



THIS PROJECT HAS RECEIVED FUNDING FROM
THE EUROPEAN UNION'S HORIZON 2020 RESEARCH
AND INNOVATION PROGRAMME UNDER GRANT
AGREEMENT N. 773718



PTIMA

Novel sensing and machine learning techniques for in field disease detection

Workshop on Adoption of Digital Technology for Data Generation for the
efficacy evaluation of Plant Protection Products - 2022-06-27/29 - Ede (NL)

Gerrit Polder, Wageningen University & Research

Wageningen University & Research



- A university plus R&D organisation for innovation in the agrifood sector.
 - Working with industry, governmental authorities and other knowledge institutes
 - 6.500 employees
 - 12.000 students
 - 100 countries
 - 65 researchers on Agro Food Robotics
- Gerrit Polder
 - 30 years at Wageningen University & Research.
 - Senior scientist computer vision for plant phenotyping
 - Background: Electronics/Applied Physics.
 - PhD on Spectral Imaging

OPTimised Integrated Pest MAnagement **for precise detection and control of** **plant diseases in perennial crops and** **open-field vegetables**

An EU research project (Horizon 2020 framework)
1 September 2018 – 30 June 2022

5 Main Objectives



- **Optimize** plant disease prediction models and **develop** advanced early disease detection methods.
- Evaluate and screen biological and synthetic PPPs and assess plant and pathogen resistance mechanisms for **successful disease control**.
- Enhance and develop **innovative** precision spraying technologies.
- Test and evaluate the proposed **new IPM elements** under field conditions.
- Assess health, environmental and socioeconomic impacts and risks of the proposed **IPM system**.

- Early disease detection methods can be utilised as a tool to measure the effect of plant protection products.
- This presentation describes research on disease detection systems, for different crops and diseases, based on colour and spectral imaging and deep learning techniques to precisely localise and quantify infections, as applied in the OPTIMA project.

Field contexts where OPTIMA activities will be carried out :

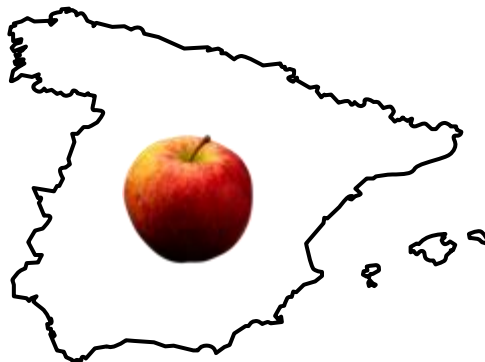
FRANCE



Crop: carrots in open field

Disease: *Alternaria leaf blight*

SPAIN



Crop: apple orchards

Disease: *Apple scab*

ITALY



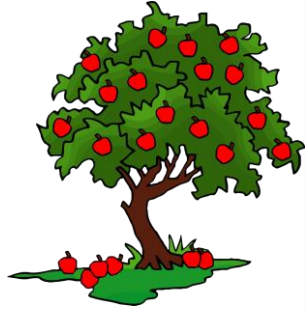
Crop: vineyards

Disease: *Grape downy mildew*

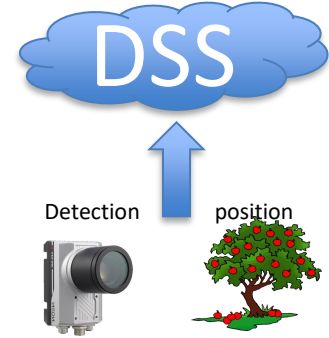
EDS DSS system overview



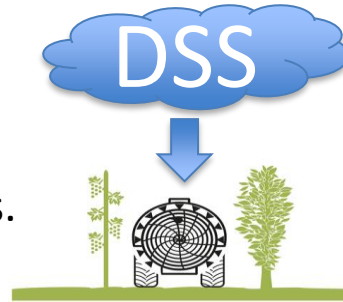
A camera based early detection system for scab in apple orchards.



