

# 1   **The Rocklea Dome 3D Mineral Mapping Test Data Set**

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## 10   **ABSTRACT**

11           The integration of surface and subsurface geoscience data is critical for efficient and  
12   effective mineral exploration and mining. Publicly accessible datasets to evaluate the various  
13   geoscience analytical tools and their effectiveness for characterisation of mineral assemblages  
14   and lithologies or discrimination of ore from waste are however scarce. The open access  
15   Rocklea Dome 3D Mineral Mapping Test Data Set ([Laukamp, 2020;](https://doi.org/10.25919/5ed83bf55be6a)  
16   <https://doi.org/10.25919/5ed83bf55be6a>) provides an opportunity for evaluating proximal and  
17   remote sensing data, validated and calibrated by independent geochemical and mineralogical  
18   analyses, for exploration of channel-iron deposits (CID) through cover. We present  
19   hyperspectral airborne, surface and drill core reflectance spectra collected in the visible-near  
20   infrared and shortwave infrared wavelength ranges (VNIR-SWIR; 350 to 2,500 nm), as well  
21   as whole rock geochemistry obtained by means of X-Ray fluorescence analysis and loss on  
22   ignition measurements of drill core samples.

23           The integration of surface with subsurface hyperspectral data collected in the frame of  
24   previously published Rocklea Dome 3D Mineral Mapping case studies demonstrated that about  
25   30% of exploration drill holes were sunk into barren ground and could have been of better use,  
26   located elsewhere, if airborne hyperspectral imagery had been consulted for drill hole planning.  
27   The remote mapping of transported Tertiary detritals (i.e. potential hosts of channel iron ore

28 resources) versus weathered in situ Archaean bedrock (i.e. barren ground) has significant  
29 implications for other areas where “cover” (i.e. regolith and/or sediments covering bedrock  
30 hosting mineral deposits) hinders mineral exploration. Hyperspectral remote sensing represents  
31 a cost-effective method for regolith landform mapping required for planning drilling programs.  
32 In the Rocklea Dome area, vegetation unmixing methods applied to airborne hyperspectral  
33 data, integrated with subsurface data, resulted in seamless mapping of ore zones from the  
34 weathered surface to the base of the CID – a concept that can be applied to other mineral  
35 exploration and mineral deposit studies. Furthermore, the associated, independent calibration  
36 data allowed to quantify iron oxide phases and associated mineralogy from hyperspectral data.  
37 Using the Rocklea Dome data set, novel geostatistical clustering methods were applied to the  
38 drill core data sets for ore body domaining that introduced scientific rigour to a traditionally  
39 subjective procedure, resulting in reproducible objective domains that are critical for the  
40 mining process.

41       Beyond the already published case studies, the Rocklea Dome 3D Mineral Mapping  
42 Test Data Set has the potential to develop new methods for advanced resource characterisation  
43 and develop new applications that aid exploration for mineral deposits through cover. The here  
44 newly presented white mica and chlorite abundance maps derived from airborne hyperspectral  
45 highlight the additional applications of remote sensing for geological mapping and could help  
46 to evaluate newly launched hyper- and multispectral spaceborne systems for geoscience and  
47 mineral exploration.

48

49 **Key words:** Channel Iron ore Deposits, regolith, hyperspectral remote sensing, hyperspectral  
50 drill core sensing, geochemistry

51 **1. INTRODUCTION**

52

53 The three dimensional (3D) geologic case history of the Rocklea Dome located in the  
54 Hamersley Province (Western Australia) targeted the use of reflectance and emission  
55 spectroscopy for measuring mineralogy and geochemistry specific to the exploration and  
56 characterisation of economic Tertiary channel iron ore deposits in a terrain obscured by  
57 weathered, transported materials. This public case history was generated by CSIRO’s Western  
58 Australian Centre of Excellence for 3D Mineral Mapping (C3DMM), which was operated from  
59 2009 to 2012 and had the primary aim of generating and demonstrating the capabilities for  
60 “scalable” 3D mineral mapping from the continental to the prospect scales (Cudahy, 2016).  
61 The Rocklea Dome project was established in collaboration with Murchison Metal Ltd, who  
62 granted C3DMM access to their drill hole dataset, consisting of 14 diamond cores and 180  
63 reverse circulation drill holes. These drill holes were designed using traditional exploration  
64 mapping technologies, such as published geology maps and geophysical data (magnetics and  
65 radiometrics).

66 Key achievements of the Rocklea Dome 3D Mineral Mapping case study include:

- 67 • Based on the kaolin crystallinity index derived from surface and sub-surface  
68 hyperspectral data (Cudahy, 2016), drill holes were identified that were sunk at surface  
69 into barren (i.e. bedrock) weathered material. If surface mineral mapping data, such as  
70 airborne hyperspectral imagery, would have been used during drill hole planning,  
71 approximately one third of the drill holes need not have been drilled or would have been  
72 located differently. This represents potential significant savings in time, money and  
73 environmental disturbance.
- 74 • Characterisation of clay mineralogy associated with distinct domains of the CID and its  
75 cover (i.e. kaolin group vs. Al-smectites vs. Fe-smectites) suggested that clay mineral

76 assemblages as well as calcrete atop buried CIDs have a different composition when  
77 compared to regolith covering adjacent areas. That could represent useful information  
78 when exploring for CIDs through regolith cover.

- 79 • Quantification of iron oxide phases and associated mineralogy derived from  
80 hyperspectral data and validated using X-ray diffractometry and geochemistry (Haest et  
81 al., 2012 a,b):
  - 82 ○ iron (oxyhydr-)oxide content: RMSE of 9.1 weight % Fe
  - 83 ○ Al clay content: RMSE 3.9 weight % Al<sub>2</sub>O<sub>3</sub>
  - 84 ○ hematite/goethite ratio: RMSE 9.0 weight % goethite
  - 85 ○ spatial characterisation of vitreous vs. ochreous goethite
- 86 • Geological modelling the iron ore resource of the Rocklea Dome CID (Haest et al., 2012  
87 a,b; Cudahy, 2016; Fouedjio et al., 2018), which was reported by Dragon Resources in  
88 2012 to be 72.6Mt (at 53% Fe cut-off) with 54.4% Fe, 7.2% SiO<sub>2</sub>, 2.7% Al<sub>2</sub>O<sub>3</sub>, 0.031%  
89 P and 11.2% LOI.
- 90 • Improvement of quality of mineral maps by application of vegetation unmixing methods  
91 (Haest et al, 2013), resulting in seamless mapping of ore zones from the weathered  
92 surface to the base of deposit (Cudahy, 2016).

93 All the above points showcase how hyperspectral data can be used for critical parts of  
94 the mining cycle, especially exploration and 3D resource characterisation.

95 This article aims to provide an overview of the publicly available hyperspectral data set  
96 of the Rocklea Dome, which ought to be used as a test data set for 1) data mining for exploration  
97 and mining, 2) integration of independent geoscience data sets (i.e. hyperspectral,  
98 geochemical), 3) resource modelling, and 4) different approaches for routine processing of  
99 hyperspectral data.

100 The geological setting of the Rocklea Dome area, as well as analytical and processing  
101 methods will be discussed first, after which the publicly available test data are listed as a table.  
102 Example applications of the geochemical and mineralogical data for exploration, 3D mineral  
103 mapping and Resource estimation are summarised briefly in the discussion.

