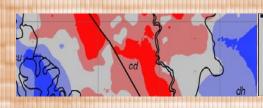


Scope to map soil management units at district level from γ -ray (remote) and EM (proximal) data

Jing D, Huang J, Banks P, Triantafilis J

Problem Definition?

Materials and Methods







Results, Discussion & Conclusions



Talk Outline

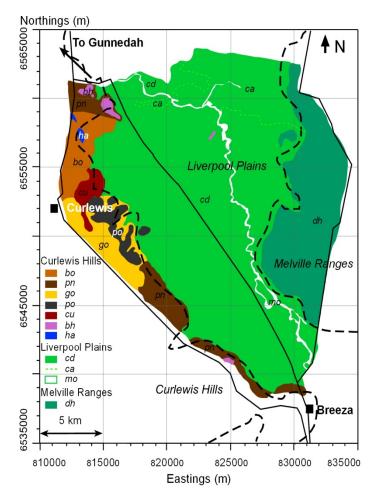
Problem Definition



- Traditionally requires
 - Field survey and morphological data and limited chemical data
 - Followed by classification of soil profile into pre-existing soil classification system
- Expensive, time-consuming, labour-intensive & not useful for management (e.g. soil colour)



Problem Definition

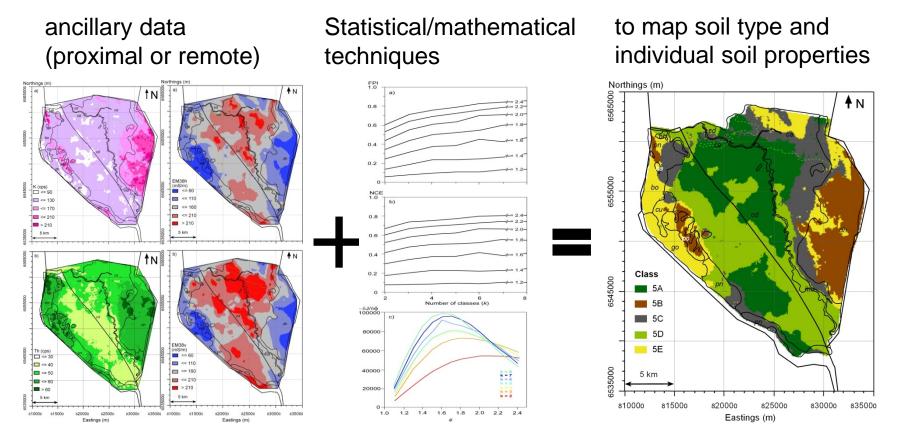


- Results in a soil map
 - Limited samples due to cost
 - Based on extrapolation using air-photos
 - Based on subjective interpretation of field surveyor to classify the profile and extrapolate and identify the soil landscape units



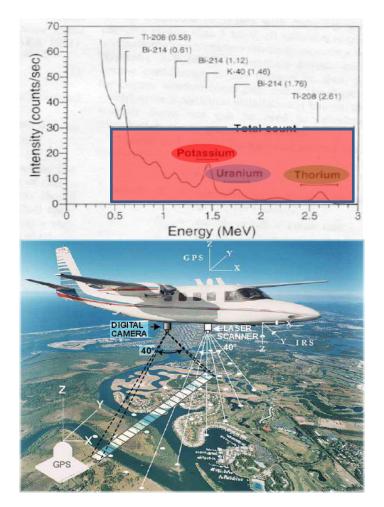
Aim: Is there an alternative?