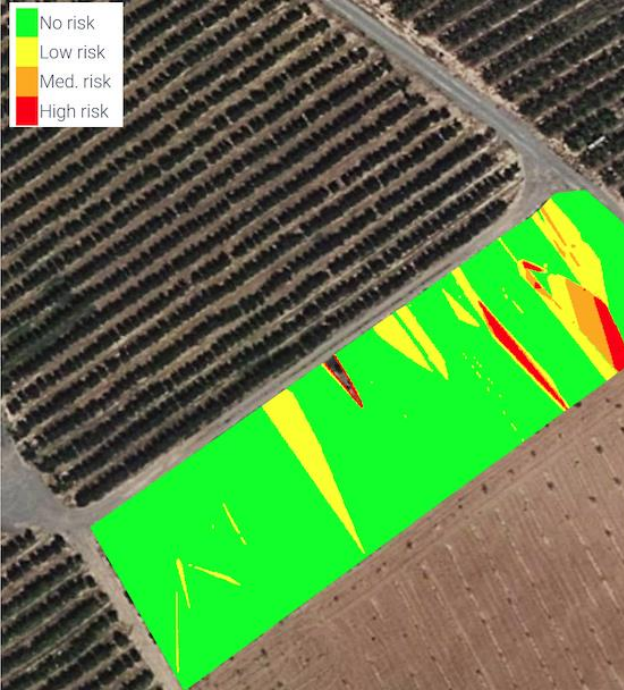
Detections are sent along with location information to a decision support system in the cloud.



The decision support system will advise on spraying actions.



DSS Prescription Maps





- RGB Imaging
 - Smart Camera
 - Deep Learning
- Spectral Imaging
 - Traditional machine learning
 - Deep learning

Deep Learning Approach

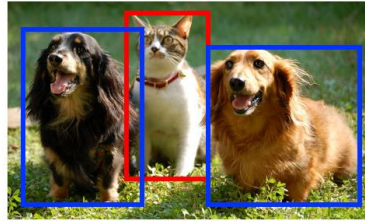


Classification



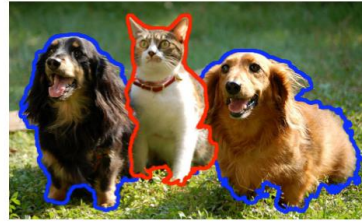
CAT

Object Detection



CAT, DOG

Instance Segmentation



CAT, DOG

- Classification (EfficientNet-B5)
 - Healthy/diseased
 - One hit, probability = confidence
- Detection (YOLO-V5)
 - Multiple spots in one image
- Fancy new thing (EfficientNet-B0):
 - Classification on image patches



Downy mildew in grapes:

