



Presentations

4th NJF Agromek EurAgEng joint Seminar

Advances and Innovations in Agricultural Engineering




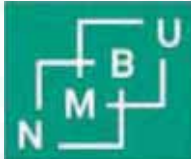

29-30 November 2022, Herning, Denmark



Nils Bjugstad
REALTEK, NMBU, Norway

NJF, EurAgEng, Agromek joint seminar. 29th to 30th of November 2022

Innovation	Sensors	Robotics	Precision agriculture	Smart farming	Environment
Design	Advances and Innovations in Agricultural Engineering The 4 th NJF - EurAgEng - Agromek Joint Seminar November 29 th - 30 th , 2022, Herning, Denmark				Climate neutral
Ergonomics					Circular economy
Soil compaction					Plant protection
Safety					Post harvest
Big data	Please visit www.nmbu.no/go/agromek for more information				Artificial intelligence
Data Science					Imaging
Drones					GIS Mapping
Navigation					SD Goals
Logistics	Advisory service	Management	Education	Networking	Horizon Europe



Keywords; Smart Farming, robots & drones, sensor technology, GIS/GNSS, sustainability, innovation

r ani in o ittee

Nils Bjugstad* (NO) - chairman. Faculty of Science and Technology, Norwegian University of Life Sciences, Ås

Claus Hermansen (DK). CEO for Agromek & Danish Agroindustry

Claus Sørensen (DK). Former president of the EurAgEng, Department of Engineering, Aarhus University

Alastair James Ward, (DK) Department of Engineering, Aarhus University

Antti Lajunen, (FI) University of Helsinki, Department of Agricultural Sciences, Helsinki

Sven Bernesson, (SE) Swedish University of Agricultural Sciences, Department of Energy and Technology, Uppsala

Sigtryggur V. Herbertsson, (IS) Icelandic Agricultural Advisory Centre

Vitalijs Osadcuks , (LV) Latvia University of Agriculture, Faculty of Engineering, Jelgava

Kęstutis Venslauskas, (LT) Vytautas Magnus University, Department of Mechanical, Energy and Biotechnology Engineering

Oliver Sada*, (EST) Institute of Technology, Estonian University of Life Sciences, Tartu

[Maibritt Kindberg, MCH Agromek](#)

Tuesday 29 November 2022

Bus to Agromek leaves 8.45 outside DGI Hotel, Herning. Please be there in advance.

People not showing up at 8.45 are expected to go to the Agromek conference venue on their own. Arrival to venue at Agromek, Herning expected to be around 9.00.

Venue: Sydsalen, Entrance Agromek West Registration from 09.00 to 11.10

Time	Topic	Presentations	Speakers
09.00 -11.10		Registration / networking/ visit Agromek on your own	
11.10 -11.30		Introduction and welcome	Nils Bjugstad, NMBU, N
		What is EurAgEng?	Claus Aage Sorensen, AU, DK
		What is NJF?	Nils Bjugstad, NMBU, N
		<i>Chair of session: Nils Bjugstad</i>	
11.30 – 12.00	Keynote	Agricultural robots for new sustainable crop productions	Hans Griepentrog, University Hohenheim, DE
12.00 - 12.20	Smart farming	Pilot project on Precision Farming	Michael Nørremark, AU, DK
12.20 - 12.40	Smart farming	Reducing the climate impact of Swedish agriculture - Field tractor electrification	Oscar Lagnelöv, SLU, S
12.40 – 13.00	Smart farming	A low-cost robot platform for swarm applications in agriculture	Robert Braunschweig, NMBU, N
13.00 – 14.00		LUNCH	
		<i>Chair of session: Alastair Ward</i>	
14.00-14.20	Smart farming	Smart Farming Sustainable arena established at NMBU	Nils Bjugstad, NMBU, N
14.20-14.40	Imaging	Impact of Sun Elevation Angle and Type of Sensor on Multispectral UAV Imagery Data	Sahameh Shafiee, NMBU, N
14.40-15.00	Imaging	Spectral imaging in crop fields- research activities at NMBU	Ingunn Burud, NMBU, N
15.00-15.20	Solar energy	Exploring solar energy combined with a fleet of electrical vehicles and precision agriculture for reduced GHG-emissions	El Houssein Chouaib Harik, NIBIO, N
15.20-15.40		Coffee & Tea break. incl. brief intro about AGROMEK	Claus Hermansen, AGROMEK, DK
15.40-16.00	Sensor technology	Sensor technology for optimal harvest in strawberry	Siv Fagertun Remberg, NMBU, N
16.00-16.20	Smart farming	Safety concept for robotic silage harvest system	Ari Ronkainen, LUKE, FIN
16.20-16.40	Smart farming	Data infrastructure requirements for Robotics and Smart Farming Research	Animari Hartikainen, LUKE, FIN
16.40-17.00	Smart farming	Experimental study of laser weed control approach with full canopy area treatment	Vitālijs Osadčuks, LBTU, LV
Ca. 17.30		Bus to Agromek dinner in Herning City (updated times during seminar)	
18.00-23.00		Agromek dinner	

Wednesday 30 November

Bus to Agromek leaves 08.45 outside DGI Hotel, Herning. Please be there in advance.

People not showing up at 08.45 are expected to go to the Agromek conference venue on their own.

Arrival to venue at Agromek, Herning expected to be around 9.00

Time	Topic	Presentations	Speakers
09.00-10.00		Short visit of Agromek on your own / Venue Sydsalen open & coffee	
		<i>Chair of session: Claus Sorensen</i>	
10.00-10.40	Keynote	Can agtech innovation dissolve goal contradictions?	Per Frankelius, LiU, SE Jonas Engström, RISE, SE
10.40-11.00	Sustainability	Measures for reductions of greenhouse gas emissions in agriculture	Sven Bernesson, SLU, S
11.00-11.20	Biogas energy	Ensiling of straw as a pre-treatment and storage method for biogas production	Alastair James Ward, AU, DK
11.20-11.40		Coffee break	
11.40-12.00	Energy efficiency	Energy efficient grain drying processes	Jens Moller Andersen, AU, DK
12.00-12.20	Soil compaction	Soil compaction from using agricultural robots in complex arable operations	Alvaro Calleja Huerta, AU, DK
12.20-12.40	Soil compaction	Prevention of subsoil compaction: Technologies and strategic planning	Mathieu Lamandé, NMBU/AU, DK
12.40-13.00	Soil tillage	Drawbar Pull Testing for Machine Soil Interaction Characterization	Ole Balling, AU, DK
13.00-13.10		Closing session	Nils Bjugstad, NMBU, N
13.10-14.00		LUNCH	
14.00-15.00		Visit Square meter farm and other involved partners	Claus Sorensen, AU, DK
15.00-16.15		Visit Agromek on your own	
16.15-17.00		How to successfully implement artificial intelligence in your business	Tomas Borovicka, CEO at Datamole ¹ , Days of future, hall E, see map
15.00-20.00	AGROMEK	Visit Agromek free of charge. Agromek closes 20.00	All

Frequently shuttle buses from Agromek fair to Billund airport.

e notes

Agricultural Robots For New Sustainable Crop Productions Systems with:



Hans Griepentrog

Professor PhD, Dipl.-Ing. university of Hohenheim,
Institute of Agricultural Engineering, Germany.

Can agtech innovation dissolve goal contradictions with:



Per Frankelius

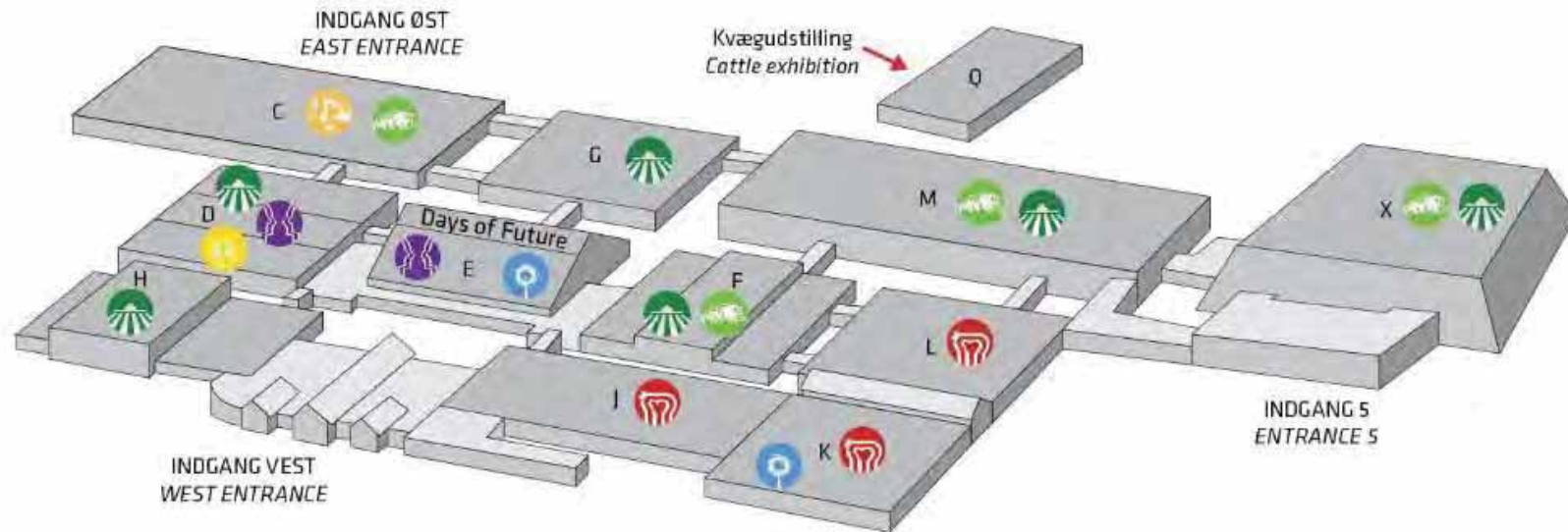
PhD, is Associate Professor (Docent) in Business
Administration at Linköping University (Sweden).



Jonas Engström

Researcher and senior project manager at RISE,
Research institutes of Sweden

Hall overview



traktorer og høstmaskiner
tractors and combine harvesters



kornhåndtering
crop management



energi
energy



viden og service
knowledge and service



markredskaber
agricultural machinery



staldmekanisering
livestock machinery



entreprenør, vej & park
construction, roads and parks





Prof. Claus Grøn Sørensen
Aarhus University, Denmark

Past President 2016-2018, EurAgEng
Current Executive, Council member

The European Society of Agricultural Engineers



The European Society of Agricultural Engineers (EurAgEng) exists to promote the profession of Agricultural and Biosystems Engineering and the people who serve it.

The Society is particularly active in Conferences, Working Groups, Publications, Networking and International lobbying.

EurAgEng is the European member of CIGR, the world wide agricultural engineering organization (International Commission of Agricultural and Biosystems Engineering)





- Founded in the 1980s. Network with over 2000 members of national societies from 23 countries.
- Biennial agricultural engineering conference (AgEng2018, Wageningen, AgEng2020, Evora, AgEng2022, Berlin, etc.)
- Partnering with German National Society, VDI-MEG, for the biennial Land.Technik-AgEng before Agritechnica in Hannover
- Partnering with CIGR for joint conferences (e.g. in Turin, 2026)



- “Biosystems Engineering” – the official scientific journal of EurAgEng
- Supports groups (Agricultural Engineering and Technology) and projects (Agriculture and Energy Efficiency; Smart-AKIS, etc).
- Standardization and harmonization of engineering curricula and student initiatives (Field Robot Event)
- Nine Fields of interest are identified. Eighteen Working Groups (WGs) within these fields of interest
- Institutional network: "European Network for Advanced Engineering in Agriculture and Environment" (ENGAGE)



Conferences



- *AgEng 2018, Wageningen, Netherlands*
- *Feb 2019: AXEMA-EurAgEng-SIMA Paris “Sustainable agriculture : An opportunity for innovation in machinery and systems”*
- *Land.Technik AgEng 2019, Hannover (Agritechnica)*
- *AgEng 2020, Evora, Portugal*
- *Land.Technik AgEng 2021, Hannover (Agritechnica)*
- *AgEng 2022, Germany*
- *Land.Technik AgEng 20223, Hannover (Agritechnica)*
- *AgEng 2024, Athens, Greece*
- *AgEng/CIGR conference, Turin, Italy, 2026*



Nordic-Baltic Association of Agricultural Scientists
Institutional EA E

Association of



- NJF is Non-governmental organization
- About 1,500 members
- Non-profit organization
- Nordic and Baltic countries
- 6 sections:
Plants, animals, environment, economics, reindeer husbandry and
TECHNOLOGY & INNOVATION



MAIN ACTIVITIES

- NJF congress, this autumn in Iceland (September 2022)
- Seminars & Workshops
- Active working group meetings
- Section board meetings
- National board meetings
- New president – Jarkko Niemi, LUKE, Finland





*In seminars working habits
differ according to needs*

*Key issues in NJF functions
are
interaction, networking &
meeting persons sharing the
same interests*



NJF MEMBERSHIP

- Want to join NJF and be a member?
- Costs about 30 € per year (some variation between countries)
- Join us now: visit <https://nordicagriculture.eu/>



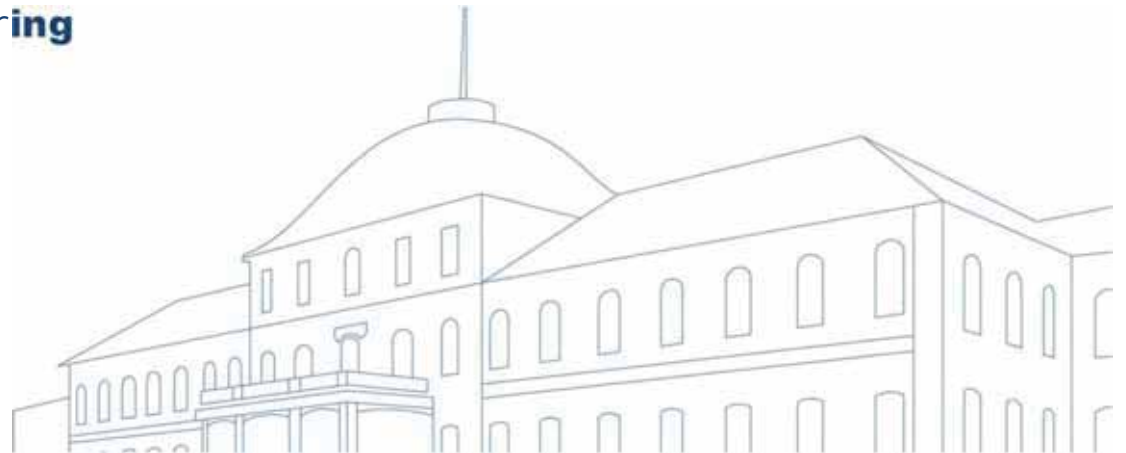
Agricultural Robots For New Sustainable Crop Production Systems

Advances and Innovations in Agricultural Engineering

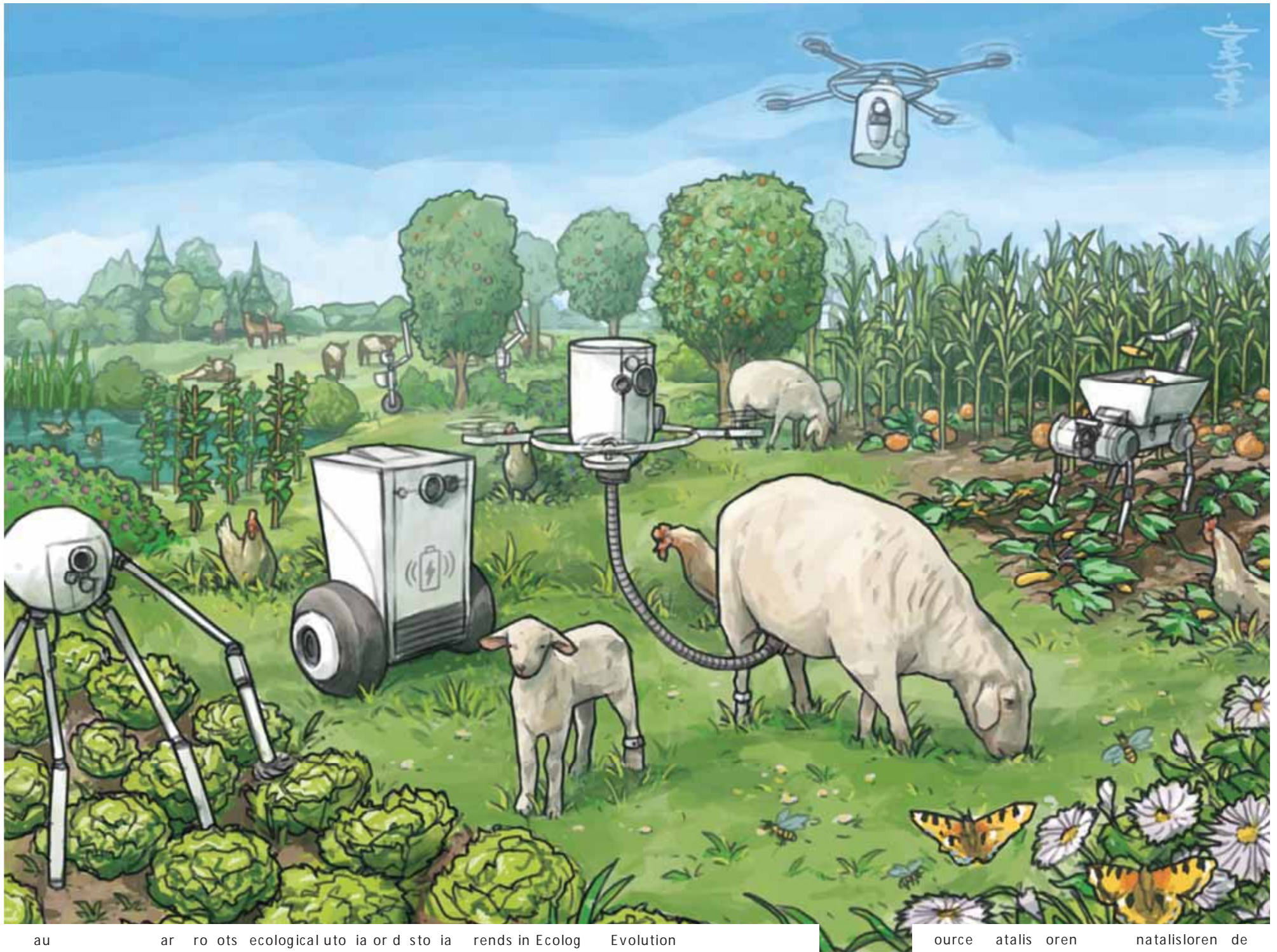
The 4th NJF - EurAgEng - Agromek Joint Seminar, 29.-30. November 2022

Hans W. Griepentrog
(Professor, PhD)

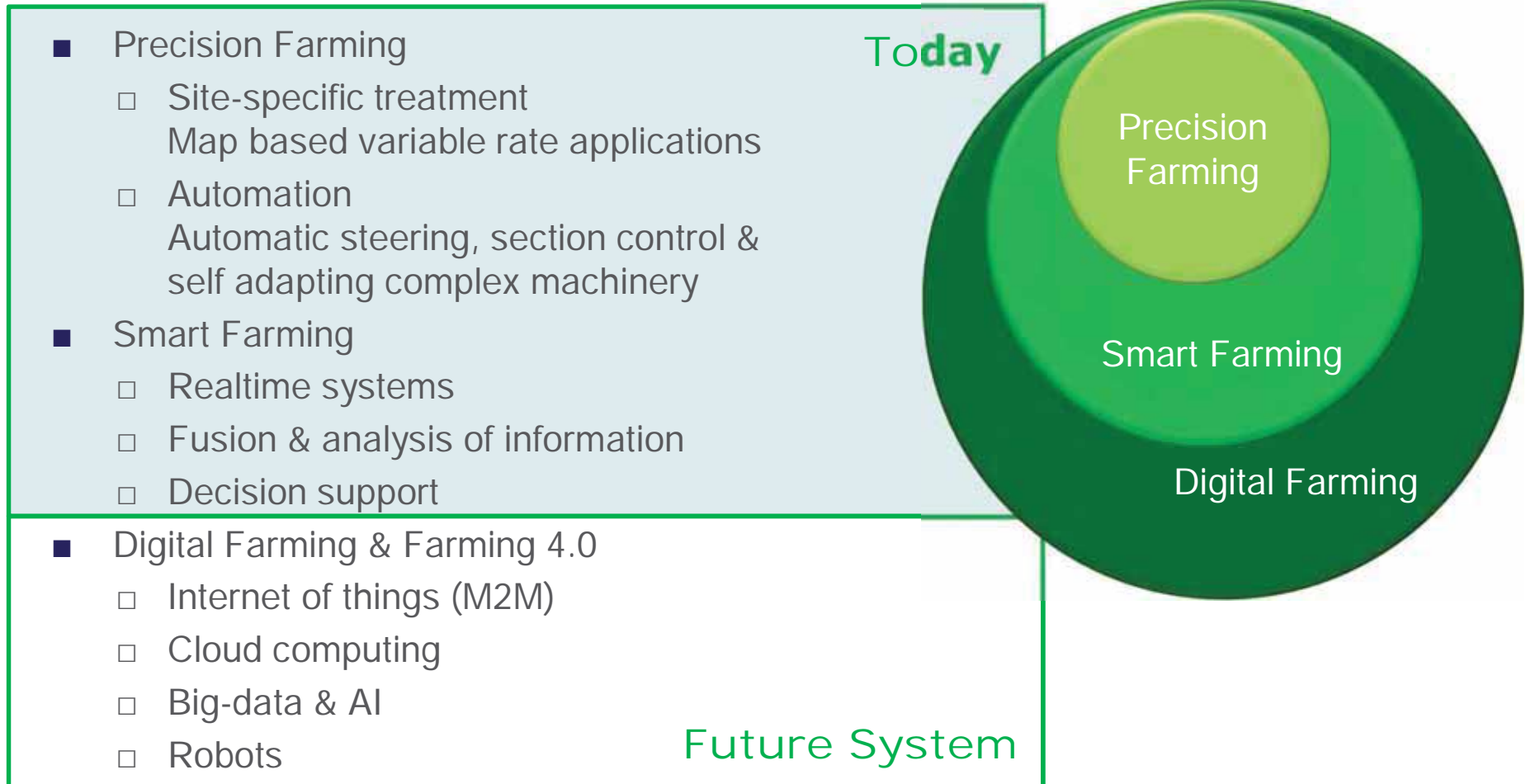
Institute of Agricultural Engineering
Technology in Crop Production
Germany





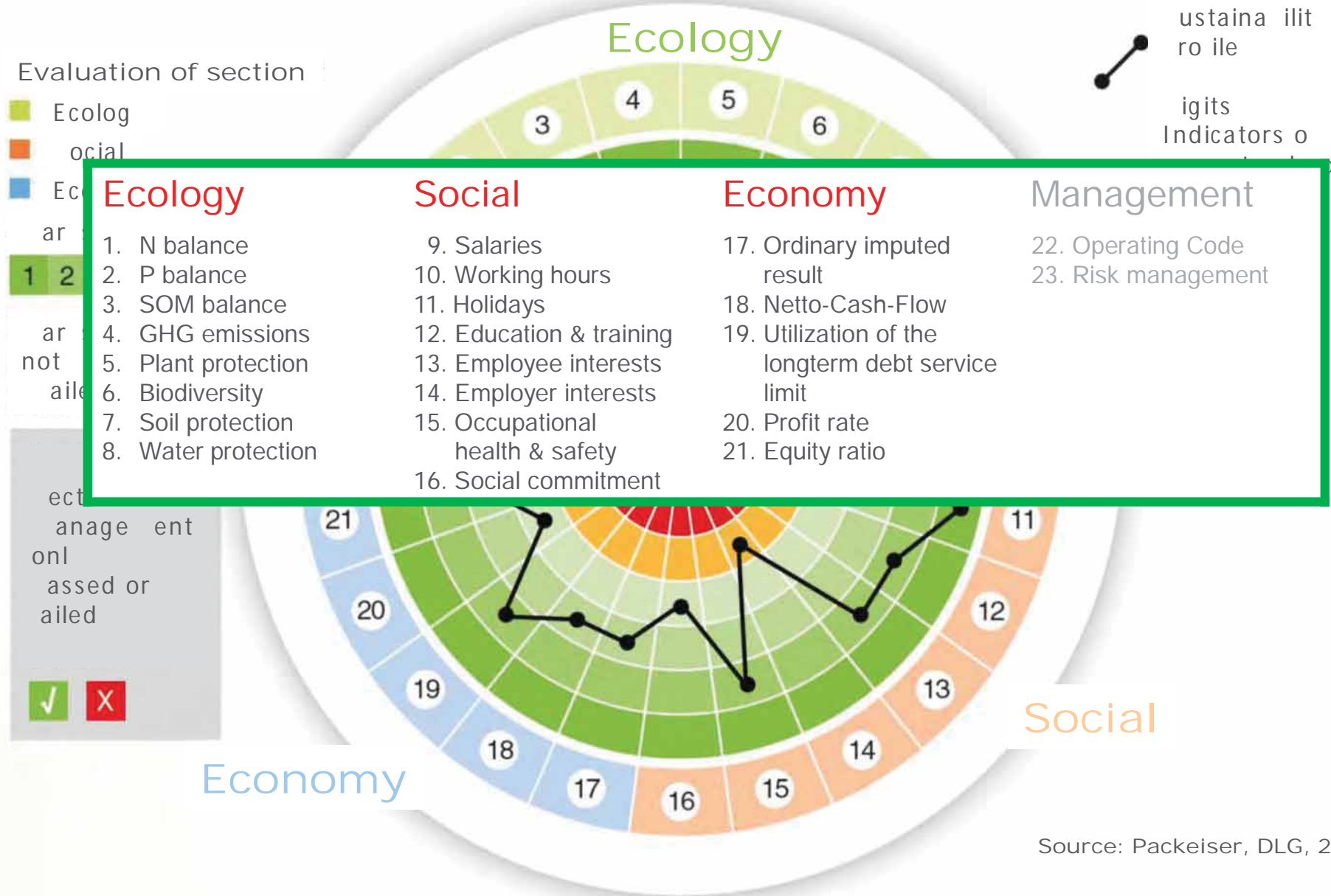


Definition of Terms





Certification of Sustainable Farming (1)





Certification of Sustainable Farming (2)

Evaluation of section

Ecolog

ocial

Econo

ar s range

1 2 3 4 5 6

ar s and
not assed
ailed

ection
anage ent
onl
assed or
ailed

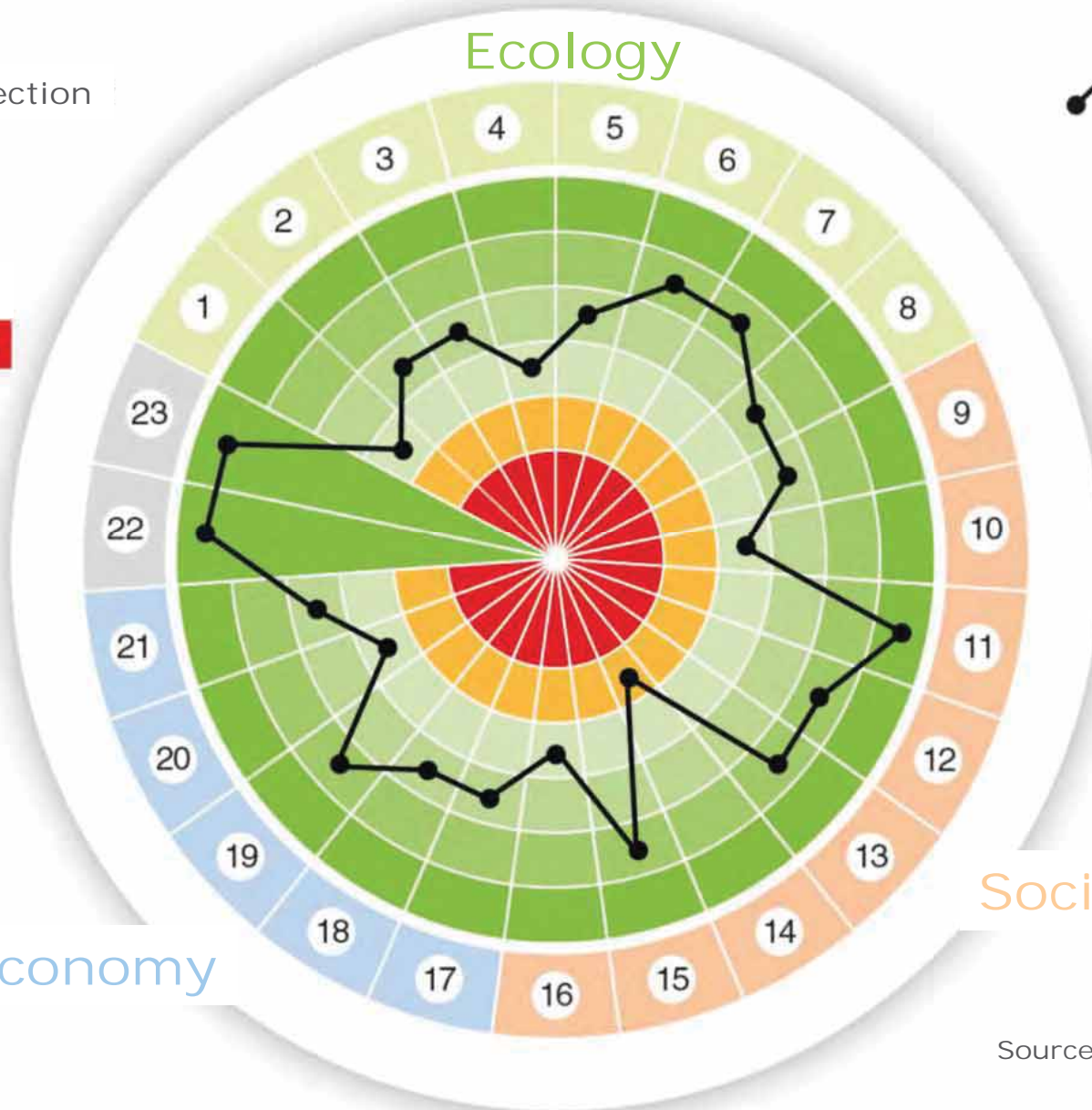


Economy

Ecology

Social

ustaina lit
ro ile
igits
Indicators o
tandard
ustaina le
ar ing





Certification of Sustainable Farming (3)

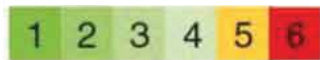
Evaluation of section

Ecology

Social

Economy

star s range



N balance - 1

SOM - 3

GHG emissions - 4

6 - Biodiversity

9 - Salaries

11 - Holidays

Indicators of sustainable farming

Data acquisition - Export from FMIS !

(Farm Management Information System - FMIS)

Economy

Social

Mechanization – Two Strategies (1)

1 - Robots for conventional implements & applications



Mechanization – Two Strategies (2)

2 - Robots for specialized implements & applications



Mechanization – Two Strategies (3) Overview



- Systems for conventional implements & applications
 - High power (100-500 kW)

New step of
mechanization

- ISOBUS data communication
- High capacity (ha/h)
- High costs / investments
- Crop management
 - Site specific (PF), auto steering, section control etc.



