

# UAS LIDAR IN AGRICULTURE

**JORDAN STEVEN BATES**

DOCTORAL RESEARCHER

INSTITUTE OF BIO- AND GEOSCIENCES: AGROSPHERE (IBG-3),

FORSCHUNGSZENTRUM JÜLICH

J.BATES@FZ-JUELICH.DE

**SUPERVISORS:**

**PROF. FRANÇOIS JONARD**

EARTH AND LIFE INSTITUTE, UNIVERSITÉ CATHOLIQUE DE LOUVAIN

F.JONARD@FZ-JUELICH.DE

**DR. CARSTEN MONTZKA**

INSTITUTE OF BIO- AND GEOSCIENCES: AGROSPHERE (IBG-3),

FORSCHUNGSZENTRUM JÜLICH

C.MONTZKA@FZ-JUELICH.DE



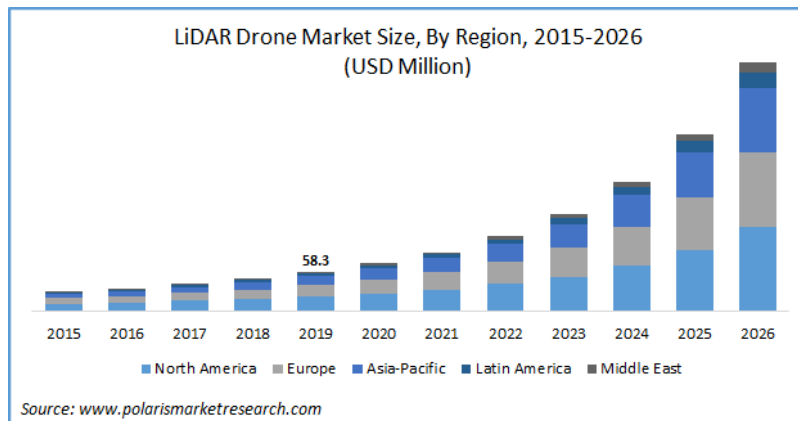
GEOSPATIAL<sup>™</sup>  
WORLD  
FORUM 

UNIVERSITÄT  **BONN**  PHENOROB

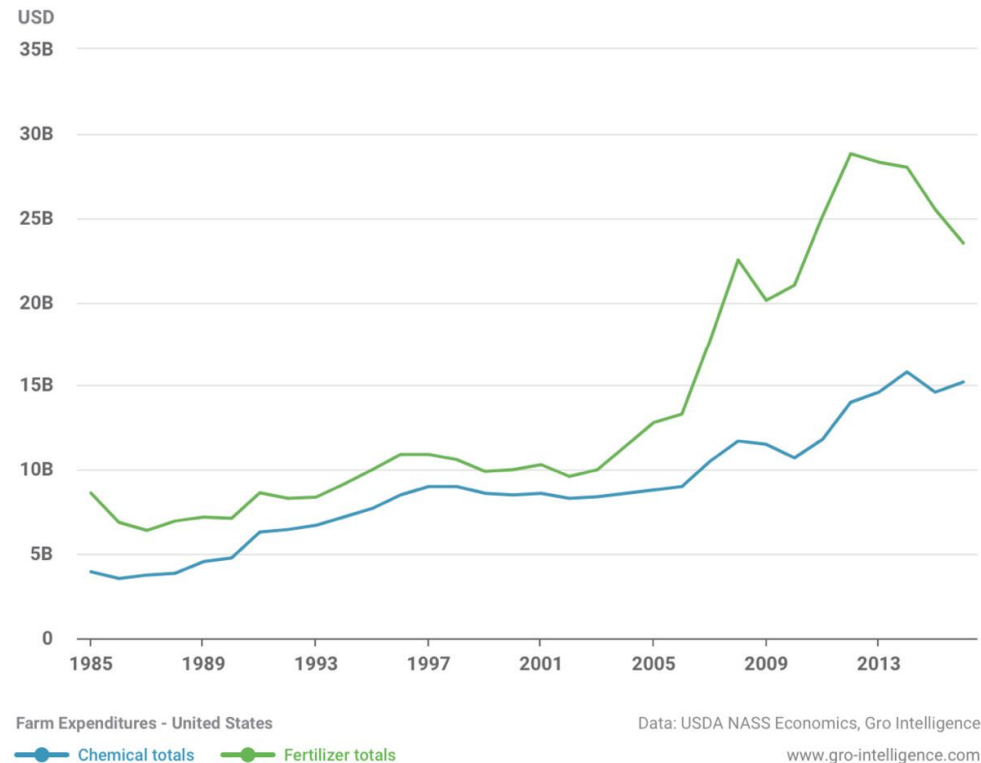
 **JÜLICH**  
Forschungszentrum

# IMPORTANCE OF DRONES IN AGRICULTURE

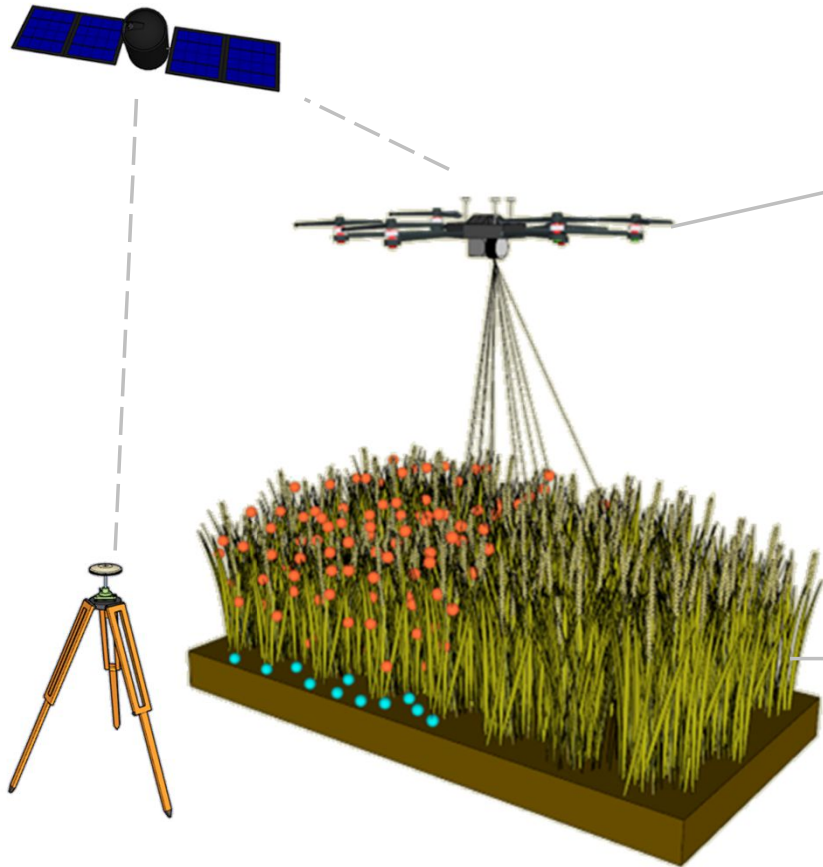