• Digital soil mapping (DSM) is the use:





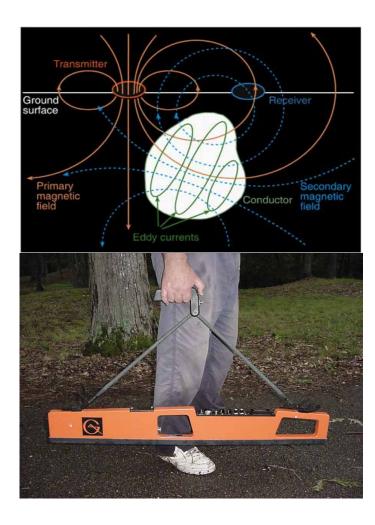
Ancillary data: γ-ray (remote)



- Natural radiation from Earth's
 - Rocks, and
 - Soil
- Commonly measures
 - K,
 - Th and
 - U counts
 - with Total Count (TC)



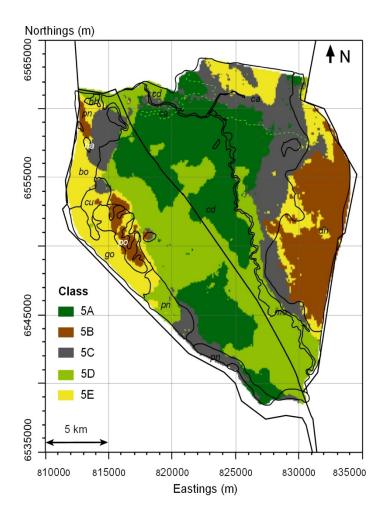
Ancillary data: EM (proximal)



- Relationship with morphological and chemical properties, for example;
 - clay content,
 - CEC,
 - salinity,
 - moisture
- Measures apparent electrical conductivity EC_a (mS/m)



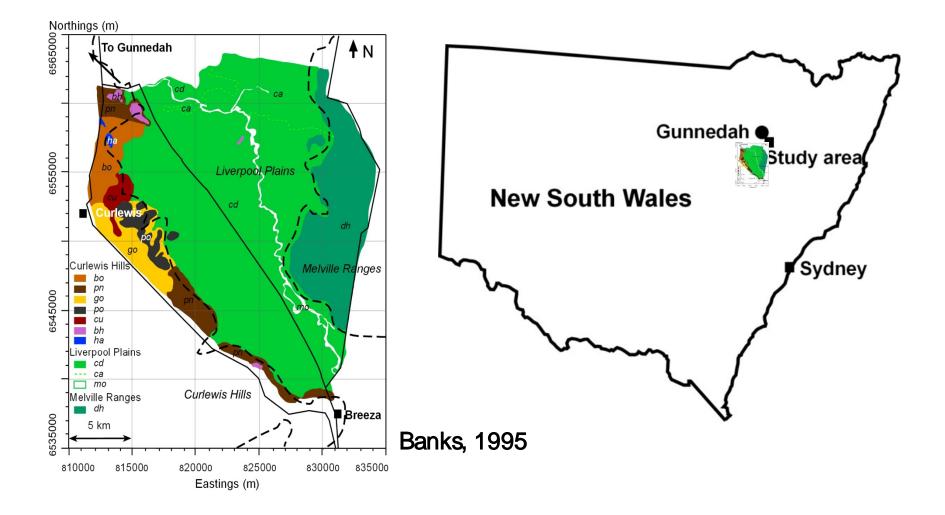
Aims



- To assess merit of ancillary data, such as remote (γ-ray) and proximal (EM), to identify soil management units using FKM
 - To minimise the MSPE of soil properties for greater accuracy in classifying landscape units
- To produce soil map direct implications for farmers

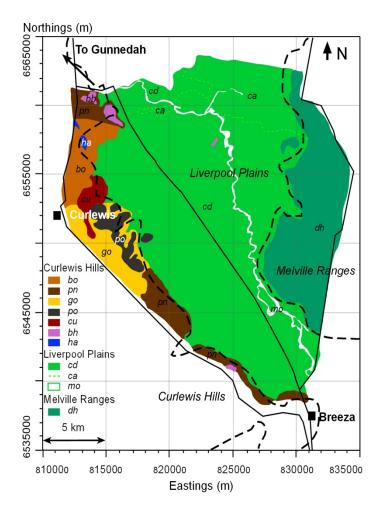


Materials & Methods: Study Area (Physiography)





