

Towards Multispectral, Multi-Sensor Indoor Positioning and Target Identification

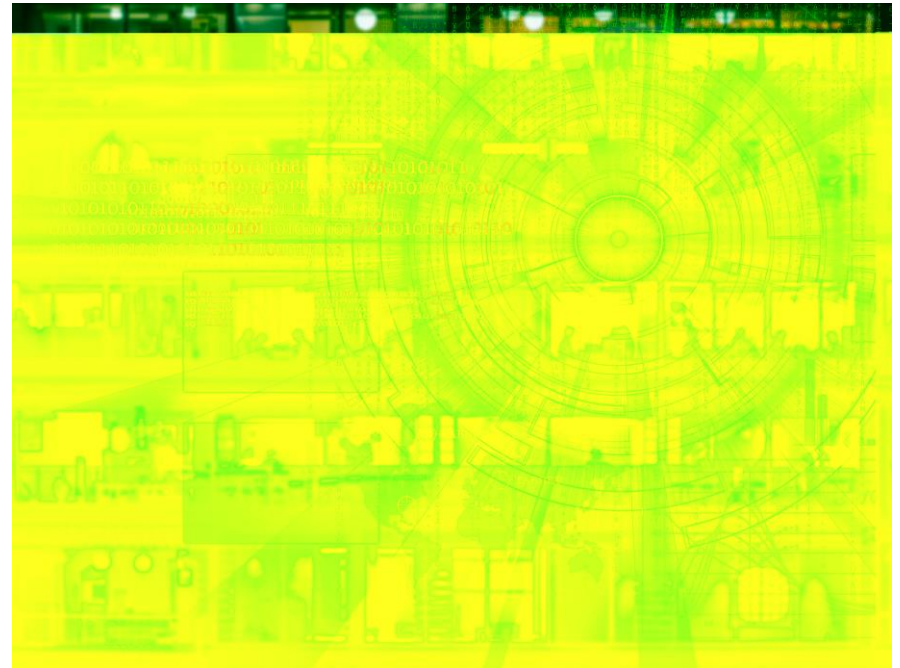
S. Kaasalainen, L. Ruotsalainen, M. Kirkko-Jaakkola, O. Nevalainen, and T. Hakala

Finnish Geospatial Research Institute (FGI)



Contents

- Motivation
- Multispectral and Hyperspectral lidar
- Target identification
- Infrastructure-free positioning
- Positioning results



Motivation

- Growing need for mobile mapping and surveillance in buildings
- Combining Light Detection and Ranging (lidar) and positioning sensors => simultaneous localization and mapping (SLAM)
- Multispectral lidar => target identification
- Multisensor fusion => infrastructure-free position solution
- Spatially resolved target identification in unknown indoor environment

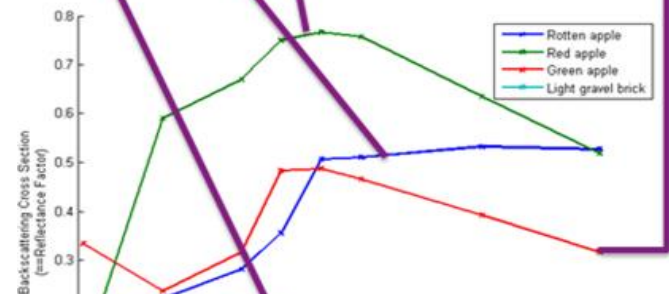
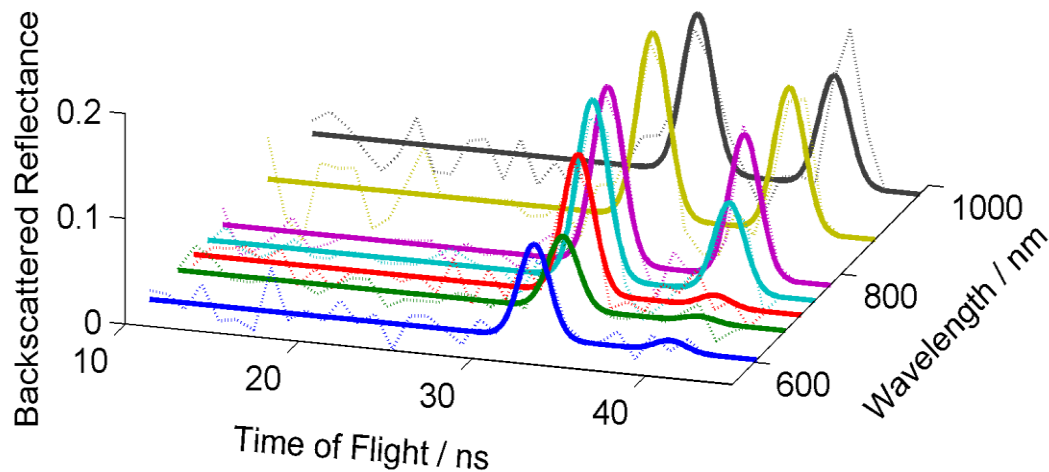
Multispectral lidar

