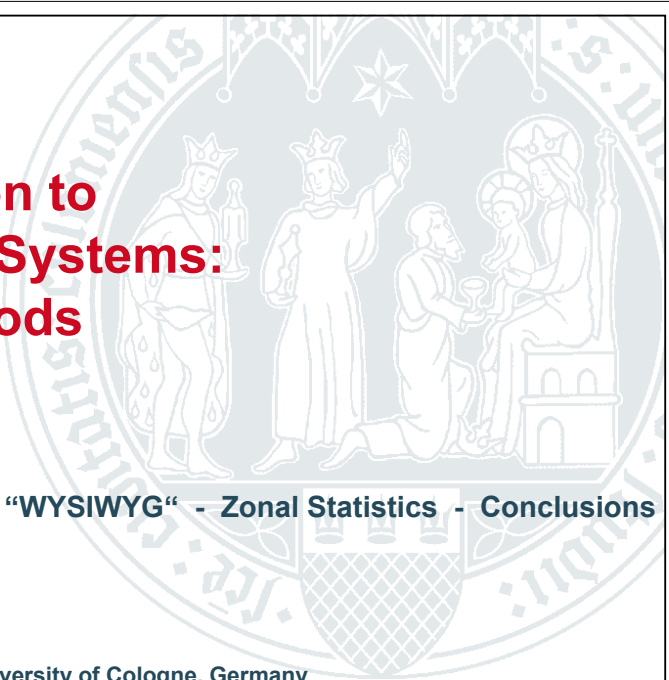


3D Data Acquisition to Monitor Cropping Systems: Sensors and Methods



Motivation - Methods - Results - "WYSIWYG" - Zonal Statistics - Conclusions

Georg Bareth

GIS & RS Group, Institute of Geography, University of Cologne, Germany

55th Photogrammetric Week, 7th - 11th September 2015, Stuttgart

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Motivation

- Precision Agriculture: need of spatial data for within-field variability
- Data demand: vitality, abiotic/biotic stress, biomass, nitrogen ...
- Limitations of PreAg: rice, maize, grassland ...
- Limitation of RS: clouds, repetition, resolution, costs ...
- Opportunities: airborne, field border, super-high resolution ...



Leeb PT 280:
Selbstfahrende Spritze mit großer
Flächenleistung

(www.horsch.com)

Objectives

- 3D-Data: plant height, plant growth, emergence, biomass
- Hyperspectral: chlorophyll, nitrogen, stress, biomass
- 3D + Hyperspec.: vitality, abiotic stress, biotic stress, biomass, nitrogen

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Nitrogen Nutrition Index - NNI

“... optimal nitrogen fertilizer application regimes in crop production have two requirements: (1) knowledge of the adequate N content for a given amount of **biomass** and (2) the development of fast, accurate methods to determine the actual N content and **biomass** (or N demand) of the crop plant”

Bodo Mistele and Urs Schmidhalter (2008): Estimating the nitrogen nutrition index using spectral canopy reflectance measurements. *Europ. J. Agronomy* 29/4: 184-190. DOI: 10.1016/j.eja.2008.05.007

$$NNI = N_{act} / N_c$$

N_{act} = **actual measured N content as a percent of the **dry matter of the canopy biomass****

N_c = **the critical N content for the crops of each plot given their amount of **dry weight****