104

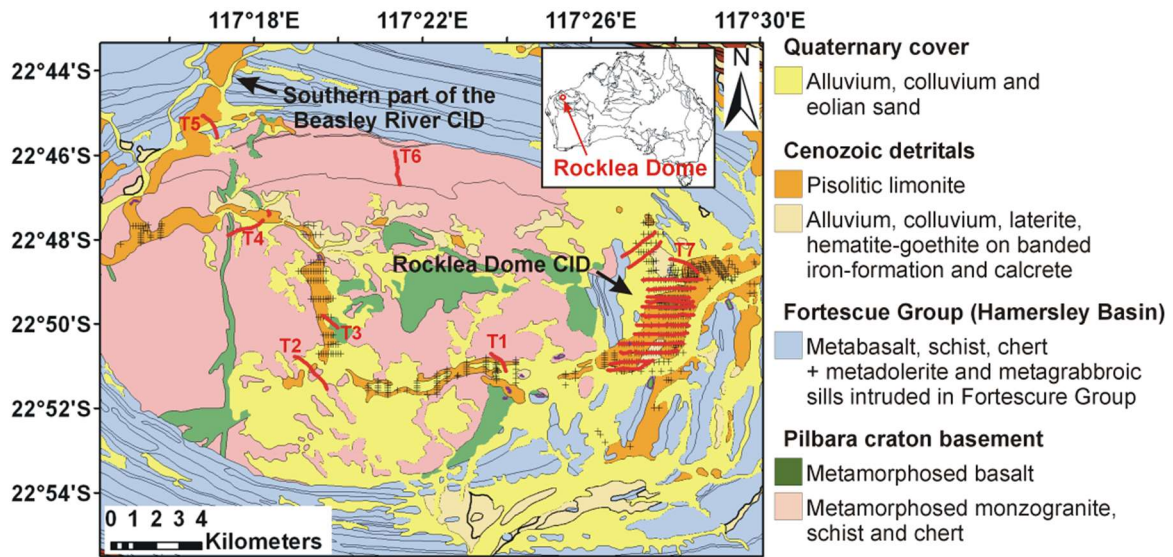
## 105 **2. GEOLOGICAL SETTING**

106

107 The Rocklea Dome Channel Iron Deposit is located in the Hamersley Province, which  
108 is the dominant source of Australia's iron ore exports. Channel Iron Deposits (CID) are  
109 economically significant formations, providing a substantial percentage of the iron ore mined  
110 in Australia. A detailed overview of the geology of the Rocklea Dome and the formation of the  
111 CID was provided by Haest et al. (2012b) and is briefly summarised here. The bedrock geology  
112 of the Rocklea Dome comprises a monzogranite pluton and cross-cutting mafic and ultramafic  
113 intrusives that form part of the Pilbara Craton. The Archean age pluton is overlain by Archean  
114 to Proterozoic metasedimentary and volcanic rocks of the Hamersley Basin, enveloping the  
115 central monzogranite dome (Fig. 1; Thorne & Tyler, 1996). Folding is attributed to  
116 development of both the Ophthalmia and the Ashburton Fold Belt (Thorne & Tyler, 1996).

117 A meandering Tertiary palaeochannel passes over the Archean and Proterozoic rocks,  
118 containing locally Channel Iron ore Deposits (CID), such as the Beasley River CID, which  
119 crosscuts the north-western part of the Rocklea Dome. Channel iron ore was also drilled along  
120 8 km strike length of a palaeochannel on the eastern side of the Rocklea Dome, which was  
121 described by Haest et al. (2012a,b; 2013) as the Rocklea Dome CID (Figure 1). The bedrocks  
122 and Tertiary channel are covered partly by regolith (e.g. Quaternary alluvium). Green  
123 vegetation and dry vegetation (mostly Spinifex grass and bushes) cover the area partly.

124 A mixture of Fe-Ox pelletoids and ferruginised wood fragments below 10 mm in size  
 125 represent the major components of CIDs (Morris & Ramanaidou, 2007). In CID systems, the  
 126 base of the paleochannel often consists of a clay horizon of variable composition. The CID is  
 127 capped in places by calcrete and silcrete.  
 128



129  
 130 Figure 1: Geological map of the core of the Rocklea Dome (Haest et al., 2013; Thorne and  
 131 Tyler, 1996). Validation transects are indicated as red lines. T1 to T7 refer to transects  
 132 described in Haest et al. (2012a). The black crosses identify the position of all reverse  
 133 circulation drill cores intersecting the palaeochannel (i.e. pisolitic limonite in map) in the core  
 134 of the Rocklea Dome.

135

### 136 3. METHODS AND MATERIALS

137

#### 138 3.1. VNIR-SWIR drill core spectroscopy

139 Reflectance spectra of 180 rock chips (RC) and 14 diamond drill cores (RKD) were  
140 measured using CSIRO's HyChips™ system (cf. Huntington et al., 2004), which comprises a  
141 TerraSpec™-based spectrometer (Malvern-Panalytical) system. In total, 7,520 reflectance  
142 spectra were collected from RC samples and 66,853 reflectance spectra were collected from  
143 RKD samples (Haest et al., 2012a). An automated X-Y table moves the drill core tray in a  
144 snake-like pattern below the TerraSpec optical fibre at a distance of ~6 to 13 cm (depending  
145 on sample type, i.e., diamond core or drill chips), while the spectral data are collected. Each  
146 sample spectrum is collected from a 1 × 1 cm area. Four light globes are positioned 40 cm  
147 above and at a small angle (off the backscatter/specular angle) to the measurement/sample  
148 point. In addition to hyperspectral data, high spatial resolution (0.1 mm pixel) images are  
149 collected from the core or chip tray and the sample height in the tray is measured using a laser  
150 profilometer. Reflectance spectra were calibrated using a Teflon/Spectralon™ panel (see Haest  
151 et al., 2012a for more details). TerraSpec™ spectra were collected in the visible to near-  
152 infrared (VNIR: 380–1,000 nm) and shortwave-infrared (SWIR: 1,000–2,500 nm), with  
153 sampling intervals of 1.4 nm in the VNIR and 2.0 nm in the SWIR and a wavelength accuracy  
154 of ±1 nm. The spectral resolution is 5 nm in the VNIR and between 11 and 12 nm in the SWIR.  
155 The TerraSpec™ radiance spectra of each sample are first converted to apparent bidirectional  
156 reflectance using the Teflon signal, which is collected at the beginning/end of each drill  
157 core/drill chip tray measurement cycle. This signal is then converted to absolute reflectance,  
158 based on the measurement of a Spectralon panel.

159

#### 160 3.2. Remote Sensing

161 Airborne VNIR–SWIR imagery was collected using the Airborne Multispectral  
162 Scanner (AMS), which is an earlier version of HyVista Corporation’s HyMap™ system (Cocks  
163 et al., 1998). The AMS system collects 96 bands over the VNIR-SWIR, excluding the  
164 atmospheric bands from ~1,000 nm to ~1,400 nm and from ~1,800 to ~1,950 nm, respectively.  
165 For each spectral band, the average spacing of collected bands is 15 nm and the average full  
166 width at half maximum is 17 nm. The AMS data over Rocklea Dome were collected in a north–  
167 south direction between the 31st of July 2000 and the 2nd of August 2000, comprising a set of  
168 14 flight lines, totalling in a combined length of ~280 km at a pixel size of approximately 7 m.  
169 Atmospheric correction was done using MODTRAN5 (Berk et al., 2004, 2006) and SODA  
170 (Rodger, 2011), based on a combination of the AMS at-sensor radiance with in-scene flight  
171 parameters (e.g. latitude, longitude, sensor height, etc.). For more details about georeferencing  
172 and mosaicking the single flight lines, see Haest et al. (2013).