- Laser scanning => multiwavelength lidar
- Output of a multispectral lidar is the pointcloud (x,y,z,I)
- I is the intensity containing multiple values of wavelength
- Used for target identification and e.g. vegetation studies

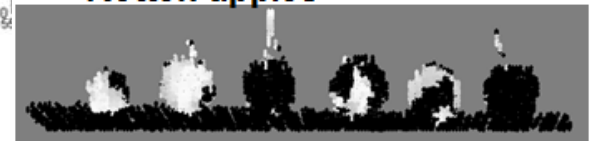


The FGI hyperspectral lidar (HSL) prototype

- Automatic 3D target identification
- One-shot topography & active hyperspectral imaging
- Spectrum directly available for each point in the laser scanning point cloud
- Algorithms for automatic classification of targets with spectral and spatial features



Rotten apples



Red apples

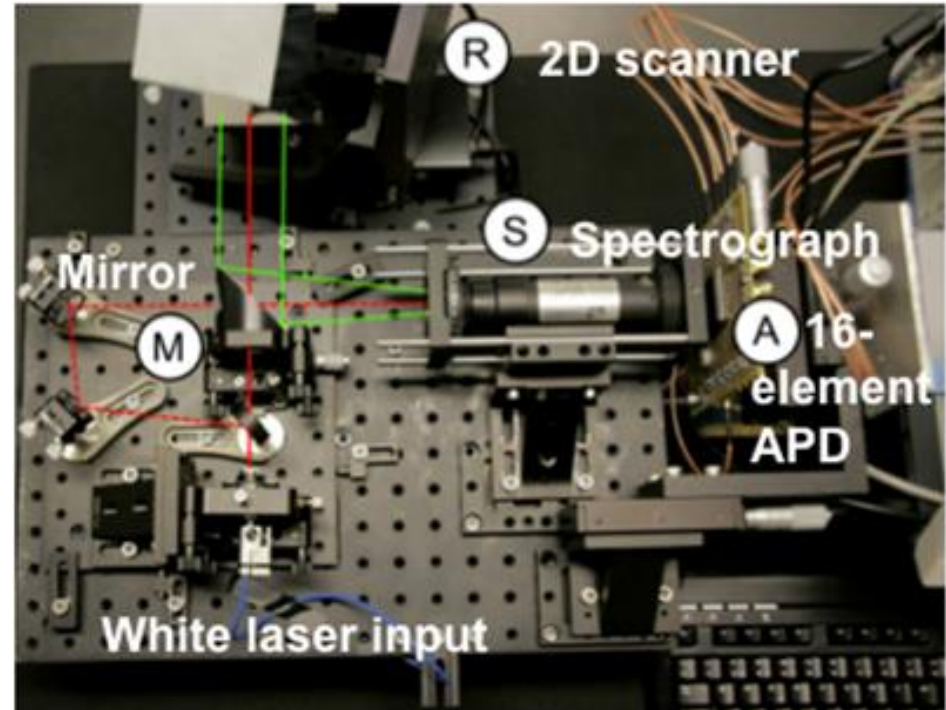


Green apples



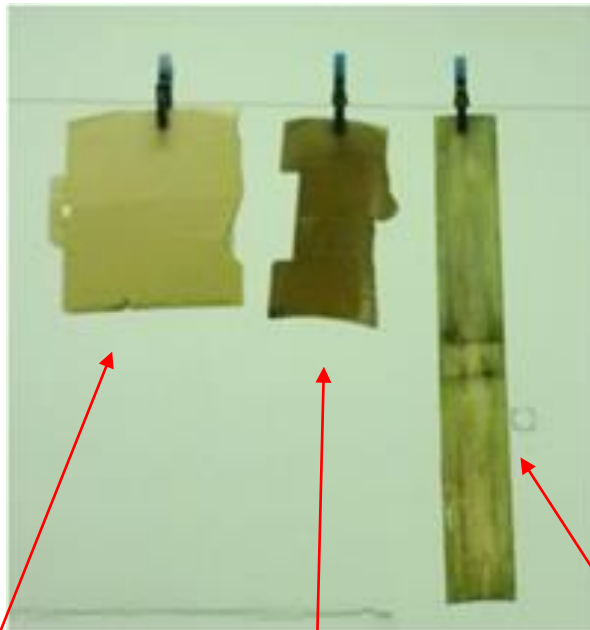
FGI HSL

- Supercontinuum laser light source
- 8 channel spectrum, 500-1000 nm for each point
- Parabolic mirror to gather returning laser pulses
- Detector
 - Spectrograph
 - 16-element avalanche photodiode
- 1 ns digitizer for data storage
- Wavelength channels selected by adjusting the spectrograph position