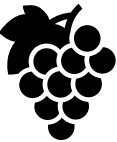
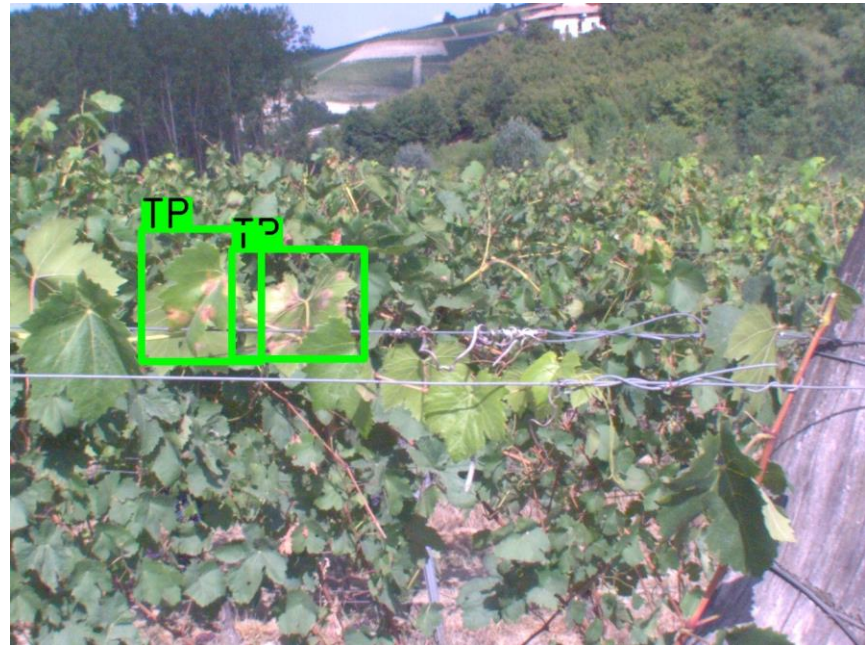
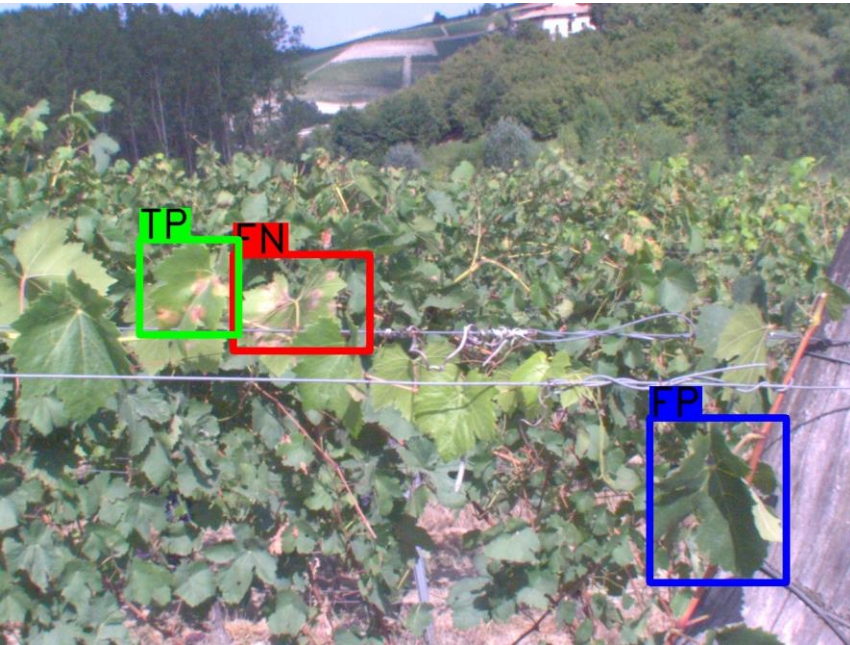
- Dataset 2019 + Dataset 2020 (IDS Campaign)
 - Training dataset: 57
 - Validation dataset: 47
 - Test dataset: 47
- Data from Greece and Italy
- Diseased spots were annotated by UNIBOX experts
- Three crop cultivars (Barbera, Moscato, Aglianico)
- Dataset 2021 (EDS Field campaign, Italy, 7007 images)
 - Training dataset: 235 (77%)
 - Validation dataset: 35 (11.5%)
 - Test dataset: 35 (11.5%)

Images taken automatically from a moving platform

Images taken in a 'controlled' way

Downy mildew in grapes:

- Detections original and after retraining on 2021 data



Downy mildew in grapes:



F1-score decreased from 86% and 67% on 2019/2020 data due to different image acquisition

More annotated data with more variation will increase this score

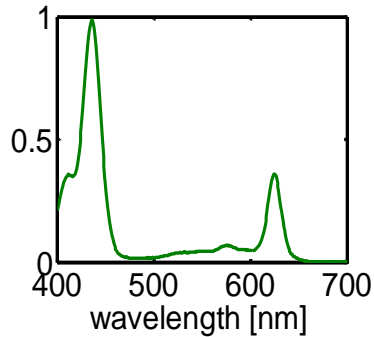
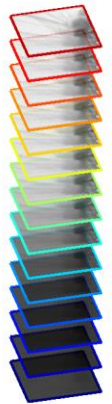
Ground-truth										
Downy-Mildew	47	58								
Healthy	118	-**	-							
Total	165	-**	-**		Total	79	-**	-**		
<i>Precision</i>	28.5%			<i>F1-score</i>	34.8%	<i>Precision</i>	77.2%		<i>F1-score</i>	66.3%