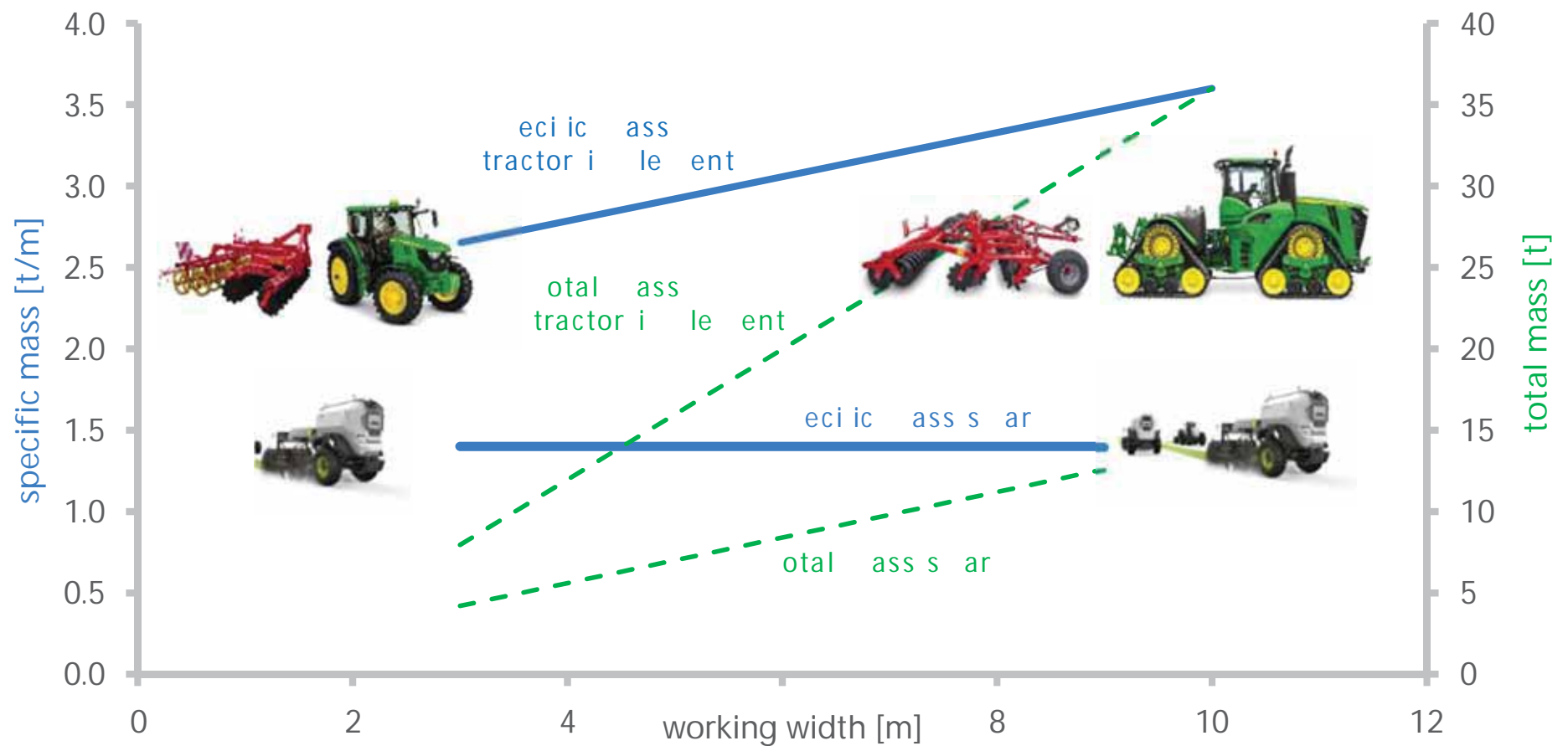
- Systems for specialized implements & applications
 - Low power (5-50 kW)

New step of
cropping system

- (scalable)
- Fits to all farm sizes
- Scalable costs
- Crop management
 - Individual plants (high spatial resolution)

Machine weights

Total and specific mass

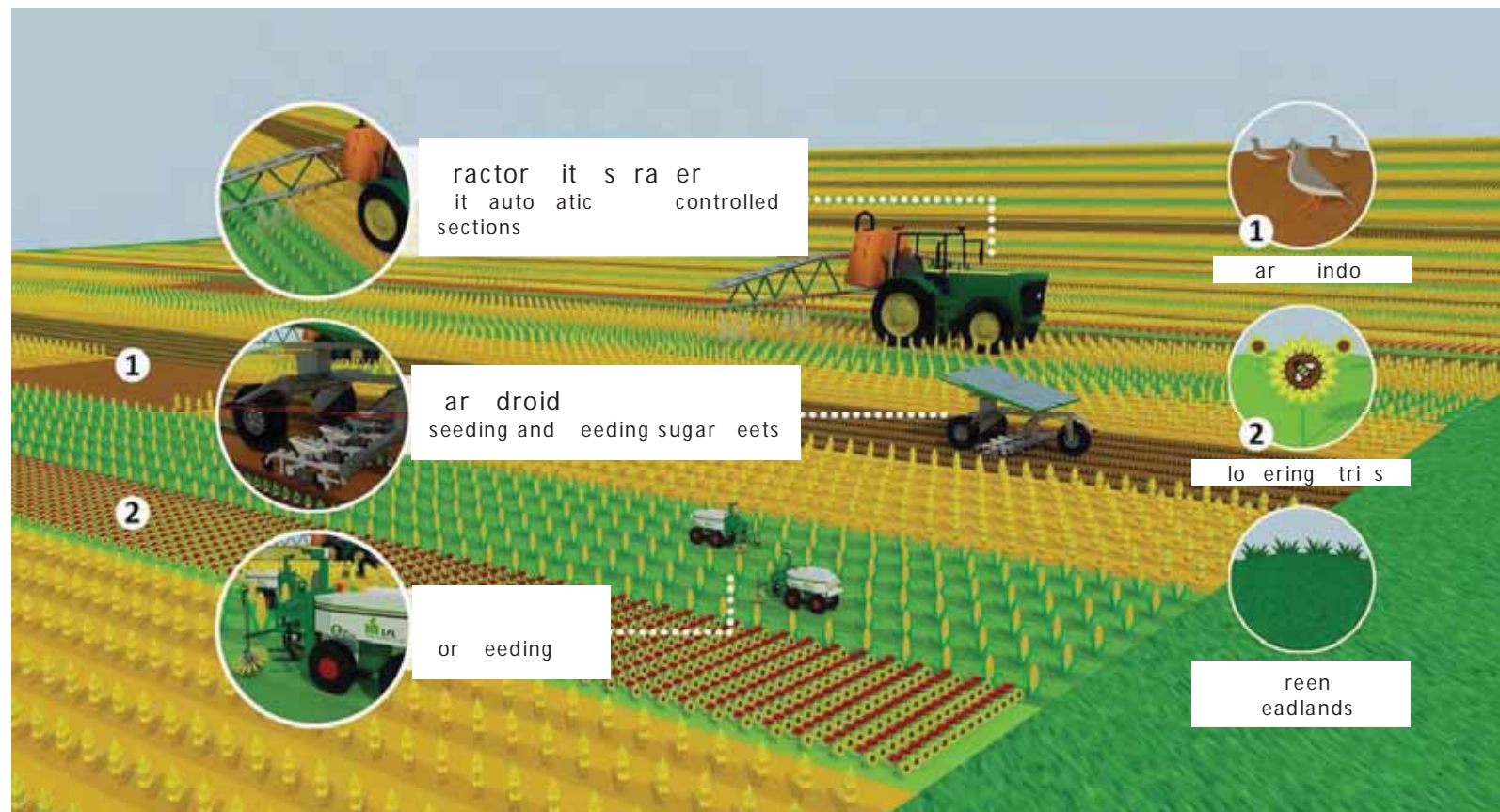


Source: T. Herlitzius, TU Dresden

Future Cropping Systems (1)

Biodivers – Soil Preventing – Digital

- Small scale & highly divers crop production system
- Climate resilient production system
- Digital and robotic solution for more opportunities
- Interdisciplinary and systems oriented



Source:
M. Gandorfer
Digital Farming Group
Bayerische
Landesanstalt für Landwirtschaft

Future Cropping Systems (2) Strip Intercropping



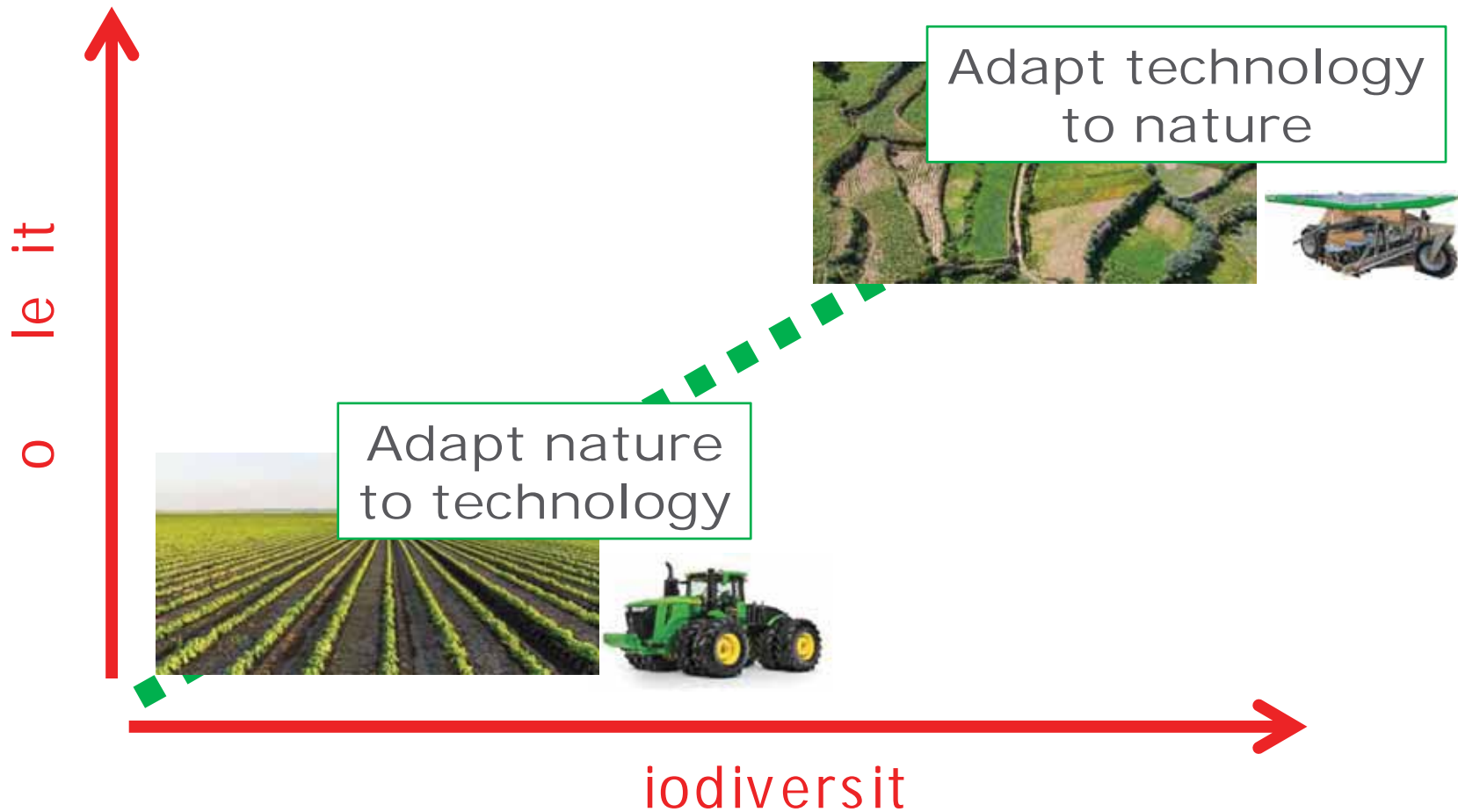
Source:
M. Gandorfer
Digital Farming Group
Bayerische
Landesanstalt für Landwirtschaft

Small Robots – New Mechanization

Slow speeds with more accurate performance

- High productivity via variable number of small machines (scalable)
- Small units highly flexible, no advantage of large cleared fields
- Higher resource efficiency (lower use of operating resources).
- Preservation or even reintroduction of landscape elements (biodiversity)
- Increased biodiversity
 - mixed crops, landscape features, contour cropping
- Selective treatments e.g. for weeding and harvesting

Paradigm Shift – Biological Intensification



Source: various data, AgEng

Outlook – Same mechanization for all? Conditions of Agricultural Production



Production for the world market; low intensities; low land prices; high automation of subsystems



Appropriate technology; medium and high intensities; low labor costs



High product qualities; high safety and environmental concerns; high labor costs, investments and intensities

Summary

- Sustainability needs more quantitative attention.
 - FMIS data and certifications can help farmers assessing their farms
- Robotics and digitalization allow new possibilities for crop production.
 - High working qualities by driving slowly
 - Soil protection through light vehicles.
 - Increased precision farming effect through high spatial resolutions and individual crop targeting.
 - No more advantages of 'big cleared out fields'.
 - Reintroduction of landscape elements and agricultural biotopes.
 - Machines are scalable in their function, thus flexible and adaptable.
- Paradigm shift: technology adapts to nature, not vice versa.

Thanks for your attention!

PILOT PROJECT ON PRECISION FARMING

ENVIRONMENTAL IMPACT CALCULATIONS OF PRECISION FERTILIZATION

By:

Michael Nørremark

Aarhus University

Department of Electrical and Computer Engineering

With contributions from:

Jacob Glerup Gyldengren, Iris Vogeler Cronin, Ingrid K. Thomsen og Peter Sørensen

Aarhus University

Department of Agroecology



AARHUS
UNIVERSITET
INSTITUT FOR ELEKTRO OG COMPUTERTEKNOLOGI

29. NOVEMBER 2022

MICHAEL NØRREMARK
SENIOR RESEARCHER



BACKGROUND

Assumption that technology and principles for precision farming have a reduction potential for leaching of nitrogen (N) in the order of 3-4 kg N/ha on arable farms and 5-6 kg N/ha on livestock farms.

Agricultural and environmental purposes: 8 -10 ha of precision fertilization for 1 ha of catch crop*

The pilot project investigated whether the potential environmental effect is the sum of a total of three sub-elements:

1. Detailed calculation of N requirements,
2. Site-specific application of N and the application of spreading equipment with section control (DAISY and APSIM based model calculations), and
3. Determination of N in livestock manure and precise prescribed amount (NLESS5 based model calculations).

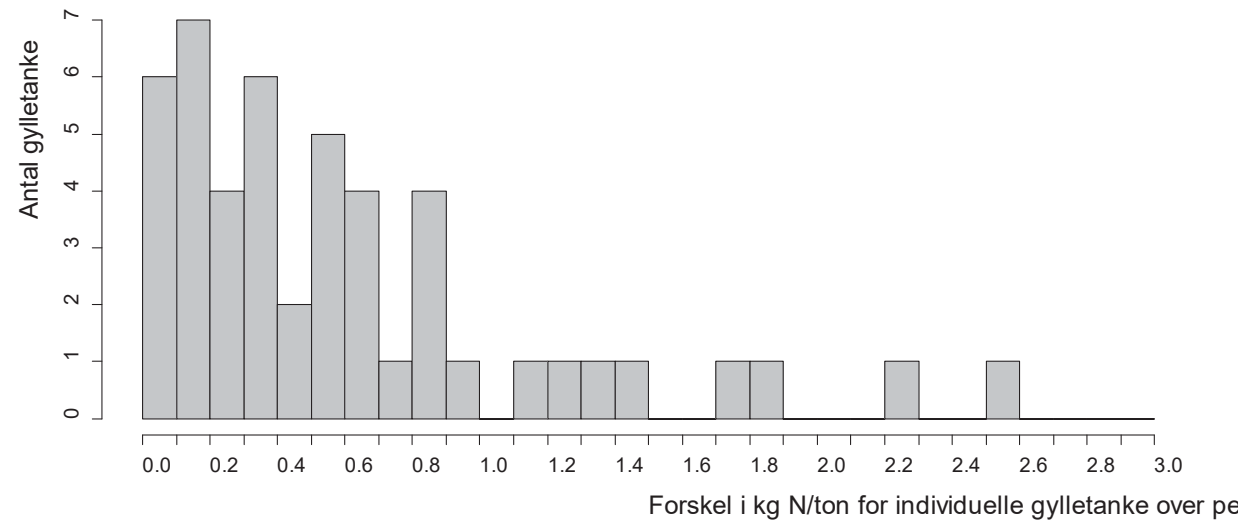
The sub-elements are delimited to only include a typical crop rotation with and without livestock manure.

As a starting point, the Danish Agricultural Agency wanted to determine conversion factors for cultivated fields w/o livestock manure fertilization.

LIVESTOCK MANURE ANALYZES- STATISTICS

Pig manure total N analysis results (data from 15 pilot project farms, 89 slurry tanks)

Kg N/ton	2018	2019	2020
Gns.:	3.56	3.44	3.41
Std. a fv.:	1.25	1.14	1.21
Minimum	1.22	1.72	1.24
Maksimum	7.00	7.64	7.44
CV in %, farms	26	23	21

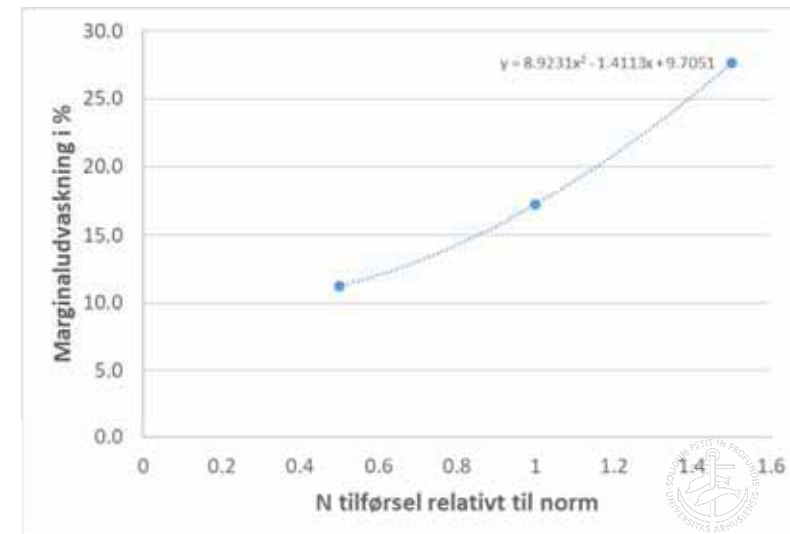


LIVESTOCK MANURE ANALYSES- ENV. IMPACT

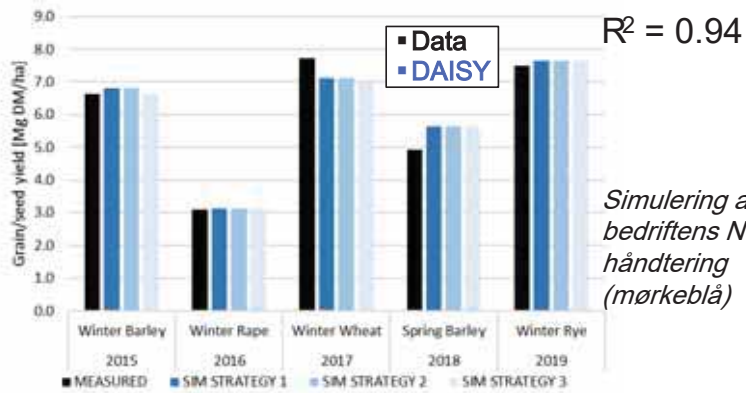
The following assumptions have been used to calculate the impact on nitrate leaching of redistribution of manure on the basis of livestock manure analyses from pilot project participants in 2018 (calculations are shown in brackets). In the calculations, it is also assumed that 50% of the manure is below the norm and redistributed to the norm, and that 50% of the manure is above the norm and redistributed to the norm.

		Assumptions and impact from redistribution of N in manure
a	CV for total N in svinøgylle [%]	26
b	Total N-application as manure [kg N/ha]	170
c	N utilisation requirements, av. for cattle- and pig manure [%]	77,5
d	Redistribution [N/ha] gns., $(a \cdot b / 100 \cdot c / 100)$, 1:1 area ratio	34,3
e	Average N-norm [kg N/ha] (all crop species) (Blicher-Mathiesen et al., 2020)	167
f	Redistribution in % of norm N $(d \cdot 100 / e)$ [%]	21
g	Marginal leaching by fertilization f% below norm [%] (cf. figure)	14,2
h	Marginal leaching by fertilization f% above norm [%] (cf. figure)	21,1
i	Average leaching for redistributed N, where 50 % of manure <u>below</u> norm are transferred to norm $(17,2 + g / 2)$ [%]	15,7
j	Average leaching for redistributed N, where 50 % of manure <u>above</u> norm are transferred to norm $(17,2 + h / 2)$ [%]	19,1
k	Altered leaching, where 50 % of manure <u>below</u> norm are transferred to norm $(d \cdot i / 100)$ [kg N/ ½ ha]	5,38
l	Altered leaching, where 50 % of manure <u>above</u> norm are transferred to norm $(d \cdot j / 100)$ [kg N/ ½ ha]	-6,55
	Environmental impact, reduction in N leaching (k+l) [kg N/ha]	1,20

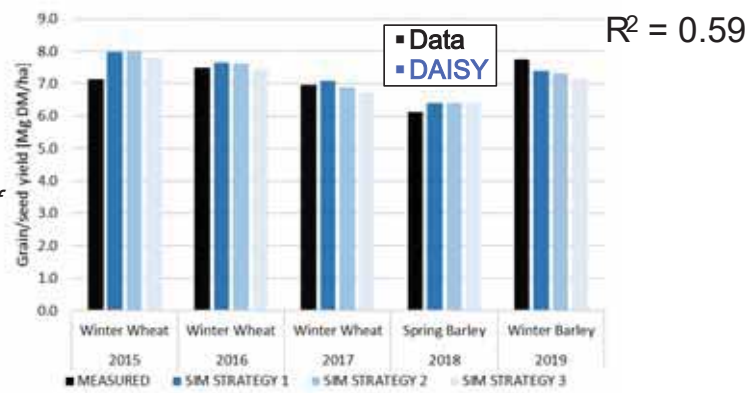
Marginal leaching in relation to N application estimated on the basis of 54 N response experiments used in calibrating the NLES5 model (Børgesen et al., 2020). The figure shows the equation that describes the average marginal leaching in the experiments relative to the N norm for a crop. The trials included a number of different crops, of which 41 were trials with cereals and winter crops



MODELING OF THE SOIL-PLANT-ATMOSPHERE SYSTEM DAISY AND APSIM



JB3 (field 31-0)



JB5 (field 37-0)



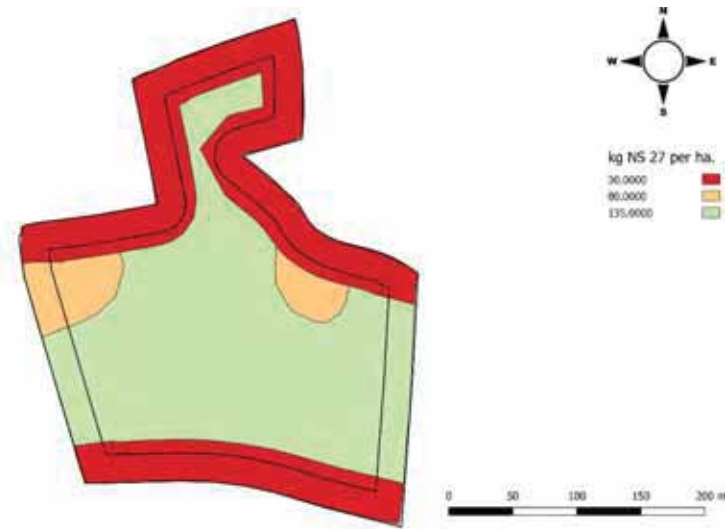
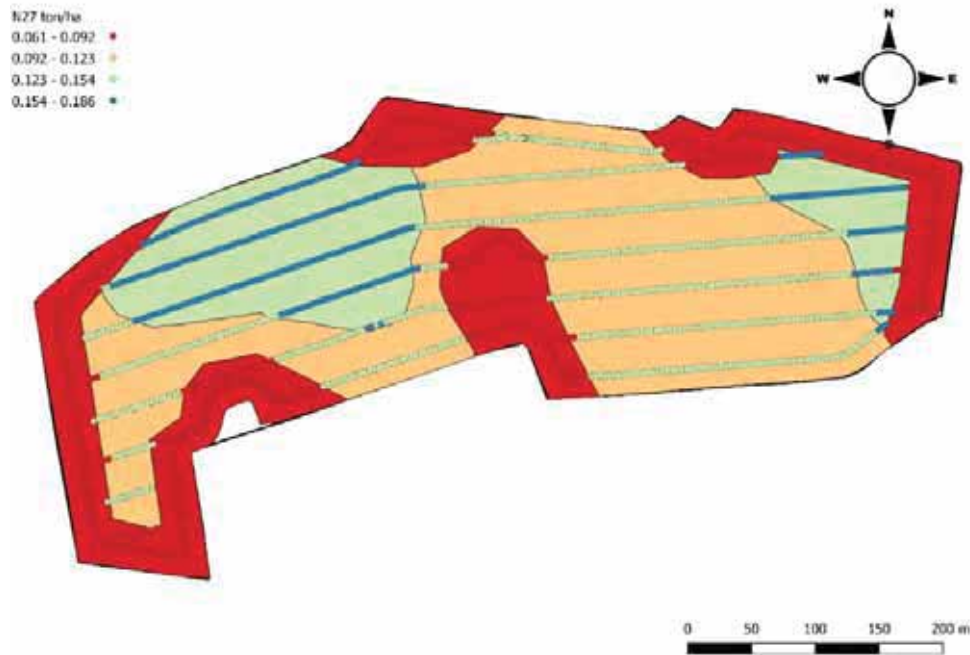
Parameters for every growing season:

Crop	Irrigation
Tillage type	Straw removal
Seed rate	Harvest date
Seeding data	Stubble cultivation
Inorganic fertilizer, 1. Date of application	Validated crop yield
Inorganic fertilizer, 2. Date of application	Farmers knowledge about field variations
Inorganic fertilizer, 3. Date of application	Soil type
Organic fertilizer, 1. Date of application	Texture
N in livestock manure	A-horizont
Utilization rate in %	B-horizont
Organic fertilizer, 2. Date of application	C-horizont
N in livestock manure	Bulk density, soil
Utilization rate in %	Weather data



Model configuration require many assumptions

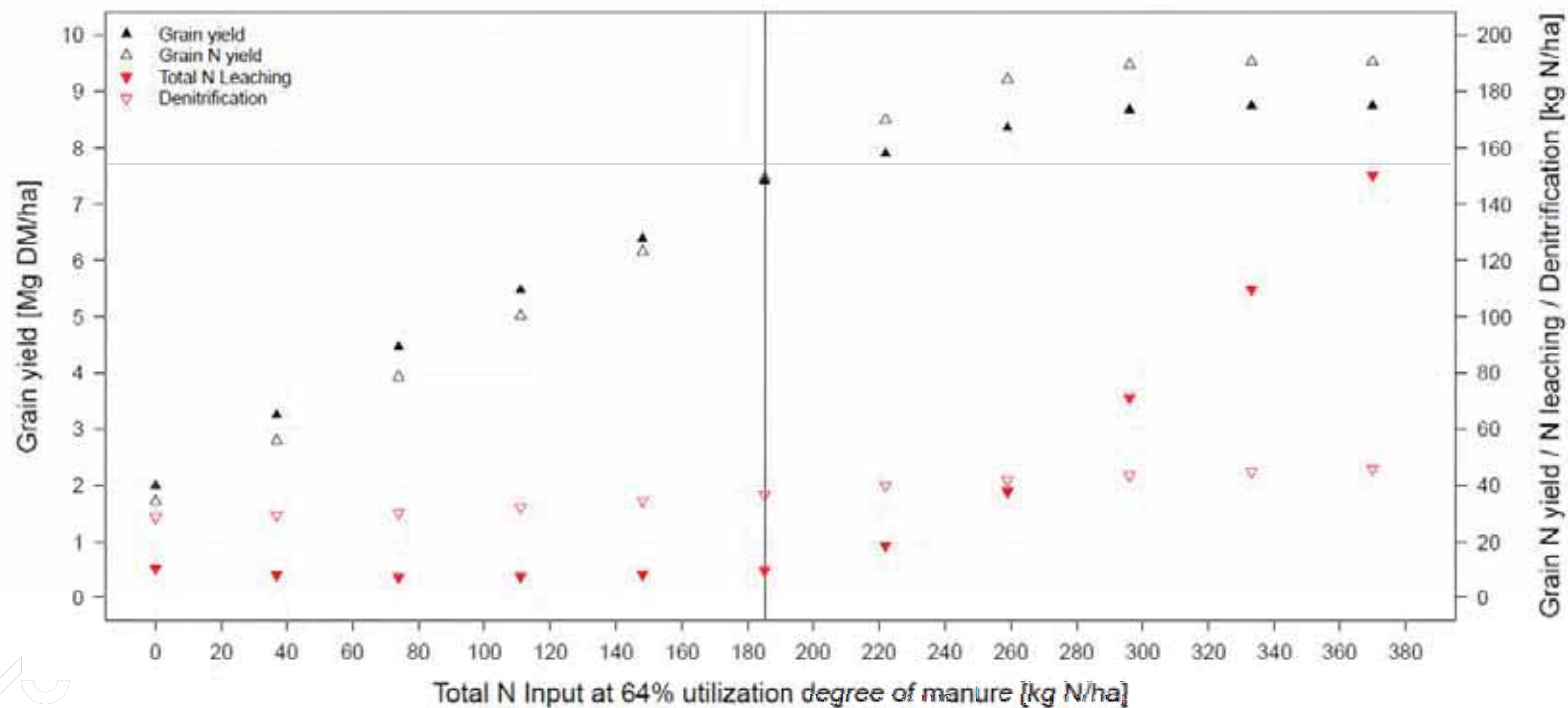
VARIABLE RATE APPLICATIONTRENDS



MODELING OF THE SOIL-PLANT-ATMOSPHERE SYSTEM DAISY AND APSIM

JB5_INFIELD

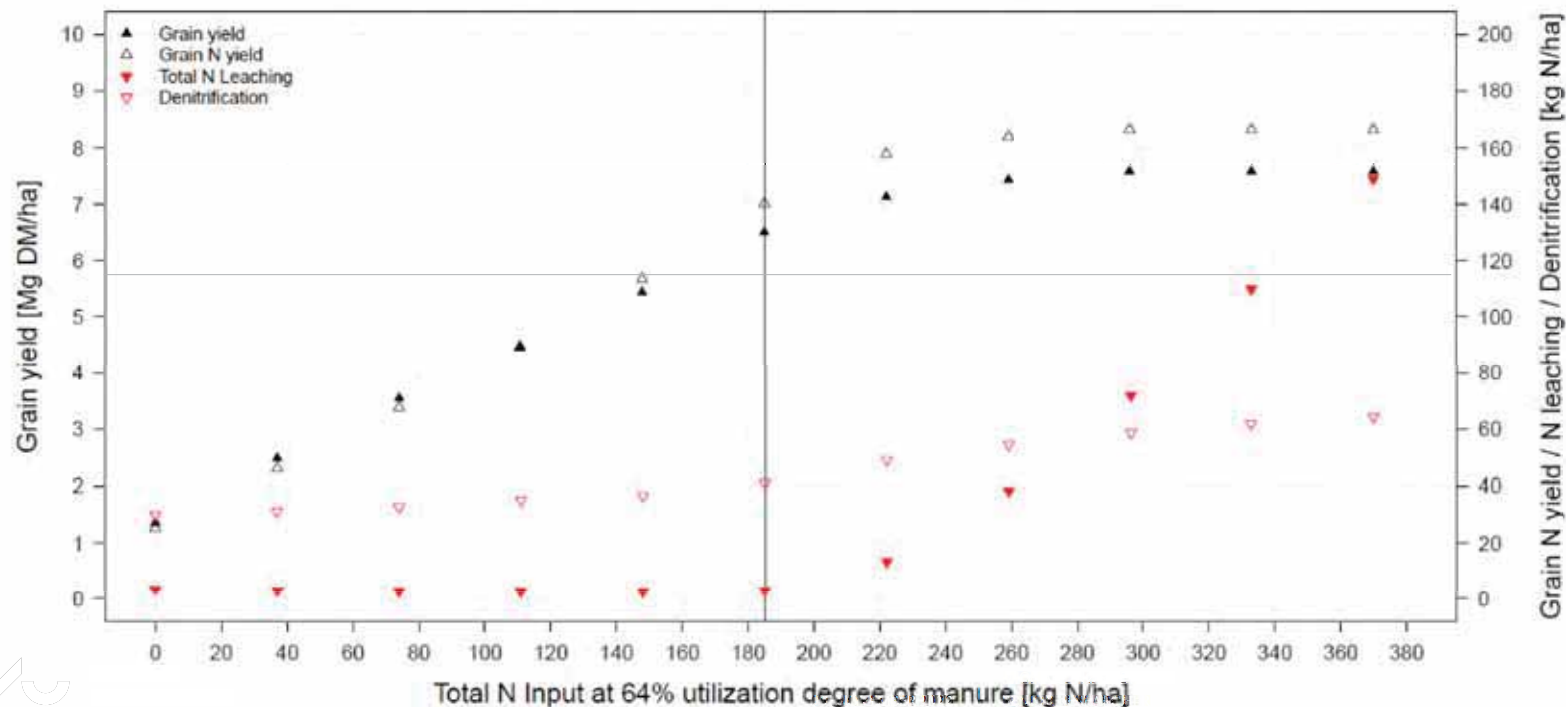
Grain yield, Grain N yield and accumulated N leaching and denitrification (September '17 – April '18)



MODELING OF THE SOIL-PLANT-ATMOSPHERE SYSTEM DAISY AND APSIM

JB5_HEADLAND

Grain yield, Grain N yield and accumulated N leaching and denitrification (September '17 – April '18)



SITESPECIFICAPPLICATIONOF N-ENV. IMPACT

Based on pilot project participants' reportings (organized by the Danish Agricultural Agency):

- Most widespread crop rotation
- Soil types
- Typical variations in site-specific N supply
- Typical partial field areas with site-specific application of N
- Regional weather data (East/ West Denmark)

Particular emphasis has been placed on headland areas, which generally have reduced yield potential, as well as depressions, where both the N dynamics and hydraulic conditions in soil differ from the flat areas.

Type 1 field

Subfield type	Area [ha]	Reference N-application	Strategy A (same total N amount)	Strategy B (reduced total N amount)
Headland	1	Norm N	- 30	- 30
Headland depression	0,5		- 30	30
Headland hilltop	0,5		- 30	- 30
Main field	14		+ 5	Norm N
Main field depression	3		- 5	- 30
Main field hilltop	2		+ 5	Norm N
Total area	21			
Total headland area	2 (9,5 %)			

Soil type	N-application for reference scenario [kg N/ha]		N-application compared to reference [kg N/ha]			
	Norm		Strategy A		Strategy B	
	1. year winter wheat (after winter rapeseed)	2. year winter wheat	1. year winter wheat	2. year winter wheat	1. year winter wheat	2. year winter wheat
JB1 w. irrigation	183	206	0	0	-13	-15
JB2 w. irrigation	183	206	0	0	-13	-15
JB3	156	179	0	0	-11	-13
JB4	162	185	0	0	-12	-13
JB5	189	212	0	+1	-14	-15
JB6	189	212	0	+1	-14	-15

Type 2 field

Subfield type	Area [ha]	Reference N-application	Strategy A (same total N amount)	Strategy B (reduced total N amount)
Headland	3	Norm N	- 30	- 30
Headland depression	0,5		- 30	- 30
Headland hilltop	0,5		- 30	- 30
Main field	50		+ 5	Norm N
Main field depression	4		-20	- 30
Main field hilltop	3		-20	Norm N
Total area	61			
Total headland area	4 (6,6 %)			

SITESPECIFIC APPLICATION OF NENV. IMPACT

All values in the table are relative to uniform distribution according to the N norm. Rolling crop rotation over the years 2015 to 2018. The table shows the difference between uniform distribution of norm N and strategies A and B respectively.

The N leaching results apply to the first leaching period after harvesting 1st year winter wheat. As the crop rotation was offset by one year in each of the rolling crop rotation scenarios, 2016 and 2017 are represented twice, once with 1st year winter wheat and once with 2nd year winter wheat.

	N leaching (strategy A)				N leaching (strategy B)			
	Type 1 field Min. Max.		Type 2 field Min. Max.		Type 1 field Min. Max.		Type 2 field Min. Max.	
JB1 w. irrigation ^a	-0,5	0,6	-0,3	0,6	-3,4	-1,7	-1,8	-0,9
JB2 w. irrigation ^a	-0,3	0,5	-0,3	0,5	-3,6	-1,7	-1,9	-0,9
JB3 ^a	-0,1	0,4	-0,1	0,6	-2,3	-1,2	-1,4	-0,6
JB4 (sand) ^a	-1,2	0,1	-1,3	0,3	-2,8	-1,3	-1,5	-0,7
JB4 (clay) ^b	-1,0	-0,1	-1,0	-0,1	-4,9	-0,5	-2,7	-0,3
JB5 ^b	-0,4	0,1	-0,2	0,2	-5,9	-0,6	-3,2	-0,3
JB6 ^b	-0,3	0,3	-0,2	1,0	-5,7	-0,6	-3,1	-0,3

^a Jynde vad weather data

^b Flakkebjerg weather data

Conclusions, summarized from notes, data from pilot project participants, latest catalog of N measures and the pilot project simulation results (report not yet published):

- Precision fertilization combined with site-specific application of N based on determination of the nutritional status of crops. N application based on either field variation data in relation to the soil and/or via sensors, in order to determine the actual fertilizer requirement.

[~1 kg N/ha]

- Precise fertilization, where the application equipment ensures that the fertilizer is applied by means of auto and section control, minimizing overlap when fertilizing

[~1 kg N/ha]

- It makes sense to analyze the long-term env. impact after the introduction of precision fertilization of N (i.e. perennial accumulation of N leaching after the introduction of precision fertilization)

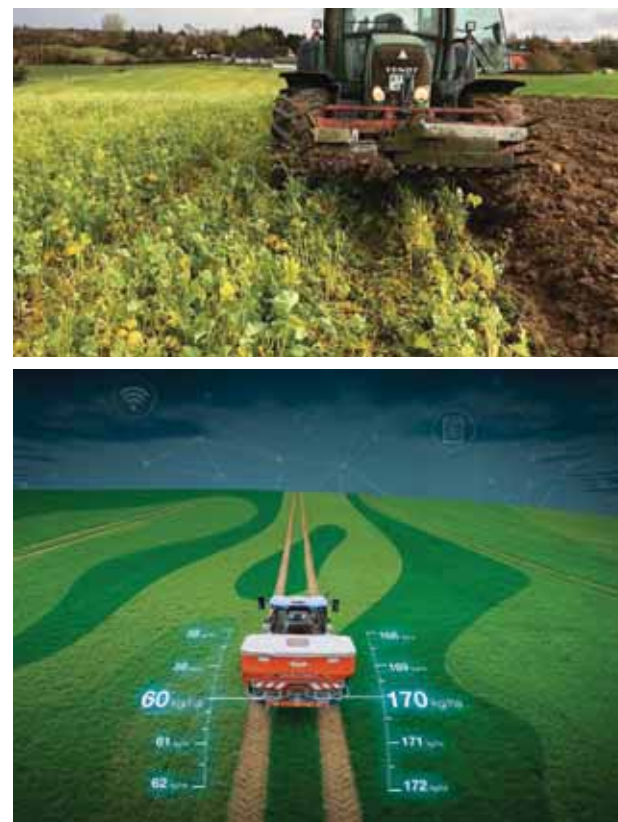
SUMMARY AND RECOMMENDATION FOR DECISION TO THE DANISH AGRICULTURAL AGENCY

	Below 80 kg N/ha in organic fertilizer	Above 80 kg N/ha in organic fertilizer
Env. impact of catch crops (kg N/ha) (Hansen et al., 2020) <i>(simple unweighted averages of effect on clay and sandy soils)</i>	22	35
Env. Impact of precision fertilization (kg N/ha)	2	3,2
Conversion factor (rounded off)	11:1	11:1

The conversion factors, with the prerequisites and assumptions used in the project in rounded off values, becomes 11:1 both below and above 80 kg N/ in livestock manure. This means that the use of precision farming on 11 ha will be able to replace 1 ha of catch crops

Now implemented in Danish regulation:

- 'Plantedækkebekendtgørelsen', BEK nr 742 af 30/05/2022, The Ministry of Food, Agriculture and Fisheries of Denmark



Thank you for your attention





SCIENCE AND
EDUCATION **FOR**
SUSTAINABLE
LIFE

Reducing the climate impact of Swedish agriculture - Field tractor electrification

Oscar Lagnelöv, PhD candidate
SLU - Swedish University of Agricultural Sciences
Department of Energy and Technology

Background

- Agricultural machinery is ~1% of global GHG emissions ^[1]
- EU: Climate neutral 2050
 - Sweden: Fossil free vehicle fleet 2030
- Agriculture must feed more people, on less land, and be sustainable & profitable



[1] Tubiello *et al.*, (2015) The Contribution of Agriculture, Forestry and other Land Use activities to Global Warming

Mistra food futures

- “...enable transformation of the Swedish food system into one that is sustainable (environmental, economic and social), resilient and delivers healthy diets.”
- WP5 - Agricultural systems with net-zero impact of greenhouse gas emissions (in 2045)
- Broad focus – GHG reduction and mitigation potential, implementation rates, system designs, societal changes.
- Broad range of subjects – Animal husbandry, energy production, ley, soil enrichment, fertilizer strategies, machinery...



