

1 **Assessment of coastal Strandveld integrity using WorldView-2 imagery in**  
2 **False Bay, South Africa**

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9

10 **Keywords**

11 Coast; ecosystem integrity; WorldView-2; remote sensing; Strandveld; False Bay

12 **Highlights**

- 13 • The Cape Flats Dune Strandveld is under heavy anthropogenic pressure
- 14 • Satellite imagery was used to assess ecosystem integrity in False Bay
- 15 • Factors impairing Strandveld integrity include vegetation loss and alien infestation
- 16 • Results provide valuable information for local environmental management

## Abstract

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18 The coastal zone as interface between land and sea faces much pressure from human  
19 activities. These pressures make it difficult for the coastal zones to fulfil their natural  
20 functions, so threatening the state of coastal environments and making them vulnerable to  
21 coastal disasters and degradation. This study aimed to test whether high resolution remote  
22 sensing imagery can be used to assess the integrity of coastal Strandveld vegetation at the  
23 high spatial resolution required as baseline information for local environmental  
24 management. The study focused on the Cape Flats Dune Strandveld vegetation in northern  
25 False Bay near Cape Town, South Africa. The approach modifies and adopts a method  
26 developed by Lück-Vogel et al. (2013) for a remote sensing derived habitat intactness  
27 assessment for the Sandveld region at South Africa's west coast.

28 A regression analysis was performed on the eight spectral bands of two WorldView-2  
29 images and five image derivatives to evaluate the most suitable bands for producing an  
30 ecosystem integrity index. Based on the results, only the RED and the NIR bands were  
31 required to perform a decision tree classification for the integrity classification, which  
32 significantly simplifies the application and opens the way for transfer to sensors with similar  
33 spectral bands.

34 The overall accuracy of the results was 80.5% with a kappa value of 0.75. This means that  
35 this approach provides results accurate enough for coastal management and conservation  
36 applications at a local scale. A further finding is the importance of seasonality to delineate  
37 natural and alien vegetation accurately.

## 38 1 Introduction

39 The coastal zone is the area where the land and sea meet (Leewis et al. 2012). Coastal  
40 habitats are highly productive areas supporting a variety of biodiversity (Costanza et al.  
41 1997). Moreover, coastal habitats are important for shoreline stabilisation, as spawning  
42 grounds for marine life and buffers against natural hazards. They also provide coastal  
43 communities with a variety of natural resources and ecosystem services (Costanza et al.  
44 1997).

45 The key pressures affecting the integrity of terrestrial coastal vegetation (i.e. vegetation  
46 above the high water mark) include unregulated public access and trampling, particularly  
47 close to urbanised areas (Nayak et al. 1989, Nayak & Bahuguna 2001), which can create  
48 pathways on dunes that eventually lead to vegetation loss (Moulis & Barbel 1999). Other  
49 pressures arise from sand mining, logging of firewood, littering and the introduction of  
50 invasive alien species that degrade coastal habitats (Nayak & Bahuguna 2001; Costanza et  
51 al. 1997).

52 One example of a vegetation type affected by these threats is the Cape Flats Dune  
53 Strandveld (in the following referred to as: Strandveld) which occurs in the greater Cape  
54 Town area at the southwestern tip of South Africa (Figure 1). Strandveld vegetation occupies  
55 habitats under the direct influence of salt spray and other factors associated with seawater  
56 (Mucina et al. 2006). It comprises of tall, evergreen, hard-leaved shrubland with abundant  
57 grasses and annual herbs growing on deep and well-drained sand along the coast (Mucina et  
58 al. 2006). Typical Strandveld plant species are *Searsia glauca*, *S. laevigata*, *Metalasia*

59 *muricata*, *Phyllica ericoides*, *Euclea racemosa*, *Pelargonium betulinum*, *Salvia africana-lutea*,  
60 *Ruschia macowanii*, *Ehrharta villosa*, *Zygophyllum flexuosum*.

61 A major handicap for coastal managers and conservationists in the region is the lack of  
62 appropriate data and information on the degree of integrity of the Strandveld ecosystem at  
63 a high spatial resolution suitable for local management.

64 According to Westra et al. (2013) in order to fully understand the degree to which  
65 ecosystems are disturbed, it is essential to first establish the “pristine” condition of the  
66 ecosystem as baseline. However, deriving information about the degree of non-pristineness  
67 or loss of integrity through detailed field surveys for mapping and monitoring of coastal  
68 environments is frequently unfeasible due to the vast extent of the coastline, physical  
69 inaccessibility and security issues or lack of funding and skilled staff. If we understand  
70 “ecosystem integrity” *sensu* Roche & Campagne (2017) and Westra et al. (2013) as the  
71 structural pristineness of a system (from which a functional intactness can be deducted to a  
72 certain extent), technical approaches for integrity assessment become more feasible.  
73 Ulanowics & Hannon (1988) postulate that integrity, as a measure of ecosystem intactness  
74 causes ecosystems to use incoming solar radiation more effectively than damaged ones.  
75 Damaged ecosystems in contrast generate latent heat (as the photosynthetic efficiency  
76 decreases). Therefore intact systems appear to be “cooler”. Both, the photosynthetic  
77 efficiency and the temperature can be assessed using satellite imagery. In the last decades,  
78 this circumstance has been used for multispectral remote sensing assessments to derive  
79 information about regional scale land cover, ecosystem functional types and biodiversity  
80 (e.g. Turner et al. 2003; Kerr & Ostrovsky 2003; Fernández et al. 2010). However, the spatial  
81 resolution of these Landsat-type satellites was not sufficient to reflect small scale patterns

82 of vegetation, biodiversity and degradation for use in local conservation (Rocchini 2007). It  
83 was only with the upcoming of high resolution satellites in the last decade that within-  
84 ecosystem analysis of local degradation gradients became feasible for ecosystems of narrow  
85 spatial extent, such as coast lines.