Sensors



www.tetracam.com

Multispectral



www.canon.com

RGB



www.oculii.com

Radar



www.rikola.fi

Hyperspec



www.agricon.de

Ultrasonic



www.pmdtec.de

TOF



www.riegl.com

Laser scanner



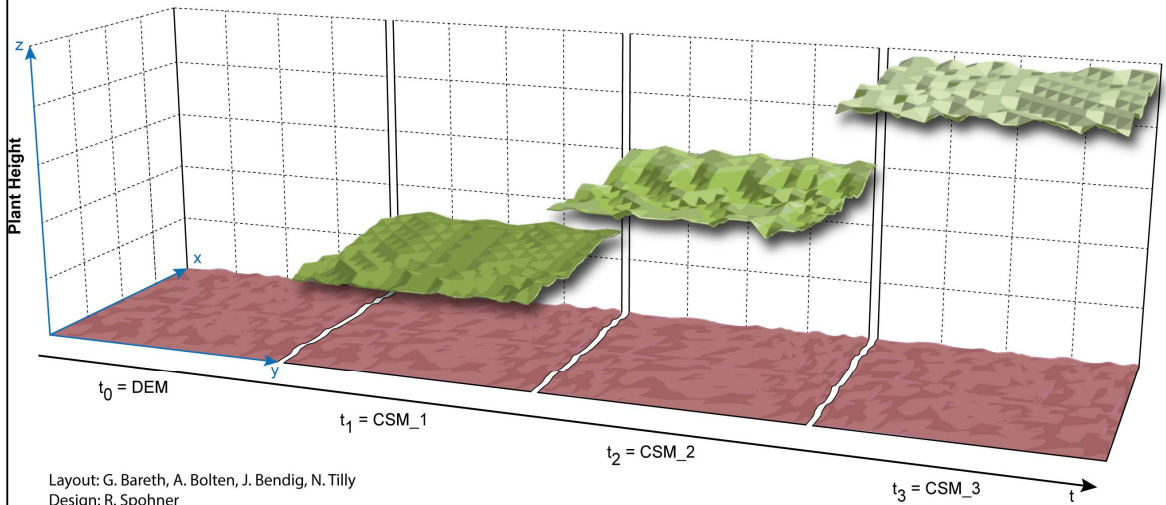
www.agricon.de



Crop Surface Model (CSM)

HOFFMEISTER et al. (2010): High resolution CSM and CVM on field level by terrestrial lasers scanning
In: Proc SPIE, Vol. 7840, 78400E:

Multi-temporal Crop Surface Models



Layout: G. Bareth, A. Bolten, J. Bendig, N. Tilly
Design: R. Spohner

$$\begin{aligned} \text{Plant Height}_{\text{total}} &= t_3 - t_0 \\ \text{Plant Height}_{\text{in-season1}} &= t_1 - t_0 \\ \text{Plant Height}_{\text{in-season2}} &= t_2 - t_0 \end{aligned}$$

$$\begin{aligned} \text{Plant Growth}_{\text{in-season3}} &= t_2 - t_1 \\ \text{Plant Growth}_{\text{in-season4}} &= t_3 - t_2 \\ \text{Plant Growth}_{\text{in-season5}} &= t_3 - t_1 \end{aligned}$$



Terrestrial Laser Scanning (TLS)



Instrumental set-up:

- (A) terrestrial laser scanner Riegl LMS-Z420i
 - (B) tractor with hydraulic platform
 - (C) ranging pole with reflective cylinder
- (Tilly et al. 2015, in print)



Unmanned Aerial Vehicle (UAV)



Sony α 5100

62.8 mm
(2.47")

109.6 mm
(4.31")

- RGB sensor
- CMOS sensor: 24,3 MP (APS-C; 23,5 x 15,6 mm)
- ISO 100-25600
- B x H x T:
109,6 x 62,8 x 35,7 mm
- weight: approx. 283 g
- format: jpeg / raw



- 20 mm / f 2,8; 69 g



Mikrokoopter UAV-platform (< 5 kg):

- rotors: 4 - 12
- payload: 250 g - 2500 g
- weight: 650 g - 1700 g
- flying time: 15 - 45 min
- distance: sight distance
- altitude: 350m
- sensors: gyroscope, acceleration, compass, GPS, barometric altitude



Low-cost UAV Approach

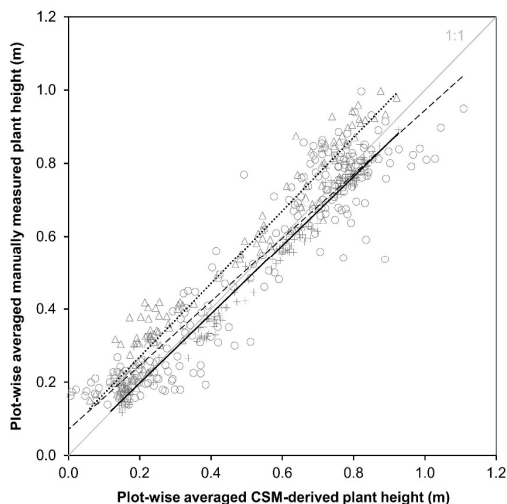
- UAV platform:
 - easy to use
 - lightweight
 - low-cost
 - uncalibrated digital camera



