

PHENET Webinar: biotic interaction in agroecosystems

Plant Health: Validated sensors and methodology applicable for biotic stresses in wheat

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29 November 2024,
Online

Objectives of this webinar

What question does this presentation address?

How to improve the crop resilience to biotic stress ?

How did we address this?

By building phenotyping tools using different optical sensors and developing models for plant disease detection with real time prediction

In particular, by the study of 5 approaches to assess wheat diseases in laboratory and in field

- *Handheld fluorometer in field (AGROSCOPE)*
- *RGB images in field (GEVES, AGROSCOPE, CRA-W)*
- *Multispectral Visible NIR imaging in laboratory and in field (GEVES)*
- *Hyperspectral NIR imaging in laboratory and in field (CRA-W)*
- *NIR Spectrometer and sorting system in laboratory (CRA-W)*

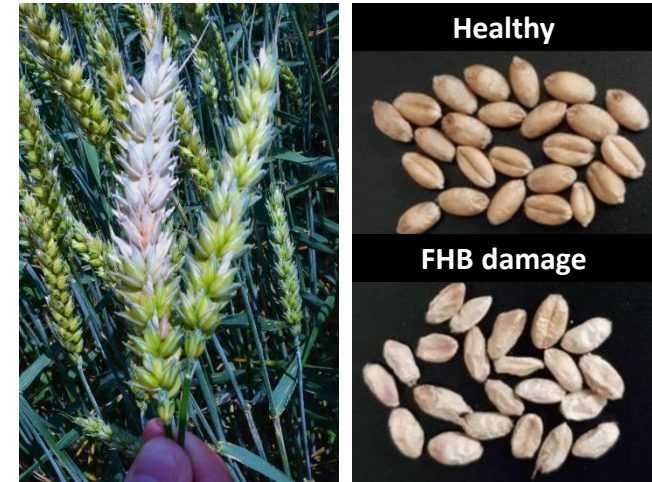
Study case: Plant Health: Fusarium Head Blight (FHB) in wheat

What is it? And why this choice?

Wheat is the major grain cereal crop cultivated in Europe

One of the major fungal diseases affecting wheat is Fusarium Head Blight (FHB)

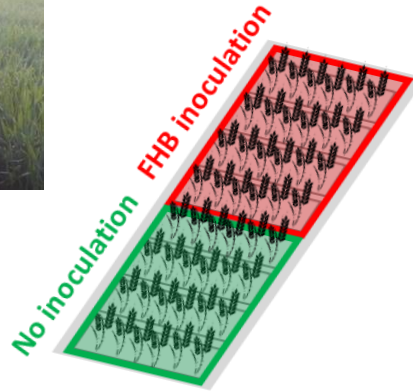
- ❑ Infects the ears at the flowering stage
 - ❑ Damages the kernels
 - yield and quality decrease
 - ❑ Can produce mycotoxins
 - risk for human and animal health
-
- ➔ Fungicide treatment is only preventive and not fully effective ...
 - ➔ Disease-tolerant varieties are needed
 - ➔ Symptoms observable at the canopy level



Source: Birr T. et al. (2020)



Study case: Plant Health: Fusarium Head Blight (FHB) in wheat Experimentation and classical method

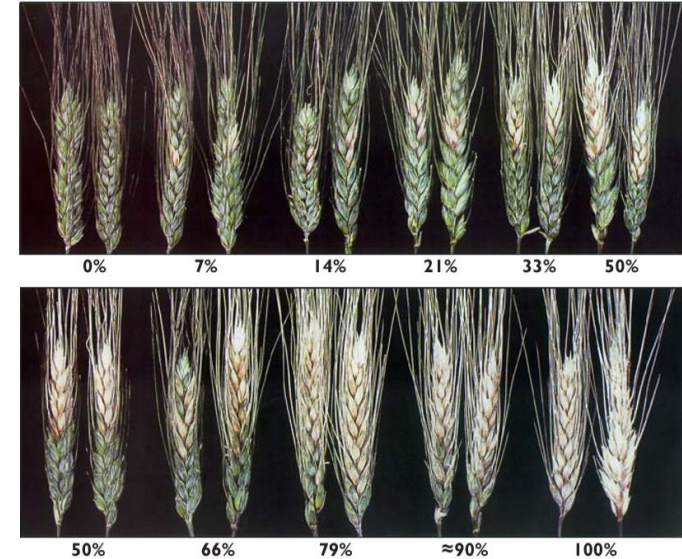


Varietal trial

Fusarium Head Blight (FHB) inoculation

Visual observation according to a reference scale
at ear level (number of spikelets infected by ear)
at plot level (number of ears infected by plot)

A Visual Scale to Estimate Severity of Fusarium Head Blight in Wheat



Source: Stack R.W. & McMullen M. (2011)



