

# Drone-Based Multi-spectral Pipeline for Detecting Abnormal Potato Strains in the Field

2024/11/25



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Agricultural and Economic Affairs Headquarters

# 01 ■ Introduction

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# I. Introduction

Frequent manual visual inspections for detecting infected and abnormal plants are often necessary to decrease their impact on potato yields.



## Objectives

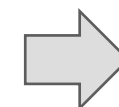
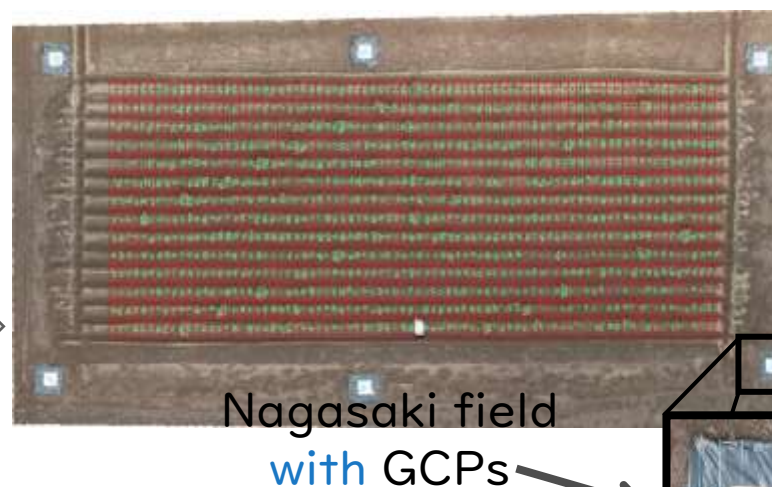
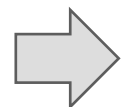
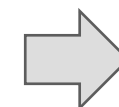
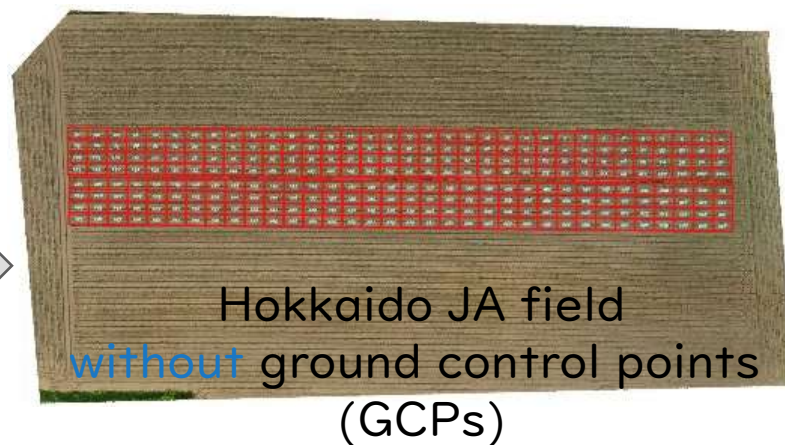
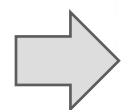
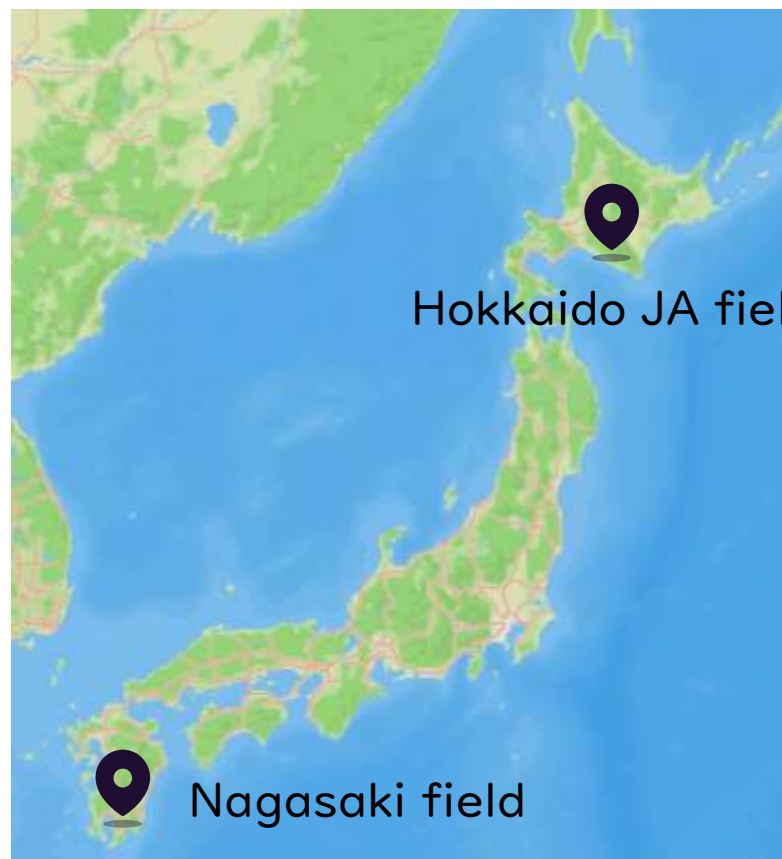
Develop a UAV-based pipeline for obtaining abnormal positions and guide the field investigation