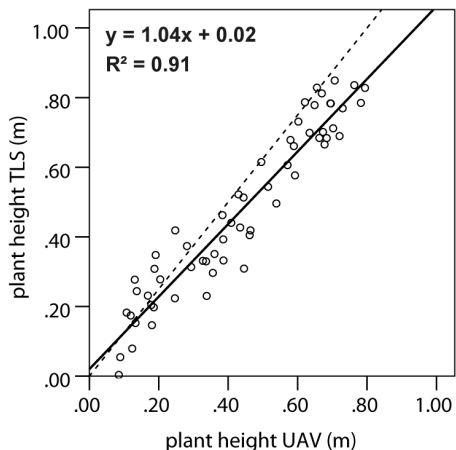
- DJI Phantom 2
- Canon Powershot 110
- Photo Scan
- ArcGIS
- GCP: RTK-referenced
- ASD FieldSpec3



CSM-derived Plant Height (TLS; UAV)



(Tilly et al. 2015, in print)

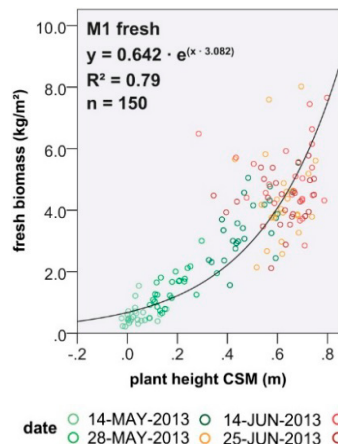


(Bareth et al. 2015, in prep.)



CSM-derived Biomass (UAV)

	Bivariate BRMs				Multivariate BRMs						
	Estimator	Whole period		Pre-anthesis		Estimator ^b	Whole period		Pre-anthesis		
		R ²	SE _E ^a	R ²	SE _E		R ²	SE _E ^a	R ²	SE _E ^a	
Dry biomass	Linear	PH	0.65	10.03	0.76	5.73					
		GnyLi	0.52	11.75	0.68	6.67	GnyLi	0.65	34.63	0.77	25.41
		NDVI	0.07	16.38	0.34	9.56	NDVI	0.69	21.49	0.76	20.73
		NRI	0.54	11.58	0.70	6.40	NRI	0.65	35.04	0.77	24.86
		RDVI	0.13	15.87	0.39	9.21	RDVI	0.69	19.18	0.76	21.40
		REIP	0.12	15.92	0.58	7.60	REIP	0.73	1933.86	0.76	258.29
		RGBVI	0.05	16.55	0.26	10.10	RGBVI	0.68	22.28	0.76	23.23
Exponential		PH	0.84	0.37	0.84	0.34					
		GnyLi	0.80	0.42	0.85	0.32	GnyLi	0.86	2.43	0.88	2.14
		NDVI	0.30	0.77	0.61	0.53	NDVI	0.85	2.84	0.88	3.99
		NRI	0.81	0.40	0.87	0.30	NRI	0.87	2.29	0.89	1.96
		RDVI	0.41	0.71	0.68	0.48	RDVI	0.85	2.52	0.88	2.84
		REIP	0.37	0.73	0.77	0.40	REIP	0.84	30.37	0.86	48.49
		RGBVI	0.23	0.81	0.48	0.60	RGBVI	0.85	2.51	0.87	2.73