- Population and food demand increasing
- Pesticide and fertilizer use increasing
- Need for technologies to increase efficiency and precision
- UAS's provide on demand high resolution data
- LiDAR UAS adaptation increasing



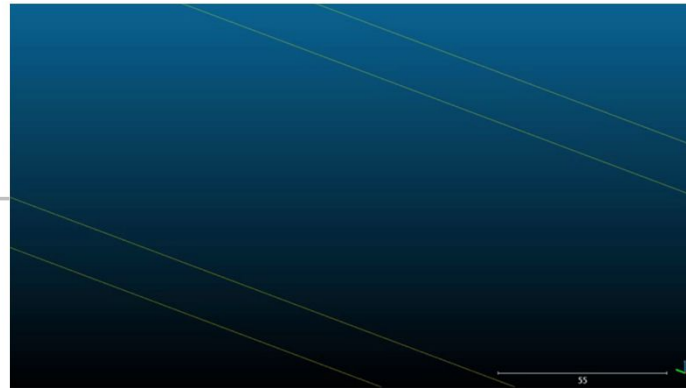
US Farmers Spend a Significant Amount on Inputs



# EQUIPMENT OVERVIEW



LiDAR	YellowScan Surveyor
Flight Altitude	50m
Flight Speed	8 m/s
Point density	85 pts/sqm
Accuracy	5 cm
Scan Angle Used	24-30 degrees



Mitglied der Helmholtz-Gemeinschaft

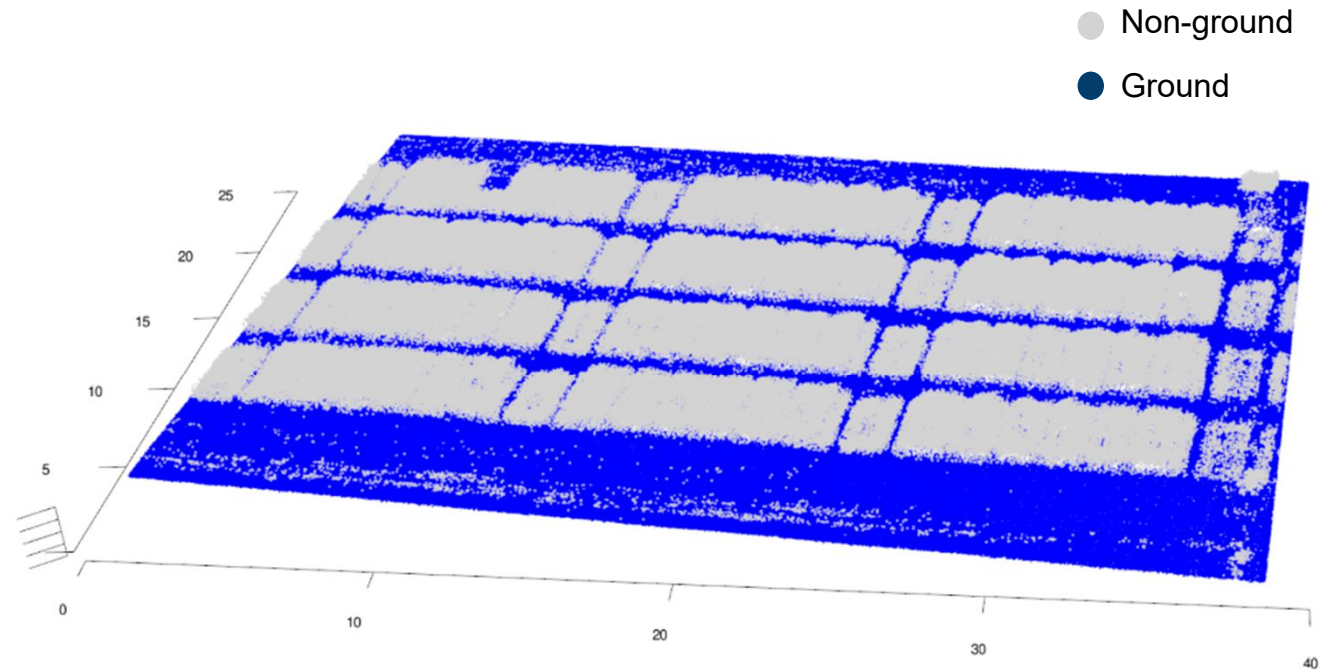
# GROUND SEGMENTATION

- Segmenting ground from vegetation is important for deriving several vegetation metrics
- Here Cloth Simulation Filter (CSF) is used. (Zhang et al. 2021).