Target identification 1/3

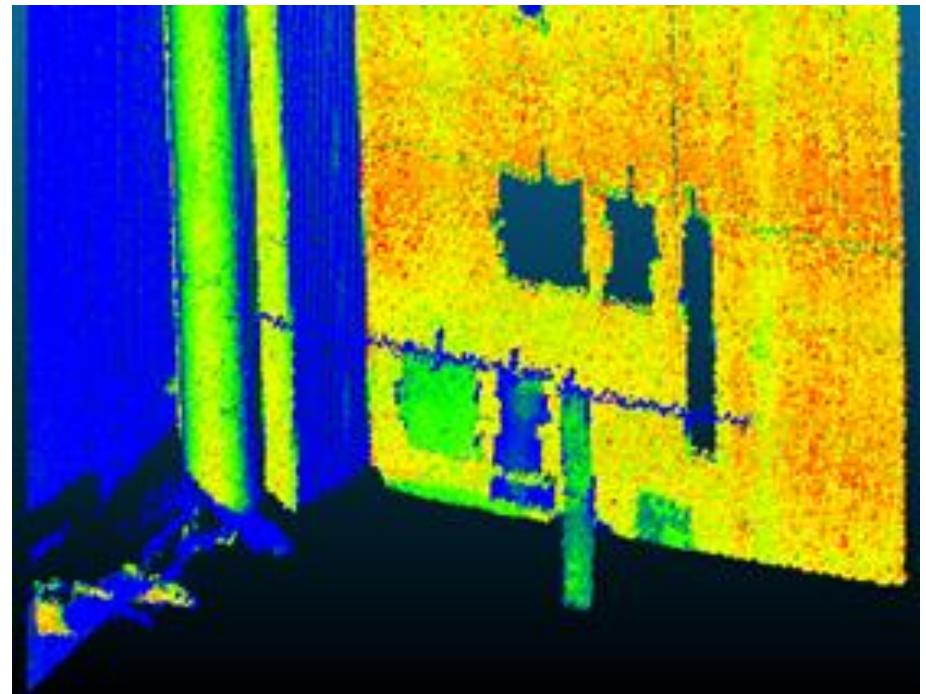
- Identifying targets with different moisture levels



Cardboard

**Cardboard
with water**

**Wood
with mold**

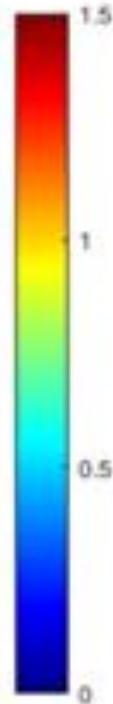
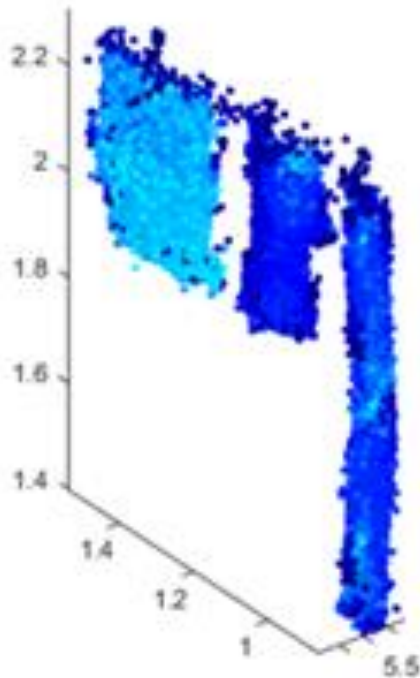


**Pointcloud of the room
showing the targets**

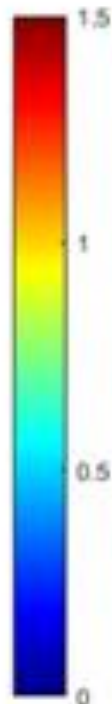
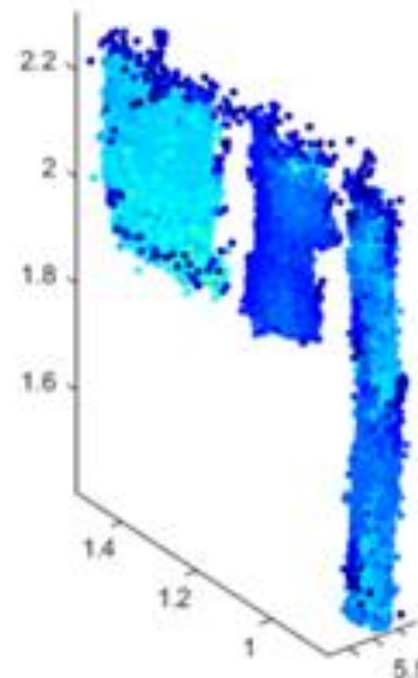
Target identification 2/3

- Targets were cropped from the point cloud => mean backscattered reflectance of all echos
- Intensity of channels 4 (720 nm) and 8 (818 nm)