173

### 174 3.3. *X-ray fluorescence analysis (XRF)*

175 XRF analysis of 11,900 RC samples (1m interval) for weight percentages of Fe, P, S,  
176 SiO<sub>2</sub>, Al<sub>2</sub>O<sub>3</sub>, Mn, CaO, K<sub>2</sub>O, MgO, and TiO<sub>2</sub> was conducted by Kalassay Ltd. (now Bureau  
177 Veritas Minerals Pty Ltd., Western Australia). A Bruker Pioneer X-ray fluorescence instrument  
178 with an end window 4 kW rhodium X-ray tube was used. Sample preparation included drying  
179 at 105°C for 12 hr or for 1 hr, depending on whether the sample was wet or dry, respectively.  
180 Samples were then crushed to a nominal 90% passing 75 µm. The sample powders were fused  
181 in a Herzog automated (RF energised) fusion furnace and cast into 40 mm diameter beads using  
182 a 12:22 flux containing 5% sodium nitrate. Matrix corrections were applied using a calculated  
183 alpha correction for this combination of flux, tube, and instrument geometry. Previously  
184 determined weight ranges were used for both the sample and the flux weight. Kalassay Ltd.  
185 used lab duplicates, internationally certified reference materials, and reference materials of the



186 same ore type as standards and reported a precision better than 0.01% for all analyses. In order  
187 to evaluate the accuracy of XRF analyses undertaken by Kalassay Ltd., duplicate samples were  
188 also sent to the Amdel laboratory in Cannington (Western Australia), with good correlation  
189 observed (Haest et al., 2012a).

190

### 191 3.4. *Loss on ignition (LOI)*

192 In order to characterise the mineral assemblages present in the samples in more detail,  
193 loss on ignition (i.e. LOI) measurements were undertaken on 11,900 RC samples to record the  
194 mass loss of samples on heating (Haest et al., 2012a). A pre-dried portion of all samples was  
195 heated in an electric furnace to 1,000°C. During this process, goethite releases its strongly  
196 bonded water and its OH groups between 260°C and 425°C (Strezov et al., 2010), organic  
197 matter completely ignites by 550°C (Dean, 1974), aluminosilicate clay materials decompose  
198 between 530°C and 605°C (Strezov et al., 2010), and inorganic carbon is oxidised and lost as  
199 CO<sub>2</sub> between 700°C and 850°C (Dean, 1974).

200

### 201 3.5. *Sample storage*

202 Drill core trays, field samples and XRF standards are all stored at the Australian  
203 Resources Research Centre (ARRC) in Kensington (Western Australia). Samples can be  
204 viewed and investigated at the ARRC, using local analytical facilities.

205

## 206 4. SOFTWARE AND PROCESSING METHODS

### 207 4.1. *Processing of hyperspectral drill core data*

208 Hyperspectral drill core data were analysed using the CSIRO's The Spectral Geologist  
209 software (TSG<sup>TM</sup>) by interpreting the abundance, composition and/or crystallinity of selected

210 mineral groups and species using the Multiple Feature Extraction Method. A list of scripts  
211 applied to the hyperspectral drill core and rock chip data can be found in Table 1.

212 Table 1: Base scripts and multiple feature extraction method scripts used for the Rocklea Dome  
 213 3D Mineral Mapping project (R = reflectance value at given wavelength)

214

Product name	Minerals detected	Base algorithm	Filters/Masks	Lower stretch limit	Upper stretch limit (only applicable for composition products)	related publication
Ferric oxide abundance (ferric_oxide_abundance.txt)	Hematite, goethite, jarosite, "limonite"	Continuum removed depth of the 900 nm absorption calculated using a fitted 2 <sup>nd</sup> order polynomial between 776 and 1,050 nm <i>900D</i>	R450 > R1650	0.04; low content	further developed on the basis of Haest et al. (2012a,b), which used a 4th order polynomial or 4 band ratio approach	
Hematite-goethite distribution (Hematite-goethite_dist.txt)	Hematite-goethite ratio (Cudahy and Ramanaidou, 1997)	Continuum removed wavelength of the 900 nm absorption calculated using a fitted 2nd order polynomial between 776 and 1,050 nm. <i>900W</i> (R920+R1650)/(R1020+R235) <i>Ferratus</i>	R450 > R1650 + 900D > 0.025	~890 nm: more hematitic ~1,005: low content	further developed on the basis of Haest et al. (2012a,b), which used a 4th order polynomial or 4 band ratio approach Laukamp et al. (2012)	
Ferrous iron abundance (Ferrous iron abundance.txt)	Fe <sup>2+</sup> in silicates & carbonates. (Fe-chlorites, Fe-amphibole, Fe-pyroxene, Fe-olivine, Fe-carbonate)	albedo @ 1,650 nm <i>1650</i>	OPAQUES_450D1650 > 0.25; albedo @ 1650 nm 1650 < 0.9%	2; low content		
opaques2 (opaques2inv.txt)	"Reduced" materials such as carbon black, sulphides and maggetite as well as Mn oxides.					
White mica and Al-smectite abundance (wmaAlsmci.txt)	Abundance of white micas (e.g. illite, muscovite, paragonite, brammalite, phengite, lepidolite, margarite) and Al-smectites (montmorillonite, beidellite)	Relative absorption depth of the 2,200 nm absorption for which the continuum is removed between 2,120 and 2,245 nm, determined using a 3 band polynomial fit around the band with the lowest reflectance. <i>2200D</i>	(R2138+R2190)/(R2156 +R2179) 2160D2190 < 1.063	0.02; low content	further developed on the basis of Sonntag et al. (2012), which used a 4th order polynomial or 4 band ratio approach	
White mica and Al-smectite composition (wmaAlsmci.txt)	Tschermak substitution of white micas, ranging from paragonite, brammalite, to illite, muscovite to phengite, and Al-smectites, ranging from beidellite to montmorillonite.	Minimum wavelength of the 2,200 nm absorption for which the continuum is removed between 2,120 and 2,245 nm, determined using a 3 band polynomial fit around the band with the lowest reflectance. <i>2200W</i>	(R2138+R2190)/(R2156 +R2179) 2160D2190 < 1.063	2,180 nm: Al-rich mica (muscovite, illite, (-)phengite) paragonite, brammalite, lepidolite	further developed on the basis of Sonntag et al. (2012), which used a 4th order polynomial or 4 band ratio approach	
Kaolin abundance index	Kaolin group minerals, namely kaolinite halloysite, dickite and nacrite	2200D (Normalized depth of a fitted 4th order polynomial between 2,120 and 2,245 nm)	2160D ((R2138+R2190)/(R2156+R2179))>1.005	0.02; low content	Sonntag et al. (2012), Haest et al. (2012a,b)	
Kaolin composition index	Composition and crystallinity of kaolin group minerals ranging from well-ordered kaolinite to halloysite to dickite (and nacrite)	[(R2138+R2173)/R2156]/[(R2156+R2190)/R2173]	2200D > 0.005	low values = low crystallinity	Sonntag et al. (2012), Haest et al. (2012a,b)	
Carbonates abundance (carb3pfit.txt)	Carbonates vs. MgOH-bearing silicates, based on left-asymmetry of CO3 feature @ 2,340 nm	Relative absorption depth of the 2,340 nm absorption for which the continuum is removed between 2,270 and 2,370, determined using a 3 band polynomial fit around the band with the lowest reflectance. <i>2340D</i>	2340D > 0.04, 2295nm<2340W<2360nm, 2250D < 0.025, 2380D<0.117*2340D+0.0002, Asymmetry of the 2340 absorption using a fitted 4th order polynomial between 2120 and 2370: 2340 left asym > 1.13	high values = high crystallinity	further developed on the basis of Sonntag et al. (2012), which used a 4th order polynomial or 4 band ratio approach	
Carbonate composition (carb3pfit.txt)	separating calcite, dolomite, siderite, ...	Minimum wavelength of the 2,340 nm absorption for which the continuum is removed between 2,270 and 2,370 nm, determined using a 3 band polynomial fit around the band with the lowest reflectance. <i>2340W</i>	2340D > 0.04, 2295nm<2340W<2360nm, 2250D < 0.025, 2380D<0.117*2340D+0.0002, Asymmetry of the 2340 absorption using a fitted 4th order polynomial between 2120 and 2370: 2340 left asym > 1.13	2,343 nm: calcite	further developed on the basis of Sonntag et al. (2012), which used a 4th order polynomial or 4 band ratio approach	
White mica (+Al-smectite) abundance, refined for airborne hyperspectral imagery	Abundance of white micas (e.g. illite, muscovite, paragonite, brammalite, phengite, lepidolite, margarite) and Al-smectites (montmorillonite, beidellite)	Relative absorption depth of the 2,200 nm absorption for which the continuum is removed between 2,120 and 2,245 nm, determined using a 3 band polynomial fit around the band with the lowest reflectance. <i>2200D</i>	(R2138+R2190)/(R2156 +R2179) 2160D2190 < 1, 2200D/2320D > 1.5	0.04; low content	further developed on the basis of Sonntag et al. (2012), which used a 4th order polynomial or 4 band ratio approach	
Chlorite (+epidote, +biotite) abundance, refined for airborne hyperspectral imagery	Abundance of chlorite (e.g. clinocllore, chamosite), as well as members of the epidote and biotite mineral groups	(R2227+R2275)/(R2241+R2259). <i>2250D</i>	2250D > 1.01, & 2300 < 2320W < 2342 & 2240 < 2250W < 2260	1.04; high content	further developed on the basis of Sonntag et al. (2012)	