86 A similar lack of quantitative and spatially extensive information on vegetation intactness  
87 has been described e.g. for the Sandveld region on South Africa's west coast as input for  
88 spatial fine scale planning and restoration purposes. The vegetation in the Sandveld region  
89 composed of different Fynbos and western Strandveld types, resembling largely medium  
90 dense, up to 2m high fine-leaved, largely evergreen vegetation. The Strandveld elements are  
91 comparable to the Strandveld vegetation, the former however characterised by more  
92 succulent elements and the latter by more grasses and annual herbs (Rebello et al. 2006).

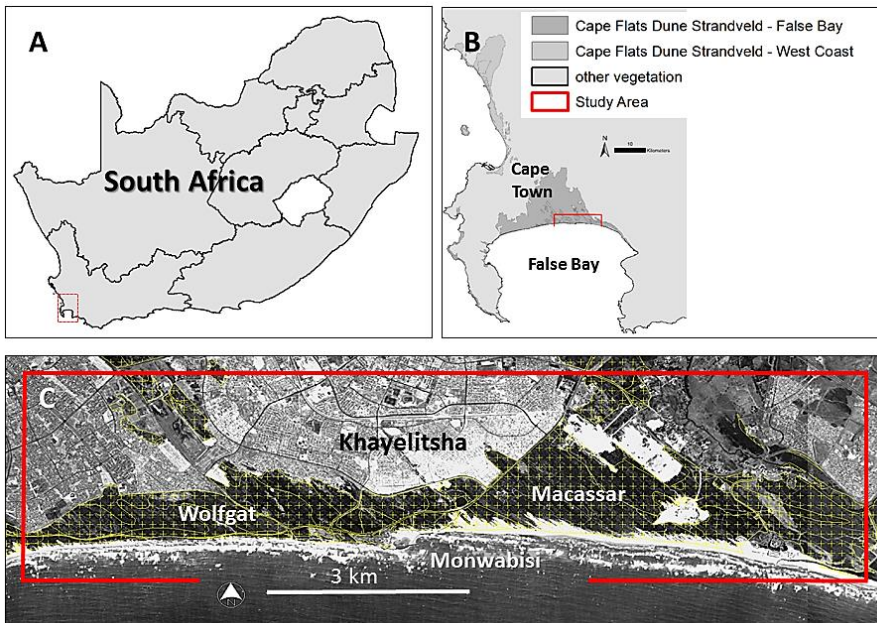
93 Lück-Vogel et al. (2013) developed a remote sensing based approach using a pansharpened  
94 15 m resolution Landsat-7 and a 20m resolution SPOT-5 image that integrated spectral,  
95 textural information and structural information derived from the imagery. The approach  
96 yielded an overall accuracy of 76.7% for the Landsat and 62.3% for the SPOT 5 imagery.  
97 However, as the approach was rolled out over the entire region composed of different  
98 vegetation types with naturally very different vegetation densities in the pristine climax  
99 condition, there was a vast over-estimation of degradation in naturally sparsely vegetated  
100 areas such as dune fields. Therefore, Lück-Vogel et al. (2013) recommended that further  
101 improvement of the results could be expected if the landscape were stratified prior to the  
102 intactness assessment, where clearly different vegetation types (e.g. reeds or open dune  
103 fields) were masked out or assessed individually.

104 The aim of this study was to test whether the technical approach presented by Lück-Vogel  
105 et al. (2013) but using high resolution satellite imagery can be used or adopted to derive  
106 reliable information for the Cape Flats Dune Strandveld vegetation in False Bay, on the  
107 southwestern tip of South Africa to provide fine-scale information on vegetation integrity  
108 for local management and conservation.

## 109 **2 Material and Methods**

### 110 **2.1 Study area**

111 False Bay is a large, partly protected bay (Figure 1), situated south of Cape Town in the  
112 Western Cape Province of South Africa (Theron & Schoones 2007). It extends between  
113 34°04' and 34°23' South and 18°26' and 18°52' East. False Bay is flanked by the Cape  
114 Peninsula mountain chain on the western and the Hottentots Holland Mountains on the  
115 eastern shore (Spargo 1991). The northern part of the Bay is relatively flat and protected  
116 from the ocean swell. The coast is therefore dominated by flat sandy beaches and dunes of  
117 tertiary or recent calcareous marine origin. However, in the eastern part of the northern  
118 bay, around Wolfgat and Macassar, cliffs and outcrops of underlying limestone rocks are  
119 present (Du Plessis & Glass 1991).



120

121 **Figure 1: Map of Study area. A: Location of study site in South Africa, B: Natural**  
 122 **distribution of Cape Flats Dune Strandveld in the greater Cape Town region, C: Map of**  
 123 **Study area. Areas of remaining Strandveld are indicated in yellow checked.**

124 False Bay has a Mediterranean climate with dry, hot summers from October to March and  
 125 wet cold winters from April to September (Clark et al. 1996). The mean temperature is  
 126 26.7°C in summer and 7.5°C in winter.