PHENWHEAT



1st approach: Fluorescence measurement: acquisition

- +/- 10000 fluorescence measurements in the field by Agroscope

Based on the chlorophyll fluorescence

Green and at maturity ears

Contact sensing ☹️

Low cost 😊

Time consuming ☹️

Point to point measurement ☹️

Not early detection, need high infection ☹️



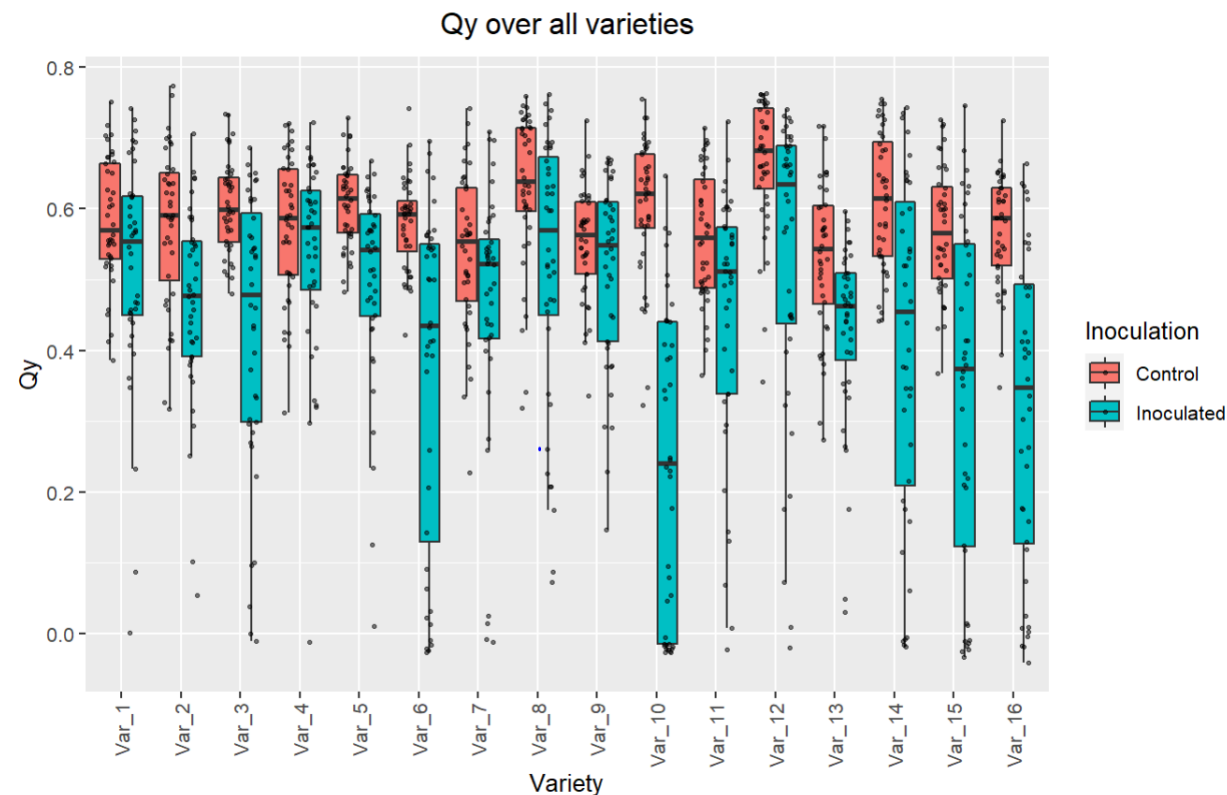
FluoroPen



1st approach: Fluorescence measurement: results

- *Quantum yield (Qy) results for measurements on ears in the field showing a very large variability in relation to strong infections*

Index	2	3	4	5	6	7
Time	10:27:54 29.3.2018	10:29:29 29.3.2018	10:31:45 29.3.2018	10:35:52 29.3.2018	10:22:44 3.4.2018	10:23:11 3.4.2018
	49° 20.3871' N 16° 28.6379' E	49° 20.3538' N 16° 28.6755' E	49° 20.2923' N 16° 28.6290' E	49° 20.2557' N 16° 28.5246' E	Qy 0.67	Qy 0.04
	Qy 0.72	Qy 0.65	Qy 0.27	Qy 0.67	Fo Backgr 378 Fo Flash 3310	Fo Backgr 897 Fo Flash 976
	Fo Backgr 299 Fo Flash 4985	Fo Backgr 378 Fo Flash 2711	Fo Backgr 89 Fo Flash 1069	Fo Backgr 438 Fo Flash 3110	Fm Backgr 398 Fm Flash 9331	Fm Backgr 864 Fm Flash 946
	Fm Backgr 299 Fm Flash 17138	Fm Backgr 418 Fm Flash 2058	Fm Backgr 92 Fm Flash 1436	Fm Backgr 418 Fm Flash 8544		
Value						



S. Treier (2025). Digital optical lean phenotyping methods in the context of wheat variety testing. Thesis in Agroscope

1st approach: Fluorescence measurement: next steps

- Assess its potential application under natural infections in a variety testing network to improve the comparability among campaigns, sites and operators



2nd approach: RGB imaging: acquisition

- +/- 3000 RGB images acquired by Agroscope, CRA-W, GEVES

Based on the color (green/white)

Green ears

Proximal sensing 😊

Low cost 😊

Random and quick acquisition 😊

Image on full ears acquired with an angle of 45°/90° 😊

Early detection when one spikelet is infected 😊

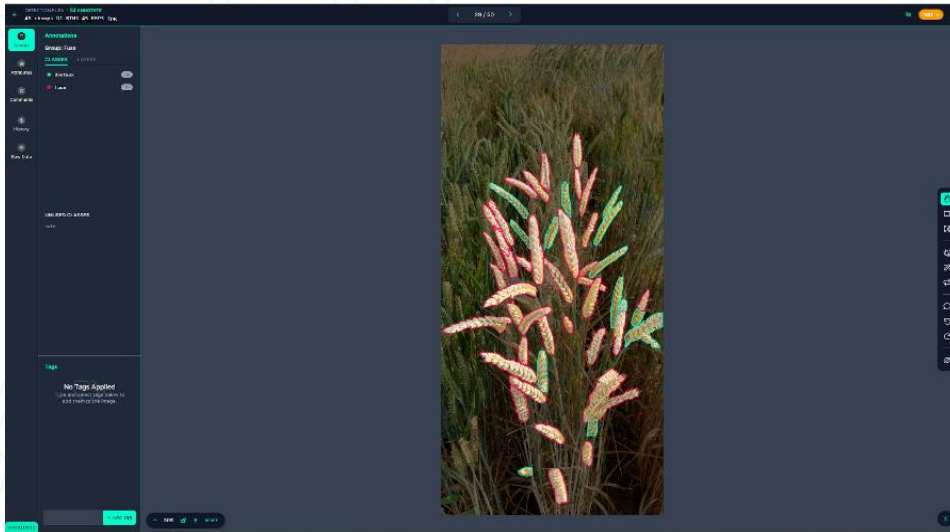


2nd approach: RGB imaging: Annotation models

- Plateforme used: **Roboflow**
- Exemple of annotations with **2 classes: healthy / Fusarium**
- 3 methods: manual, **semi-automatic model (« SAM »)** and **development of an automatic model**
- Nb images annotated by GEVES: ≈ 800 images=239 from CRA-W + 561 from GEVES

239 Total Images

[View All Images →](#)



2nd approach: RGB imaging: method of prediction

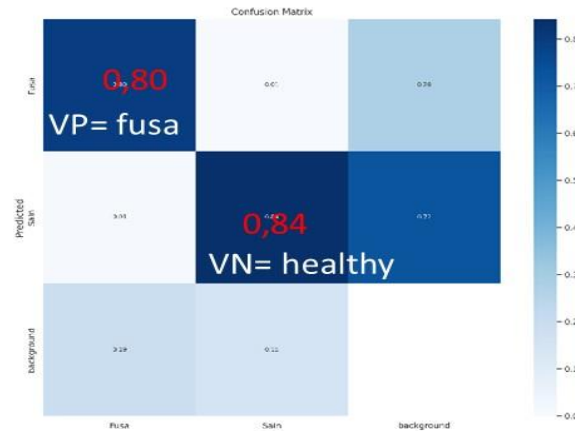
- 1 model selected upon 17 models of Deep learning



