



Modernisation of agricultural statistics

Overview of the Agri Sisa project and our future plans

Remco Paulussen
October 2nd, 2023



Project Agri Sisa

- Agriculture System Integration and Spatial Analysis
- Acronym is 2020-NL-AGRI-SISA, or short Agri Sisa
- Eurostat grant: Modernisation of Agricultural Statistics (ESTAT-2020-PA8-E-AGRI)
- Together with LUKE, Finland
- Contact at Eurostat: Johan Selenius
- Budget is 260K Euro: 219 CBS and 41 LUKE
- Reservation of 40K Euro for services by external parties
- Duration from November 1st, 2021 until December 31st, 2023

Work Packages and Organisation

WP1

Ger

System-to-System data communication applied to official crop yield surveys

WP2

Patrick



Spatial analysis of drought effects on crop yield

WP3

Arthur



Farm Management Information Systems

WP4

Remco

Martin (admin)

Maria (LUKE)

Coordination

(bi)monthly project meeting



WP1: System-to-System data communication applied to official crop yield surveys

Ger Snijkers



General idea

4.4.1 Fosfaatproductie in dierlijke mest per landbouwgebied in 2016

kg fosfaat per ha



7h1 Vaste mest: werkresultaat mesttoediening			
Geef per werkresultaat het percentage vaste mest aan op bouwland en grasland.		% van totaal toegesende vaste mest op grasland	% van totaal toegesende vaste mest op onbebouwd bouwland
	Bouwland	bewergronds toegesend, daarna ondergewerkt	n.v.t.
	Grasland en bouwland	mest ligt verdeeld over perceel na bewergronds toedienen	<input type="checkbox"/>
	Totaal		<input type="checkbox"/>

7h2 Drijfmeest: werkresultaat mesttoediening			
Geef per werkresultaat het percentage drijfmeest aan op bouwland en grasland.		% van totaal toegesende drijfmeest op grasland	% van totaal toegesende drijfmeest op bouwland
	Bouwland	bouwlandinjecteur: d.m.v. injectiestanden direct in de grond gebracht	n.v.t.
	Grasland en bouwland	de mest ligt verdeeld over het perceel zoals bij breedwerpig bewergronds toedienen of bij een machinever boven de grond houden	<input type="checkbox"/>
	Grasland	de mest ligt op de grond in strookjes zoals bij juist gebruik van een sleepvoet-machine of bij gebruik van een sleepfouder of zodenbester die geen sleepje maakt of niet srijdt	<input type="checkbox"/>
	Grasland	de mest ligt gedeeltelijk in strookjes in de grond en gedeeltelijk op de grond zoals bij gebruik van een sleepfouder of bij ondiep werken met een zodenbester	<input type="checkbox"/>
	Grasland en bouwland	de mest is geheel in de grond gebracht in strookjes zoals bij juist gebruik van een zodenbester	<input type="checkbox"/>
	Totaal		<input type="checkbox"/>



“Why do I still have to do this manually?”

Pre-filling:
How to make this work?

Pilot with
 JOHN DEERE

Pre fill

Technoboer heeft de toekomst:



