

# 1. Back ground

**INASTEK is engaged in the indoor strawberry cultivation business, which can automatically adjust the temperature, humidity and light in the room to grow strawberries even in hot regions.**



Strawberries grown indoors



Machines that regulate the indoor environment

# 1. Back ground

**Many farmers often lack knowledge of strawberry diseases, leading to incorrect treatments that can further harm plants. AI vision technology offers early, accurate detection and precise treatment guidance to ensure optimal plant health.**

**Although some diseases are easy to detect with the eye, other diseases look similar so it is difficult to differentiate one disease from another.**

## 2. Our Tasks

Develop a web application and AI Vision to detect strawberry diseases in photographs using a photo as an input image

Original Image



1573030

Disease Detection Result



Angular Leafspot 0.69

1573030

← Example of a web app to be created

Diseases

Angular Leafspot (69%)

[Process Image](#)

[Download](#)

# 3. System Architecture

The web application employs two main approaches:

## 1. Two-Stage Detection:

First, a YOLO-based model detects key plant parts (leaf, fruit, flower), and then the corresponding regions (ROIs) are cropped for further analysis by a dedicated disease detection model.

## 2. Direct Disease Detection:

Diseases are detected directly from the entire input image without segmenting plant parts.

### Strawberry Disease Detection

**Detection Method**

Detect parts first, then diseases  
 Detect diseases directly

**Part Detection Model**

Strawberry Tuned

**Disease Detection Model**

Leaf Blight Detector

**Upload Image**

Supported formats: JPEG, JPG, PNG

# 4. Methodology

## Data Collection

- Data Sources:

Datasets were gathered by searching for keywords such as “strawberry disease” and the specific disease names on platforms like Roboflow.

- Variety in Annotations:

The collected datasets feature diverse annotation styles. Some datasets annotate only the diseased areas, while others include the outer boundaries of the plant parts (leaf, fruit, or flower).



Angular Leafspot



Blossom Blight



Powdery Mildew Leaf



Powdery Mildew Fruit



Gray Mold



Leaf Spot



Anthracnose Fruit Rot



mosaic viruses



Anthracnose Stem Rot



Leaf Blight

# 4.Methodology

## Variety in Annotations:

The collected datasets feature diverse annotation styles. Some datasets annotate only the diseased areas, while others include the outer boundaries of the plant parts (leaf, fruit, or flower).



## Dataset Annotation

## Review of Existing Annotations:

- Existing datasets were reviewed to ensure high-quality annotations and diversity.

## Manual Annotation:

- Some dataset that doesn't annotate correctly, manually annotated.
- For diseases that are underrepresented in the strawberry datasets (e.g., Leaf Blight), manual annotations were created by outlining the infected areas using polygon shapes.
- In the case of the Mozaic Virus, annotations were adapted from a tomato plant dataset and manually refined for strawberry images.

# 4. Methodology

## Data Cleaning and Integration

Datasets, particularly the 7-disease dataset and the Leaf Blight dataset, were merged to create a more comprehensive dataset.

Annotations were standardized across the datasets to ensure consistency during model training.

## Model Training

### Training Resources:

Model training was conducted using Google Colab with an A100 GPU or on a university server equipped with an Nvidia RTX 3070.

### Hyperparameters:

Epochs: Between 100 and 150

Batch Sizes: 8, 16, or 32

Optimizers: SGD and AdamW

Data Augmentation: Techniques such as vertical flipping (flipud), mosaic augmentation, ,HSV, etc were applied to increase data diversity and reduce overfitting.

### Part Detection Models:

- strawberry\_part\_detection.pt (YOLOv8n)
- strawberry\_tuned.pt (YOLOv11m)

### Disease Detection Models:

- best\_strawberry\_model.pt (YOLOv8x) – detects 7 diseases
- leafblight.pt (YOLOv8n) – detects 7 diseases plus Leaf Blight

### Evaluation Metrics

The model performance was assessed using a validation dataset provided by INASTEK.

# 5. Results and Analysis

## Detect diseases directly

model: 7 Diseases + Leaf Blight Detection

### == Accuracy results ==

- Angular Leaf Spot: 66.67% (2/3)
- Anthracnose Fruit Rot: 100.00% (2/2)
- Anthracnose Stem Rot: 0.00% (0/4)
- Blossom Blight: 0.00% (0/3)
- Gray Mold: 16.67% (1/6)
- Leaf Blight: 100.00% (2/2)
- Leaf Spot: 40.00% (2/5)
- Mozaic Virus: 0.00% (0/5)
- Powdery Mildew Fruit: 100.00% (3/3)
- Powdery Mildew Leaf: 100.00% (2/2)
- Overall: 40.00% (14/35)

model: Best 7 Disease Detection

### == Accuracy results ==

- Angular Leaf Spot: 100.00% (3/3)
- Anthracnose Fruit Rot: 100.00% (2/2)
- Anthracnose Stem Rot: 0.00% (0/4)
- Blossom Blight: 0.00% (0/3)
- Gray Mold: 33.33% (2/6)
- Leaf Blight: 0.00% (0/2)
- Leaf Spot: 60.00% (3/5)
- Mozaic Virus: 0.00% (0/5)
- Powdery Mildew Fruit: 100.00% (3/3)
- Powdery Mildew Leaf: 100.00% (2/2)
- Overall: 42.86% (15/35)

