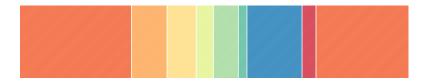
Coupling of phenological information and synthetically generated time-series for crop types as indicator for vegetation coverage information

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Outline

- Introduction
 - Motivation
 - Objectives
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 - PHASE
 - STARFM
 - Sampling
- Results
 - Mapping results and FVC regression model
 - Modelled phenological phases
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Soil erosion, phenology and vegetation coverage



- Soil erosion is an event-based process which is mainly controlled by the simultaneous occurrence of heavy rain events and low soil coverage.
- Extensive phenological information can be used as an indicator for time frames of high erodibility.

Coupling of crop phenology and NDVI profiles



- Automatic interpolation of phenological phases
- Generating of NDVI profiles derived from time-series of high temporal resolution
- Development of a phase-specific fractional vegetation coverage model

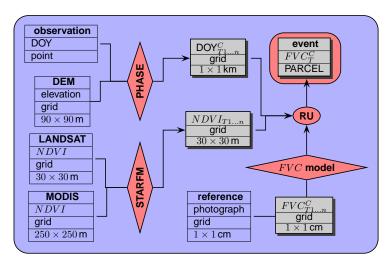


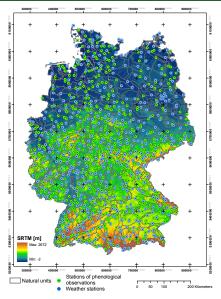
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Workflow





Point data sets

- Phenological observation network of the German Weather Service (DWD)
 - phenological observations sampled by \approx 1200 volunteers
 - positional accuracy: 2 5 km
- DWD meteorological observation network
 - $\bullet~\approx 500~stations$
 - daily mean temperatures

Spatial data

 SRTM DEM 90 × 90 m resampled to 1 × 1 km

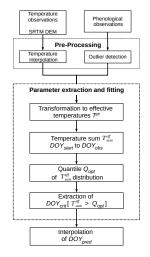


Reported phenological phases for eight typical field crops in Germany

crop	scientific name	ID	reported phases
Winter wheat	Triticum aestivum L.	202	10, 12, 15, 18,19, 21, 24
Winter rye	Secale cereale L.	203	10, 12, 15, 5, 6, 18, 21, 24
Winter barley	Hordeum vulgare L.	204	10, 12, 15, 18, 21, 24
Oilseed rape	Brassica napus L.	205	10, 12, 14, 67, 17, 5, 22, 24
Oat	Avena sativa L.	208	10, 12, 15, 66, 19, 21, 24
Maize	Zea mays L.	215	10, 12, 67, 5, 65, 19, 20, 21, 24
Potato	Solanum tuberosum L.	234	10, 12, 5, 24
Sugar beet	Beta vulgaris subsp. vulgaris	253	10, 12, 13, 24

5 – Beginning of flowering | 6 – full flowering | 10 - Beginning of sowing | 12 - emergence | 13 - closed stand | 15, 67 - beginning of shooting/stem elongation | 17 - beginning of bud formation | 18 - beginning of heading | 19 - beginning of milk ripening | 20 - beginning of early dough ripening | 21 - beginning of yellow ripening | 22 - beginning of full ripening | 24 - harvest | 65 - beginning of tassel emergence

Model structure



Growing Degree Day approach

$$GDD = 0.5 \times (T_{max} - T_{min}) - T_B$$

Model core

$$T_{sum}^{eff}[j] \ge Q_{opt} \left[\sum_{i=DOY_{start}}^{DOY_{obs}} \left((\overline{T}_{i,j} - T_B) \times \frac{DL_i}{24} \right) \right]$$

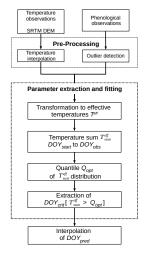


Gerstmann, H., Möller, M., Doktor, D. & Gläßer, C. (20??). PHASE: A geostatistical model for the Kriging-based spatial prediction of crop phenology using public phenological and climatological observations. Computers and Electronics in Agriculture, submitted.

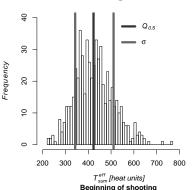


Möller, M., Möller, S., Doktor, D., Gläßer, C. (2011): Automatic interpolation of phenological phases in Germany. In MultiTemp2011, 6th International Workshop on the Analysis of Multi-temporal Remote Sensing Images (pp. 37-40), Trento/Italy,

Model structure

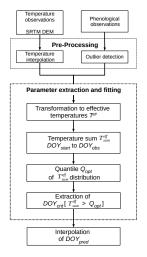


Statistical filtering

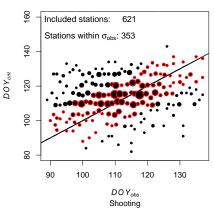


Distribution of T_{sum}^{eff} calculated for all phenological stations from sowing to the observation date for shooting of Winter Wheat in 2011