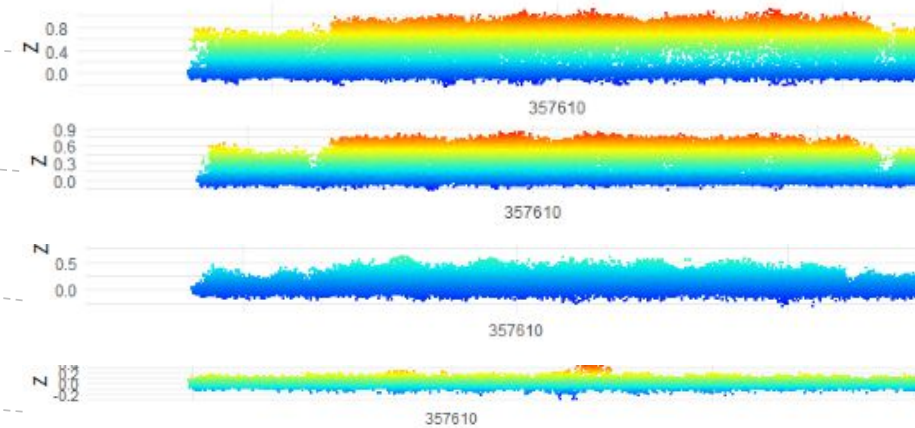
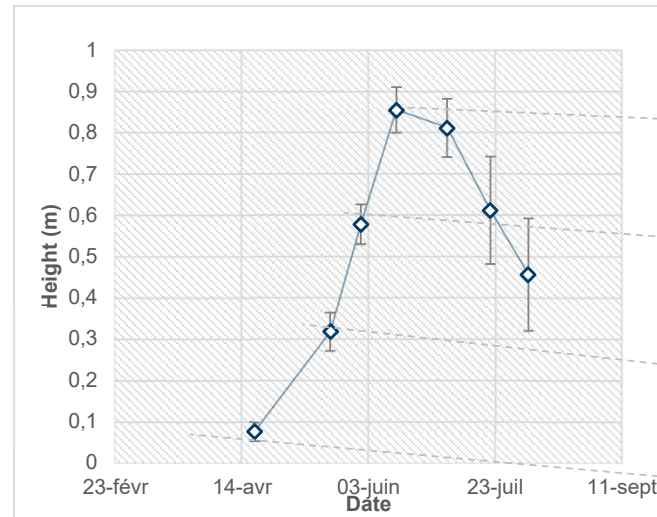
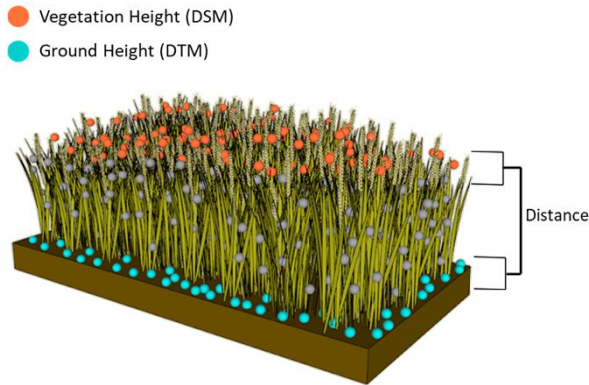
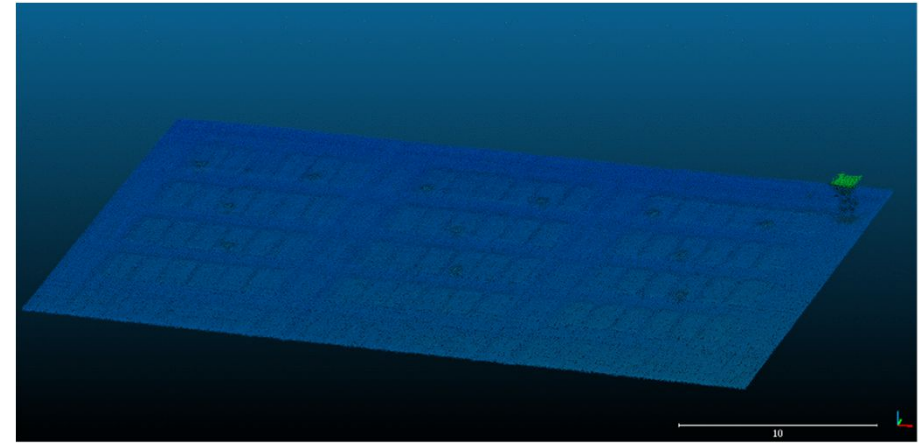
Campus Klien Altendorf, Germany  
PhenoRob Central Experiment

Mitglied der Helmholtz-Gemeinschaft



# CROP HEIGHT

- Most common crop metric with LiDAR
- Crop Height Models (CHM) with Difference of DEMs (DoD) most common
- Other methods also used such as Cloud to Cloud (C2C) with 3D point clouds

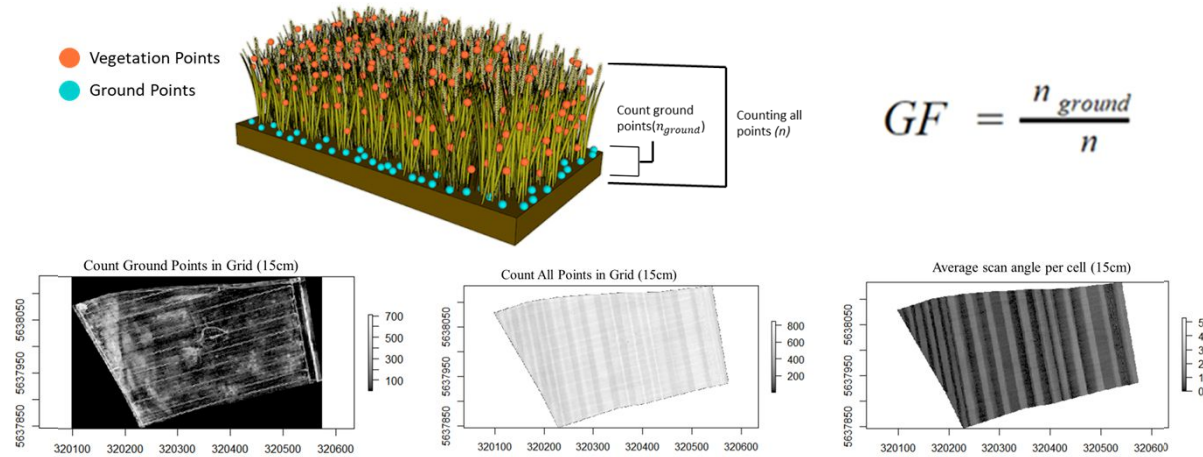


# GAP FRACTION – CANOPY DENSITY

- Used LiDAR gap fraction to estimated canopy density
- Similar method to hemispherical cameras used in forestry
- Modified Beer-Lambert equation used to relate the rate of laser penetration through the canopy to **Plant Area Index (PAI)**.

