



## Low-altitude / high-resolution (drone based) remote sensing for Field-Phenotyping

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- Phenotyping many genotypes across multiple (natural) environments is a lot of work
- Plant growth is a dynamic process
- Objective measures of traits are needed
- Automated, reproducible procedures would help





#### UAV remote sensing for field-phenotyping workflow



Aasen, H., Honkavaara, E., Lucieer, A., Zarco-Tejada, P., 2018. Quantitative Remote Sensing at Ultra-High Resolution with UAV Spectroscopy: A Review of Sensor Technology, Measurement Procedures, and Data Correction Workflows. Remote Sensing



## Outline





## Outline







## **Mission planning**

- Selection of equipment
- Flight planning
- (Legislation, weather, security & health measures)
- Can be quite complex
  - Data product (point cloud, digital surface model, orthophoto)
  - Sensor (point, line or 2d imager)
  - Data type (RGB, spectral, thermal ...)
  - Coverage (flight time, flight speed, altitude)
  - Ground sampling distance (altitude, resolution, motion blur ~ flying speed + integration time)
  - Focus distance and depth of field
  - GCP placement





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a)

. . .



## Flight planning

ground sampling distance 

#### motion blur

ı)					b)					
Flight speed		1 m s <sup>-1</sup>			Shutter speed		1/500 s			
Shutter speed		1/2500 s			Flight height		46 m			
Flight height	19 m	28 m	46 m	93 m	GSD		5 mm			
GSD	2 mm	3 mm	5 mm	10 mm	Flight speed	4 m s <sup>-1</sup>	8 m s <sup>-1</sup>	10 m s <sup>-1</sup>	15 m s <sup>-1</sup>	
Motion blur	20%	13%	8%	4%	Motion blur	160%	320%	400%	600%	
Flight direction			X+	X+	Flight direction					
	During our literature review we found only a few publications are stating									

these quality indicators





- **GCP** placement
- image overlap

L. Roth, A. Hund, and H. Aasen, 2018, "PhenoFly Planning Tool - Flight planning for high-resolution optical remote sensing with unmanned areal systems," Plant Methods.



## Flight planning

#### http://phenofly.net/PhenoFlyPlanningTool



L. Roth, A. Hund, and H. Aasen, 2018. "PhenoFly Planning Tool - Flight planning for high-resolution optical remote sensing with unmanned areal systems," *Plant Methods*."





## **Flight planning**



Helge Aasen | 05.07.2019 | 9



## **Flight planning**





L. Roth, A. Hund, and H. Aasen, 2018. "PhenoFly Planning Tool - Flight planning for high-resolution optical remote sensing with unmanned areal systems," *Plant Methods*."





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  - GCP placement
- Think of it even before you by your equipment

Headwall Nano-Hyperspec **® VNIR** 

Cubert UHD 185

(Aasen et al., 2015)

2015

## **Spectral sensors for UAS RS**



#### Simple consumer oriented systems



Parrot Sequoia / **Micasense Red-Edge** Mutli-spectral 2D imager

Rikola FPI - NIR/SWIR (1100 -1600 nm) 2D Hyperspectral sequential 2D imager (Honkavaara et al., 2016)

۹ TetraCam mini-mca

2013

Multispectral 2D imager (Berni et al., 2009) (Kelcey and Lucieer, 2012)

2009

2014 **Rikola FPI – VNIR** 

2D Hyperspectral sequential imager

(Honkavaara et al., 2013)

Imec filter-on-chip Hyperspectral snapshot 2D

2016

Ø

CUDED

2D Hyperspectral snapshot imager

#### **High-quality systems**

2018



SPECIM FX10

2017



Headwall micro-HyperSpec

2012

Hyperspectral line-scanner (Zarco-Tejada et al., 2012) (Lucieer et al., 2014)

**OceanOptics STS** Hyperspectral points-pectrometer (Burkart et al., 2014, 2015)

Aasen, H., Honkavaara, E., Lucieer, A., Zarco-Tejada, P., 2018. Quantitative Remote Sensing at Ultra-High Resolution with UAV Spectroscopy: A Review of Sensor Technology, Measurement Procedures, and Data Correction Workflows. Remote Sensing

2019



## (spectral) 2D imagers



#### SfM + GCPs (and/or imu + gnss)



Drawings kindly provided by Stefan Livens (VITO)