215

#### 216 4.2. *Image Processing*

217 The processing strategy for generating geoscience products from AMS data, such as the  
218 Kaolin Crystallinity (Table 1) builds on the quality control of the acquired data (Cudahy et al.,  
219 2008). Well calibrated radiance-at-sensor or surface reflectance data are required for the  
220 processing of airborne hyperspectral imagery. Commonly applied levelling and statistics-based  
221 methods were avoided as these introduce undesirable scene-dependencies, making a  
222 comparison of image products from different areas impossible. Physics-based reduction  
223 models were applied to the remote sensing data, using the image processing software ENVI™.  
224 Complicating effects were removed in their order of development (i.e. 1. instrument, 2.  
225 atmospheric, 3. surface effects) through either normalization or offsets.

226

#### 227 4.3. *The Multiple Feature Extraction Method*

228 In hyperspectral proximal (e.g. HyLogging™) and remote (e.g. AMS) sensing  
229 technologies, the VNIR, SWIR, and thermal infrared (TIR: ca. 6,000 – 14,500 nm) wavelength  
230 ranges are used to infer abundance and composition of various rocks and minerals in a wide  
231 range of sample types, including drill core, rock chips and pulps. The relative intensity and  
232 wavelength position of absorption features in the reflectance spectra relate to the  
233 physicochemical characteristics of the various minerals. Feature extraction methods can be  
234 used to determine the mineralogy of a sample material (Cudahy et al., 2008). The advantage of  
235 the multiple feature extraction method (MFEM) is that the associated scripts are not based on  
236 a training dataset or spectral reference libraries, but are based only on the visible and/or infrared  
237 active functional groups of minerals (see Laukamp et al., 2011, for more details). As they are  
238 instrument independent, the same scripts can be applied to remote sensing and proximal  
239 hyperspectral data, easing the integration of, for example, surface (e.g. HyMap) and subsurface

240 data (e.g. HyLogging™) for the purpose of visualisation in 3D or advanced data analytics.  
241 Interferences of mineralogical information with other surface materials such as vegetation can  
242 be evaluated by using a multiple linear regression model for unmixing vegetation from  
243 hyperspectral remote sensing data (Rodger & Cudahy, 2009; Haest et al., 2013). Other  
244 complications, such as spectrally overlapping materials, are removed by the application of  
245 thresholds.

246

## 247 5. DATA PRODUCTS AND APPLICATIONS

248 Publicly accessible data of the Rocklea Dome 3D Mineral Mapping project can be  
249 found on CSIRO's Data Access Portal:  
250 <https://data.csiro.au/collections/#collection/Cicsiro:44783>  
251 (<https://doi.org/10.25919/5ed83bf55be6a>) and are listed in Table 2. Data and other content on  
252 this site are scientific research data collected by CSIRO and third parties and are made available  
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263 information or material contained in it.

264

265 Table 2: Publicly accessible data of the Rocklea Dome 3D Mineral Mapping project

266 (<https://data.csiro.au/collections/#collection/CIcsiro:44783>;

267 <https://doi.org/10.25919/5ed83bf55be6a>)

main directory	sub directory	file name	type of data	source/IP	
DTM		<u>DTM_Rocklea_50k.00t</u>	Digital terrain model	GSWA	
		DTM_Rocklea_50k.dxf			
		DTM_Rocklea_50k.evf			
		DTM_Rocklea_50k.zip			
			Hardey_HR_DTM.00t	Digital terrain model of 100K mapsheet Hardey 2252	GSWA
			Hardey_HR_DTM.dxf		
			Hardey_HR_DTM.evf		
			Topography_ENVI	Digital elevation model	GSWA
			Topography_ENVI.hdr		
			dem_plus_collars.csv	Digital elevation model	GSWA
drill hole data	RC_hyperspectral_geochem	GeoscienceProductDescriptions_ProximalHyperspectral.xlsx	table describing multiple feature extraction scripts applied to hyperspectral data for interpretation of mineralogy	CSIRO	
		RC_data.tsg	TSG-file	CSIRO	
		RC_data.ini	TSG-file	CSIRO	
		RC_data.bip	TSG-file	CSIRO	
		RC_data_cras.bip	TSG-file	CSIRO	
			RC_data_tsgexport.CSV	spectral and geochemical data exported from TSG	CSIRO
	RKD		RKD5-7-9.tsg	TSG-file	CSIRO
			RKD5-7-9.ini	TSG-file	CSIRO
			RKD5-7-9.bip	TSG-file	CSIRO
			RKD5-7-9_cras.bip	TSG-file	CSIRO
remote sensing data	GeoTIFF_AMS/	<u>2200D_Mstd.tfw</u>	AMS product "2200D", showing the relative abundance of Al-clays	CSIRO	
		<u>2200D_Mstd.tif</u>		CSIRO	
		<u>2200WAR_2190-2205.tfw</u>	AMS product "2200W", indicating compositional changes of Al-smectites and white micas (Al <sup>VI</sup> Al <sup>IV</sup> (Fe,Mg) <sub>1</sub> Si <sub>1</sub> )	CSIRO	
		<u>2200WAR_2190-2205.tif</u>		CSIRO	
		<u>2250_MStd.tfw</u>	AMS product "2250D", showing the relative abundance of chlorite, epidote and/or biotite	CSIRO	
		<u>2250_MStd.tif</u>		CSIRO	
		<u>2330_2250-2380.tfw</u>	AMS product "Carbonate abundance", showing the relative abundance of carbonates	CSIRO	
		<u>2330_2250-2380.tif</u>		CSIRO	
			<u>KC_NoSM_22D+216DM_3MeFi.tfw</u>	AMS product "Kaolin crystallinity"	CSIRO
			<u>KC_NoSM_22D+216DM_3MeFi.tif</u>		CSIRO
	TXT_AMS/		<u>2320D_vegunm.txt</u>	AMS product "2320D", vegetation unmixed	CSIRO
			<u>AlOHAbVegunm.txt</u>	AMS product "Al-clay abundance index", vegetation unmixed	CSIRO
			<u>FeOxVegUnm.txt</u>	AMS product "Ferric Oxide Abundance Index", vegetation unmixed	CSIRO
			<u>SRTM_RockleaDome+HardeyRiver.txt</u>	Digital elevation model	GSWA

Rocklea Dome exercise	StudentExercises_Rocklea.docx	Exercises for analysis of HyLogging data	CSIRO
	Answers_CIDexercises.docx	Suggested answers to exercises for analysis of HyLogging data	CSIRO
	MinSpec_Workshop_7RockleaDomeTSG_HandsOn.pptx	PPT-presentation summarising Rocklea Dome exercise and results	CSIRO

268

269           The following chapters briefly describe examples of how the provided hyperspectral  
270 and geochemical proximal and remote sensing data sets can be used to address challenges for  
271 the mineral resources sector.