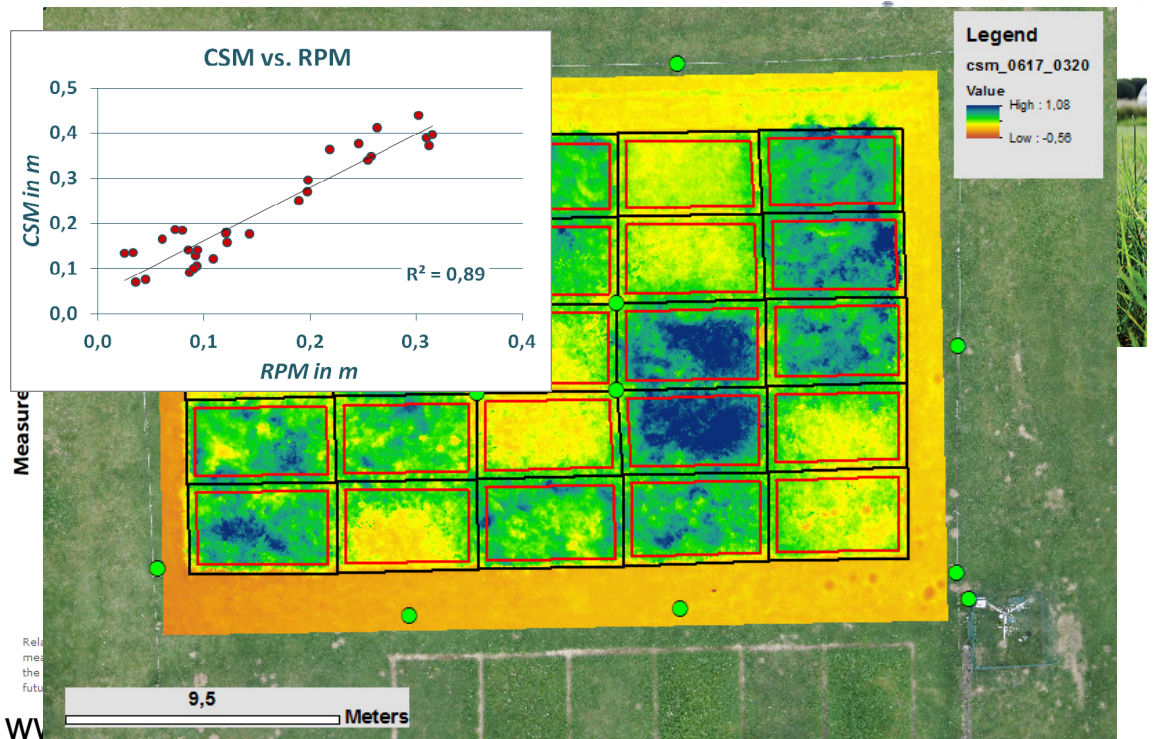


(UAV: Bendig et al. 2014)

(TLS: Tilly et al. 2015, in print)

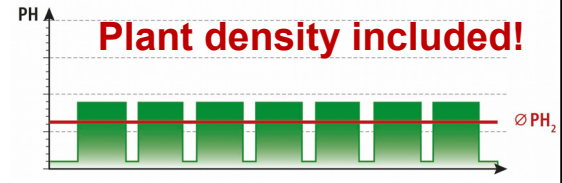
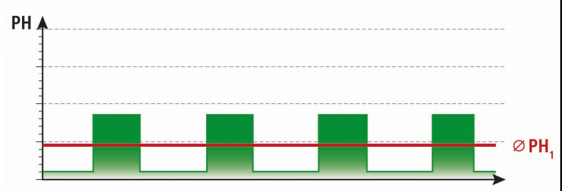
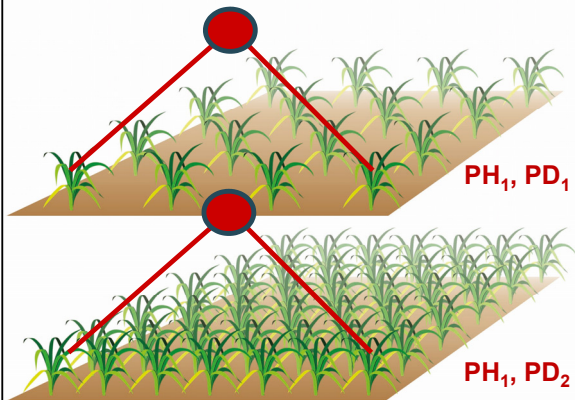


