

# Mapping grassland age at a national scale using multidecadal satellite time series

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## Klima-Fern Project:

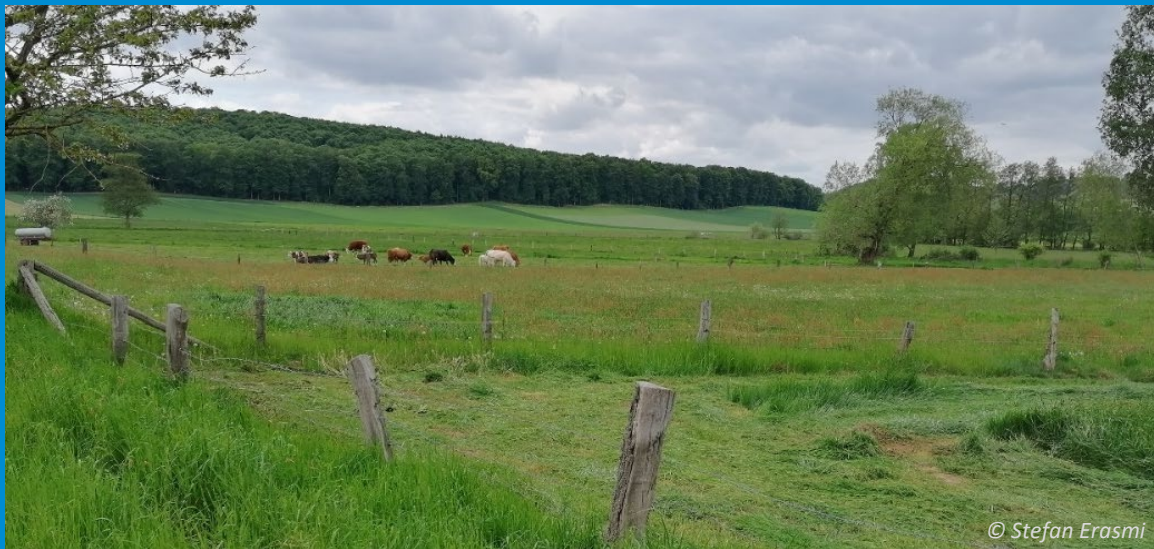
*Remote sensing of  
agricultural land use for  
enhanced climate reporting*

With support from



Federal Ministry  
of Agriculture, Food  
and Regional Identity

by decision of the  
German Bundestag



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# Importance of grassland history

- Grassland type and management intensity directly affect ecosystem function and service provision (Bengtsson et al., 2019)
- Grassland resilience, carbon storage or species composition develops within the initial 15-60 years (Ballesteros et al., 2024; Emde et al., 2024; Snoo et al., 2012; Wagg et al., 2022)
- Still, spatial explicit information on grassland age is rarely available



# Objectives

**What is the spatial distribution of grassland-age throughout Germany?**

**We define “grassland-age” as years since initial grassland establishment after land use (LU) change.**

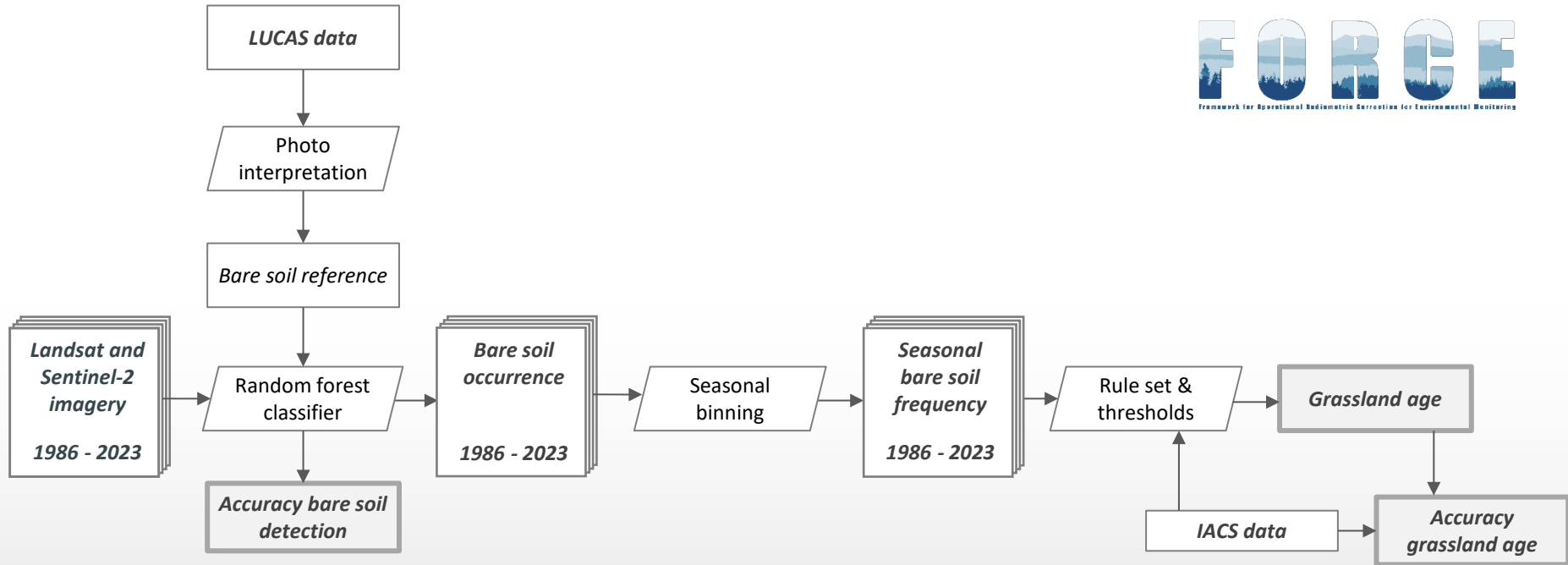
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We define “grassland-age” as years since initial grassland establishment after land use (LU) change.

→ One approach is to go via the **detection of bare soil** to capture **grassland conversion**.

# Workflow



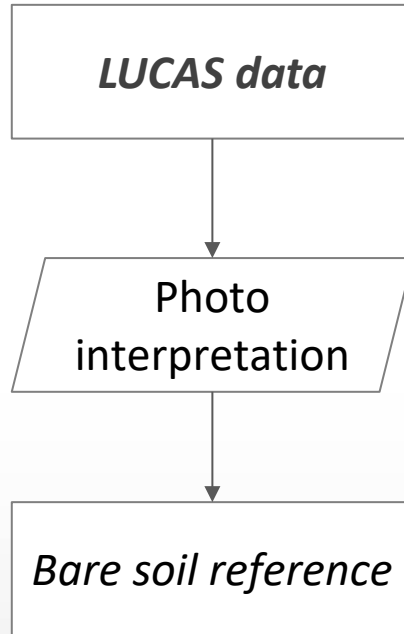


***Landsat and  
Sentinel-2  
imagery***

***1986 - 2023***

- **Landsat 5-9 and Sentinel-2 for 1986–2023**
- Preprocessed using FORCE (Fantz et al., 2019)
- Landsat adjusted to reflectance of Sentinel-2 following Okujeni et al. (2024)

# Workflow



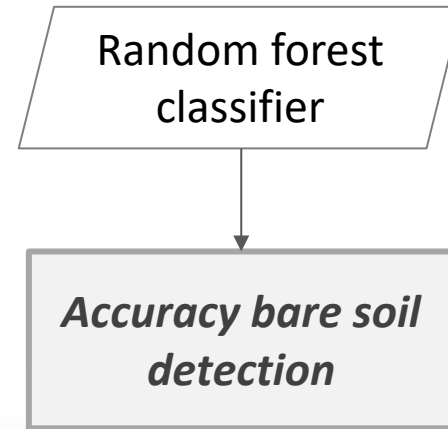
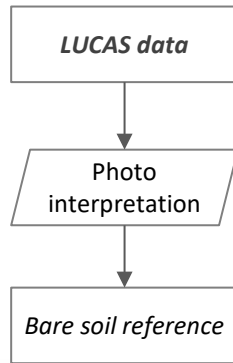
*Landsat and  
Sentinel-2  
imagery*