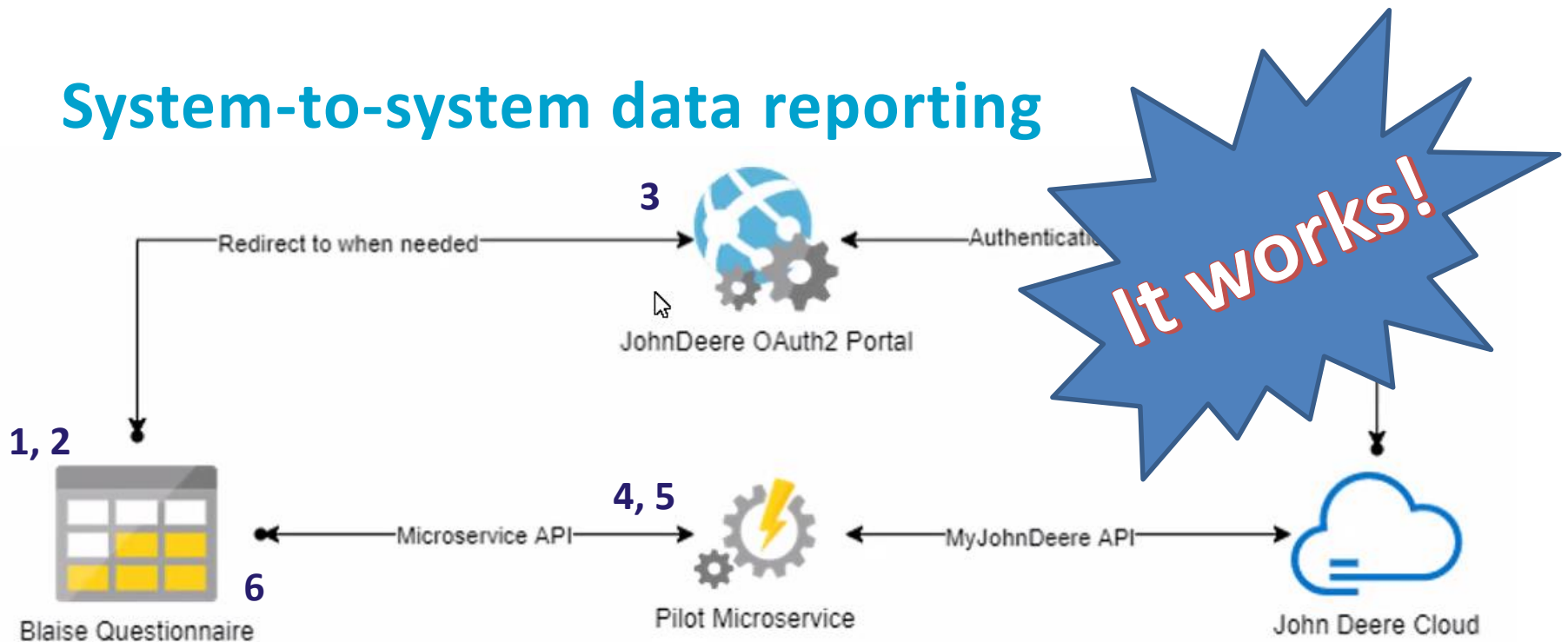
©John Deere

200+ registries

Techno farmer has the future

Smart industries
Smart farming

System-to-system data reporting



Steps in process:

1. Online Q login
2. MyJohnDeere?
3. Authentication
4. Online Q <-> Microservice <-> John Deere
5. Prefilled questionnaire
6. Submit?



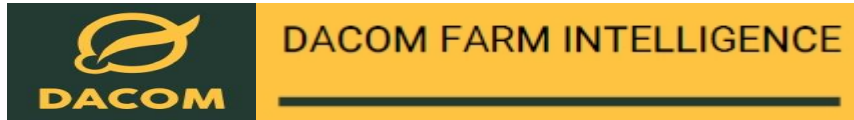
WP3: Farm Management Information Systems

Arthur Denneman



FMIS vendors The Netherlands

- Two most-used systems in Netherlands:



- Connection made using AgroConnect (EDI-Crop Standard)
- Data of some farmers analysed; results are very promising

FMIS vendors Finland

- Suonentieto
 - Agrineuvos
 - AgriSmart

} focus on arable farming
- Mtech
 - Visu
 - WebWisu
 - MobiWisu

} focus on dairy production
- Softsalon
- Peltotuki Pro
- Datatech
 - Aktiivipelto
 - Aktiivikirjanpito



Perustiedot Kasvulohko Työkäsit Tulosteet Kavoukusi Asetukset Tutkimukset Ohje

Yhtäyksiön Tien tiedot Viheruusi Tavutukset Muuttopöytä Tarkastelu Viikajärven AgriSmart Visu Dokumentit Tulokset Sijla

