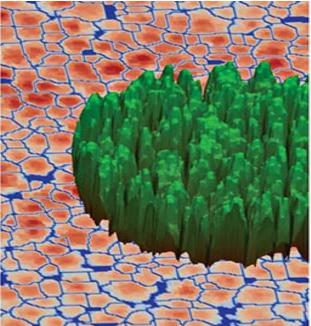


Use of hyperspectral imagery within forestry to detect nutritional deficiencies

Michael Watt, Grant Pearse, Jonathan Dash, Nathanael Melia













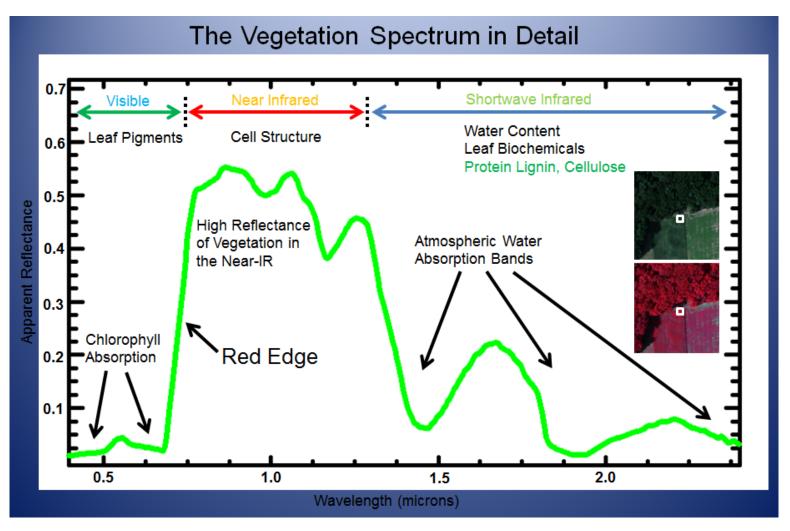
Outline

- What is hyperspectral imagery?
- Platforms that could supply imagery
- Potential uses within NZ forestry focus on detection of nutritional deficiencies
- · Further research in this area