Materials & Methods: Study Area (soil associations)



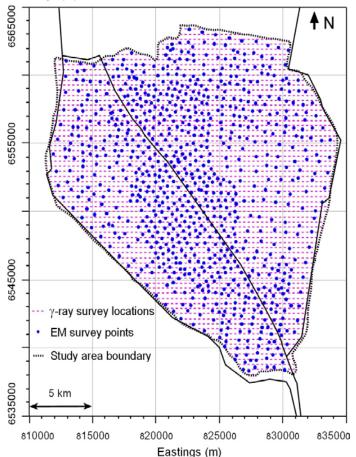


- Curlewis Hills
 - bo Booloocooroo alluv., sandstone & basalt
 - *cu* Curlewis Swamp deep clays under peat
 - bh Battery Hill shallow stoney basaltic soil
 - ha Hartfell
 rhyolitic soil
- Liverpool Plains
 - *cd*–Conadilly Floodplain alluv. from basalt
- Melville Ranges
 - E.g. *dh* Dead horse
 - Diverse, undifferentiated T and Q alluvial



Materials & Methods: ancillary data collection

Northings (m)



- 400m spacing flight lines;

– K, Th, U & TC

• EM38

γ-ray



- 500m spacing in irrigated and 1km on dryland;
- Kriged
 - Common 100m grid



Materials & Methods: Clustering

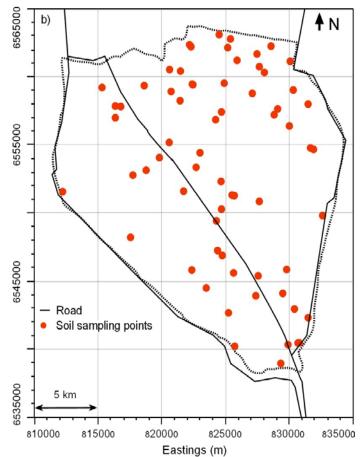
📅 FuzME Input Interface	e ©	ММШ АСРА		
FuzME	Save	ot Discrim. Exit		
Files	Fuzzy Clustering	Bootstrap/Jackknife		
Write distance	Fuzzy Exponent (>= 1) 1.30 Multiple runs Start End Step 1.10 2.00 0.10 ✓ Perform fuzzy discriminant analysis ✓ Plot Princ Component when clustering	Image: Fuzzy k Means Maximum Iterations 300 Stopping criterion 0.0001		
Mumber of Classes	Algorithm	Fuzzy k Means with Extragrades		
Min no. of classes (>1)	© Fuzzy k Means 2 C Fuzzy k Means with Extragrades	Fix Alfa to 1.0 Alfa Iterations 60		
Max no. of classes ((<100)	Fuzzy k Means equal area F Fkm with Fuzzy Covariance Matrix	Tolerance 0.0001 Define average extragrade membership		

- Fuzzy k-means (FKM) analysis of ancillary data only
- FuzME program (Minasny et al., 2012)
 (ver. 3.5c)
- *k* = 2, 3, 4, 5, 6, 7 and 8



Materials & Methods: : laboratory analysis

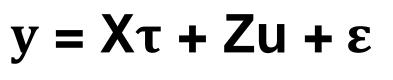
Northings (m)



- Soil sampling locations: 71
- Two representative depths;
 - topsoil 0-0.3m, and
 - subsoil 0.9-1.2m
- Texture: Clay, silt & sand %;
- EC_e and pH;
- Exch. cations Ca, Mg, K, Na
- CEC; and ESP



Materials & Methods: statistics and mathematics



Response Fixed effect Random effect

Residual

 $\mathbf{y} = \mathbf{X}\mathbf{\tau} + \mathbf{Z}\mathbf{U} + \mathbf{\varepsilon}$ • Residual Maximum Likelihood Analysis (REML)

