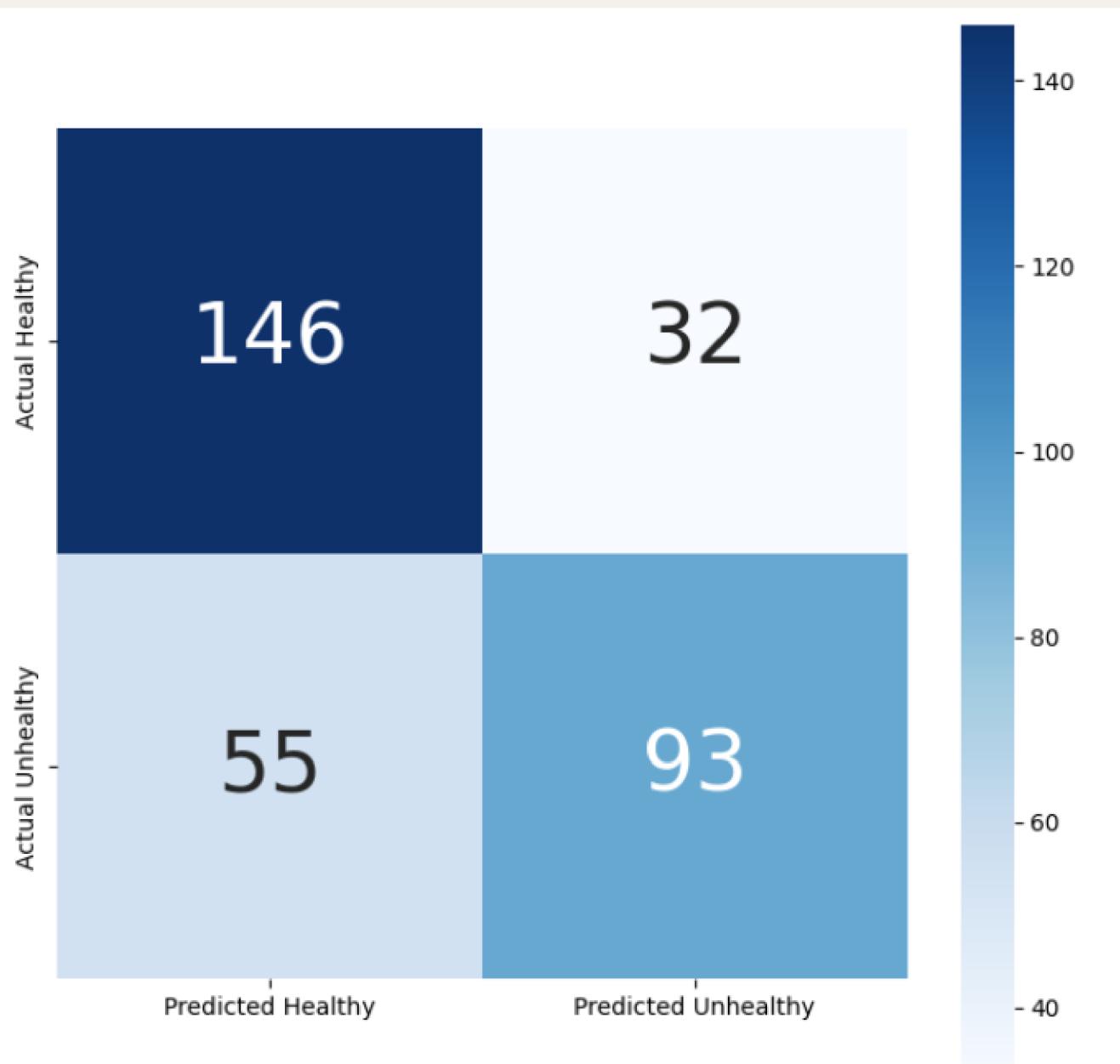
# Decision Tree

**Accuracy: 0.7331**

**F1 Score: 0.7300**

**Recall: 0.7331**

**Precision: 0.7344**



The higher accuracy, F1 score, recall, and precision indicate that the Random Forest model is more reliable and effective in distinguishing between healthy and unhealthy cases.

This is visually supported by the confusion matrices, where the Random Forest has fewer false positives and false negatives compared to the Decision Tree and KNN.