*1986 - 2023*

## Land Use and Cover Area frame Survey 2006-2018 (d'Andrimont et al. 2020)

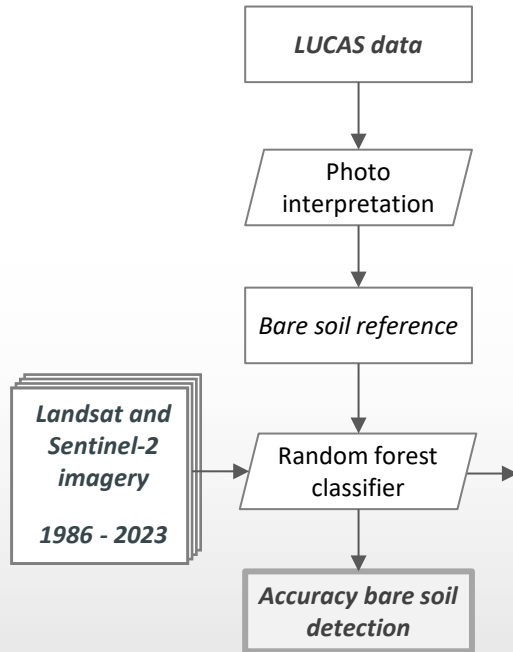


# Workflow

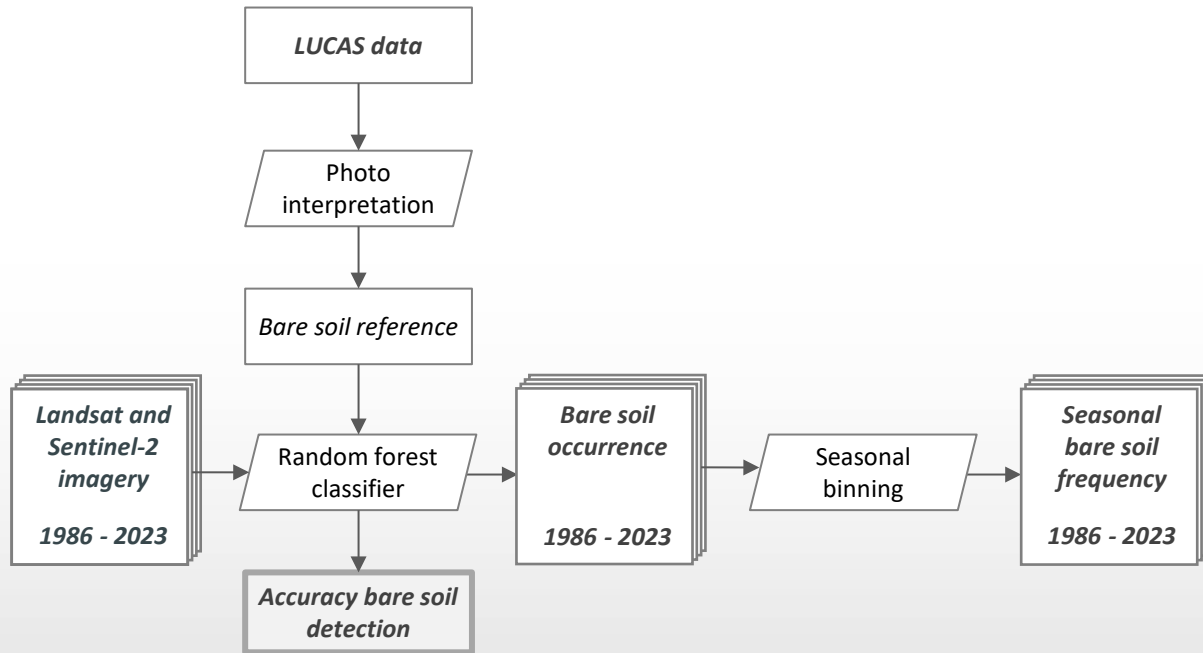


- Random forest model for bare soil detection on each clear sky observation
- Overall Accuracy: **91.87 ± 0.6%**

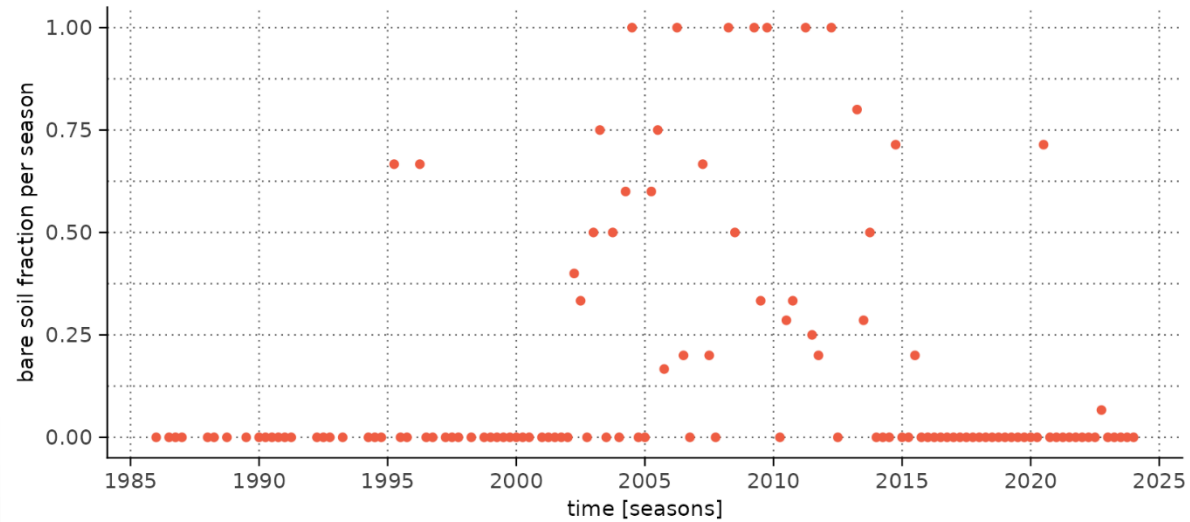
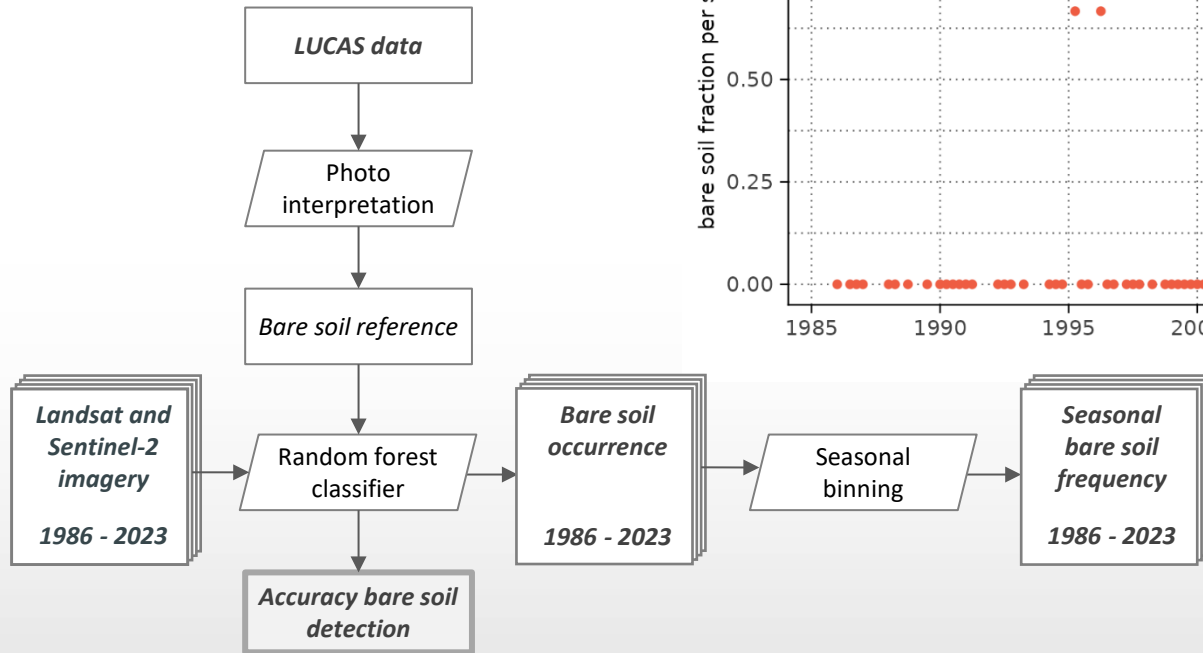
# Workflow