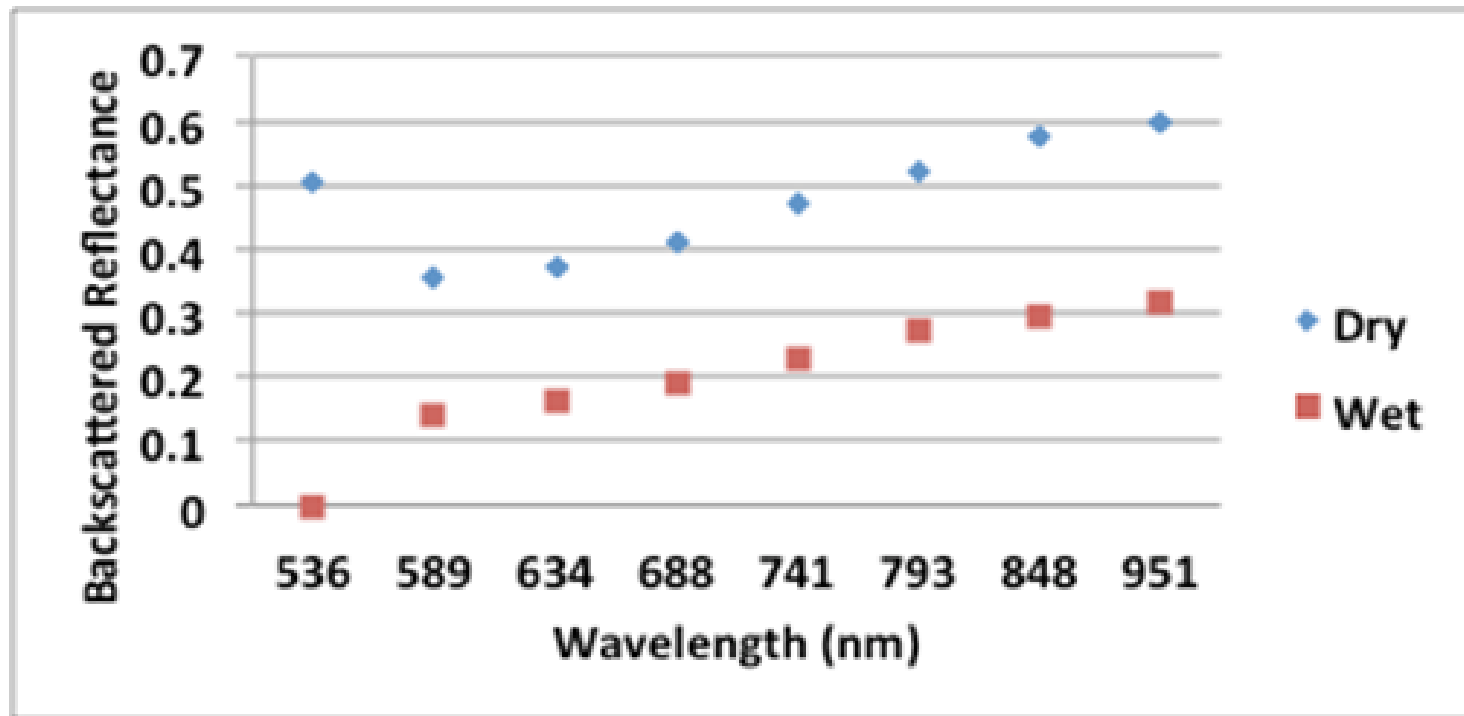
Color by channel: 4 = 720.2902 nm



Color by channel: 6 = 818.002 nm

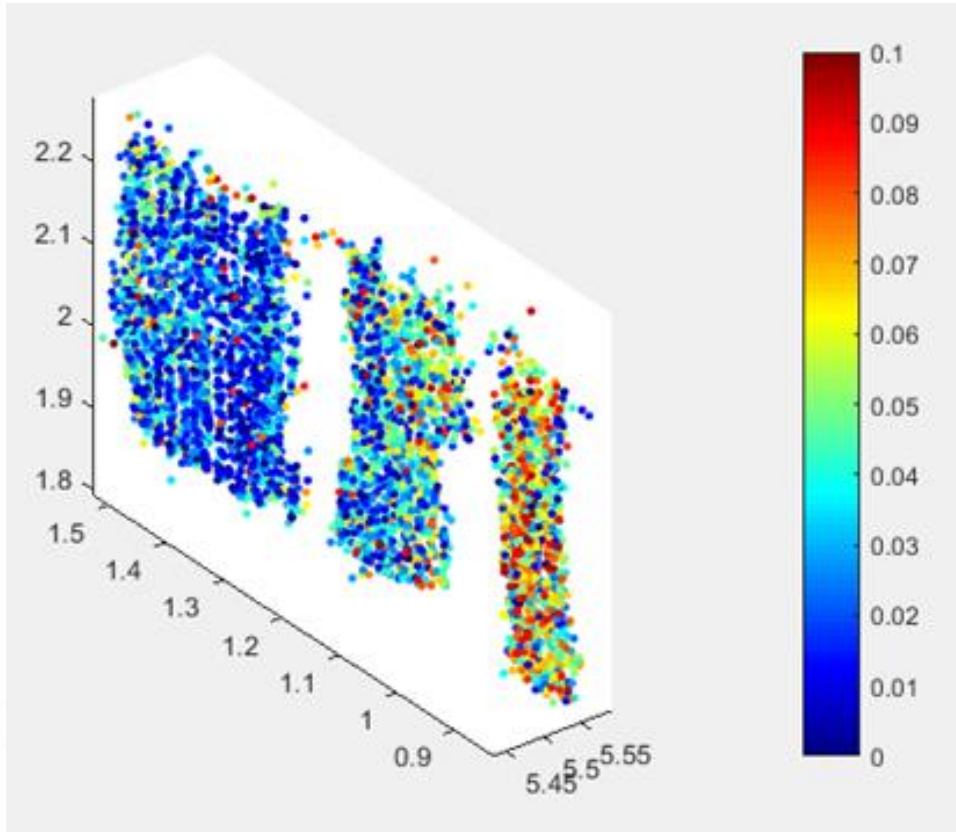


Target identification 3/3



- Spectra of the wet and dry cardboard panels
- Averaging the intensity of all points at each wavelength channel
- 6 % error in reflectance measurement
- Difference in the intensity visible at all wavelengths