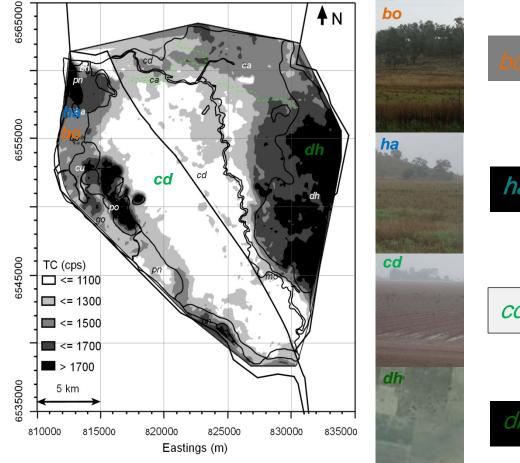
- Calculate Mean Squared Prediction Error (MSPE) of soil properties of each of the numerical classes
- Minimise MSPE



Results: γ-ray (Total counts – cps)



Northings (m)





ha > 1,700- high

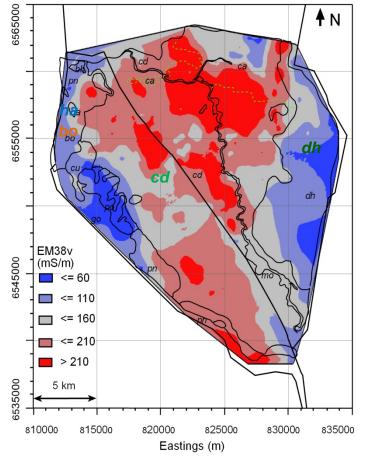
cd: < 1,100 - small

dh > 1,700- high



Results: EM38v (mS/m)

Northings (m)





bo: 110 - 160 – Intermediate

ha < 60 – 110 low/medium

cd: 160 – 210 high/very high

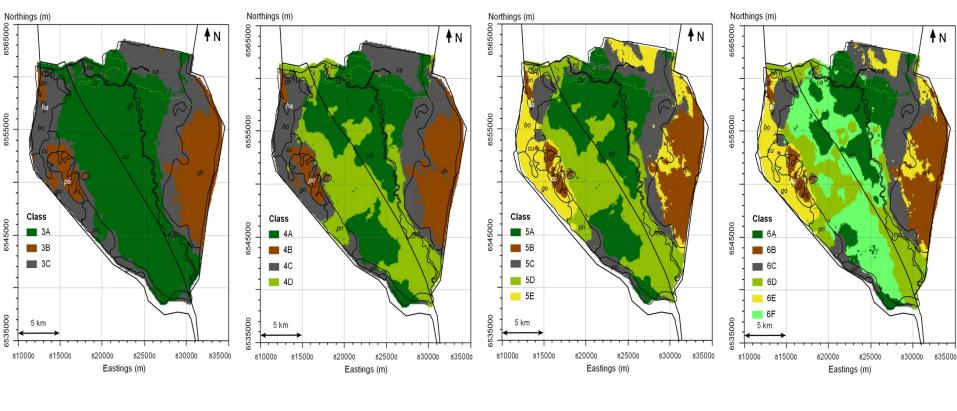
dh: 60 - 110 low/intermediate







Results: Cluster maps



k = 3

= 4

= 5

= 6



Results: MSPE for topsoil

No. of Classes (k)	2	3	4	5	6	7	8
Clay	159.97	106.68	114.72	93.37	115.16	104.84	97.26
Sit	57.31	49.15	53.12	47.62	49.40	52.15	51.73
Sand	213.46	66.88	65.39	60.68	71.30	62.31	62.57
ECe	2.07	2.15	2.17	2.18	2.22	2.35	2.36
рН	0.33	0.35	0.35	0.34	0.34	0.36	0.37
CEC	364.88	174.49	179.93	136.75	177.66	138.80	169.46
Ca	102.74	84.04	90.95	81.07	85.85	76.52	90.30
Mg	70.48	59.90	46.49	35.20	53.08	37.28	42.63
К	0.64	0.59	0.64	0.56	0.64	0.58	0.65
Na	7.51	7.47	7.48	7.49	7.96	7.76	7.94
ESP	37.42	37.21	38.09	39.05	40.47	39.20	41.22
No. of lowest MSE	2	2	0	6	0	1	0