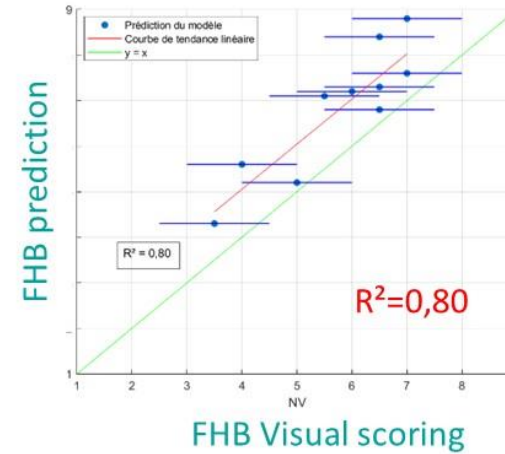
Segmentation

Class	Precision	Rappel	mAP50	mAP50-95):
all	0.794	0.728	0.806	0.474
Fusa	0.823	0.72	0.808	0.481
Sain	0.766	0.736	0.804	0.468

Confusion matrix

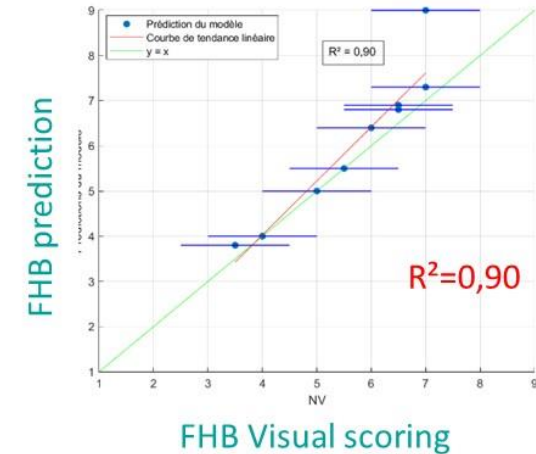


Without detection of spikes



With detection of spikes

(model coming from Global Wheat Challenge 2021)



High correlation between FHB predictions & visual scorings in field



2nd approach: RGB imaging: method of prediction

RGB image
in perspective



FHB prediction by Deep learning (Yolo V8)

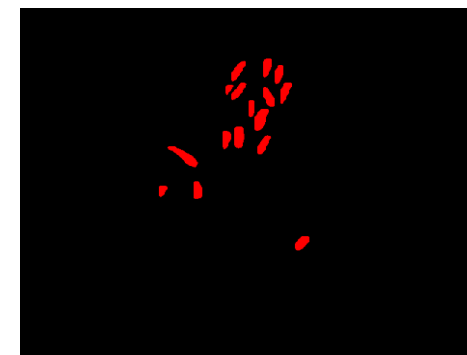
prediction



All Mask



Fusa mask



% FHB area=Fusa mask/all mask ; Ex:9.03 %



2nd approach: RGB imaging: comparisons of DL performance between sensors



Camera
Sony



Smartphone
Samsung A 54



Smartphone
Google Pixel 8

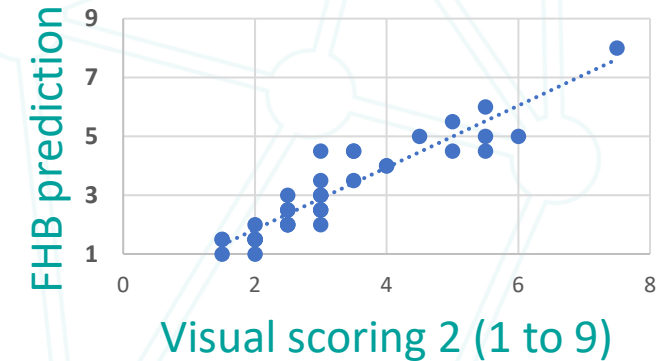


Groupe	R ²	REP1	REP2	REP3	MOY
BTH1	D1NV1	0,62	0,60	0,46	0,56
	D2NV2	0,27	0,69	0,76	0,58
BTH2	D3NV2	0,87	0,77	0,74	0,79
BDH	D3NV2	0,58	0,60	0,66	0,61

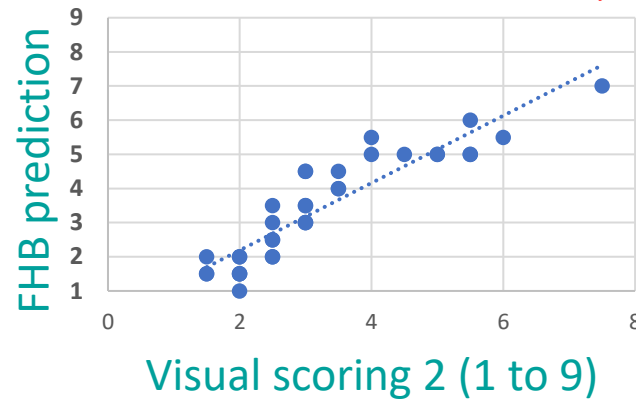
Groupe	R ²	REP1	REP2	REP3	MOY
BTH1	D1NV1	0,61	0,37	0,37	0,45
	D2NV2	0,46	0,66	0,41	0,51
BTH2	D3NV2	0,83	0,73	0,74	0,77
BDH	D3NV2	0,81	0,55	0,68	0,68

Groupe	Groupe	R ²	REP1	REP2	REP3	MOY
BTH1	BTH1	D1NV1	0,85	0,55	0,35	0,58
		D2NV2	0,38	0,69	0,44	0,50
BTH2	BTH2	D3NV2	0,89	0,88	0,81	0,86
BDH	BDH	D3NV2	0,59	0,82	0,91	0,77

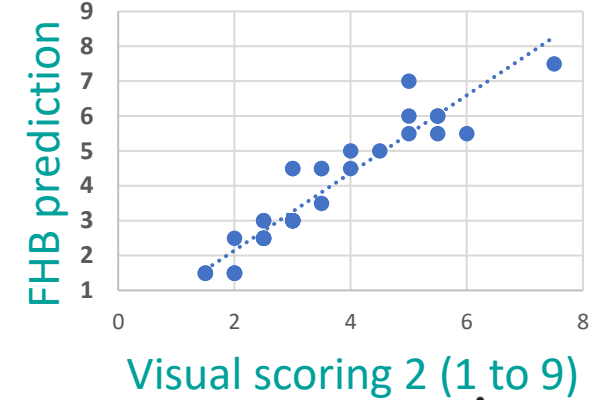
BTH2: D3 NV2 rep1 $R^2 = 0,87$



BTH2: D3 NV2 rep1 $R^2 = 0,83$



BTH2: D3 NV2 rep1 $R^2 = 0,89$



The same Deep Learning model adapted to different low cost RGB sensors



2nd approach: RGB imaging: user interface to collect the % FHB area/image

Analysis for 1 several image

Détection d'objets YOLOv8 et Visualisation des Masques

Analyse d'une seule image Analyse de plusieurs images

Analyse d'une seule image

% FHB area
ex : 64,8%

Choisissez une image...