Automatizing target identification



Normalized Water Index (NWI)

- Spectral identification using spectral indices
 - Water concentration index WI
 - Normalized water index NWI
 - Both use
 - Water absorption band 970 nm
 - Reference wavelength
 - $WI(\text{dry}) = 0.96$, $WI(\text{wet}) = 0.94$
 - $NWI(\text{dry}) = 0.02$, $NWI(\text{wet}) = 0.33$

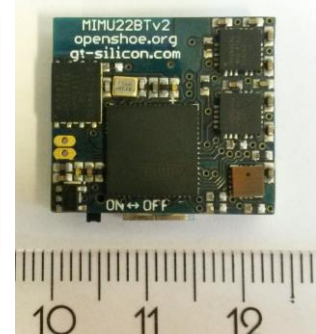
Infrastructure-free positioning

- GNSS positioning is unavailable indoors
- Other radio positioning requires infrastructure and preparation
- Self-contained sensors provide motion measurements
 - Inertial sensors, magnetometers, odometers, ...
 - Measurement errors deteriorate the solution
- Multisensor fusion



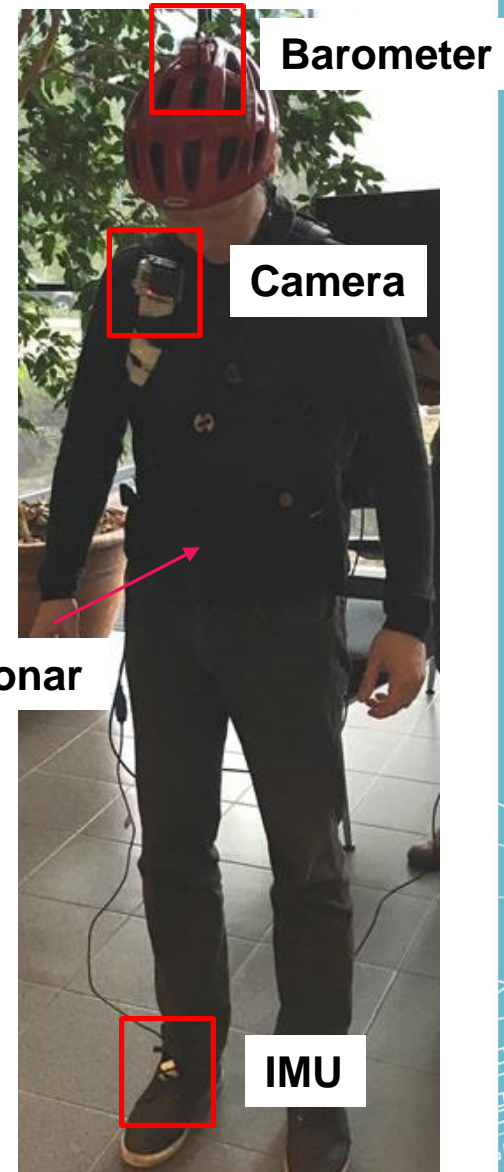
Multisensor position solution

- Sensors in our system:
 - 3 Inertial Measurement Units (IMUs)
 - one in foot (positioning),
body and helmet (motion recognition)
 - A camera
 - Sonar
 - Barometer for vertical position
- Measurements fused using Particle filtering
- **INTACT**- a project funded by the Ministry of Defence in Finland 1.1.2015-30.11.2017



Measurement models

- Inertial sensors mounted to the foot for horizontal positioning
 - Zero velocity updates improve the result
 - Calibrated using vision-aiding
- Barometer for vertical positioning
 - Computes height using pressure and temperature
 - Calibrated using sonar