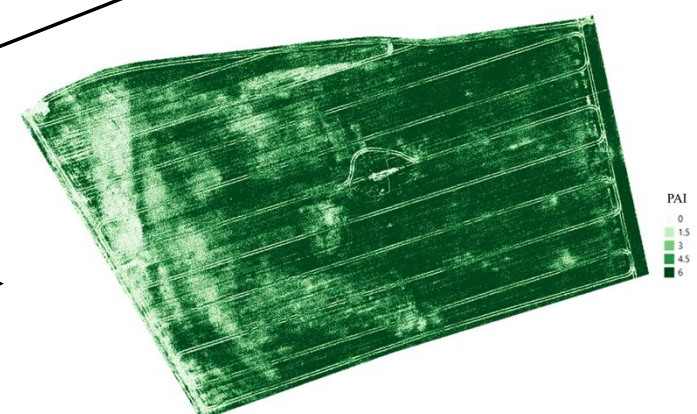
(Jonckheere et al. 2005)

(Richardson et al. 2009)



$$PAI_{LiDAR} = - \frac{\overline{\cos(\theta)} \times \ln(GF)}{k}$$

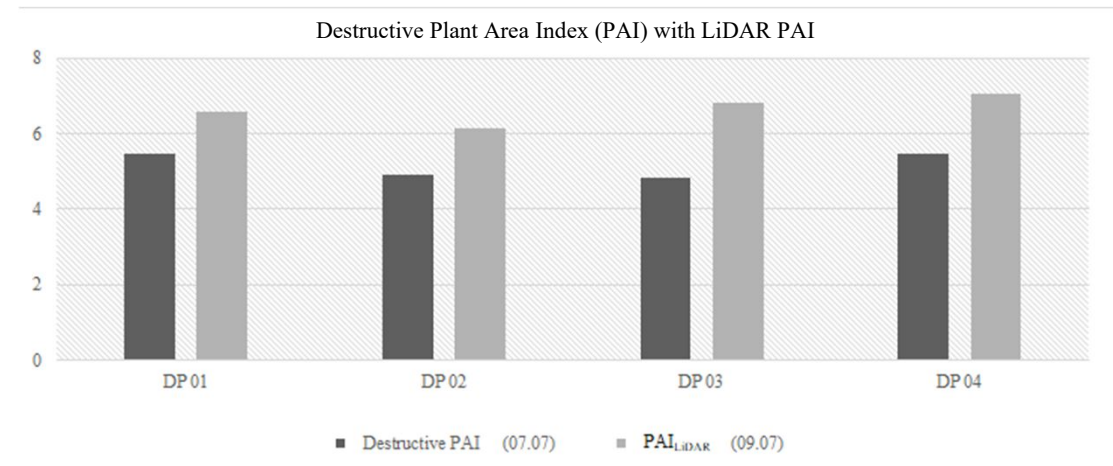
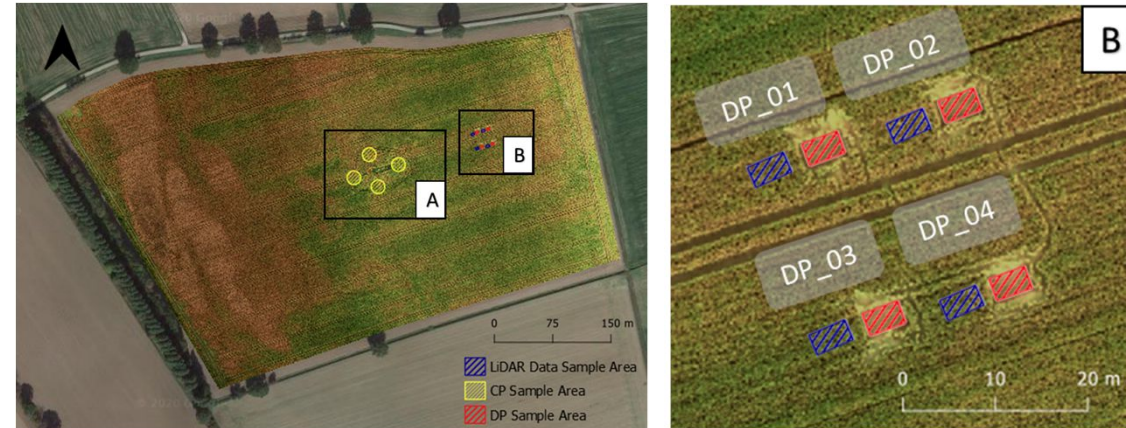
Modified Beer-Lambert Equation



# LEAF AREA INDEX (LAI)

- Compared LiDAR PAI to ground destructive measurements (B).
- **28% error** from LiDAR in overestimating later in the growing season. Better performance earlier in the growing season (**3-12%** before senescence with comparison to ceptometer).
- Provides evidence that LiDAR can be used to calculate PAI or LAI like metrics for winter wheat
- Can be **improved** upon with flight parameters and processing variables