Drag and drop file here
Limit 200MB per file • JPG, JPEG, PNG

Browse files



Analysis for several images

Analyse d'une seule image Analyse de plusieurs images

Analyse de plusieurs images

Choisissez des images...

Drag and drop files here
Limit 200MB per file • JPG, JPEG, PNG

Browse files

- crop_49-Sony-D2-BTH1-41-REP3.JPG 0.8MB
- crop_49-Sony-D2-BTH1-40-REP3.JPG 1.0MB
- crop_49-Sony-D2-BTH1-39-REP3.JPG 0.7MB

Showing page 1 of 14

	Image	Pourcentage de Fusa
0	crop_49-Sony-D2-BTH1-1-REP3.JPG	64.8226
1	crop_49-Sony-D2-BTH1-2-REP3.JPG	86.5061
2	crop_49-Sony-D2-BTH1-3-REP3.JPG	80.0881
3	crop_49-Sony-D2-BTH1-4-REP3.JPG	16.9305
4	crop_49-Sony-D2-BTH1-5-REP3.JPG	14.4327
5	crop_49-Sony-D2-BTH1-6-REP3.JPG	88.2476
6	crop_49-Sony-D2-BTH1-7-REP3.JPG	86.6072
7	crop_49-Sony-D2-BTH1-8-REP3.JPG	78.1664
8	crop_49-Sony-D2-BTH1-9-REP3.JPG	77.9965
9	crop_49-Sony-D2-BTH1-10-REP3.JPG	98.9945

Télécharger les résultats en Excel

% FHB area loaded on Excel

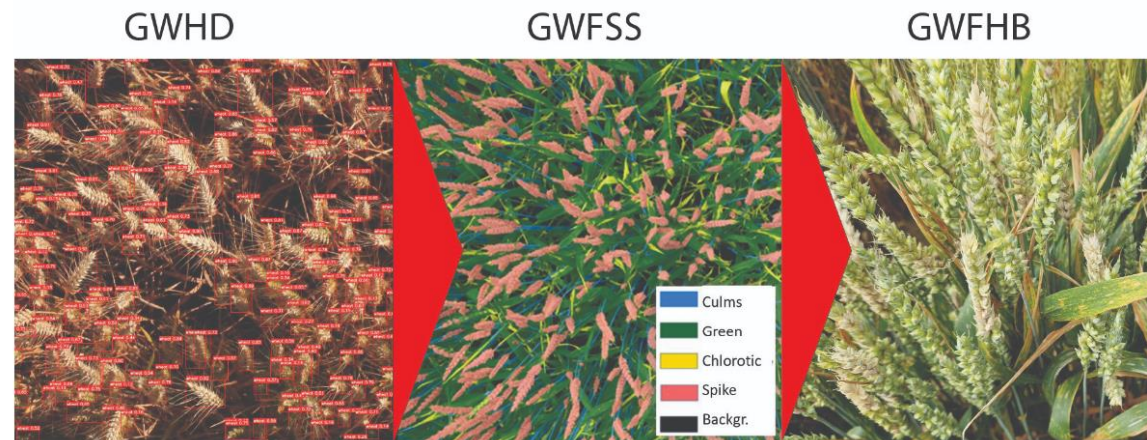
Image	Pourcentage de Fusa
crop_49-Sony-D2-BTH1-1-REP3.JPG	64,82
crop_49-Sony-D2-BTH1-2-REP3.JPG	86,51
crop_49-Sony-D2-BTH1-3-REP3.JPG	80,09
crop_49-Sony-D2-BTH1-4-REP3.JPG	16,93
crop_49-Sony-D2-BTH1-5-REP3.JPG	14,43
crop_49-Sony-D2-BTH1-6-REP3.JPG	88,25
crop_49-Sony-D2-BTH1-7-REP3.JPG	86,61
crop_49-Sony-D2-BTH1-8-REP3.JPG	78,17
crop_49-Sony-D2-BTH1-9-REP3.JPG	78,00
crop_49-Sony-D2-BTH1-10-REP3.JPG	98,99



2nd approach: RGB imaging: results



- Contribution to the GWFHB: Global Wheat Fusarium Head Blight



Head
Detection

Full Semantic
Segmentation

Fusarium
Head Blight

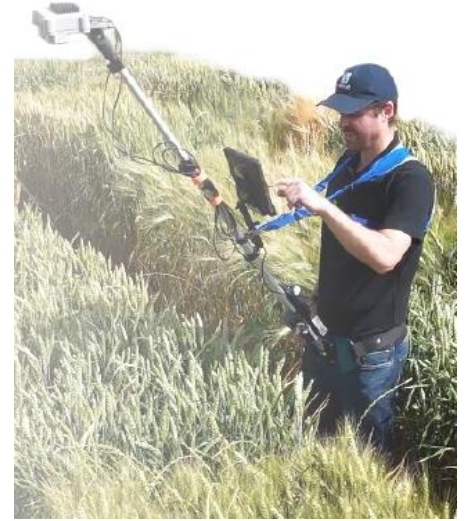


<https://www.global-wheat.com/gwfss.html>



2nd approach: RGB imaging: next steps

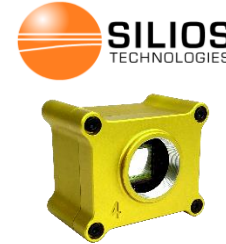
- Guidelines document for images
(semi-automatic) annotating to assess biotic stress
- Use of the annotated images
for model development by GEVES
- Use of this model to extend the portfolio of agronomics traits
on existing phenotyping device such as:
 - the Mobile-based Rapid Phenotyping (MoRPH) application developed by WUR
 - the Literal stick developed by Hiphen



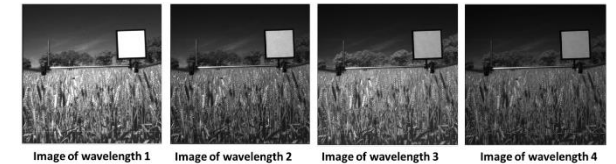
3rd approach: VNIR multispectral imaging: acquisition



- 383 images (1658 spikes) acquired in laboratory by GEVES
- 560 images acquired on 7 sites in field by GEVES