# 02 ■ Methods and Materials

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## 2. Methods & Results

### 1) Experimental fields



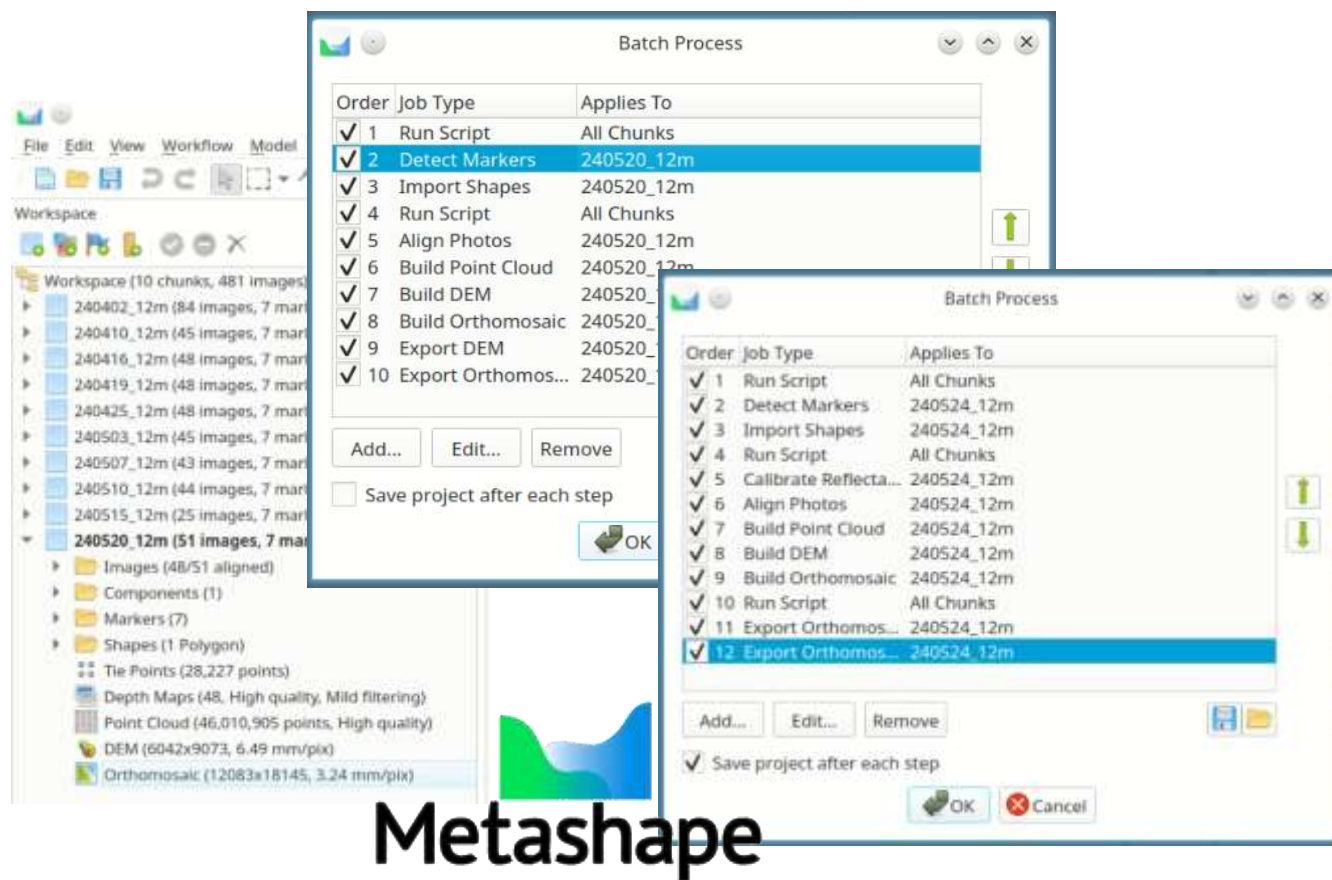


## 2. Methods & Results

### 2) Image acquisition & 3D reconstruction



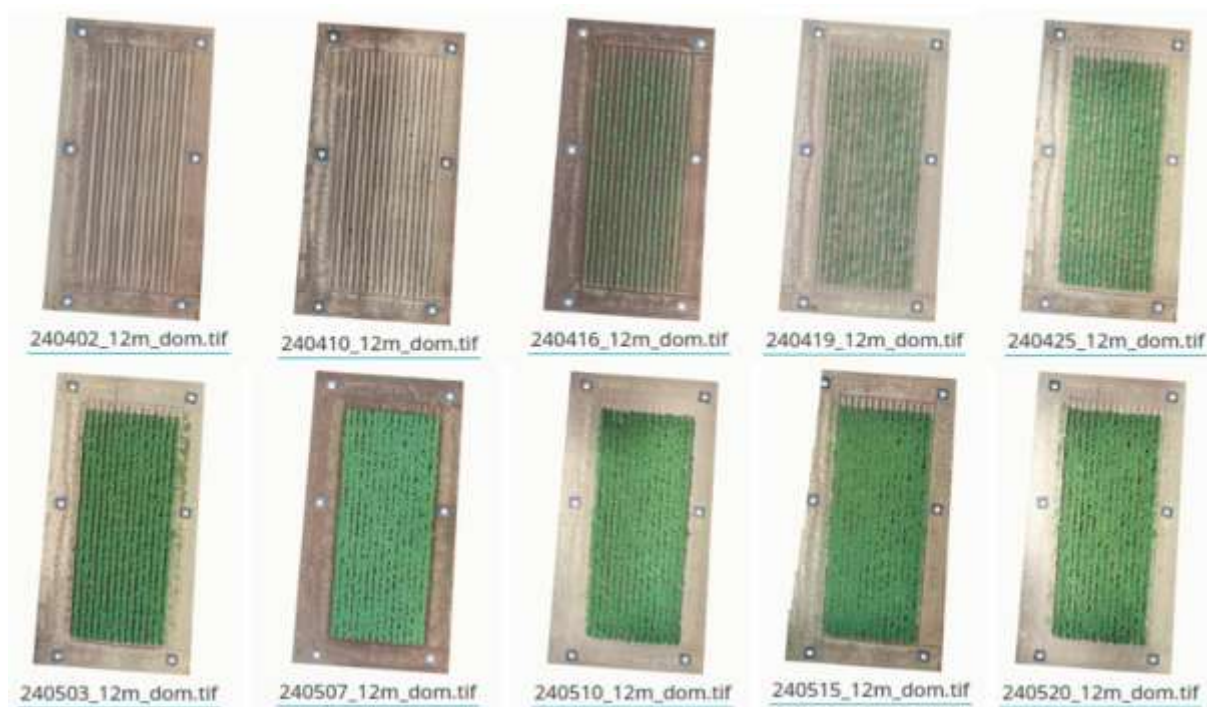
Collect time-series RGB and multi-spectral images by DJI P4 UAV at 12m