272

### 273 *5.1. Drill core mineralogy and geochemistry*

274           Reflectance spectra collected from rock chips (RC) and diamond drill cores (RKD)  
275 using CSIRO’s HyChips™ system presented a cost-effective way to spatially map the major  
276 ore (i.e. goethite +/- hematite) and gangue minerals (i.e. kaolinite, smectite, carbonate), apart  
277 from quartz, in detail. To achieve this, the relative intensity of mineral-diagnostic absorption  
278 features was calculated using a suite of batch scripts (“Geoscience Products” in Haest et al.,  
279 2012a). The relative intensity of the respective absorption features correlates with the relative  
280 abundance of the respective mineral, whereas the wavelength position of key absorption  
281 features relates to mineral speciation (e.g. ochreous versus vitreous goethite) or determining  
282 the mineral chemistry. For example, the relative abundance of iron oxides was calculated from  
283 the relative depth of the ferric iron-related absorption at around 900 nm (Cudahy &  
284 Ramanaidou, 1997), whereas goethite was distinguished from hematite by tracking the  
285 wavelength position of the same absorption feature (Table 2 in Haest et al., 2012a).

286           Whole rock geochemistry obtained from the same drill core material showed significant  
287 correlations with the Geoscience Products. Haest et al. (2012) determined an RMSE of 9.1  
288 weight % Fe for the correlation between the hyperspectrally-derived iron oxide abundance and  
289 the XRF weight % Fe data and an RMSE of 3.9 weight % Al<sub>2</sub>O<sub>3</sub> for correlation between the

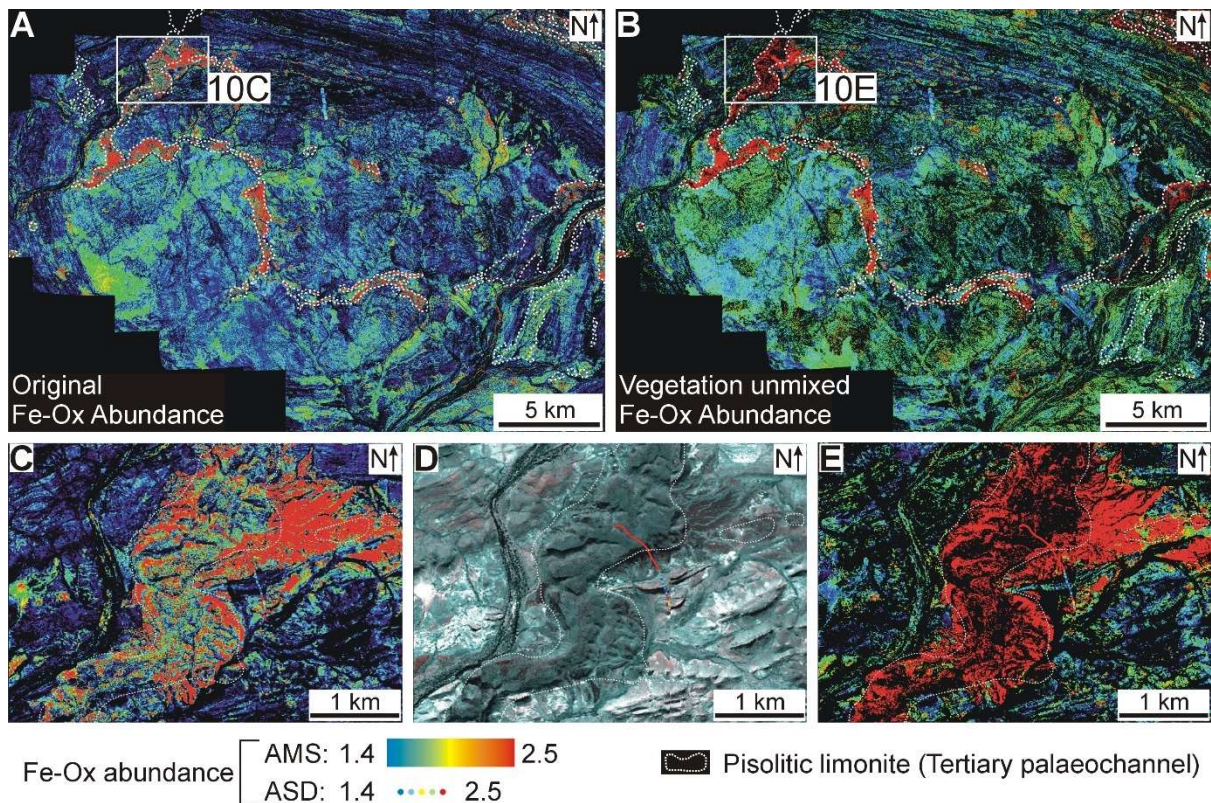
290 hyperspectrally-derived Al-clay abundance and the XRF weight % Al<sub>2</sub>O<sub>3</sub> data. The errors  
291 associated with the correlations were found to be due to a combination of grain size variations  
292 and the transopaue behaviour of iron oxides and/or different amounts of silica, causing  
293 variations in the optical depth of sample material.

294

## 295 5.2. *Surface mineral mapping*

296 Airborne hyperspectral surveys provide spatially contiguous mineralogical information  
297 of the Earth's surface at high spatial resolution (down to circa 1 m). The relative intensity of  
298 mineral diagnostic absorption features and their wavelength positions can be used to infer the  
299 relative abundance of the respective minerals and even variations of single mineral species in  
300 terms of their cation composition, crystallinity and hydroxylation. The Rocklea Dome case  
301 study data set was used by Haest et al. (2013) to demonstrate how quantitative mineral maps  
302 can be produced by validation of airborne hyperspectral data against field data, including  
303 reflectance spectra and XRF data collected from surface samples. The effect of both green and  
304 dry vegetation cover was unmixed at the pixel-level using the Normalised Difference  
305 Vegetation Index (NDVI; e.g. Tucker, 1979) and the continuum-removed depth of the  
306 cellulose-lignin absorption centred at around 2,100 nm, respectively. The resulting mineral  
307 mapping products have a higher spatial continuity, as well as higher accuracy of, for example,  
308 mineral abundance or composition values shown in single pixels. This proved to be especially  
309 useful in areas with outcropping CID, which appeared to be sub-economic from the original  
310 iron oxide abundance mineral maps but showed as potentially economic CID resources when  
311 the vegetation cover was unmixed (Figure 2).