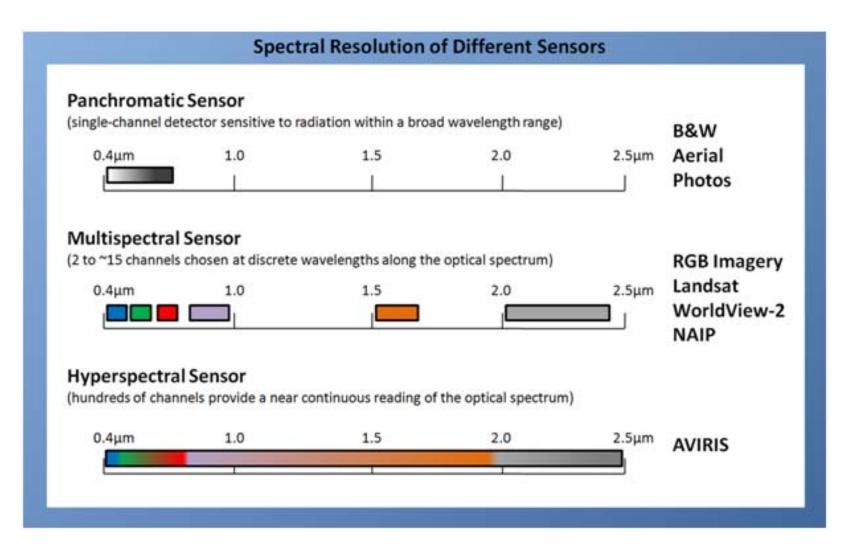




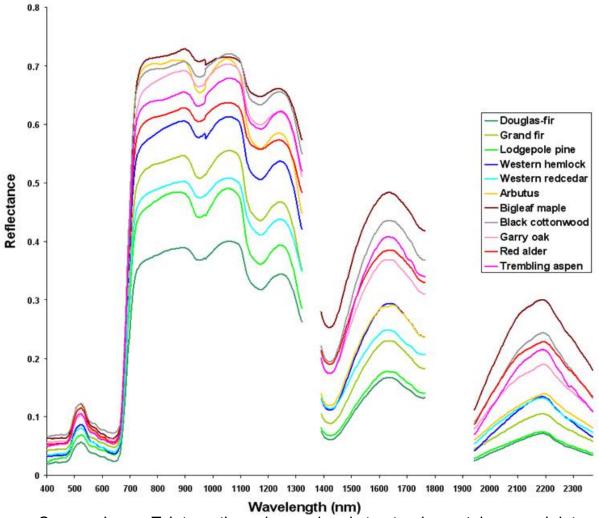




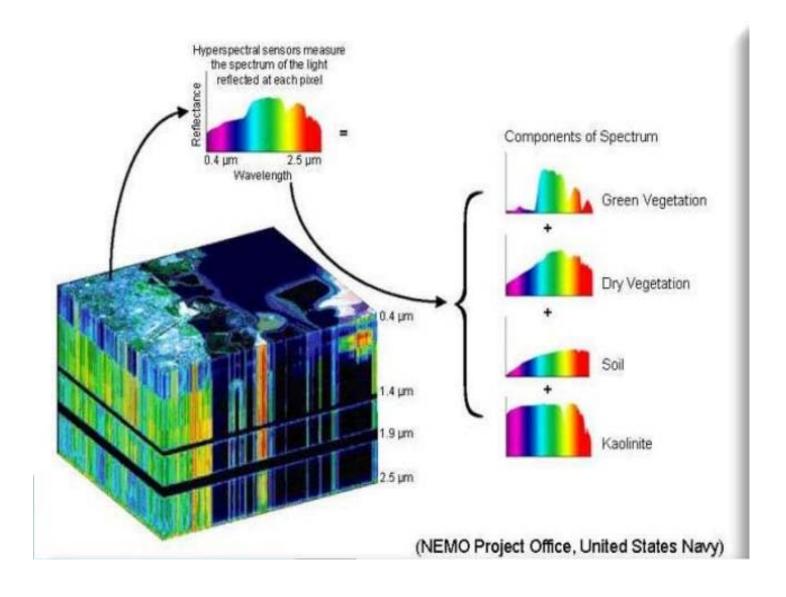
Source: Elowitz, Mark R. "What is Imaging Spectroscopy (Hyperspectral Imaging)?". Retrieved from www.markelowitz.com/Hyperspectral.html



Source: Vegetation Analysis: Using Vegetation Indices in ENVI. Retrieved from http://www.harrisgeospatial.com



Source: Jones, T. Integrating advanced and structural remotely sensed data to improve vegetated terrestrial ecosystem mapping http://irsslab.forestry.ubc.ca/research/data-fusion/



Comparison between platforms

	Satellite	Aircraft	UAV
Spectral range	400 – 2500 nm (Hyperion)	380 – 2,500 nm (AISA FENIX)	Mainly 400 – 1000 nm, with separate sensors covering 1000 – 2500 nm
Spectral resolution	10 nm	3.5 nm (VNIR) 10 nm (SWIR)	2 – 10 nm, ca. 5 nm
Spatial resolution	18 m (Proba-1) – 30 m (Hyperion)	1 m	< 1m potential
Cost	Moderate	\$10 – 15/ha	High cost/ha but small areas possible





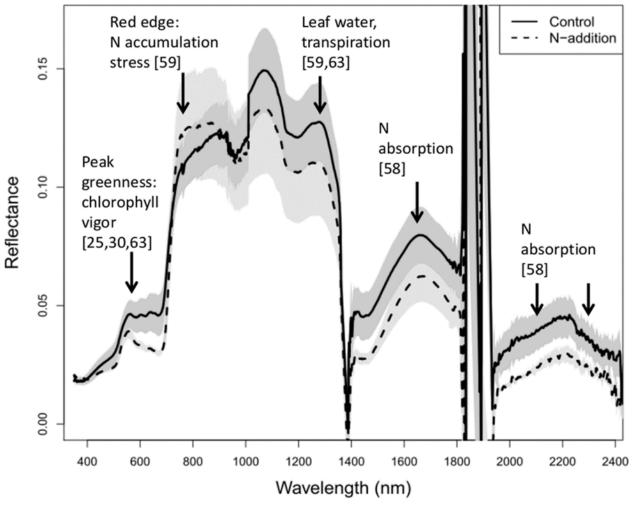


Use of hyperspectral imagery to detect nutrient deficiency

- Fertilisation important practice provides significant growth gains at mid-rotation
- Stands respond well to N and P additions
- Identifying stands where there will be a cost effective response is difficult
- Use of hyperspectral imagery could assist with this as symptoms are often quite visible