Reports



Scientific publications

Röös, E., Wood, A., Säll, S., Abou Hatab, A., Ahlgren, S., Hallström, E., Tidåker, P. & Hansson, H. (2023). Diagnostic, regenerative or fossil-free – exploring stakeholder perceptions of Swedish food system sustainability. *Ecological Economics*. Accepted for publication.

von Greyerz K, Tidåker P, Karlsson J, Röös E. (2023). A large share of climate impacts of beef and dairy can be attributed to ecosystem services other than food production. *Journal of Environmental Management* 325:116400

Adamie, B., Uehleke, R., Hansson, H., Musshof, O., Hüttel, S. (2022). Dairy cow welfare measures: can production economic data help? *Journal of Sustainable Production and Consumption*, 32, 296-305.

Guo, A., Bryngelsson, S., Strid, A., Bianchi, M., Winkvist, A., Hallström, E. (2022). Choice of health metrics for combined health and environmental assessments of foods and diets: a systematic review of methods. *Journal of Cleaner Production*, 365, 132622.

Macura, B., Ran Y., Persson, U.M., Abu Hatab, A., Jonell, M., Lindahl, T and E. Röös. (2022). What evidence exists on the effects of public policy interventions for achieving environmentally sustainable food consumption? A systematic map protocol. *Environmental Evidence* 11(1):17.

Hammar T, Hansson P-A, Röös E. (2022). Time-dependent climate impact of beef production – can carbon sequestration in soil offset enteric methane emissions?. *Journal of Cleaner Production* 2022:331.

Sieber, P., Ericsson, N., Hammar, T., Hansson, P.-A. (2022). Albedo impacts of current agricultural land use: Crop-specific albedo from MODIS data and inclusion in LCA of crop production, *Science of the Total Environment*, 2022:835, 155455

Martinsson, E., Hansson, H. (2021) Adjusting eco-efficiency to greenhouse gas emissions targets at farm level – The case of Swedish dairy farms. *Journal of Environmental Management*, Volume 287, 1 June 2021, 112313.

Bennett, E.M., Biggs, R., Peterson, G.D. and Gordon, L.J., (2021). Patchwork Earth: Navigating pathways to just, thriving, and sustainable futures. *One Earth*, 4(2), pp.172-176.

Röös, E., Bajzelj, B., Weil, C., Andersson, E., Bossio, D. and Gordon, L.J., (2021). Moving beyond organic – A food system approach to assessing sustainable and resilient farming. *Global Food Security*, 28, p.100487

Lagnefö, O., Larsson, G., Larsolle, A., & Hansson, P.-A. (2021). Life Cycle Assessment of Autonomous Electric Field Tractors in Swedish Agriculture. *Sustainability*, 2021:3, 11285

Electric and Autonomy

Electric

- + High driveline efficiency
- + Fuel independance
- + Low GHG emissions
- + Reduced maintenace
- ~~Low energy carrying capacity~~
- ~~Frequent refuelings~~
- ~~Heavy vehicles~~

Autonomous (self-driving)

- + 24h-operation
- + Frees up qualified labour
- + Lower vehicle weight – less compaction
- + Good economic potential
- Legally a grey area
- Technical complexity
- ~~No emission reduction~~

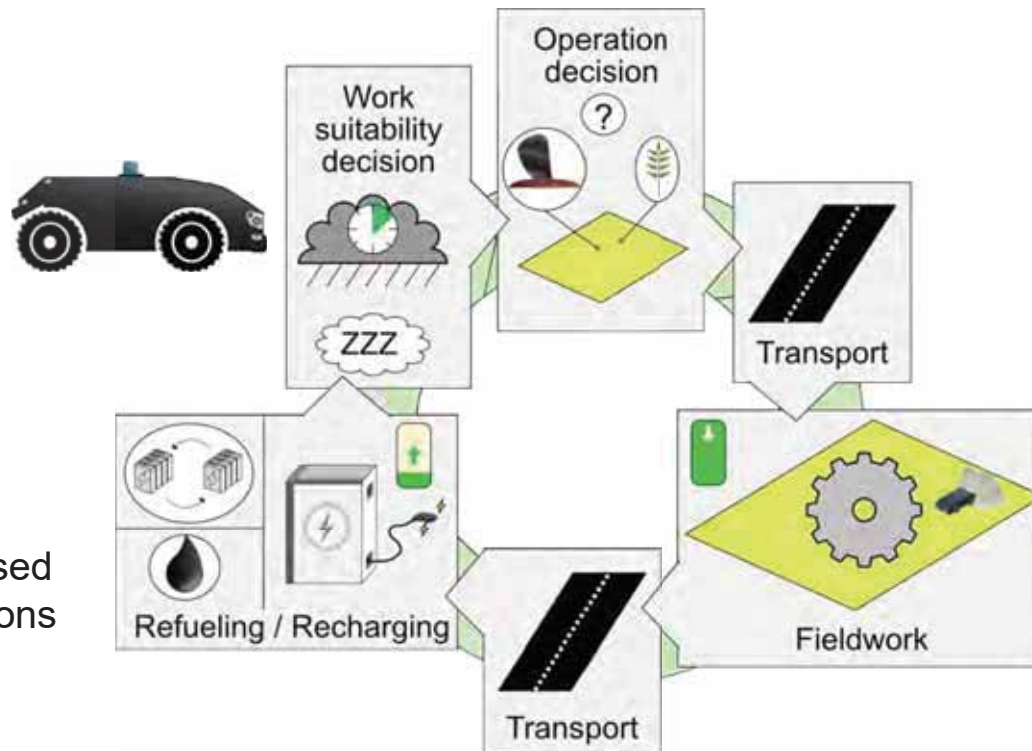
Synergetic effects



Simulation model

Inputs

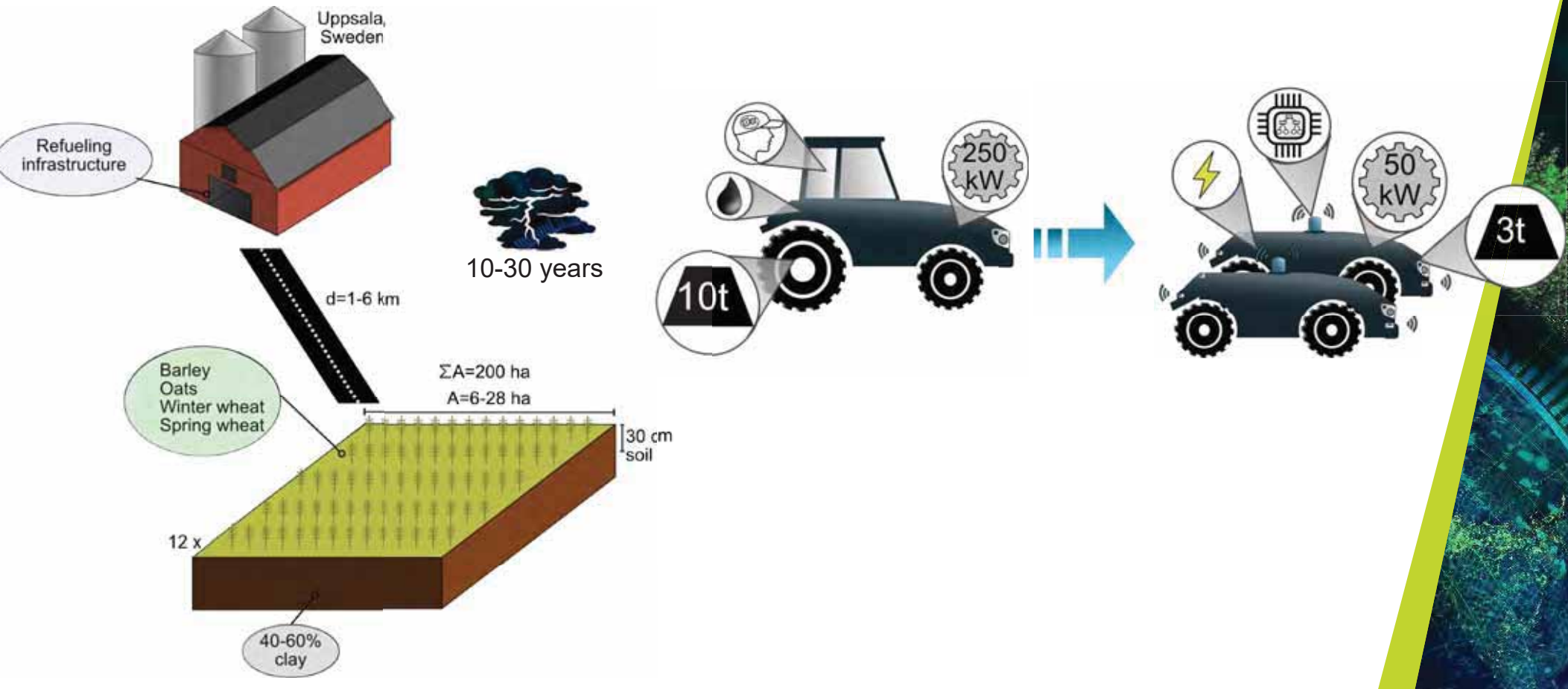
- Vehicle...
- Battery...*
- Charger...*
- Field...
- Weather...
- Soil compaction
- Modules were based on ag.eng. equations



Outputs

- Energy use
- Number of refuelings
- Time required
- Operational cost
- Environmental impacts
- ...and much more

Inputs

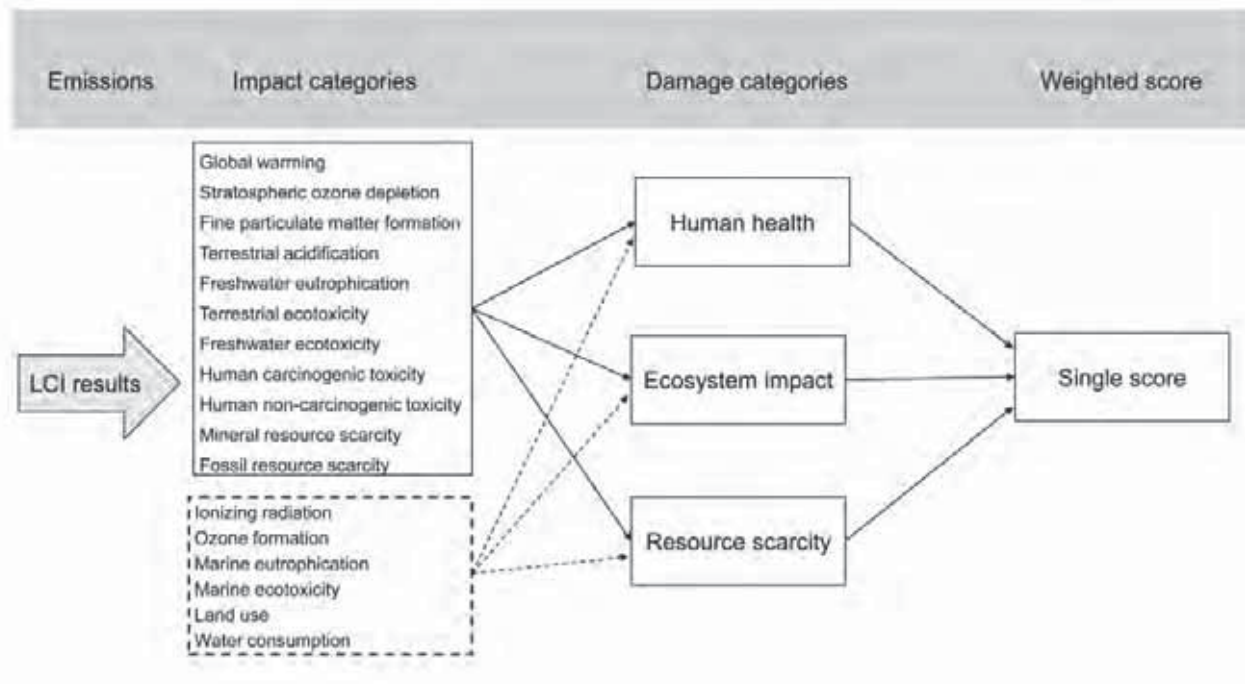


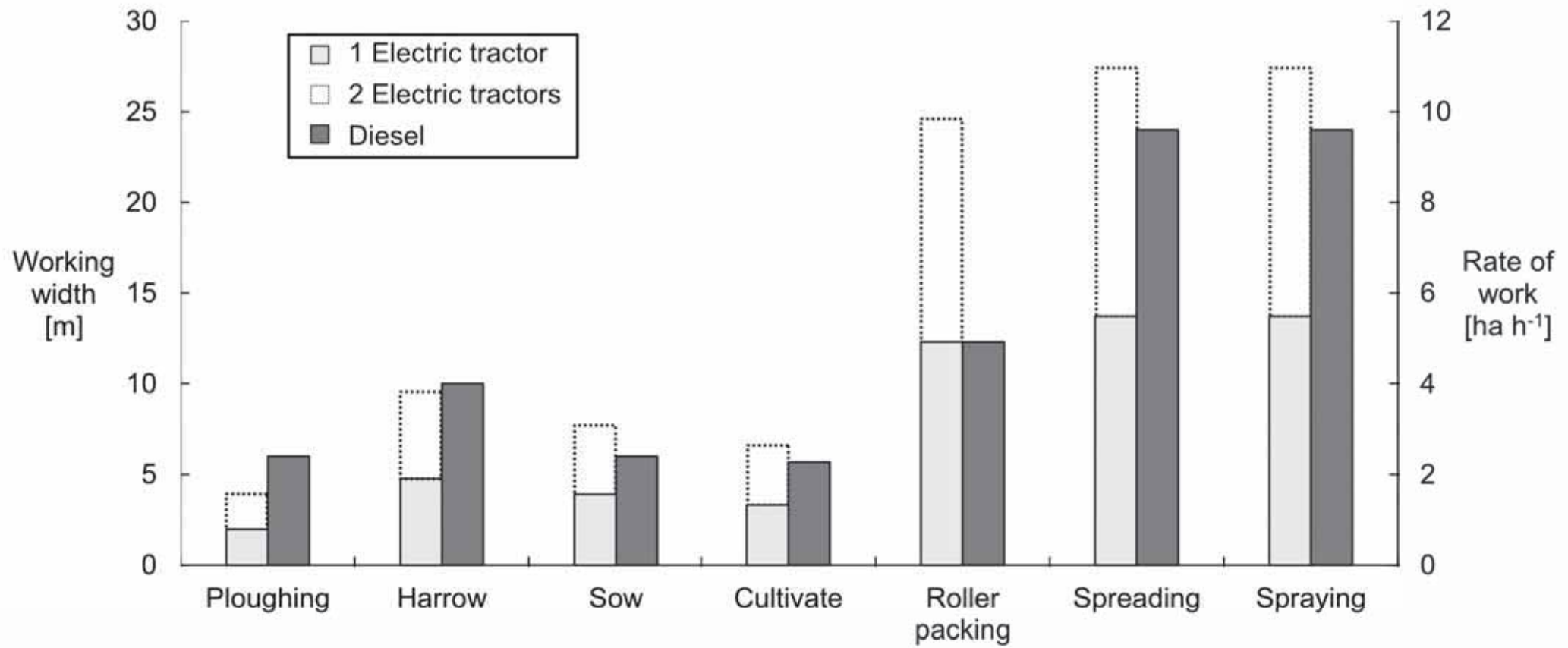
LCA

Production & assembly

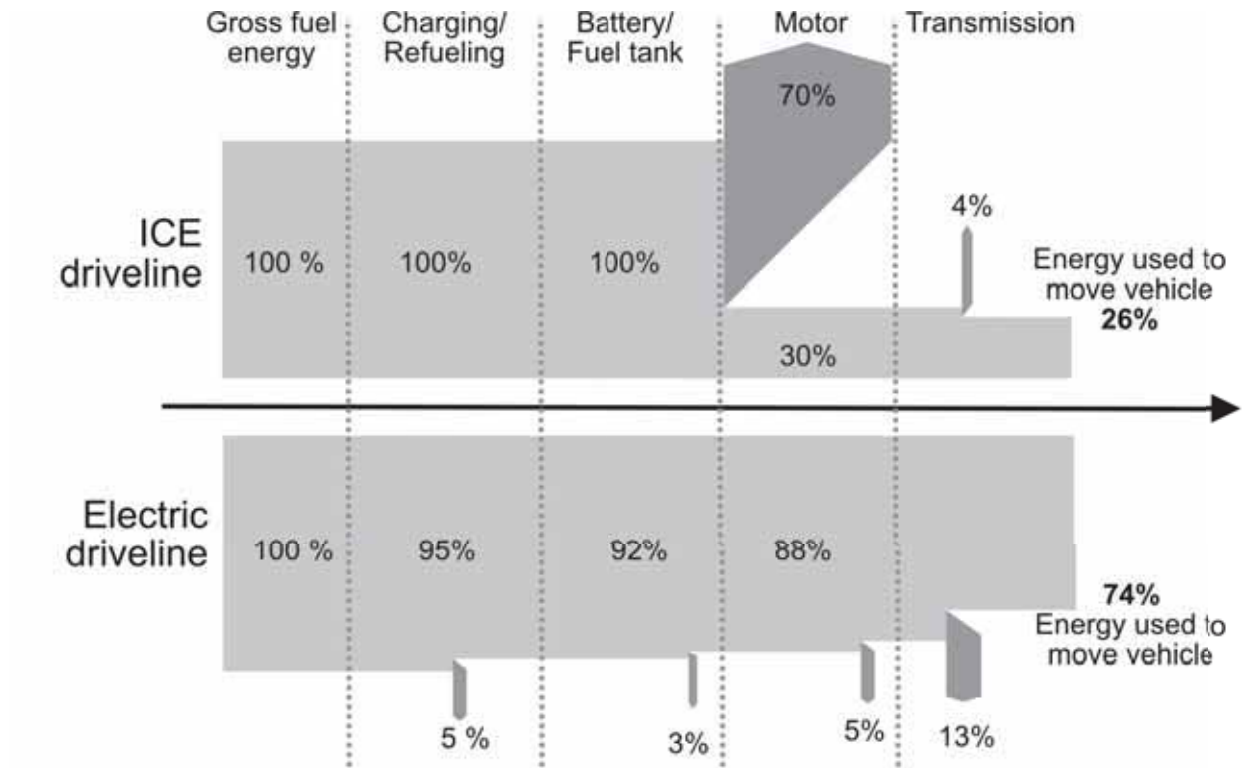
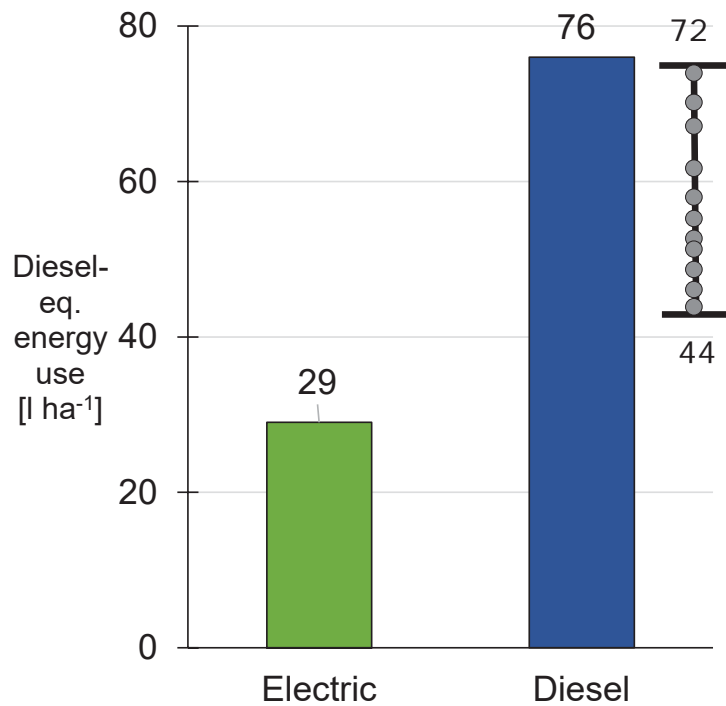
Use phase

End-of-life





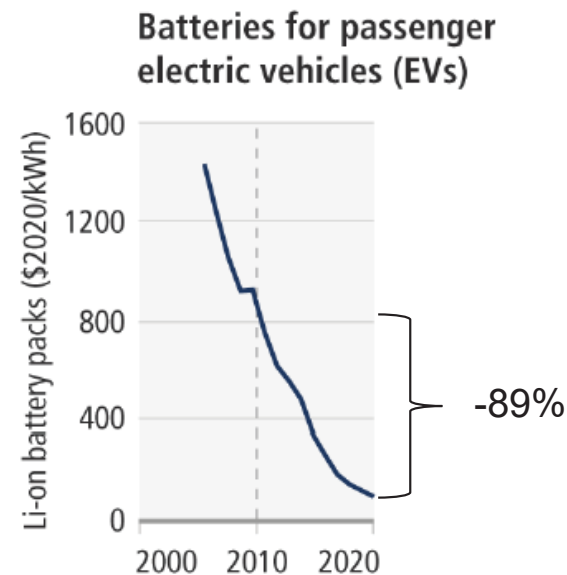
Energy use



Economy

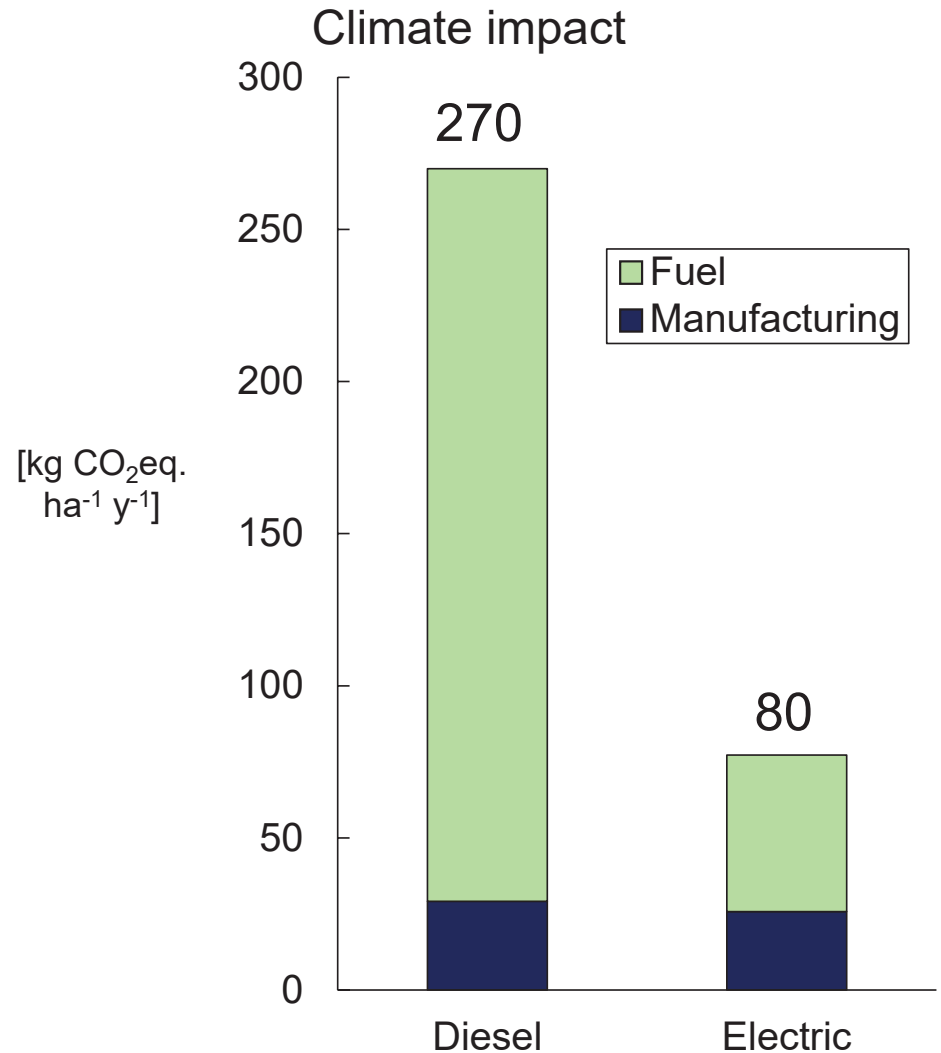
- Higher investment costs
 - Batteries
 - Infrastructure
 - Autonomous drive – costs uncertain
- Lower operational costs
 - Operator
 - Fuel
- 32-37% reduction in total annual cost
 - Compared to conventional tractor

Operator	↓
Fuel	↓
Maintenance	↓
Battery cost	↑
Timeliness	↓↑
Investment	↑
Yearly cost	↓



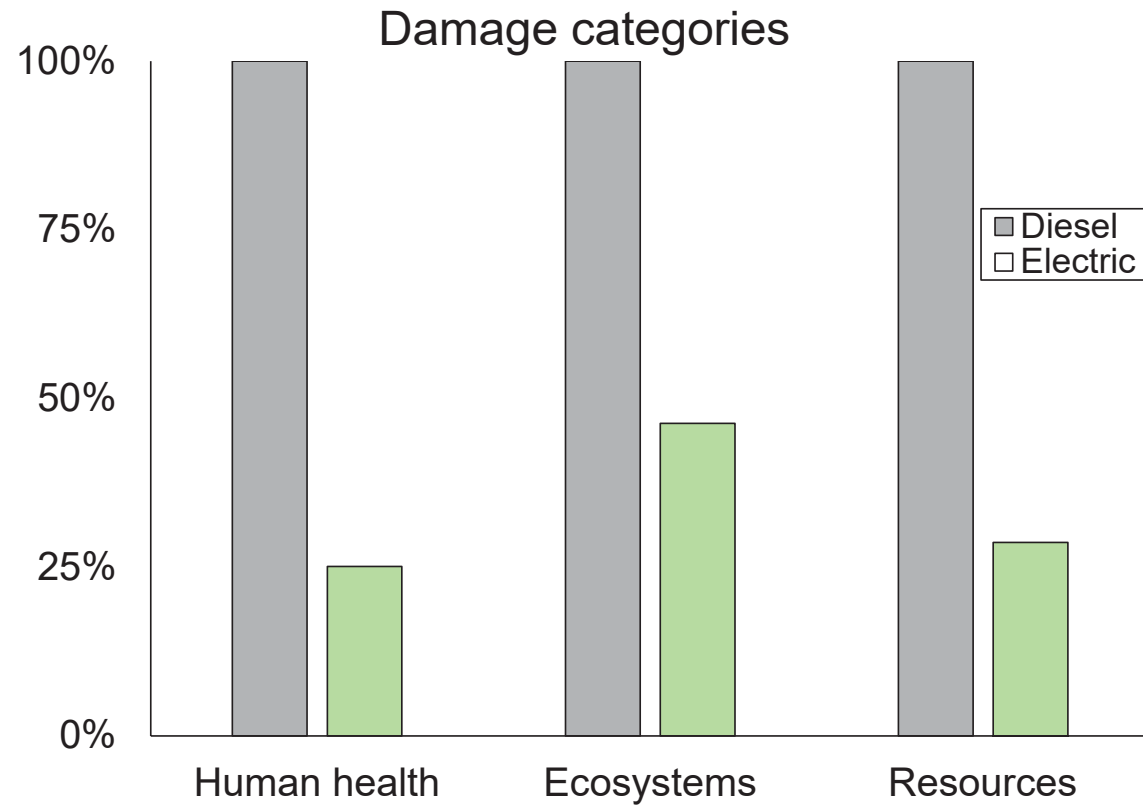
Climate impact

- The electric tractor had 71% lower impact
- Battery production had a strong impact.
 - As did heavy vehicles
- Fuel choice was the most important factor
 - Diesel 100% (no mix-in)
 - Swedish marginal electricity mix
 - 41 % natural gas,
 - 35 % wind,
 - 24 % biomass



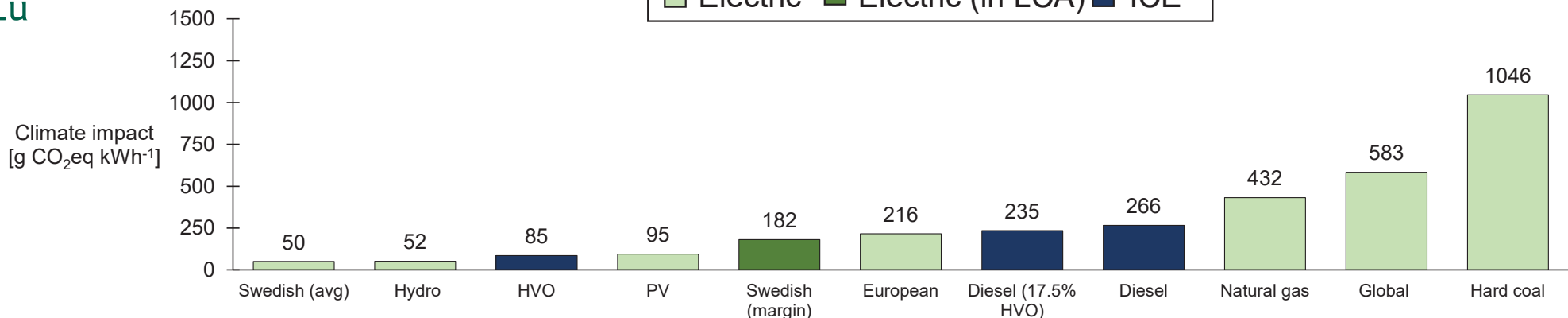
Damage impact

- Electric tractors showed big reductions in all categories
- Fuel very impactful
- The same reduction was seen in the single point score



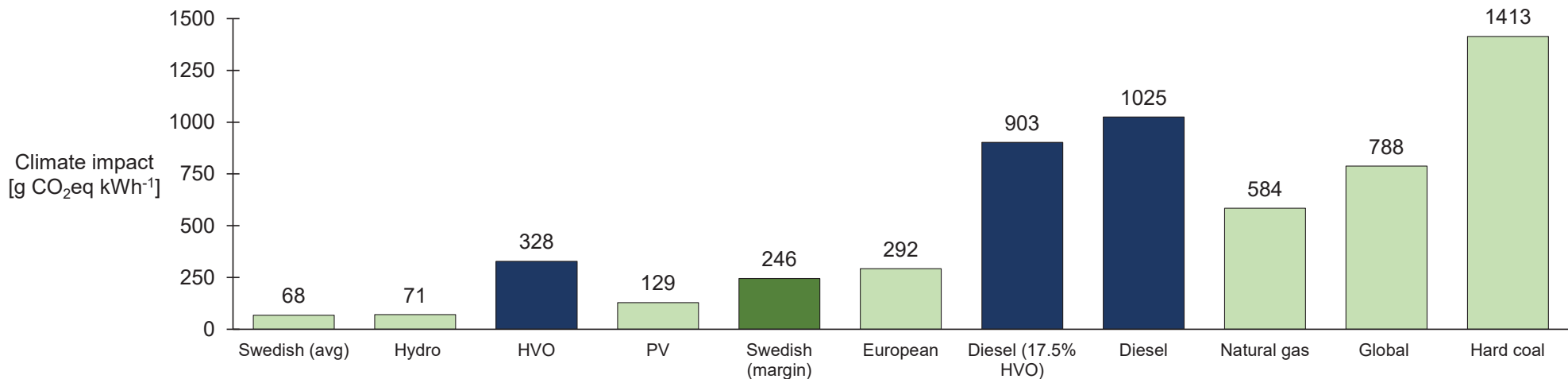


Energy climate impact (before driveline losses)

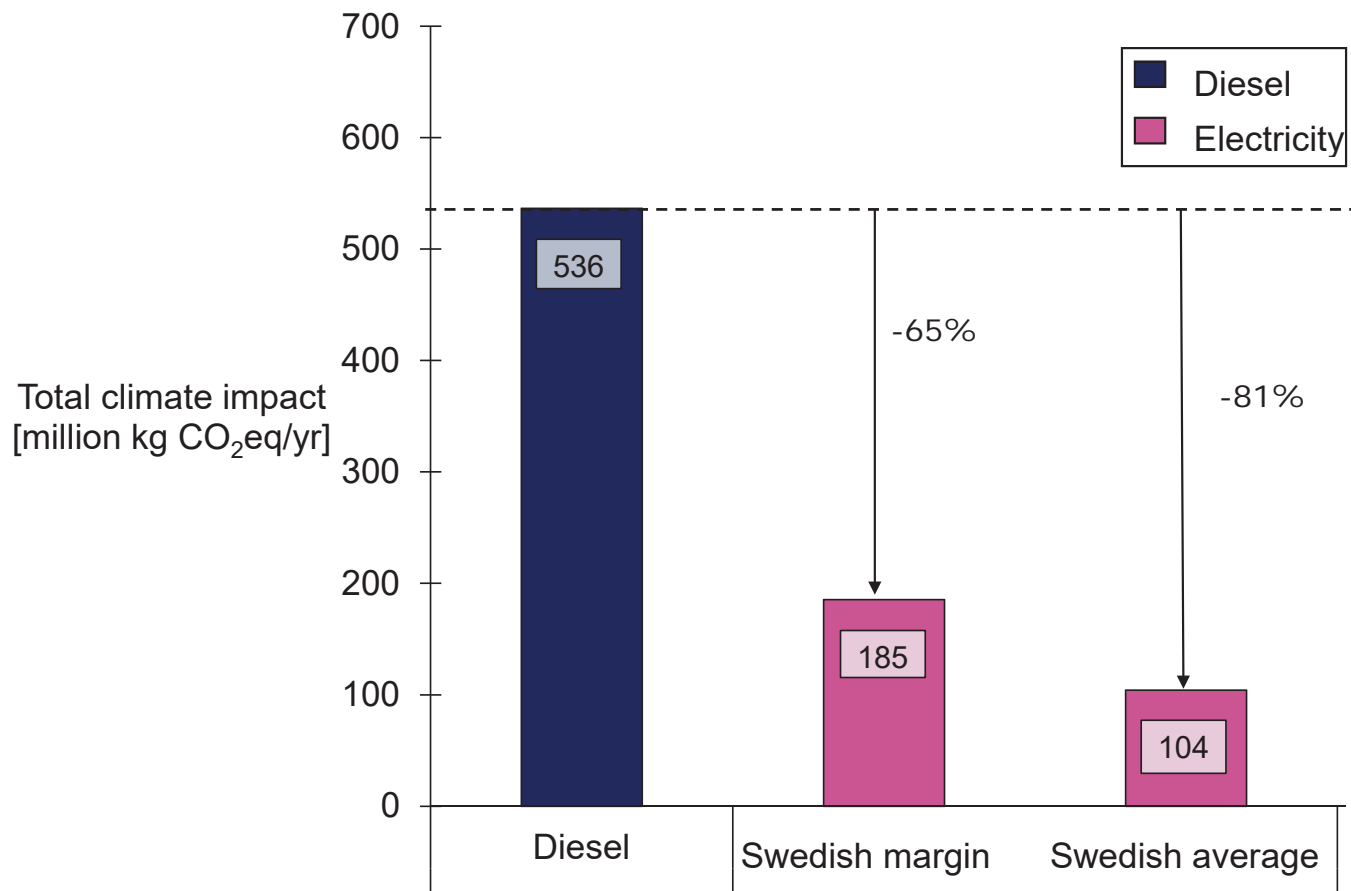


Driveline efficiency losses

Energy climate impact (at wheel)



National potential



- Results scaled up
 - Ley, grains & other
 - Clay and sand soils
- 2-3% of Swedens transport GHG emissions

Conclusions

- Electric tractors requires larger investments, but have lower operational cost
 - Economically (-1/3) & environmentally (-2/3)
- Fuel is the largest climate factor for both diesel and electric tractors
 - "Cleaner" fuel & reduced energy use possible
- Battery optimizations are important factors
 - Not one size fits all
- Autonomy is a key driver, but electric tractors can compete and provide benefits.
 - Many "hidden" benefits





Thank you for you attention!