Aasen, H., Honkavaara, E., Lucieer, A., Zarco-Tejada, P., 2018. Quantitative Remote Sensing at Ultra-High Resolution with UAV Spectroscopy: A Review of Sensor Technology, Measurement Procedures, and Data Correction Workflows. Remote Sensing





#### **Structure from Motion**



Aasen, H., Burkart, A., Bolten, A., Bareth, G., 2015. Generating 3D hyperspectral information with lightweight UAV snapshot cameras for vegetation monitoring: From camera calibration to quality assurance. ISPRS Journal of Photogrammetry and Remote Sensing





## Spectral (/thermal) digital surface model



A spectral digital surface model is a representation of the surface in 3D space linked with spectral information emitted and reflected by the objects covered by the surface

Aasen, H., Burkart, A., Bolten, A., Bareth, G., 2015. Generating 3D hyperspectral information with lightweight UAV snapshot cameras for vegetation monitoring: From camera calibration to quality assurance. ISPRS Journal of Photogrammetry and Remote Sensing



## **Track plant growth with 3D information**



- H. Aasen, A. Burkart, A. Bolten, and G. Bareth, "Generating 3D hyperspectral information with lightweight UAV snapshot cameras for vegetation monitoring: From camera calibration to quality assurance," *ISPRS Journal of Photogrammetry and Remote Sensing*, vol. 108, pp. 245–259, Oct. 2015.
- J. Bendig et al., "Combining UAV-based plant height from crop surface models, visible, and near infrared vegetation indices for biomass monitoring in barley," International Journal of Applied Earth Observation and Geoinformation, vol. 39, pp. 79–87, Jul. 2015.
- N. Tilly, H. Aasen, and G. Bareth, "Fusion of Plant Height and Vegetation Indices for the Estimation of Barley Biomass," *Remote Sensing*, vol. 7, no. 9, pp. 11449–11480, Sep. 2015.
- H. Aasen and A. Bolten, "Multi-temporal high-resolution imaging spectroscopy with hyperspectral 2D imagers From theory to application," *Remote Sensing of Environment*, vol. 205, pp. 374–389, Feb. 2018.
- H. Aasen and G. Bareth, "Ground and UAV sensing approaches for spectral and 3D crop trait estimation," in *Hyperspectral Remote Sensing of Vegetation Volume II: Advanced Approaches and Applications in Crops and Plants*, Second Edition., P. Thenkabail, J. G. Lyon, and A. Huete, Eds. Taylor and Francis Inc., "accepted."
- L. Kronenberg, K. Yu, A. Walter, and A. Hund, "Monitoring the dynamics of wheat stem elongation: genotypes differ at critical stages," *Euphytica*, vol. 213, no. 7, Jul. 2017.





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## **Data processing**





## **Imaging spectroscopy with 2D imagers**





## Imaging spectroscopy with 2D imagers



## Single (most nadir) image



## Mosaic, blending: disabled



### Mosaic, blending: average





#### A: single image













## Imaging spectroscopy with 2D imagers





## 'pixel' in digital representation



Aasen, H., Honkavaara, E., Lucieer, A., Zarco-Tejada, P., 2018. Quantitative Remote Sensing at Ultra-High Resolution with UAV Spectroscopy: A Review of Sensor Technology, Measurement Procedures, and Data Correction Workflows. Remote Sensing



## Outline







## **PhenoFly mission statement**

- The PhenoFly team develops sensing systems and analysis procedures that deliver quantitative data to capture reliable information about vegetation
- Our vision is to bring (high-throughput) phenotyping approaches from large facilities to the field
- We aim to understand the interaction of plants with their environment to facilitate a more sustainable use of resources.



## Phenofly<sup>\*</sup> Low-altitude / high-resolution remote sensing at the Crop Science Group



platform FIP: a cable-suspended multi-sensor system. Functional Plant Biology

@PhenoFly | Helge.Aasen@usys.ethz.ch

27



## Plant research station Eschikon, ETH Zurich



FIP field 360°



## **Example: canopy temperature**



## PhenoFly multi-sensor payload





## **Canopy temperature workflow**



CROP SCIENCE Phenofly







## Influences on canopy temperature

- Genotype
- Soil
- Management
  - ...
- Measurement conditions (illumination fluctuations)
- Measurement device

#### Flight pattern



## **Canopy temperature workflow**



[1] M. X. Rodríguez-Álvarez, M. P. Boer, F. A. van Eeuwijk, and P. H. C. Eilers, "Correcting for spatial heterogeneity in plant breeding experiments with P-splines," *Spatial Statistics*, vol. 23, pp. 52–71, Mar. 2018.