Original

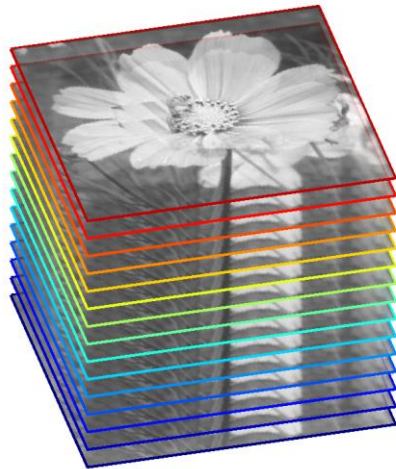
Retrained on 2021

Spectral Imaging

Spectroscopy



Spectral
Imaging



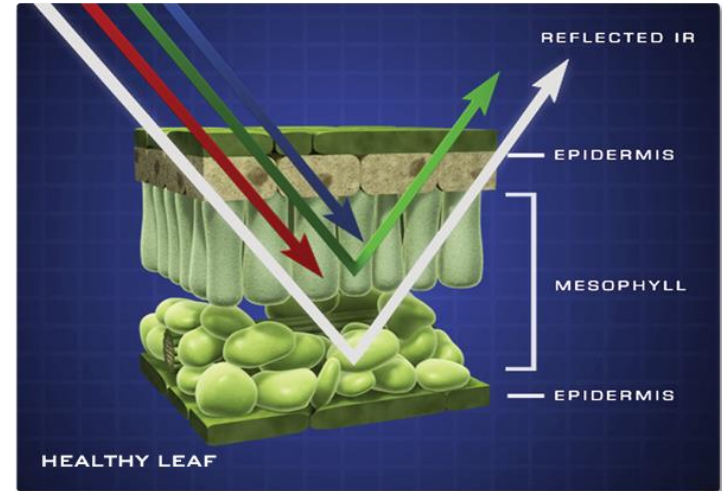
Imaging



Spectral Imaging

- Why Spectral Imaging?
 - Imaging access plant structure
 - Spectroscopy provides chemical composition
 - VIS (400-700 nm): pigments
 - NIR (700 – 2500 nm): moisture and internal structure of plant leaves

@Humboldt state University

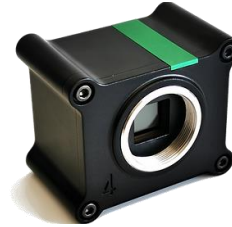


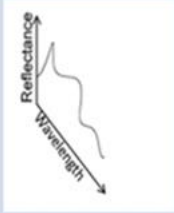
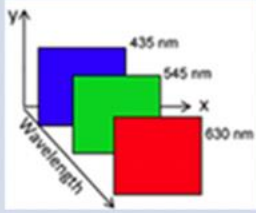
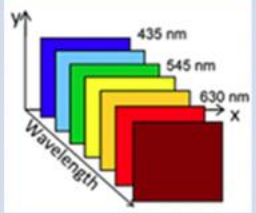
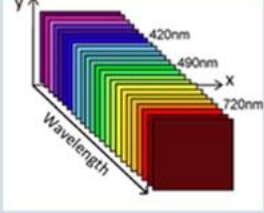
- Mishra, Puneet, et al. "Close range hyperspectral imaging of plants: A review." *Biosystems Engineering* 164 (2017): 49-67.