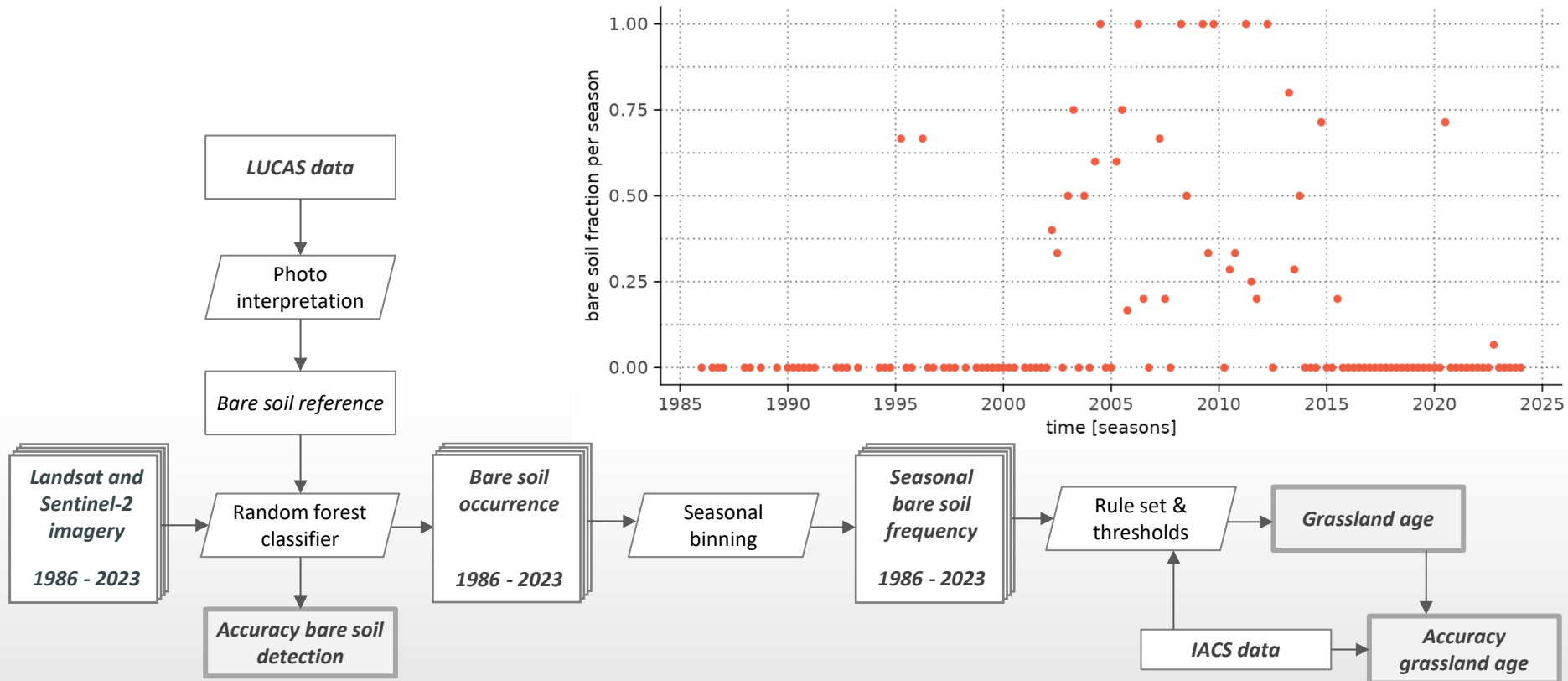
# Workflow



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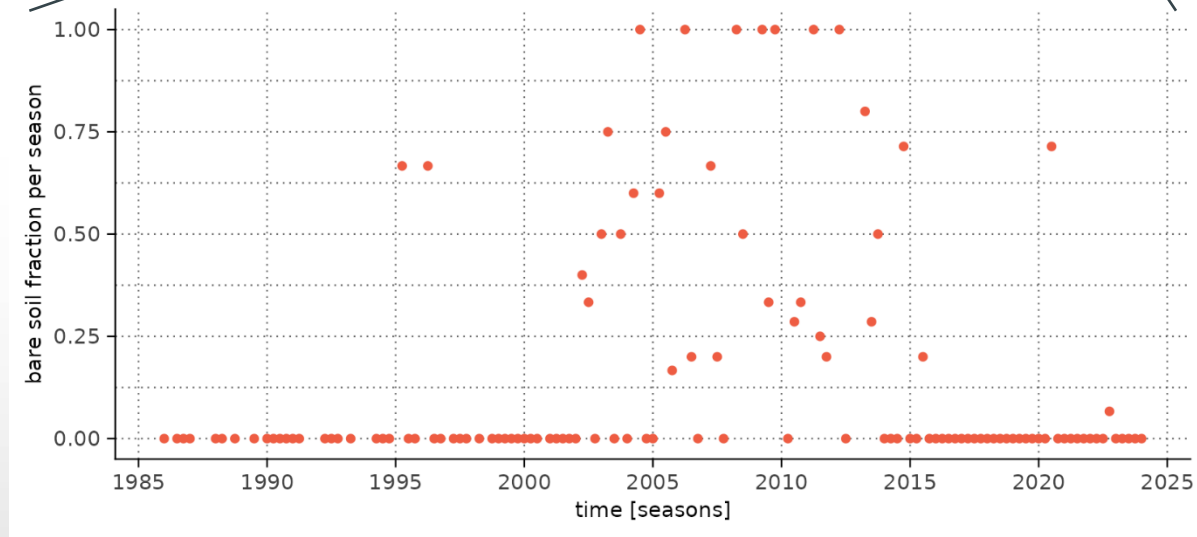


# Workflow



# Age estimation

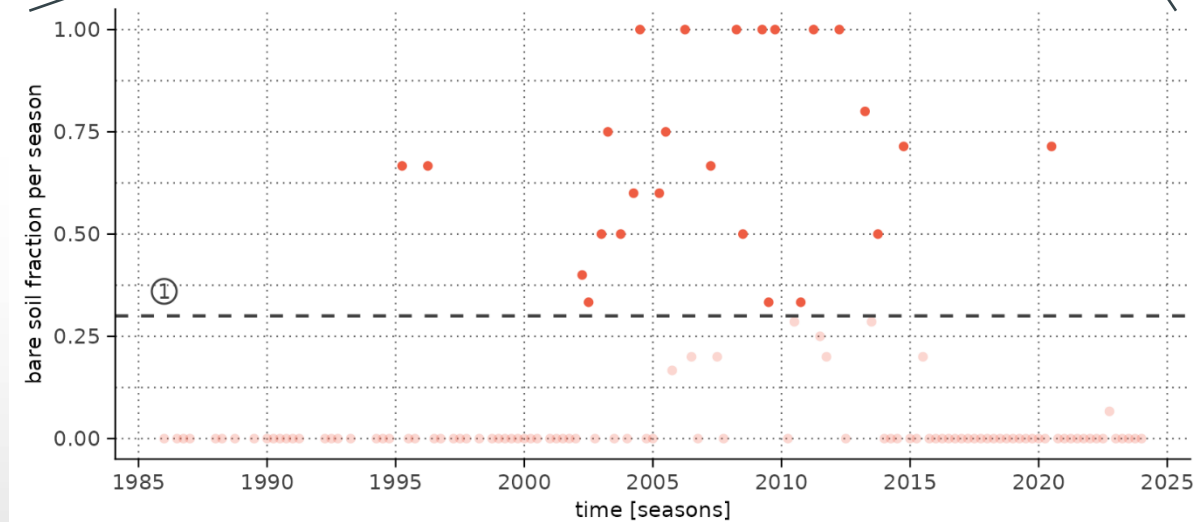
Rule-set to derive grassland age:



# Age estimation

## Rule-set to derive grassland age:

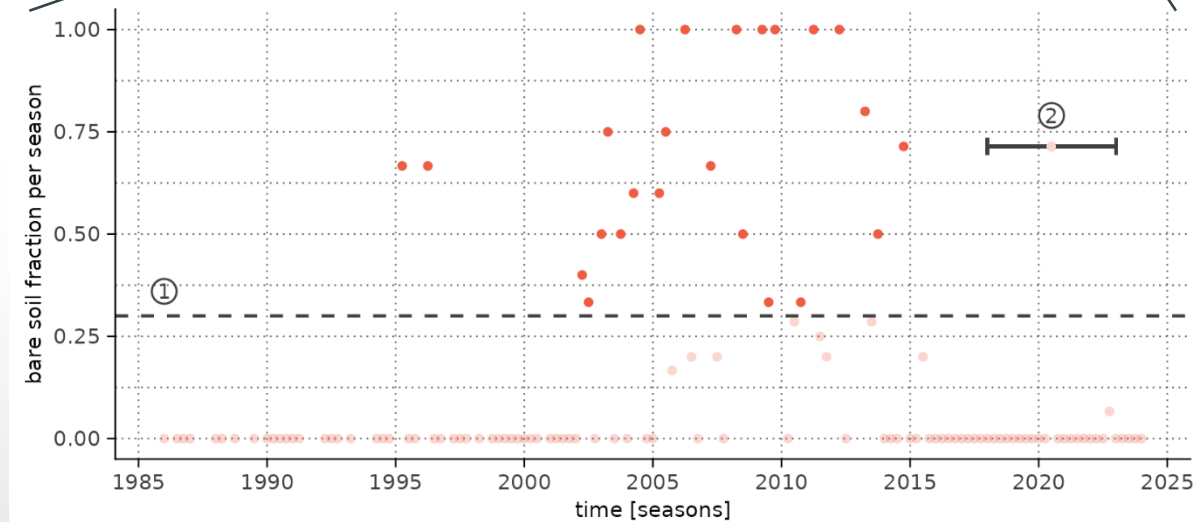
- 1) Bare soil fraction threshold to minimize noise



# Age estimation

## Rule-set to derive grassland age:

- 1) Bare soil fraction threshold to **minimize noise**
- 2) **Exclude management events** by seasonal bare soil count in moving window

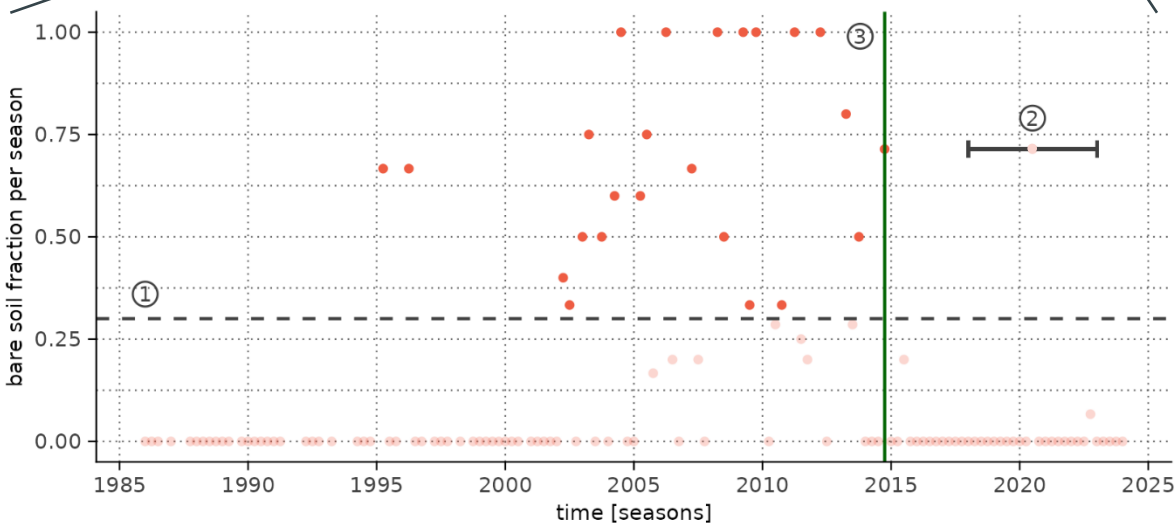
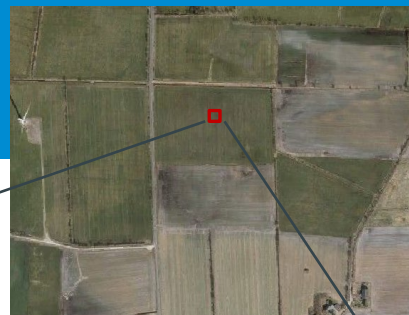