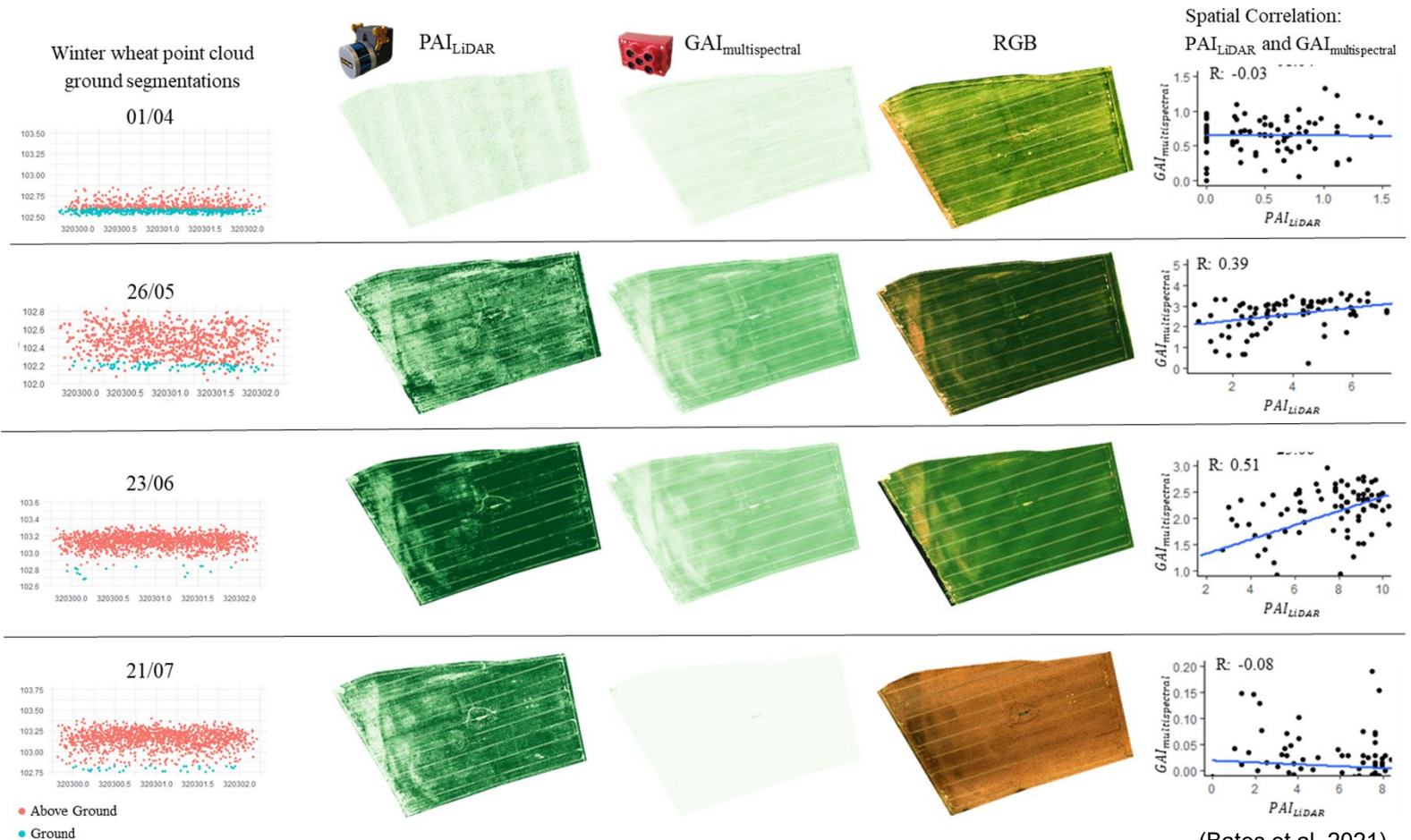
Selhausen, Germany FZJ Test Site



(Bates et al. 2021).

# LIDAR VS. PASSIVE OPTICAL SENSORS

- The LiDAR variability was compared to multispectral **Green Area Index (GAI)**
- GAI is based on **leaf pigment** while LiDAR PAI method is sensing **plant structure**
- The LiDAR is still able to sense variation in canopy structure after full senescence



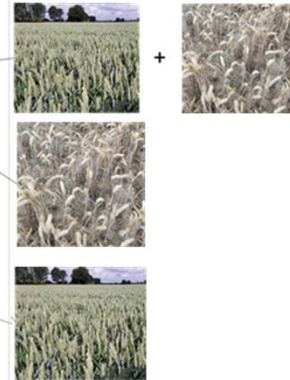
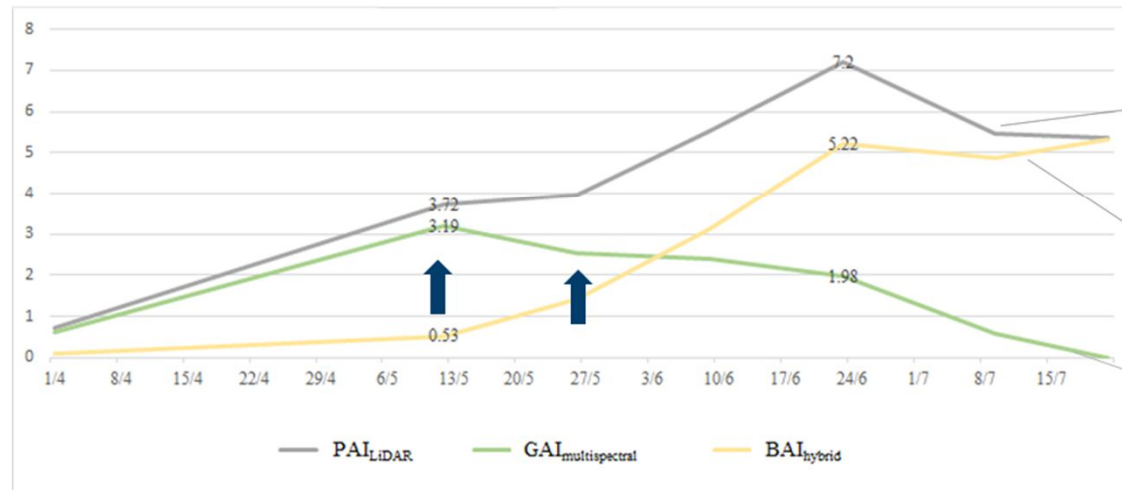
(Bates et al. 2021).