Spectral Imaging



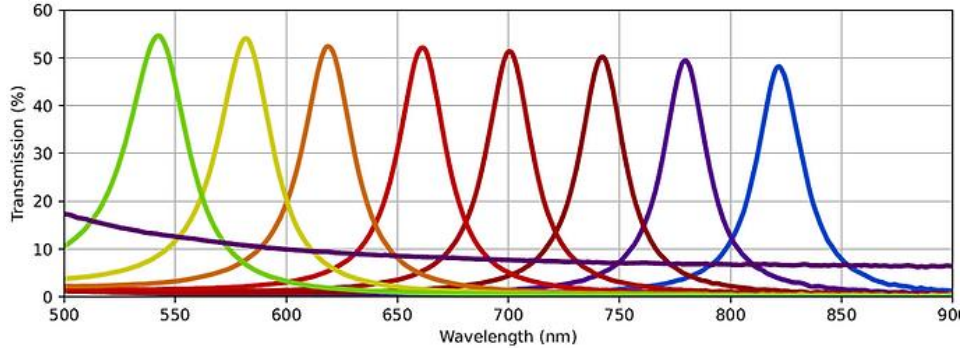
- Camera types
- Spectral/spatial resolution



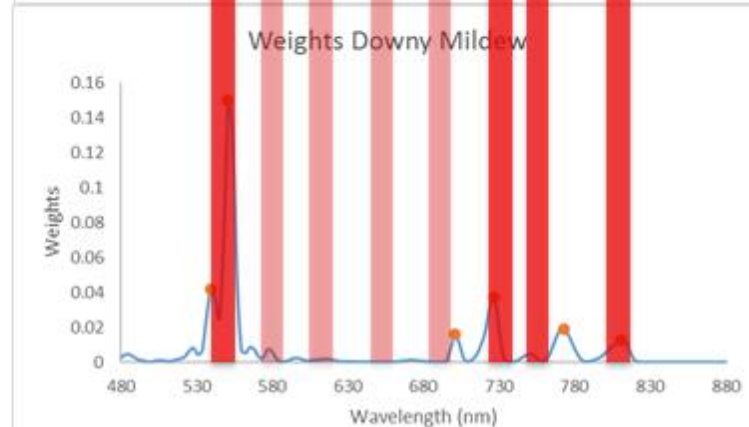
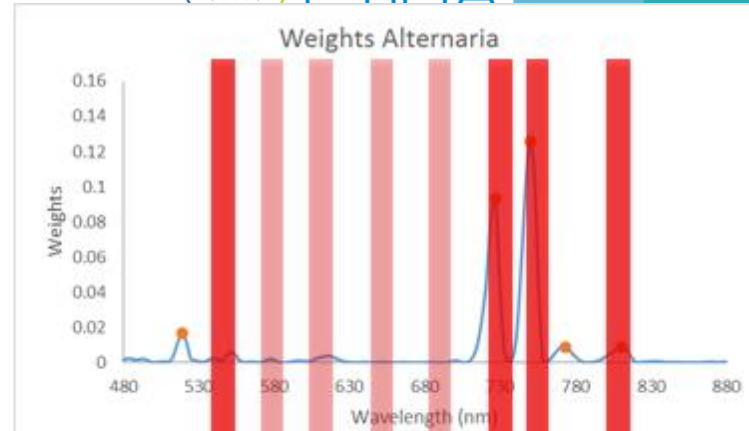
			
High spectral but zero spatial info Fast	High Spatial but very low spectral info Fast	High spatial and moderate spectral info Fast	High spatial AND spectral info Slow

Source: (<http://www.spectricon.com/spectral-imaging/>)

Spectral Camera Selection



- **SILIOS CMS-V 4**
 - 8 wavelength MS Camera
 - 682 x 682 resolution





Multispectral DL Dataset



- Performed on annotated Apple Scab data.
- Spectral Images converted into false color using three most discriminating Wavelengths (545 nm, 743 nm, 824 nm).
- RGB Yolo Neuro

