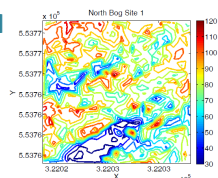


Early results from the first full waveform LiDAR survey over a lowland ombrotrophic peatland, and synthesis with hyperspectral Eagle-Hawk data

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Doreen Boyd
Ted Milton
Cheyne Hadley
Christopher Koster



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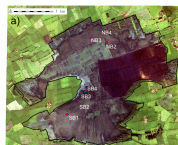
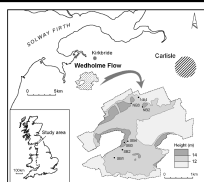


Aims of project

- Determine information content of FWL data
 - context of short sward vegetation
- Test the processing chain for FWL data
- Consider operational data processing and handling issues
 - Develop methods for processing
 - Feedback to ARSF-DAN
- Compare with existing LiDAR data from a well-instrumented site
 - Peatland site with demonstrated links between top of canopy structure and underlying hydrology / biodiversity

Study site

- Wedholme Flow, Cumbria
- Ombrotrophic peatland
- Range of condition classes
 - Wet, intact, diverse *Sphagnum*
 - Dry, degraded, cut over, shrub-dominated
- Good test bed for short-sward vegetation
- NERC-funded work on LiDAR DSMs and vegetation structure showed relevance of ALS technique



Anderson, K., J. J. Beattie, et al. (2010). "Combining LiDAR and IKONOS data for ecological classification of a lowland ombrotrophic peatland." *Journal of Environmental Quality* 39: 1-14.

Anderson, K., J. J. Benzie, et al. (2009). "Laser scanning of fine-scale pattern along a hydrological gradient in a wetland ecosystem." *Landscape Ecology* **24**: 477–492.

Relationship to NERC strategy

- Biodiversity resource
- Spatial data required for ecosystem service evaluation
- Hydrology, ecology and C sequestration linked
- “compost bomb” instability in peatlands under c.c.*
- Many relationships to NERC-banners including *climate system, biodiversity, technologies*



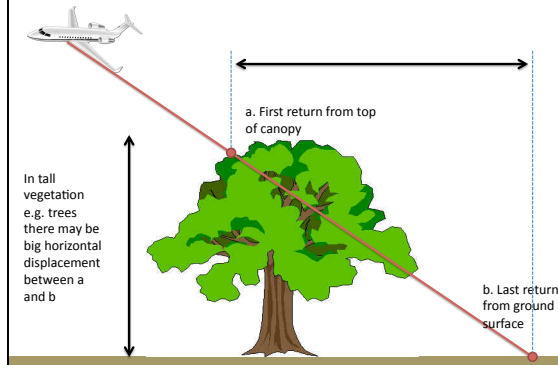
C. M. Luke et al. **Soil carbon and climate change: from the Jenkinson effect to the compost-bomb instability.** *European Journal of Soil Science*, 2010

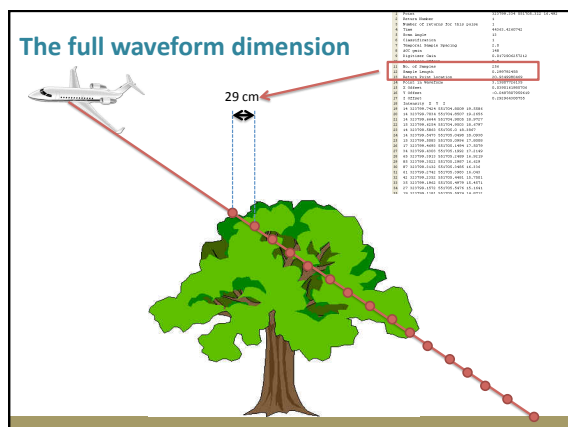
1. Data format

- Text file for every point on the land surface

[illegible]

What do the data mean? Standard ALS...