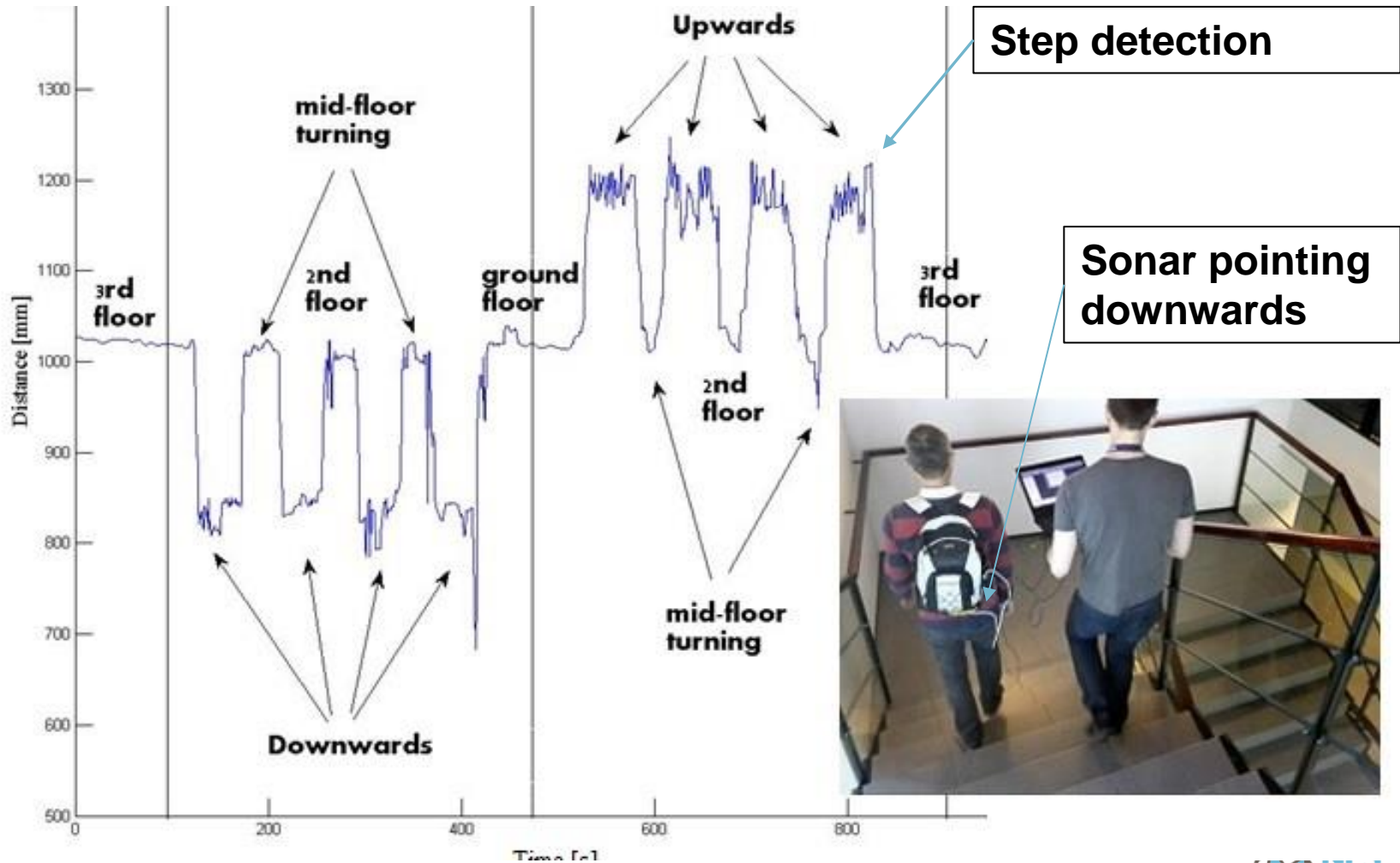
Vision-aided positioning

- Tracking features in consecutive images => motion of the camera may be computed
- Used as additional gyroscope and accelerometer
- Different error sources => complete inertial sensors



Ruotsalainen I., Vision-Aided Pedestrian Navigation for Challenging GNSS Environments, Doctoral Dissertation, 2013, TUT

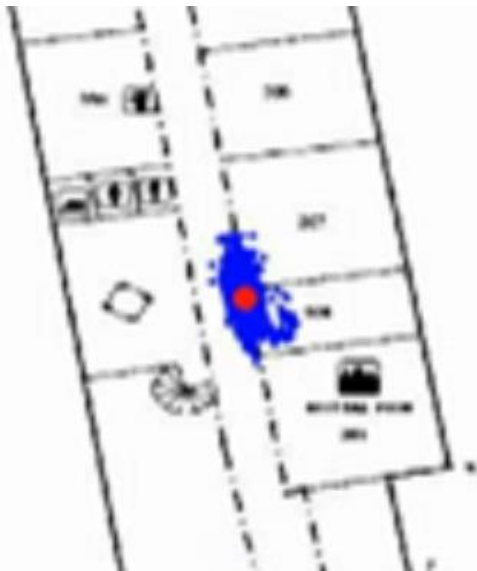
Sonar and barometer integrated for robust vertical positioning



Fusing measurements using Particle filtering

$$\mathbf{x}_k = [E \ N \ H \ \dot{\psi} \ \psi \ S]_k \quad \text{State vector}$$

$$\mathbf{z}_k = [\dot{\psi}_{footIMU} \ \dot{\psi}_{visual} \ S_{footIMU} \ S_{visual}, H_{baro}, H_{ultrasonar}] \quad \text{Measurement vector}$$



- Particle filtering is Bayesian estimation method for non-linear and non-Gaussian measurements
- Resampling of particles based on weights
- State at each epoch k is the weighted mean of particles