Implement batch processing scripts for automatic time-series reconstruction

## 2. Methods & Results

### 2) Image acquisition & 3D reconstruction



Time-series RGB ortho-maps  
(Nagasaki field)



Time-series NDVI maps  
(Nagasaki field)

## 2. Methods & Results

### 3) Traits calculation

Fraction of  
Coverage

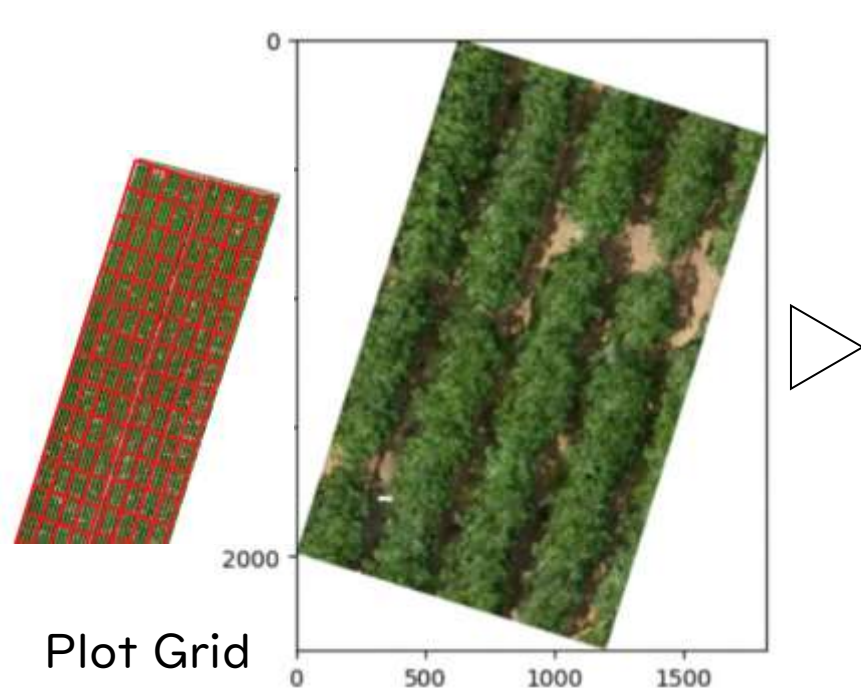
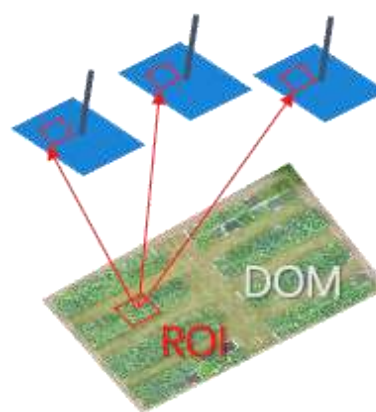
Canopy  
Height

Canopy  
NDVI

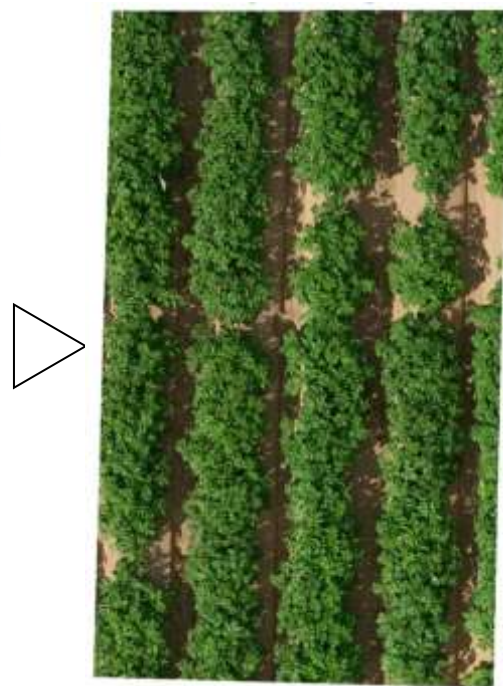


## 2. Methods & Results

### 3) Traits calculation - Coverage



Ortho-maps often  
has low quality



Use EasyIDP.v2 to  
obtain corresponding  
place on raw UAV image



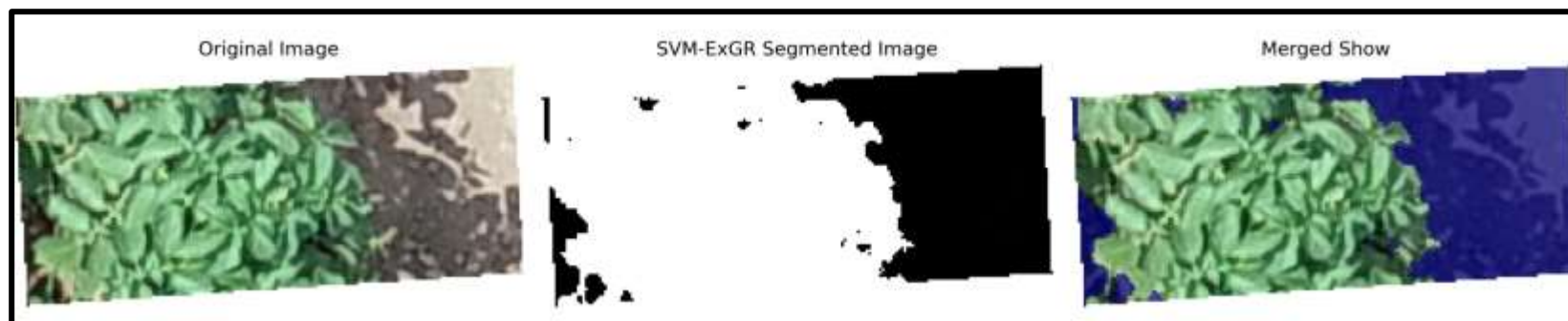
Train EasyPCC.v3

SVM-based segmentation machine  
learning model  
(input features: ExG | ExR | HSV | Lab)