2019 921 03103-51 Ruukkosuo 4,90 ha P-tasaus päättyy xx Pärsti Nurmi

Kasvulohkot

Tunnus	Nimi	Ala	Kasvi	Käyttö	Laatu
1-B	Ruukkosuo	0,51	Kaura (S)	Siemet	Niläis
2-A	Etupelto	2,90	Kevätkasvi	Olyntuotanto	Syntha
3-A	Talopelto	2,15	Nurmet (Doruustaan)	Säilörehunurmi	Nurmeso
4-A	LÄTOHANZO	0,32	Nurmet (Doruustaan)	Säilörehunurmi	Nurmeso
5-A	Pihapello	1,18	Nurmet (Doruustaan)	Laidunurmi	Nurmeso
6-A	PAJAPELTO	0,77	Nurmet (Puuustaan)	Laidunurmi	Sien soppakana
6-B	PAJAPELTO	1,84	Nurmet (Doruustaan)	Laidunurmi	Nurmeso
7-A	Iso tarha	0,53	Nurmet (Doruustaan)	Laidunurmi	Apilapönnön seos
8-A	Pikutarha	0,41	Nurmet (Doruustaan)	Laidunurmi	Apilapönnön seos
9-A	Lähdäkierros	1,13	Nurmet (Puuustaan)	Laidunurmi	Sien soppakana
10-A	Ruivo	0,28	Nurmet (Doruustaan)	Mitturuus (Suuhäntä)	Nurmeso
11-A	Hajälkä	1,30	Nurmet (Doruustaan)	Mitturuus (Suuhäntä)	Nurmeso
12-A	Aluska	0,32	Kaura (S)	Siemet	Niläis
13-A	Pikäranta	0,37	Nurmet (Doruustaan)	Laidunurmi	Nurmeso
14-A	Alantarha	0,98	Nurmet (Doruustaan)	Laidunurmi	Nurmeso
15-A	Talokaspieno	1,24	Nurmet (Puuustaan)	Laidunurmi	Sien soppakana
17-A	Lehtomäki	1,98	Nurmet (Doruustaan)	Säilörehunurmi	Nurmeso
17-B	Lehtomäki	1,18	Nurmet (Doruustaan)	Säilörehunurmi	Nurmeso
18-A	Hakurietty	1,90	Suogryhkyt (pt. alk. 2...)	Säilörehunurmi	Nurmeso
19-A	Tarvetus	0,70	Nurmet (Doruustaan)	Säilörehunurmi	Nurmeso
20-A	Rippelto	0,39	Suogryhkyt (pt. alk. 2...)	Siemet	Niläis
21-A	Halla-ohi	1,41	Kaura	Siemet	Niläis
22-A	Rippelto	0,85	Kaura	Siemet	Niläis
22-B	Rippelto	0,42	Kaura	Siemet	Niläis
23-A	NOKOLA 1	0,95	Kaura	Siemet	Niläis
24-A	NOKOLA 2	0,52	Suogryhkyt (pt. alk. 2...)	Siemet	Niläis

6,2386,22 ha 2142 9,11 %

Kartta

Historia | Ihteritänen | Kasvupeteyty | Kasvatit, LPH, Mönnytuotus | Kasvujaksot

2015	2016	2017	2018	2019
A Ruukkosuo 1,60	A Ruukkosuo 4,89	A Ruukkosuo 4,90	A Ruukkosuo 4,90	A Ruukkosuo
Nurmet (Doruustaan)	Kaura	Ohra	Ohra	Nurmet (Doruustaan)
Apilapönnön seos	Niläis	Niläis	Terve	Niläis
Mitturuus (Suuhäntä)	3200 kg	3200 kg	3200 kg	3206,04
Terve m. L. seos	3200 kg	3200 kg	3200 kg	3206,04

6,2386,22 ha



Lot of information, e.g. Suonentieto

Agrineuvos	AgriSmart
<ul style="list-style-type: none">• Cultivation plan and cultivation notes• Soil sample and fertility studies• Parcel nutrient accumulation and nutrient demand (environmental commitment, nitrate regulation)• Farm inputs (manures, fertilizers, plant protection agents, seeds and crop products)• Plants and varieties• Industrial fertilizers• Domestic and industrial manures (manure analyses)• Soil conditioners• Plant protection agents	<ul style="list-style-type: none">• Farming notes (planned/completed tasks, related supplies and parcel information)• Supplies (manures, fertilizers, plant protection agents, seeds and crop products)• Warehouses (material flows related to the location and the supplies in them)• Customers (connected to work area or route)• Work areas (field parcels, forest blocks and other work areas)• Notes (with location and pictures)• Yield maps• Satellite images• Work machines and their routes (tractors, connected devices, types of work)• Treatment zones (formation of ISOBUS tasks)



Conclusions and next step

- Results are very promising: crop statistics, pesticides (becomes annual), manure (to be analysed)
- System-to-system communication works!
- FMIS vendors are willing to support integration
- FMIS vendors have different interfaces
- FMIS landscape is scattered; many vendors European-wide
- NSI's together are a *bigger fish* to support

Our proposal: let's form a consortium to take this further as part of MAS2023



WP2: Spatial analysis of drought effects on crop yield

Patrick Bogaart



Main objective and information sources

Develop methods to quantify and map drought stress related crop yield reductions, using approaches that are consistent with the current agricultural crop yield statistics.