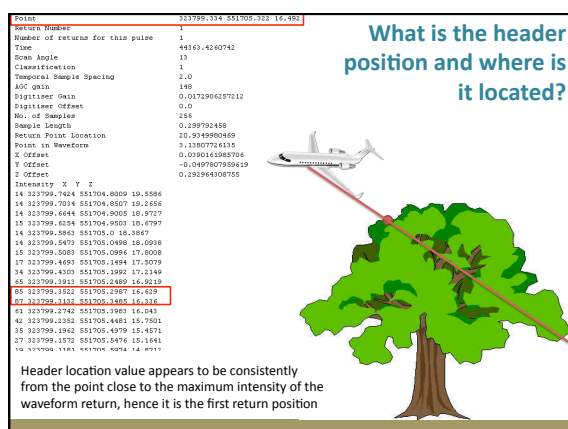
Data format

- X,Y,Z co-ordinate in header for the "point"

```

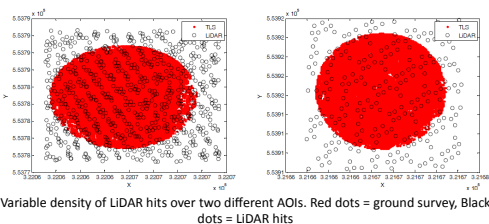
Point 323799.314 551705.322 16.480
Scanner 101213.9132 609954.0656 1550.1159
Time 42436.52221
Channel 1 count 256
Sample length 0.29979
Point 3.31792
Vector x -1.316537E-005
Vector y 2.242169E-005
Vector z 1.475535E-004
Channel 1 samples
  
```

Point	Intensity	X	Y	Z
1	0.017906157112	323799.314	551705.322	16.480
2	0.017906157112	323799.314	551705.322	16.480
3	0.017906157112	323799.314	551705.322	16.480
4	0.017906157112	323799.314	551705.322	16.480
5	0.017906157112	323799.314	551705.322	16.480
6	0.017906157112	323799.314	551705.322	16.480
7	0.017906157112	323799.314	551705.322	16.480
8	0.017906157112	323799.314	551705.322	16.480
9	0.017906157112	323799.314	551705.322	16.480
10	0.017906157112	323799.314	551705.322	16.480
11	0.017906157112	323799.314	551705.322	16.480
12	0.017906157112	323799.314	551705.322	16.480
13	0.017906157112	323799.314	551705.322	16.480
14	0.017906157112	323799.314	551705.322	16.480
15	0.017906157112	323799.314	551705.322	16.480
16	0.017906157112	323799.314	551705.322	16.480



2. Density of points

- Note that density of LiDAR points will vary depending on data capture
- Flightline overlap regions will be denser



Data volumes

- Health warning
- 256 FW values per point (high redundancy)
- In 10 x 10 m AOI = up to 750 text files
- For entire peatland (5 x 7 km) = several million files
- Large data processing burden
- Raw text files not ready for use with standard IP software
- Requires coding capabilities (matlab, R, C etc.) to digest efficiently



Point	Intensity	X	Y	Z
1	0.017906157112	323799.314	551705.322	16.480
2	0.017906157112	323799.314	551705.322	16.480
3	0.017906157112	323799.314	551705.322	16.480
4	0.017906157112	323799.314	551705.322	16.480
5	0.017906157112	323799.314	551705.322	16.480
6	0.017906157112	323799.314	551705.322	16.480
7	0.017906157112	323799.314	551705.322	16.480
8	0.017906157112	323799.314	551705.322	16.480
9	0.017906157112	323799.314	551705.322	16.480
10	0.017906157112	323799.314	551705.322	16.480
11	0.017906157112	323799.314	551705.322	16.480
12	0.017906157112	323799.314	551705.322	16.480
13	0.017906157112	323799.314	551705.322	16.480
14	0.017906157112	323799.314	551705.322	16.480
15	0.017906157112	323799.314	551705.322	16.480
16	0.017906157112	323799.314	551705.322	16.480