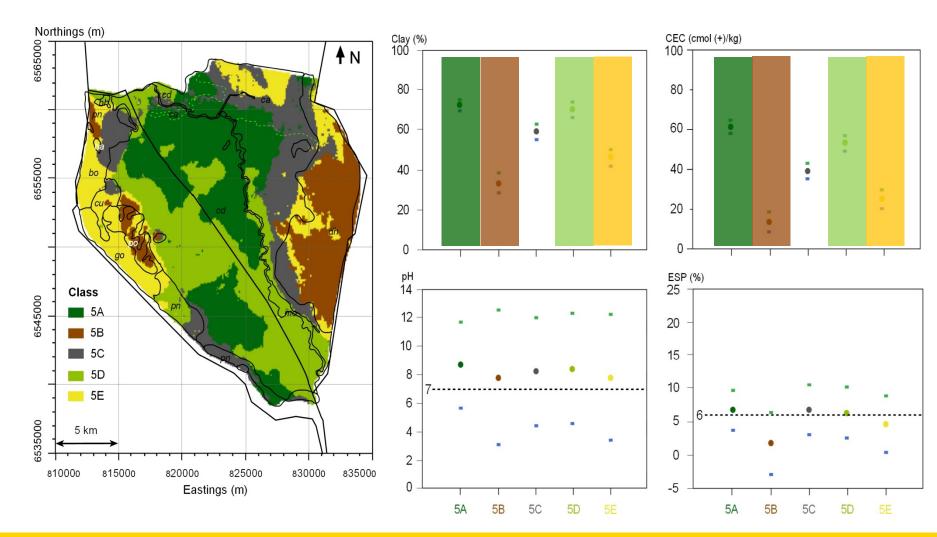


Results: MSPE for subsoil

No. of Classes (k)	2	3	4	5	6	7	8
Clay	248.79	203.58	202.30	168.72	455.98	176.93	183.97
SIt	89.05	84.24	87.60	75.64	91.51	74.61	79.27
Sand	208.65	127.28	122.02	115.45	123.38	121.96	128.71
EC _e	11.91	12.36	12.89	12.73	12.52	13.54	13.32
рН	0.36	0.34	0.36	0.35	0.37	0.31	0.35
CEC	122.50	123.88	126.17	128.76	137.91	137.06	144.36
Ca	77.36	62.42	64.32	54.06	61.33	53.28	55.12
Mg	0.39	35.11	0.40	0.36	0.40	0.37	0.40
К	19.51	18.96	19.37	19.02	20.77	19.80	20.18
Na	63.59	60.49	62.39	60.14	63.79	60.40	63.97
ESP	289.23	254.89	255.51	242.44	281.78	244.08	261.83
No. of lowest MSE	2	1	0	5	0	3	0