We focus on two countries that have been experiencing a cool to temperate climate in the past, but now start to experience serious droughts: The Netherlands and Finland

Information sources

Hydrometeorological data

- Precipitation, (potential) evapotranspiration, etc.
- Station; gridded

Agricultural statistics

- Crops planned, yields;
- Local; Regional

Earth Observation

- Biomass; NPP; Soil moisture?
- Yield predictions

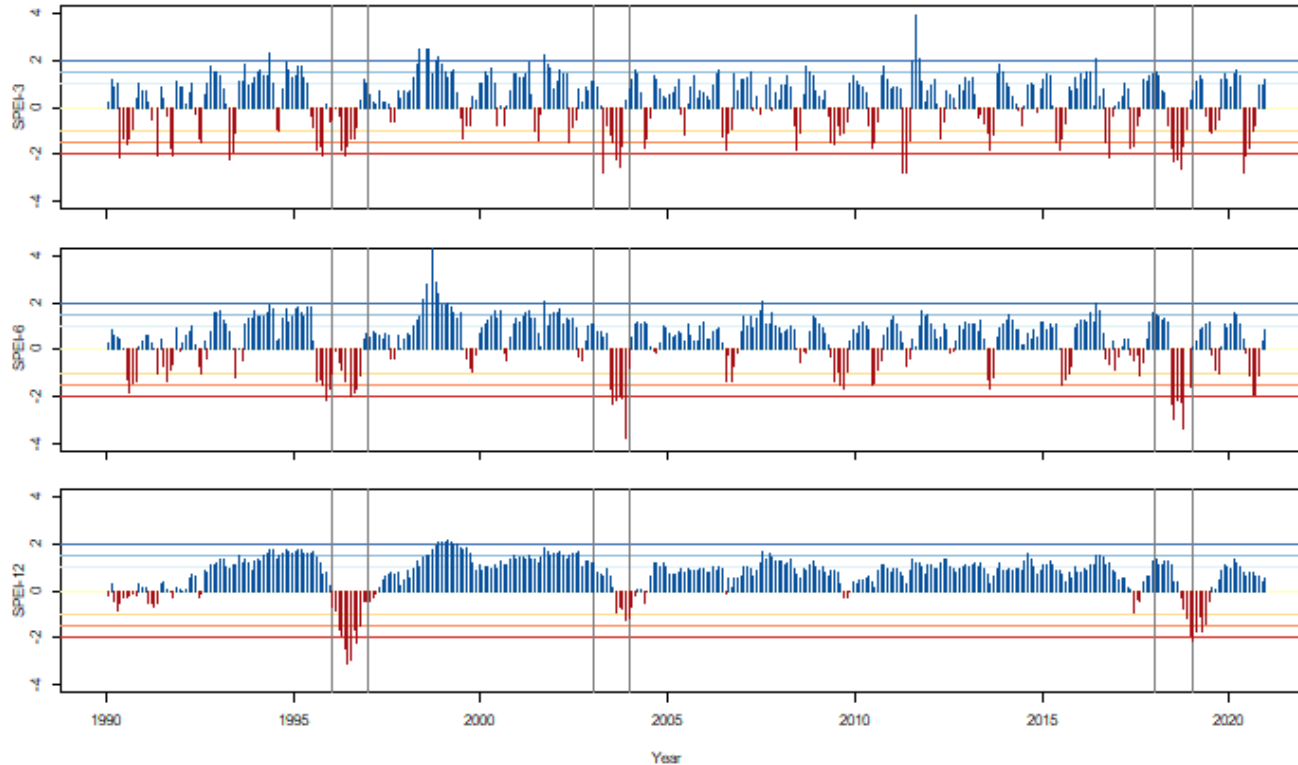
Mechanistic Crop-growth modelling

- SWAP / WOFOST



Standardized Precipitation-Evapotranspiration Index (SPEI)

Normalized ($\mu=0$; $sd=1$) and smoothed (moving avg) difference between precipitation and potential evapotranspiration)