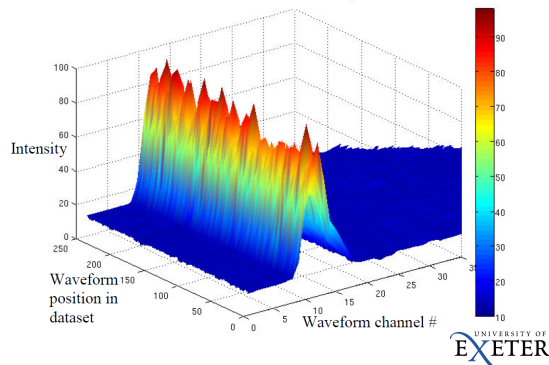
Results 1: header elevation value Validation against EA LiDAR

- N=25
- 8 sites distributed across peatland

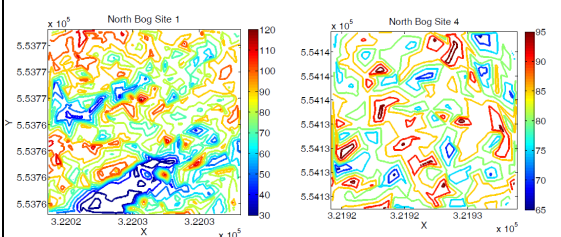
Site	Mean difference (m)
North bog 1	-0.04
North bog 2	0.08
North bog 3	-0.10
North bog 4	-0.02
South bog 1	-0.10
South bog 2	-0.14
South bog 3	-0.25
South bog 4	-0.11
Mean	-0.09



Results 2: Basic waveform analysis



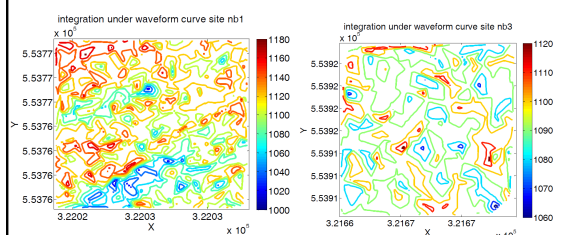
Using Matlab, it is possible to plot the spatial patterns in waveform characteristics



- Contour plots of peak intensity values in the waveform.

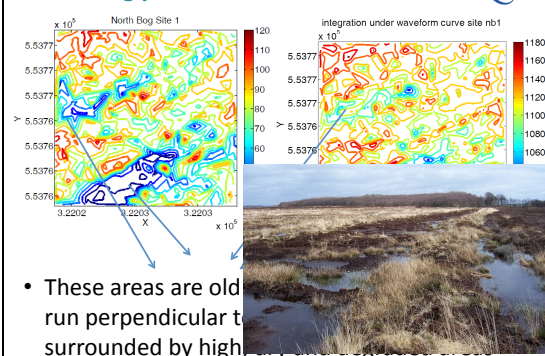
- Lower values = blue, higher values = red

Using Matlab, it is possible to plot the spatial patterns in waveform characteristics