127 The northern part of False Bay extending from Strand to Muizenburg is dominated by Cape  
 128 Flats Dune Strandveld (Mucina et al. 2006). On the seaward side, this area borders on sandy  
 129 beaches and limestone cliffs. In the north it borders on the township Kayelitsha.

130 According to Mucina et al. (2006) and Holmes et al. (2012), Cape Flats Dune Strandveld is  
131 highly endangered by the invasion of alien plants such as *Acacia cyclops* (Rooikrans) and  
132 *Acacia saligna* (Port Jackson wattle), informal footpaths (access to the beach), illegal logging  
133 (fire wood), illegal sand mining, livestock overgrazing and illegal waste disposal, pollution  
134 and urban sprawl, largely resulting from the close proximity of the dense low-to-medium  
135 income, formal and informal settlements of Kayelitsha that are heavily encroaching the  
136 Macassar-Monwabisi-Wolfgat dune area from the north (especially since 2014). The  
137 introduction of alien invasive plants also increased the susceptibility of the area to veld fires.  
138 In the Municipality of Cape Town where most of the Strandveld vegetation is located, about  
139 55% of this vegetation is lost to transformation and about 36% of the remaining vegetation  
140 is formally protected as a Nature Reserve (City of Cape Town, no date).

141 This study focused on the remaining Strandveld occurring in the 12 x 3 km wide area  
142 between Macassar and Monwabisi. This area was selected as study area because it was  
143 accessible for collecting data in the field. Further, False Bay is a prime case of activities to be  
144 experienced on South Africa's coasts. Third, the processes impacting on vegetation integrity  
145 in the area are representative for the larger region.

## 146 **2.2 Input data**

### 147 **2.2.1 WorldView-2 (WV-2) satellite data**

148 Two partly overlapping WorldView-2 (WV-2) images acquired on 25 February 2014 and 11  
149 October 2014 respectively were used for the assessment of vegetation integrity in the study  
150 area. The WV-2 sensor has eight spectral bands (a coastal band in the short-wave blue  
151 region, a "common" blue band, a green, yellow and red band, and bands in the red-edge,



152 near-infrared1 (NIR1) and near-infrared2 (NIR2) range  
153 (<http://www.landinfo.com/WorldView2.htm>). The multispectral bands have a 2x2 m spatial  
154 resolution and a 16-bit data range which provides a high degree of spectral detail. The WV-2  
155 level 3A data used in this study were acquired from the South African National Space Agency  
156 (SANSA). The projection of the images is Universal Transverse Mercator (UTM) zone 34  
157 south. The satellite images were provided in a tile format.

### 158 **2.2.2 Biodiversity Network data**

159 The City of Cape Town, under whose government the study area falls, provided Biodiversity  
160 Network data in form of ESRI Shape (SHP) files that contained the natural vegetation types  
161 and subtypes in 2015. The distribution of the main vegetation type was largely identical with  
162 Mucina et al. (2006). At a subtype level the GIS layer further distinguished between  
163 Strandveld growing on sand or on lime stone. The data further contained information on  
164 built up or otherwise transformed areas. These data were used for masking out areas that  
165 were not of immediate interest to the study.

### 166 **2.2.3 Ground reference data**

167 Two field visits were conducted at the beginning of the project and at a later stage for the  
168 collection of ground reference data. During the first trip in August 2015, the collected and  
169 GPS-referenced information on vegetation integrity and factors impacting it was used for  
170 identification of the main impacting phenomena and the set up and validation of the  
171 integrity index approach.

172 In addition to the initial GPS-referenced field data, a further 100 points were collected on  
173 the screen, based on the general observations made in the field, resulting in a total of 130

174 reference points. They were used to set up a classification scheme and to perform an  
175 accuracy assessment of the results.

176 Given the continuous succession and recovery after disturbance, as well as fire events that  
177 occurred after the WV-2 image acquisitions, the judgement whether the classification for  
178 this point was correct or not, needed some “historical” information from the City of Cape  
179 Town’s field rangers which have been working in the area over the whole period. Therefore,  
180 a second joint field validation was conducted in June 2016.

### 181 **2.3 Image pre-processing**

182 The WV-2 images acquired on 25 February 2014 and 11 October 2014 used in the study  
183 were delivered in 5 and 6 individual tiles respectively. In order to enhance further  
184 processing, all the tiles per date were mosaicked into one image per date using ERDAS  
185 IMAGINE (version 2014).

186 The mosaicked images were then radiometrically corrected to remove illumination and  
187 atmospheric effects to ensure the transferability of the classification rule set from one  
188 image to the other. The software used was ATCOR2 embedded in ENVI IDL. The output was  
189 top-of-canopy reflectance (Richter 2014).

190 Both images were subset to the extent of the coastal area of interest. Further all areas other  
191 than untransformed Strandveld (i.e. urban areas, water bodies, roads and other vegetation  
192 types) were masked out, using the Biodiversity Network data. These subsets were created  
193 to minimise computation time and to avoid biases in the accuracy of the integrity index as  
194 suggested by Lück-Vogel et al. (2013).