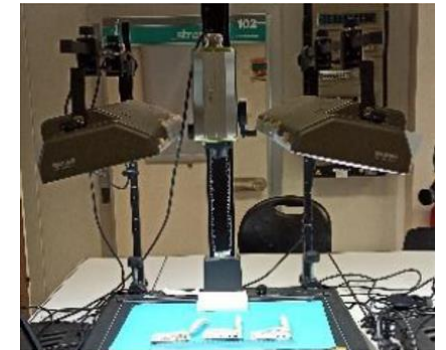
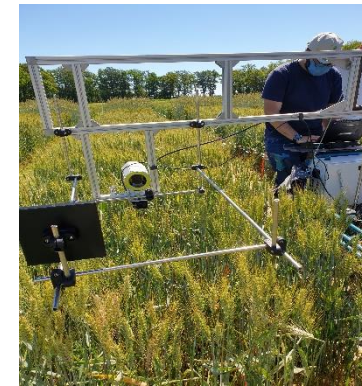


Multispectral imaging - plane scan
CMS 4: 4 wavelengths for FHB detection (3 RGB & 1 NIR bands)



Based mainly on the Chlorophyll content
Green and yellow ears (until 550°day post inoculation)
Proximal sensing 😊
Rather specific of FHB (≠Microdochium or Yellow rust) 😊
High cost 😞
Time consuming 😞
Image on full ears acquired in frontal view (1 row of ears) 😊
Possible early detection when one spikelet is infected 😊

In laboratory



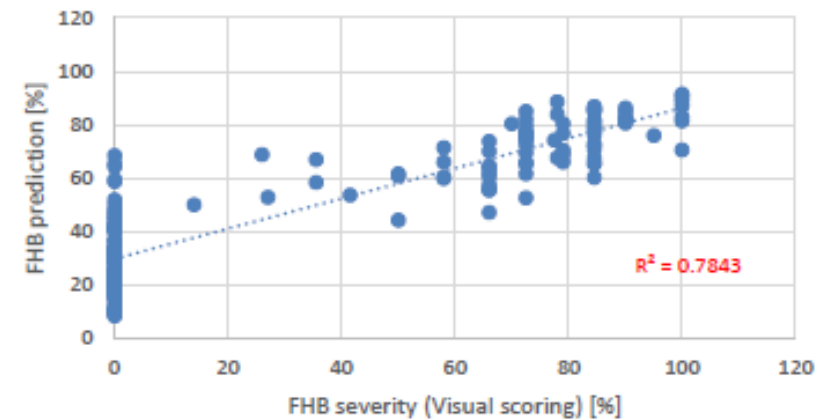
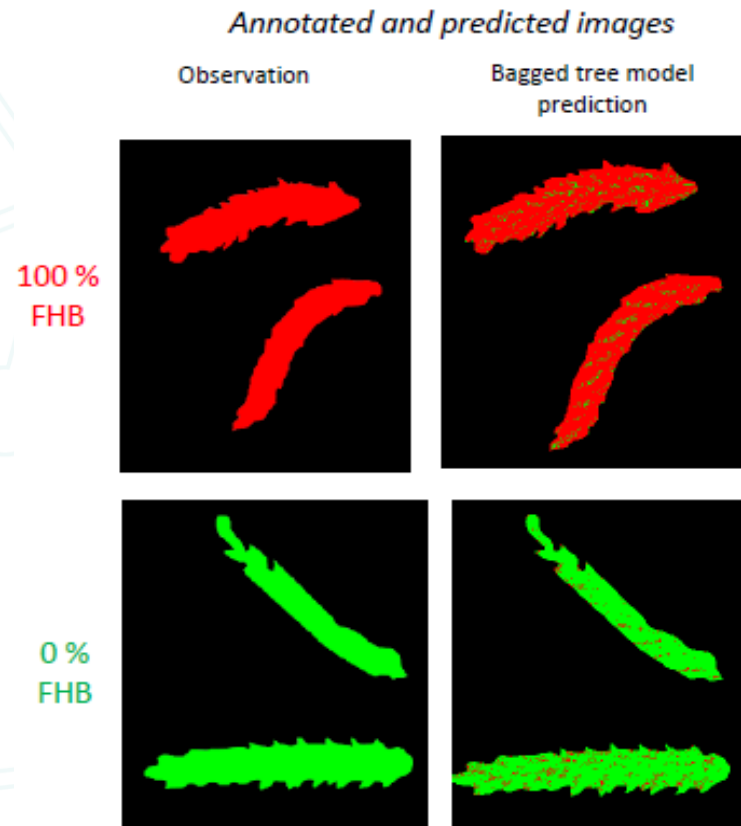
In field



3rd approach: VNIR multispectral imaging: results



In laboratory



H. Garbougé (2022). Deep learning applied to multi-component imagery for variety testing problems. Thesis in U.Angers



3rd approach: VNIR multispectral imaging: results



In field in frontal view

Annotations

Annotations of the 1st row of ears



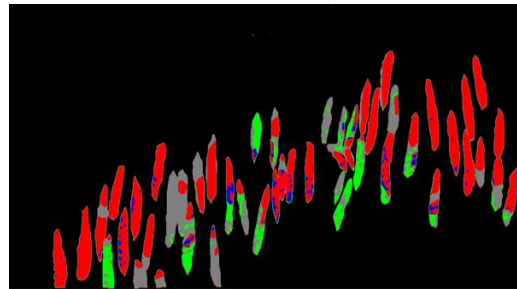
Creation of 2 algorithms :

Ear segmentation (U-NET)



H. Garbousse (2022).
Thesis in U.Angers

Fusarium quantification
prediction (Machine learning)

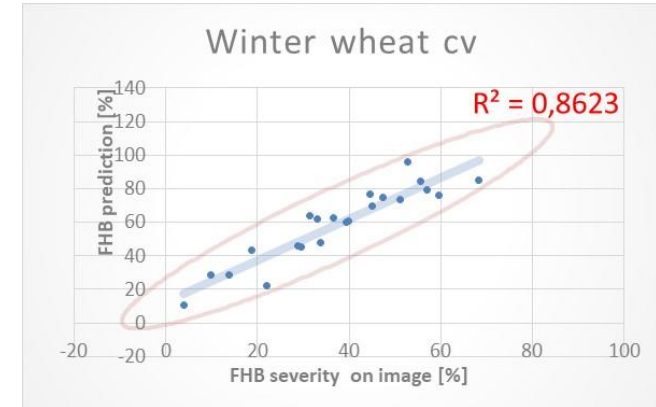


Fusarium annotations



Fusarium pixels

FHB prediction results



- Usable for assesment of FHB cv resistance
- 🤖 **High correlation with annotations:**
in case of good acquisition conditions & high FHB pressure.