Oscar Lagnelöv

PhD Candidate

Swedish University of Agricultural Sciences

Department of Energy and Technology

+46(0) 70 221 79 79

Oscar.lagnelov@slu.se

SCIENCE AND
EDUCATION
**FOR
SUSTAINABLE
LIFE**



SCIENCE AND
EDUCATION **FOR**
SUSTAINABLE
LIFE

A low-cost robot platform for swarm applications in agriculture

Robert Braunschweig

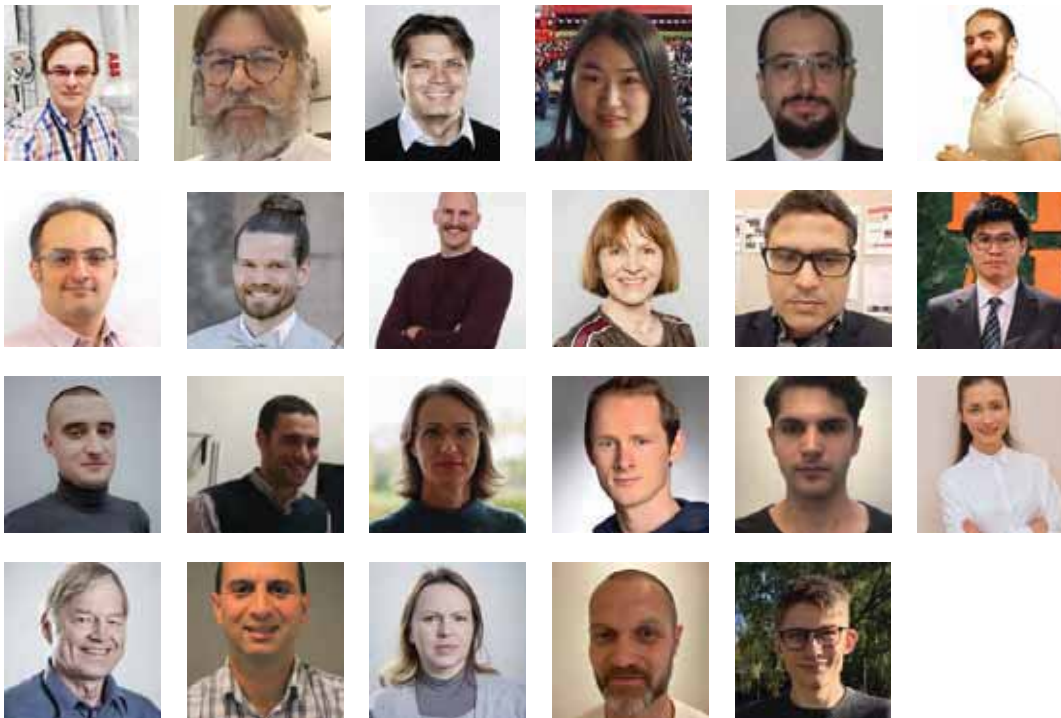
29.11.2022



Agenda

1. NMBU Robotics Group
2. Background and Motivation
3. Swarm Robot Platform
4. Future Research

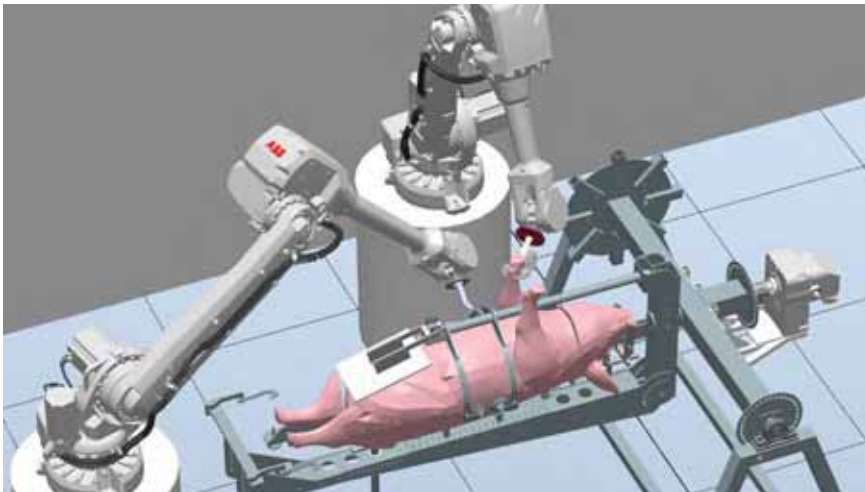
1. NMBU Robotics Group



- 2 main themes in the agrifood value chain *on farm* and *food processing*
- Currently ca. 25 members including academic staff, PhD students and administrative/technical support
- Research underpins NMBU's 5-year master programme in *applied robotics*

1. NMBU Robotics Group

RoBUTCHER – A Robust, Flexible and Scalable Cognitive Robotics Platform



Thorvald – Modular Robot Platform for Agriculture



Open Fields

Polytunnels

Greenhouses

2. Background and Motivation

Robots in Agriculture

UGV

UAV

Special



Fendt






Panasonic



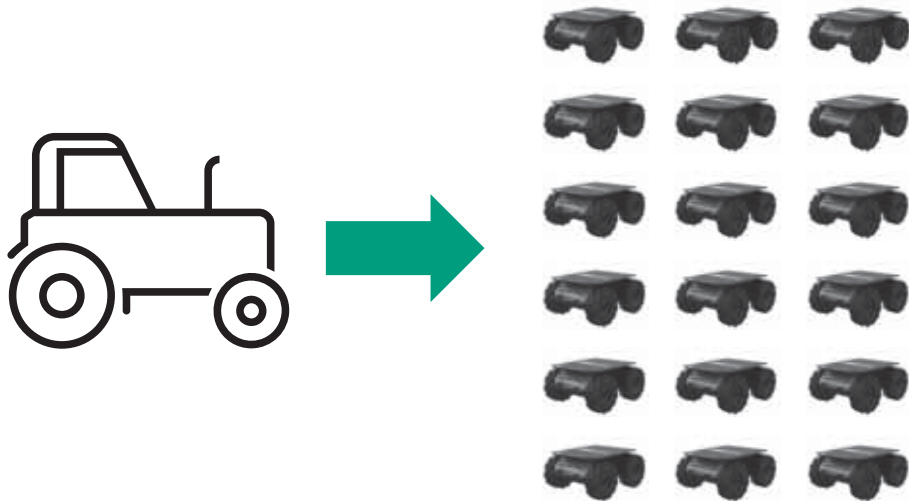
NMBU

2. Background and Motivation

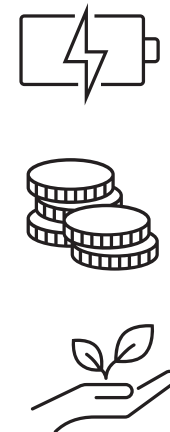
		
<p>Small: EarthSense TerraSentia</p> <ul style="list-style-type: none"> - Commercial sensing and phenotyping platform - RTK GPS, Camera, IMU, Encoder, 2D/3D Lidar - Light, low-cost, easy to use 	<p>Medium: Saga Robotics Thorvald</p> <ul style="list-style-type: none"> - Robust, commercially available products - Weight: ca 200 kg, Payload: ca 150-250 kg - RTK GPS, Camera, - IMU, Encoder, 2D/3D Lidar 	<p>Large: AutoAgri IC Series</p> <ul style="list-style-type: none"> - Pilot and research platform - Weight: ca. 2500kg, Payload: ca. 2000 kg - RTK GPS, Cameras, Flexible mounting of own equipment (tractor replacement)

2. Background and Motivation

Can a swarm of small robots replace a tractor or bigger robot?



Improvements in costs, reliability, sustainability?



2. Background and Motivation

- Inspiration for project:
 - MARS/XAVER project led by Fendt
 - Research since 2015, now commercializing
 - Focus on precision seeding
 - <40 kg, ca. 50x40 cm



Copyright: Agco GmbH

3. Swarm Robot Platform

Pros

- **Costs:**
 - Low HW costs, scalability
 - Low energy costs, lightweight
 - Less human labour
- **Sustainability:**
 - No soil compaction
 - Fully electrified
 - Site-specific operation
 - Redundant system
- **Safety:**
 - Low weight and power

Cons

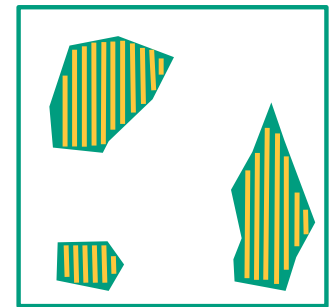
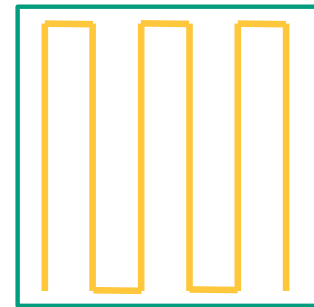
- Software complexity
- Limited number of applications
- No standard equipment



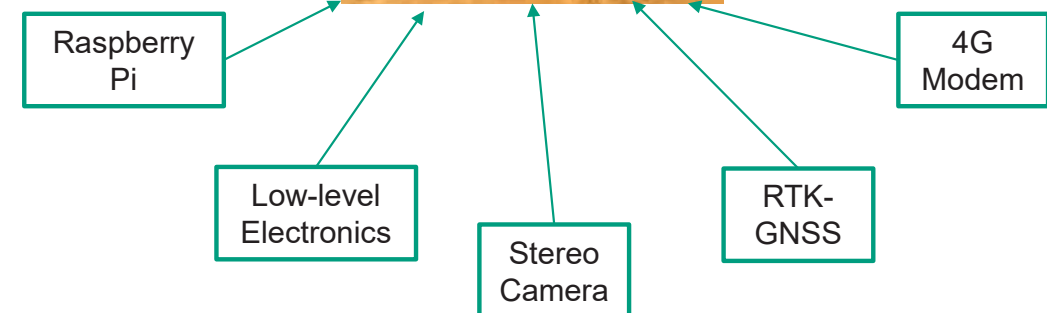
3. Swarm Robot Platform

Applications

- Precision seeding (time/space dependent)
- Mechanical weeding
- Sensing/phenotyping
- Site-specific treatment



3. Swarm Robot Platform



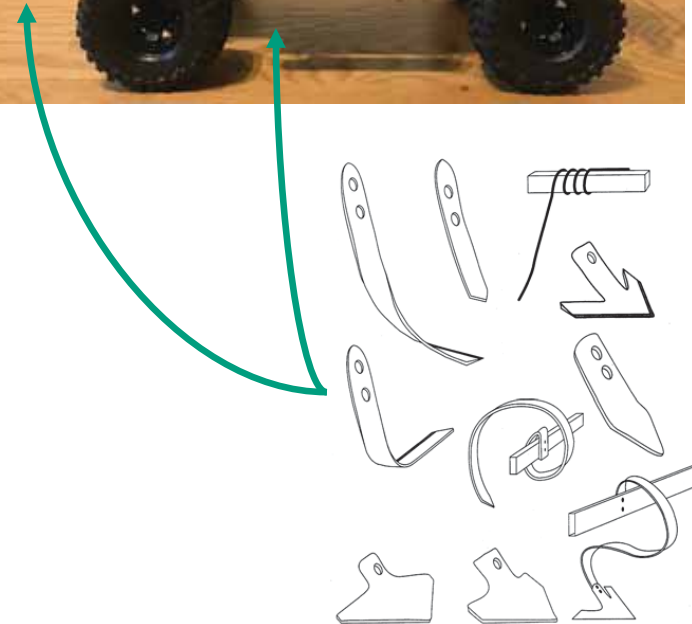
- Mobile ground robot based on Traxxas RC car:
Small **V**ehicles for **A**utonomy (SVEA), KTH Stockholm
- Adapted for outdoor application
- Pros:
 - Cheap and robust mechanics
 - Off-the shelf components
 - ROS compatible

Size (LxWxH)	55 x 25 x 24 cm
Weight	3.5 kg
Price	~ 1500 €

3. Swarm Robot Platform

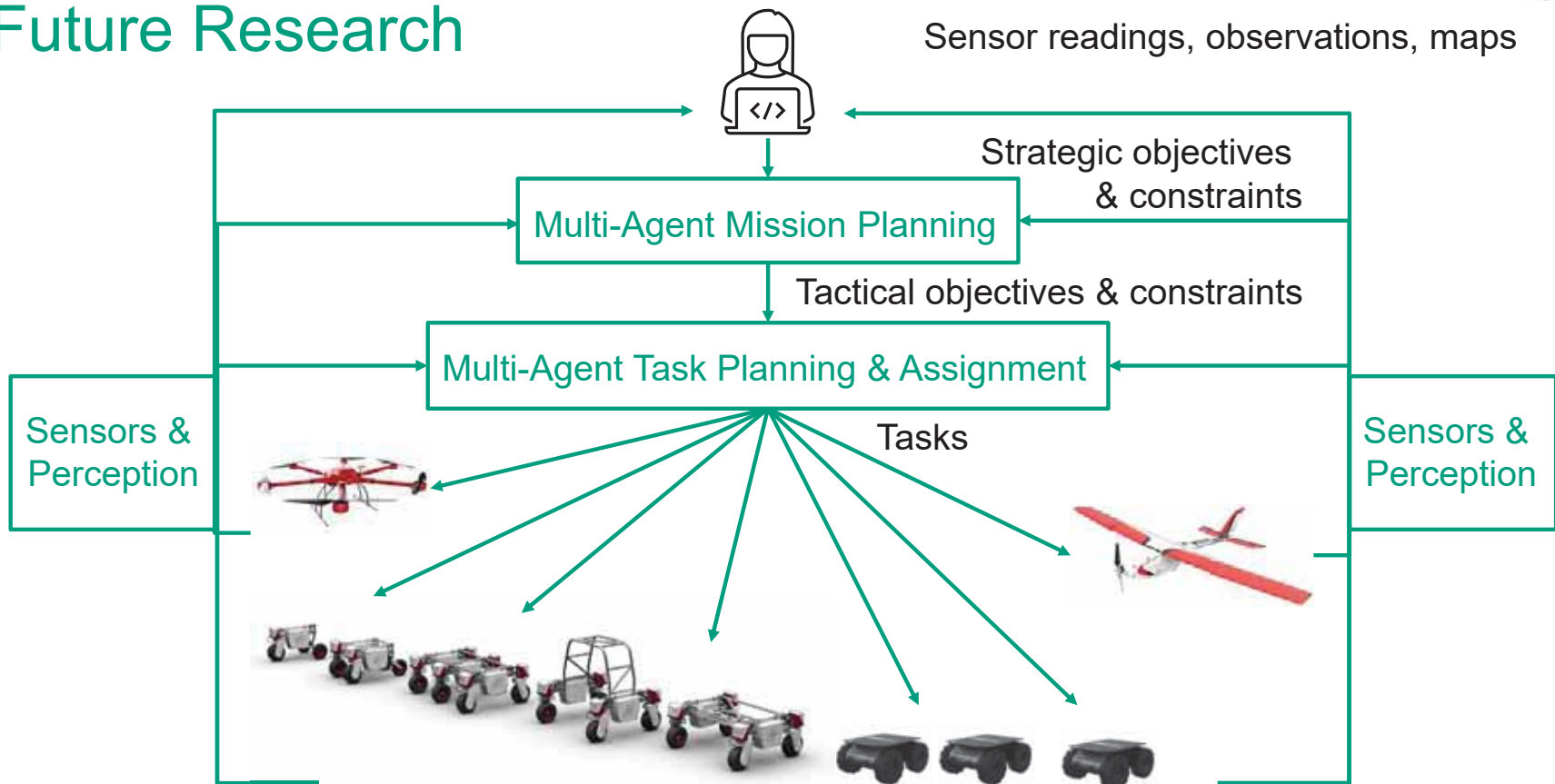
Ongoing work

- Robust design
- Mounting option for weeding tools
- Traction, speed, endurance tests
- Energy monitoring
- Navigation by GPS and camera



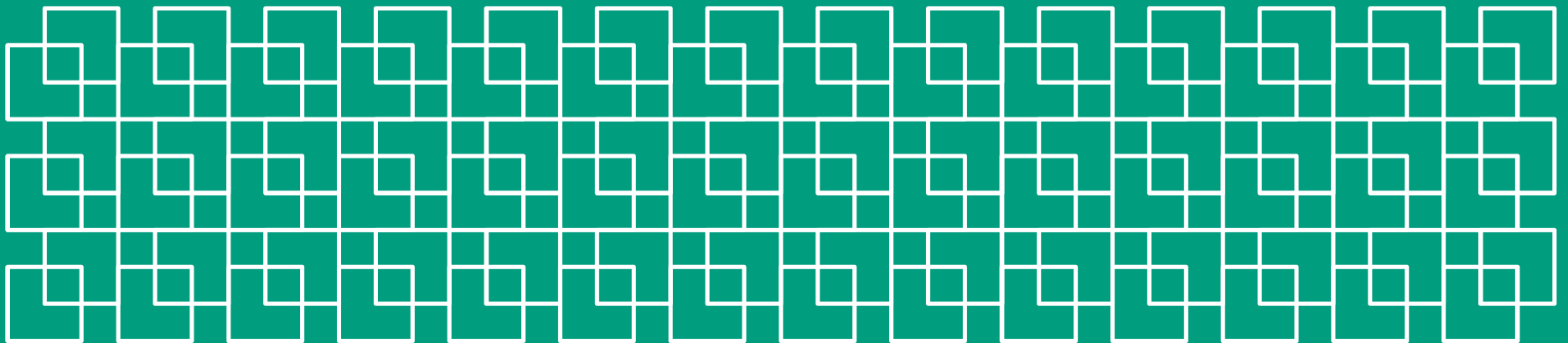


4. Future Research





Thank you for your attention



Smart Farming and Green Innovation - a platform for sustainable agriculture

Nils Bjugstad
REALTEK NMBU



Bærekraftsarena
Smart Farming



Smart Farming | Grønn Innovasjon
Green Innovation Student LAB



Norges miljø- og
biovitenskapelige
universitet

NMBU.no







Outline

The objectives of the platform

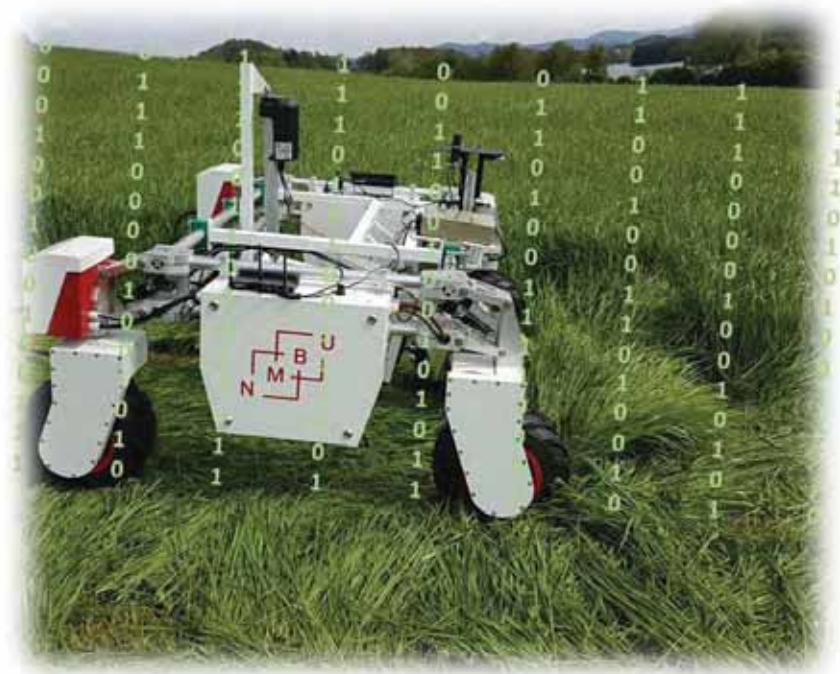
Interdisciplinary topics within education and research

Field robotics in Norway

Green Innovation Student LAB

GrasRobotics

Introduction to other projects within Smart Farming





Funding– «Smart farming and green innovation»



Year				
NOK	700.000	700.000	700.000	2.800.000



Research groups within, robotics, drones, precision agriculture, GIS and GNSS, sensors, image analysis, computer science, phenotyping, plant sciences, fertilizer sciences and soil sciences are included in the group, which can also be expanded as needed.





Goals within «Smart farming and green innovation»

Main goals:

Develop a strong and complementary professional environment within "Smart Farming and Green Innovation" at NMBU. The academic community at NMBU will be a national leader with interdisciplinary knowledge and have well-developed networks with top international environments. An active and targeted collaboration will be developed with the business sector to increase the use of "Smart Farming", increased innovation and strengthened training of students and candidates. NMBU as a sustainability university will be strengthened.

Research:

Over a four-year period, the group will have initiated several strong research projects. The goal is to develop an SFI (Centre for Research-based Innovation), or equivalent a larger long-term research programme, together with user organisations and other related academic communities in Norway.

Education:

The goal is to develop the **Green Innovation Student Lab** (GISL) where interdisciplinary students and researchers at NMBU together with the business community create new innovative solutions for sustainable agriculture. GISL holds both demo/experimental fields, access to sensors and other research infrastructure. Students will participate in research early in their studies.

Master's theses, internships etc. are linked to innovative solutions and Smart Farming in cooperation with the business community. Relevant master's and PhD students come from robotics, image analysis and physics, computer science, geomatics, plant, soil and economics sciences.

Innovation:

The aim is to strengthen innovation activities for students and employees both by having outstanding research and teaching, but also through systematic development of cooperation with business and external actors. New knowledge and research-based ideas will be developed and commercialised through a strengthened culture of entrepreneurship for the benefit of business and society.

DEVELOPMENT OF AGRICULTURE

TECHNOLOGICAL REVOLUTIONS

1.0



WORK-INTENSIVE

- Work intensive
- Low productivity
- 1/3rd of population involved-required

2.0



GREEN REVOLUTION

- Artificial fertilizers
- pesticides
- More efficient equipment
- Productivity dramatically increased

3.0



PRECISION AGRICULTURE

- Precision operations within crops
- Individual treatment of animals vs total flock
- Automatic steering with 10mm precision
- Sensors and controls

4.0



SMART FARMING

- Internal and external network integration of agricultural operations
- Cloud service usage, large data sets processed
- Cheap and advanced sensors
- Big data analytics
- New algorithms that transform raw data into insight

5.0



ROBOT FARMING

- Operations without human presence
- Artificial intelligence, self-learning systems
- Production systems adapted to plant/animal needs
- Food production, consumer needs fulfilled
- Controls of ingredients/ internal components



Science and Technology master studies at NMBU related to SF

Geomatics

Applied robotics

Engineering, processing and product development

Data science

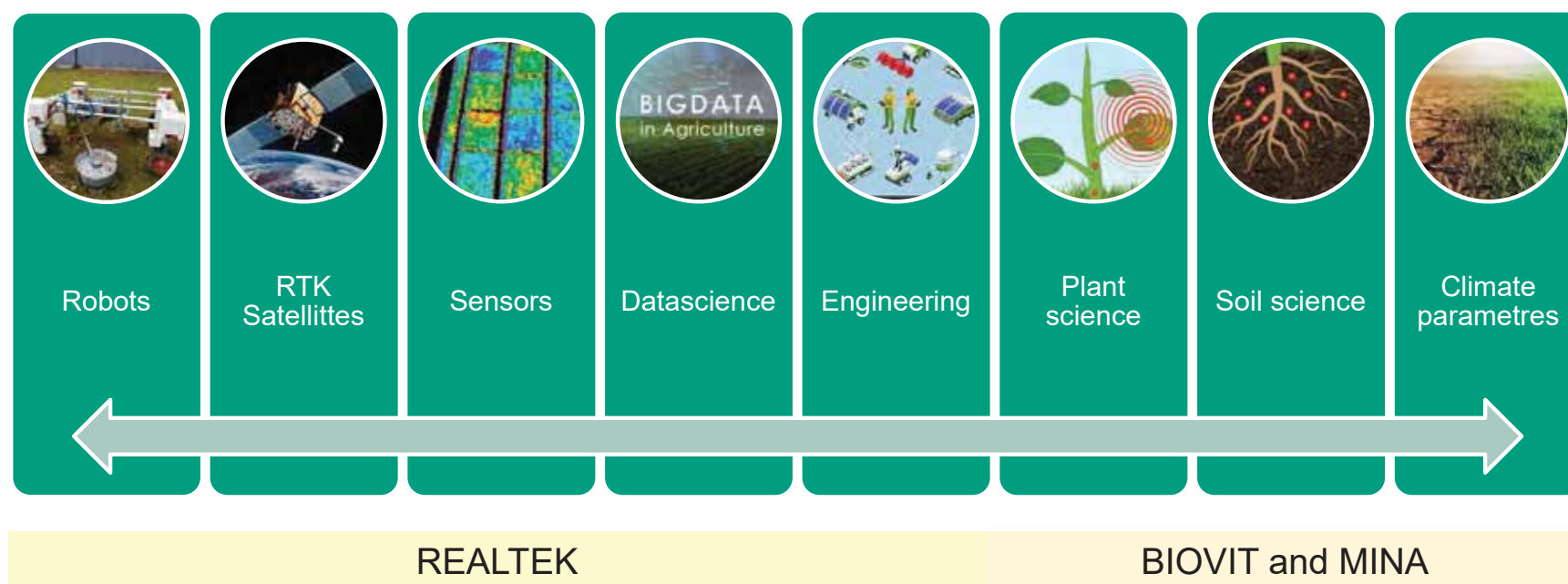
Physics (especially imaging and energy technology)

Plant science

Soil science

Others

Research groups and topics



SKP – Centre for Plant Research in Controlled Climate (senter for klimaregulert planteforskning)

SHF – The livestock Production Research Centre (senter for husdyrforskning)

What is **reen Innovation Student LAB?**





Pilot areas for education and research at NMBU

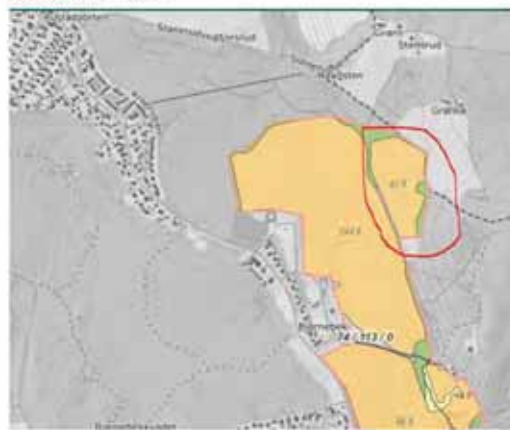


20 daa Kjærringjordet



14 daa Vallebekk

Grønalia - Bjørnabekk - 42 - 1



Grønalia - 42 daa
Høybråtan - 20 daa

Høybråtan - Sørås - 42 - 1



A FarmBot robotic system is shown in a garden bed. The robot is a white metal frame with a vertical mast on the right side, equipped with a camera and a watering arm. It is positioned over a raised garden bed filled with dark soil and small green seedlings. The background is a dense line of trees with green and yellow foliage, suggesting an autumn setting. The text "FarmBot" is overlaid in the center of the image.

FarmBot

Take Back Control

[WATCH THE VIDEO](#)

[ORDER NOW](#)



DAT



Benefits of precision spraying

Conserve herbicide

Experiments have shown that a typical barley or oats farm can reduce its herbicide consumption by 50%.

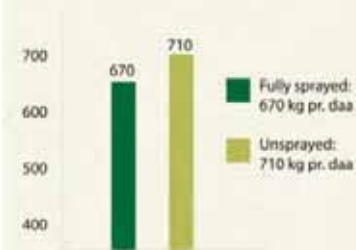
Provided a herbicide cost of NOK 300 / EUR 30 per ha, a 500 ha farm will save NOK 150 000 / EUR 15 000 per season.

Increase yields

We have observed that non-sprayed areas of barley and oats have a yield increase of 6-7%.

This will result in increased sales value of NOK 500 / EUR 20 per ha. For a 500 ha farm: NOK 250 000 / EUR 10 000.

Field trials in barley and oats indicates 6-7 % yield increase.



Harvest quantities registered by





Projects within smart farming to be presented at this seminar



Presentations Smart Farming Sustainable Plattform NMBU		Speakers
A low-cost robot platform for swarm applications in agriculture		Robert Braunschweig, NMBU, N
Impact of Sun Elevation Angle and Type of Sensor on Multispectral UAV Imagery Data		Sahameh Shafiee, NMBU, N
Spectral Imaging in Crop Fields – Research Activities at NMBU		Ingunn Burud, NMBU, N
Sensor technology for optimal strawberry harvesting		Siv Fagertun Remberg, NMBU, N
Prevention of subsoil compaction: Technologies and strategic planning		Mathieu Lamandé, NMBU/AU, DK

Impact of Sun Elevation Angle and Type of Sensor on Multispectral UAV Imagery Data

Sahameh Shafiee

Ingunn Burud

Morten Lillemo

Norwegian University of Life Sciences



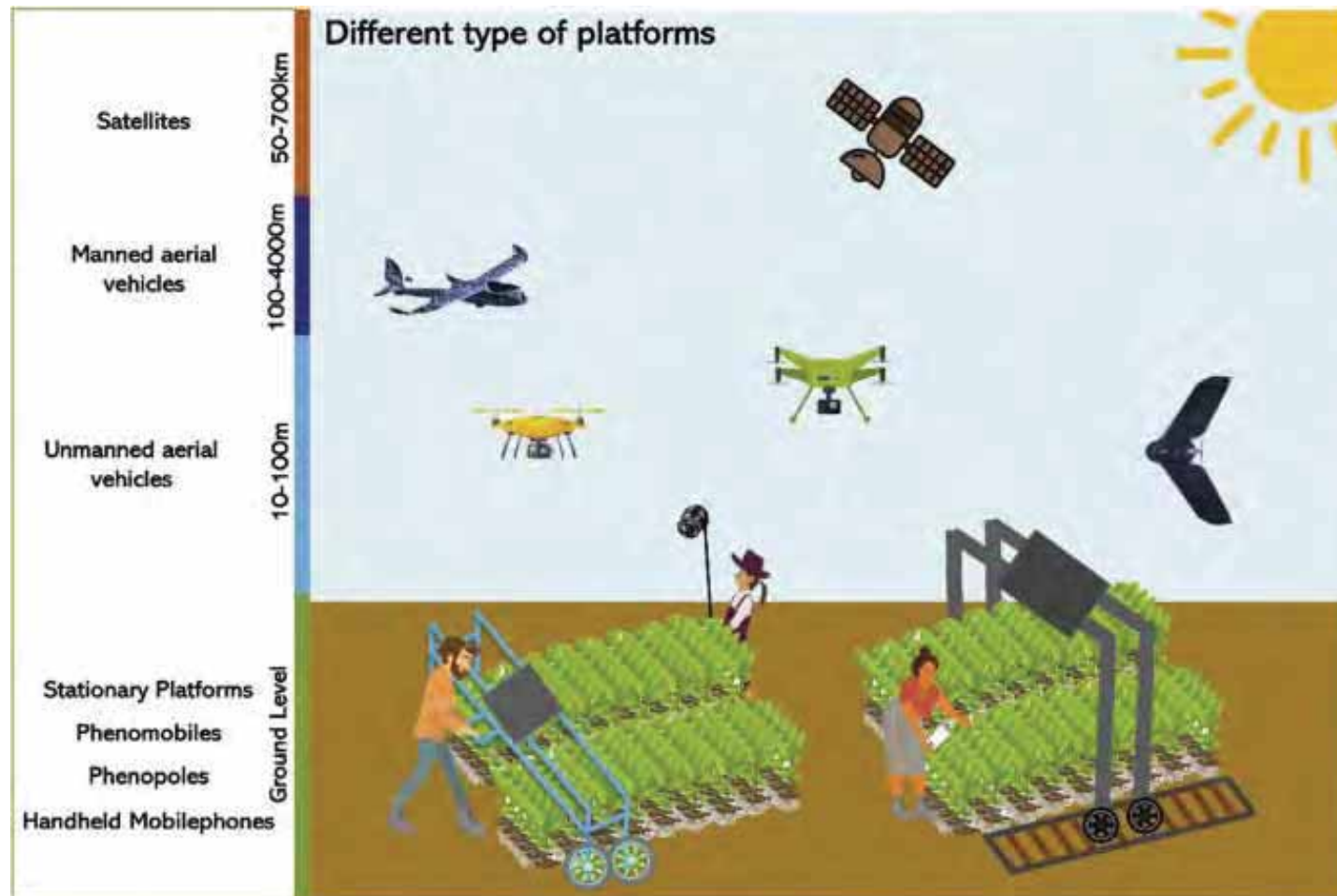
NJF Conference, Herning. 29.11.2022

UAV Applications in Agricultural



- Plant Phenotyping
- Planting seeds
- Field Monitoring
- Disease control
- Pest and herbs control
- Crop damage assessment (lodging)
- Planting seeds
- Livestock management
- And lastly pollination drones



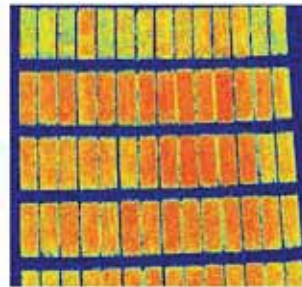


[Field Based High Throughput Phenotyping Rapidly Identifies Genomic - gn.racesociety.com](http://gn.racesociety.com)

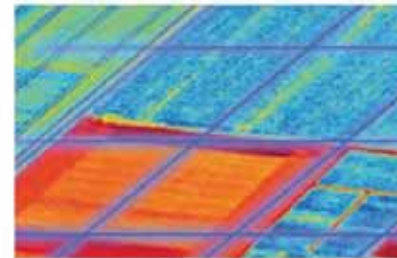
What we can do with a UAV?