195 In order to derive textural and structural characteristics in addition to spectral parameters  
196 for spatial features in the satellite images, the images were segmented using eCognition  
197 Developer (version 9.0) software (Darwish et al. 2003; Definiens 2007). The result was a  
198 polygon vector layer outlining homogeneous areas in the image from which five image  
199 derivatives (i.e. Brightness, Compactness, Area, NDVI and NIR1 standard deviation) were  
200 generated from both images. Brightness and NDVI (Normalised Difference Vegetation Index)  
201 provide spectral information complementing the original 8 spectral WV-2 bands (Pettorelli  
202 et al. 2005). The compactness is an indicator of the “squareness” of a landscape feature.  
203 Natural landscape features appear more frayed/fractal than man-made features such as  
204 land use parcels or roads with straight boundaries. The polygon area adds structural  
205 information. According to Lück-Vogel et al. (2013) the NIR1 standard deviation can be used  
206 as proxy for the spectral within-polygon heterogeneity of the image object because natural  
207 vegetation, in contrast to e.g. agricultural crops and plantations, is characterised by a  
208 species- and age structure-diverse canopy that is expressed in the image objects in a higher  
209 local spectral variability. All five derivatives were calculated on both WV-2 images using  
210 ERDAS 2014 and eCognition software.

211 The segmented eight spectral bands and five image derivatives (Brightness, Compactness,  
212 Area, NDVI and NIR1 standard deviation) were exported from eCognition as smoothed  
213 polygon Shape files. They were then rasterised in ArcGIS (version 10.1) with a 2 m pixel size  
214 to match the original WV-2 resolution. All 13 bands were then unified in one multispectral  
215 layerstack for further processing in ERDAS.

216 **2.4 Identification of processes impacting vegetation integrity**

217 During the first field trip in August 2015, with the aid of GPS-referenced points, the  
218 condition of the Strandveld vegetation and its degree of integrity were established on the  
219 WV-2 imagery. Ecological integrity in this case was defined as “the state of being sound,  
220 unimpaired/unaltered by human activities and its complementary anthropogenic pressures  
221 and disturbances” (Roche & Campagne 2017; Westra et al. 2013). In contrast to the actual  
222 functional ecological intactness which is rather process-focussed (Roche & Campagne 2017),  
223 the integrity in terms of unimpairedness, expressed as natural density and diversity of the  
224 vegetation cover and deviations thereof, can more readily be assessed using remote  
225 sensing. The field trip helped to identify areas which are as close as possible (in a  
226 Metropolitan vicinity) to its undisturbed natural condition. Using these areas as baseline,  
227 the processes leading to an increasing impairment of integrity were identified.

228 Two types of largely intact woody climax Strandveld vegetation occurred in that area. On  
229 deep sandy soils, 2-4m high dense shrubby vegetation dominated, while on shallow soils on  
230 limestone rocks a heath-like, hardy shrub vegetation up to 2m height dominated. Both  
231 aspects were assumed to represent the “unimpaired” or intact condition.

232 Further, some areas were dominated by indigenous herbaceous vegetation. This vegetation  
233 is a natural community of the Strandveld vegetation on shallow soils and was thus  
234 considered as an intact aspect of Strandveld as well.

235 Apart from these largely undisturbed areas, various processes were observed that impacted  
236 the integrity of the Strandveld to various degrees. Patches of bare soil, i.e. areas that were

237 depleted of all vegetation were considered the least intact state. Those areas included for  
238 instance footpaths and (illegal) drive ways.

239 The other extreme of impaired integrity were large areas where dense stands of alien  
240 *Acacia cyclops* and *Acacia saligna* had evicted most of the indigenous vegetation. Areas  
241 invaded by alien vegetation are also more prone to fires than intact Strandveld due to the  
242 higher degree of succulence of the latter (Rebelo et al. 2006).

243 Apart from these two extreme cases of impaired integrity, two forms of less extreme  
244 degradation were observed, both related however to the occurrence of alien vegetation.  
245 Firstly, scars in the (frequently previously alien infested) Strandveld from fires during years  
246 preceding the image acquisition presented various stages of biodiverse vegetation regrowth  
247 and were considered less impaired, as it was the indigenous vegetation recovering towards  
248 a higher degree of "unimpairedness".

249 Secondly, areas that were cleared of alien vegetation recently consisted of some remaining  
250 larger indigenous shrubs with frequently unconsolidated open soil where acacias had been  
251 removed physically. Undergrowth was largely absent as the dense cover of aliens had  
252 impeded its development. Those areas were thus considered as heavily impaired or  
253 removed from the natural integrity of the system.

254 The vegetation categories that were observed are summarised in Table 1, ranked according  
255 to an integrity gradient from 1 (highly impaired) to 4 (largely undisturbed).

256 **Table 1: Identified vegetation type relating to the level of integrity**

Vegetation category	Assigned habitat integrity index	
Description	Integrity Value	Description
Bare Soil	1b	Highly impaired (bare)
Alien vegetation thickets	1a	Highly impaired (alien)
Recently cleared alien vegetation	2	Moderately impaired
Strandveld recovering from veld fires	3	Lightly impaired
Natural pristine herbaceous and woody Strandveld vegetation on limestone or sand	4	Largely unimpaired