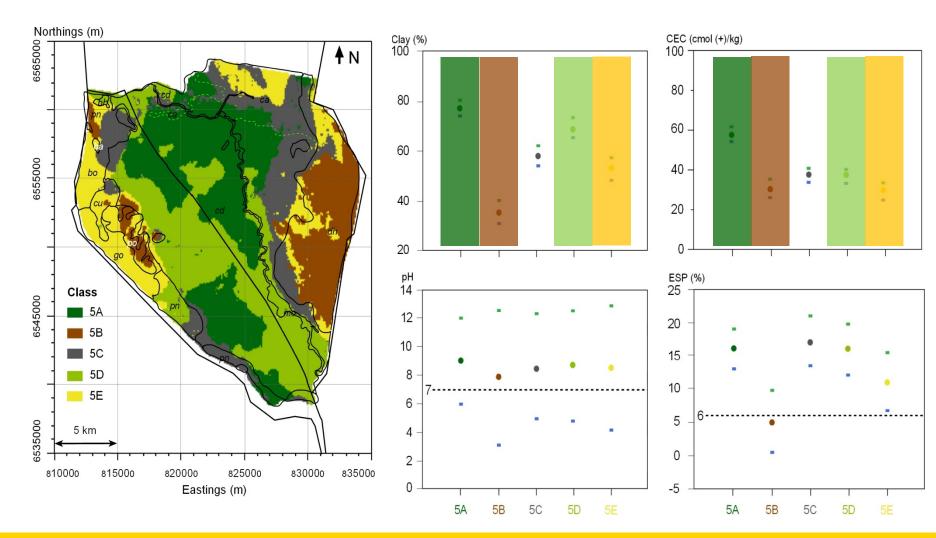


Results: topsoil properties w/ mean + SD



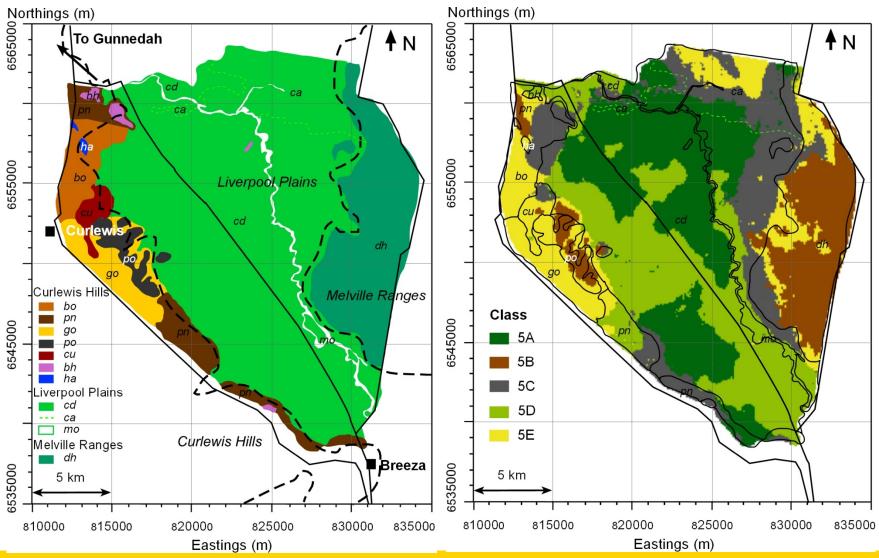


Results: subsoil properties w/ mean + SD



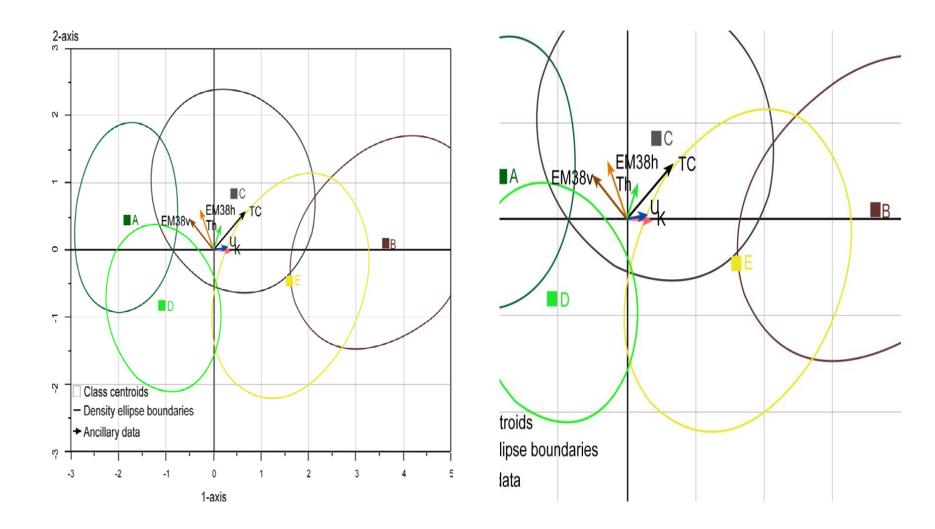


Discussion: Comparison



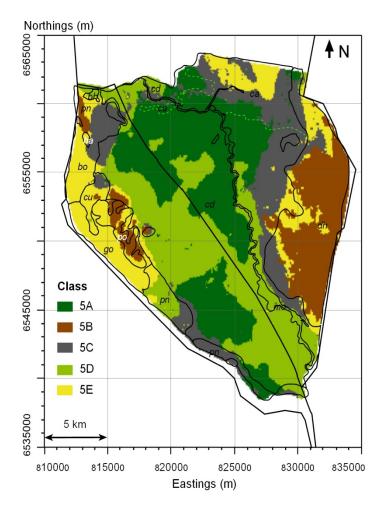


Results: Canonical analysis





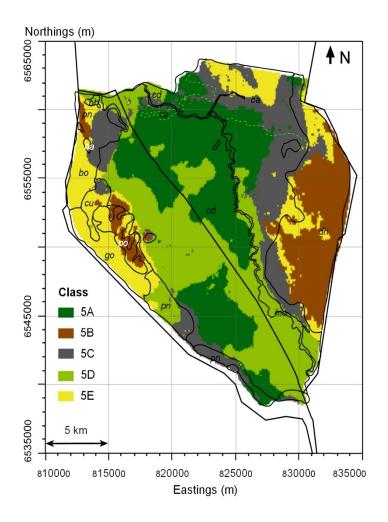
Conclusions: Key findings



- MSPE minimised when k = 5
- DSM consistent with major physiographical units
- DSM revealed subtle differences in highly productive Liverpool Plains physiographic unit



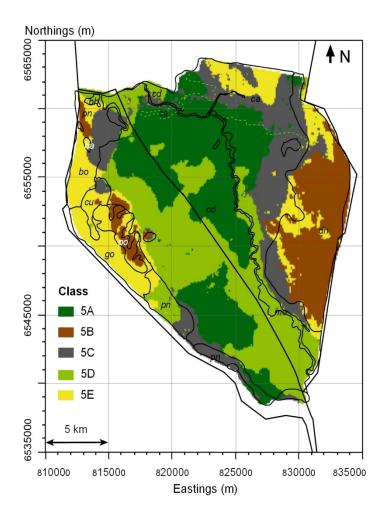
Conclusions: Key limitations



- Some smaller soil landscape units not reflected in DSM (e.g. *CU*, *ha*)
 - Add in more ancillary data such as DEM?