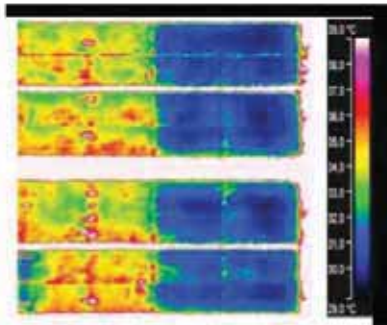
RGB images



Multispectral images



Hyperspectral images

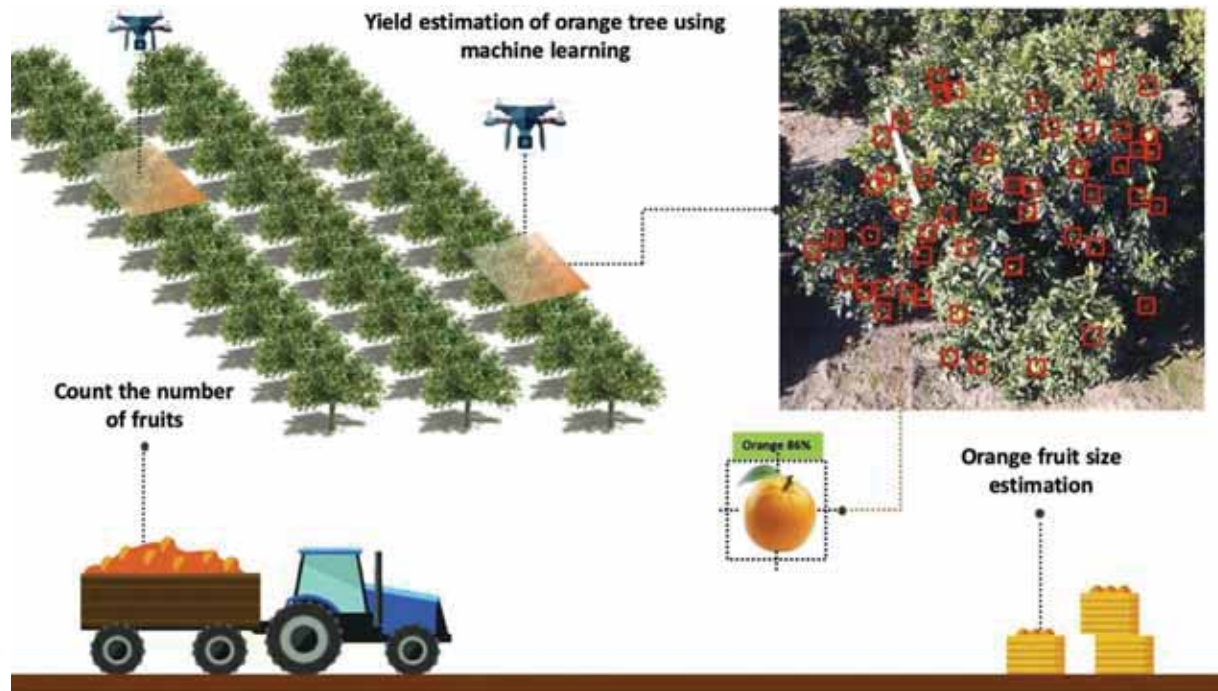


Thermography (IR)



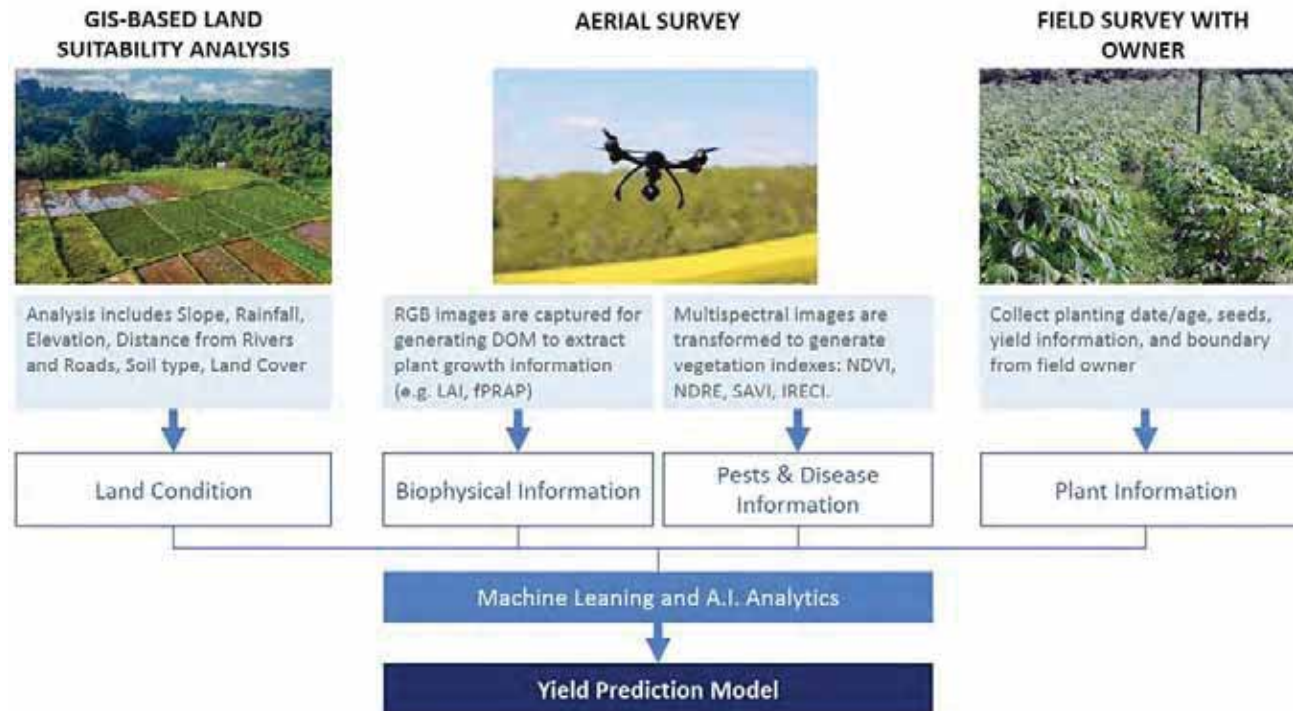
3D models

Yield Estimation



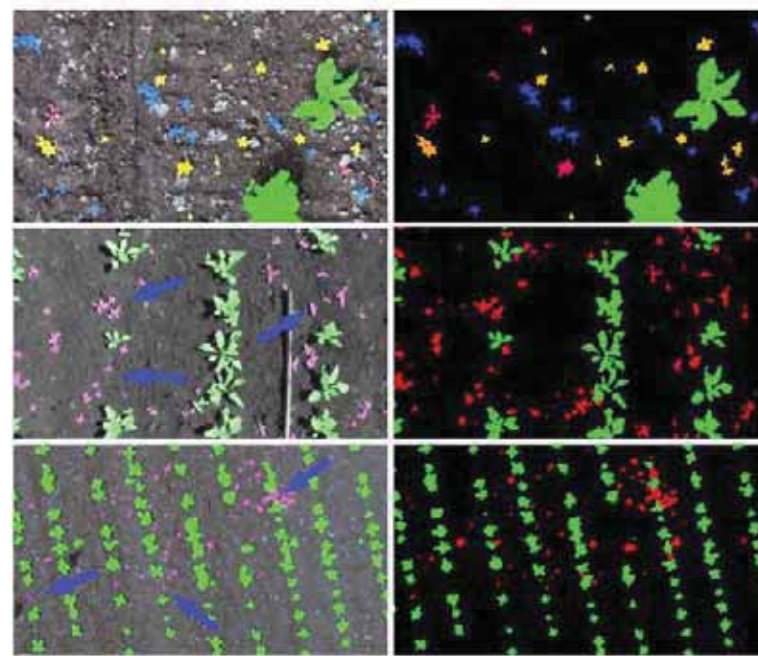
<https://www.sciencedirect.com/science/article/pii/S1161030120300381#fig0055>

Development of Cassava Yield Prediction Model



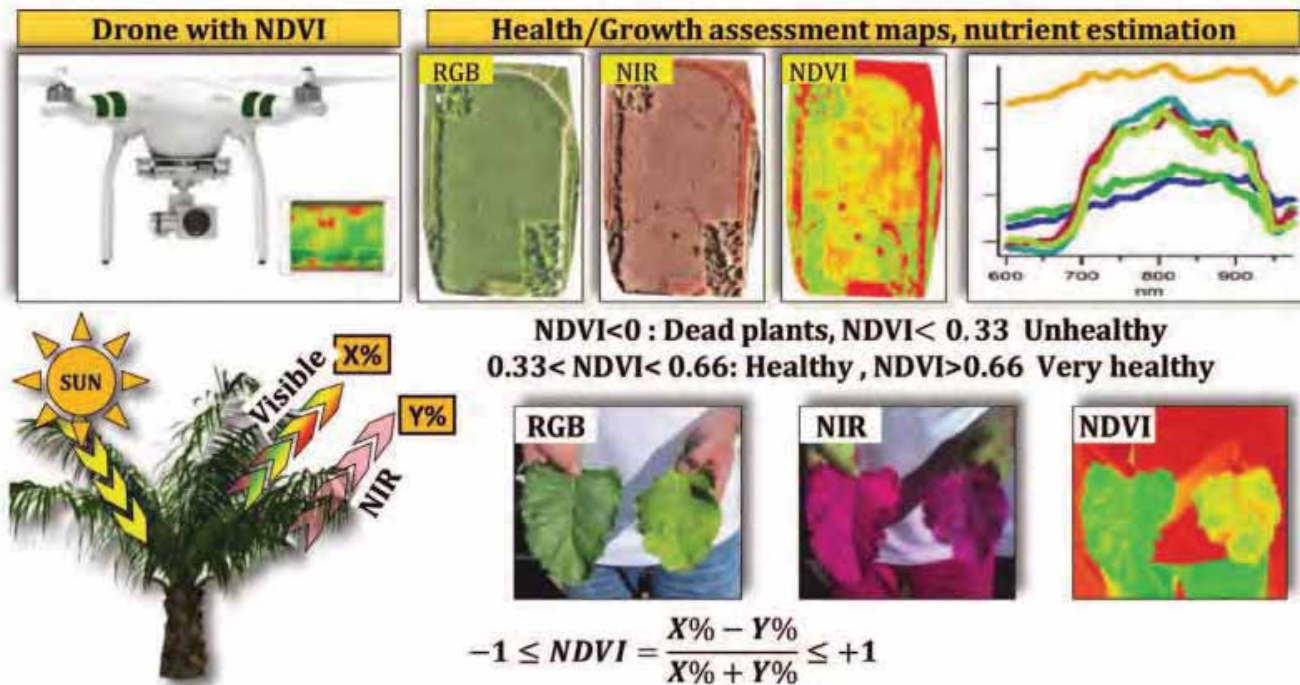
<https://anavision.com/blog/cassava-yield-prediction/>

Weed Detection



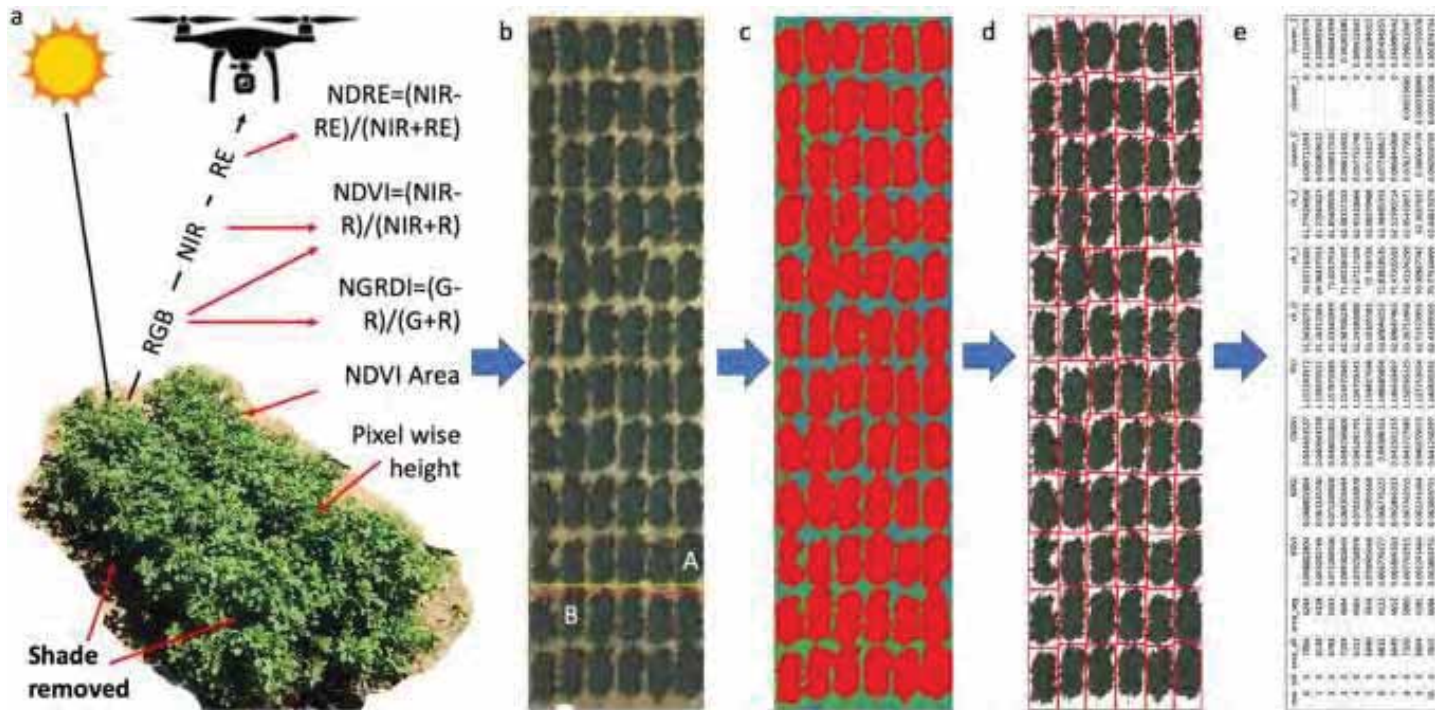
<https://robohub.org/uav-based-crop-and-weed-classification-for-future-farming/>

Disease detection



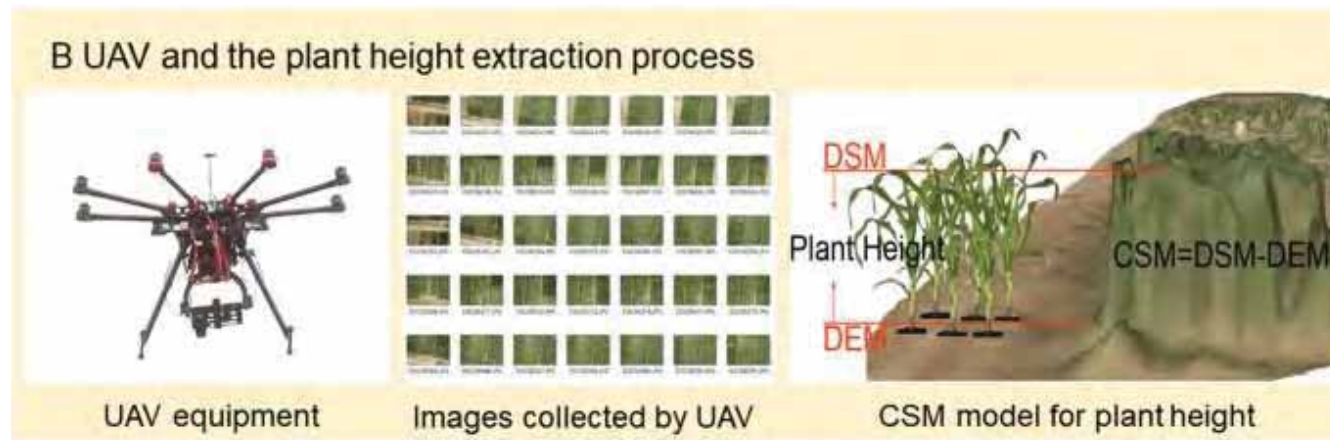
UAV sample NDVI mapping for health assessment and disease UAV sample NDVI mapping for health assessment and disease detection (Shamshiri et al., 2018).

Biomass Estimation



Validation of UAV-based alfalfa biomass predictability using photogrammetry with fully automatic plot segmentation | Scientific Reports (nature.com)

Plant Height Measurement

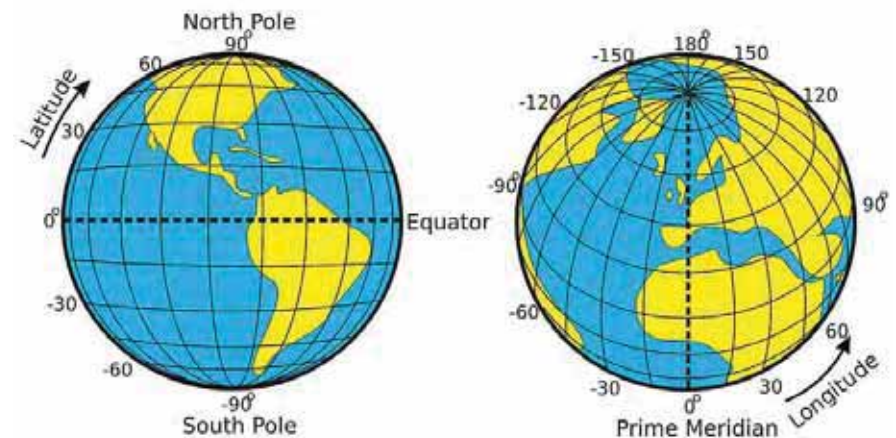


[Dynamic plant height QTL revealed in maize through remote sensing phenotyping using a high-throughput unmanned aerial vehicle \(UAV\)](#)
[| Scientific Reports \(nature.com\)](#)

Is it possible to use the same protocol for UAVs?

Case study: UAVs in high latitudes

- 1) How does the time of flight or sun elevation angle affect data gathering flexibility and quality by UAVs?
- 2) Does the sun elevation angle have a significant effect on trait prediction using UAV data?
- 3) How does the type of sensor could affect data quality and model prediction performance?



[The Distance Between Degrees of Latitude and Longitude \(thoughtco.com\)](http://thoughtco.com)

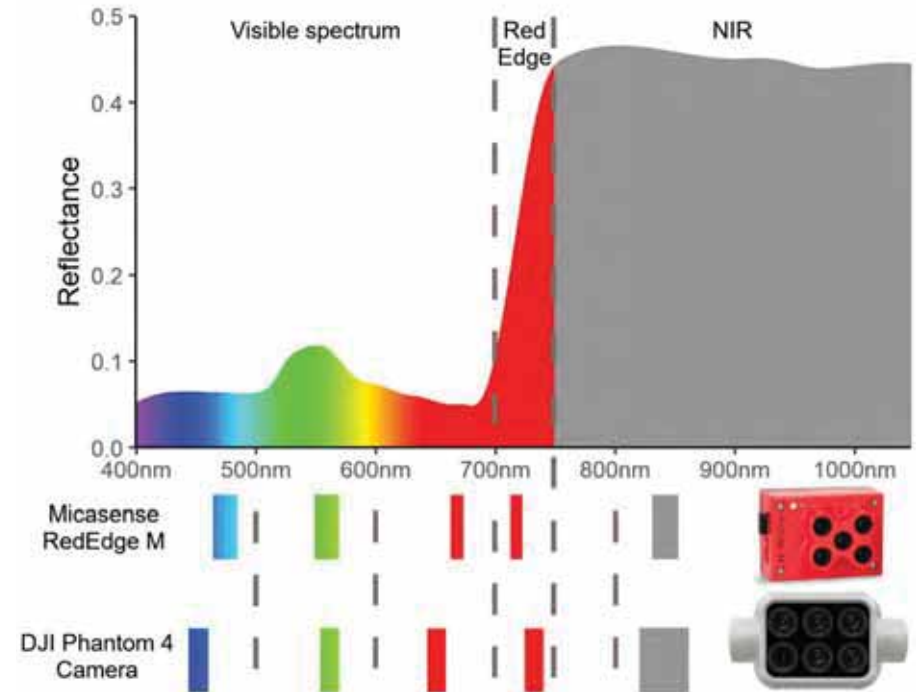
What types of sensors?



Phantom 4
Multispectral
(P4M)



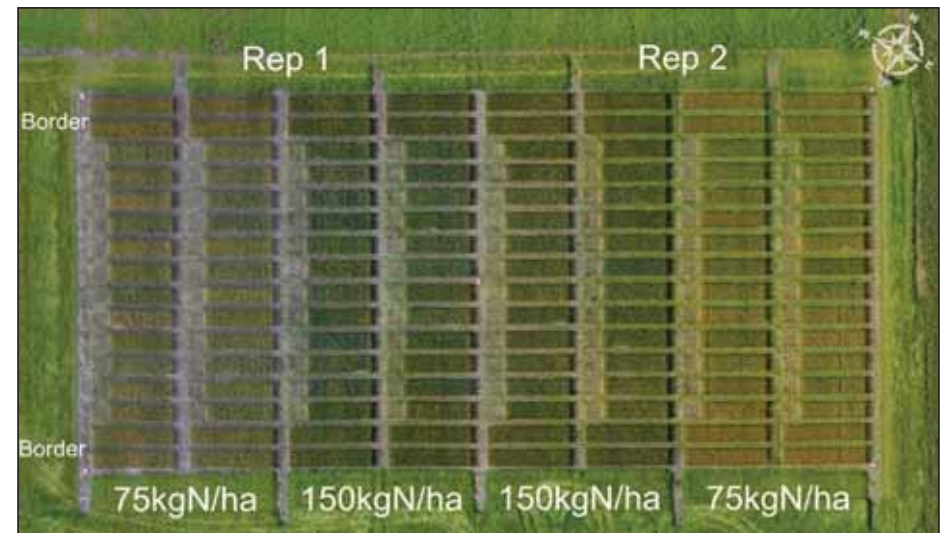
Micasense
RedEdge-M



	P4 Multispectral	Micasense RedEdge M
Red	650 ± 16 nm	668 ± 5 nm
Green	560 ± 16 nm	560 ± 10 nm
Blue	450 ± 16 nm	475 ± 10 nm
Red Edge	730 ± 16 nm	717 ± 5 nm
NIR	840 ± 26 nm	840 ± 20 nm

Trials and flights

- A collection of 24 historical and modern spring wheat cultivars planted in Ås, south-eastern Norway, 59°39'N, 10°45'E, in 2020 and 2021.
- 5 parallel mission flights:
 - Beginning of June – tillering stage
 - Last week of June – stem elongation
 - Beginning of July – heading stage
 - End of July – the onset of maturity
 - End of July – maturing progress



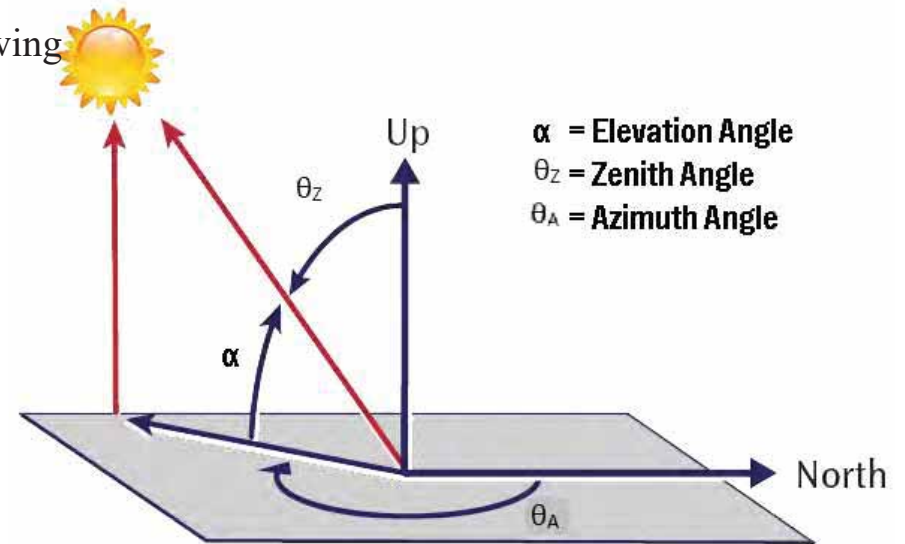
Field Overview

What is the impact of the sun's elevation angle?

To study the sun elevation angle effect:

Seven flights were conducted by both cameras at the following time points and corresponding solar elevation angles:

- 09:00 (28.59°),
- 10:00 (37.75°),
- 11:00 (42.04°),
- 12:00 (46.78°),
- 13:00 (49.23°),
- 14 (48.92°),
- 15 (45.89°)

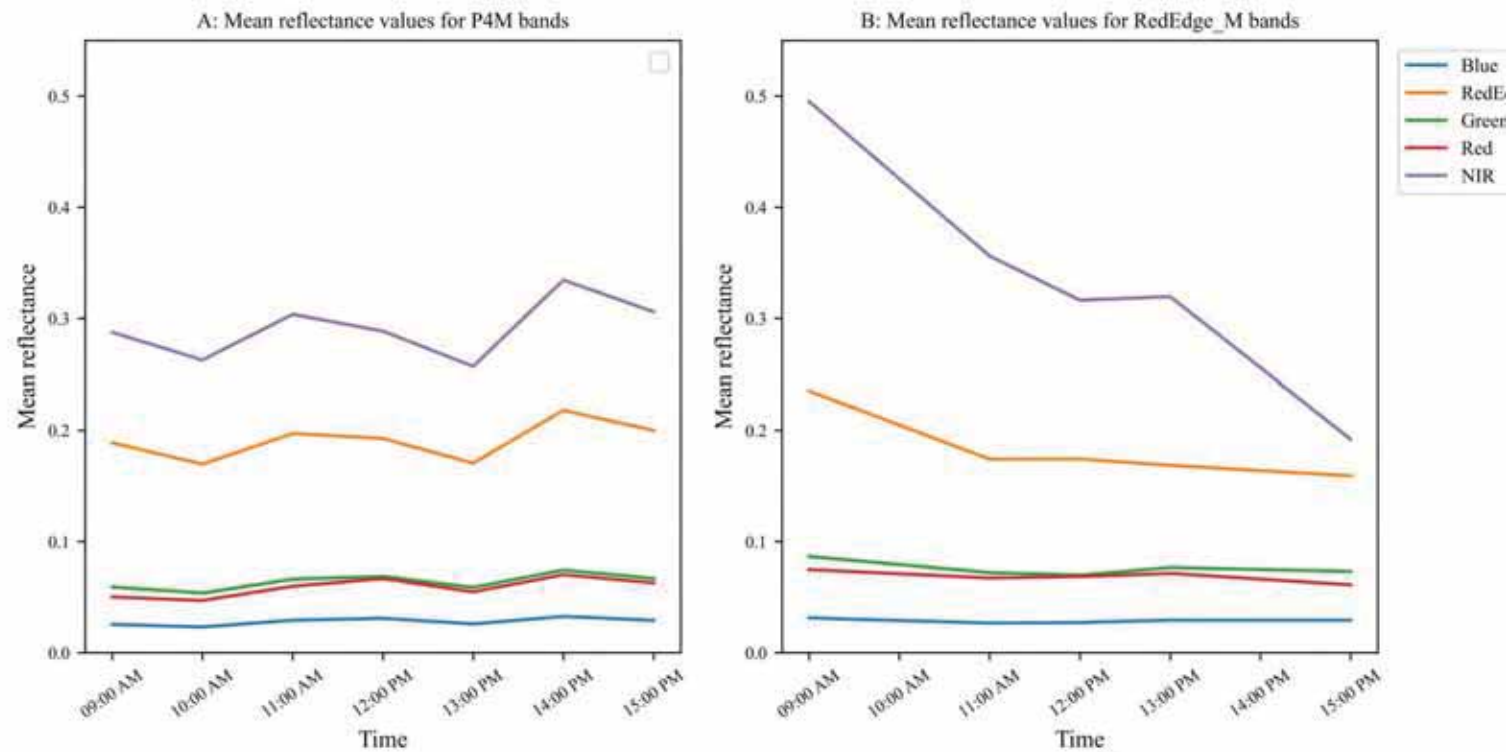


Trait prediction using two models

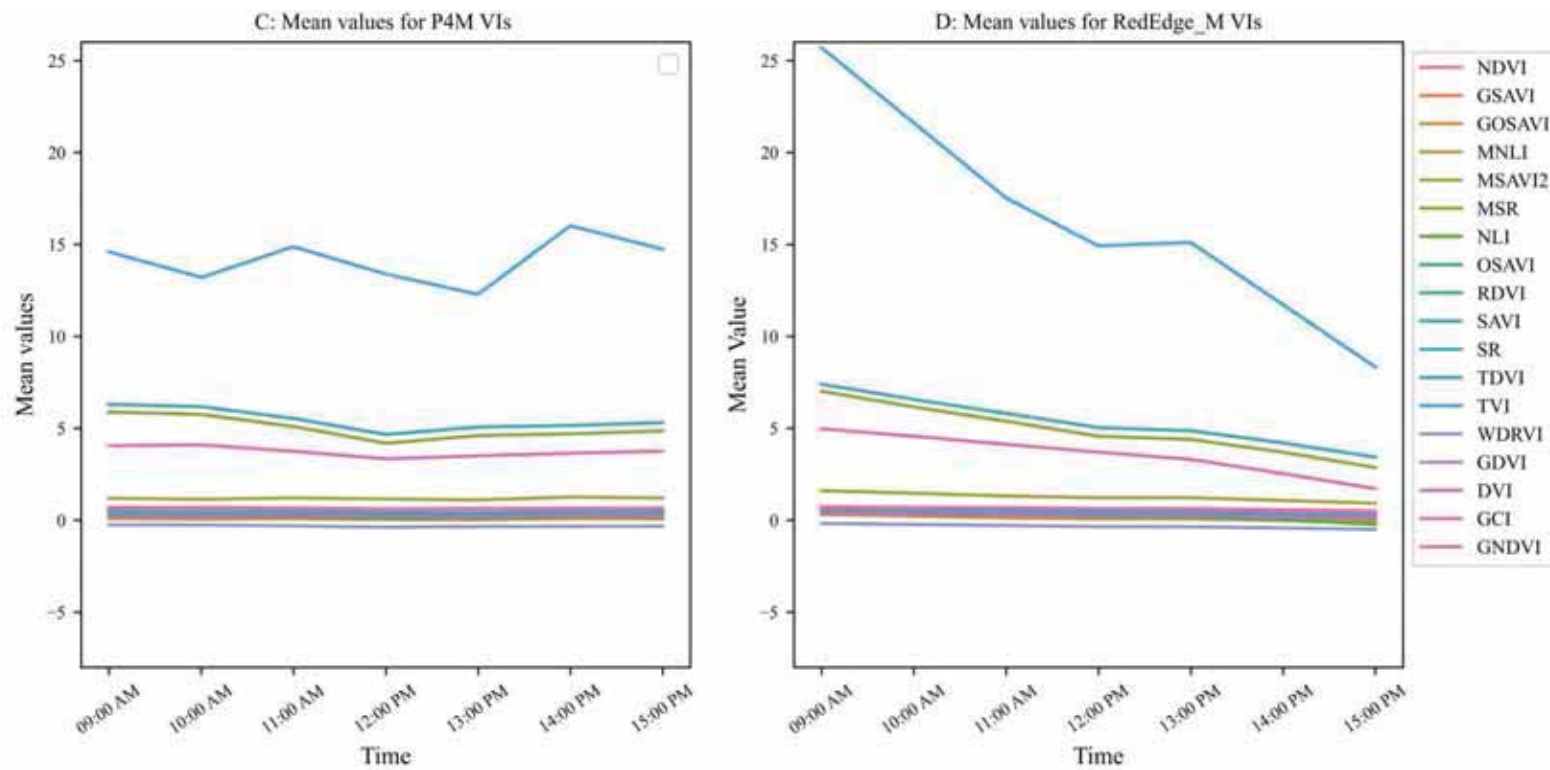
The overall between-camera differences for the corresponding variables were estimated using the mean difference (MD).

$$MD_i = \frac{\sum_1^j (Var_{ij}^{S1} - Var_{ij}^{S2})}{j}$$

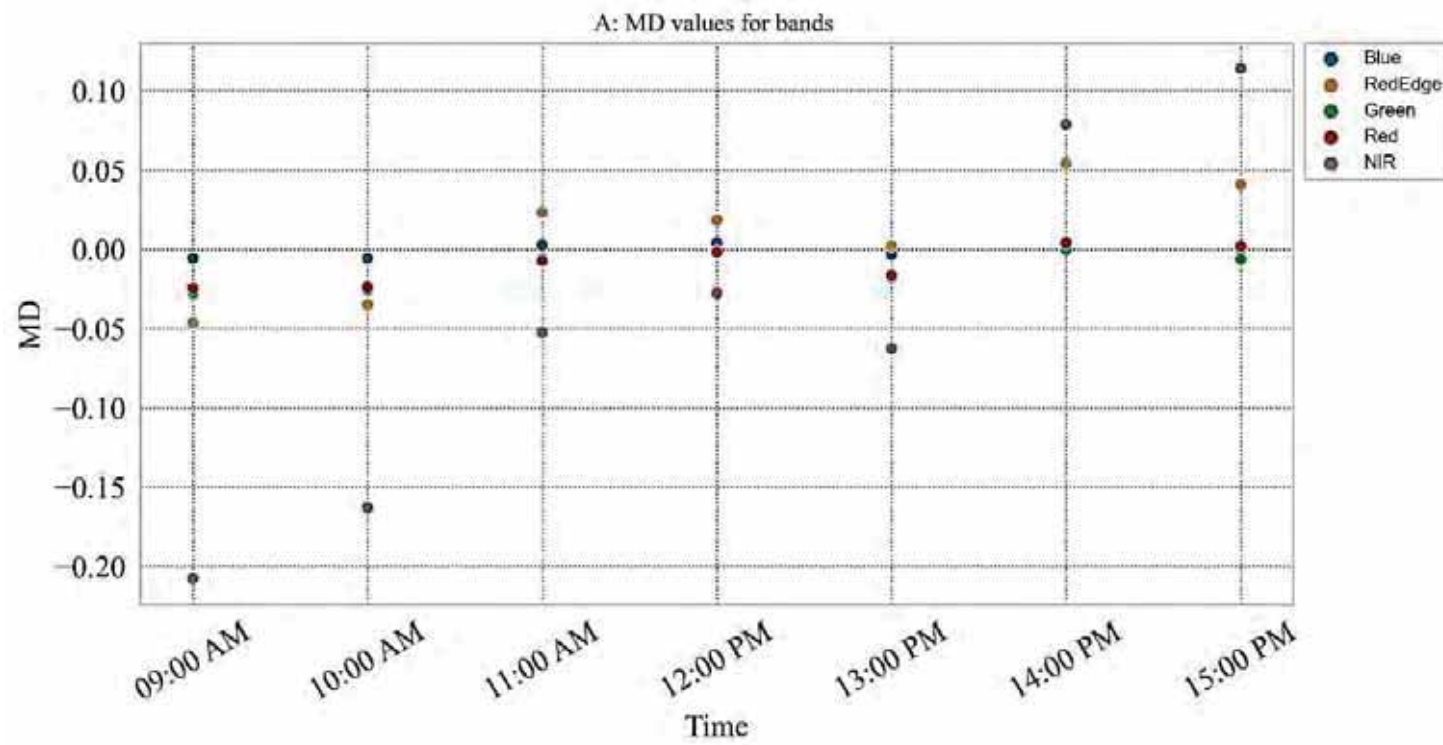
Where Var_{ij}^{S1} and Var_{ij}^{S2} are corresponding values of sample j ($j=1,2, 3, \dots, n$) for the variable i for cameras S1 and S2 respectively.