Rising Plate Meter (RPM) vs. UAV-CSM in Grassland

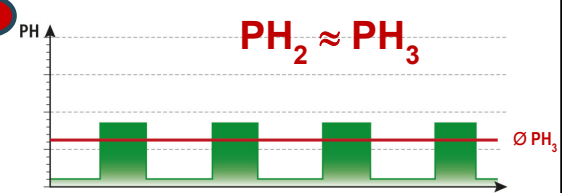
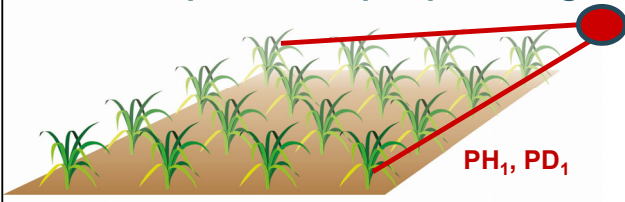


"What you see is what you get!" or "What do we see?"

- Nadir view: equal plant height but different plant density.

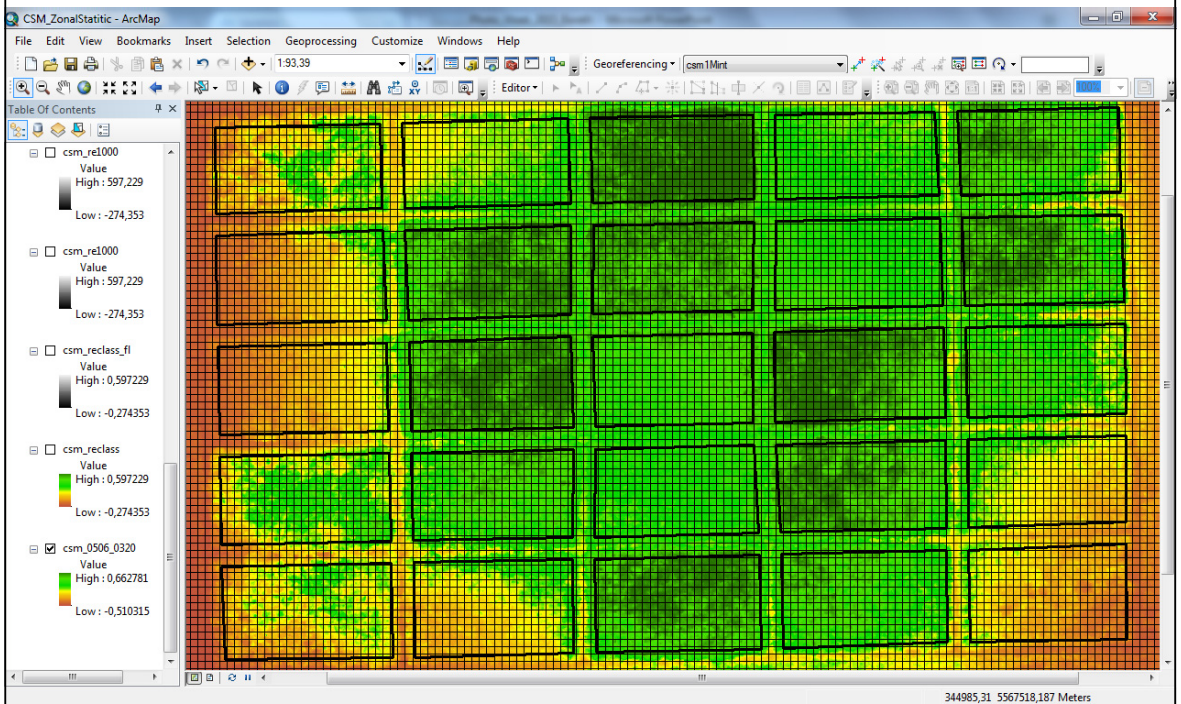


- Oblique view: equal plant height but different plant density.