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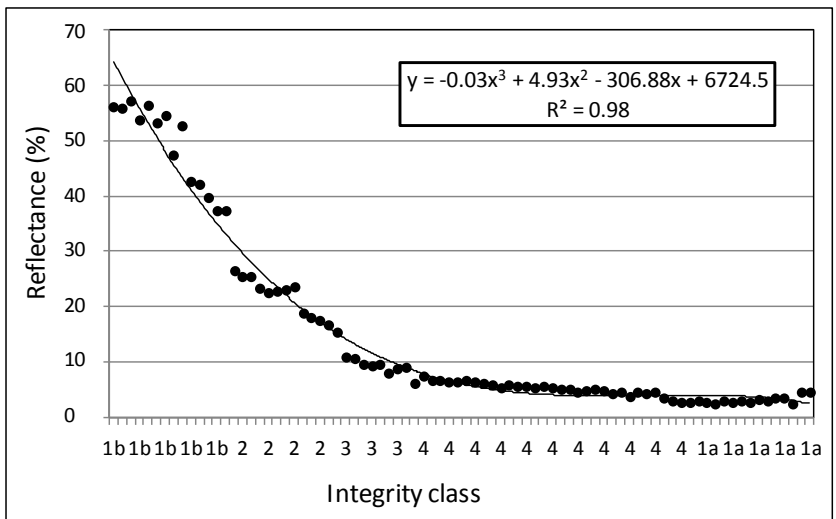
258 All available 130 ground reference points were assigned the respective habitat integrity  
 259 value. Two thirds of the points were used for training the decision tree classification, and  
 260 the remaining one third was kept for later validation purposes.

261 **2.5 Decision tree classification**

262 For each of the training points, the values from all 13 spectral WV-2 bands and derivatives  
 263 were extracted and exported to MS Excel. Then the correlation between the 13 bands and  
 264 the respective integrity class was assessed using a 3<sup>rd</sup>-order polynomial regression analysis  
 265 (coefficient of determination R<sup>2</sup>). This procedure was performed on the 25 February 2014  
 266 WV-2 image only.

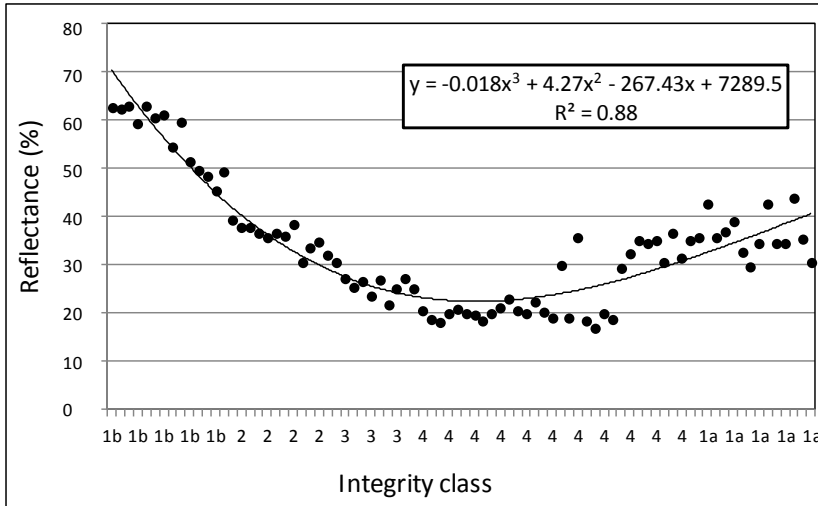
267 For the spectral bands, including NDVI and Brightness, two general patterns of relationship  
 268 were visible. These are illustrated in the regression graphs in Figure 2 and Figure 3 which  
 269 show the relationship between the five integrity classes on the X-axis and the mean  
 270 reflectance values for the red band and the NIR1 respectively on the Y-axis. The integrity

271 classes were sorted from highly degraded on the left with an increase in intactness to the  
272 right (with exception of class 1a: alien vegetation).



273  
274 **Figure 2: Relationship between levels of integrity and reflectance in the red band from the**  
275 **February 2014 image. For description of the integrity classes refer to Table 1.**

276



277

278 **Figure 3: Relationship between levels of integrity and reflectance in the NIR1 band from**  
 279 **the February 2014 image. For description of intactness class refer to Table 1.**

280 For the WV-2 bands 1 – 5 in the visible range (coastal, blue, green, yellow and red), a  
 281 continuous decrease in reflection from the classes of highest degradation (intactness class  
 282 1b) to highest integrity (and alien infestation as additional degradation class; integrity  
 283 classes 4 and 1a respectively) similar to Figure 2 was observed. The Infrared bands 6 – 8  
 284 (red-edge, NIR1 and NIR2) and the brightness followed the trend visible in Figure 3 where  
 285 the Alien infested areas (class 1a) can be distinguished from the most pristine class 4 by  
 286 slightly higher reflectance.

287 As these two spectral patterns were the same in all spectral bands, for simplicity reasons it  
 288 was decided to build the classification based on thresholding in the RED and the NIR1 bands  
 289 only as these bands are available in most other earth observation sensors, should the



290 transfer of the approach to another sensor be desired. The classification was applied on  
291 both the 25 February and 11 October 2014 image.

292 It was found that structural and textural derivatives, i.e. the compactness, Area and NIR1  
293 Standard Deviation, in contrast to Lück-Vogel et al.'s (2013) work, did show no strong  
294 correlations to the integrity classes. The  $R^2$  values for the 3<sup>rd</sup> order polynomial regression for  
295 these bands ranged between 0.09 and 0.18. The reason for this might be that the structural  
296 and textural derivatives are very useful to distinguish between different vegetation types  
297 but not within a vegetation type, as in this study all other vegetation and land cover was  
298 masked out from the beginning (see section **Error! Reference source not found.**). In the  
299 NDVI band, there was a positive linear relation visible between NDVI and integrity (apart  
300 from class 1a). However, given the extra processing step for calculating the NDVI, it was  
301 decided to use the original spectral input bands, red and NIR1 instead for the development  
302 of the decision tree.