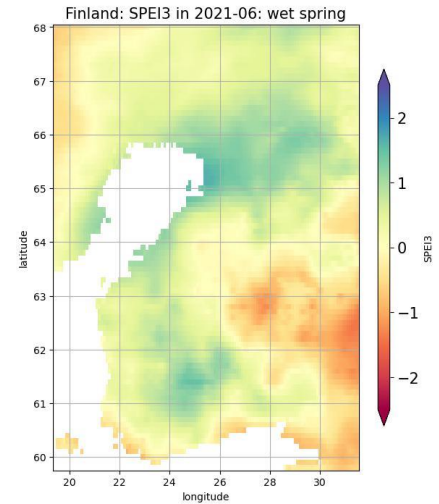
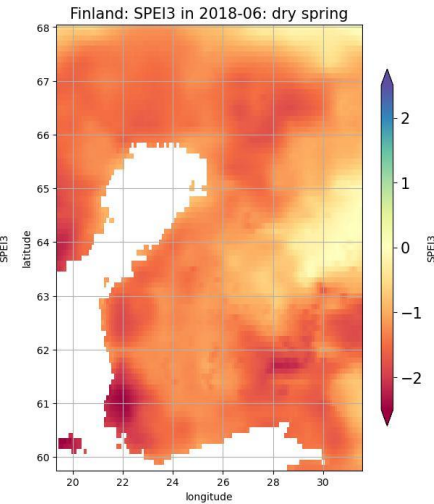
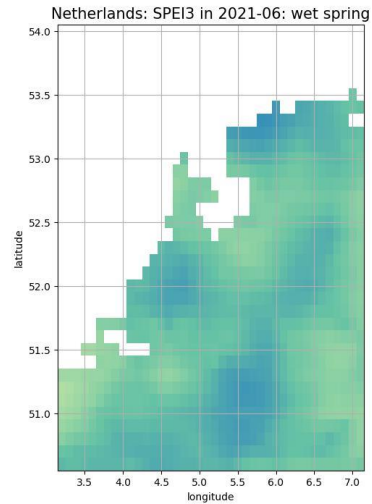
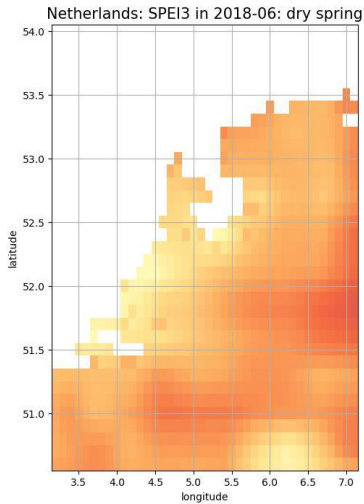
De Bilt region (NL)

Value	Classification
2.0 or more	Extremely Wet
1.5 to 1.99	Very Wet
1.0 to 1.49	Moderate Wet
-0.99 to 0.99	Normal
-1.0 to -1.49	Moderate Dry
-1.5 to -1.99	Very Dry
-2.0 or less	Extremely Dry



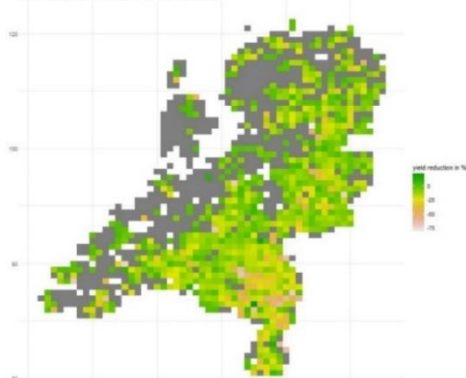
SPEI maps for the Netherlands and Finland

Data from (P) and based on (Etpot) ERA-5 reanalysis



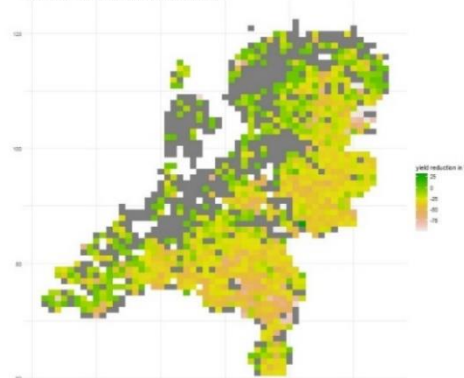
Yield reductions for Maize, 2017 - 2022

Yield reduction for maize in 2017 using the quantile method



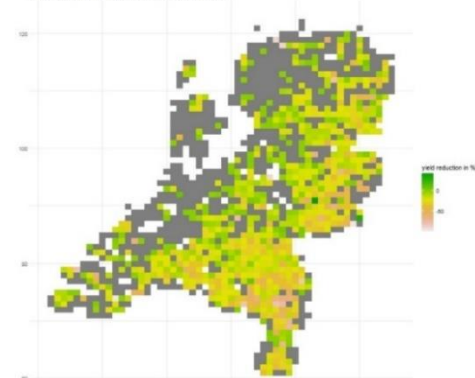
Summer 2017: from extremely hot to cooler temperatures