Model structure



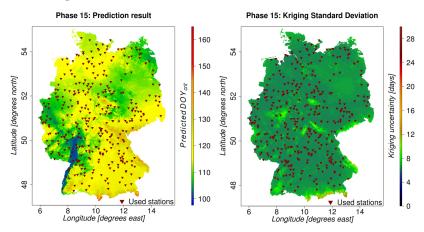
Statistical filtering



Density plot of DOY_{crit} and DOY_{obs} for shooting of Winter Wheat in 2011

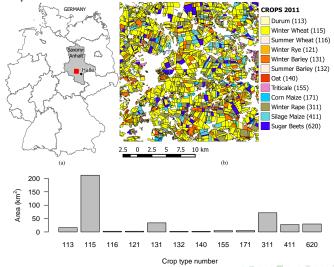


Shooting of Winter Wheat in 2011

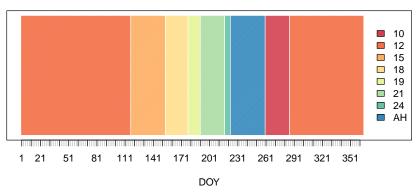


Accuracy metrics

Plant	Phase	T_B	Q_{opt}	nmax	$RMSE_{mod}$	R^2_{mod}	$RMSE_{raw}$	R_{raw}^2
Winter wheat	emergence	0	0.35	50	12.66	0.91	10.90	0.18
	shooting	0	0.55	25	6.11	0.59	10.46	0.03
	heading	8	0.50	40	2.75	0.69	5.51	0.24
	milk ripening	6	0.45	50	3.33	0.58	9.25	0.03
	yellow ripening	0	0.45	13	3.88	0.72	7.29	0.13



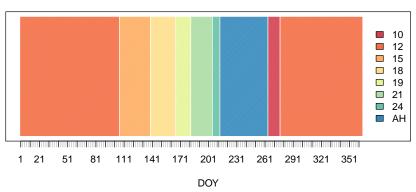
2010



10 - sowing | 12 - emergence | 15 - shooting | 18 - heading | 19 - milk ripening | 21 - yellow ripening | 24 - harvest



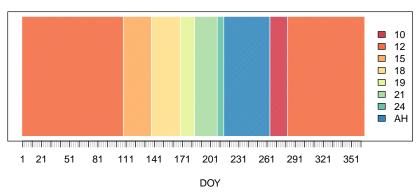
2011



 $10-sowing \mid 12-emergence \mid 15-shooting \mid 18-heading \mid 19-milk \ ripening \mid 21-yellow \ ripening \mid 24-harvest$



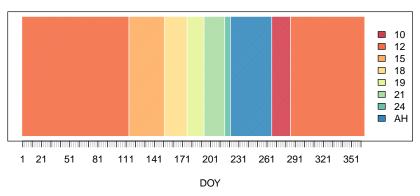
2012



10 - sowing | 12 - emergence | 15 - shooting | 18 - heading | 19 - milk ripening | 21 - yellow ripening | 24 - harvest



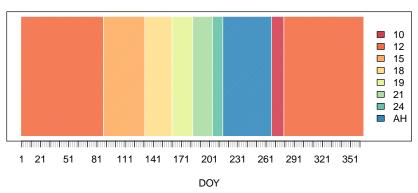
2013



10 - sowing | 12 - emergence | 15 - shooting | 18 - heading | 19 - milk ripening | 21 - yellow ripening | 24 - harvest



2014

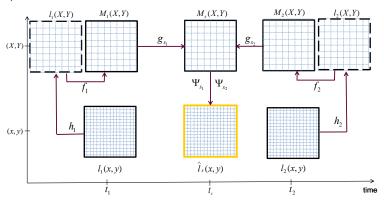


 $10-sowing \mid 12-emergence \mid 15-shooting \mid 18-heading \mid 19-milk \ ripening \mid 21-yellow \ ripening \mid 24-harvest$



Spatial and Temporal Adaptive Reflectance Fusion Model

Spatial resolution



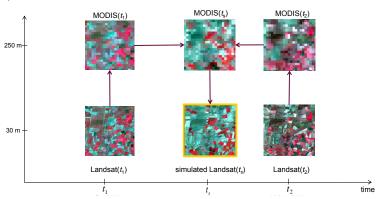


Gao, F., Masek, J., Schwaller, M., Hall, F., 2006. On the blending of the Landsat and MODIS surface reflectance: Predicting daily Landsat surface reflectance. IEEE Transactions on Geoscience and Remote Sensing 44, 2207–2218.