 - Due to significant local relief these units occupy
 - Because of resistant igneous outcrops of these units (e.g. basalt, rhyolite)



Conclusions



- γ-ray and EM can be combined to identify soil mgt classes;
- DSM was able to highlight subtle differences in diverse physiography;
- Addition of DEM may improve identification of smaller units;
- Potential for cheap, fast, objective,
 accurate and meaningful DSM for
 farmers on districts scales





Scope to predict soil type using proximally sensed gamma-ray spectrometer and EM induction data





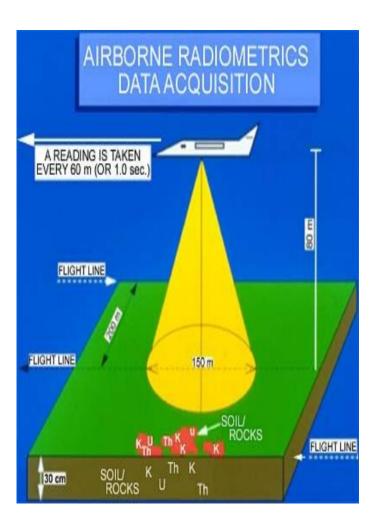
Huang, J., Lark, R.M., Robinson, D.A. Lebron, I., Keith[,] A.M., Rawlins, B., Tye, A., Kuras, O., Raines[,] M., Triantafilis, J.