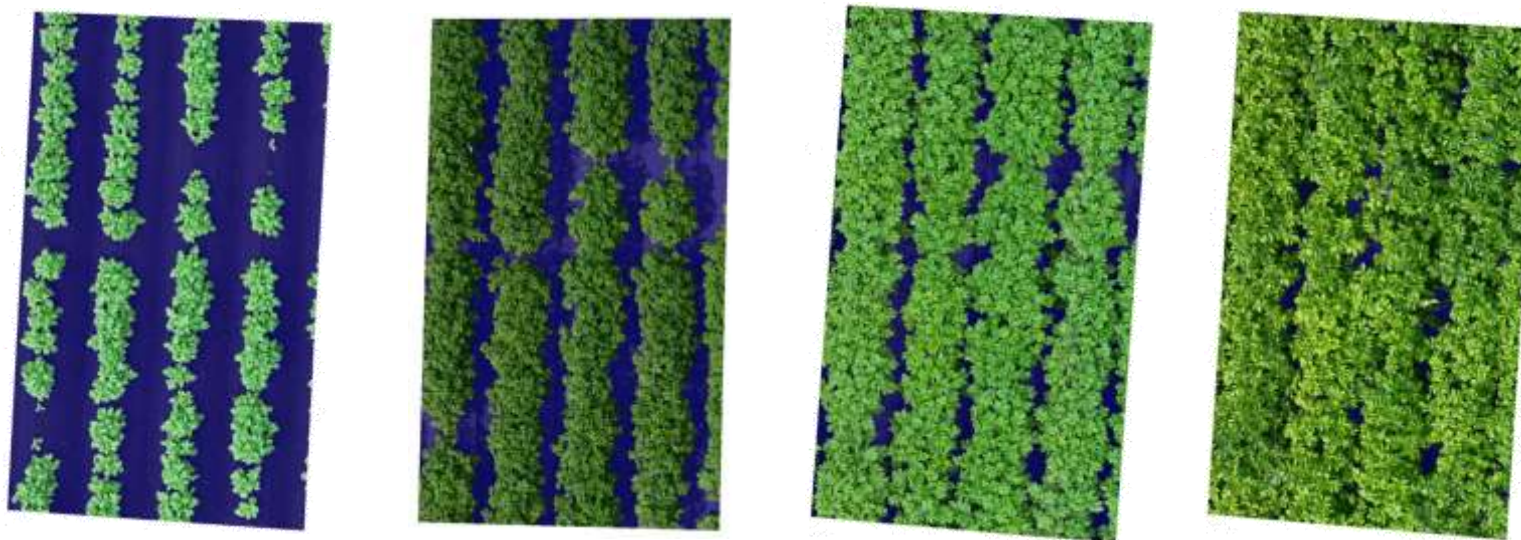
## 2. Methods & Results

### 3) Traits calculation – Coverage

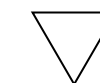
Nagasaki  
field



Hokkaido  
JA field



Processing  
speed:  
2s per plot



Fraction of  
Coverage

## 2. Methods & Results

### 3) Traits calculation

Fraction of  
Coverage

Canopy  
Height

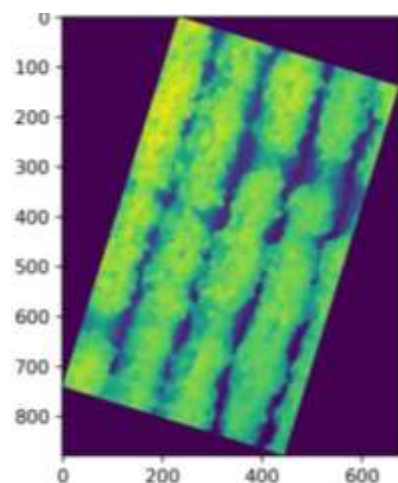
Canopy  
NDVI



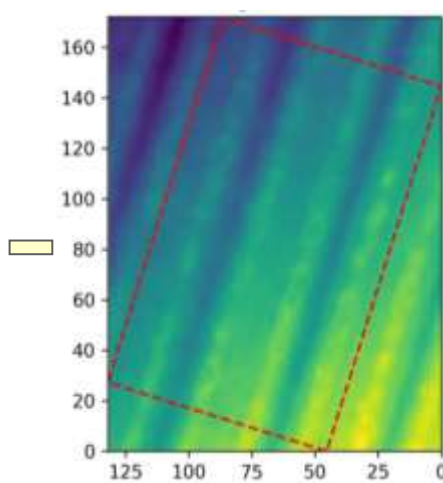
# 2. Methods & Results

## 3) Traits calculation - Height

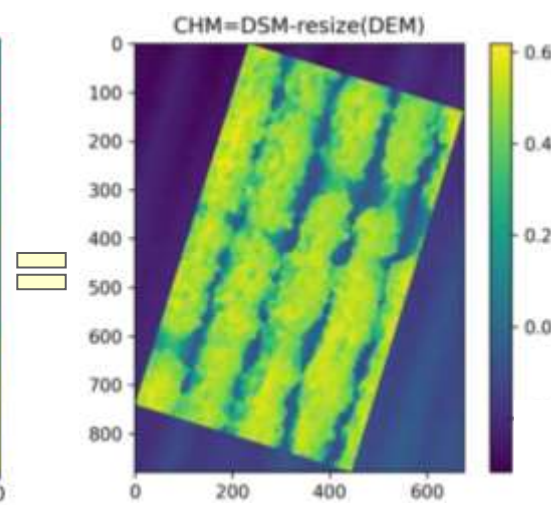
EasyIDP plot region extraction



Surface model of specific date  
( $DSM_i^{plot}$ )



Bare soil surface model at early stage  
( $DSM_0^{plot}$ )