## 303 2.6 Validation of results

304 Two approaches were used to assess the accuracy of the classification results. The first one  
305 made use of the 41 ground reference points described in section 2.2.3 to generate an error  
306 matrix and deducted accuracy parameters (Congalton & Green 2009) using ERDAS's  
307 classification accuracy tool. An error matrix is a comparison between remote sensing results  
308 and ground truth data (reference data). It identifies misclassification errors and their  
309 quantity.

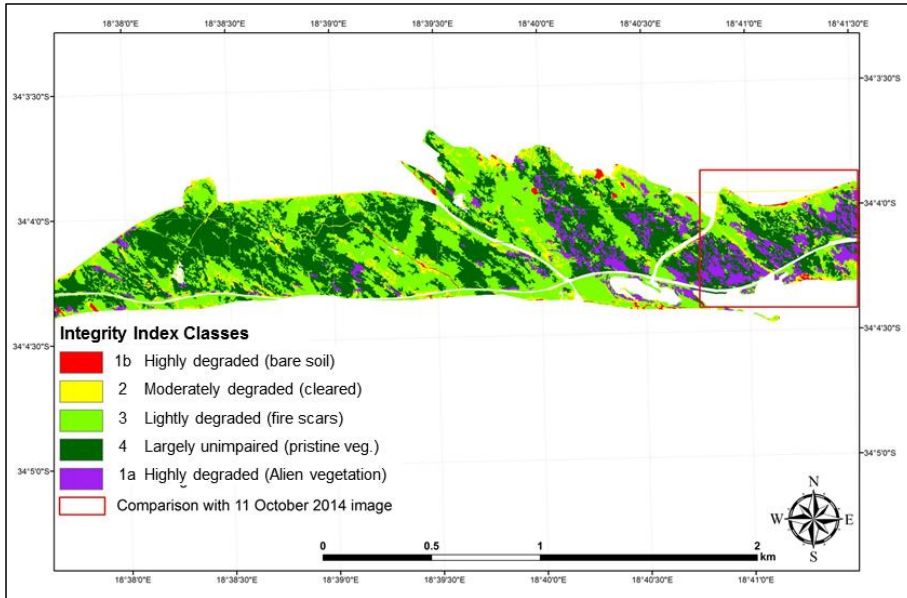
310 Additionally, a further field trip was conducted in June 2016 to verify the results. The  
311 printed map results of both the February and October 2014 images were taken along for the

312 selection of points of interest for validation. Investigated points were marked with GPS. A  
313 total of 18 field validation points was captured. The GPS points were later converted to a  
314 shapefile for further GIS analysis. Correspondence between appearance in the field and the  
315 classification output was analysed.

### 316 **3 Results and Discussion**

#### 317 **3.1 Derived integrity map for February 2014**

318 Figure 4 shows the integrity map derived from the WV-2 image captured on 25 February  
319 2014, representing the dry summer season. The classification identified areas ranging from  
320 highly degraded to high integrity. Highly degraded areas (red) are mainly bare areas of  
321 footpaths, and informal roads (could not be masked out initially) and other open patches  
322 without vegetation. Moderately degraded areas (yellow) indicate areas of cleared alien  
323 vegetation. Lightly degraded areas (light green) mainly represent regrowth of vegetation  
324 recovering from fires. Intact indigenous vegetation of the Strandveld is indicated in dark  
325 green. Purple indicates areas infested with alien invasive vegetation, mainly *Acacia cyclops*.



326

327 **Figure 4 Integrity map derived from 25 February 2014 image (dry season).**

328

329 **3.1.1 Accuracy based on error matrix**

330 Table 2 below shows the results of the accuracy assessment for the classification of the  
 331 February image in form of an error matrix. From the error matrix several metrics were  
 332 derived to assess the quality of the classification. The overall accuracy and the Kappa  
 333 coefficient are measures of overall agreement between the remotely sensed classification  
 334 and the reference data (Congalton & Green 2009). The producer's accuracy indicates how  
 335 well the training points per class are classified, while the user's accuracy indicates the  
 336 probability of a pixel being assigned the same class we actually find on the ground (Lillesand  
 337 et al. 2004).

338 **Table 2: Error matrix for the classification results of the 25 February 2014 WV-2 image**  
 339 **based on field informed reference points. Shaded cells: correctly classified points.**

		Ground truth reference					classified total	User's Acc.	
		1b	2	3	4	1a			
Classified values	1b Highly degr. (bare)	4	0	0	0	0	4	100.0	
	2 Mod. degraded	2	4	0	0	0	6	66.7	
	3 Lightly degraded	0	1	6	0	0	7	85.7	
	4 Pristine	0	0	0	12	0	12	100.0	
	1a Highly degr. (alien)	0	0	0	5	7	12	58.3	
	Reference points total	6	5	6	17	7	41		
Producer's Accuracy		66.7	80.0	100.0	70.6	100.0			
<b>Kappa</b>		<b>0.75</b>							
<b>Overall Accuracy</b>		<b>80.5 %</b>							

340

341 Table 2 indicates that 33 points out of the total 41 reference points were classified correctly.

342 Four out of 6 reference points for highly degraded (class 1b) were correctly classified; 4 out

343 of 5 reference points for moderately degraded (class 2) were correctly classified; all 6

344 reference points for lightly degraded (class 3) were correctly classified; 12 out of 17

345 reference points for pristine vegetation (class 4) were correctly classified; and all 7 alien

346 reference points (class 1a) were correctly classified in the WV-2 image for 25 February 2014.