VP='Fusarium' pixels predicted as 'Fusarium'

FP='Healthy' pixels predicted as 'Fusarium'

FN='Healthy' predicted 'Fusarium' pixels

FN='Healthy' pixels predicted as 'Healthy'

3rd approach: VNIR multispectral imaging: next steps

- Develop a semi-automatic annotation method to assess biotic stress



4th approach: VNIR hyperspectral imaging: acquisition



- 614 images acquired in laboratory by CRA-W
- 100 images acquired on 1 site in field by CRA-W

Hyperspectral imaging - line scan (Specim)
(400-1000, 1000-1700, 1000-2500 nm)

Based on the chlorophyll and water content
Green and at maturity ears

Proximal sensing 😊

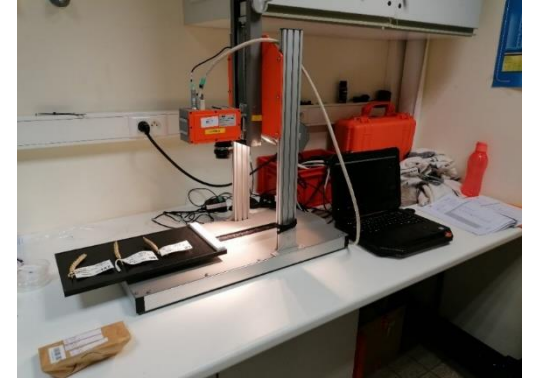
High cost 😞

Time consuming 😞

Image on ears acquired in vertical view (line scan camera) 😊

Possible detection on a wider area (100 ears or 1m²/plot) 😊

In laboratory



In field

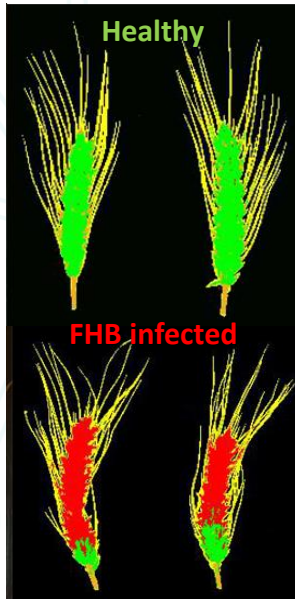


4th approach: VNIR hyperspectral imaging: results



In laboratory

MODEL VALIDATION (PLSDA)



NIR-HSI predicted images

Infected or not

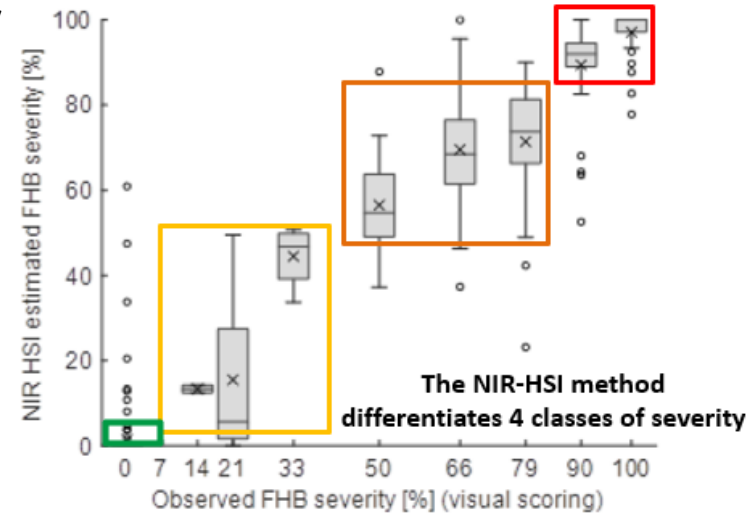
	Actual class	
	FHB-infected	Healthy
Predicted as FHB-infected	152	13
Predicted as Healthy	2	141

Sensitivity: 98,7%



Specificity: 91,6 %

Severity



D. Vincke et al. (2023). Near infrared hyperspectral imaging method to assess Fusarium Head Blight infection on winter wheat ears. *Microchemical Journal* 191, 108812

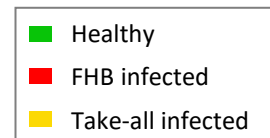
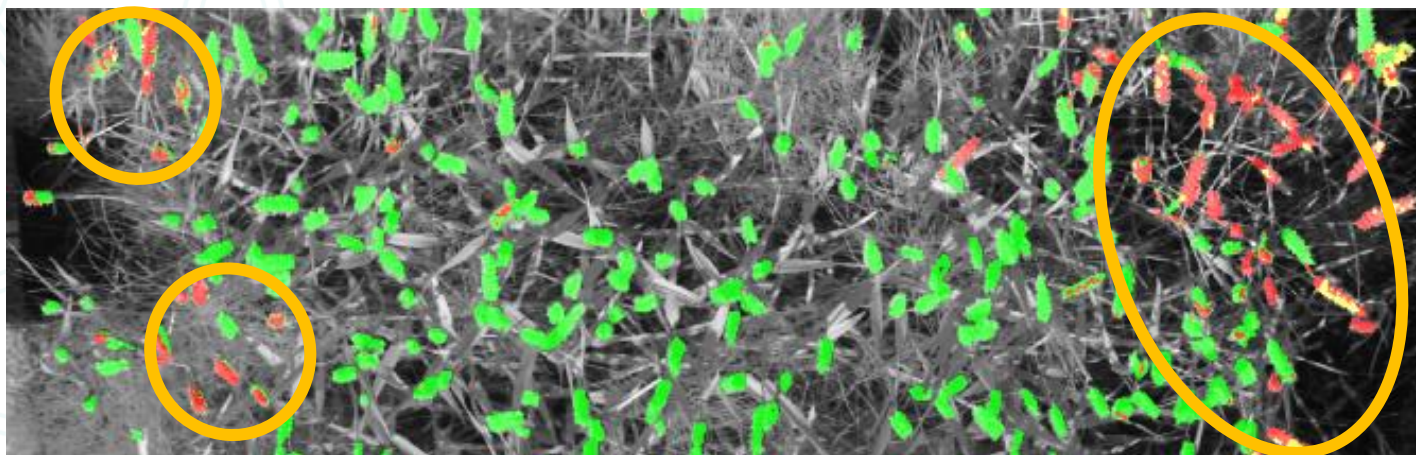


4th approach: VNIR hyperspectral imaging: results



In field

Predicted images



Date	FHB		Take-all		Overall stress	
	RMSE	R ²	RMSE	R ²	RMSE	R ²
28-06-22	0,7	0,08	4,1	0,39	2,0	0,85
04-07-22	0,6	0,26	7,8	0,56	4,2	0,87
14-07-22	0,5	0,42	5,4	0,79	1,8	0,98

D. Vincke (2024). Evaluation of fusarium head blight infection on winter wheat using near infrared hyperspectral imaging from the laboratory to the field. Thesis in U.Liège Gembloux Agro-Bio Tech

The method can assess the overall stress of the ears but ...
It is not specific enough to differentiate two diseases with similar symptoms.