Canopy surface model

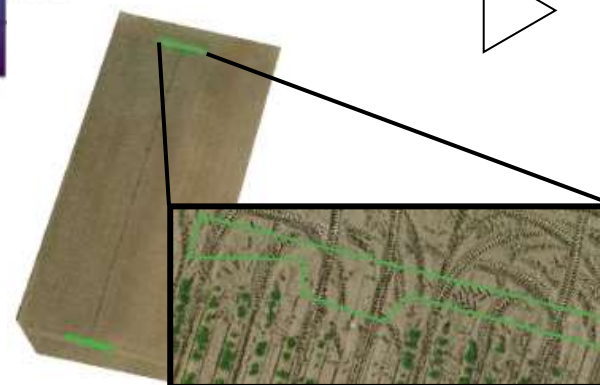
Nagasaki plot is stabilized by GCPs



95p meanHt

Means of over 95 percentile of CHM

Hokkaido JA plot without GCPs



Height offset calibration by bare region

$$= \overline{DSM_i^{cali}} - \overline{DSM_0^{cali}}$$

## 2. Methods & Results

### 3) Traits calculation

Fraction of  
Coverage

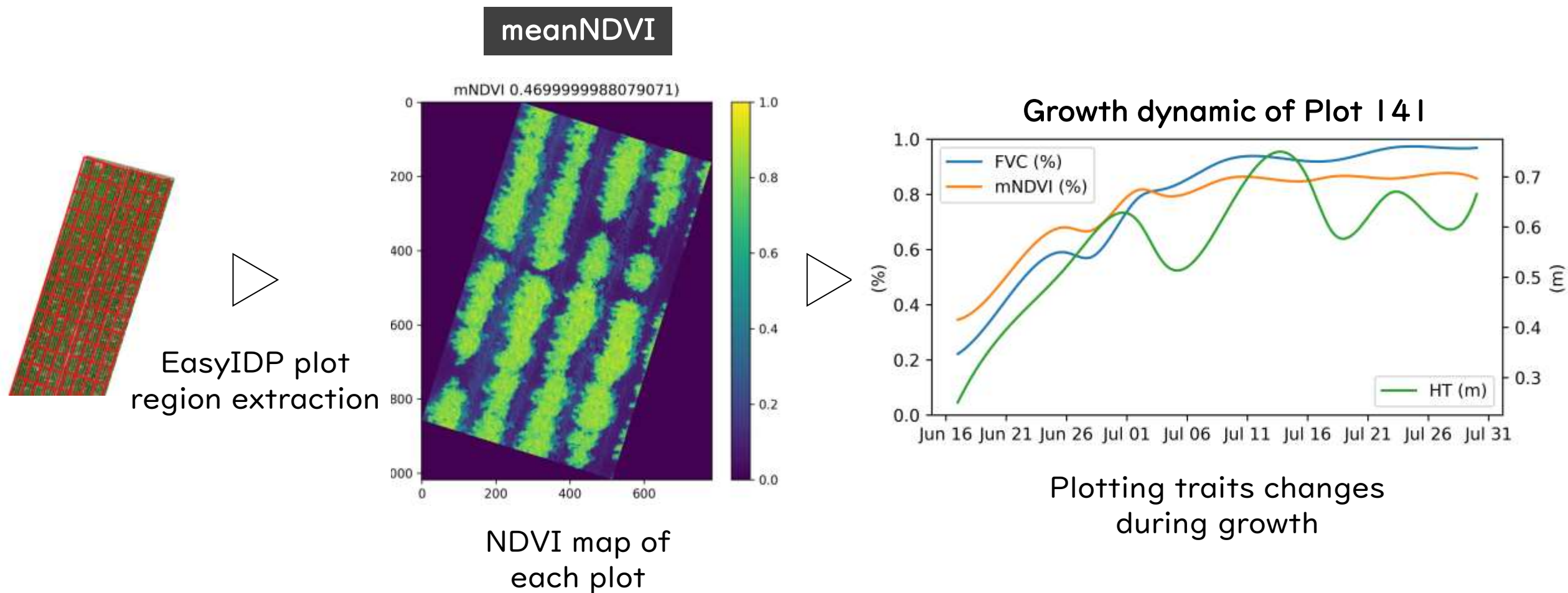
Canopy  
Height

Canopy  
NDVI



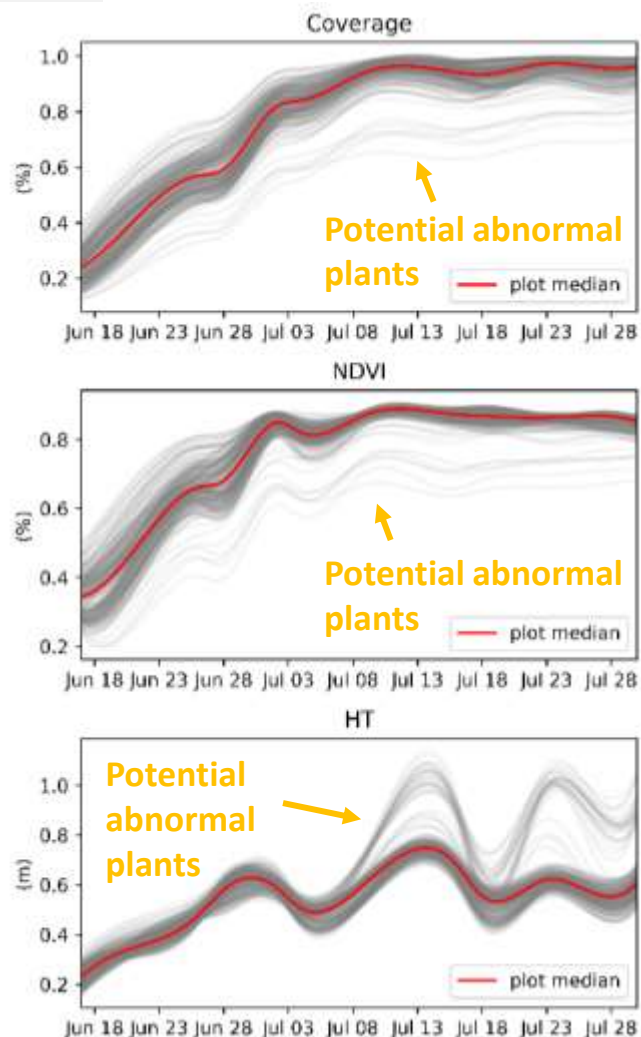
## 2. Methods & Results

### 3) Traits calculation – NDVI

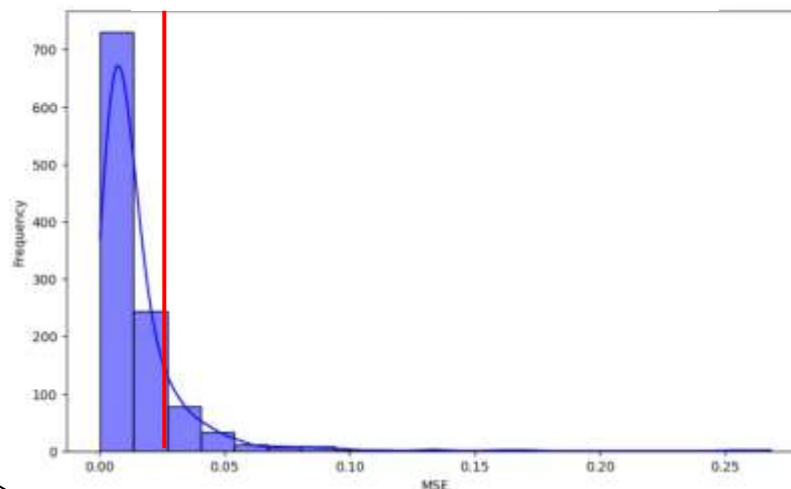


# 2. Methods & Results

## 4) Abnormal plot acquisition



Distribution of coverage MSE

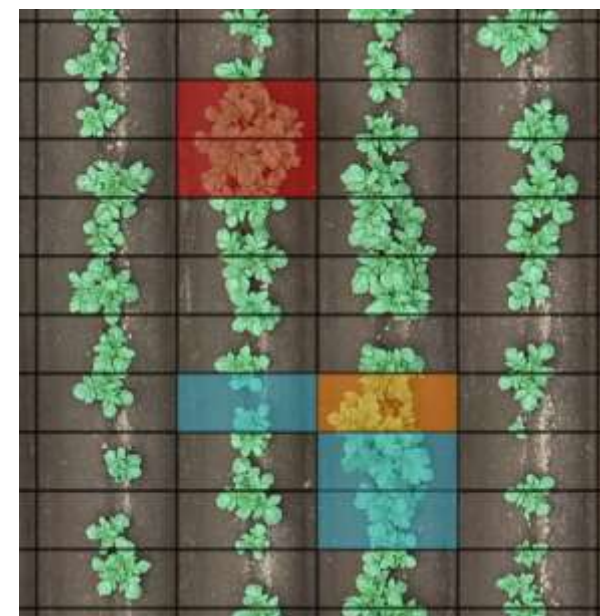


Estimate MSE of each plot to median

IQR outlier thresholding

$$= 75 \text{ percentile} - 25 \text{ percentile}$$