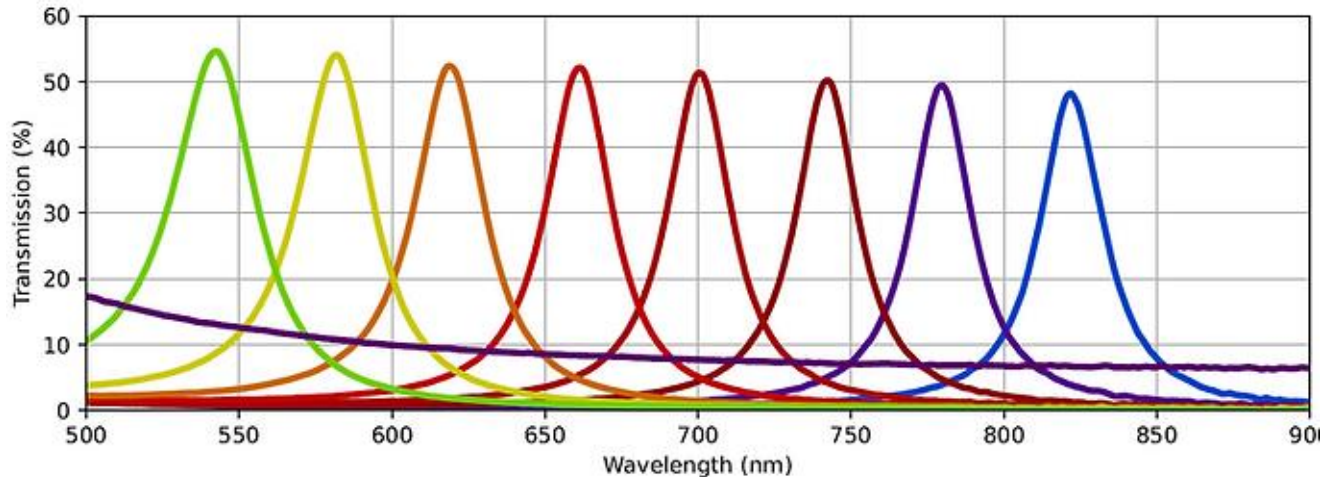
Ground-truth	
Apple scab	
Healthy	
Total	
Precision	100.0%
	76.9%

But our camera
has 8 bands



- Research question:

What is the effect on the YOLOv5 performance when using images with more channels (6 and 9) compared to using 3-channel images?



YOLOv5 multispectral network



- Standard YOLOv5 network from Ultralytics:
<https://github.com/ultralytics/yolov5>
- Dataloader added that can process the multichannel images (basically by stacking multiple 3-channel images into a bigger image tensor)
- Transfer-learning: duplicate the COCO weights of the first network layer (for the number of image channels):
 - 3 channel image: 1x
 - 6 channel image: 2x
 - 9 channel image: 3x