# MULTI-SENSOR USE

- Difference between PAI LiDAR and GAI multispectral methods allowed for hybrid estimation of Brown Area Index (BAI)
- Concept of multi-sensor use in combination with the advantages of UAS high spatial and temporal resolutions that provide a more complete picture for farming management
- For more info, please check out our publication.

Time-series of field average PAI, GAI, and BAI



Open Access Article

## Estimating Canopy Density Parameters Time-Series for Winter Wheat Using UAS Mounted LiDAR

by [Jordan Steven Bates](#)<sup>1,\*</sup>, [Carsten Montzka](#)<sup>1</sup>, [Marius Schmidt](#)<sup>1</sup> and [François Jonard](#)<sup>1,2</sup>

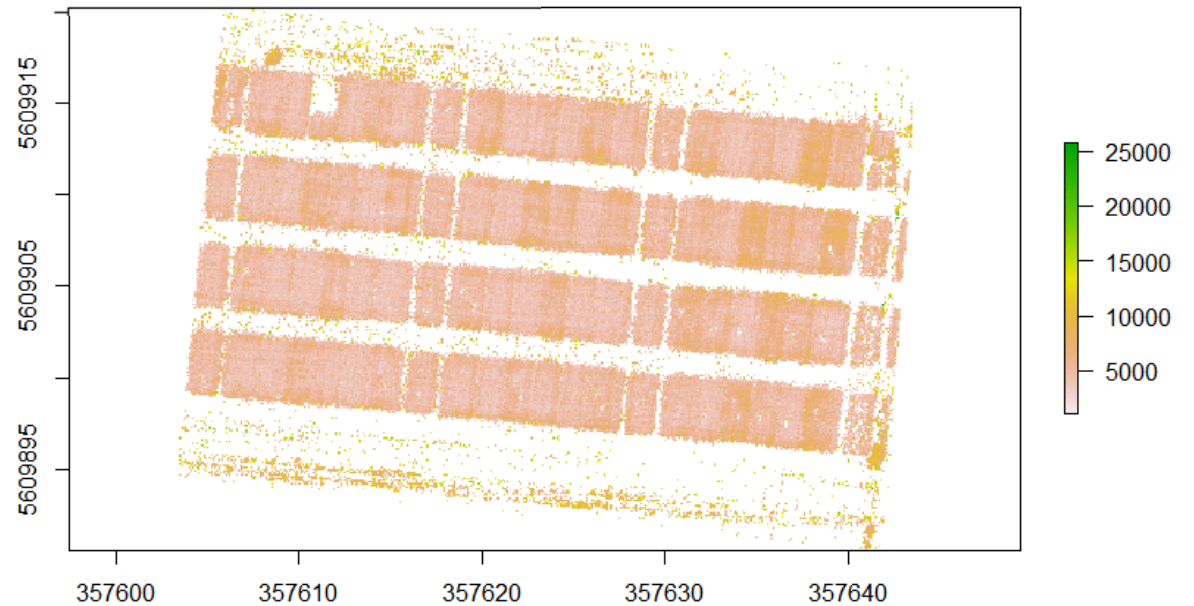
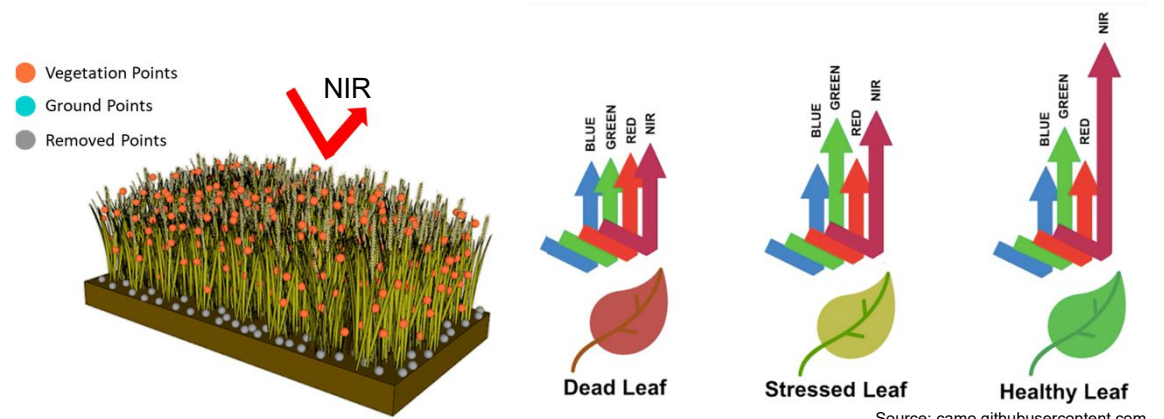
<sup>1</sup> Institute of Bio- and Geosciences: Agrosphere (IBG-3), Forschungszentrum Jülich GmbH, 52428 Jülich, Germany

<sup>2</sup> Earth and Life Institute, Université catholique de Louvain, 1348 Louvain-la-Neuve, Belgium

\* Author to whom correspondence should be addressed.