Spatial and Temporal Adaptive Reflectance Fusion Model

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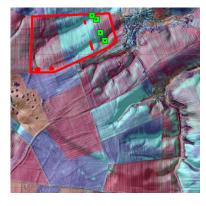




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Photo spots

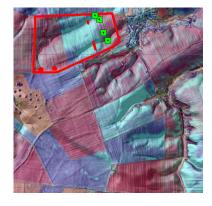


Mapping App



http://www.umgeodat.de/?q=node/8

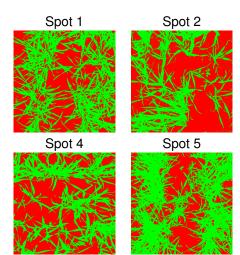
Photo spots





Winter Wheat 2014



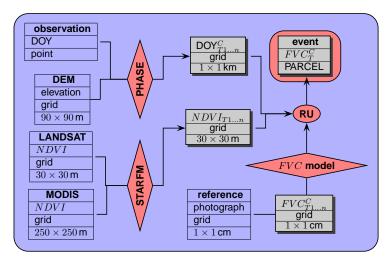


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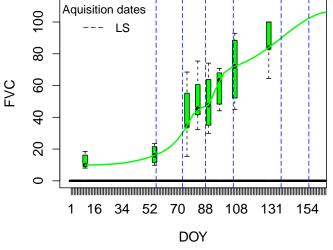
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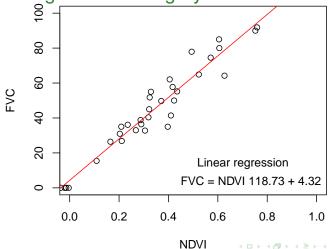
Workflow



Mapping results: Winter Wheat in spring 2014

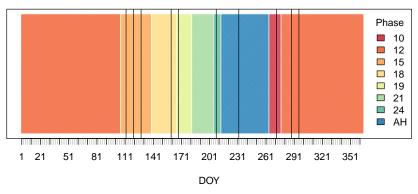


FVC model based on samples of 2013/2014 and corresponding Landsat imagery



Winter Wheat 2011

available Landsat 7 imagery

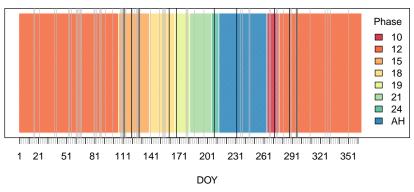


 $10-sowing \mid 12-emergence \mid 15-shooting \mid 18-heading \mid 19-milk \ ripening \mid 21-yellow \ ripening \mid 24-harvest$



Winter Wheat 2011

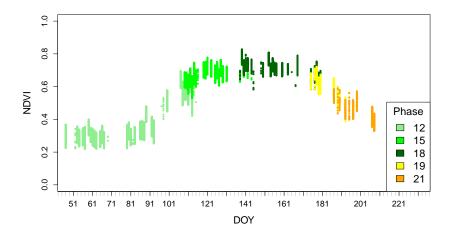
simulated Landsat 7 imagery (cloud cover < 70 %)



 $10-sowing \mid 12-emergence \mid 15-shooting \mid 18-heading \mid 19-milk \ ripening \mid 21-yellow \ ripening \mid 24-harvest$



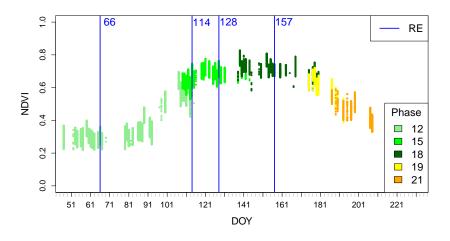
25 % and 75 % NDVI quantiles





Coupling of crop phenology and simulated NDVI profiles

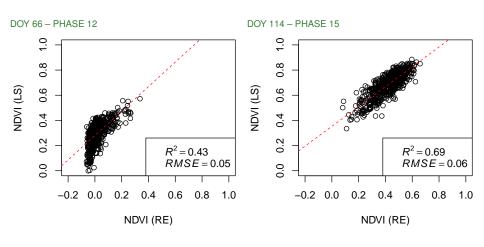
Validation: available RapidEye imagery



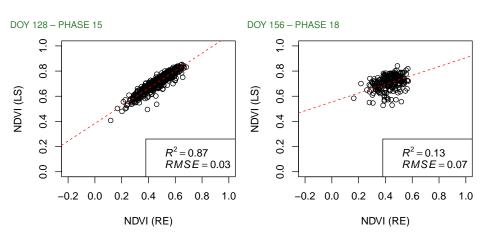


Coupling of crop phenology and simulated NDVI profiles

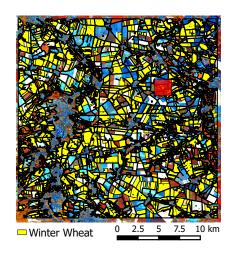
Validation: scatterplots

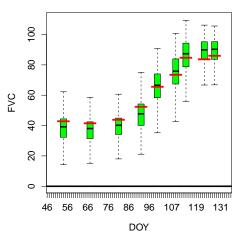


Validation: scatterplots



FVC prediction for phase 12 and 15 in 2011







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Conclusion

- Vegetation indices show a high correlation to fractional vegetation cover during green phenological phases.
- We have shown how such relevant time periods can be identified by coupling phenological phases and simulated NDVI time series over a specific vegetation period.

Outlook

- effective sampling
- application for crop residue coverages
- implementation in a WebGIS environment



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Thank you for your attention!

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Temporal and spatial C-factor derivation by using Sentinel 2 imagery for the localization of soil erosion events by water http://paradigmaps.geo.uni-halle.de/dynac



Phenological structuring of high temporal Sentinel-2 satellite imagery to improve land cover classification http://paradigmaps.geo.uni-halle.de/phenos