Image visualized - RGB



Image visualized – SegN image

Result of
Segmentation
network based
on 9 band
spectral Image
data

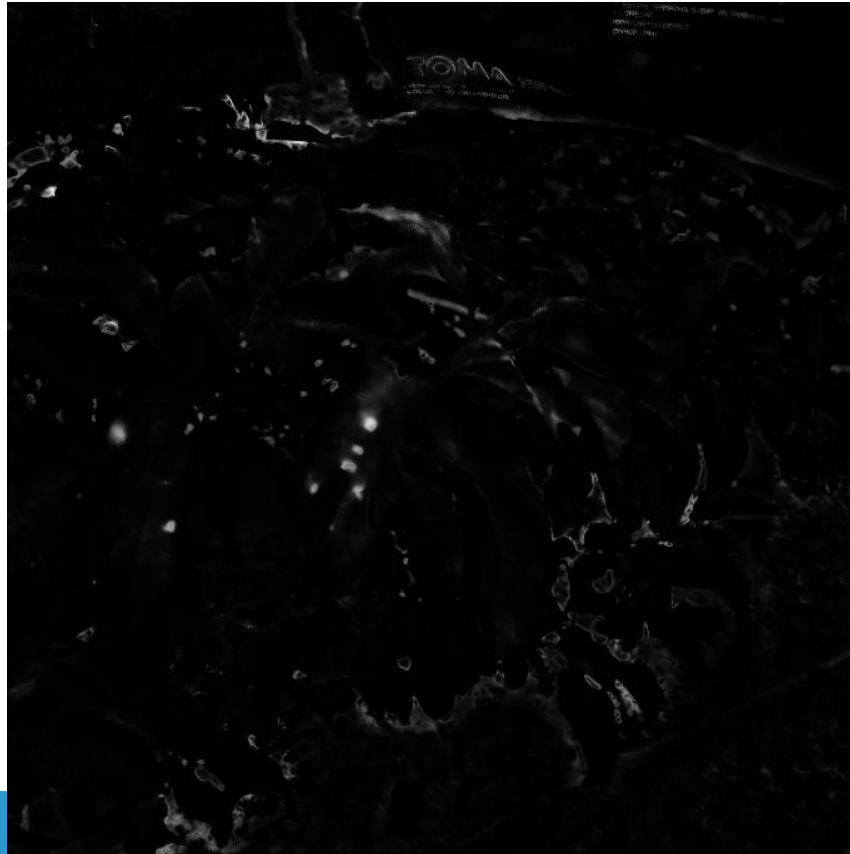


Image visualized – MS image 7,3,1



Image visualized – MS image 7,3 + SegN



PTIMA



Image visualized – MS image 7,3,1,5,2,6



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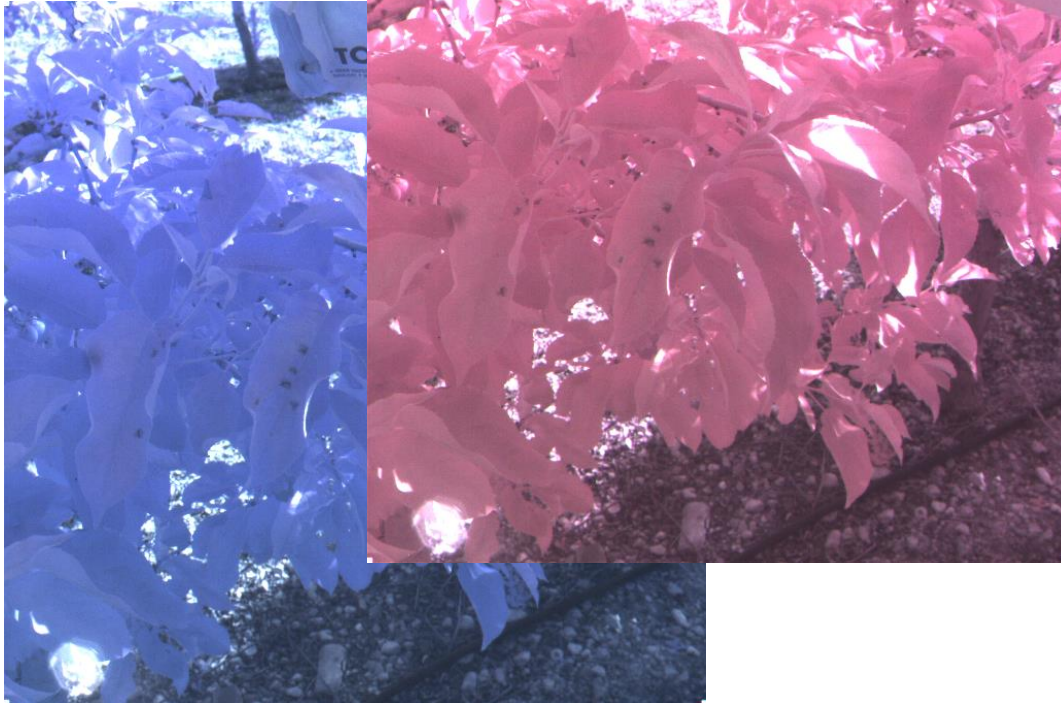


Image visualized – MS image

7,3,1,5,2,6,8,4,0



Preliminary results show that although symptoms are better visible in MS, classification performance is not significantly better

Take home message



- Disease detection as used in integrated pest management, can be utilised as a tool to measure the effect of plant protection products.
- Data is the key.
variation/annotation/pretraining
- More bands that are beneficial require a much larger annotated MS dataset.



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time for questions



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