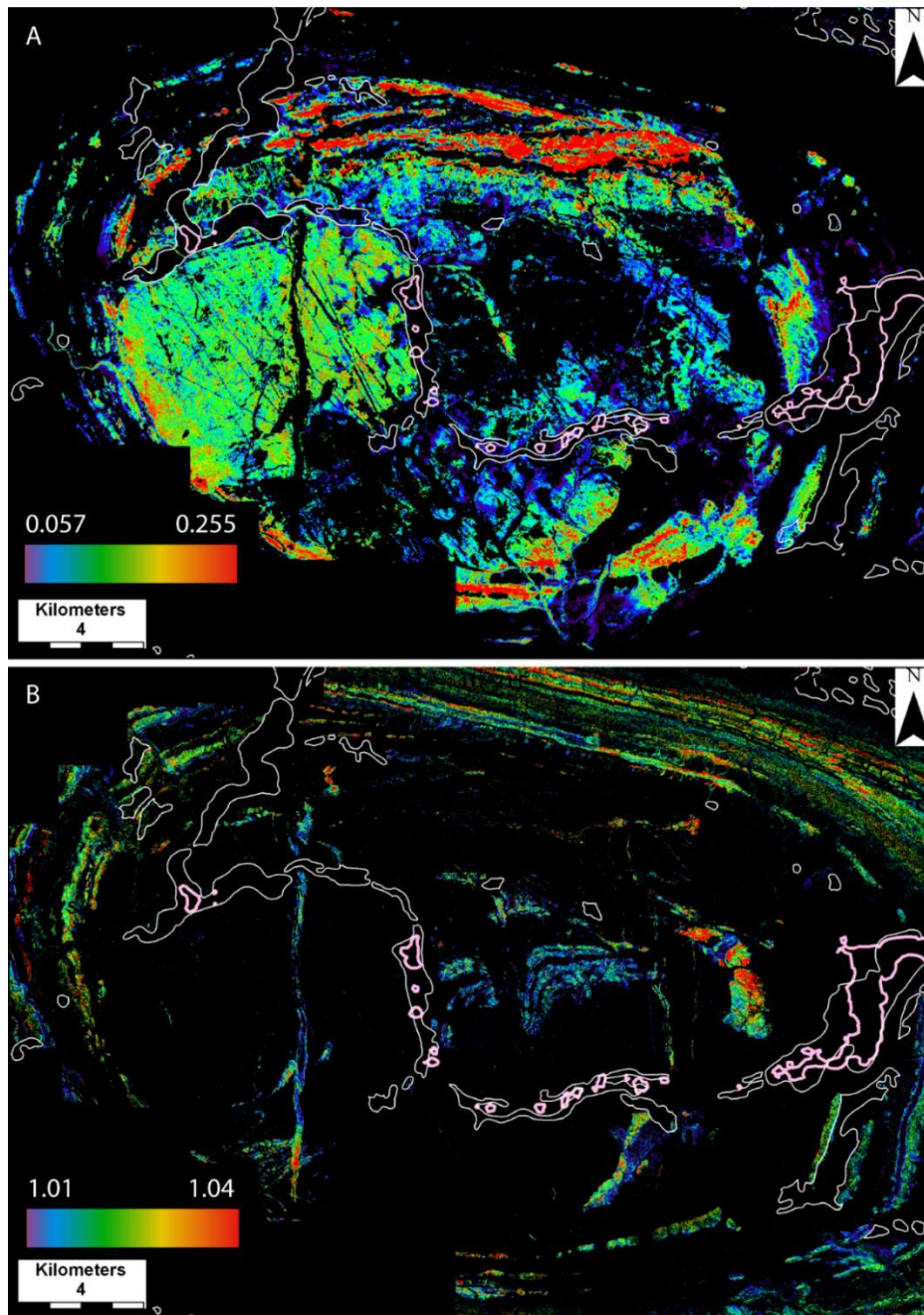
312

313 Figure 2: A–B: Fe-(oxyhydr-)oxide (Fe-Ox) abundance maps of the Rocklea Dome without  
 314 (A) and with (B) vegetation unmixing. C–E: Fe-Ox abundance maps of the southern part of  
 315 the Beasley River CID with (C) and without (E) vegetation removal and the false colour  
 316 image of this area (D). The Beasley River CID has a plateau like surface expression, with the  
 317 edges of the plateau clearly visible in the false colour image. These edges were mapped by  
 318 the Geological Survey of Western Australia as representing the boundary of the pisolitic  
 319 limonite (Fe-rich palaeochannel; white stippled line) (the Fe-Ox abundance measurements  
 320 collected along transects 1 to 7 with the TerraSpec<sup>TM</sup> are also shown for reference).

321

322 Beyond the iron oxide, kaolin and carbonate mineral maps published by Haest et al.  
 323 (2013), airborne hyperspectral data can be used to create numerous additional mineral mapping  
 324 products that can be used to address other geoscientific questions. For example, the Rocklea  
 325 Dome presents a wide variety of igneous units that are part of the Proterozoic basement of the  
 326 Pilbara Craton (Figure 1). These include 1) metamorphosed monzogranite, schist and chert, 2)  
 327 metamorphosed basalt, and 3) amphibolite dykes. According to the white mica abundance  
 328 derived from airborne hyperspectral data (green in Figure 3a), the metamorphosed  
 329 monzogranite contains less white mica, when compared to the metamorphosed schists which

330 are striking East-West and occur in the northern part of the Rocklea Dome (red in Figure 3a).  
331 In the eastern half of the Proterozoic basement in the Rocklea Dome, white mica is much less  
332 abundant to absent. This coincides with elevated amounts of chlorite (folded lithologies in the  
333 centre of Figure 3b), which map out metamorphosed basalt (Figure 1). The North-South  
334 striking occurrence of chlorite in the Western half of the Proterozoic basement traces an  
335 amphibolite dyke. Both the white mica abundance and chlorite abundance maps can also be  
336 used to map out different lithologies in the metasediments and metabasalts of the Fortescue  
337 that crop out to the North and South of investigated area, demonstrating how the airborne  
338 hyperspectral data can be used to map out all major lithologies occurring in the Rocklea Dome  
339 case study area.



340

341 Figure 3: A & B: White mica (+Al-smectite) and chlorite (+epidote, +biotite) abundance  
 342 maps of the Rocklea Dome area in A and B, respectively, calculated from airborne  
 343 hyperspectral data using algorithms described in Table 1. Warm colours represent high  
 344 abundance and cool colours low abundance of the respective minerals. Black pixels have  
 345 been masked out as relative intensity of the absorption feature mapped in the respective  
 346 mineral map is below a given threshold (Table 1) and/or because of non-mineralogical effects  
 347 (e.g. vegetation, clouds). A shows monzogranites in the western part of the dome in green  
 348 colours and the Fortescue Group in the northern fringe of the dome in red colours. B  
 349 highlights Archean metamorphosed basalts in the eastern part of the dome structure and an N-  
 350 S trending amphibolite dyke in the western part of the dome. White lines indicate the surface  
 351 extension of the Tertiary paleochannel as mapped by Thorne & Tyler (1996). Pink lines

352 indicate the horizontal extension of the Tertiary paleochannel as mapped by the hyperspectral  
353 data.