Counting outlier frequency of each trait



Appear

- 1 time Slight Abnormal
- 2 times Mid Abnormal
- 3 times Abnormal

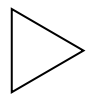
Growth dynamic of full plot

# 2. Methods & Results

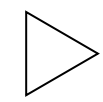
## 5) Guide field operation by Google Earth App

Export result to KML file  
for Google Earth App

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- Slight Abnormal
- Mid Abnormal
- Abnormal



Field check  
and remove

# ■ Results and Discussion

## 03

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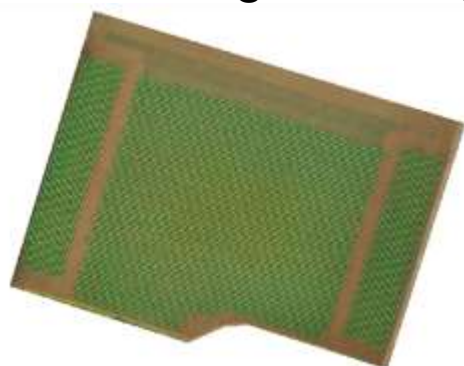


# 3. Results & Discussion

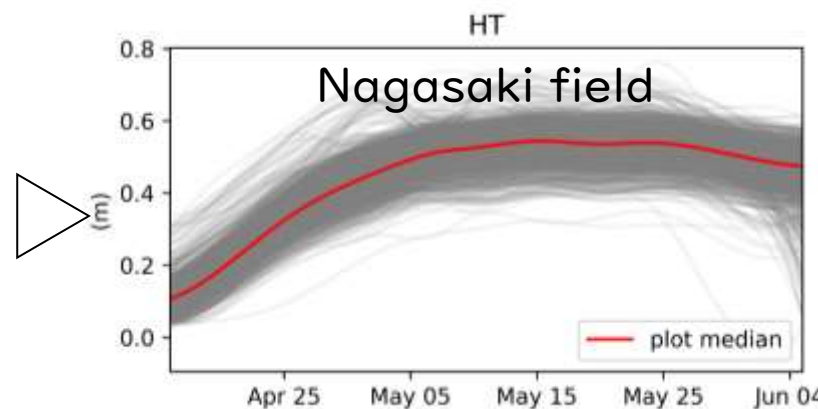
## 1) The effects of setting ground control points (GCP)



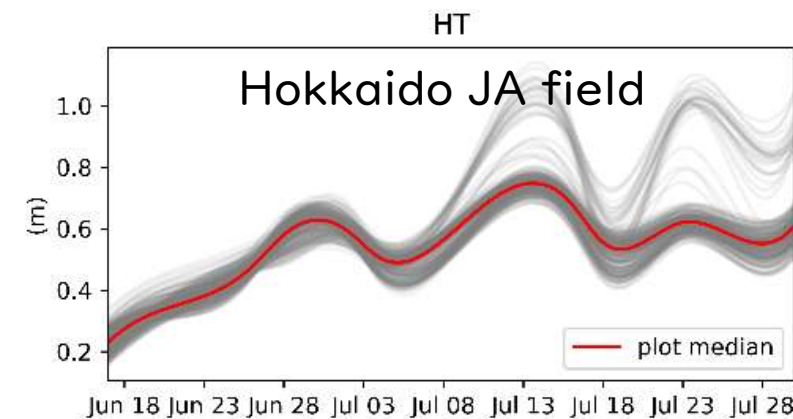
In pre-experiment,  
**Nagasaki field** sometimes  
involves large shifting