Yield reduction for maize in 2018 using the quantile method



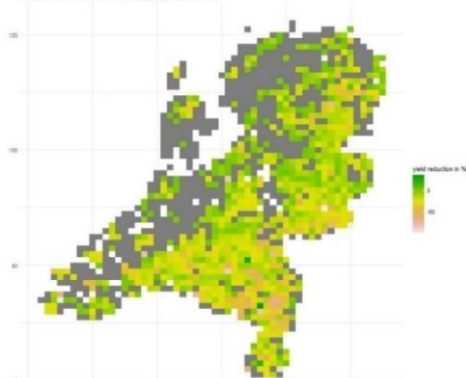
Summer 2018, extremely hot, very sunny and very dry

Yield reduction for maize in 2019 using the quantile method



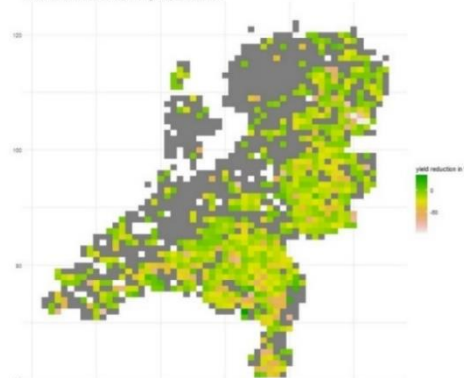
Summer 2019: extremely hot, very sunny and dry

Yield reduction for maize in 2020 using the quantile method



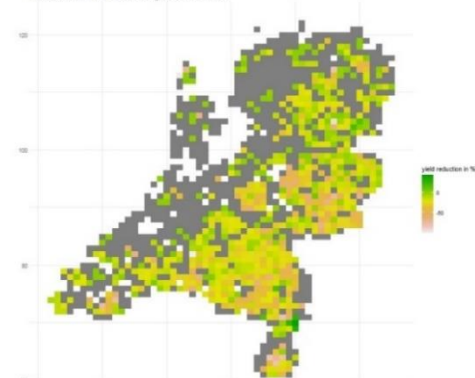
Summer 2020: extremely hot, very sunny, normal precipitation

Yield reduction for maize in 2021 using the quantile method



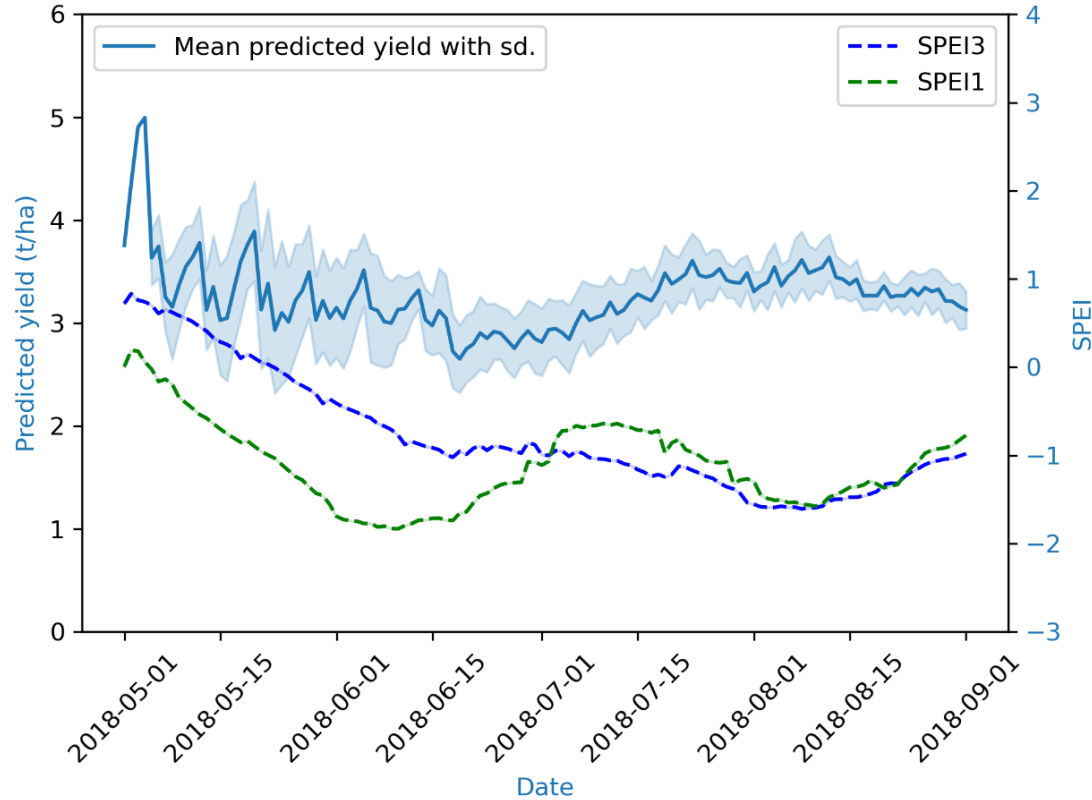
Summer 2021: normal summer, but a bit more precipitation than normal