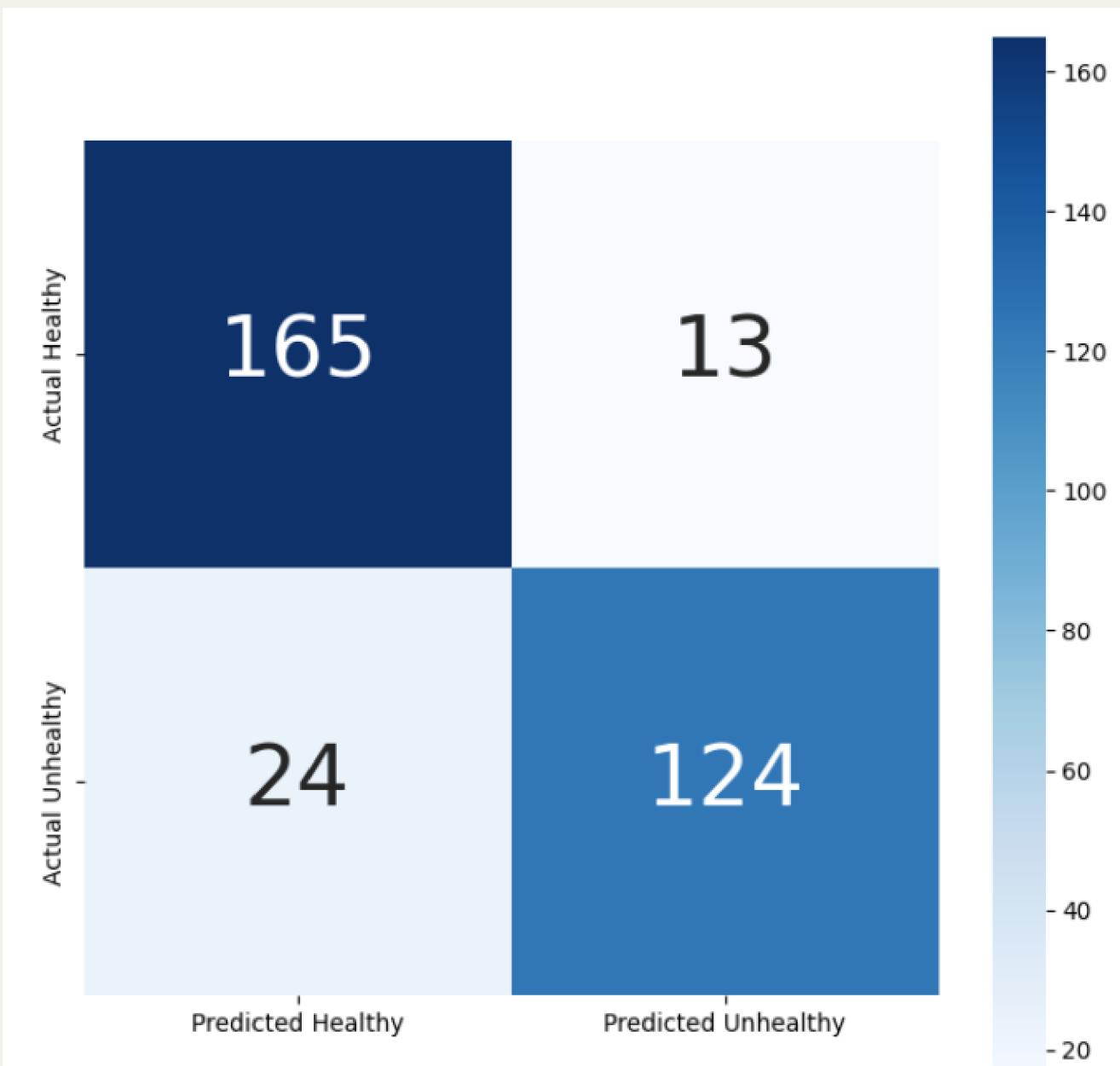
# Random Forest

**Accuracy: 0.8865**

**F1 Score: 0.8860**

**Recall: 0.8865**

**Precision: 0.8876**



# Support vector machine

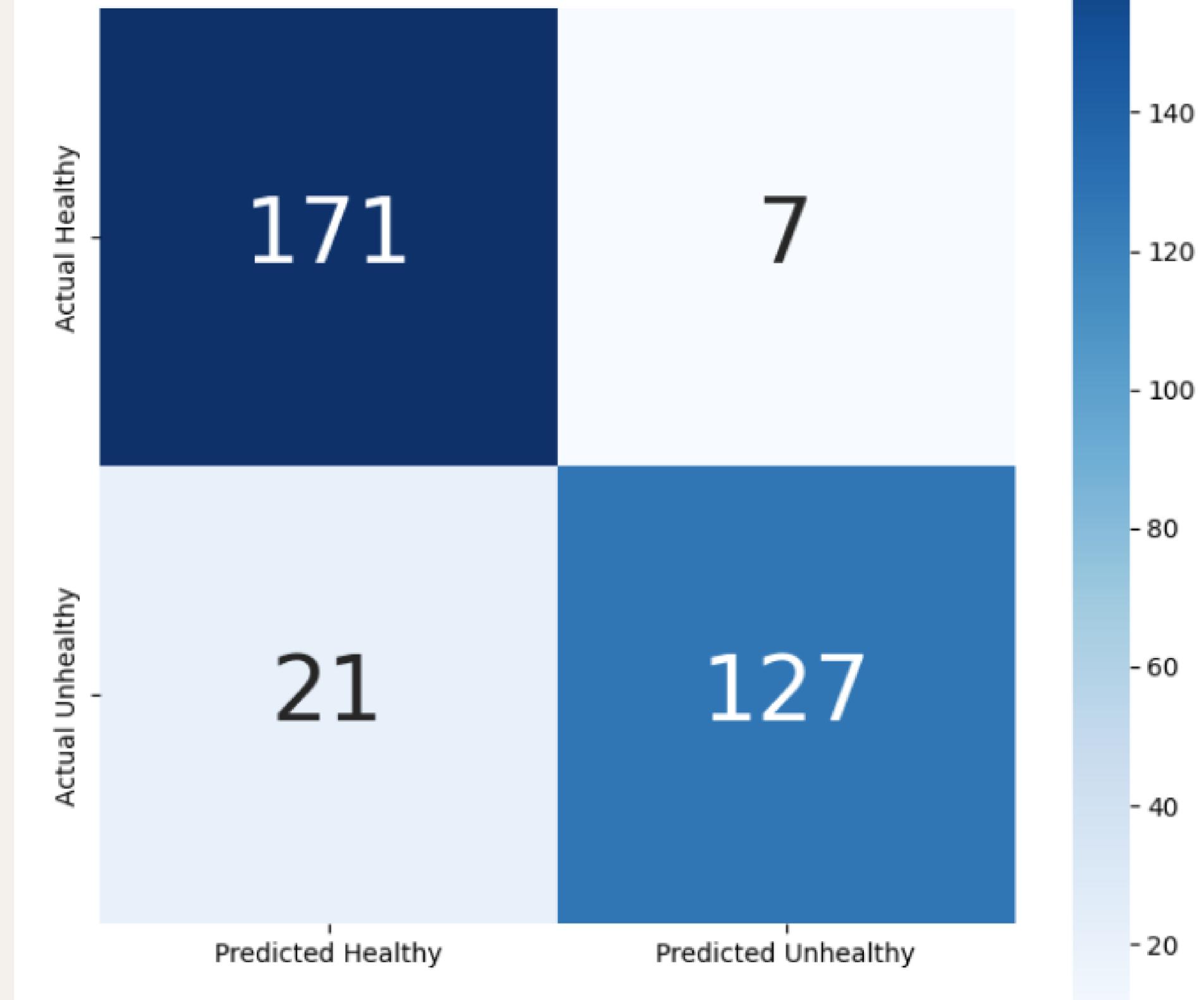
Accuracy: 0.9141

F1 Score: 0.9136

Recall: 0.9141

Precision: 0.9166

- High Performance: our model's accuracy, precision, recall and F1 score values are all over 90%. This indicates that our model performed quite well on our dataset and succeeded in the classification task.
- The best confusion matrix result is in SVM



The Support Vector Machine model has the potential to be used in disease diagnosis and improving product quality in the field of agriculture.

# CONCLUSION

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- In this project, an artificial intelligence model that can detect healthy and unhealthy conditions of pepper plants was developed. The developed model was evaluated using various performance metrics and satisfactory results were obtained.
- We achieved the best result in CNN, but if we want to use a machine learning method, SVM will be the best option.
- The performance of the model is directly related to the quality and diversity of the data set. Future work may aim to further improve the overall performance of the model by using larger and more diverse data sets.



**THANKS FOR LISTENING**