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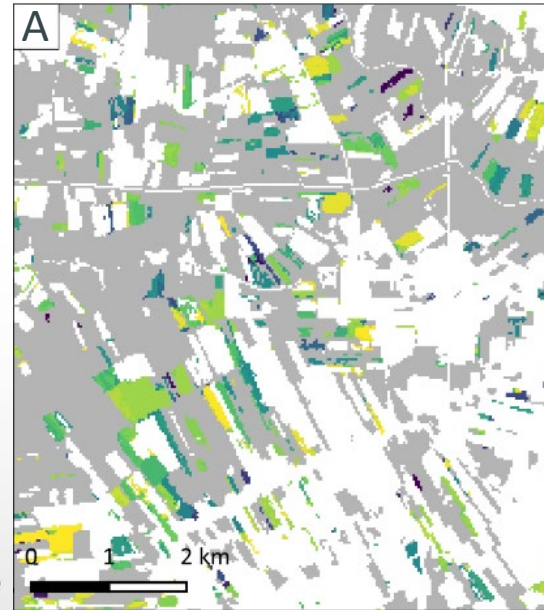
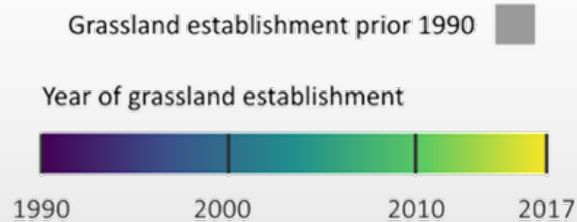
- 1) Bare soil fraction threshold to **minimize noise**
- 2) **Exclude management events** by seasonal bare soil count in moving window
- 3) **Assign grassland establishment** to last bare soil occurrence

Rule-set parametrization empirically against up to 18 years of LPIS data



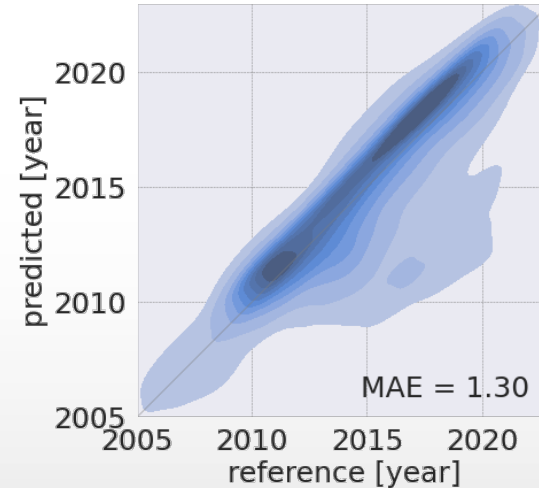
# Grassland age accuracy

- Accuracy for the differentiation of persistent grassland and recent grassland:  **$99.21 \pm 0.02\%$**
- F-score of recent grassland:  **$78.92 \pm 0.64$**

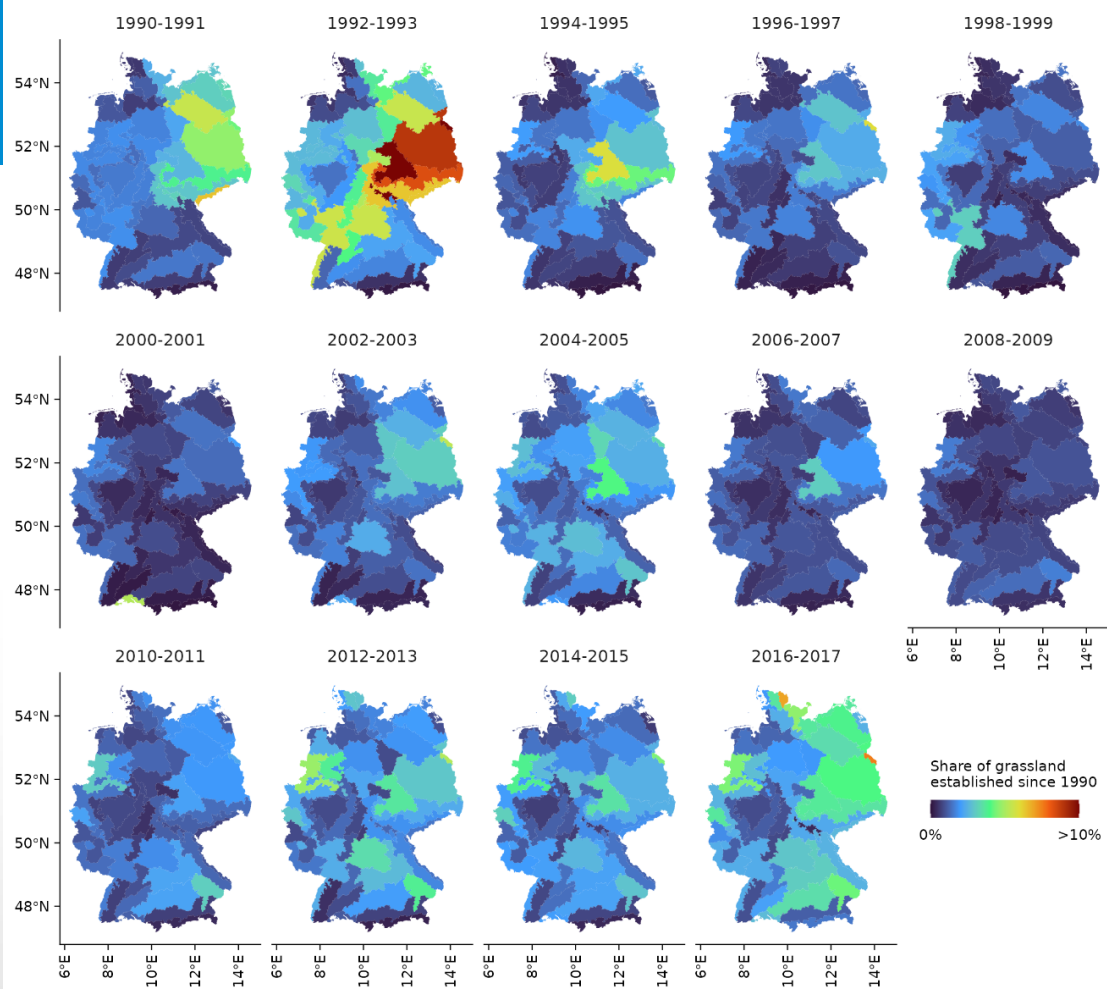


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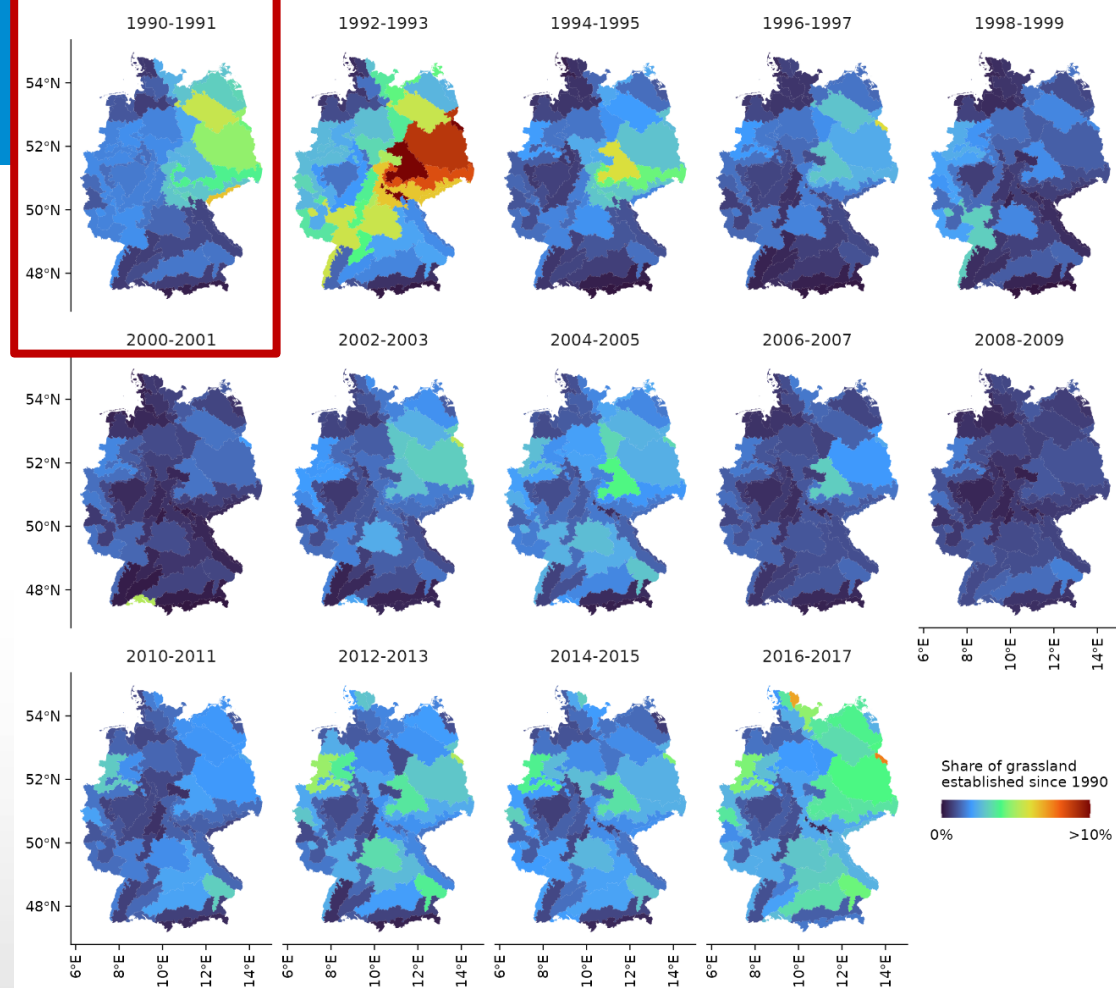