# 5. Results and Analysis

**Detect parts first, then diseases**

**Part Detection : Strawberry Part Detection**

**Disease Detection : 7 Diseases + Leaf**

**Blight Detection**  
**7 Diseases + Leaf Blight**  
**Detection**

**== Accuracy results ==**

- Angular Leaf Spot: 100.00% (3/3)
- Anthracnose Fruit Rot: 100.00% (2/2)
- Anthracnose Stem Rot: 0.00% (0/4)
- Blossom Blight: 0.00% (0/3)
- Gray Mold: 33.33% (2/6)
- Leaf Blight: 100.00% (2/2)
- Leaf Spot: 40.00% (2/5)
- Mozaic Virus: 0.00% (0/5)
- Powdery Mildew Fruit: 100.00% (3/3)
- Powdery Mildew Leaf: 100.00% (2/2)
- Overall: 45.71% (16/35)

**Part Detection : Strawberry Part Detection**

**Disease Detection : Best 7 Disease Detection**

**== Accuracy results ==**

- Angular Leaf Spot: 100.00% (3/3)
- Anthracnose Fruit Rot: 100.00% (2/2)
- Anthracnose Stem Rot: 0.00% (0/4)
- Blossom Blight: 0.00% (0/3)
- Gray Mold: 83.33% (5/6)
- Leaf Blight: 0.00% (0/2)
- Leaf Spot: 60.00% (3/5)
- Mozaic Virus: 0.00% (0/5)
- Powdery Mildew Fruit: 100.00% (3/3)
- Powdery Mildew Leaf: 100.00% (2/2)
- Overall: 51.43% (18/35)

# 5. Results and Analysis

Detect parts first, then diseases

Part Detection : Strawberry Part

Detection v2

Disease Detection : 7 Diseases + Leaf

Blight Detection

==Accuracy results ==

- Angular Leaf Spot: 100.00% (3/3)
- Anthracnose Fruit Rot: 100.00% (2/2)
- Anthracnose Stem Rot: 0.00% (0/4)
- Blossom Blight: 0.00% (0/3)
- Gray Mold: 66.67% (4/6)
- Leaf Blight: 100.00% (2/2)
- Leaf Spot: 40.00% (2/5)
- Mozaic Virus: 0.00% (0/5)
- Powdery Mildew Fruit: 100.00% (3/3)
- Powdery Mildew Leaf: 100.00% (2/2)
- Overall: 51.43% (18/35)

Part Detection : Strawberry Part Detection v2

Disease Detection : Best 7 Disease Detection

== Accuracy results ==

- Angular Leaf Spot: 100.00% (3/3)
- Anthracnose Fruit Rot: 100.00% (2/2)
- Anthracnose Stem Rot: 0.00% (0/4)
- Blossom Blight: 0.00% (0/3)
- Gray Mold: 100.00% (6/6)
- Leaf Blight: 0.00% (0/2)
- Leaf Spot: 40.00% (2/5)
- Mozaic Virus: 0.00% (0/5)
- Powdery Mildew Fruit: 100.00% (3/3)
- Powdery Mildew Leaf: 100.00% (2/2)
- Overall: 51.43% (18/35)

# 5. Results and Analysis

## Best Overall Performance:

- Three methods ("Strawberry Part Detection - Best 7 Diseases", "Strawberry Part Detection v2 - Best 7 Diseases", and "Strawberry Part Detection v2 - Leaf Blight") tied at 51.43% (18/35).
- Two-stage (part-based) approaches outperformed direct detection methods.

## Disease-Specific Performance:

- 100% Detection: Angular Leaf Spot, Anthracnose Fruit Rot, and Powdery Mildew (fruit & leaf).
- Variable Detection: Gray Mold ranged from 16.67% to 100%, with best performance using Strawberry Part Detection.
- 0% Detection: Anthracnose Stem Rot, Blossom Blight, and Mozaic Virus were not detected.

## Model Specialization:

- "7 Diseases + Leaf Blight Detection" models were the only ones to detect Leaf Blight.
- "Best 7 Diseases Detection" models generally achieved better results for 7 diseases.

## Considerations:

- The "Strawberry Part Detection v2" approach with "Best 7 Diseases" appears most promising.

# 6.Drawbacks

## Lack of dataset



**mosaic viruses**



**Anthracnose Stem Rot**

## Low accuracy

- **Unable to accommodate light color changes.**
- **Unable to accommodate more complex images.**
- **Flower often not detected**
- **Weak to detect blossom blight disease**
- **Weak to detect leaf part and the diseases of the bottom of the leaf**

# 7. Suggestion

## **Suggestion for further development.**

- Provide better resources to be able to conduct more experiments.
- Do better data augmentation and cleaning before training.
- Get more variant dataset.

## **Suggestion for using the web app.**

- Take a closer photo of the affected part and then use detection method that detect disease directly.
- Take better and simpler photos.