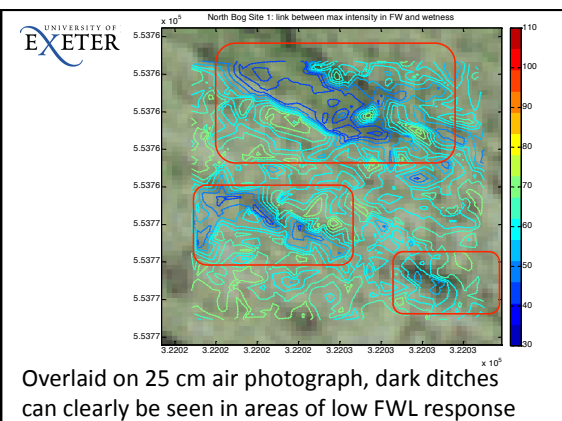


- Contour plots of waveform integral
- Lower values = blue, higher values = red

Interesting patterns at NB1

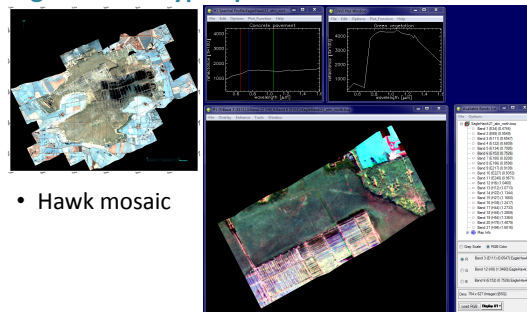


- These areas are old run perpendicular to surrounding high ground

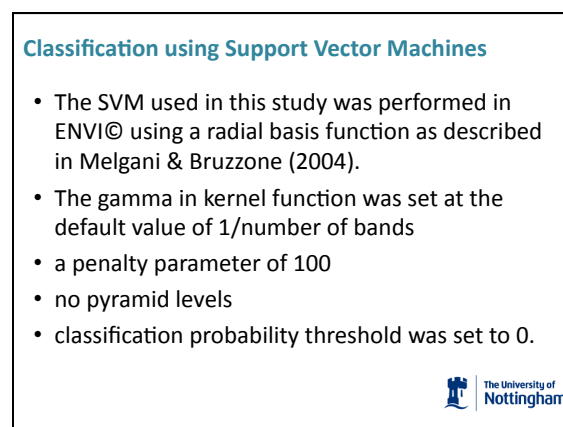
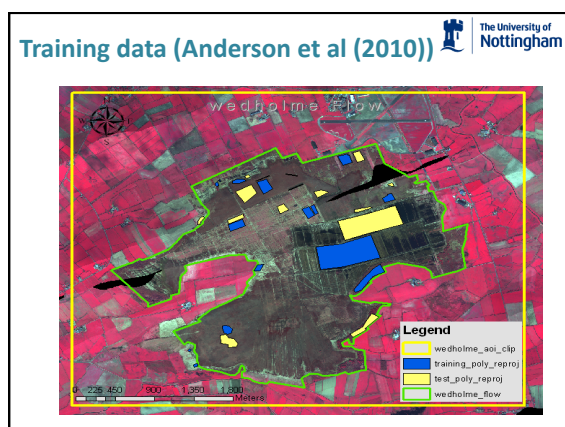
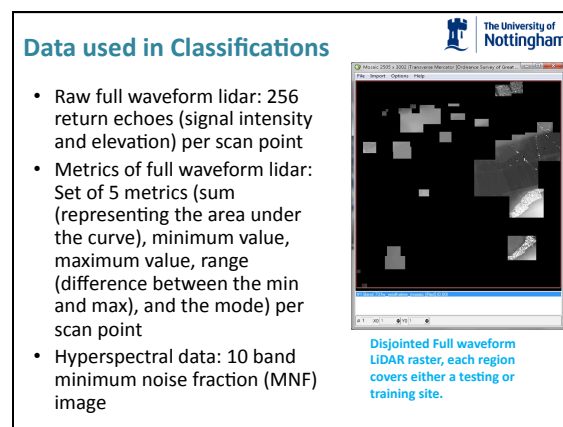
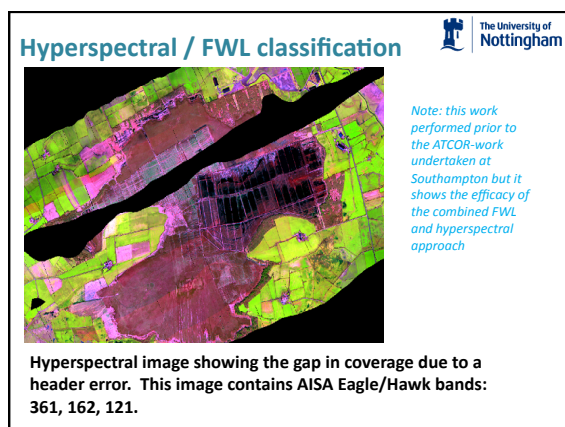
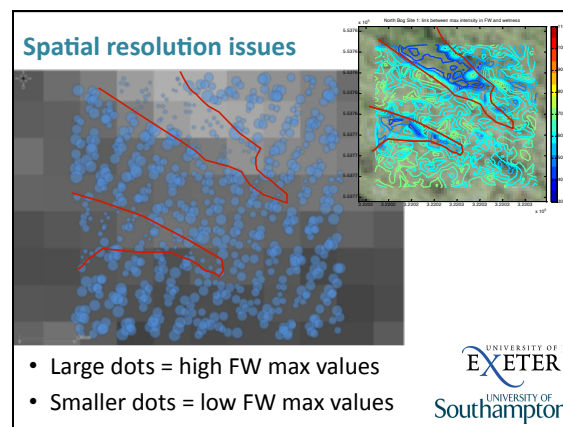
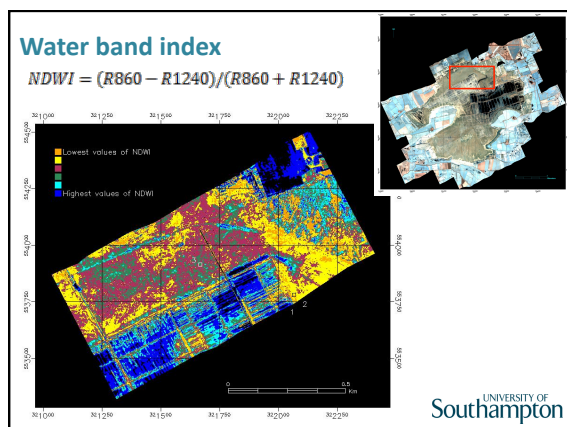