dT 18-06-20 13:59:41







dT 18-06-20 10:11:10

[AT\* 100]









![](_page_39_Picture_0.jpeg)

![](_page_39_Picture_1.jpeg)

Correlations between the SpATS corrected plant traits and the canopy cover temperature for the solar noon measurements

![](_page_39_Figure_3.jpeg)

![](_page_40_Picture_0.jpeg)

![](_page_40_Picture_1.jpeg)

## Example: Extracting leaf area index using viewing geometry effects

![](_page_41_Picture_1.jpeg)

![](_page_41_Picture_2.jpeg)

![](_page_42_Picture_0.jpeg)

![](_page_42_Picture_1.jpeg)

#### FIP field –plant research station Eschikon

- RGB orthophoto and DSM (> 0.003 m)
- Mapped 1-3 times a week

![](_page_42_Picture_5.jpeg)

#### CROP SCIENCE Phenofly.net

## EnHzürich Extracting leaf area index using viewing geometry effects

![](_page_43_Figure_2.jpeg)

L. Roth, H. Aasen, A. Walter, and F. Liebisch, "Extracting leaf area index using viewing geometry effects—A new perspective on high-resolution unmanned aerial system photography," *ISPRS Journal of Photogrammetry and Remote Sensing*, 2018.

## ETH zürich Extracting leaf area index using viewing geometry effects

![](_page_44_Figure_1.jpeg)

L. Roth, H. Aasen, A. Walter, and F. Liebisch, "Extracting leaf area index using viewing geometry effects—A new perspective on high-resolution unmanned aerial system photography," *ISPRS Journal of Photogrammetry and Remote Sensing*, 2018.

![](_page_44_Picture_4.jpeg)

# Entracting leaf area index using viewing geometry effects

![](_page_45_Picture_1.jpeg)

GSD 0.007 m

Phenofly<sup>\*</sup>

![](_page_45_Figure_3.jpeg)

L. Roth, H. Aasen, A. Walter, and F. Liebisch, "Extracting leaf area index using viewing geometry effects—A new perspective on high-resolution unmanned aerial system photography," *ISPRS Journal of Photogrammetry and Remote Sensing*, 2018.

![](_page_46_Picture_1.jpeg)

## Conclusion

- High-resolution (drone based) remote sensing approaches offer great potential for field-phenotyping
- Mission planning is absolutely crucial
  - Know what you want to measure and plan accordingly
  - Think of efficiency and reliability: set up your site accordingly
- Know what you do and let others know

![](_page_47_Picture_1.jpeg)

## Conclusion

- High-resolution (drone based) remote sensing approaches offer great potential for field-phenotyping
- Mission planning is absolutely crucial
  - Know what you want to measure and plan accordingly
  - Think of efficiency and reliability: set up your site accordingly
- Know what you do and let others know
  - Keep track of metadata
  - Report important information in publications
- For the future:
  - I believe that high-resolution remote sensing will revolutionize field-phenotyping
  - Common protocols are needed (comparability / effectiveness) 05.07.201

![](_page_48_Picture_1.jpeg)

## Thank you for your attention

![](_page_48_Picture_3.jpeg)

#### Schweizerische Eidgenossenschaft Confédération suisse Confederazione Svizzera Confederaziun svizra

Schweizerische Eidgenossenschaft Confédération suisse Confederazione Svizzera Confederaziun svizra Bundesamt für Landwirtschaft BLW

Innosuisse - Schweizerische Agentur für Innovationsförderung

**ETH** zürich

![](_page_48_Picture_9.jpeg)

![](_page_48_Picture_10.jpeg)

![](_page_48_Picture_11.jpeg)

#### CROP SCIENCE Phenofly

## EnHzürich Extracting leaf area index using viewing geometry effects

![](_page_49_Figure_2.jpeg)

L. Roth, H. Aasen, A. Walter, and F. Liebisch, "Extracting leaf area index using viewing geometry effects—A new perspective on high-resolution unmanned aerial system photography," *ISPRS Journal of Photogrammetry and Remote Sensing*, 2018.

![](_page_50_Picture_1.jpeg)

# Extracting leaf area index using viewing geometry effects

![](_page_50_Figure_3.jpeg)

L. Roth, H. Aasen, A. Walter, and F. Liebisch, "Extracting leaf area index using viewing geometry effects—A new perspective on high-resolution unmanned aerial system photography," *ISPRS Journal of Photogrammetry and Remote Sensing*, 2018.