Geoderma 232-234, 69-80

Case Study

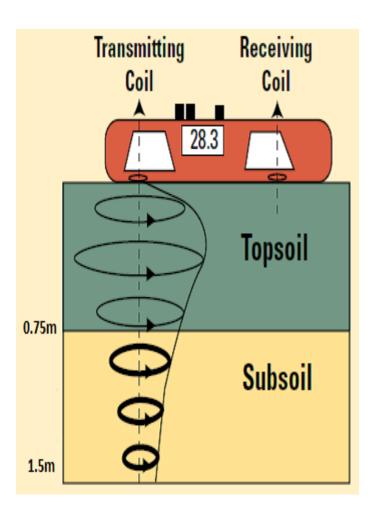
Ancillary data: γ-ray (remote)



- Relationship with underlying;
 - geology, and
 - parent materials (f{ weathering})
- For example soil derived from;
 - Basalt = low signatures,
 - Granite = high signatures
- Airborne
 - helicopter
 - fixed-wing
- Depths up to 0.3-0.4 m



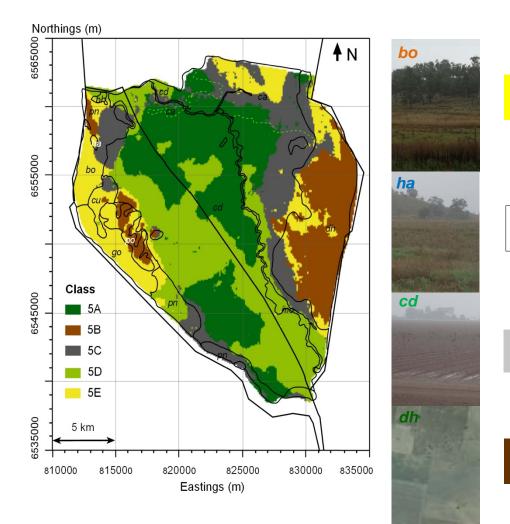
Ancillary data: EM (proximal)



- Mobile can be mounted behind tractors or handheld
- Depths up to
 - 0.75m for horizontal mode
 - 1.5m for vertical mode



Discussion: Identification of soil associatons



bo: resolved to some extent

ha. not identified (too small)

cot actually 2 units maybe 3

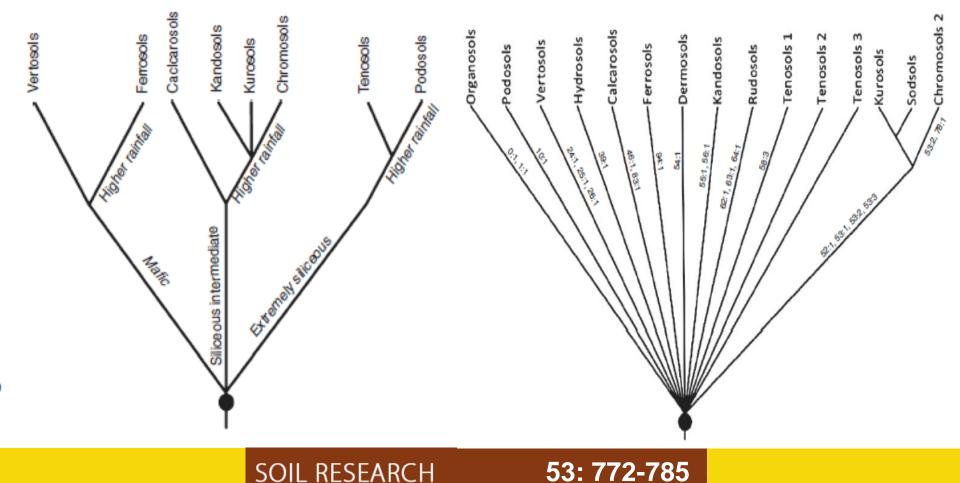
ah resolved to some extent





Assessing the Australian Soil Classification using cladistic analysis

Miltenyi GPL, Malte MC, **Triantafilis J**



SOIL RESEARCH