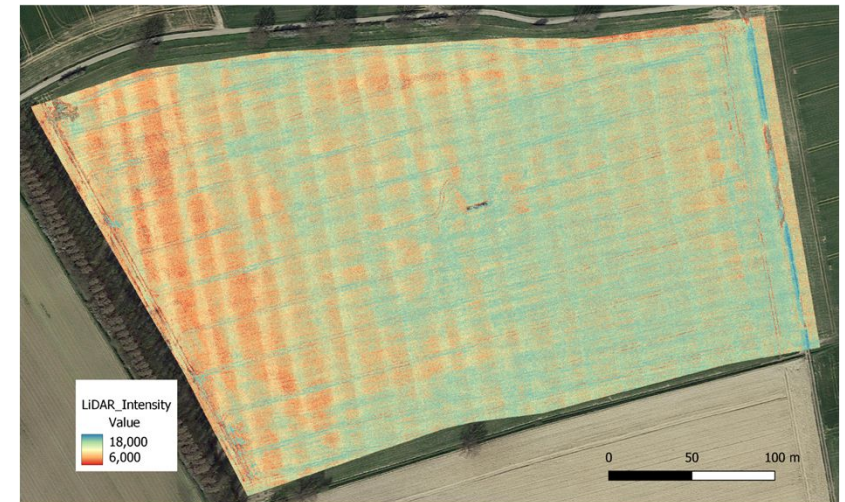
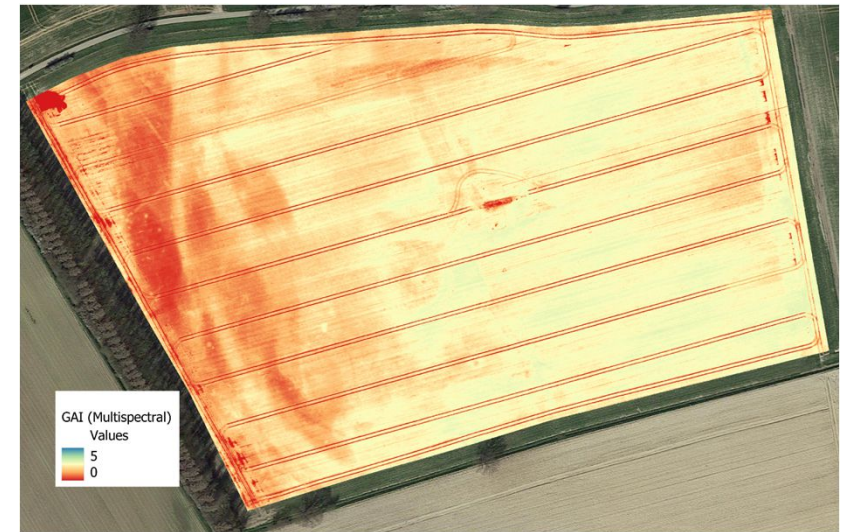
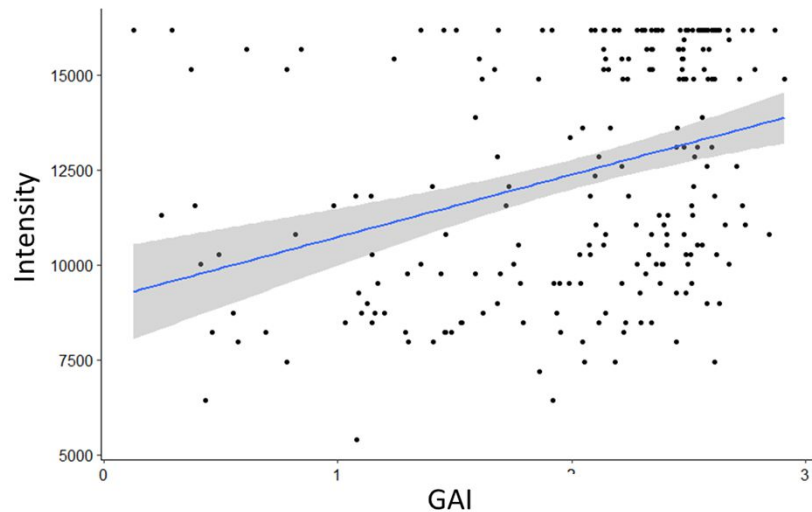
# LIDAR INTENSITY

- The recorded strength of the returning signal
- YellowScan Surveyor's electromagnetic spectrum bandwidth is positioned around **903nm – NIR**
- Ground intensity values between dead and healthy winter wheat
- Once ground is removed, information on vegetation chlorophyll status appears