Yield reduction for maize in 2022 using the quantile method



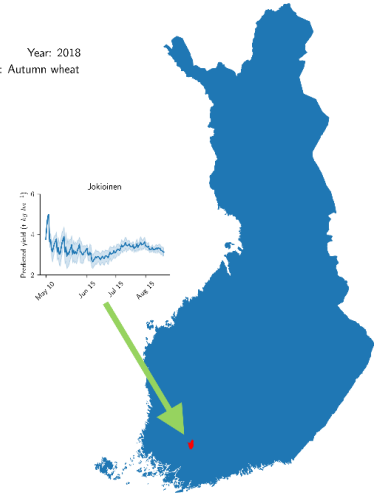
Summer 2022: extremely hot, very sunny and very dry

Using Machine learning to continuously predict yield



Jokioinen region (FI)

Year: 2018
Crop: Autumn wheat

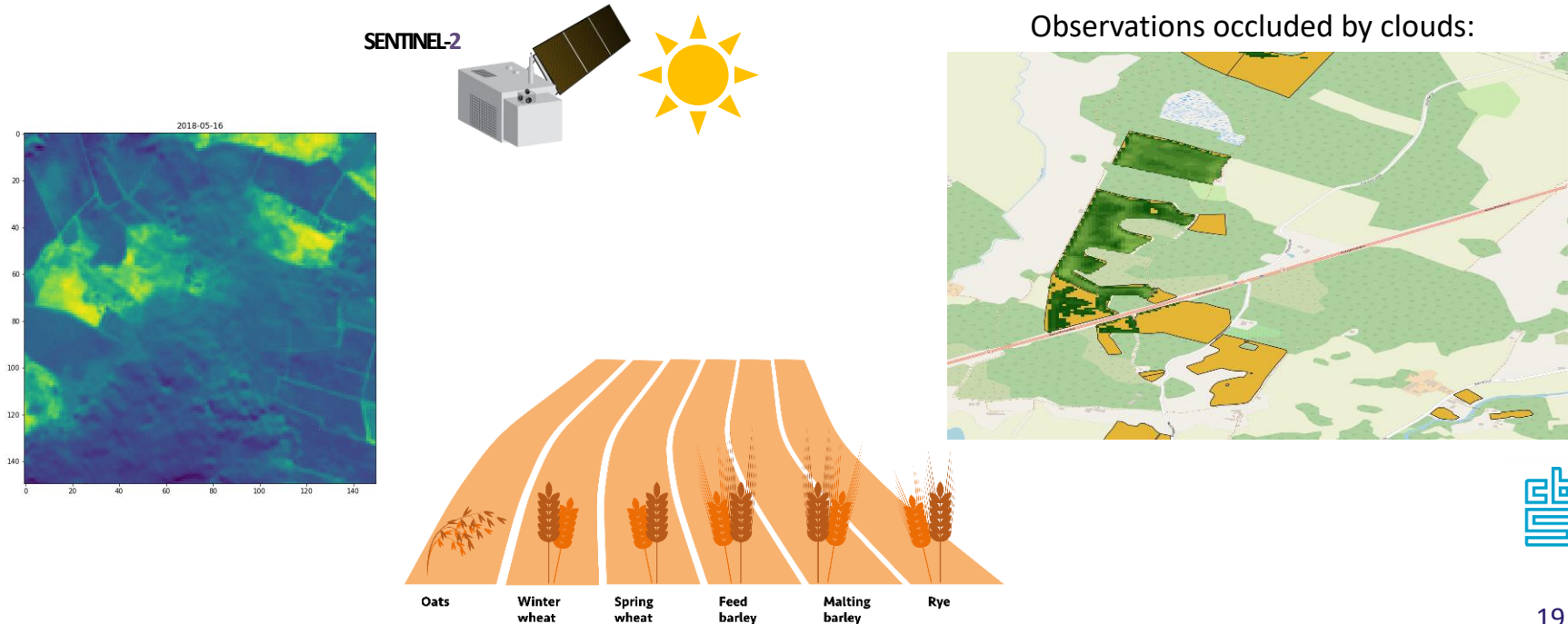


Conclusions and next steps (1)

