

# Assessing hydraulic and morphological conditions using UAS

Christian Haas
Philipp Thumser
Dr. Jeffrey Tuhtan





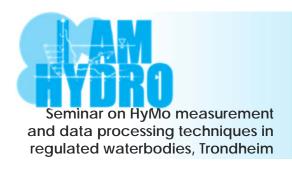






## Outline





# Why UAS?

#### **Current Monitoring strategies:**

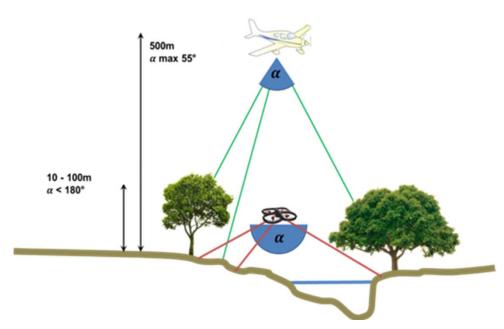
- No monitoring
- Modelling
- Single Point/ Data



Requirements of Monitoring:

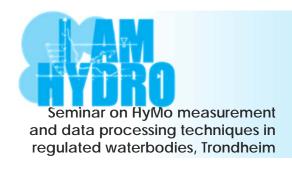
Monitoring strategies/ tools need to picture natural processes in time and scale!

New/ additional Monitoring tools and strategies required:



Low altitudes/ fast repeatability

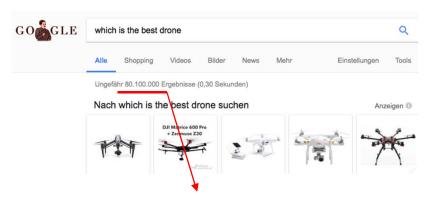
- → High resolution (spatial and temporal)
- → Better learning/ understanding?



## Which UAS?

#### ...lets ask Google:

http://lmgtfy.com/?q=which+is+the+best+drone

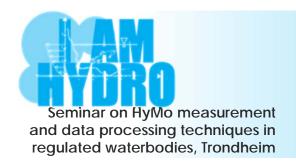


80.100.000 results

What does experience show?

- There is no best drone in general!
  - Hardware must be reliable
  - Data must be sufficient
- The investment is not correlated to the outcome but to knowledge!

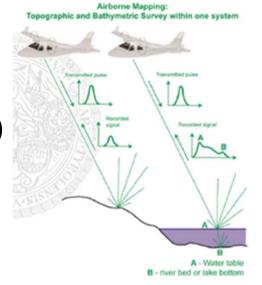
→ inexpensive equipment + knowledge = high quality output



## Sensors

- Cameras
  - Post Processing
    - RGB
    - NIR
    - Thermal
  - Real Time Data
    - SLAM Simultaneous Localization and Mapping
    - RAPTOR Real Time Particle Tracking
- LIDAR (Light Detection and Ranging)
  - IR LIDAR
  - Green LIDAR

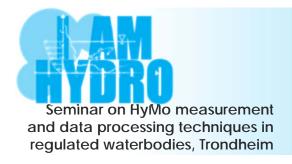






## 3D Reconstruction



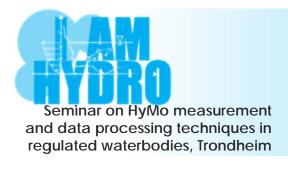


# Georeferenced Imagery



## Hydropeaking on a gravel bar at river Lech, Germany

- temporal and spatial measurements of up- and down ramping events
- Control and investigate altered ramping rates for mitigation

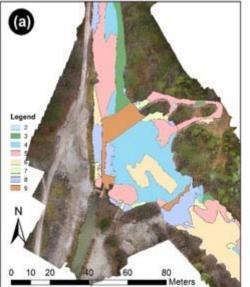


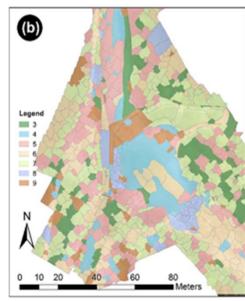
## Classification of UAV Imagery

#### Classification:

- Supervised & unsupervised classification
- Threshold classification
- Masking
- Segmentation based classification
- Validation and analysis

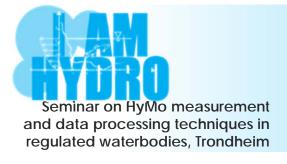
Software for this approach: ERDAS IMAGINE, Matlab, eCognition & Arc GIS





Comparative illustration of substrate classified map showing spatial distribution of the sediment types:

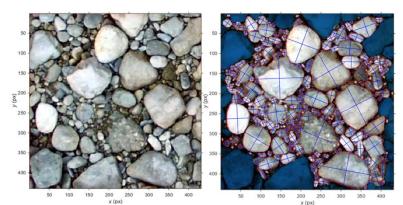
- a) Manually mapped substrate
- b) Automatically mapped substrate



# Sediment Mapping & Grain Size

#### **Photo Sieving**

BASEGRAIN (Detert & Weitbrecht, 2013)



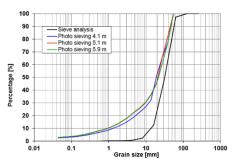
**BASEGRAIN:** 

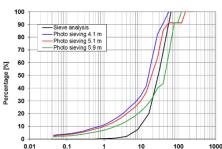
MATLAB-based automated object detection software tool (Detert & Weitbrecht, 2013)

#### Comparison with lab sieving curve of sample







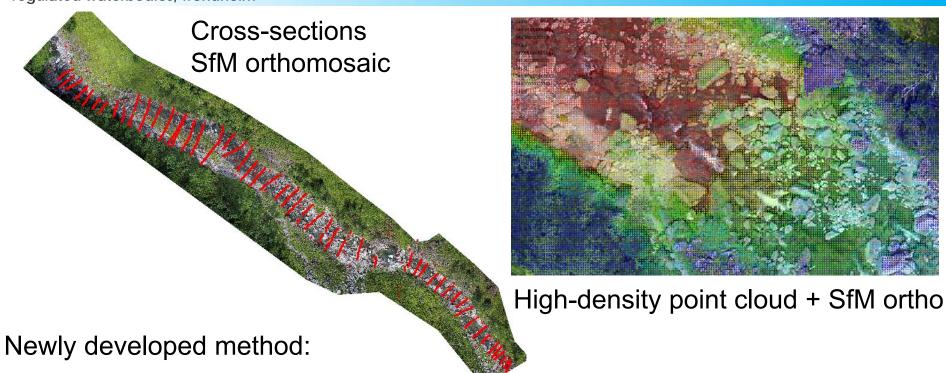


Good results in dry areas

More variation in submerged areas and areas with biofilm on sediment



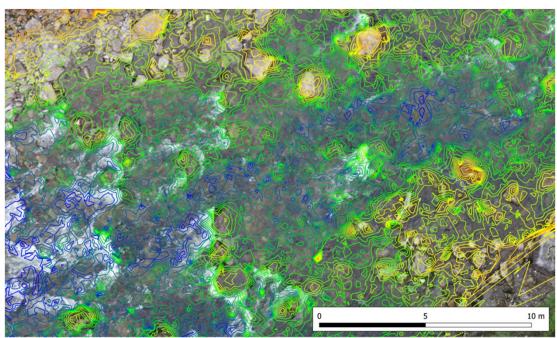
# Bathymetric Reconstruction Taigdalselva, Norway



- Uses standard cross-section measurements + Structure from Motion (SfM) point cloud
- 2) Multivariate model of submerged regions from SfM + cross-section data
- 3) High-resolution (10 cm/pt) bathymetric reconstruction (error < 10 cm)



# UAS-based bathymetric reconstruction



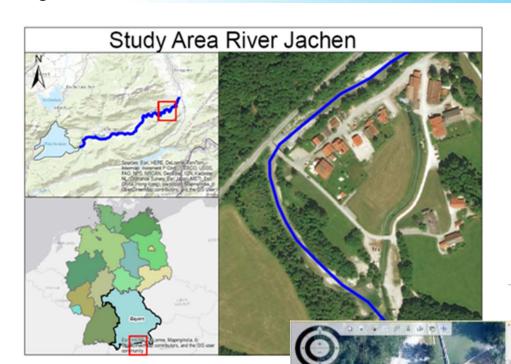
Overlay of contours and 2 cm/px orthomosaic

Can be used to generate highly accurate numerical models and DEM change maps

- Uses pre-existing data + UAS imagery products
- Low cost (~2k EUR / km)
- Same accuracy or better than LiDAR
- Applicable for reaches 50 m to 20 km
- Requires field measurements of submerged regions

# Seminar on HyMo measurement and data processing techniques in regulated waterbodies, Trondheim

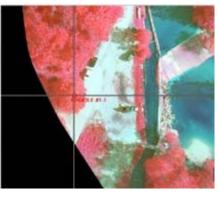
# NIR imagery



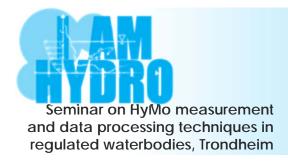
Low-cost multicamera system (Haas et al. 2016)