Stage 3: enter hyperspectral...

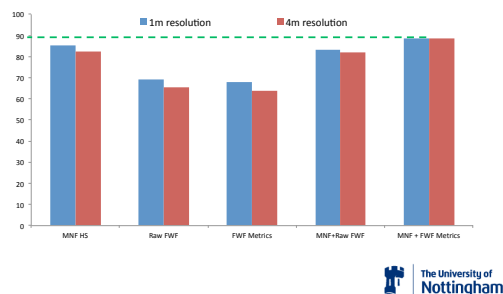
Southampton



- Atmospherically corrected using ATCOR-4



Overall Classification Accuracies (%)



Analysis per class (4m SVM classification)

	Optimal Dataset	Classification accuracy (%)
Active raised bog	MNF HS + FWF metrics	98.0
Active raised bog, degraded	MNF HS + Raw FWF	96.9
Drained and degraded (Calluna dominated)	Raw FWF	32.0
Drained and degraded (Erica dominated)	MNF HA + FWF metrics	25.0
Carr woodland	MNF HS + Raw FWF	98.7
Bracken	MNF HS + Raw FWF	35.3
Tussock grassland (Molinea dominated)	MNF HA + FWF metrics	63.8
Milled unvegetated peat	MNF + FWF metrics	92.6
Eriophorum bog	Raw FWF	97.0

In each case the optimal dataset for classification always includes a full wave form dataset, either in combination with the hyper-spectral dataset or exclusively.



Conclusions

- Full waveform LiDAR data have high volume
 - Processing effort / requirement beyond standard RS
- We have presented results based on a few focused sites
 - Rolling this out over full extent is difficult
- Resolution difference between LiDAR point and hyperspectral data
- Initial results suggest that FWF intensity is proxy for near surface moisture

Recommendations to ARSF

- I would suggest that higher level product retrieval from FWL data could be an ARSF-DAN focus for the future
 - x, y, z product of maximum intensity
 - X, y, z product of waveform integral
 - X, y, z product describing waveform shape & extent
- For many users this would be the most efficient way of integrating this additional information into classifications etc.

Acknowledgements

- ARSF-DAN
 - Emma Carolan
 - Mike Grant
- MSc students
 - Cheyne Hadley (Notts)
 - Chris Koster (Exeter)