347 Of the 6 reference points for highly degraded two were incorrectly classified as moderately

348 degraded. A main reason for this error is probably the similarity of the spectral properties of

349 the two classes in cases where alien clearing was conducted just prior to the image

350 acquisition hence leaving the ground largely bare. Furthermore, confusion between

351 moderately and lightly degraded occurred, probably due to the continuous gradient of

352 degradation severity between both classes. However, about 30% of the pristine vegetation

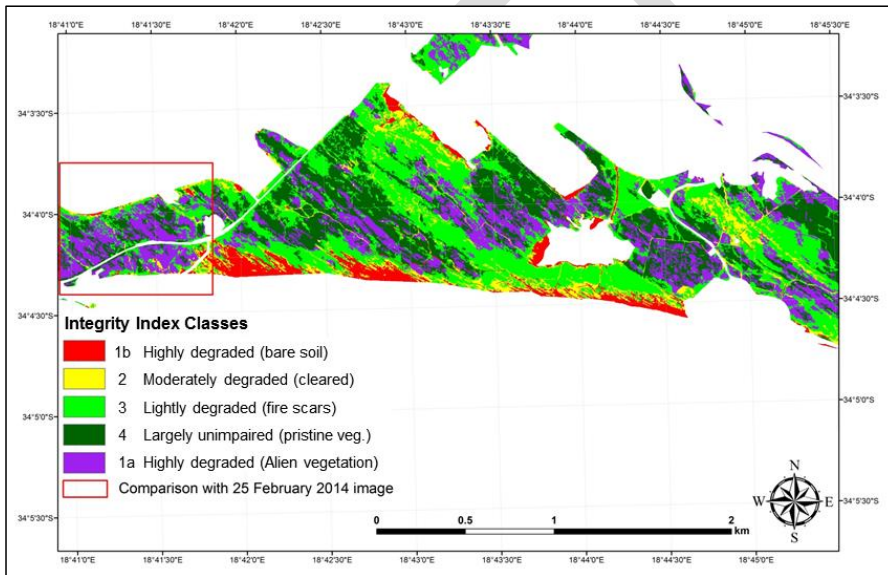
353 reference points were wrongly classified as alien vegetation. This means, the occurrence of  
354 alien vegetation was over-estimated. This confusion might have occurred in cases where  
355 pristine vegetation appeared on exceptionally moist places (leading to above-average  
356 vegetation vigour) or in places where patches of milkwood forest (*Sideroxylon inerme*)  
357 occurred which are not classified separately for that area in the Biodiversity Network data.  
358 Milkwood would have a higher vegetation density and would thus appear more like alien  
359 vegetation in the classification.

360 The best producer's accuracy was achieved for alien vegetation and lightly degraded at  
361 100%, with intact vegetation scoring 70.6%. The least producer's accuracy was registered for  
362 highly degraded at 66.7%. Satisfactory user's accuracies were achieved for pristine  
363 vegetation at 100%. This means that pristine vegetation should be found in 100% of the  
364 cases where it is indicated in the map. The second highest user's accuracy was found for  
365 lightly degraded at 85.7%. The minimum user's accuracy was achieved for alien vegetation  
366 at 58.33% which is considered to be below the desired level for satisfactory accuracy.  
367 However, the experienced over-estimation of alien vegetation is still preferable to an  
368 underestimation, as the over-estimation still has merit for the land manager has a pre-  
369 screening tool.

370 The high overall accuracy of 80.5% and a kappa of 0.75 for the February classification  
371 indicate substantial agreement (Congalton & Green 2009; Landis & Koch 1997) between the  
372 classification results and the reference data. The high overall accuracy is ascribable to the  
373 high spatial resolution of 2 metres and the high spectral resolution of the bands of the WV-2  
374 image, as well as the exclusion of transformed areas and non-Strandveld vegetation as per  
375 recommendation by Lück-Vogel et al. (2013).

376 **3.2 Derived integrity map for October 2014**

377 Figure 5 shows the integrity map derived from the WV-2 image of 11 October 2014,  
378 representing the wet winter season. The general occurrence and description of the classes is  
379 the same as in the February image. However, in the October map naturally bare coastal  
380 areas on the coast (southern fringe of map), such as dunes are (wrongly) classified as highly  
381 degraded (red), too. Apparently here the masking of non-Strandveld vegetation was  
382 inaccurate.