# Large scale patterns



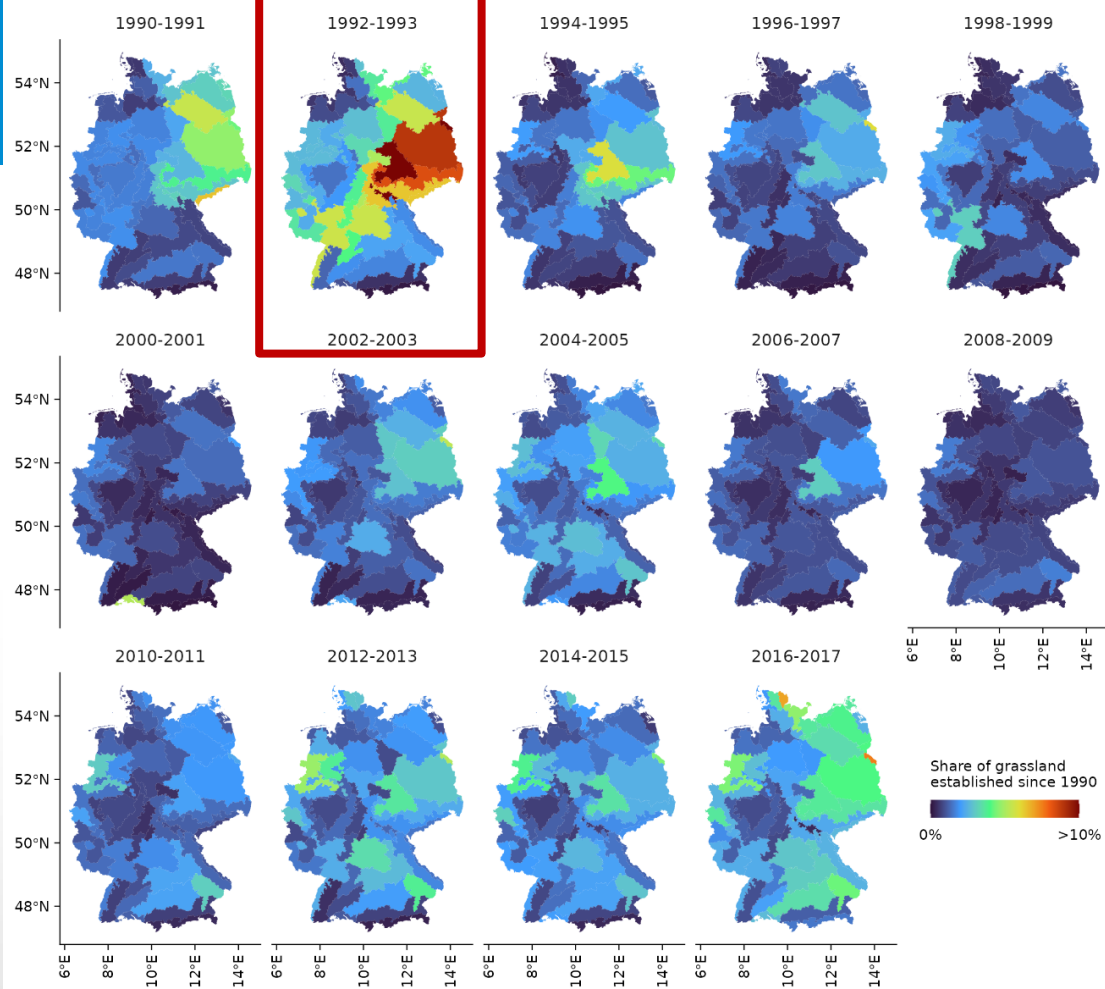
# Large scale patterns

- 1990: higher rates of land abandonment/decline in harvested areas after the collapse of the GDR



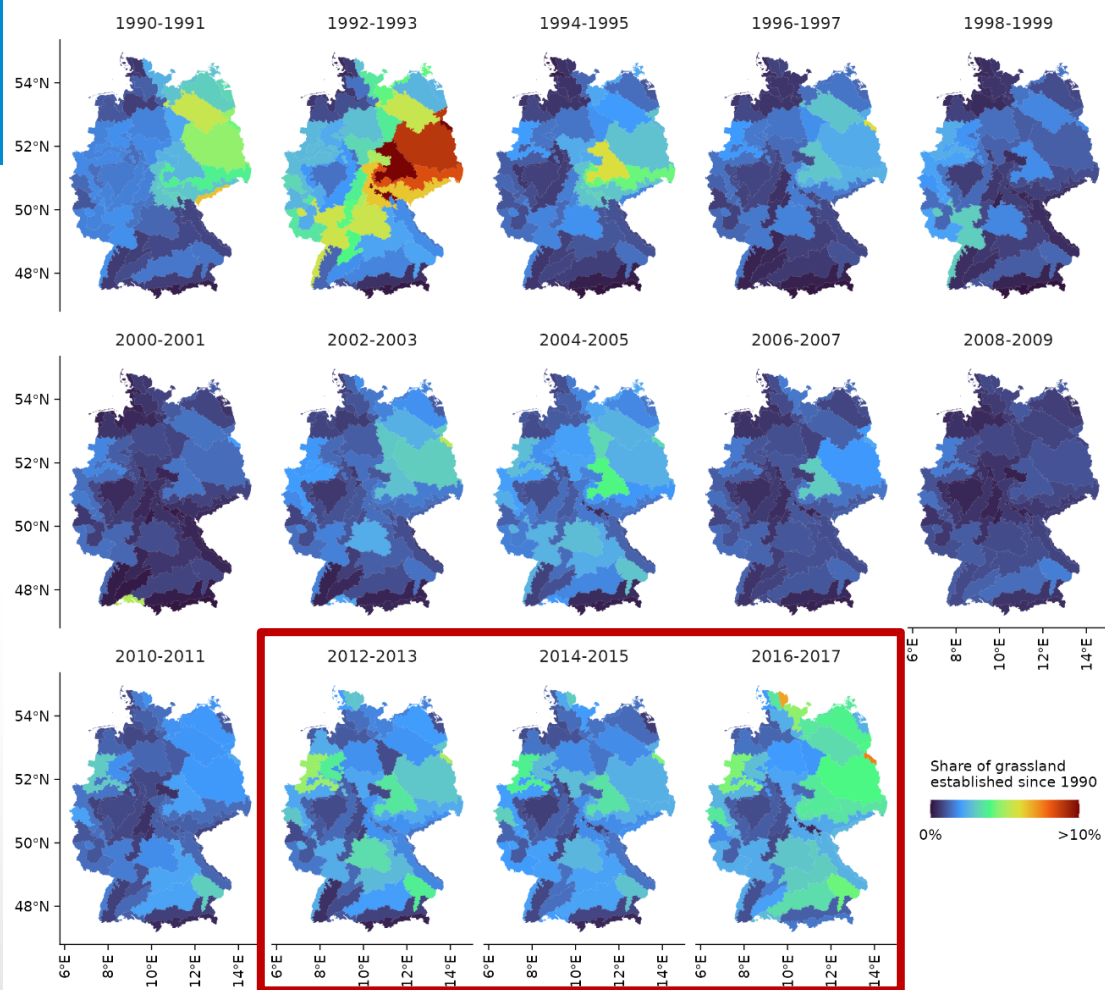
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- 1990: higher rates of land abandonment/decline in harvested areas after the collapse of the GDR
- 1992: CAP McSharry-Reform, exclude arable land from intensive production
- 2013 CAP reform, measures to limit the conversion of grassland to arable land



# Limitations

- The accuracy of grassland age **estimates depends on clear-sky observations** before LU change
- Differentiating bare soil and vegetation land cover is crucial; highest error rates are expected in areas dominated by **non-photosynthetically active vegetation and dry soil**
- Approach **is focused on** grassland establishment within **the agricultural landscape**

# Concluding:

- High accuracy and spatial consistency, even in observation-sparse periods
  - **Potential data for climate reporting from 1990 onwards**
- Distinct spatial and temporal grassland age patterns
  - **Spatially explicit data important**
- Extendable to soil management events
  - **Reliable reference data needed**



# Looking forward to discussions! Feel free to get in touch:

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[www.thuenen.de/theo](http://www.thuenen.de/theo)



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# References

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## Remote sensing of agricultural land use for enhanced climate reporting