- Quantitative analysis of yields and yield reduction in terms of explanatory variables
 - Using stepwise linear regression
 - Using Temperature, SP(E)I (1,3,6,12), for each month
 - Two data sets:
 - Recent years: spatially distributed, 5x5km scale
 - Longer time series: regional scale
- First results indicate feasibility of approach

Illustrate one approach: Earth Observation

Earth Observation AI/ML model created by LUKE



Each crop type a separate prediction model

Input data: Sentinel-2

- Reflectance values from 10 bands
- Spatial resolution 10m
- Temporal resolution 2-3 days

Trained on farm-level yields

- From crop production survey

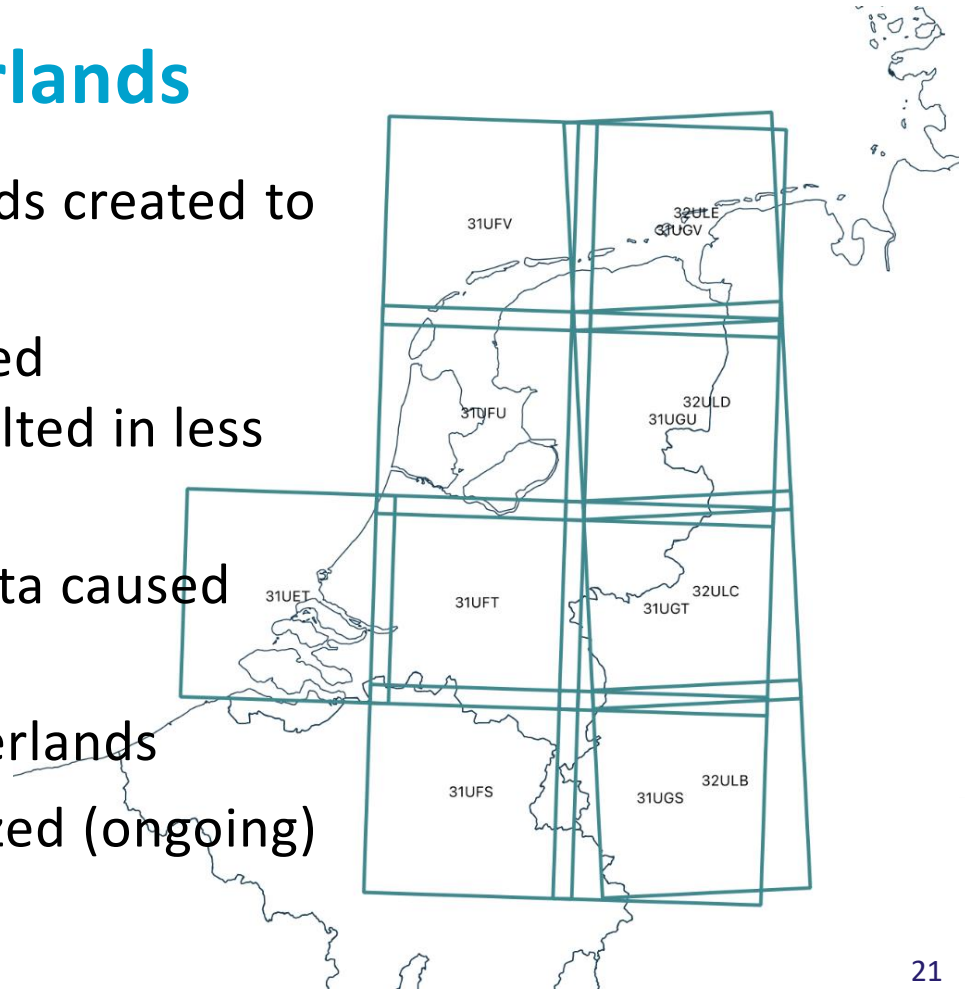
Output: Field-level predictions

- Field or any homogeneous spatial object



Applied to The Netherlands

- Dataset with farm-level yields created to re-train the model
- Dataset statistically protected (competition sensitive) resulted in less quality
- Training with less quality data caused model to perform less
- Model applied to The Netherlands
- Results retrieved and analyzed (ongoing)



Conclusions and next steps (2)

- Besides the interesting results on drought effects on crop yield, we see potential in generalizing and applying existing EO AI/ML models on other countries and time frames
- Proposal for AI/ML grant: One Stop Shop, WP7 on use case Earth Observation (start date April 1st, 2024)
- Idea is to use the Polish crop yield model (among others)
- Consortium created, consisting of AT, DK, FR, IE, IT (co-lead), NL (lead), PL and PT



Q&A