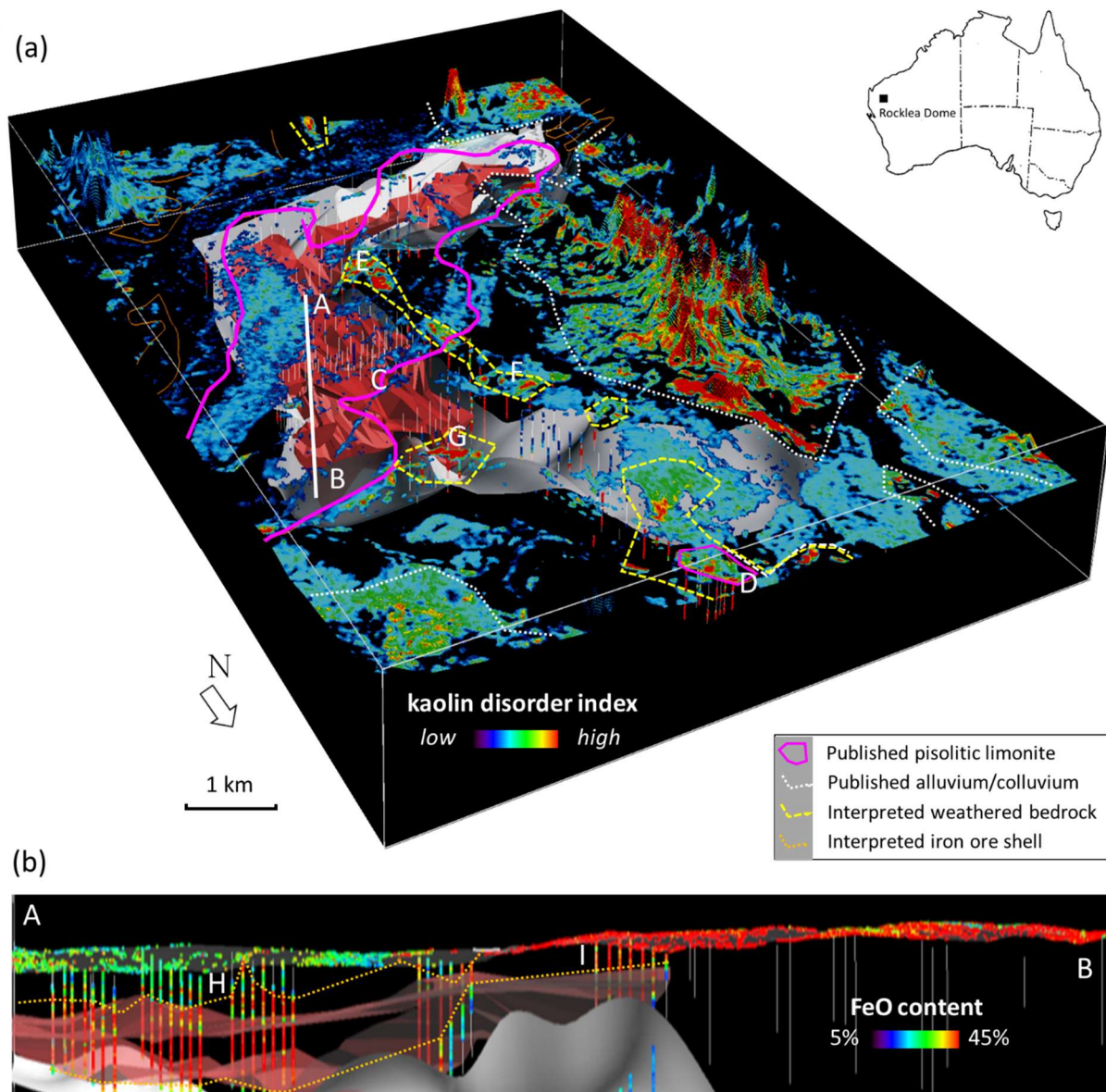
354

### 355 5.3. *3D Mineral Mapping*

356 The hyperspectral drill core data can be combined with airborne hyperspectral data into  
357 a seamless 3D mineral model of the Rocklea Dome CID (Figure 4). For this, all hyperspectral  
358 data were resampled to the same spatial resolution and imported into the 3D modelling software  
359 GoCad/SKUA™. The channel basement contact that was delineated at depth using the kaolin  
360 crystallinity products could also be delineated at the surface from the airborne hyperspectral  
361 image. A combination of both provided a seamless surface of the channel bottom (grey surface  
362 in Figure 4) that separates the basement characterised by well-crystalline kaolinite from the  
363 tertiary channel sediments characterised by poorly-crystalline kaolinite. The here identified  
364 channel basement contact deviates at the surface significantly from the area mapped by the  
365 geological survey as palaeochannel. This suggests that drilling patterns could have been much  
366 better defined if the airborne hyperspectral-based surface outline would have been available  
367 prior to drilling (Cudahy, 2016).

368 As part of their 3D Geomodel Series, the Geological Survey of Western Australia  
369 (GSWA) provides access to 3D models of the Rocklea Dome area via their online portal:  
370 [https://dmpbookshop.eruditetechnologies.com.au/product/rocklea-inlier-2016-3d-geomodel-](https://dmpbookshop.eruditetechnologies.com.au/product/rocklea-inlier-2016-3d-geomodel-series.do)  
371 [series.do](https://dmpbookshop.eruditetechnologies.com.au/product/rocklea-inlier-2016-3d-geomodel-series.do). The data can be viewed in three different formats (3D PDF, Geoscience Analyst,  
372 GOCAD).

373



374

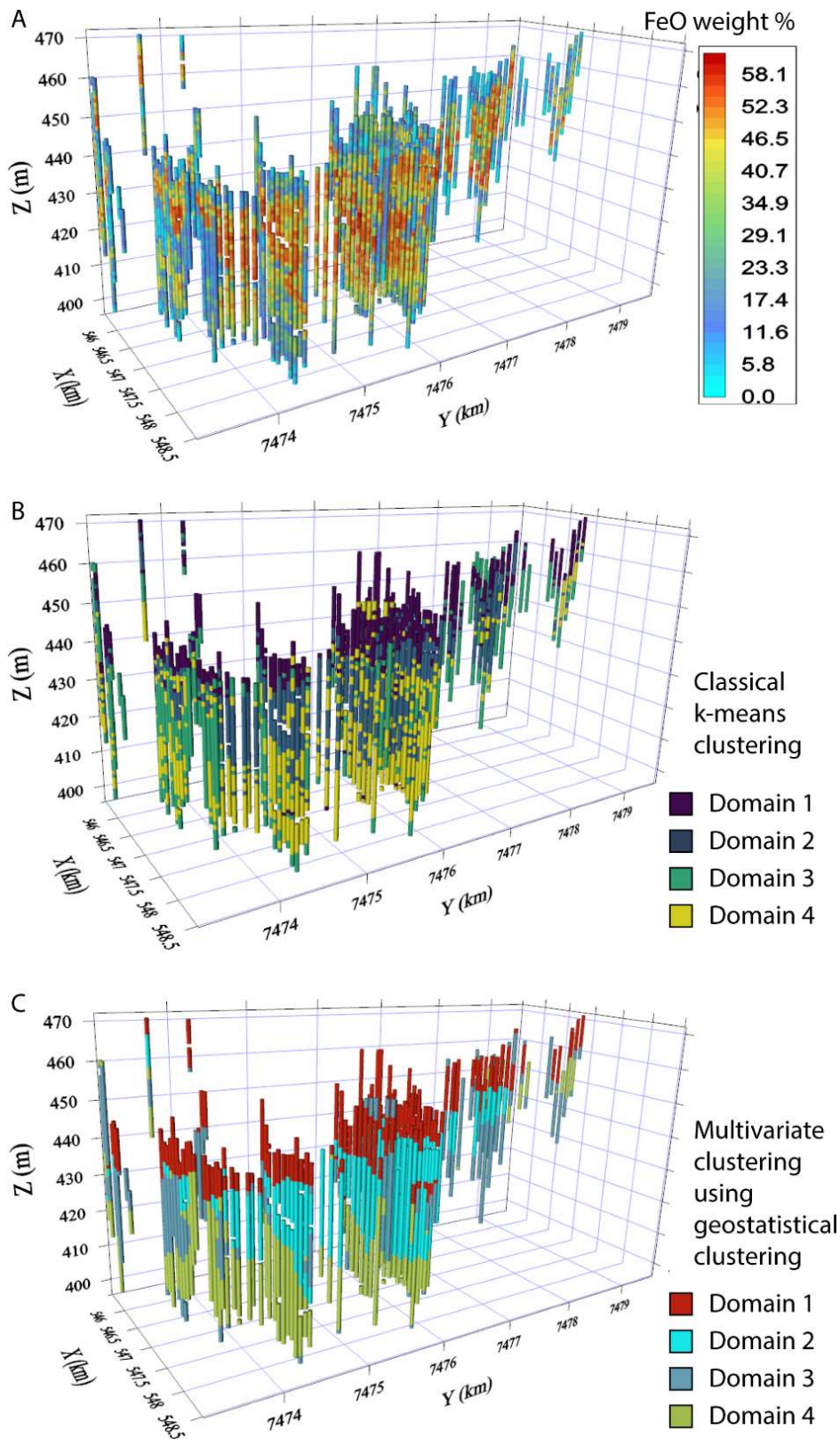
375 Figure 4: 3D mineral models of the Rocklea Dome area (Cudahy, 2016). Scene centre is  
 376 approximately 22.8216° latitude 117.4652° longitude. (a) A southwest oblique 3D view of  
 377 the Rocklea Dome study area showing kaolin disorder measured using airborne HyMap™  
 378 (surface) and drill core HyLogger™ (coloured vertical pegs) reflectance spectra. Warmer  
 379 colours (well-ordered kaolin) relate to weathered, in situ bedrock, while cooler colours  
 380 (poorly-ordered kaolin) relate to transported (alluvium/colluvium) materials. The interpolated  
 381 model of the base of the channel iron system calculated using the 3D kaolin crystallinity map  
 382 is shown by the shaded grey surface. The CID, which was calculated from the XRF-derived  
 383 % FeO (Haest et al., 2012a), is shown by a shaded red volume (C). Areas of weathered  
 384 bedrock (Haest et al., 2012b; Cudahy, 2016) are highlighted by yellow-coloured hashed lines  
 385 and highlight which drill cores were sunk into barren ground (D, E, F, G). A white straight  
 386 line shows the location of the cross-section (A–B) presented in (b); (b) Cross section A–B in  
 387 (a) of the % FeO measured from the drill core and airborne imagery, which was vegetation  
 388 unmixed (Haest et al., 2013). Orange, dotted polygon indicates the shell of iron ore, which  
 389 extends from under cover of ~20 m of alluvium (H) to exposed at the surface (I).