Gefördert durch:



Bundesministerium  
für Ernährung  
und Landwirtschaft

aufgrund eines Beschlusses  
des Deutschen Bundestages

### SP1: Humus retention and build-up in arable land



### SP2: Conservation of permanent grassland



### SP3: Soil carbon content in arable land



# Reference data - LUCAS

## Land Use and Cover Area frame Survey (LUCAS)

- 2006, 2009, 2012, 2015, 2018
- Based on systematic sampling grid and harmonized (d'Andrimont et al. 2020)
- Systematic photo collection of survey point (d'Andrimont et al. 2022)



**Dominant LC:**

bare soil<sup>[1]</sup>

photosynthetic vegetation<sup>[2]</sup>

non photosynthetic vegetation<sup>[3]</sup>

other<sup>[4]</sup>

**Optional tags:**

on edge<sup>[5]</sup>

uncertain decision<sup>[6]</sup>

not relevant<sup>[4]</sup>

**Secondary LC:**

bare soil<sup>[4]</sup>

photosynthetic vegetation<sup>[4]</sup>

non photosynthetic vegetation<sup>[4]</sup>

none<sup>[5]</sup>

**additional notes:**

**Minor LC:**

bare soil<sup>[4]</sup>

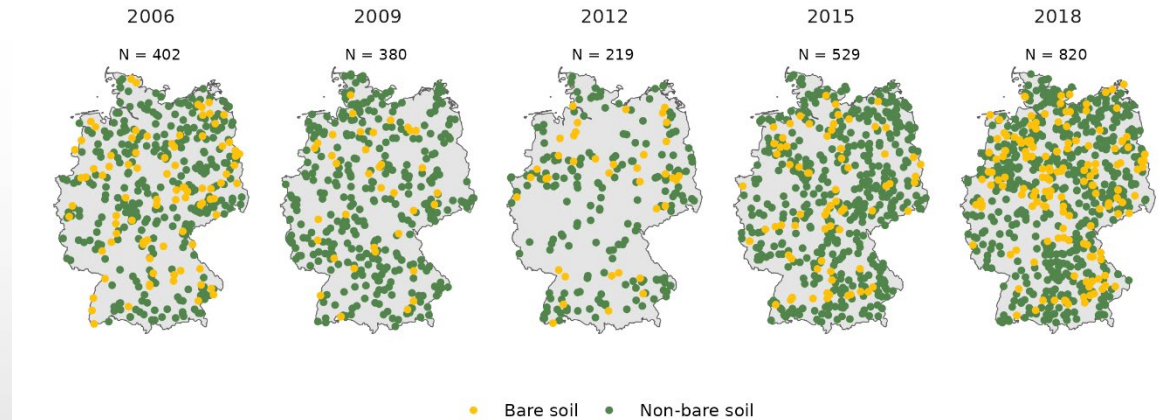
photosynthetic vegetation<sup>[4]</sup>

non photosynthetic vegetation<sup>[4]</sup>

none<sup>[5]</sup>

# Reference data - LUCAS

- Random sample of 2.500 survey plots with landscape photo
  - Grassland and Cropland land cover type
  - Clear sky observation within 2 weeks of survey date



# Sentinel-2 and Landsat data

- Landsat 5-9 and Sentinel-2 for 1986–2023
- Preprocessed using FORCE (Fantz et al., 2019)
- Landsat adjusted to reflectance of Sentinel-2 following Okujeni et al. (2024)
- Feature selection following Azzari et al. (2019)

→ Random forest model for bare soil detection on each clear sky observation

Indices	Formula	Source
NBR2	$(SWIR1 - SWIR2)/(SWIR1 + SWIR2)$	Broeg et al. (2024), Zepp et al. (2023), Azzari et al. (2019), Sorenson et al. (2021)
NDVI	$(NIR - RED)/(NIR + RED)$	Broeg et al. (2024), Sorenson et al. (2021), Mzid et al. (2021)
BCC	$BLUE/(RED + GREEN + BLUE)$	Broeg et al. (2024)
RG_diff	$RED - GREEN$	Broeg et al. (2024)
NDWI	$(GREEN - NIR)/(GREEN + NIR)$	Sorenson et al. (2021)
NDI7	$(NIR - SWIR2)/(NIR + SWIR2)$	Sorenson et al. (2021), Azzari et al. (2019)
PV_BLUE	$((NIR - RED)/(NIR + RED)) + ((NIR - BLUE)/(NIR + BLUE))$	Zepp et al. (2023)
PV_IR2	$((NIR - RED)/(NIR + RED)) + ((NIR - SWIR2)/(NIR + SWIR2))$	Zepp et al. (2023)
BSI	$((SWIR1 + RED) - (NIR + BLUE))/((SWIR1 + RED) + (NIR + BLUE))$	Mzid et al. (2021)
EVI	$2.5 * (NIR - RED)/(NIR + 6 * RED - 7.5 * BLUE + 1)$	Azzari et al. (2019)
GCVI	$(NIR/GREEN) - 1$	Azzari et al. (2019)
NDI5	$(NIR - SWIR1)/(NIR + SWIR1)$	Azzari et al. (2019)
CRC	$(SWIR1 - GREEN)/(SWIR1 + GREEN)$	Azzari et al. (2019)
STI	$SWIR1/SWIR2$	Azzari et al. (2019)

# Results – Quality of Bare Soil Detection

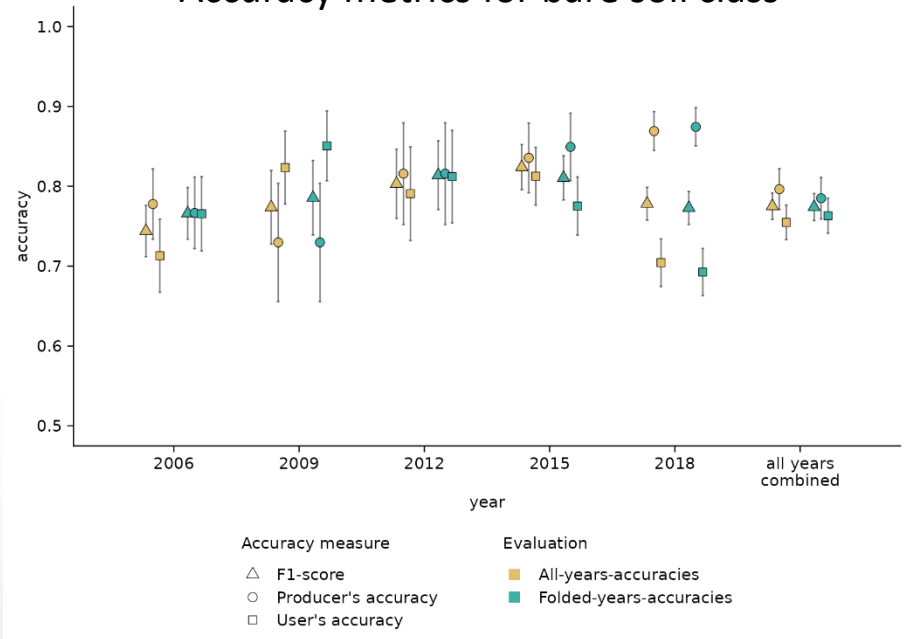
## Final features:

NBR2, RG\_diff, NDWI, and NDI5

**Overall Accuracy:  $92.34 \pm 0.6\%$**

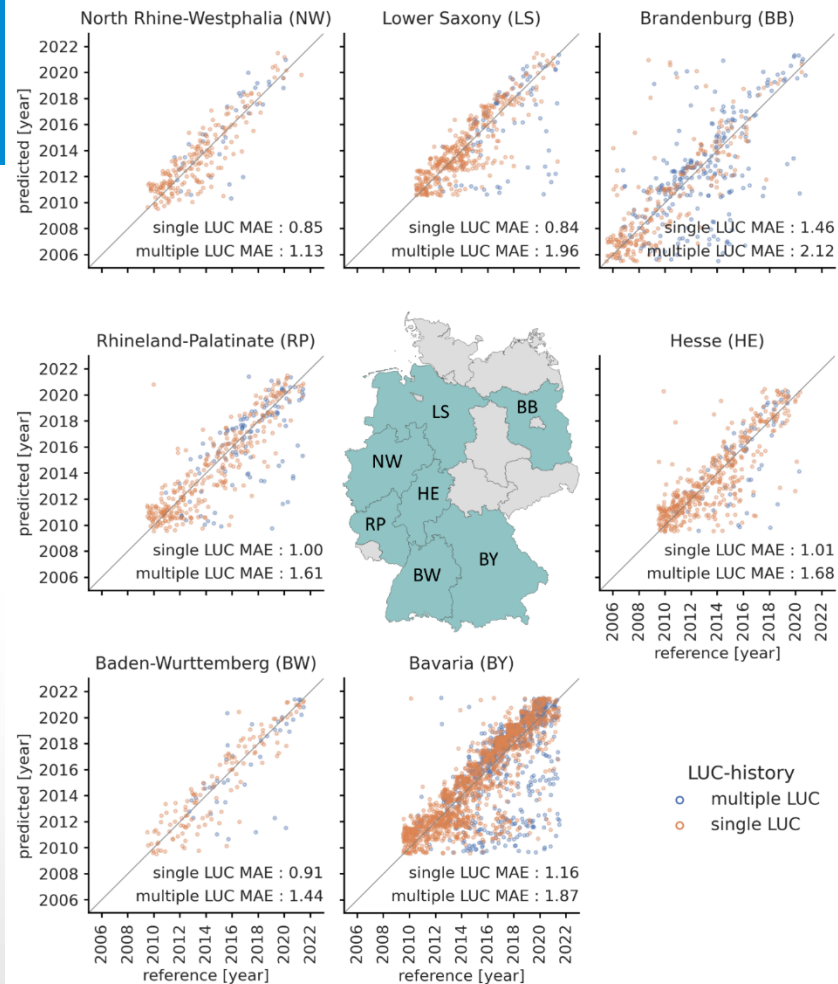
**Overall Accuracy – folded:  $91.87 \pm 0.6\%$**