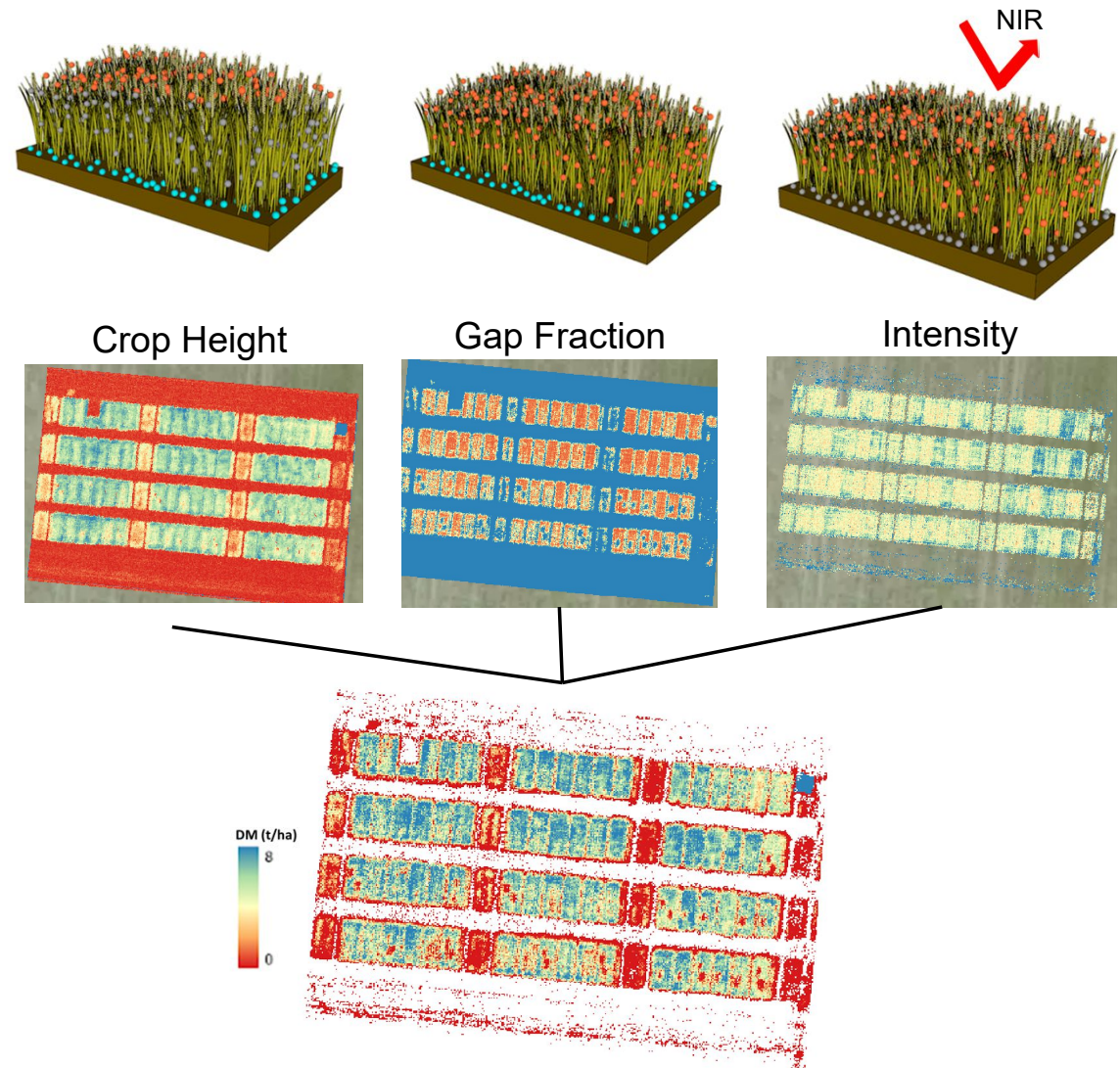
# INTENSITY POTENTIAL

- Coloration between LiDAR intensity and multispectral GAI showing potential to further implement LiDAR for crop status
- Plans to use the intensity in place of multispectral methods to identify and quantify senescence locations



# BIOMASS

- Plans to use **crop height (vertical information)**, **gap fraction (lateral density)**, **intensity (chlorophyll content)**, and several other LiDAR metrics as predictors in for machine learning models
- Current method delivered a standard error of .5 (t/ha)



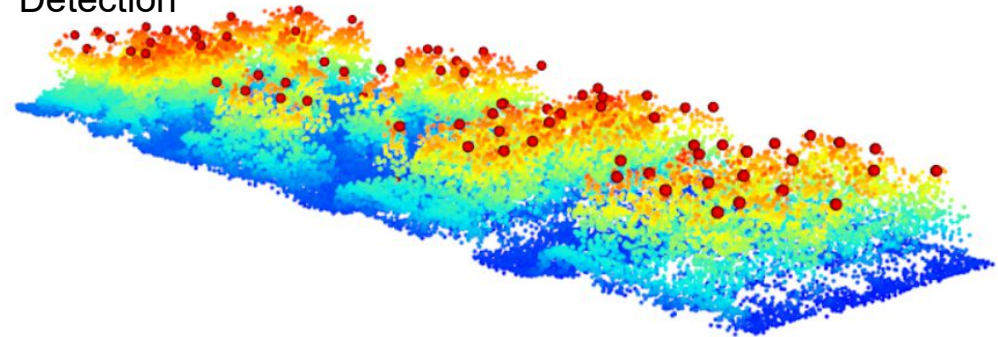
# CROP DETECTION

- The height information can be used for crop detection and segmentation for certain crops.
- Can be intercepted with multispectral and thermal IR data.
- Can show which and how many plants are affected by water stress and etc.

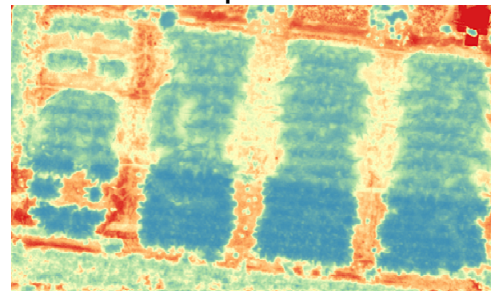
Example: Maize @ CKA



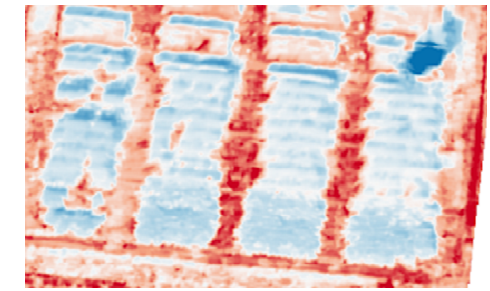
Individual Maize Detection



NDVI: Multispectral



LST: Thermal IR



# OUTLOOK

- LiDAR has the potential to provide crop height, canopy density (PAI), chlorophyll content (GAI), Biomass, and crop detection.
- FZJ is testing fully automatic UAV collections for farming applications.
- On-site housing and powering hubs further the practicality this technology with high temporal collections while lessening logistics and specialized training.



# REFERENCES

- Bates, J. S., Montzka, C., Schmidt, M., & Jonard, F. (2021). Estimating Canopy Density Parameters Time-Series for Winter Wheat Using UAS Mounted LiDAR. *Remote Sensing*, 13(4), 710.
- Jonckheere, I., Nackaerts, K., Muys, B., & Coppin, P. (2005). Assessment of automatic gap fraction estimation of forests from digital hemispherical photography. *Agricultural and Forest Meteorology*, 132(1), 96–114.
- Mattia, F., Dente, L., Satalino, G., & Le Toan, T. (2005). *Sensitivity of ASAR AP Data to Wheat Crop Parameters*.
- Richardson, J. J., Moskal, L. M., & Kim, S.-H. (2009). Modeling approaches to estimate effective leaf area index from aerial discrete-return LIDAR. *Agricultural and Forest Meteorology*, 149(6), 1152–1160.
- Zhang W, Qi J, Wan P, Wang H, Xie D, Wang X, Yan G. An Easy-to-Use Airborne LiDAR Data Filtering Method Based on Cloth Simulation. *Remote Sensing*. 2016; 8(6):501.

# THANK YOU FOR YOUR ATTENTION!

CONTACT ME AT  
EMAIL: [J.BATES@FZ-JUELICH.DE](mailto:J.BATES@FZ-JUELICH.DE)