4th approach: VNIR hyperspectral imaging: next steps

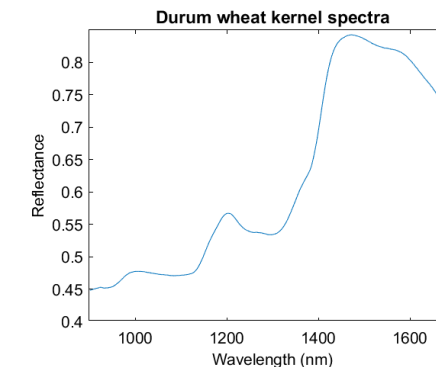
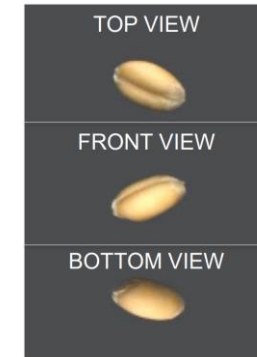
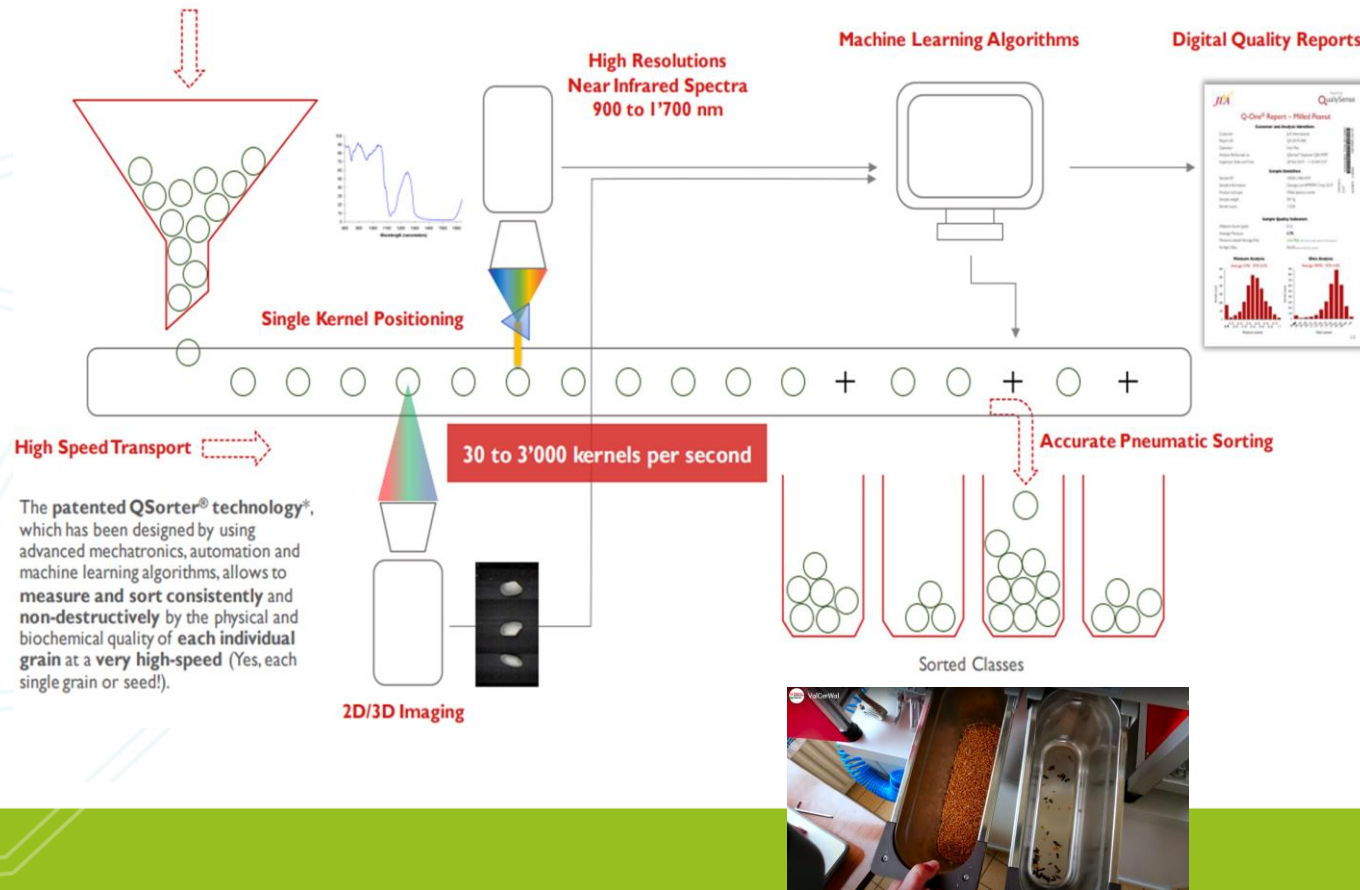
- Transfer to a tractor platform
& adaptation
with blackout box (natural light control)
with angle 45 or vertical view
- Model optimisation:
specificity
prediction in real time



5th approach: NIR + RGB imaging for sorting grains

➤ Single-kernel analyzer: Qsorter explorer (Qualisense)

- NIR range 900-1700 nm
- 3D RGB imaging
- Information acquired from each individual grain



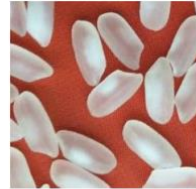
5th approach: NIR + RGB imaging for sorting grains

➤ High-speed pneumatic sorting system

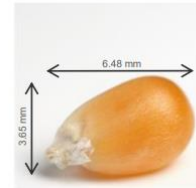
- Sorting based on **physical** and/or **biochemical** characteristics of grains
- Quality inspection in real time at **high throughput**
- In-depth batch **characterization** and better contaminant **management**
- **Individual grain** phenotyping for the development of new varieties



INSECT DAMAGE



BROKEN KERNELS



GEOMETRY



DISEASES
(FUSARIUM)