RGB orthomosaic



NIR orthomosaic

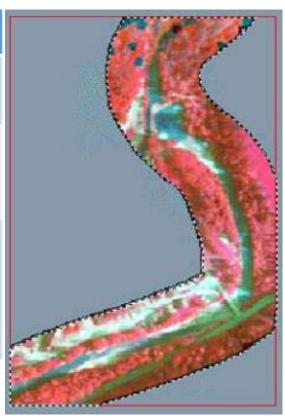


# NIR imagery

#### Vegetation indices from multispectral imagery\*

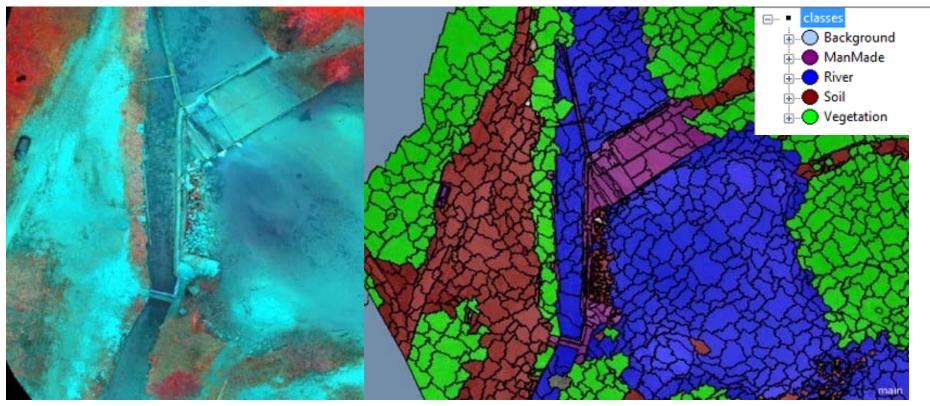
Index	Formula	Description
DVI	DVI = NIR - RED	<b>Difference Vegetation Index</b> , it is used to separate soil from vegetation. It is sensitive to illuminations conditions.
SR	$SR = \frac{NIR}{REd}$	Simple Ratio Index shows high values for vegetation and lower ones for soil.  Compensates differences in lighting conditions
NDVI	$NDVI = \frac{NIR - RED}{NIR + REd}$	Normalized Difference Vegetation Index used to estimate amount of vegetation, good to distinguish vegetation from soil.  More Robust applicable to both reflectance and Radiance. Reduces the effect of no uniform illumination.

(\*Jones and Vaughan, 2010)



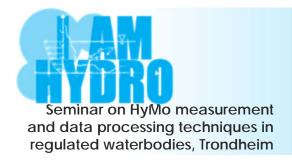


# Multi-band imagery



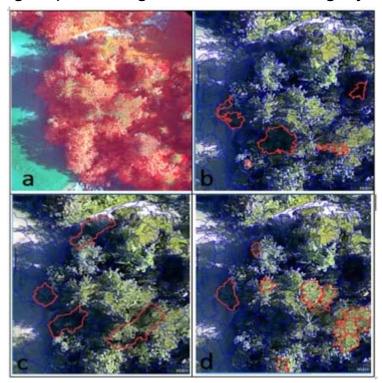
Multi-band image combines visible RGB and NIR

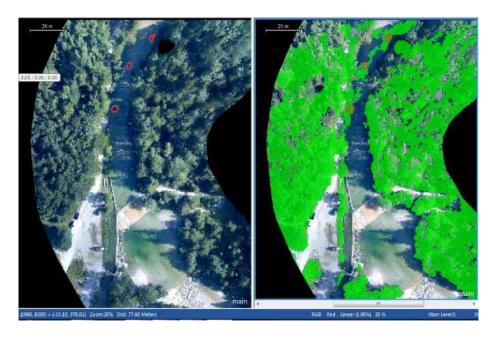
**Example usage:** automated classification of the landscape into user-defined groups