390

391 5.4. *Resource estimation*

392 Resource estimation of base and precious metal deposits requires the grouping of drill  
393 hole data into domains that represent zones of homogenous properties for accurate grade  
394 estimation and practical exploitation purposes. In practice, this is more than often performed  
395 through a subjective time-consuming manual interpretation of sample analytical data.  
396 Traditional automated clustering techniques, such as multivariate clustering and k-means, tend  
397 to show poor spatial contiguity of domains in a mineral deposit. Fouedjio et al. (2017) used the  
398 Rocklea Dome drill core data set to showcase how geostatistical clustering methods can take  
399 spatial dependency into account (Figure 5). By integrating whole rock geochemistry and  
400 hyperspectral drill core data, Fouedjio et al. (2017) revealed two distinct domains in the  
401 Rocklea Dome Channel Iron Ore Deposit that are mainly characterised by four geochemical  
402 variables (FeO, Al<sub>2</sub>O<sub>3</sub>, SiO<sub>2</sub> and TiO<sub>2</sub>) and two mineralogical variables derived from  
403 hyperspectral data (ferric oxide abundance and kaolinite abundance). Ore body domaining  
404 through geostatistical clustering represents a method for objective samples clustering that  
405 introduces scientific rigour to a traditionally subjective procedure. The robust domaining is  
406 based in genuine multivariate geostatistics combining all available data. The flexible and  
407 reproducible automatic domaining technique saves time, improves the understanding of  
408 domains critical for exploitation of the ore and allows an easy integration of new data sets.

409



410

411 Figure 5: a) Spatial plot of FeO distribution in the Rocklea Dome Channel Iron Deposit. (b)  
 412 Classical k-means clustering method using 4 domains. (c) Geostatistical spectral clustering  
 413 using 4 domains. Z-axis was scaled to ease visualisation. Modified from Fouedjio et al.  
 414 (2017).

415

416

417

#### 418 5.5. *Teaching material*

419 The publicly available Rocklea Dome data set provided an opportunity to compile  
420 training and teaching material about the application of hyperspectral drill core and chips data  
421 for iron ore resource characterisation using The Spectral Geologist Software (TSG<sup>TM</sup>;  
422 <https://research.csiro.au/thespectralgeologist/>). Student exercises and example answers, as well  
423 as a ppt for teaching are part of the data package:

424 Exercise: StudentExercises\_Rocklea.docx

425 Answers: Answers\_CIDexercises.docx

426 PPT for teaching: MinSpec\_Workshop\_7RockleaDomeTSG\_HandsOn.pptx

427

## 428 6. CONCLUSIONS AND OUTLOOK

429 We have established an open-access dataset comprising drill core, surface and airborne  
430 hyperspectral data of the Rocklea Dome area in the Hamersley Basin of Western Australia,  
431 which features a wide variety of lithologies and morphologies and is prospective for channel-  
432 hosted iron ore resources. The proximal and remote sensing data, together with associated  
433 whole rock geochemistry are ideal for researching the geology of this economically significant  
434 area and allow a thorough comparison of different geoanalytical techniques and their  
435 effectiveness for resource characterisation. Combining the surface and subsurface data into 3D  
436 mineral maps provides a better visual understanding of the geological environment.

437 In addition to the already published surface and subsurface mineral mapping products,  
438 many more Geoscience Products can be generated to better understand this geologically  
439 complex area. The here newly presented white mica and chlorite abundance maps clearly



440 highlight the potential for mapping out different sections of the Archaean monzogranitic  
441 basement as well as different generations of mafic intrusives. Of particular interest are the  
442 contact zones between the mafic dykes and their host rocks, as they could help to better  
443 understand the intensity of alteration within the dyke and within the host granite as well as the  
444 associated fluid-rock interaction processes.

445 The teaching material provided together with this open-access dataset aims to support  
446 training of geoscience graduates and post-graduates in the potential applications of  
447 hyperspectral proximal and remote sensing data for mineral exploration and resource  
448 characterisation.

449 All analytical technologies used for collection of the geoscience data, as well as  
450 software packages used for processing the data, are commercially available. However, it should  
451 be noted that the HyChips™ system is now superseded by HyLogger3, which collects thermal  
452 infrared wavelengths (TIR; 6000 to 14500 nm) in addition to the VNIR-SWIR data. The  
453 collection of the TIR wavelength range enables the characterisation of major rock forming  
454 minerals such as quartz, which are of major importance for characterisation of iron ore  
455 resources, but were not detectable with HyChips™. The HyLogger3 technology is in operation  
456 at the six nodes of the Australian National Virtual Core Library  
457 (<https://www.auscope.org.au/nvcl>), which provides online open access to more than 3,500 drill  
458 cores from the Australian continent.

459

## 460 7. DATA AVAILABILITY

461 The supplement related to this article is available online at:  
462 <https://doi.org/10.25919/5ed83bf55be6a> (Laukamp, 2020). A 3D model of the Rocklea Dome  
463 data set is also available from the Geological Survey of Western Australia:  
464 <https://dasc.dmp.wa.gov.au/DASC?productAlias=Rocklea3D>.

465

466 **8. AUTHOR CONTRIBUTIONS**

467 CL, TC and MH contributed equally to the manuscript preparation. CL is the custodian  
468 of the Rocklea Dome data set stored on the CSIRO's Data Access Portal.

469

470 **9. COMPETING INTERESTS**

471 The authors declare that they have no conflict of interest.

472

473 **10. ACKNOWLEDGEMENTS**

474 The Rocklea Dome 3D Mineral Mapping project was funded by the Western Australian  
475 Government through their support to the Western Australian Centre of Excellence for Three-  
476 dimensional Mineral Mapping (C3DMM) in Kensington and by Murchison Metals Ltd. M.  
477 Cardy, A. Hackett, and S. Travaglione are acknowledged for the acquisition of the infrared  
478 spectroscopic data. This work profited from fruitful discussions with CSIRO colleagues C.  
479 Ong, A. Rodger, E. Ramanaidou, and M. Wells and with Murchison Metals Ltd. geologists J.  
480 Johnson and S. Peterson. E. Ramanaidou (CSIRO) as well as staff of Murchison Metals were  
481 instrumental in securing this site for public demonstration. The Geological Survey of Western  
482 Australia covered part of the costs for the diamond drilling through their Exploration Incentive  
483 Scheme co-funded Exploration Drilling program. Two anonymous reviewers are thanked for  
484 their constructive feedback.

485

486

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