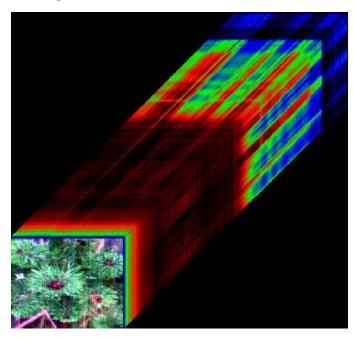
O'Connell, J. L., Byrd, K. B., & Kelly, M. (2014). Remotely-sensed indicators of N-related biomass allocation in *Schoenoplectus acutus*. *PloS one*, *9*(3), e90870

Precision of models (R^2) developed for predicting foliar element concentration

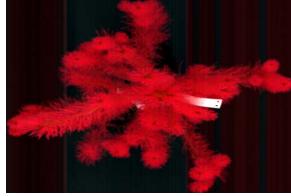
											. Prodiction using in situ data for
Species	N	Р	Ca	K	Mg	Zn	Fe	В	Mn	Cu	 Prediction using in-situ data for
In situ											N, P - high strength
Slash pine	0.99	0.88									ri, r - mgn suengui
Loblolly pine	0.99										
Mixed species	0.99	0.94	0.95								
Aleppo pine	0.94	0.95									
Maple	0.92										 Using fixed wing airborne data
Loblolly pine	0.81	0.70	0.42	0.68	0.51						9
Tropical	0.81	0.68	0.69	0.55	0.61	0.27	0.48	0.43	0.29		predictions low. – high strength
Tropical	0.76	0.62									predictions low. Thigh strength
Balsam fir	0.66										
Scots pine	0.37	0.32	0.31	0.16	0.04	0.33	0.07	0.11	0.24	0.04	
Airborne Eucalyptus Mixed sp. Mixed Oak Range Mixed sp.	0.88 0.87 0.85 0.83 0.79	0.23	0.63	0.56	0.64						 Using satellite data predictions low – high strength
Temperate Mixed sp. Mixed sp. Mixed sp. Mixed sp. Mixed sp.	0.79 0.76 0.72 0.63 0.53	0.20									 Predictions other elements lower strength than N
Satellite Mixed Oak Range Eucalyptus Temperate Radiata pine	0.98 0.82 0.62 0.60 0.44	0.28		0.68		0.14	0.41	0.56		0.45	UAV data has not been used

Scion Hyperspectral Pilot study

- Specim FX10 VNIR Sensor (400-1000 nm)
- Collaboration with University of Trier, Germany
- Herbicide induced mortality of Pinus contorta
- UAV mountable sensor





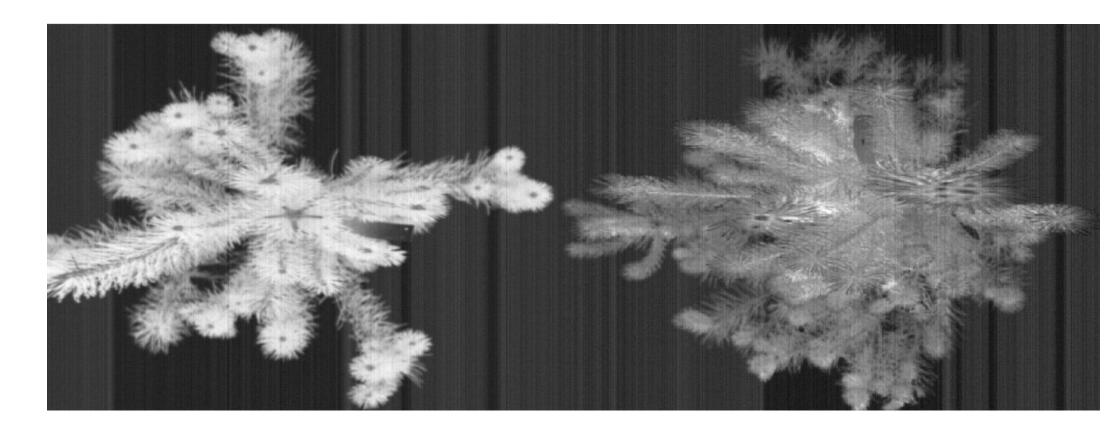








Modified Chlorophyll index – Artificially induced stress (right)









Experiment at Scion

- Plants will be grown in pots for two years under range of N and P
- Hyperspectral data obtained from fixed platform and UAV
- Models of N, P and chlorophyll content will be developed
- Should provide insight into viability of detecting nutrient deficiencies
- Further research will look at scaling models to larger areas using satellite (EnMAP) data





Michael Watt Research Leader – Geomatics michael.watt@scionresearch.com

April 2018

www.fgr.nz www.gcff.nz www.scionresearch.com