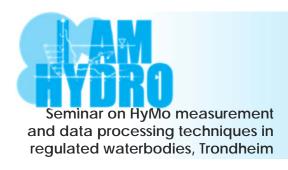
# Multi-band imagery

# Vegetated regions identified and grouped using NIR and RGB imagery





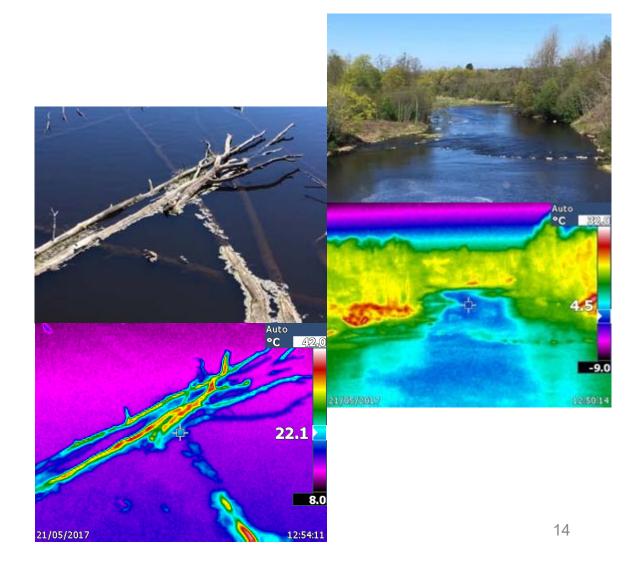
- Data used to compare changes in vegetation over different periods
- interpretation aided by overlay with aerial imagery



# Thermal imagery

#### Used for:

- Thermopeaking
- Discharge
- Upwelling and downwelling (heat stress)
- Persist "memory effects" of shading and illumination
- Effects of heat sinks (walls, boulders)

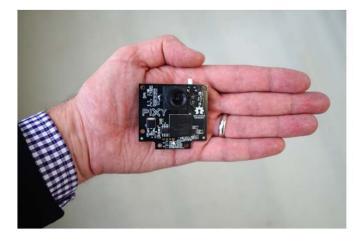


# RAPTOR – Real-Time Particle Tracking of Surface Flows Using Unmanned Aerial Vehicles

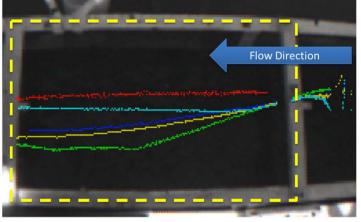


UAS tracks illuminated particles on surface

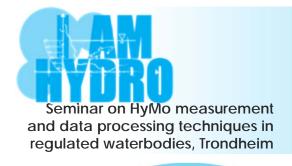
2D velocity vector, real-time



Light-weight embedded color vision controller



System can also be used in lab to track sediment, fish etc.



# **UAS Applications**

#### **Hydrology** and Hydraulics

- digital elevation models
- waterlines

#### Morphology and **Morphodyna**<sup>r</sup>

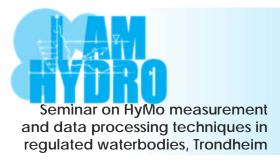
- grainsi
- - deposition

# less modelling,

- Habitat mapping
- plant and macrophyte

#### Intersection: Ecohydraulics

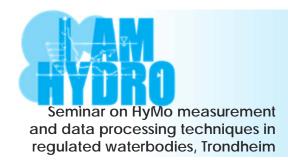
- → very usefull tool for many of flow and morphology related tasks and questions
- → monitoring of **dynamic processes** with high spatial and temporal resolution



#### Some recent research...

- Alfredsen, K., Haas, C., Tuhtan J.A., Zinke, P., 2018. Brief Communication: Mapping river ice using drones and structure from motion. In: The Cryosphere
- Haas, C., Thumser, P., Tuhtan, J., 2017. Unmanned Aerial Systems (UAS) New opportunities for measuring, mapping and modelling rivers and lakes. In: Joint Workshop IEA Hydropower TCP

   European Commission DG RTD, Brussels, Belgium.
- Arif, M. S. M., Gülch, E., Tuhtan, J.A., Thumser, P., Haas, C., 2016. An
   on image processing techniques for substrate classification based on
   dominant grain size on RGB images from UAV. In: International Journal of
   Sensing.
- Thumser, P., Haas, C., Tuhtan, J.A., Fuentes-Pérez, J.F., Toming, G., 2017.
   RAPTOR-UAV: Real-Time Particle Tracking in Rivers using an Unmanned Aerial Vehicle. In: Earth Surface Processes and Landforms.
- Haas, C., Thumser, P., Seitz, L., 2016. UAV based determination of grain size distribution at River Jachen, Germany. Proceedings of the 13th International Symposium on River Sedimentation, Stuttgart, Germany.



#### Outlook

- The internet of things (IOT) allows for rapid integration of custom sensing systems (e.g. cameras, depth, temperature)
- Long-term, persistent monitoring of aquatic variables including depth, temperature, salinity, suspended load is becoming very inexpensive (< 1500 EUR / site)</li>
- Aggregation and real-time assessment of large, multimodal unstructured data sets using statistical learning methods (e.g. deep learning, tensor flow, support vector machines)
- Increasing availability of geospatial measurements will in some cases begin replacing models (e.g. stream flow, satellite-based precipitation, ARGO network)