Accuracy metrics for bare soil class

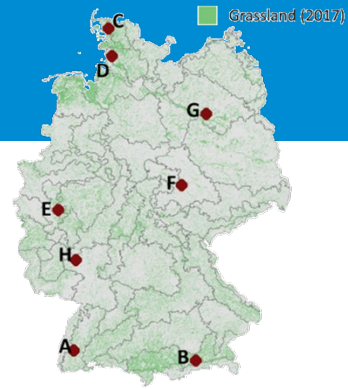
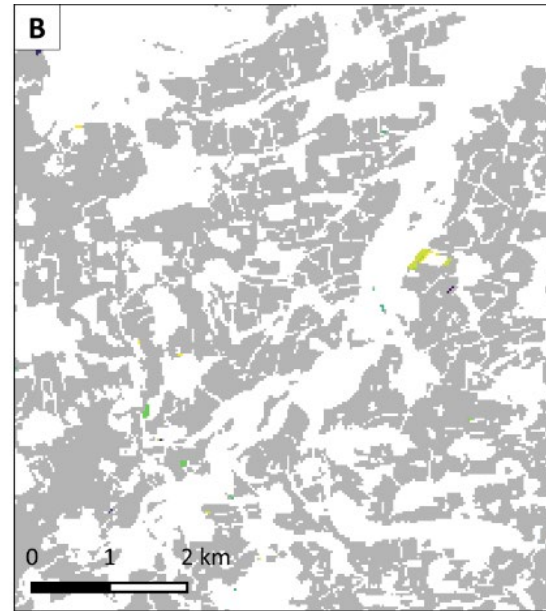
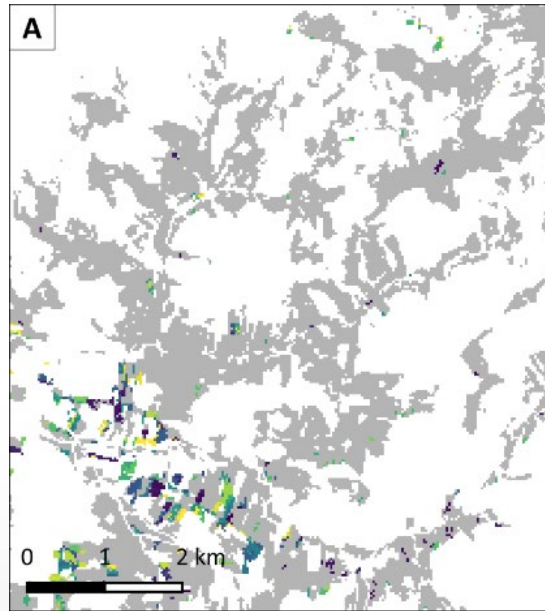


# Grassland age accuracy

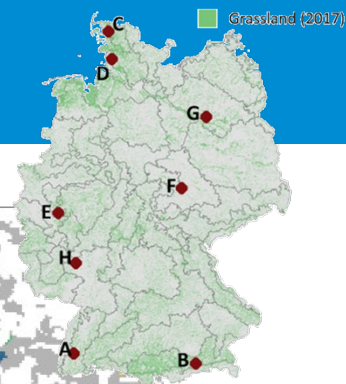
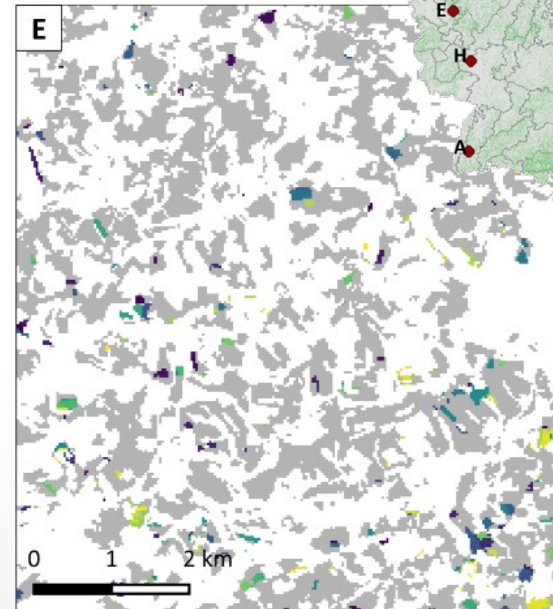
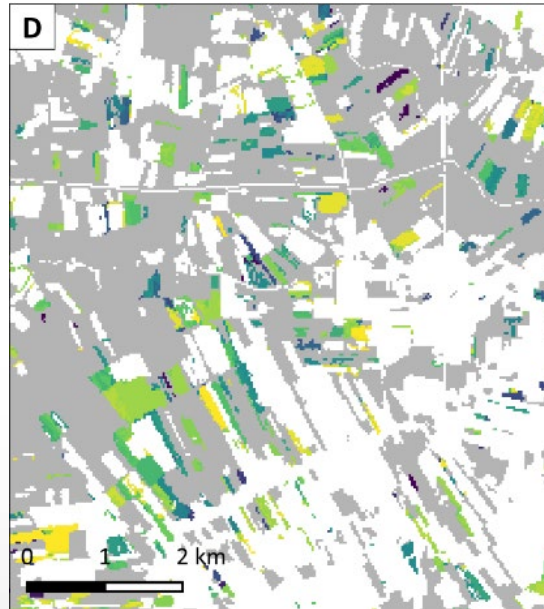
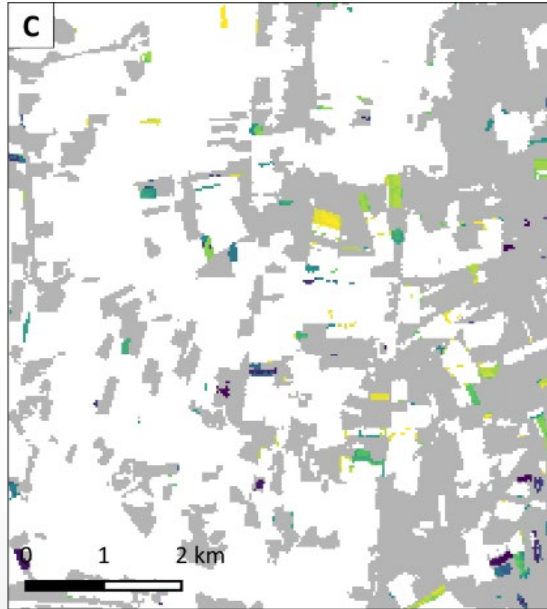
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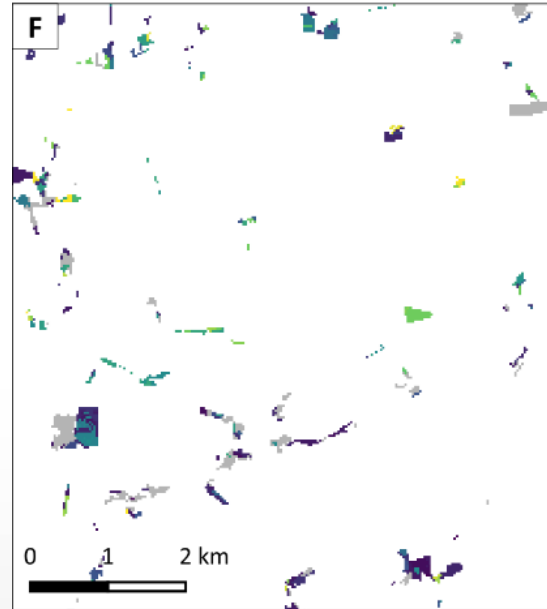
# Mountainous landscape