Polygon Grids - Zonal Statistics

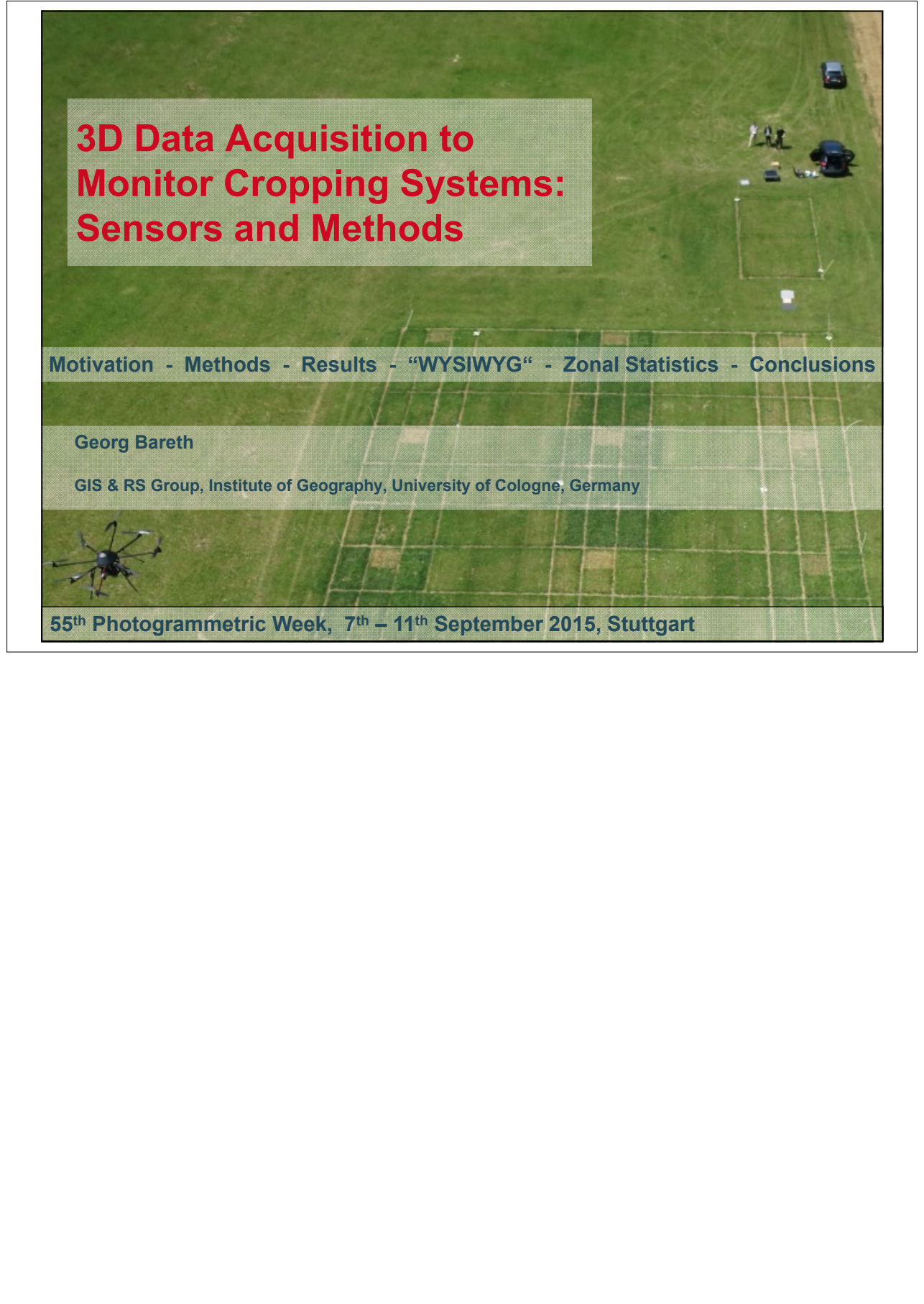
- Zonal statistics is a standard GIS/RS software tool for descriptive statistics



Conclusion & Outlook

- plant height is a robust estimator for biomass
- TLS-/UAV-derived CSMs work well for biomass estimation
- low-cost UAV imaging system also works well
- polygon grids / zonal statistics preserve plant height variability and plant density
- UAV: precise direct georeferencing
- TLS -> MLS -> ALS
- approach works for hyperspectral frame cameras: 3D hyperspectral data (Aasen et al. 2015, in print)





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