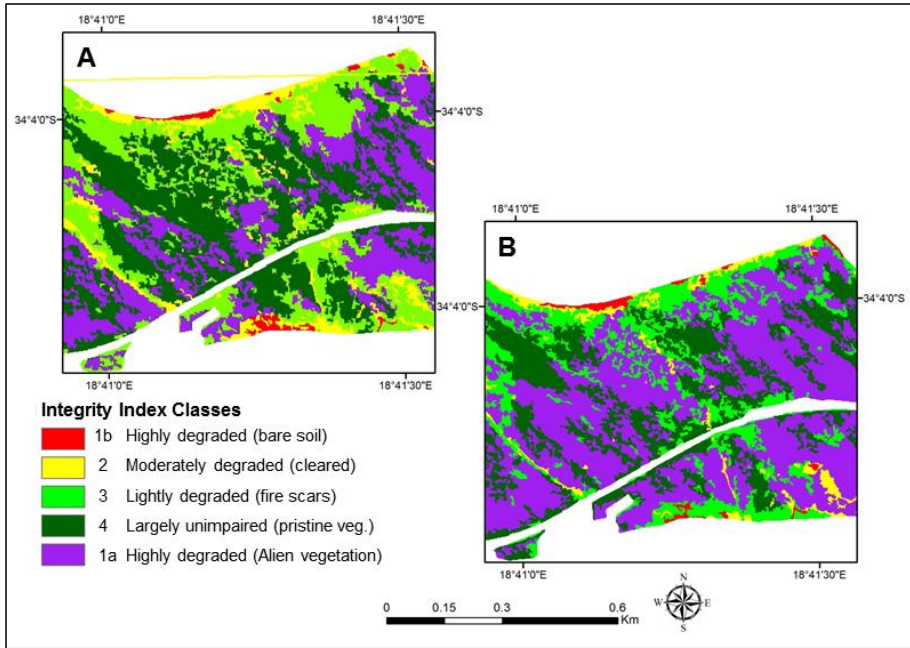


383

384 **Figure 5: Integrity map derived from 11 October 2014 image (wet season).**

385 **3.3 Comparison between February and October integrity maps**

386 Figure 6 compares the area where the February and the October maps overlap (indicated by  
387 the red frames in Figure 4 and Figure 5).



388

389 **Figure 6: A) Integrity map for the 25 February 2014 image; B) for the same area for the 11**  
 390 **October 2014 image.**

391 The October classification shows a significant increase in alien vegetation compared to the  
 392 classification from February. However, field validation (see below) showed that this  
 393 apparent increase is not caused by a true increase in alien vegetation but a phenological  
 394 effect of vegetation growth given different seasons in which the images were taken. It  
 395 appears that during the dry season, indigenous vegetation reduces their photosynthetic  
 396 activity to a minimum, while the alien vegetation still remained highly photosynthetic active.  
 397 In the rainy season, the indigenous vegetation is highly photosynthetically active too, thus  
 398 wrongly being classified as alien vegetation. These results give important insights on how

399 the seasonality influences the accuracy of alien vegetation detection in a Strandveld  
400 environment.

### 401 **3.4 Field validation**

402 As mentioned in section 2.2.3, one additional field trip was conducted in June 2016 in order  
403 to gain “historical” information about the dynamics and changes that took place in the area  
404 during the time between the WV-2 image acquisition and the conduction of this project in  
405 2016. Based on the information provided by the Nature Reserve staff, the mapping results  
406 conducted for both 25 February 2014 and 11 October 2014 images were interpreted. Out of  
407 the 18 investigated points, for 13 points field observation and classification coincided.

408 Out of the wrongly classified 5 points, three confused pristine vegetation with alien  
409 vegetation. This happened either in areas where alien vegetation did not come to complete  
410 dominance yet, or on areas which were only classified as alien in the October image  
411 (phenological effects, see above). For one point, a moderate degradation was confused with  
412 a light degradation where a fire occurred just prior to the image acquisition, and in one  
413 place, azonal, dense riverine vegetation was confused with pristine vegetation. This non-  
414 Strandveld vegetation patch should have been masked out in the initial phase of the  
415 analysis. For a detailed description of the field observations refer to Mbolambi (2016).

## 416 **4 Conclusions and recommendations**

417 The presented paper aimed at identifying degrees of vegetation integrity within the Cape  
418 Flats Dune Strandveld vegetation in the north-eastern part of False Bay, Cape Town, South  
419 Africa using WorldView-2 multispectral satellite imagery. Partly overlapping imagery from



420 February and October 2014 with a spatial resolution of 2x2m was used. Based on field  
421 observations, a decision tree classification for both images was conducted. Regression  
422 analysis between the individual spectral bands and the degradation classes allowed to  
423 identify the two most suitable spectral bands for the decision tree classification, i.e. the red  
424 and the NIR1 band. Accuracy assessment was conducted based on additional field data. The  
425 overall accuracy of the results was 80.5% with kappa statistical value of 0.75. The accuracies  
426 thus show that the masking of all other vegetation or land cover types prior to image  
427 classification does increase the classification accuracy, as stipulated by Lück-Vogel et al.  
428 (2013). The results showed that with satisfactory accuracy it was possible to distinguish  
429 areas with high vegetation integrity from areas which are highly, moderately and lightly  
430 degraded. The distinction between intact vegetation and dense alien *Acacia* vegetation  
431 turned out to be less accurate and highly influenced by the actual acquisition date of the  
432 satellite imagery. The study results revealed that Images sensed in the dry, summer season  
433 better detected alien invasive vegetation because then the signal from the indigenous  
434 vegetation is significantly lower than the alien vegetation signal. During the wet season  
435 however, indigenous vegetation is more active and easily confused with alien vegetation.  
436 For future approaches, it is also advised to conduct field work as closely to the image  
437 acquisition date as possible, to avoid inconsistencies between field and image data due to  
438 natural vegetation succession and human impact such as de-bushing or fires happening in  
439 the interim.

440 The high spatial resolution of the WorldView-2 image allowed for the detection of very small  
441 scale degradation patterns, e.g. caused by illegal footpaths. The highly detailed maps

442 therefore provide a suitable tool for environmental managers to put in place local  
443 management actions and to monitor their success.

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455

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