IDENTIFY IMPURITIES

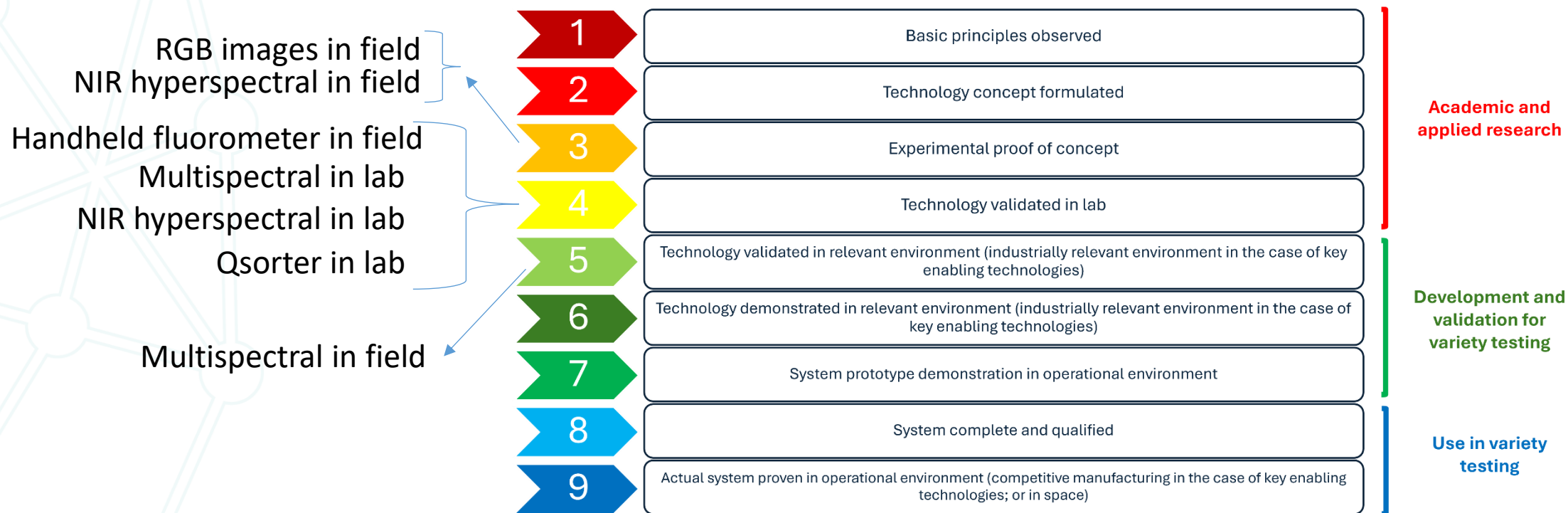


Image	Spectra	Protein [%]	Moisture [%]	Length [mm]	Width [mm]	Area [sqmm]	Elongation Factor [-]	Prediction Index	Label
		12.23	12.34	9.21	2.45	22.58	4.58	1.02	Oat
		13.65	11.21	8.76	2.76	24.21	4.48	1.12	Oat
		11.42	12.13	7.78	2.57	20.00	3.85	1.08	Oat
		15.35	11.05	7.25	3.97	28.78	3.97	0.92	Contaminant
		12.78	10.98	11.58	3.12	36.17	4.72	1.01	Oat

Various compositional properties:
protein, fatty acids, sugar...



Readiness of the tools



"Technology readiness levels (TRL); Extract from Part 19 - Commission Decision C(2014)4995" (PDF). ec.europa.eu.

Next steps

Extension to other diseases

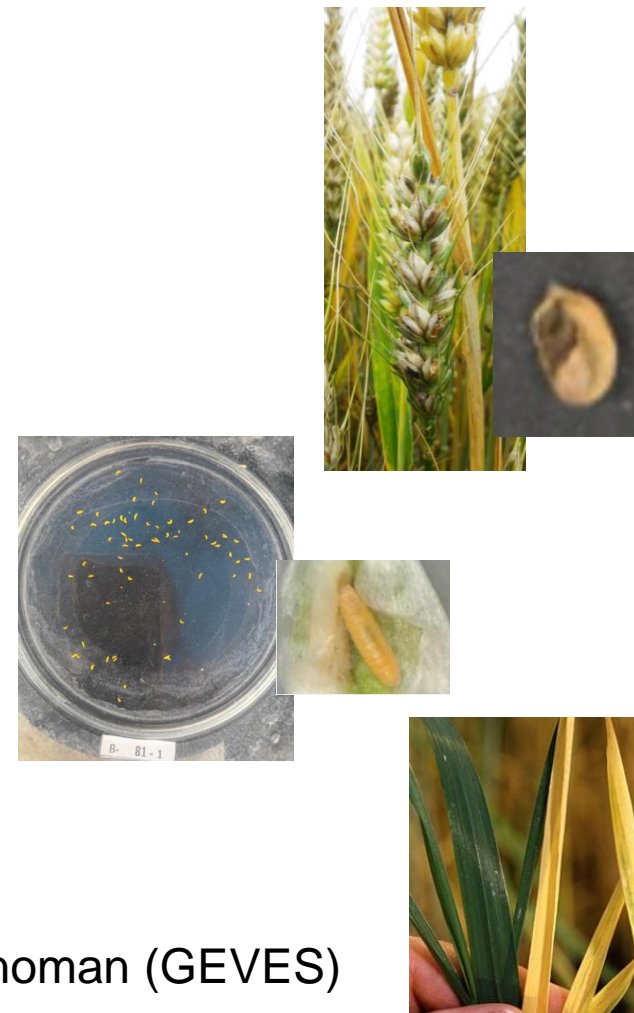
☐ Head diseases

Bunt (*Tilletia caries*): contamination of grain lots by spores
transfer methodology to other head disease (CRA-W)

Orange wheat blossom midge larvae – *Cecidomyia* – Yield loss
counting of larvae from RGB image (CRA-W, GEVES)

☐ Leaf diseases

Barley yellow dwarf virus (BYDV)- Vector: aphids - Yield loss
assess BYDV symptoms on leaves with the perch Phenoman (GEVES)



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François Laurens (INRAE)
David Rousseau (UA)
Joseph Peller (WUR)



Philippe Vermeulen
Damien Vincke
Corentin Demoitie
Damien Eylenbosch
Vincent Baeten



Andreas Hund (ETHz)

PHENWHEAT

Benoît Mercatoris (GxABT)



Joseph Peller
Dan Rustia

VALCERWAL

Bruno Godin (CRA-W)

Fus'eye

Valérie Cadot (GEVES)



David Rousseau
Hadhami Garbougé