**Hokkaido JA field** has better RTK  
devices, acceptable on X-Y plane



**Setting GCP** to stabilize both  
X-Y plane and Z-axis shifting



**Without GCP** to has large  
Z-axis shifting



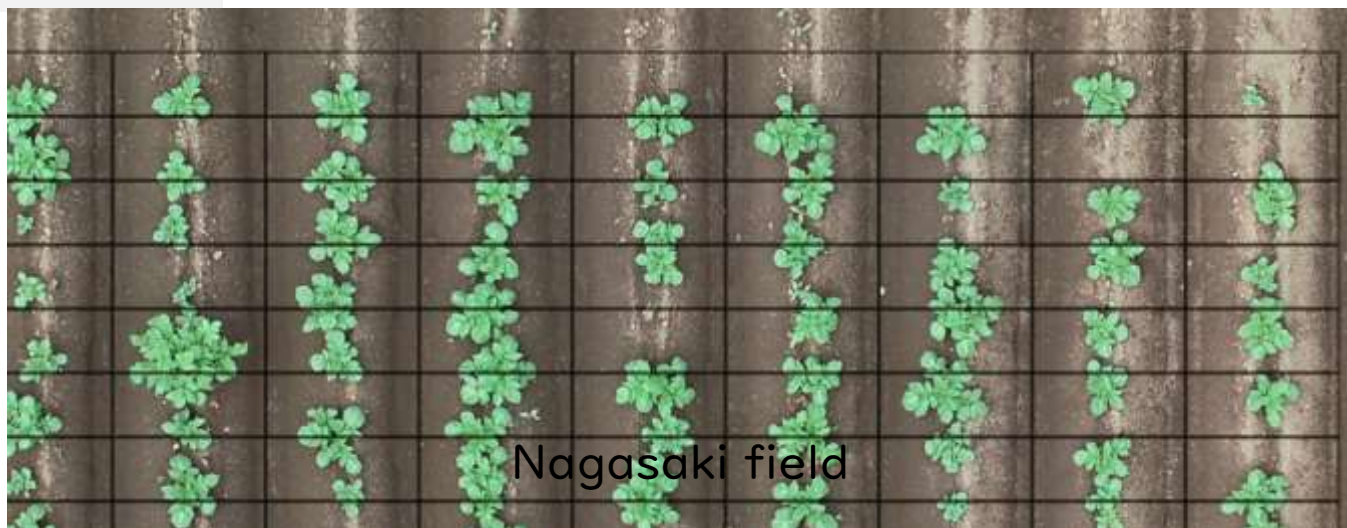
**Without GCP** produces  
variable plant height even with  
calibration by bare area





# 3. Results & Discussion

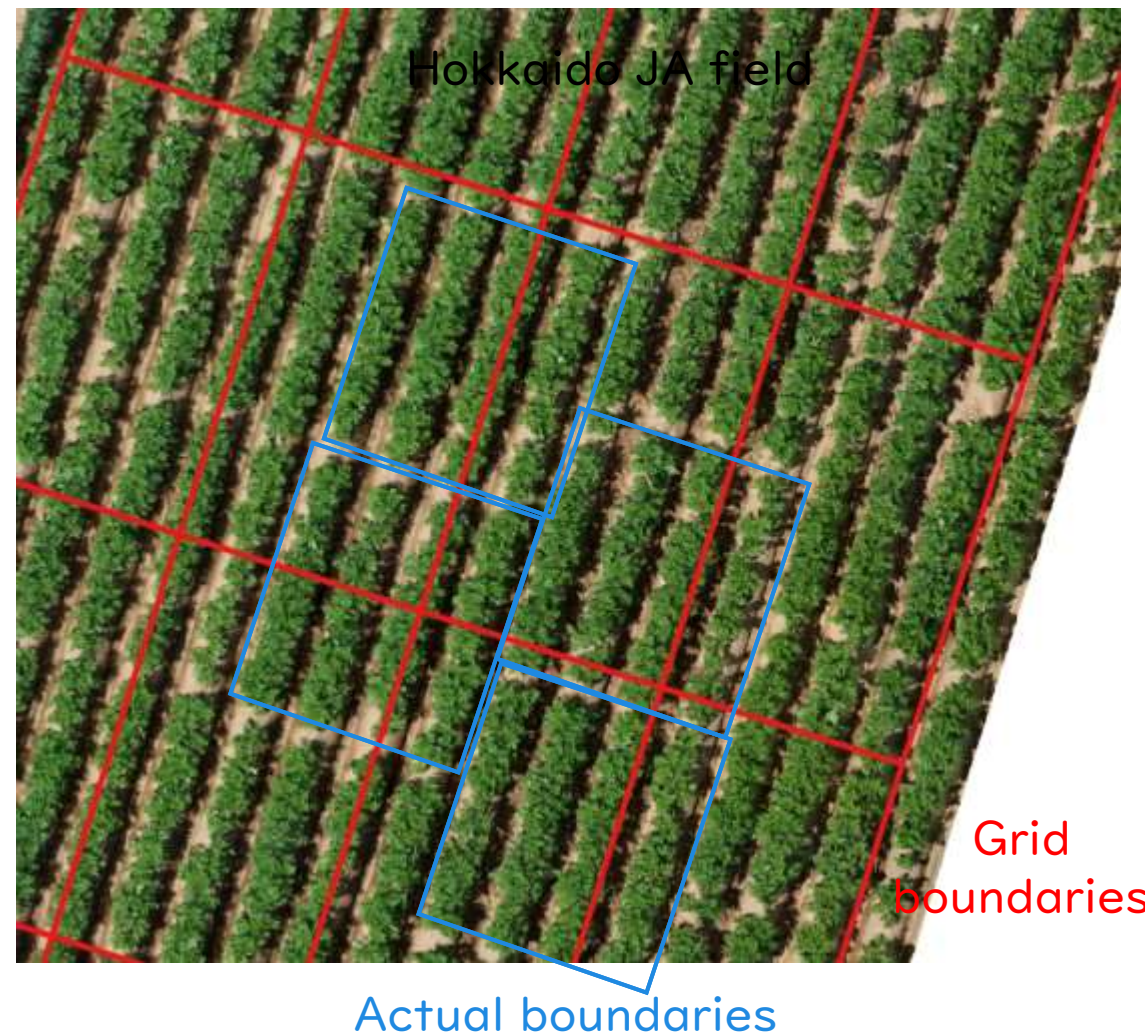
## 2) The necessity of root position detection



For quickly testing the feasibility of full pipeline,  
we simply generated grids not perfectly covering  
individuals and groups.