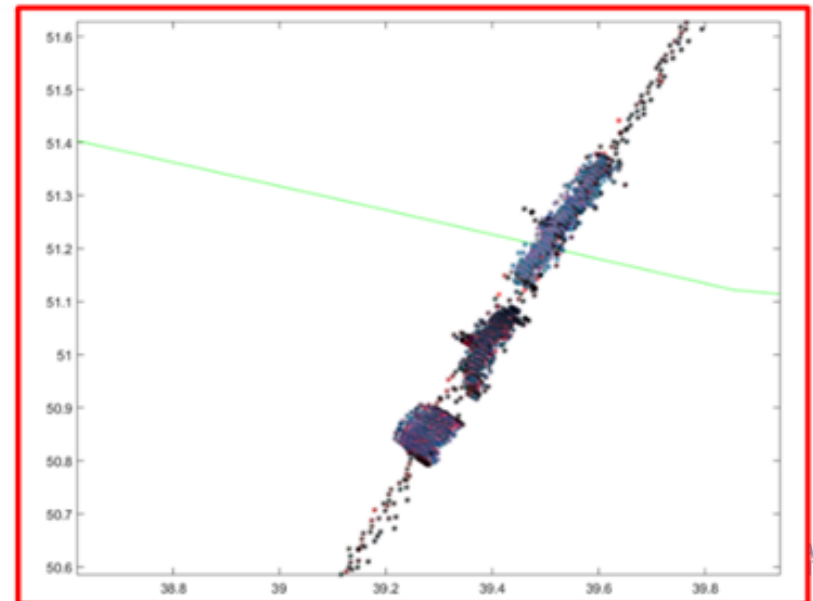
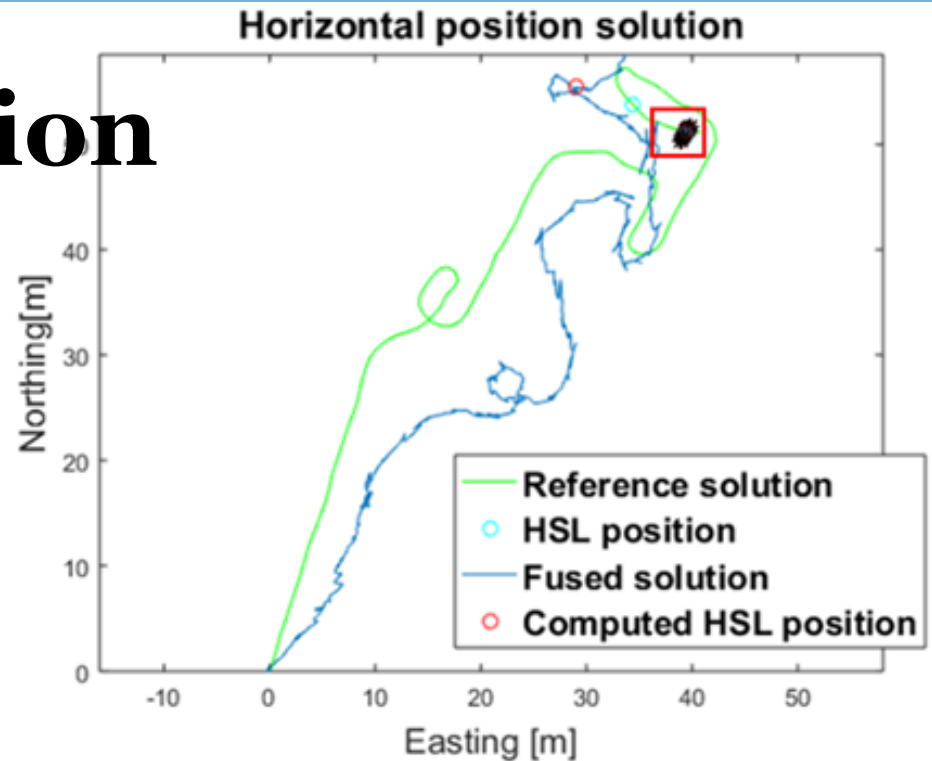
Positioning results



- Position solution from outdoors to the HSL location
- Route 200m
- Challenging environments including spiral stairs etc
- Equipment:
 - GoPro camera
 - Osmium MIMU22BT IMU
 - XSENS Mti-G-700 barometer
 - HRUSB-MaxSonar
 - Novatel SPAN for reference