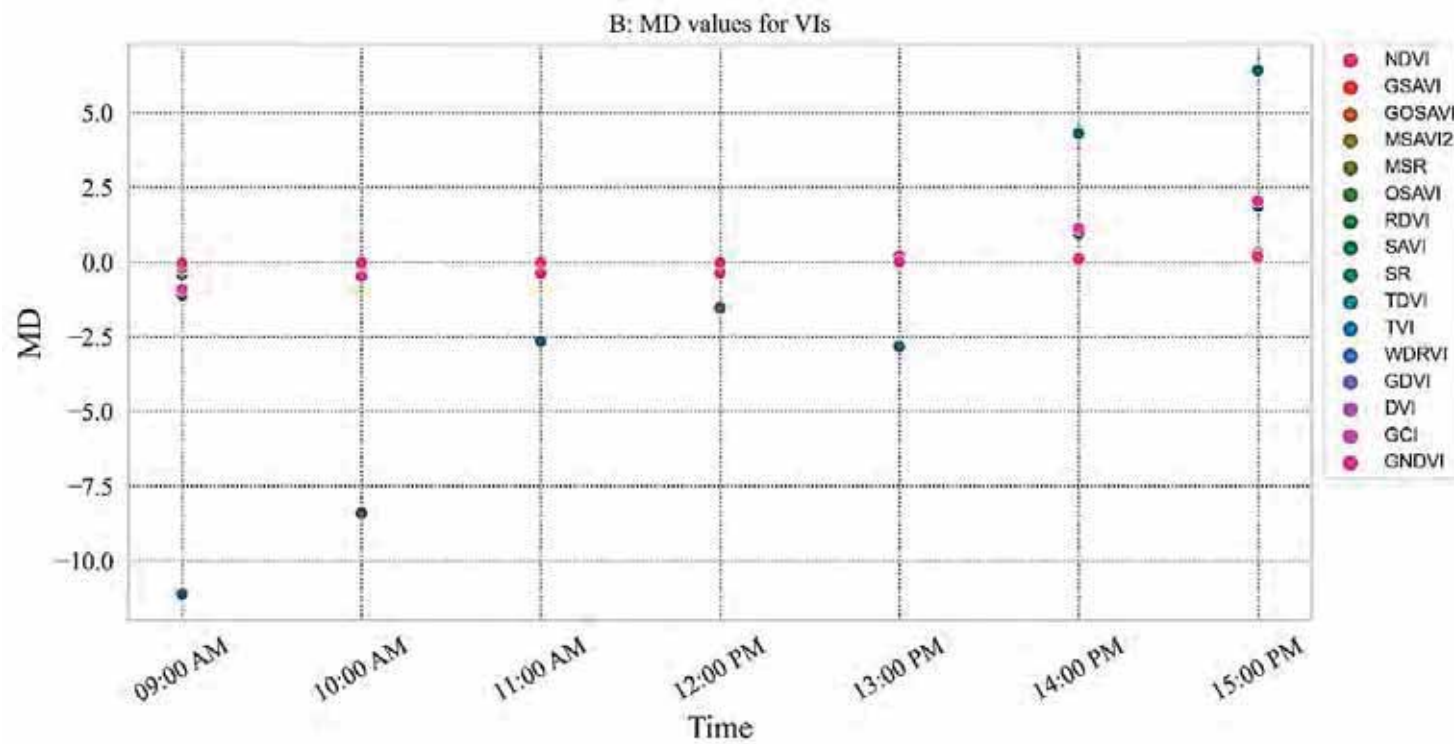
Average band reflectance (A, B) values during the day generated from both cameras



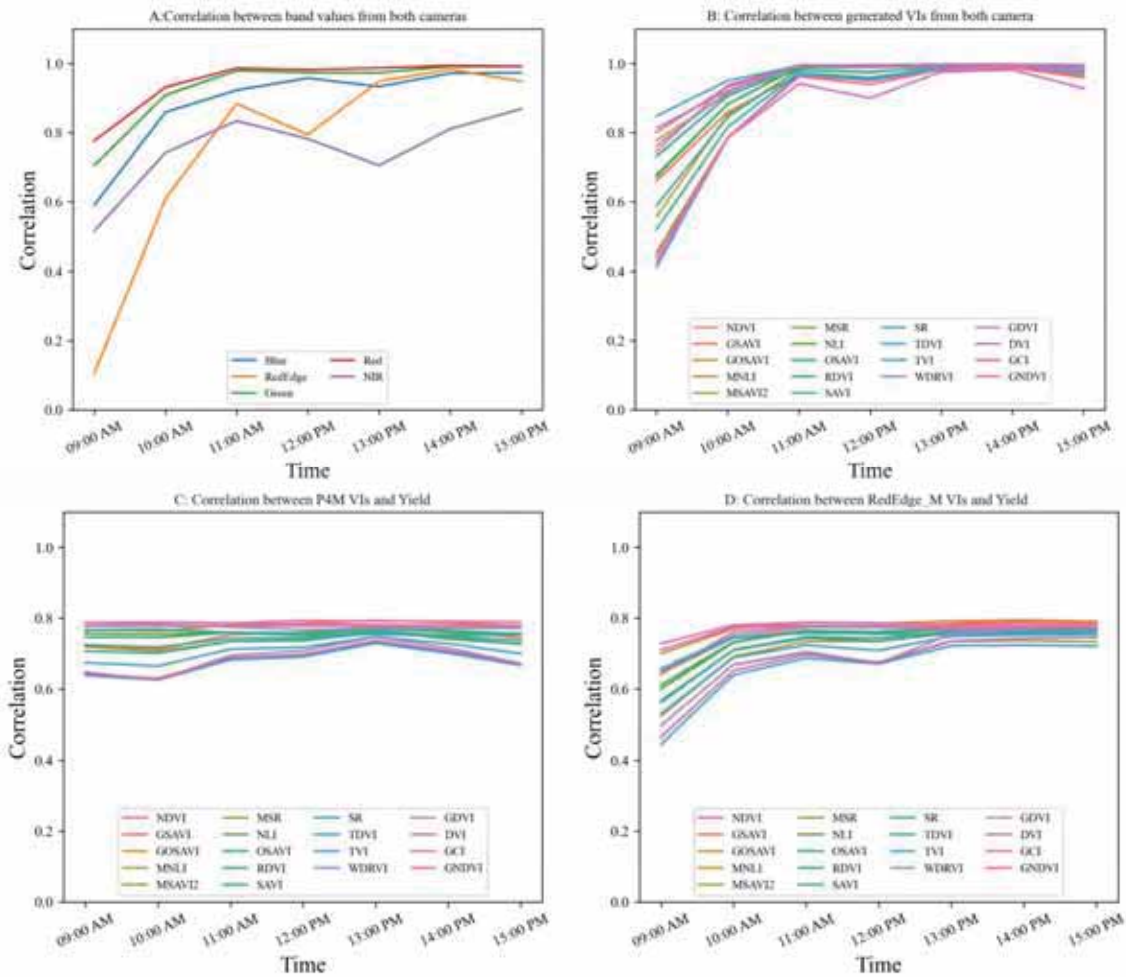
Average band reflectance VIs (C, D) values during the day generated from both cameras



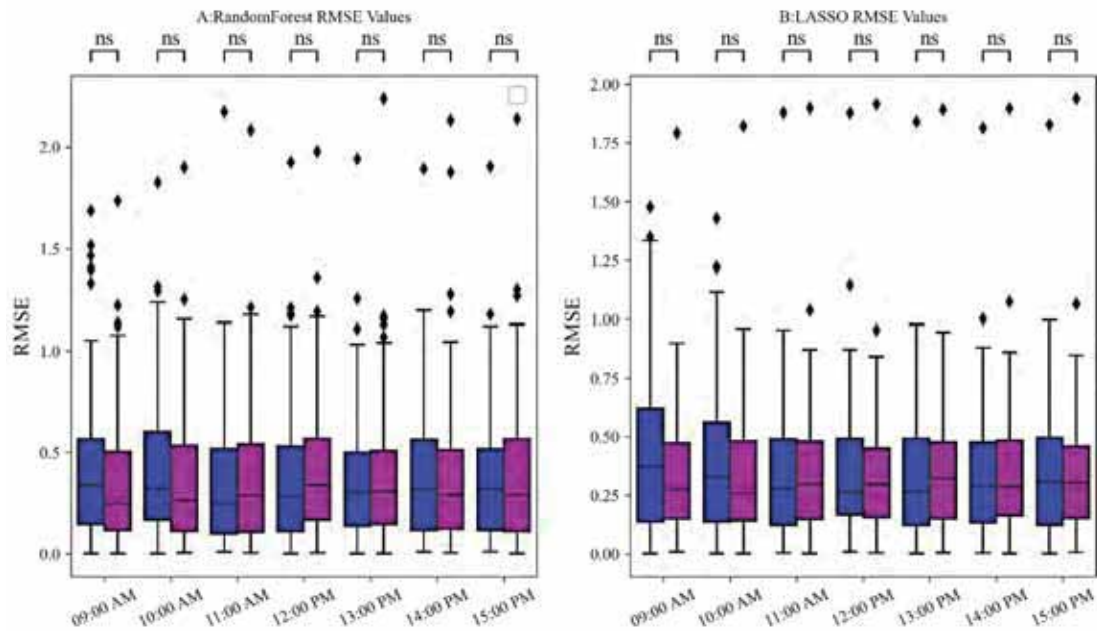
Mean difference (MD) values of captured bands by both cameras A) Bands B) VIs



Mean difference (MD) values of captured bands by both cameras VIs

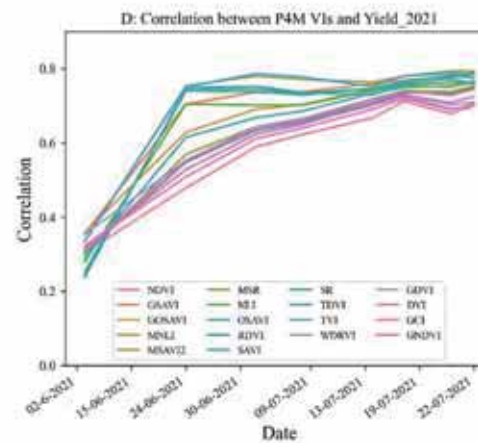
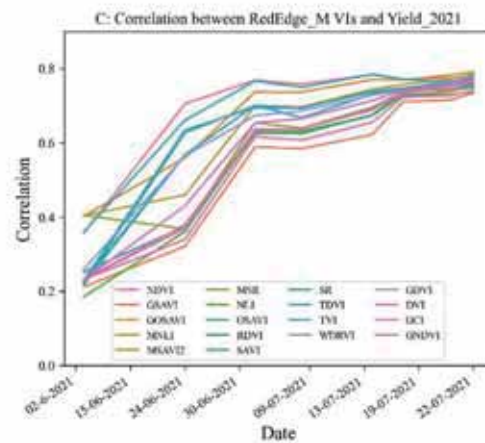
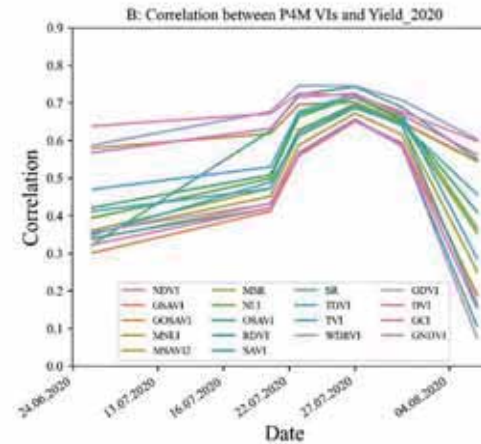
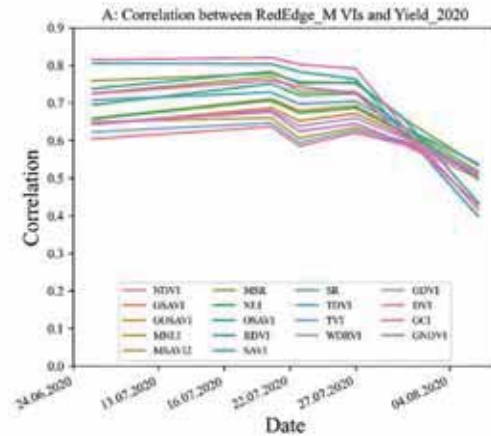


Correlation between band values and VIs generated from both cameras (A, B), the correlation between VIs by both cameras, and yield (C, D).

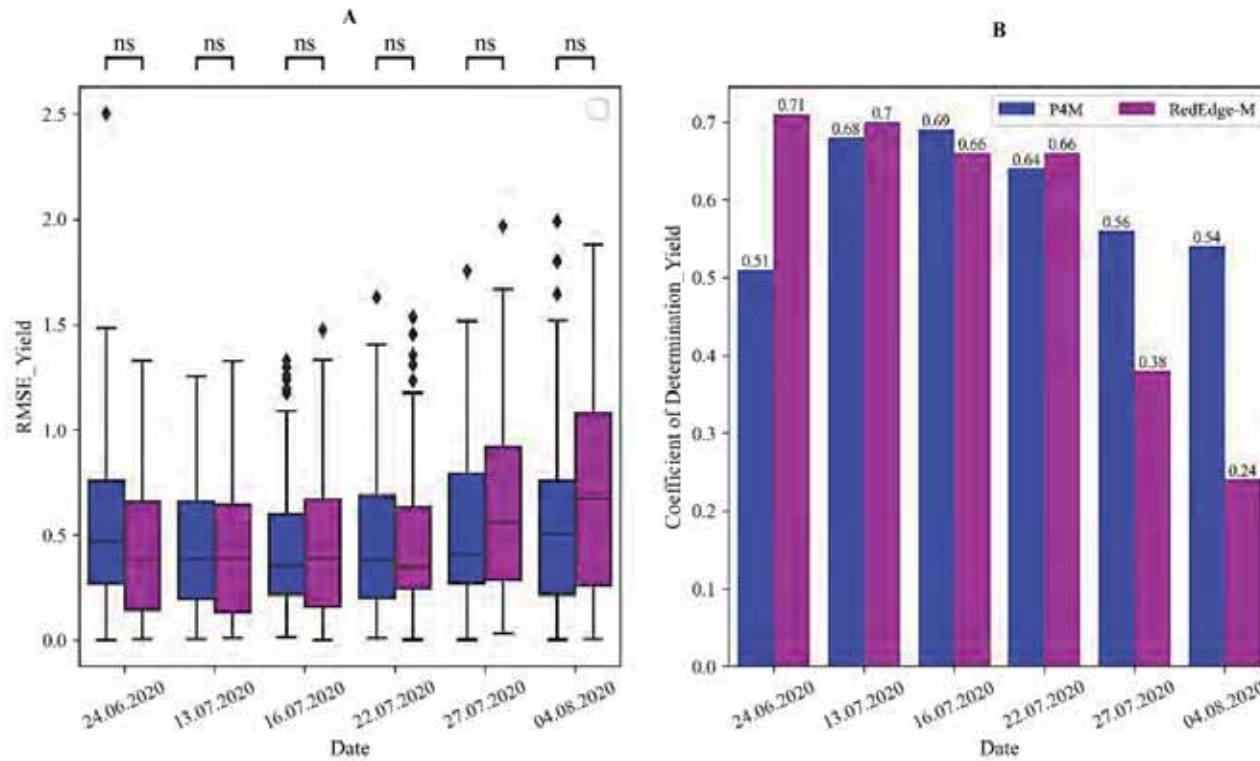


Time	LASSO		RF	
	P4M	RedEdge-M	P4M	RedEdge-M
09:00 AM	0.63	0.45	0.59	0.44
10:00 AM	0.66	0.54	0.57	0.47
11:00 AM	0.64	0.65	0.57	0.56
12:00 PM	0.64	0.65	0.51	0.52
13:00 PM	0.65	0.65	0.53	0.56
14:00 PM	0.65	0.65	0.51	0.56
15:00 PM	0.64	0.65	0.52	0.56

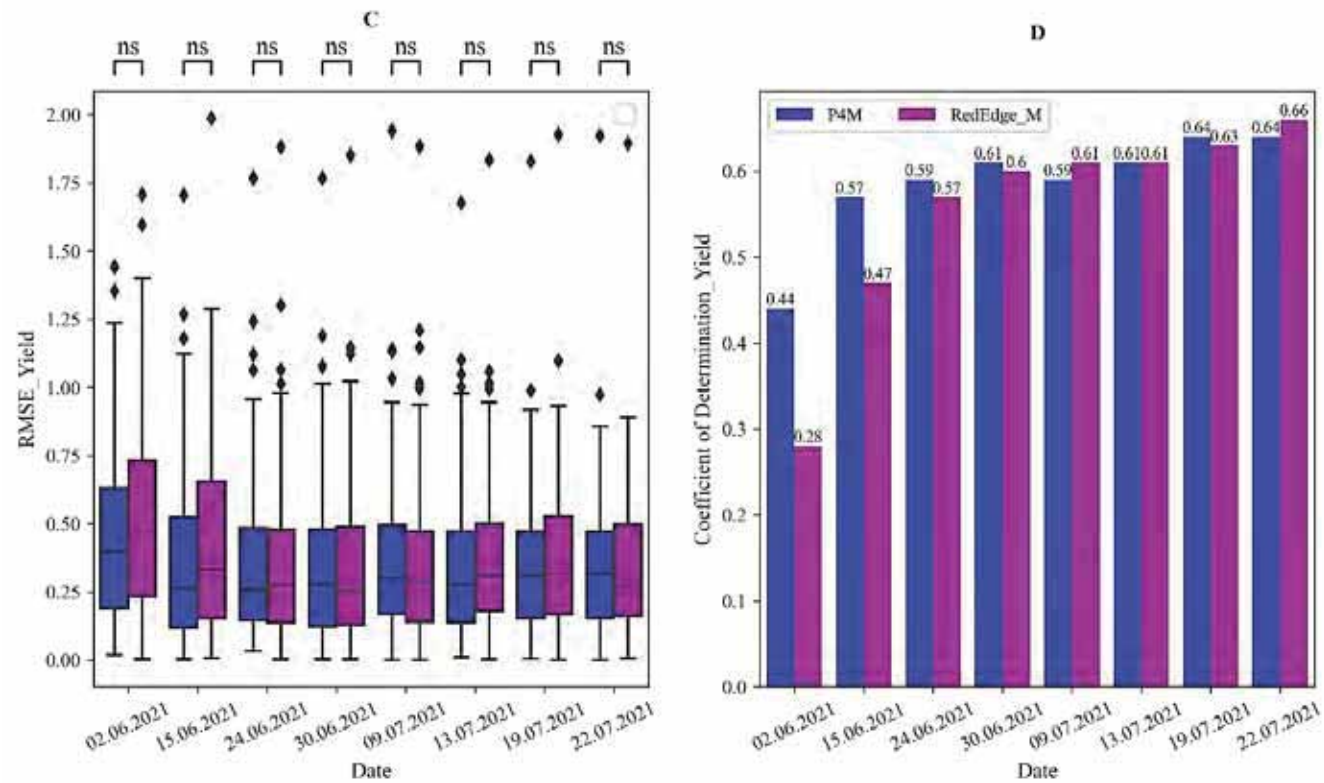
Models' performance and t-test results to compare the two cameras during the course of the day



Pearson correlation between yield and generated VIs by both cameras A, B (2020), C, D (2021)



Model performance and t-test results to compare the two cameras during the 2020 season



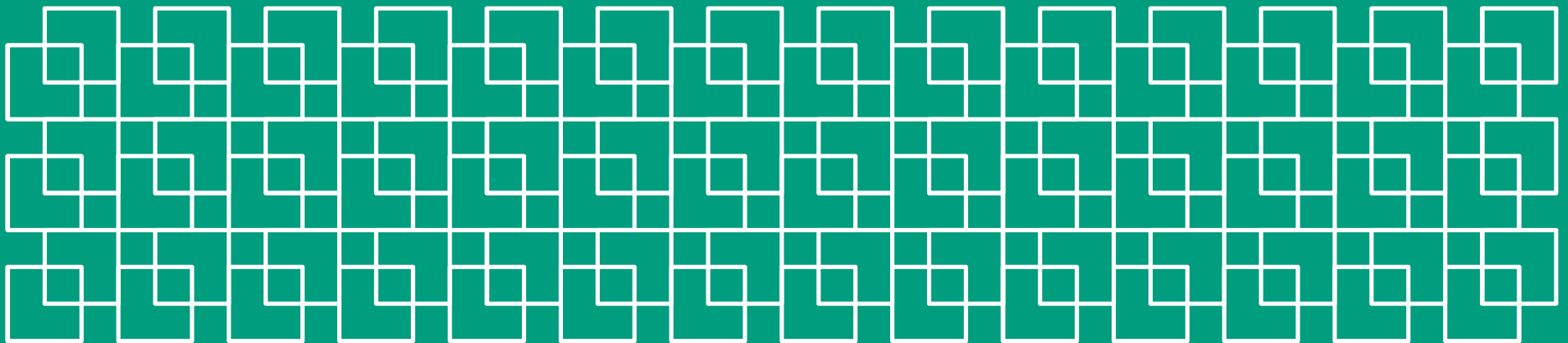
Model performance and t-test results to compare the two cameras during the 2021 season

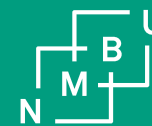
Conclusion:

- The reflectance values changed with changes in the sun elevation angle.
- The reflectance values measured by both cameras are most correlated around local noon when the sun angle is highest.
- Some vegetation indices measured by P4M multispectral camera can compensate for the sun elevation angle effect and generate more consistent values during the day.
- The research results showed that the P4M camera has a better performance and consistency to generate different vegetation indices for application in plant phenotyping and precision agriculture.
- In overall, the difference between both sensors wasn't significant for trait prediction.



Thank you
sahameh.shafiee@nmbu.no



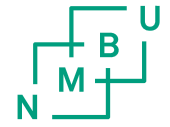


SPECTRAL IMAGING IN CROP FIELDS – RESEARCH ACTIVITIES AT NMBU

Ingunn Burud, Sahameh Shafiee, Tomasz Mroz, Morten Lillemo
Norwegian University of Life Sciences

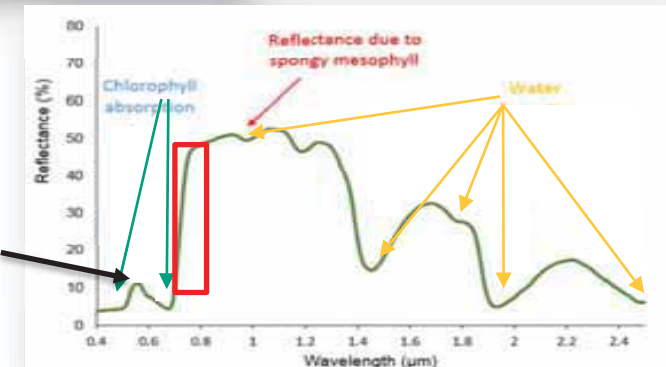
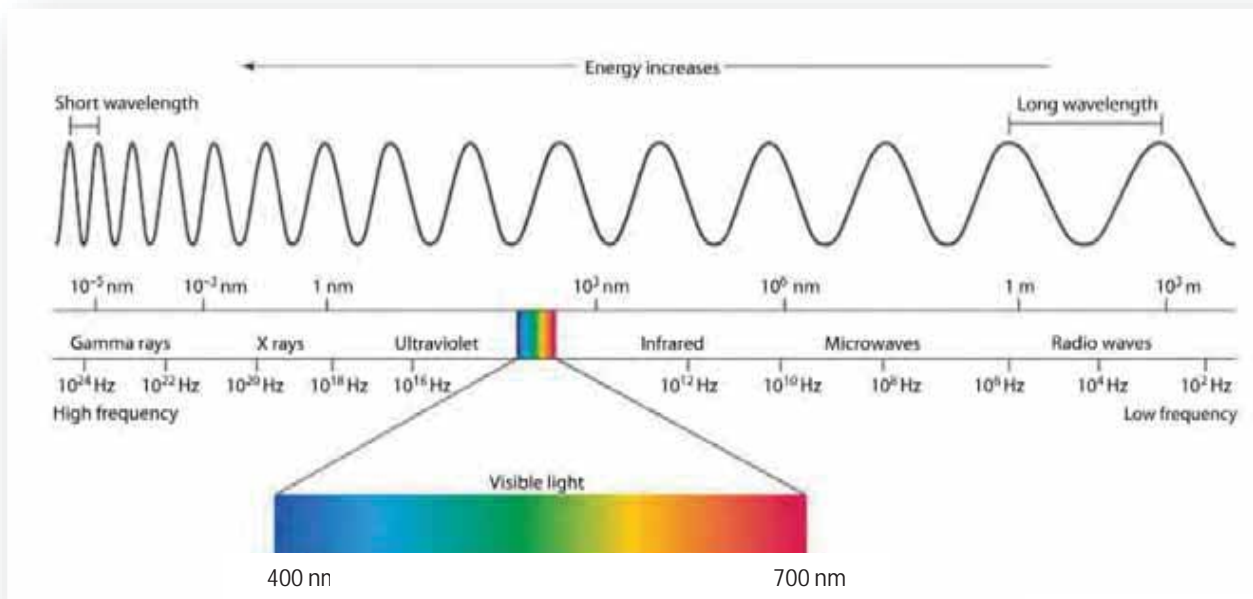


NIR imaging of the Earth



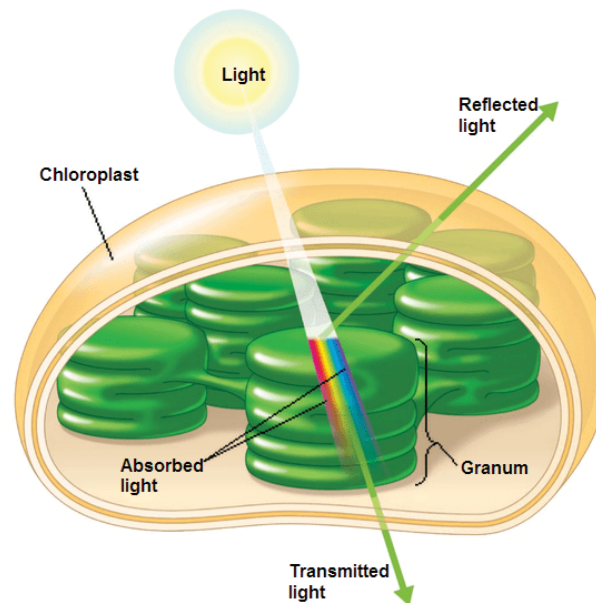
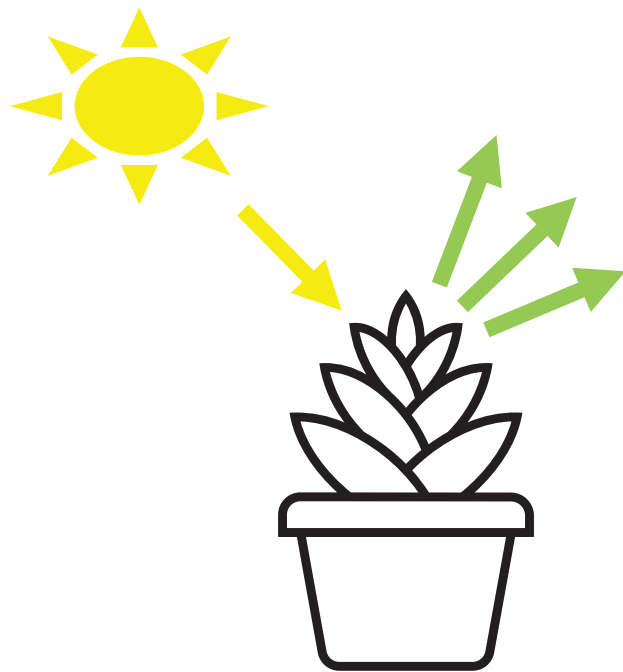
$$\text{NDVI} = (\text{NIR} - \text{R}) / (\text{NIR} + \text{R})$$

The electromagnetic spectrum

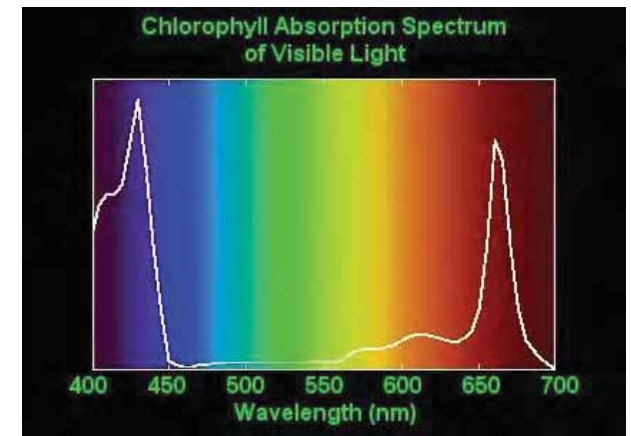


Red edge:
scattering due to leaf internal structure in the NIR region

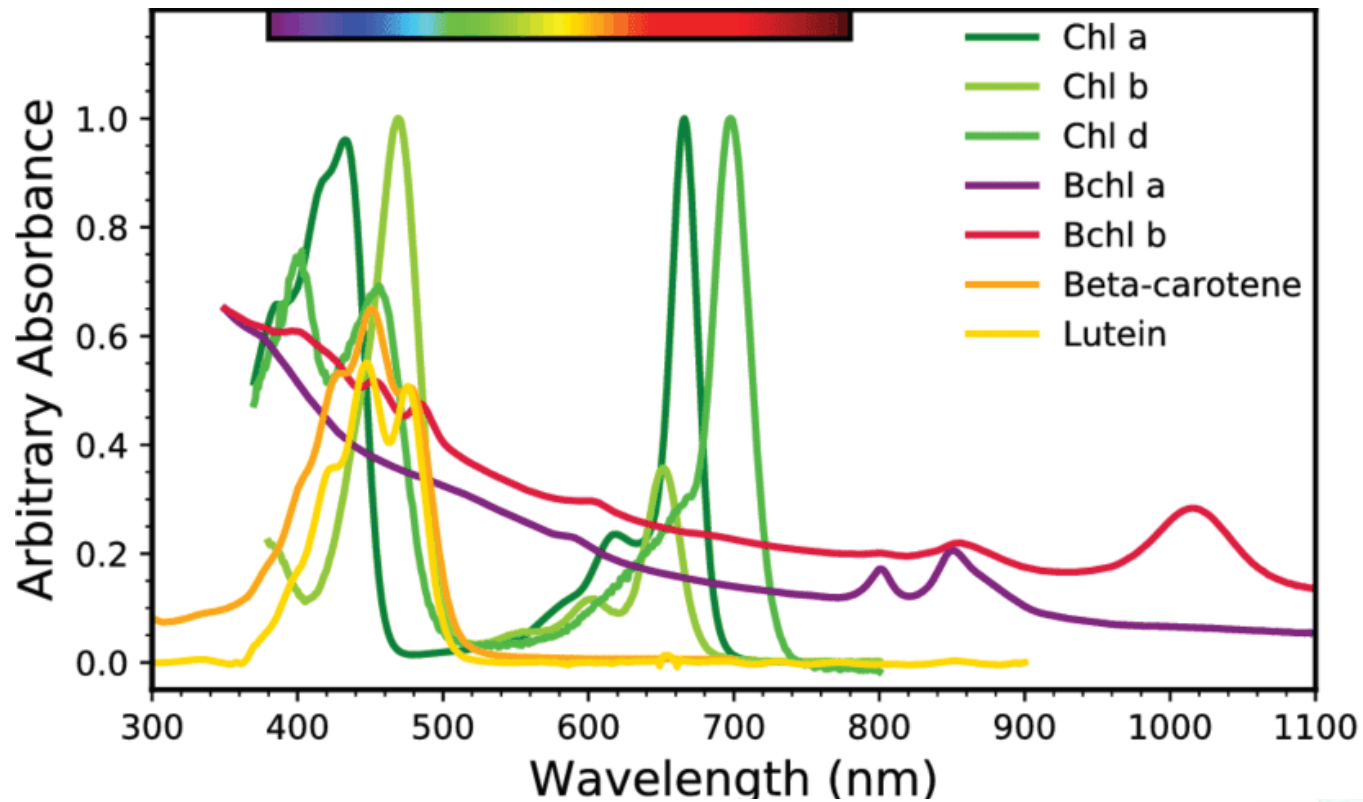
Plant Reaction to Stress



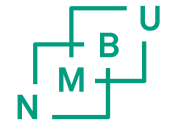
Copyright © 2005 Pearson Education, Inc. Publishing as Pearson Benjamin Cummings. All rights reserved.



Leaf Spectra



RGB images



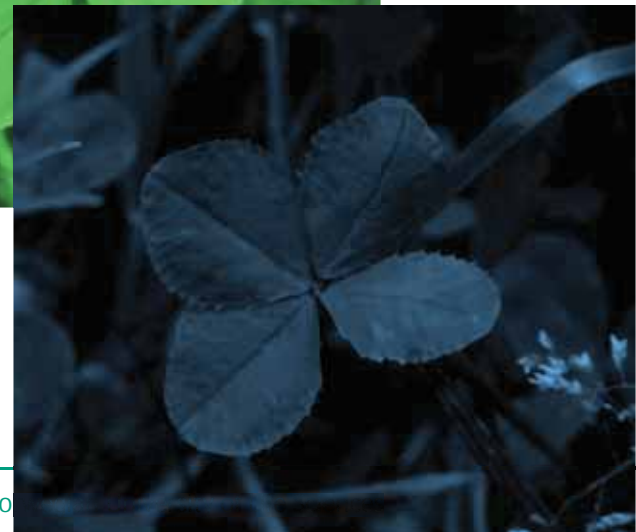
Red ≈ 645 nm



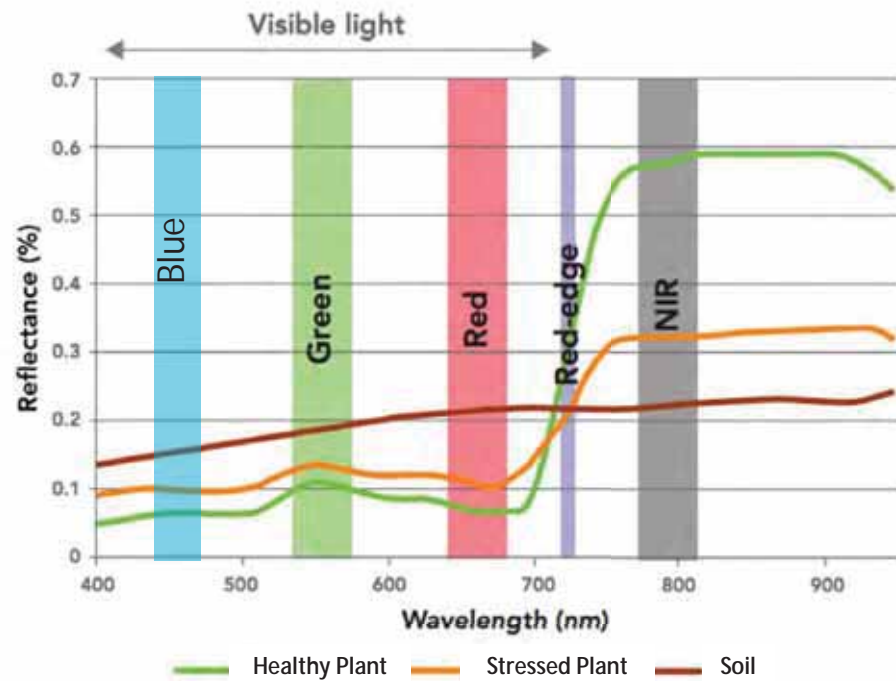
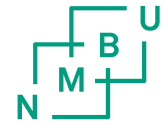
Green ≈ 510 nm



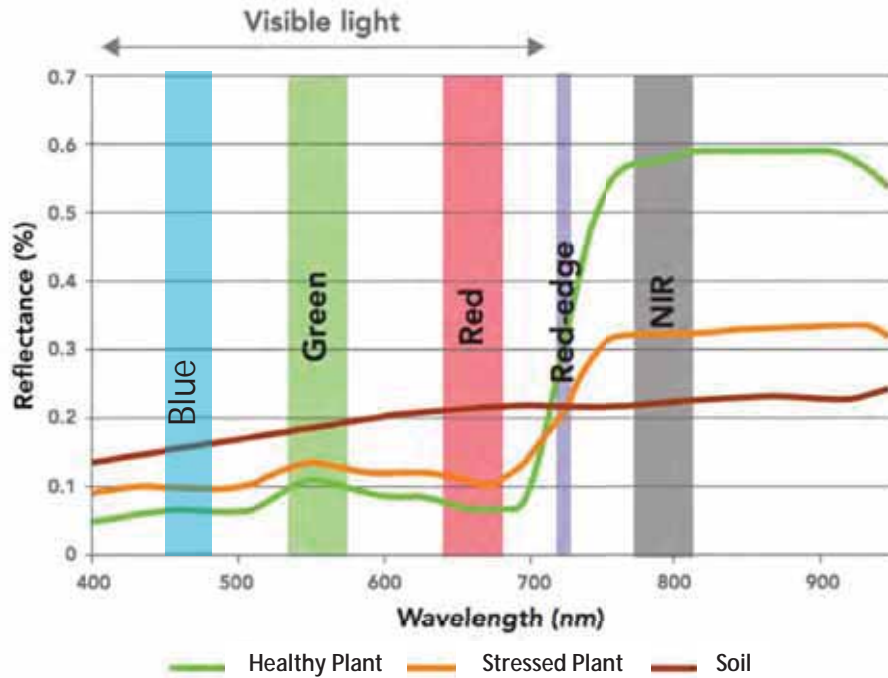
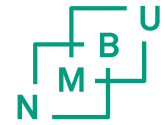
Blue ≈ 400 nm