Need developing individual root position / group  
boundary extraction method for better analyzing results

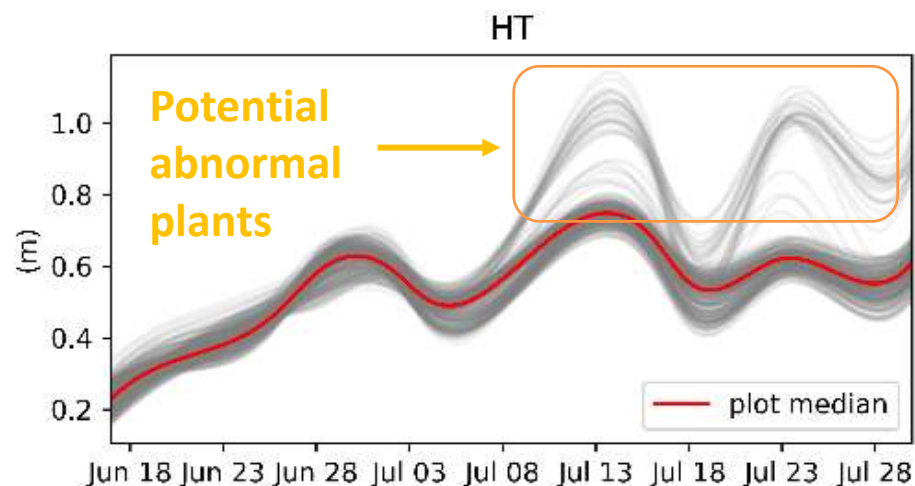




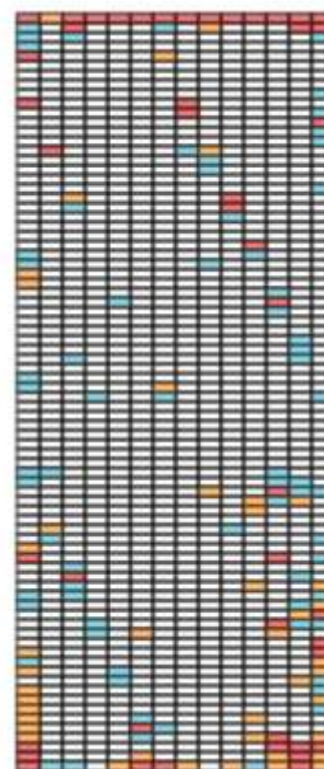
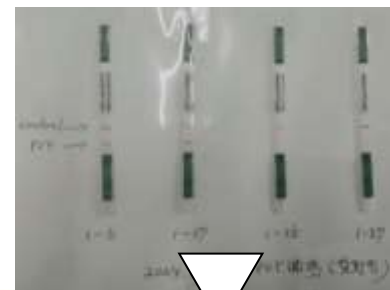
# 3. Results & Discussion

## 3) Abnormal miss detection for diseases

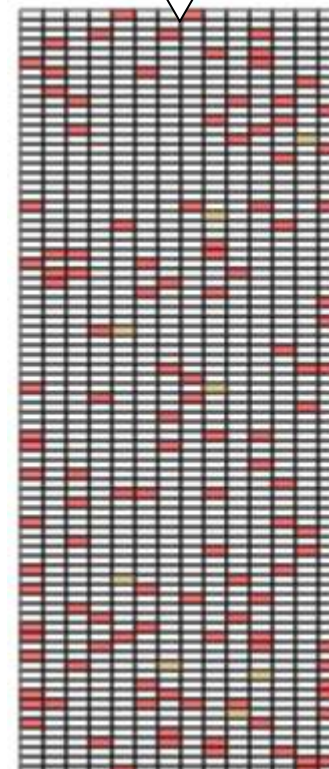
Current abnormal detection is statistical based “outliers”



The outlier abnormal results not catch with ground disease investigation well



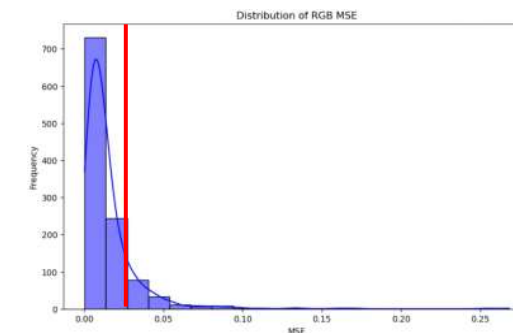
Pipeline outliers abnormal



Ground disease investigation



- Grid boundary not perfectly cover each plants



- Non-perfect outlier thresholding selection
- Only using HT, coverage, NDVI may not enough to reflect disease abnormal

Fusing RGB+Multi (3+4 layers) & Train time-series supervised machine learning classification models

# ■ Conclusions

## 04

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## 4. Conclusions

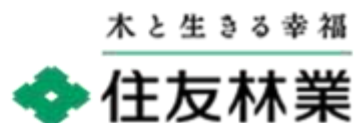
### Tested pipeline for detecting abnormal potato strains, including:

- Time-series UAV photogrammetry and batch processing workflow
- Coverage, NDVI and HT extraction by EasyIDP.v2 and EasyPCC.v3
- Abnormal detection based on statistical outliers
- Mobile app support for in-field operation

### Future works

- Better root position & group boundary detection
- Train image-based supervised machine learning models for abnormal detection
- Integrate full pipeline to GUI-based software

# Acknowledgement







# Thank You !

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<https://lab.fieldphenomics.com>