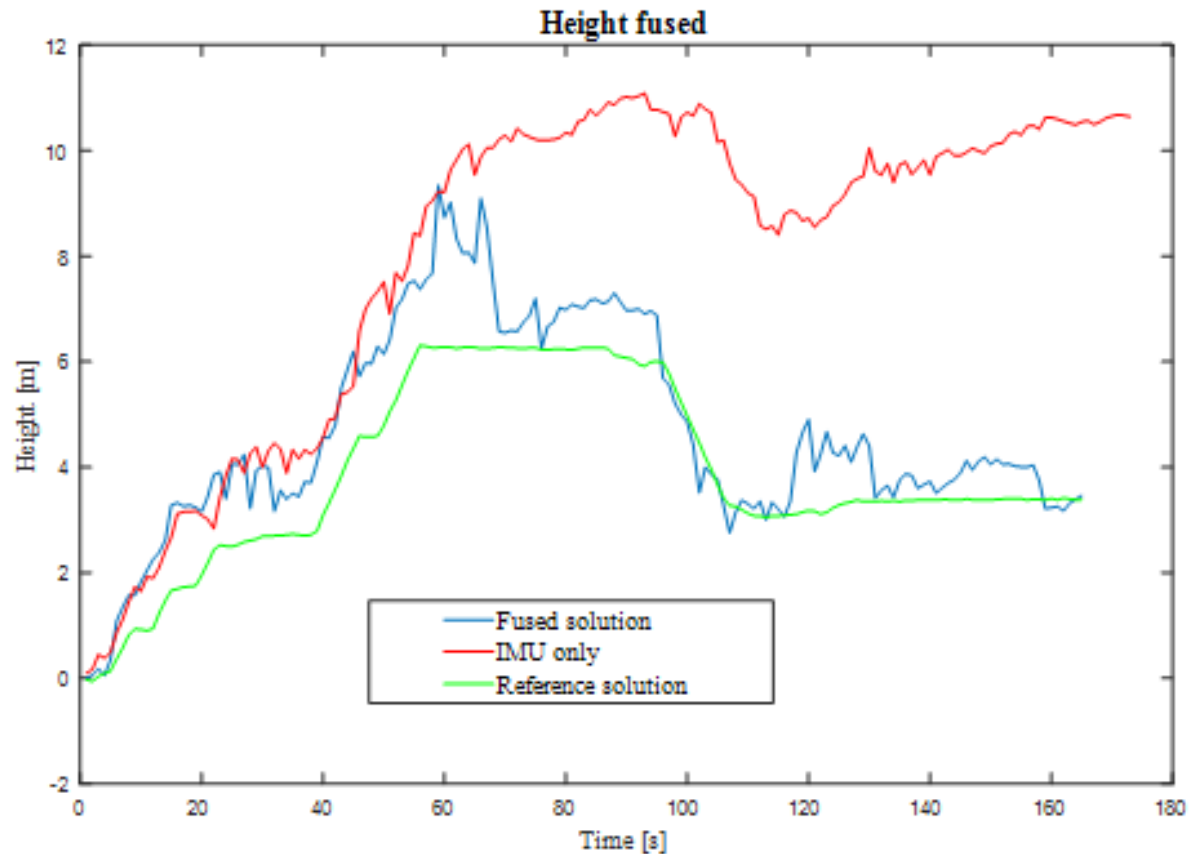
Horizontal position solution

- Distance Root Mean Squared (DRMS) error was 3.4 m
- IMU only DRMS 6 m
- Below close-up showing the target



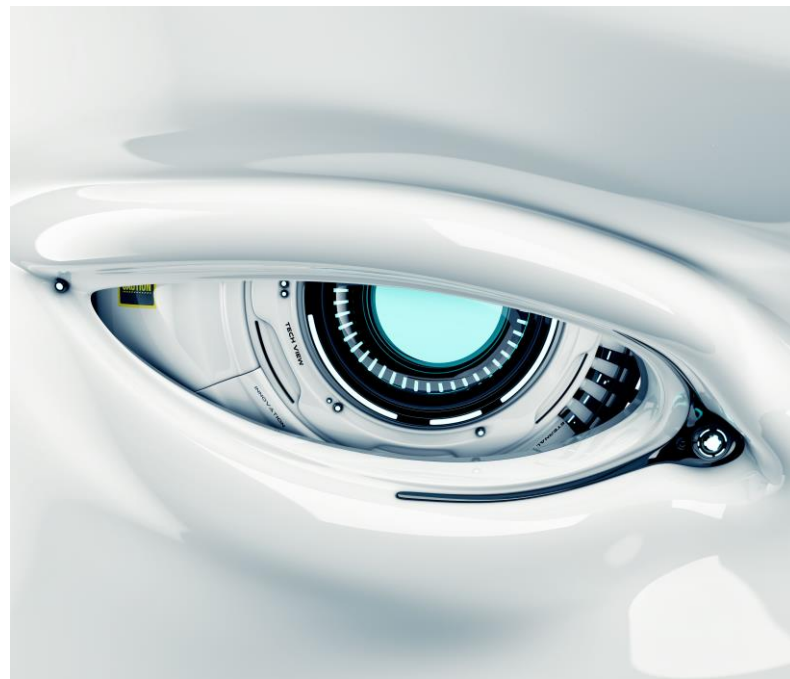
Vertical Position Solution

- Mean error for fused solution 1 m, std 0.6 m
- Mean error for IMU only 4.2 m, std 2.1 m



Conclusions

- Normalized Water Index is feasible for automatic target detection
- Accuracy of the fused position solution seems feasible for indoor positioning for many applications
- In future these two will be fused for more accurate and target identification enhanced SLAM solution
- Long term goal is to obtain a method for autonomous surveillance for e.g. tactical and security applications



More information on our website



www.fgi.fi

Follow us



[@fgi_nls](https://twitter.com/fgi_nls)



Finnish Geospatial Research Institute



Finnish Geospatial Research Institute (FGI), NLS

Contact us



firstname.lastname@nls.fi

Thank you!