Multispectral images



Vegetation indices

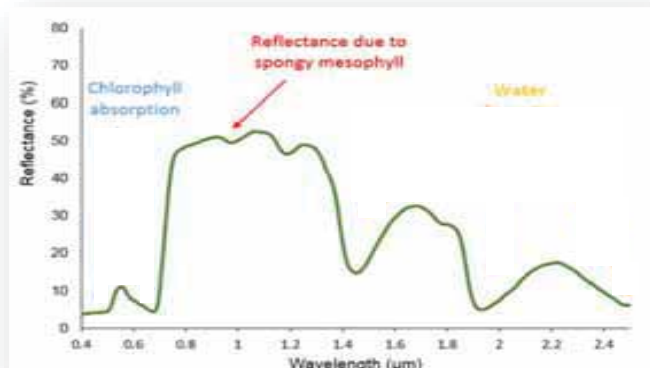
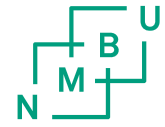


$$NDVI = \frac{NIR - Red}{NIR + Red}$$

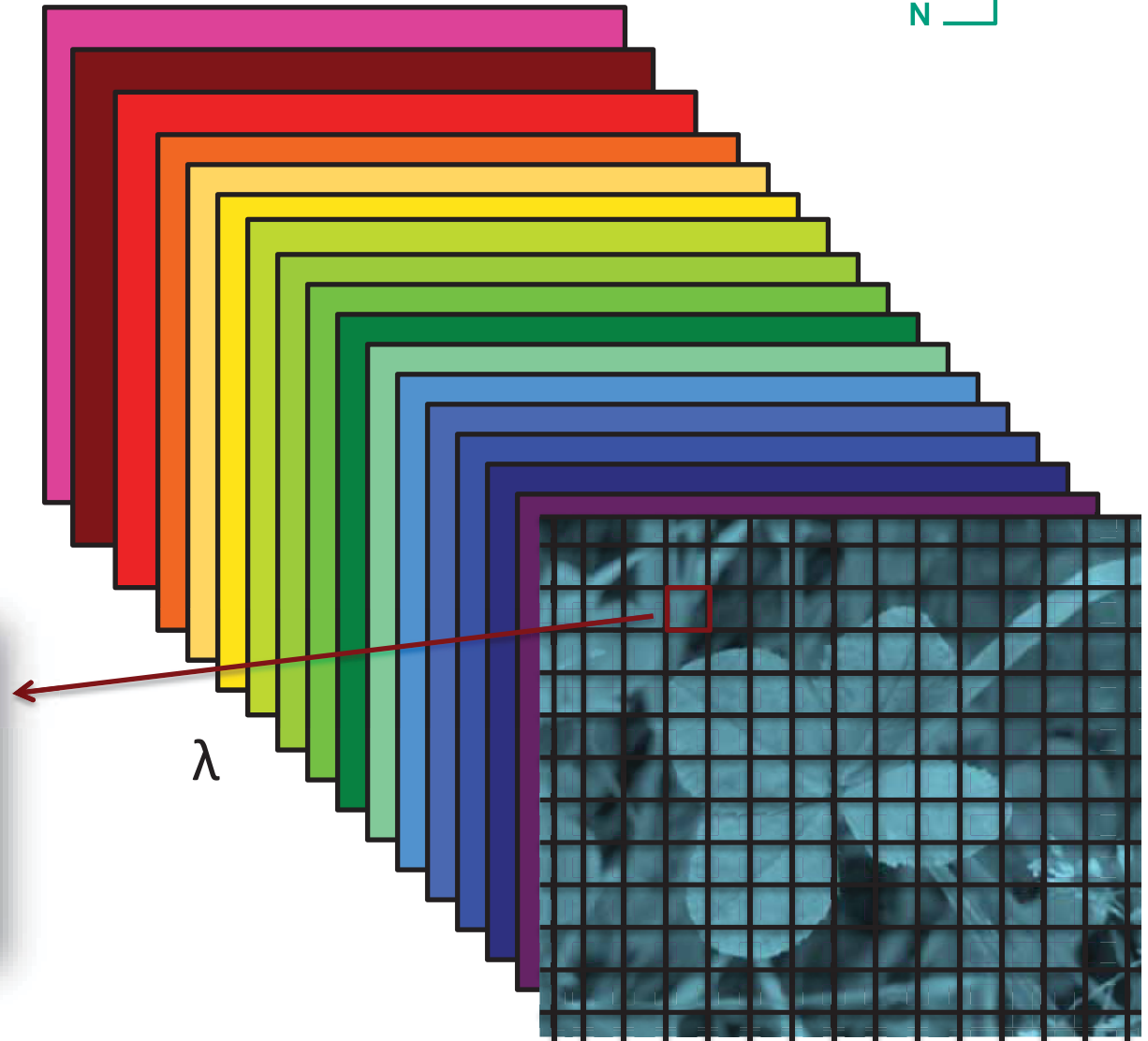
$$MTCI = \frac{NIR - Rededge}{Rededge - Red}$$

$$EVI = 2.5 * \frac{(NIR - Rededge)}{(NIR + 6 * Red - 7.5 * Blue + 1)}$$

Hyperspectral imaging



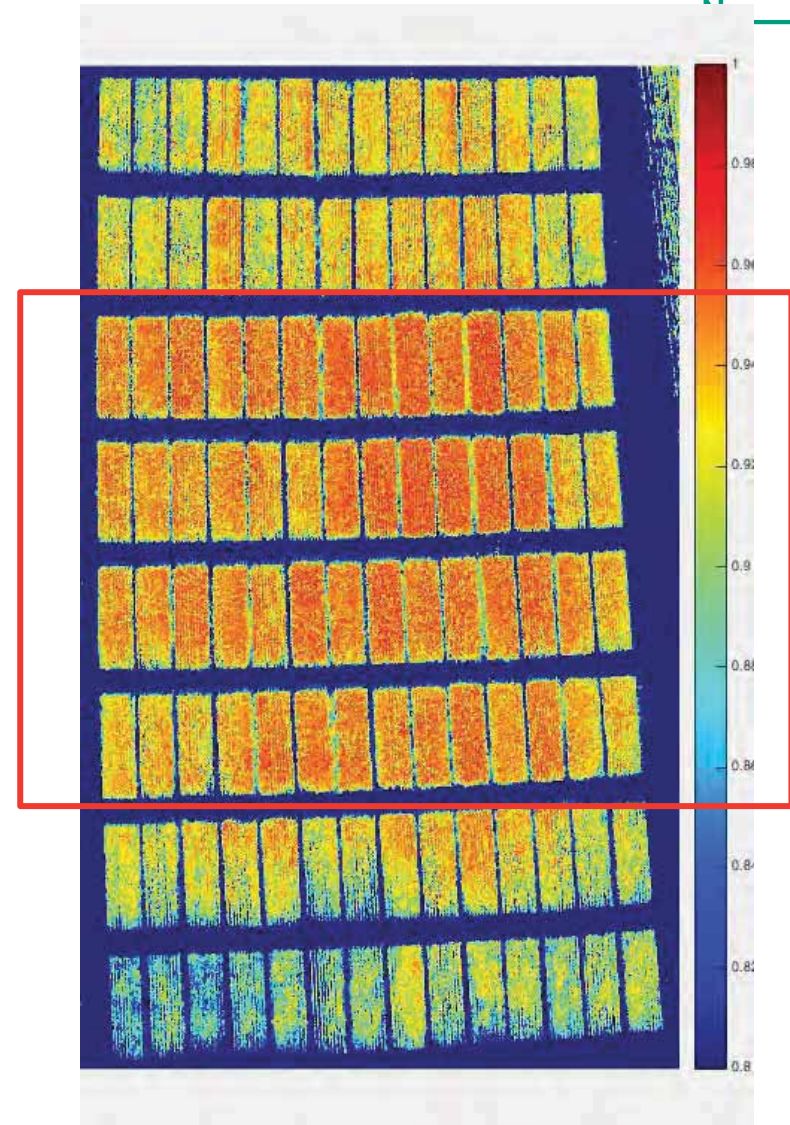
Full spectrum in each pixel



Phenotyping of wheat

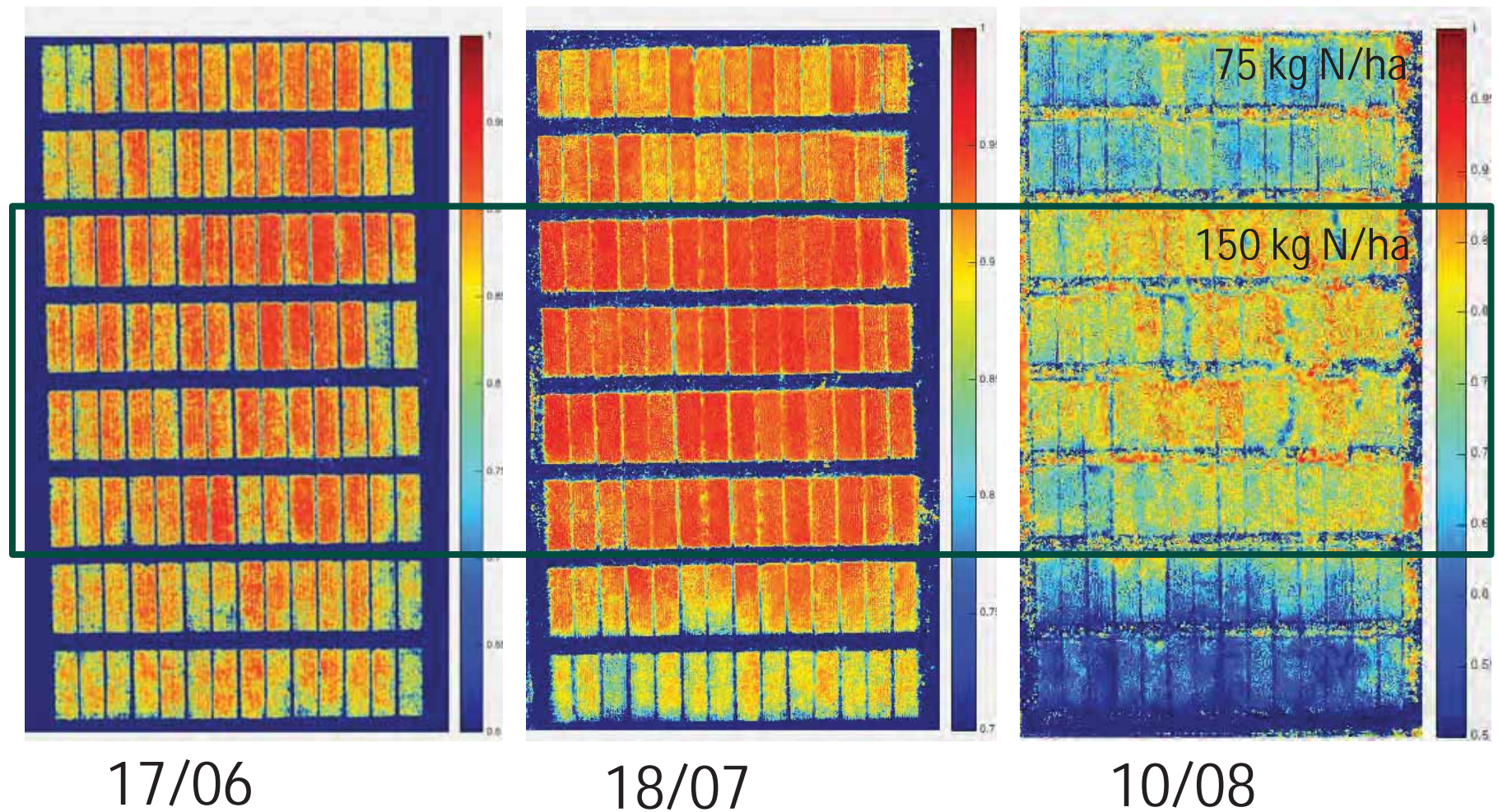


Chlorophyll production

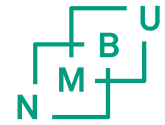


- Plant cover
- Stress
- Diseases
- Fertilizers

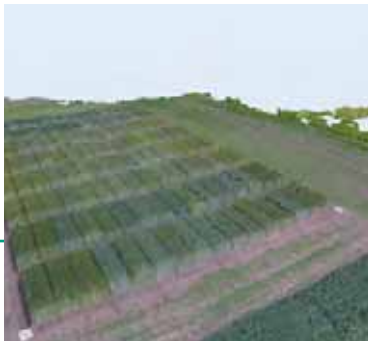
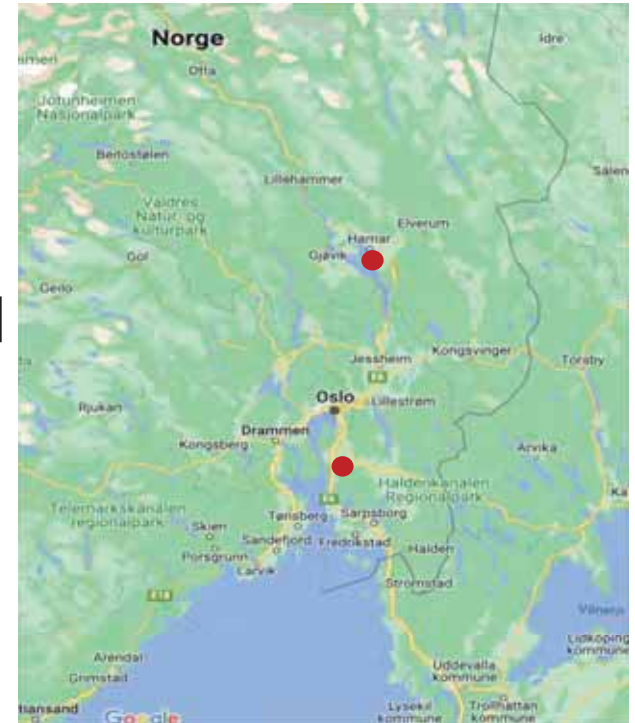
NDVI maps



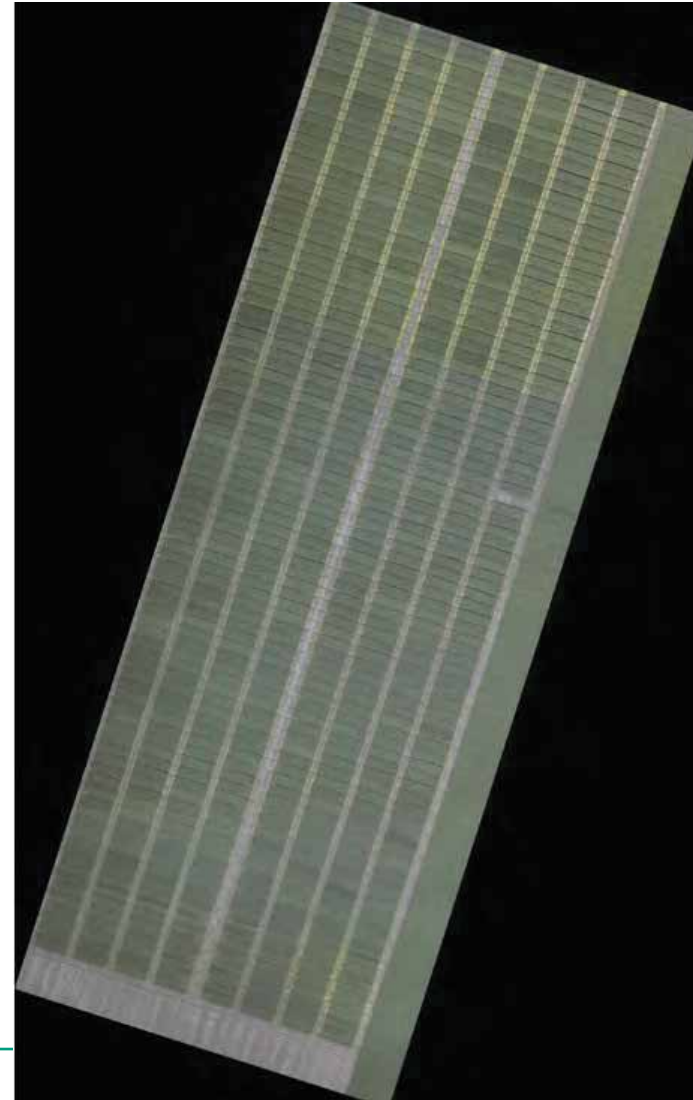
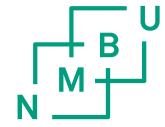
Phenotyping experiment



- Field trials at two locations in Norway:
Ås and Stange
- 300 spring wheat lines, 396 cultivars
- Measurements of agronomic traits and grain yield
- Multispectral images from UAV

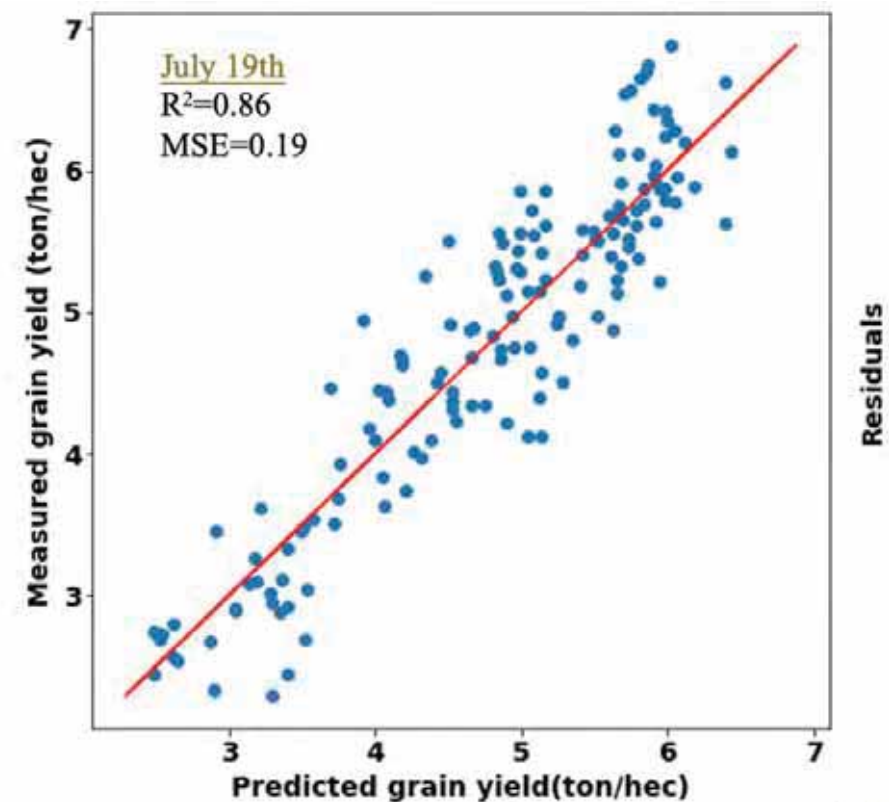


Phenotyping experiment



Yield prediction

- Machine learning was applied for yield prediction (Sequential Forward Selection and LASSO)
- [NDVI](#) showed good prediction ability of grain yield in final grain filling stages.
- MTCI and EVI are more important grain yield predictors during early grain filling
- Best predictions resulted from combining vegetation indices from multitemporal data



Computers and Electronics in Agriculture
 Volume 182, April 2021, 106036

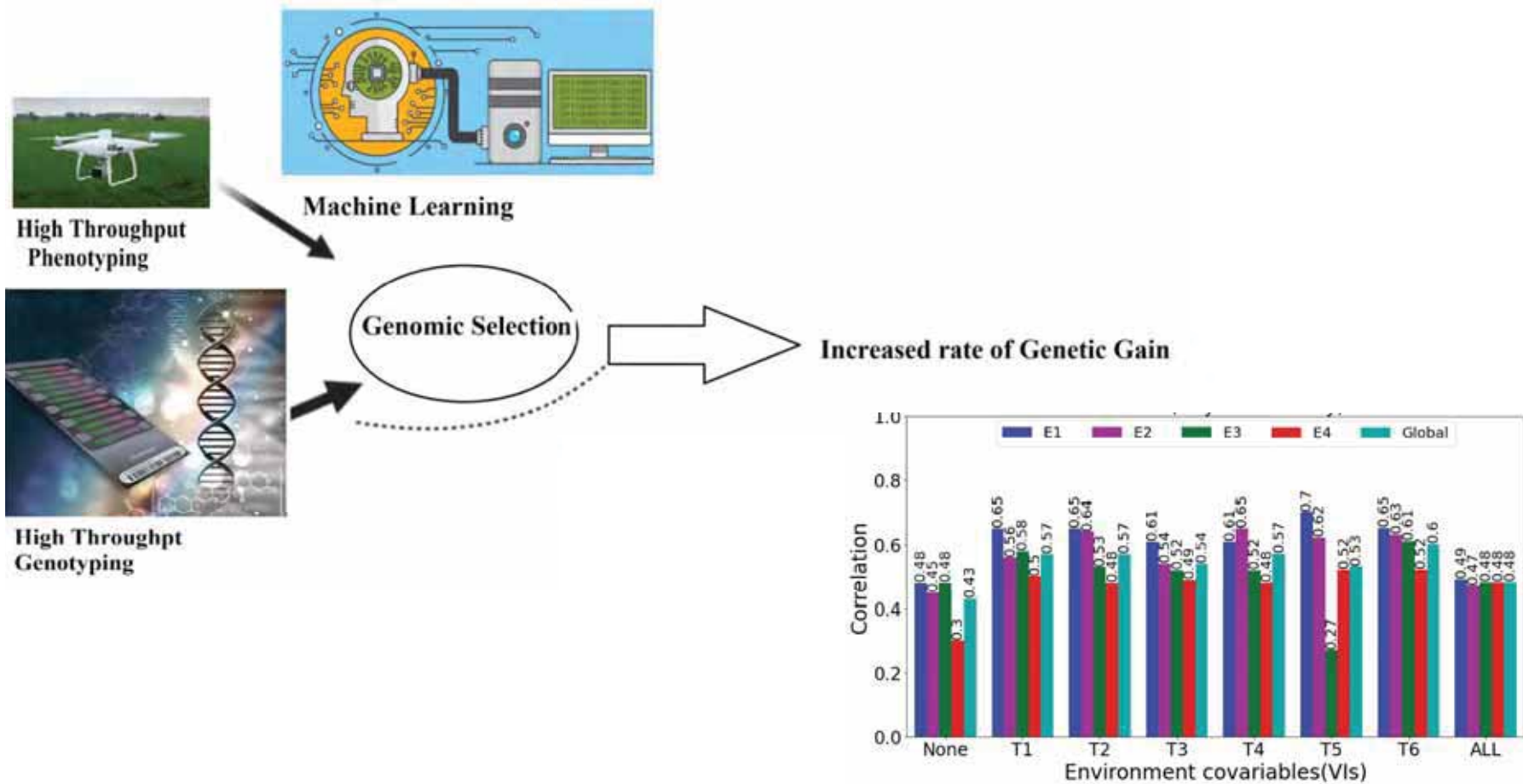
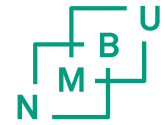


Original paper

Sequential forward selection and support vector regression in comparison to LASSO regression for spring wheat yield prediction based on UAV imagery

Saharwan Shafiq^{a,*}, Lian-Ming Liu^b, Jaganath Suresh^c, Jin-Ahn Doh^d, Muzib-Allah^e, P. Manoj
^aDepartment of Agriculture, University of Agriculture, Faisalabad, Pakistan

Genomic prediction



Comparison of multispectral cameras

- In 2020 and 2021 we performed weekly parallel testing of the P4M and the Micasence camera
- P4M camera was more stable than Micasence for the derived vegetation indices and less sensitive to sun angles
- P4M seems particularly well suited for phenotyping

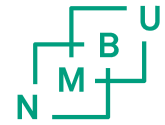


P4 Multispectral DJI

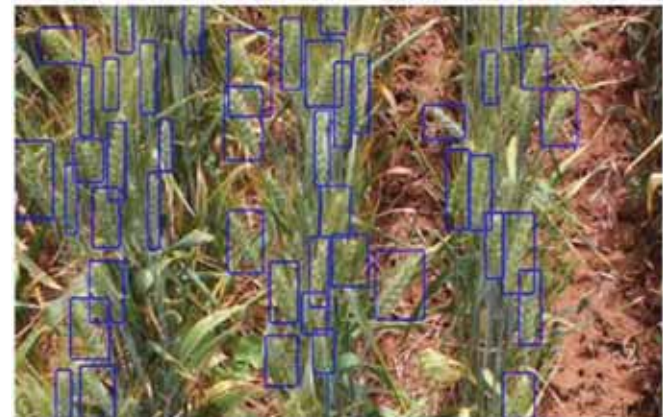


Micasense RedEdge-M

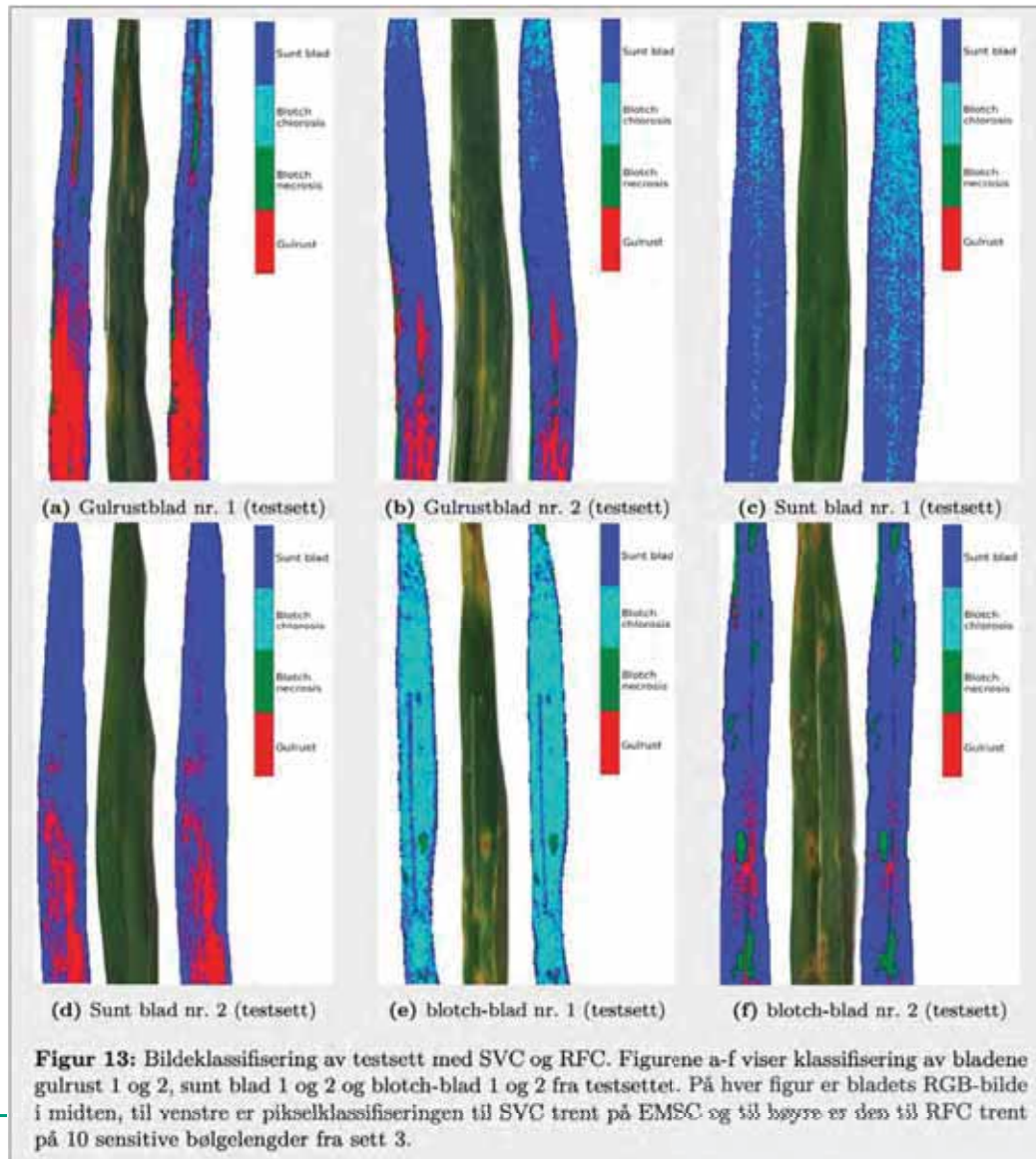
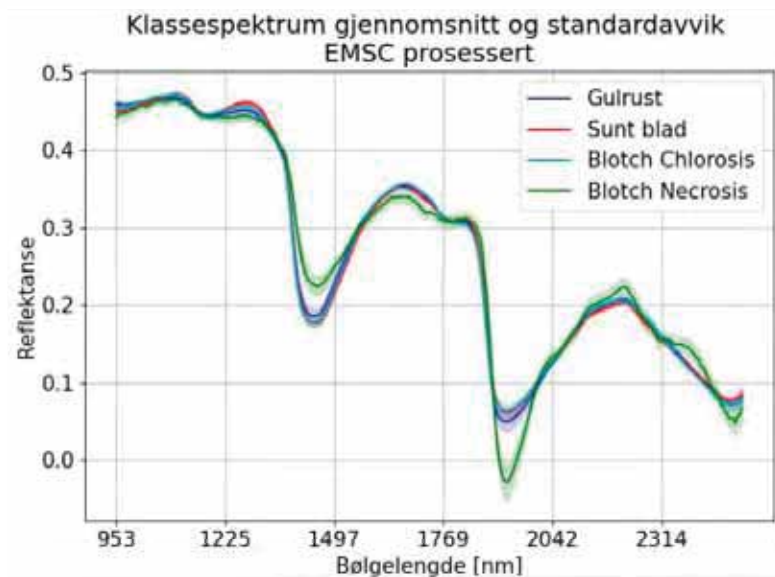
Robot images



- Number of spikes is related to yield
- Automatic spike detection from robot images
- Deep learning: annotating, annotating, annotating,...
- Kaggle competition
(<https://www.kaggle.com/c/global-wheat-detection>)

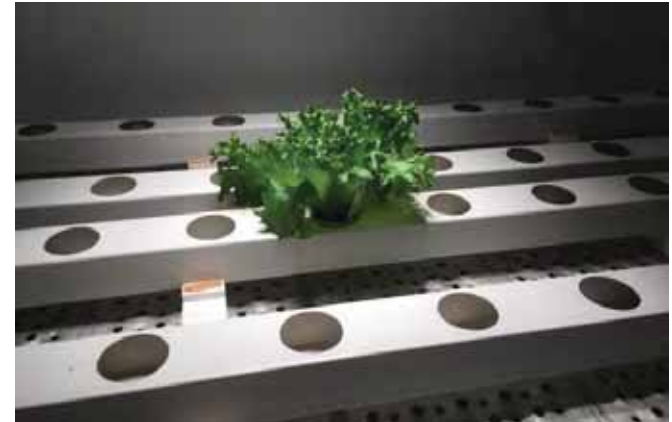


Hyperspectral imaging – wheat rust

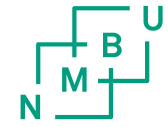


Hyperspectral imaging

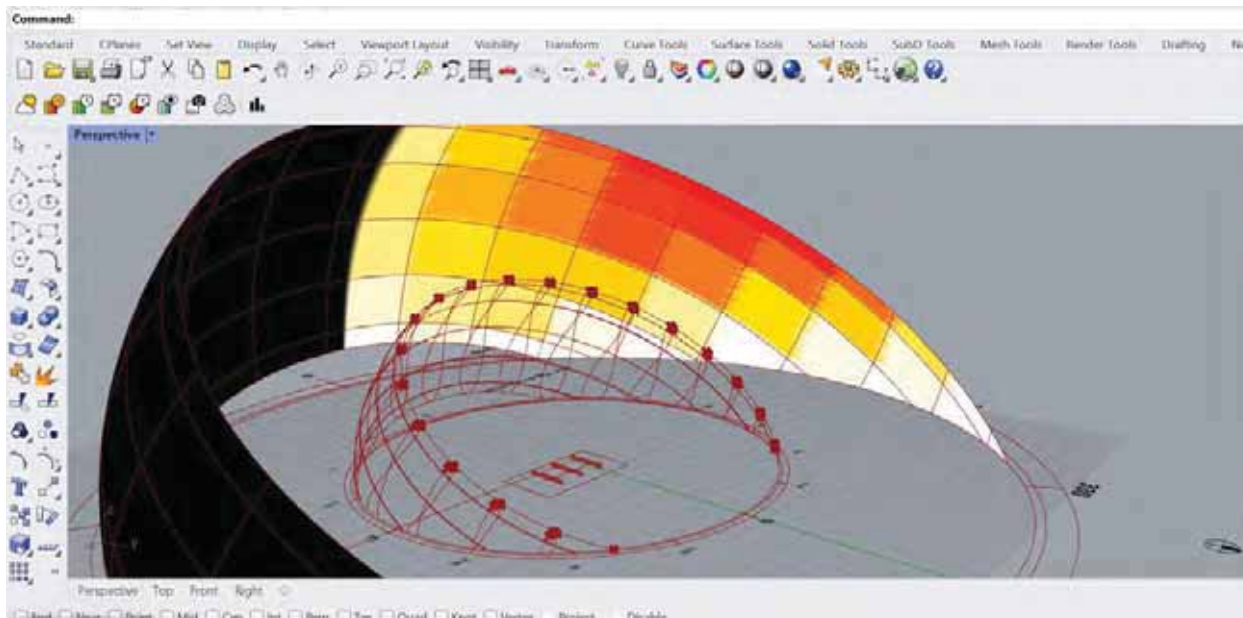
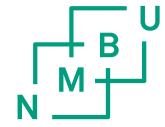
Detecting tipburn in lettuce



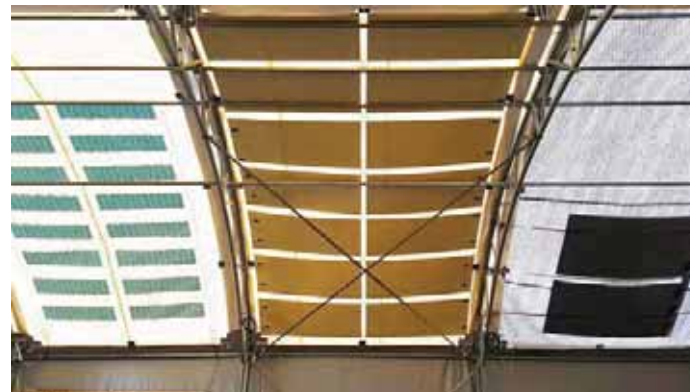
AgriPV – combining agriculture and solar energy production



AgriPV – SunFarming at NMBU



AgriPV – SunFarming at NMBU







NIBIO

NORSK INSTITUTT FOR
BIOØKONOMI

Exploring solar energy combined with a fleet of electrical vehicles and precision agriculture for reduced GHG-emissions (SolarFarm)

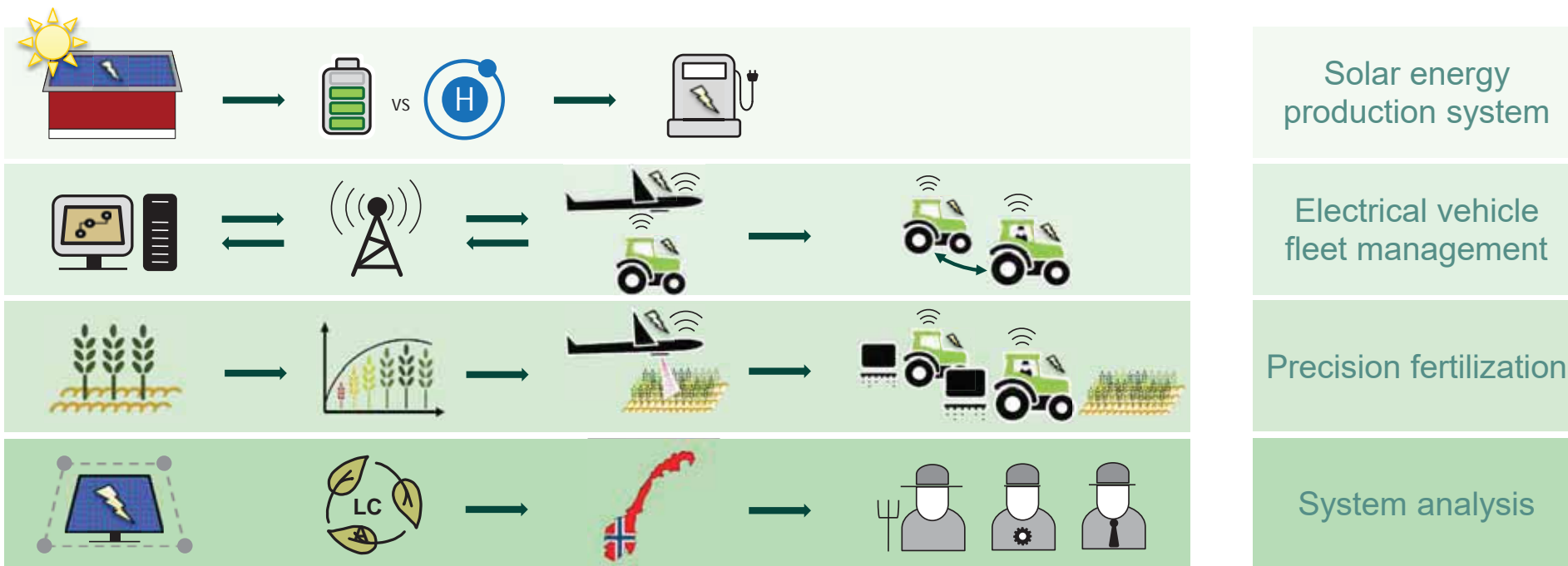
Advances and Innovations in Agriculture. The 4th NJF –EurAgEng – Agromek Joint seminar

November 29th , 2022

El Houssein Chouaib Harik / NIBIO



SolarFarm – at a glance



“The low hanging fruits”

1. Develop a fleet of electrically powered machinery as a **non-fossil based alternative**.
2. Improve nitrogen use efficiency to **reduce nitrogen losses to the environment**.



Reduce the emissions from tractor related activities on the farm

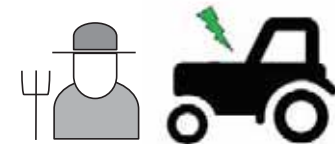
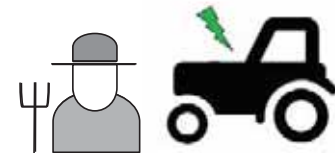
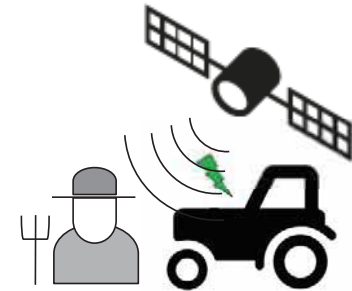
Replace diesel tractors
with electrical tractors



150 HP Diesel tractor



3 x 50 HP Electric tractor



How to drive 3 tractors at the same time?