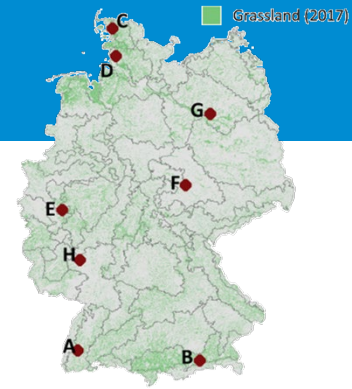
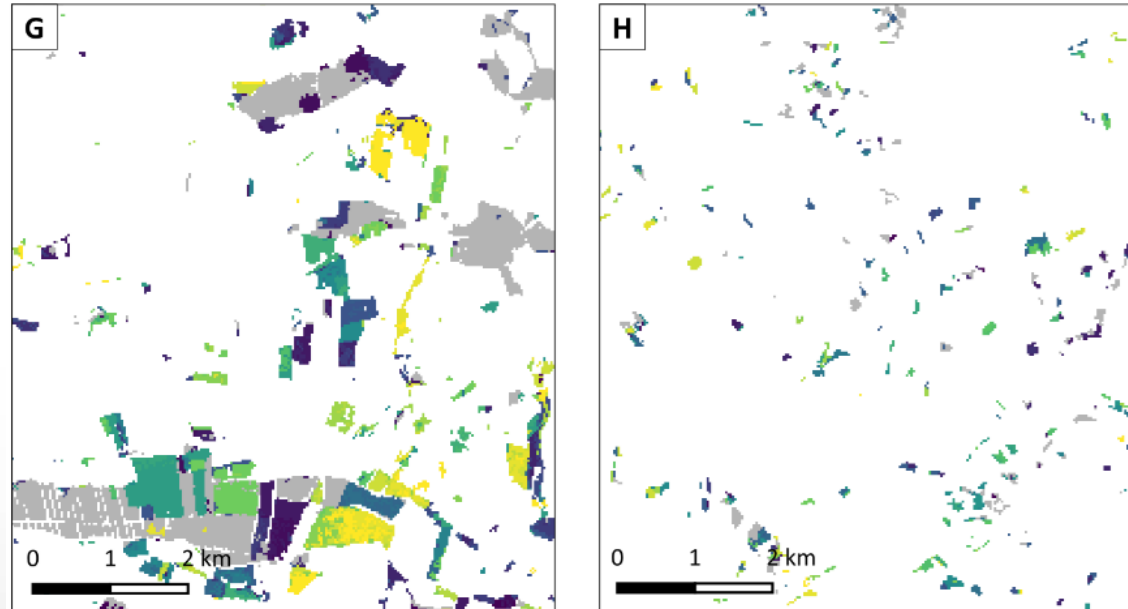
# Costal + low-lying landscape

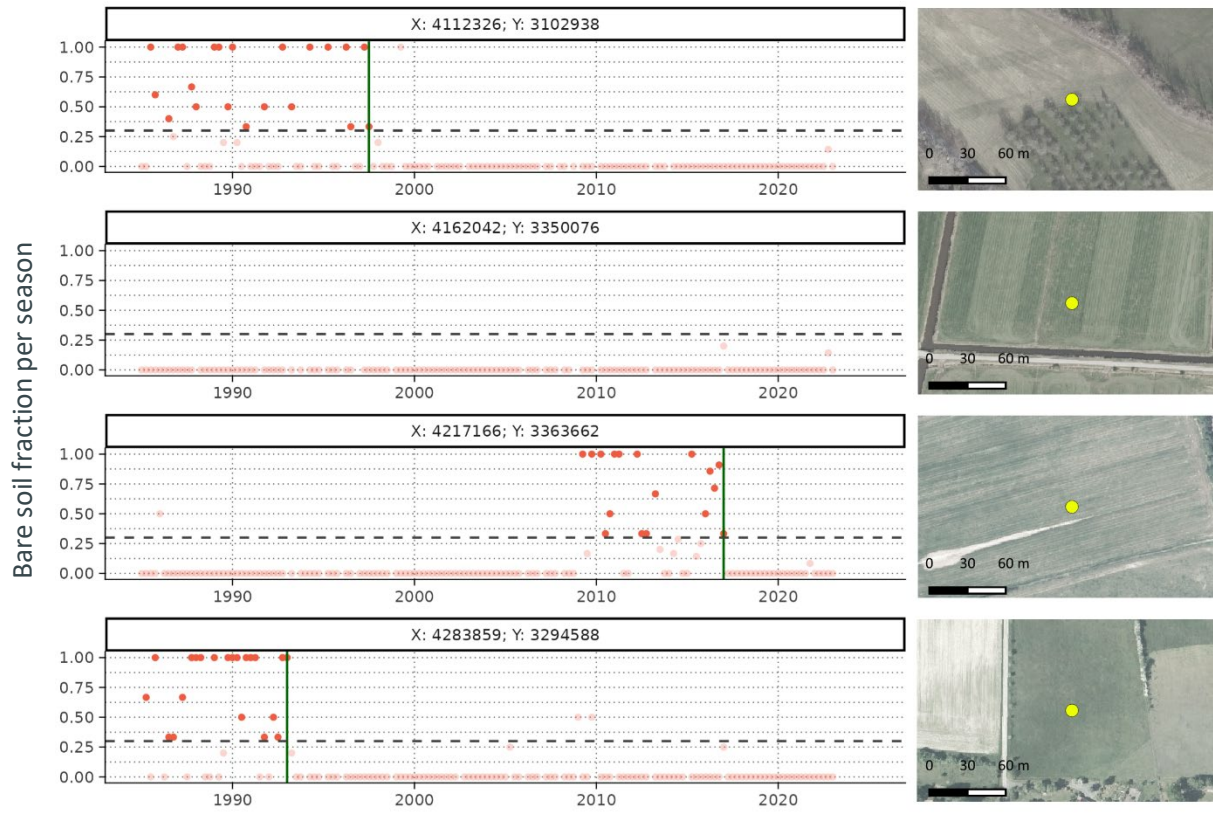


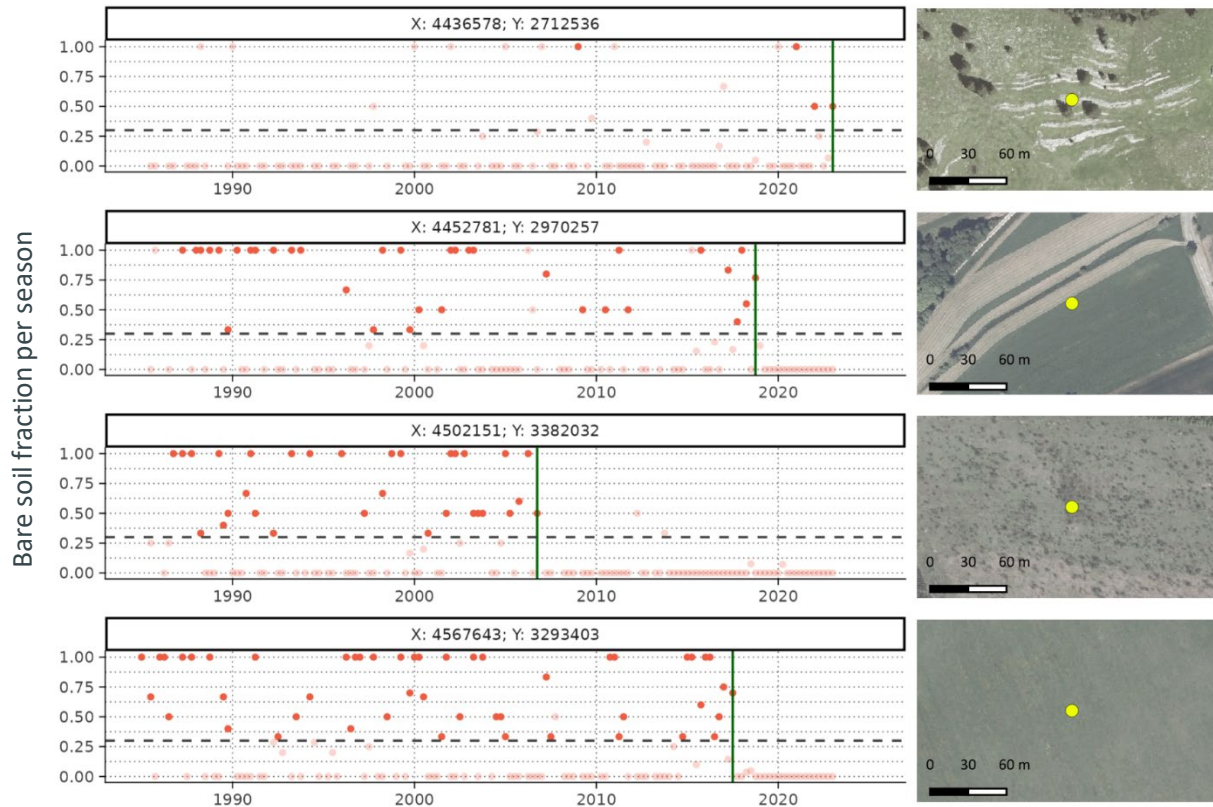
# Cropland dominated region



# Regions with variable field sizes







# IACS reference data

Federal state	Years	IACS-training sample <i>Grassland persistent/recent</i>	IACS-validation sample <i>Grassland persistent/recent</i>
Brandenburg	2006–2022	72,910 / 3,951	31,247 / 1,693
Baden-Württemberg	2010–2022	81,500 / 1,180	34,929 / 506
Bavaria	2010–2022	324,210 / 9,660	138,947 / 4,140
Hesse	2010–2021	127,609 / 3,602	54,690 / 1,544
Lower Saxony	2011–2022	110,073 / 2,055	47,174 / 880
North Rhine-Westphalia	2010–2022	68,267 / 1,250	29,257 / 536
Rhineland-Palatinate	2010–2022	52,064 / 3,270	22,313 / 1,402

# Data availability

	1986-1998	1999-2011	2012	2013-2014	2015-2016	2017-2020	2021-2023
Satellite	Landsat 5	Landsat 5 Landsat 7	Landsat 7	Landsat 7 Landsat 8	Landsat 7 Landsat 8 Sentinel-2A	Landsat 7 Landsat 8 Sentinel-2A Sentinel-2B	Landsat 8 Landsat 9 Sentinel-2A Sentinel-2B
Ø CSO-count:							
Dez. - Feb.	0.6	0.9	0.2	1.2	1.9	5.1	5.2
Mar. - May	2.3	3.6	2.4	3.6	6.4	14.1	15.9
Jun. - Aug.	2.9	4.1	1.4	5.5	9.3	14.6	16.5
Sep. - Nov.	1.7	3.4	1.9	2.8	6.6	10.7	10.9