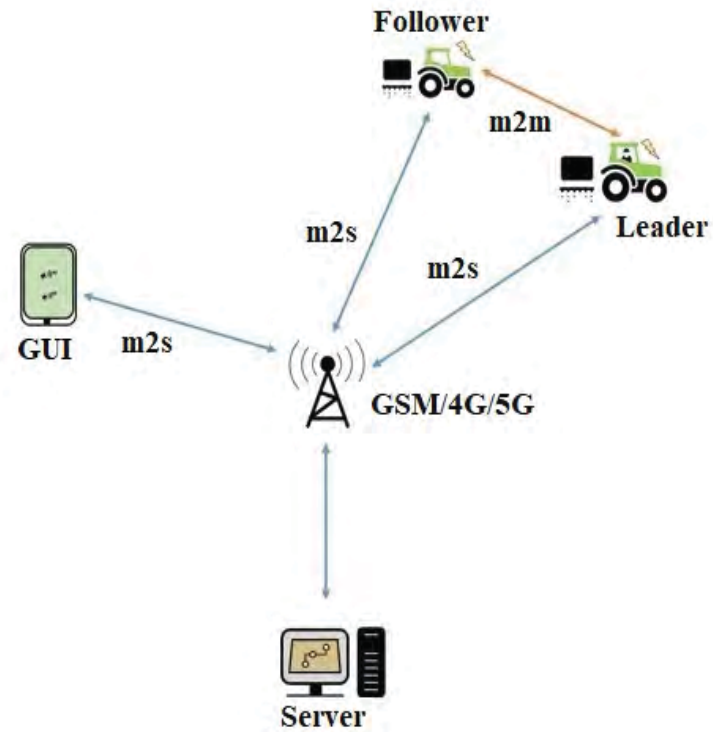
Solution 2

To reach this:

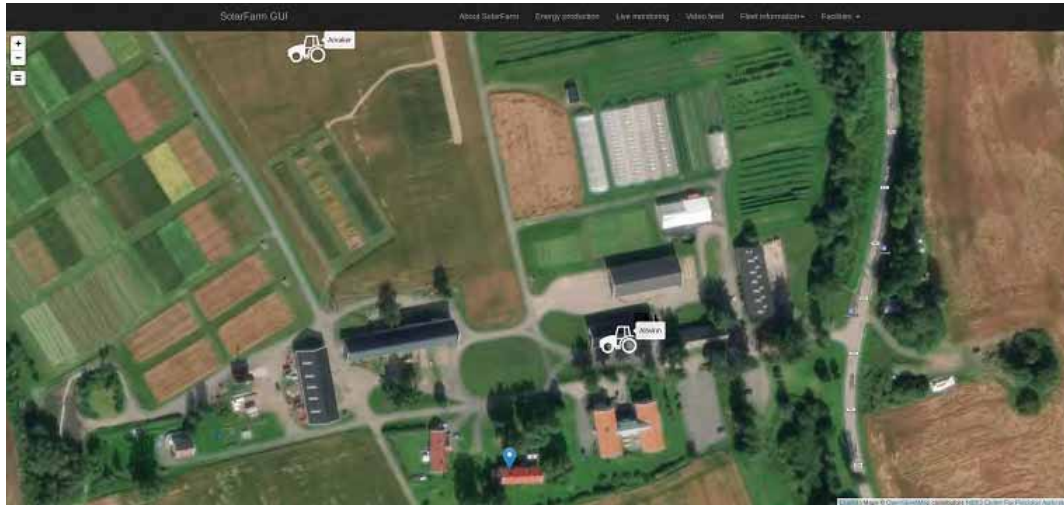
We want to "*Design a control and communication system for semi-autonomous on-farm fleet management.*"



Communication scheme



The user interface



The GUI

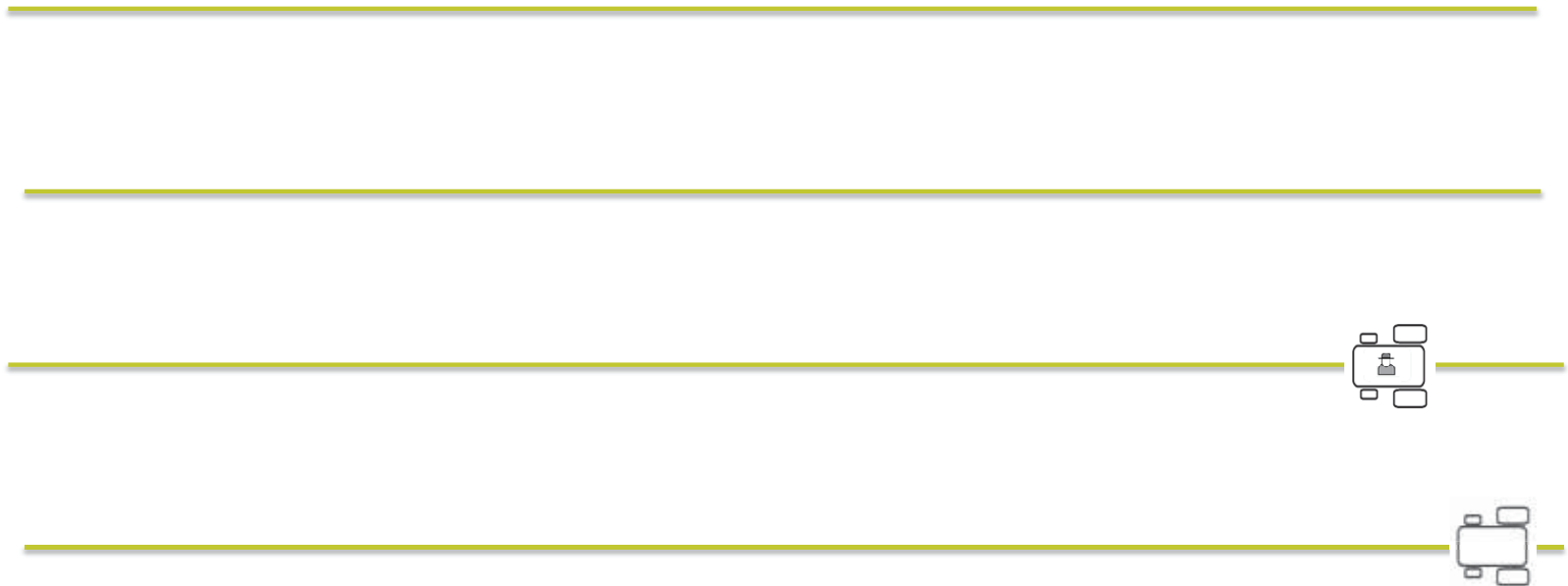


Contextual menu of the robot

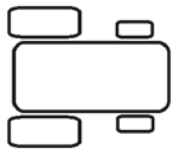
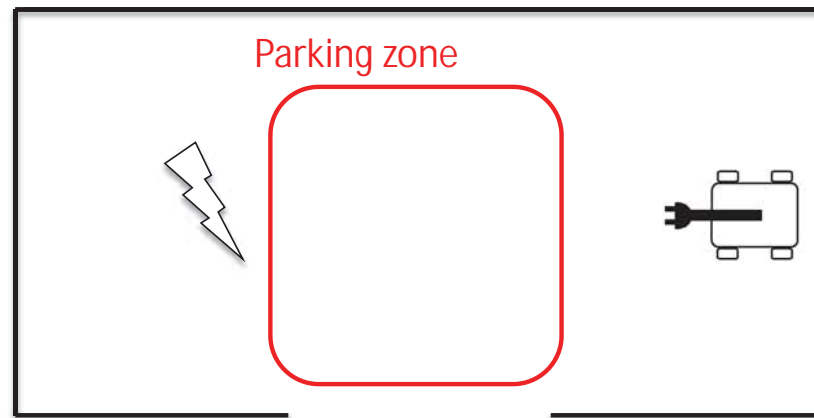


Task allocation example

Operation of vehicles in a small team



The Charging station



Reduce the emissions related to sub-optimized
use of inorganic nitrogen fertilizers

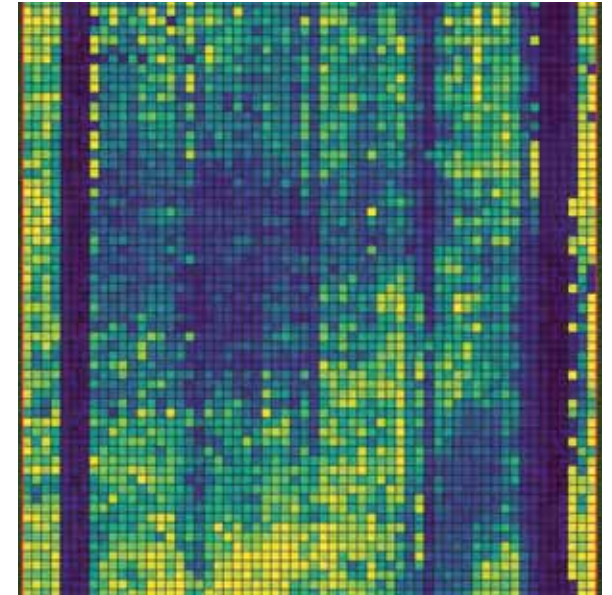
The path from drone measurement to the field application



An UAV equipped with an RGB and a hyperspectral camera

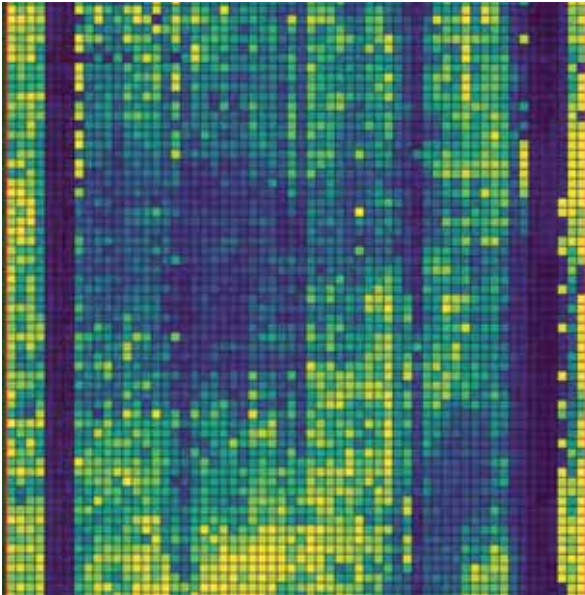


A UAV taken image (BBCH49)

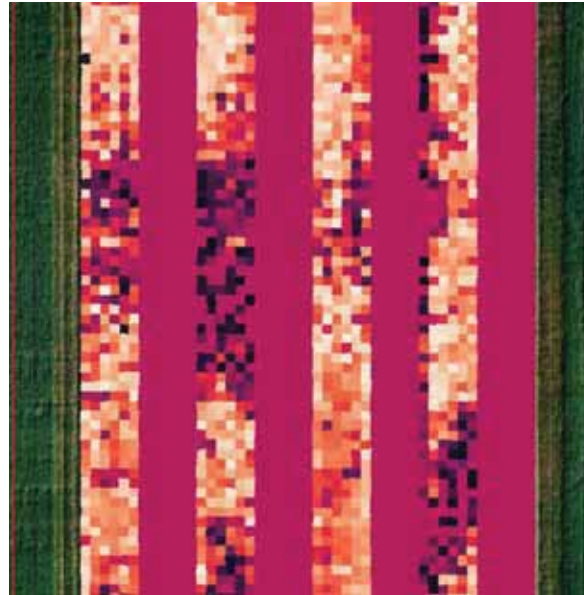


Prediction of the N uptake (BBCH49)
(blue = low, green = middle, yellow = high)

The path from drone measurement to the field application



Prediction of the N uptake (BBCH49)
(blue = low, green = middle, yellow = high)



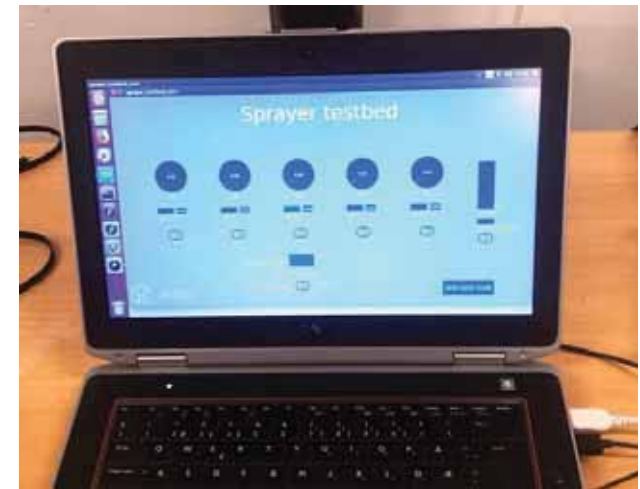
Recommendations for variable N application (BBCH49)
(dark = low, rose = middle, light = high)



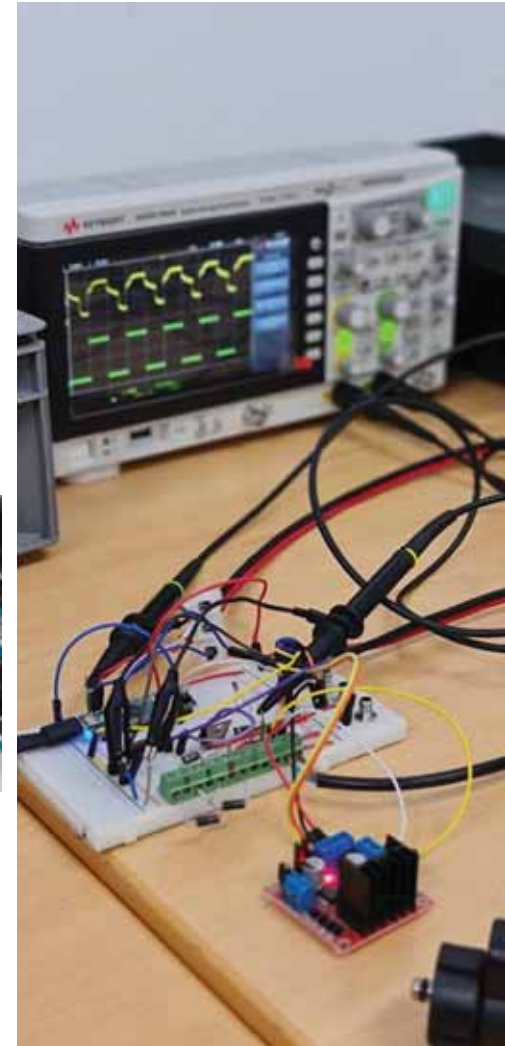
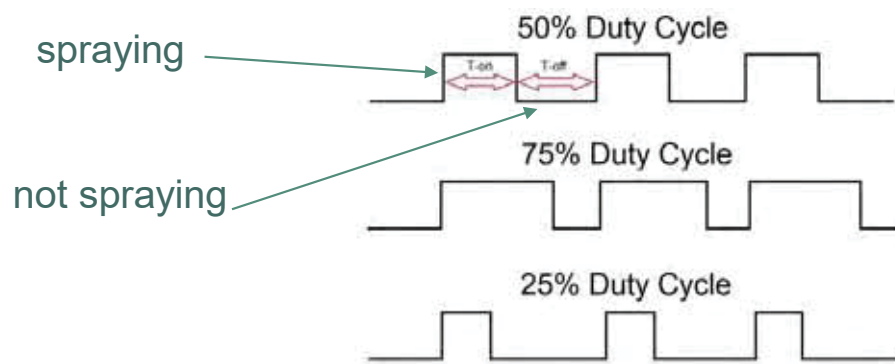
Precision application of liquid mineral N fertilizer using the developed prototype

Development of a high spatial resolution liquid fertilizer applicator

Test bench to analyze the performance of the valves and controllers and calibrate the PWM signal with the weight/volume of the liquid discharged at each nozzle.



Driver development



The first functioning prototype of the precision spreader based on a Hardi sprayer.



Thank you for your attention

El Houssein Chouaib Harik
chouaib.harik@nibio.no
www.precisionag.no



NIBIO

NORSK INSTITUTT FOR
BIOØKONOMI



NIBIO_no



NIBIO.no



NIBIO_no

www.nibio.no



Jakob Geipel
Research Scientist



El Houssein Chouaib Harik
Research Scientist



Krzysztof Kusnierek
Research Scientist



Anne-Grete Roer Hjelkrem
Research Scientist



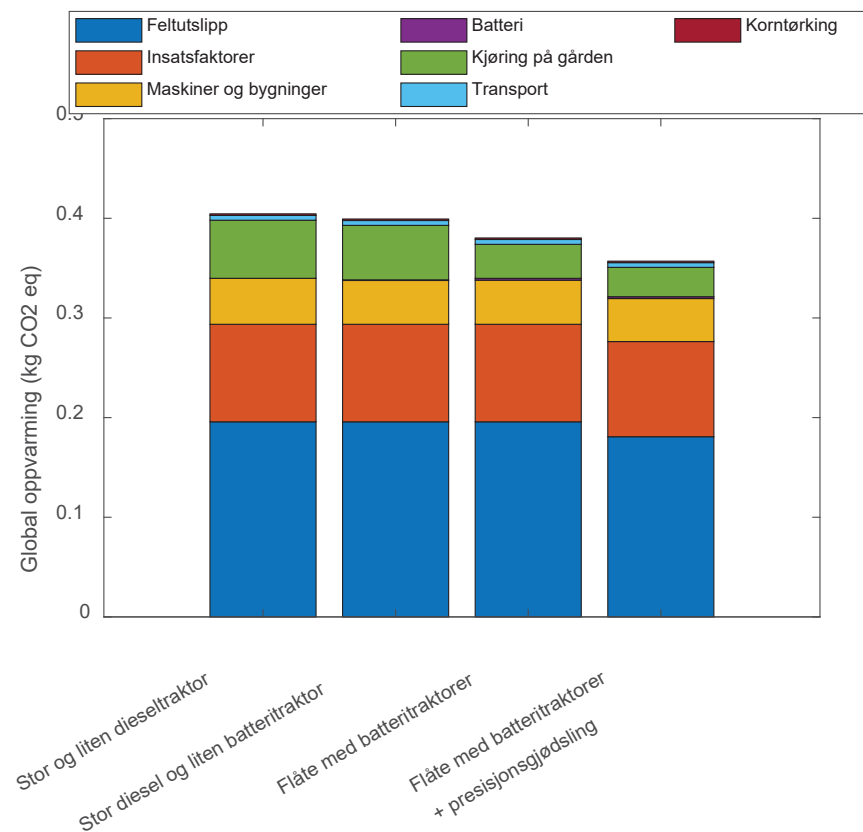
Kristian Rindal
Senior Engineer



Some numbers...

Lokalitet: Sørstørge	Stor korngrørd		Stort melkebruk		Stor kombinasjonsgård gris og korn	
Totalt forbruk	Diesel	Elektrisitet	Diesel	Elektrisitet	Diesel	Elektrisitet
Stor og liten dieseltraktor	4 910 l	-	8 250 l	-	4 380 l	-
Stor diesel- og liten batteritraktor	4 250 l	2 500 kWh (18 m ²)	8 080 l	600 kWh (5 m ²)	3 900 l	1 800 kWh (13 m ²)
Stor hydrogen- og liten batteritraktor	-	50 500 kWh (376 m ²)	-	94 200 kWh (702 m ²)	-	46 500 kWh (346 m ²)
Stor og liten hydrogen traktor	-	55 700 kWh (415 m ²)	-	95 600 kWh (712 m ²)	-	50 300 kWh (374 m ²)
Flåte med små batteritraktorer	-	10 200 kWh (76 m ²)	-	-	-	-

- Stor korngård lokalisert i Sørøst-Norge
- Funksjonell enhet: 1 kg (85% tørrstoff) med bygg, vårhvete og høsthvete ved gårdsgrinda
- Klimagassutslippene synker med omkring 6% pr kg produsert korn ved å bytte fra dieseltraktorer til batteritraktorer som lades med solenergi fra gården.

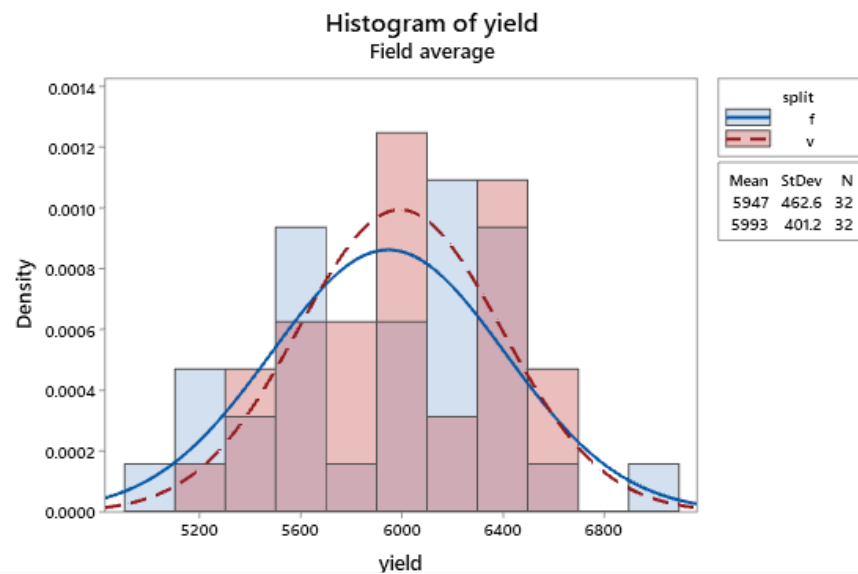


Avlingsregistrering

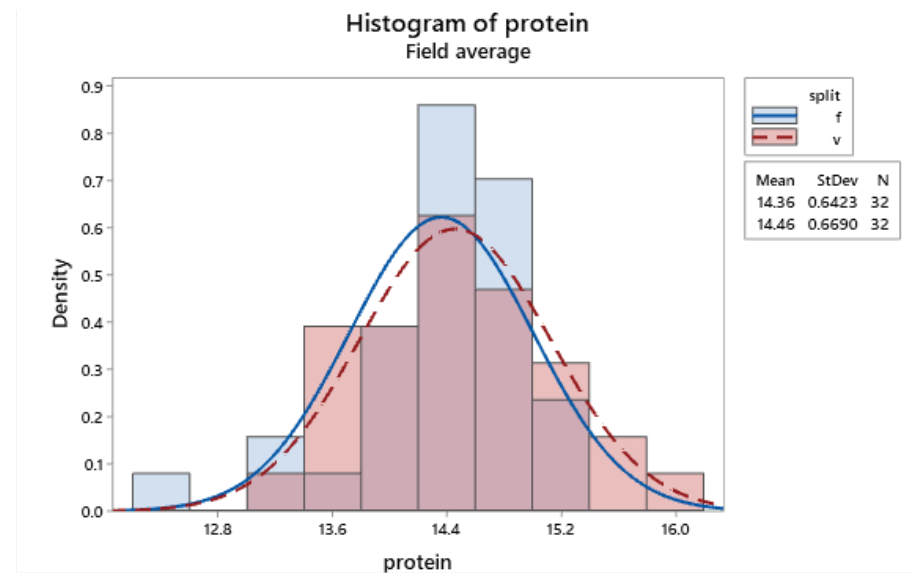
- 96 høsteruter satt ut i de forskjellige blokker/striper i feltet
- Kornprøver tatt med Haldrup forsøkestresker og registrert med samme rutiner som vanlige feltforsøk
- Avling og kvalitet analysert på Apelsvoll med samme rutiner som vanlige feltforsøk. Relativt lav avling med omtrent 600 kg daa-1.
- Jordvariasjon påvirket avlingsresultat og var ikke fullstendig eliminert av variabel N-tildeling basert på drone og regressiv agronomisk tildelingsmodell



Avlingsresultat



Relativt lav avling men forbedret homogenitet med variabel tildeling (v) sammenlignet med konvensjonell tildeling (f)



Forbedret protein-innhold men litt redusert homogenitet med variabel tildeling (v) sammenlignet med konvensjonell tildeling (f)

Sensor technology for optimal harvest in strawberry

Siv Fagertun Remberg

Associate professor

Norwegian University of Life Sciences (NMBU)



Advances and Innovations in Agricultural Engineering

The 4th NJF – EurAgEng – Agromek Joint Seminar, Herning, Denmark, 29.-30. November 2022

A photograph showing strawberry plants growing in a tunnel cultivation system. The plants are supported by a metal frame, and several ripe, red strawberries are visible hanging from the stems. The background shows a gravel path and green foliage.

Strawberryproduction (in Norway and many other countries)

- Open fields
- Greenhouses
- Tunnels
 - 1) Tunnel cultivation with plants on soil ridges
 - 2) Tunnel cultivation (high tunnels) with plants on table-tops

A changed production results in changed production methods -but give new opportunities!

- The climatic conditions are changed!
 - Temperature, light (?), precipitation/moisture
- Plants out of the ground and up in the air!
 - Require more adapted equipment (irrigation, fertilization..)
 - A production adapted to all kind of plants and cultivars?



A changed production results in changed production methods -but give new opportunities!

- Extended growing and production season
 - Both 'pre- and post' ordinary season (seasonal flowering cvs.)
- Produce (straw)berries in novel areas
 - Geographically
 - Growing technique not dependent on growing in the ground



Strawberries – plant types and type of cultivation



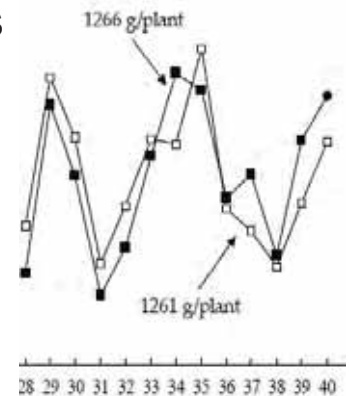
- Field production:

- Seasonal flowering cultivars
= short days are needed for flowerbud induction
- Flowering the following year
- Harvesting for 4 weeks
- Different cultivars mature at different times
- In Norway: strawberry fruit harvest June-August



- Tunnel:

- Everbearing cultivars
- Flower induction in long days
- Yield (on the same plants) the whole growing season
 - Usually 3 crop tops in Norway



Norwegian University of Life Sciences

Strawberry cultivation

- All strawberry cultivars are different!
 - Growth and development, fruit shape, taste and flavour, chemical content, susceptibility to insects and diseases, frost tolerance etc.
- There are certain requirements for strawberry cultivars used in commercial production (which is not decided by the producer!)
- Different requirements and regulations for production in open fields compared to greenhouse/tunnel
- Possible changes due to cultivation practice:
 - Pests and diseases, berry quality, pollination etc.







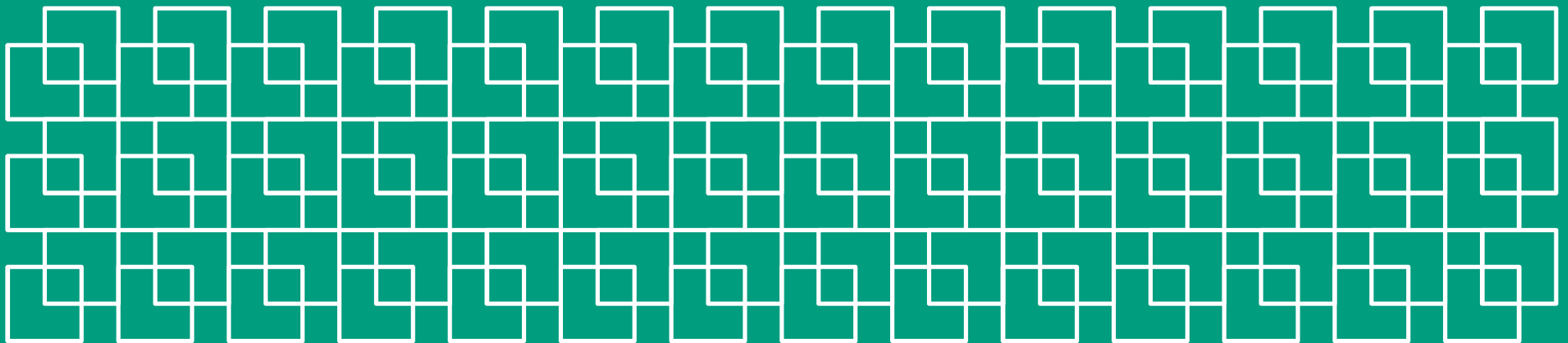
Case from NMBU: strawberry tabletop tunnel cultivation

- Cooperation: Faculty of Science and Technology (Realtek) and Biosciences (Biovit) and Centre for Plant Research in Controlled Climate (SKP), NMBU
- Funding from NMBU, FORREGION Vestfold and Saga Robotics at an early stage
- Tunnel (Haygrove, UK (Myhrene AS) and tabletop built in 2019, automatic irrigation and fertigation system (Elceta), technical help from SKP and Realtek



Innovation project (SAGA Robotics, RCN/NFR):

Sensors for automatic strawberry precision picking
Sensorer for automatisk presisjonsplukking av jordbær



MålBær

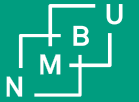
- Aims to develop sensors for estimating:
 - Size and shape
 - Ripeness
 - Chemical indicators
 - Diseases?
 - With the use of an autonomous (strawberry-picking) robot
-



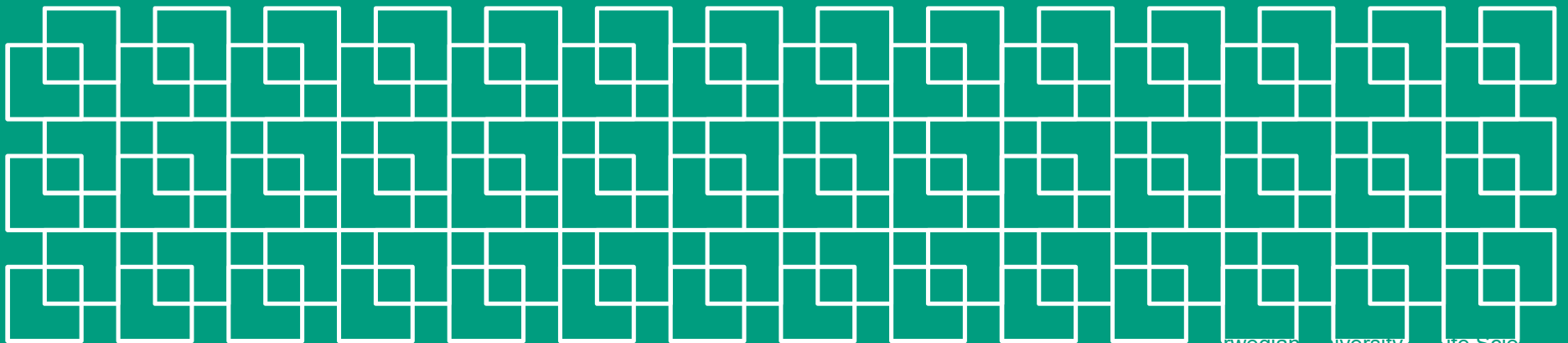
MålBær

- The value potential of a strawberry-picking robot will be realized when:
 - The sensors will give reliable information on berry quality at harvest
- Destructive measurements are used to measure various quality attributes
- Correlated to non-destructive measurements (different sensors)





Technology can help to solve
the food production in the future!



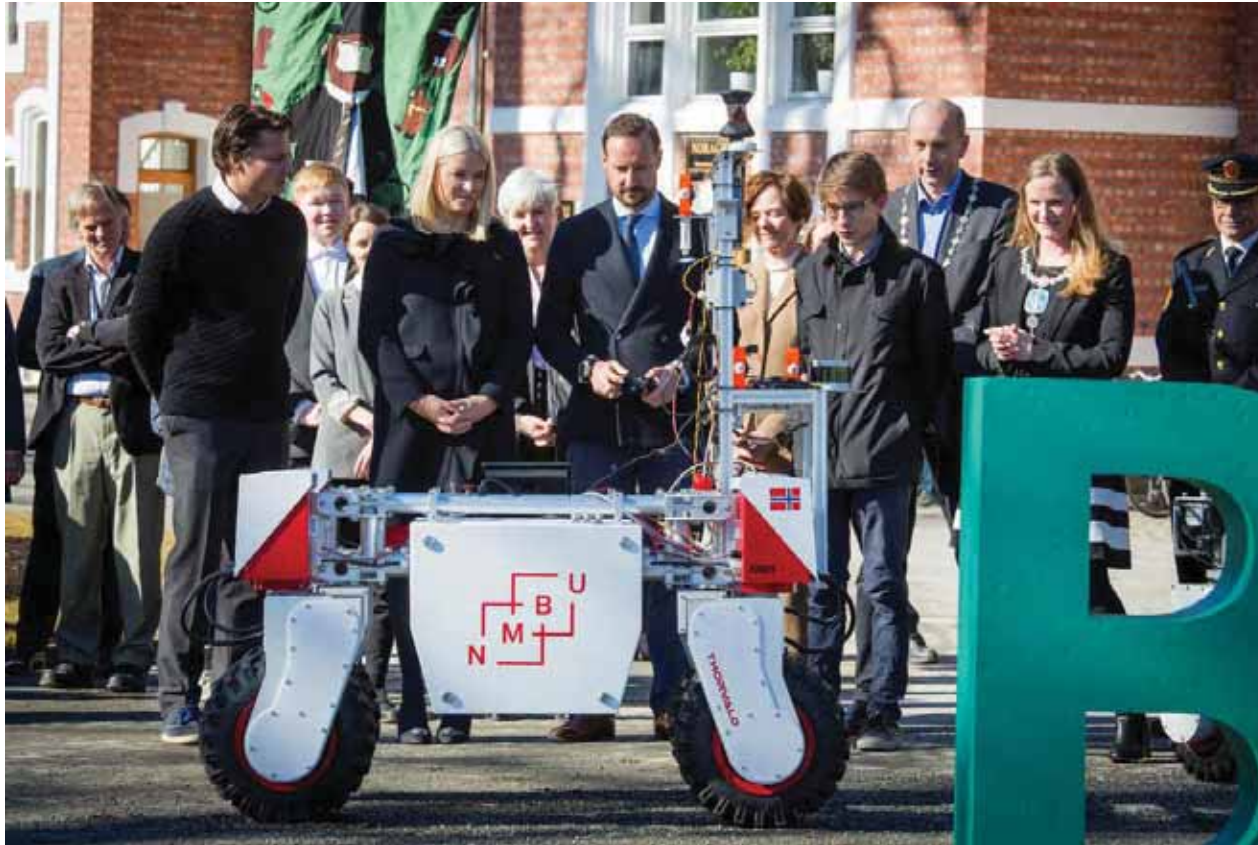
Thorvald – the robot!



Photo: nmbu.no



Thorvald – the famous robot!



Royal visit to NMBU in March 2017

Photo: Håkon Sparre, NMBU

Thorvald as a platform –and different technical solutions on various challenges



- Powdery mildew in strawberries
- Efficient, autonomous, sustainable and cost-effective treatment
- UV-C light treatments every night
 - Exposing the plants to shortwave light weekly at night
 - A substitute for conventional (chemical) fungicide application
 - Controls powdery mildew
 - Autonomous application
 - Environmental benefits





Strawberry picker

- A unit attached on Thorvald platform
- This unit can track and seek out single berries
 - Can identify red colour
 - This does not mean berries of the correct ripening stage!
 - Succession rate of 70-90%, with a low rate of picking unripe berries
- It can also follow and register berry development





Strawberry picker

- It has a potential for including several sensors
- Sensordata can optimize production and harvest and will provide yield predictability valuable for the market
- To develop this picking unit further:
 - Distinguish between unripe berries, and berries of various ripening stages
 - Identify berries with different symptoms
 - Diseases, deformed berries
 - In the end: sort the different qualities to ensure berries of good quality in the punnets





Strawberry picker



- Tested in UK and Norway in 2020
- In traditional strawberry production, labor-costs are estimated to approx. 40% of the production costs
- In Europe, shortage of labor is a problem
 - Automatization will then play an important role for this industry
- Customers are willing to pay good money for berries of good quality
- The sensors developed in this project can also be used on other crops, e.g. apples



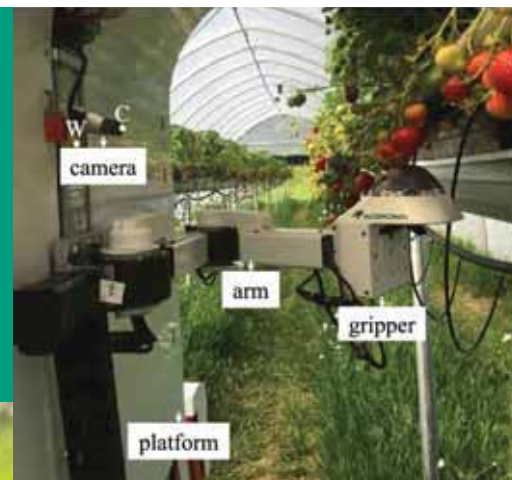
Project 2021&2022



- 2021:
 - Growing strawberries in NMBU-tunnel in a table-top system
 - Berry quality analyses
 - “traditional” destructive analyses
 - Develop and test sensors in lab
 - Outdoor and indoor measurements with multiple sensors
 - Sensory panel (human taste/flavour perception)
- 2022: Further testing of the robot
- 2023: Develop sensors on Thorvald, further and final testing!



Promising
preliminary results!





Thank you for listening!