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2 **Is the patch size distribution of vegetation a suitable**  
3 **indicator of desertification processes?:**

4 **Comment**

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21 Manuscript type: Comment

22 With ongoing climate change, the search for indicators of imminent ecosystem shifts is  
23 attracting increasing attention (e.g. Scheffer et al. 2009). Recently, the spatial organization of  
24 ecosystems has been suggested as a good candidate for such an indicator in spatially  
25 structured ecosystems (Rietkerk et al. 2004, Kéfi et al. 2007a, Guttal and Jayaprakash 2009).  
26 Arid ecosystems are well-known for the spatial organization of their vegetation cover, which  
27 is often characterized by clumps of vegetation in an otherwise bare soil matrix. Two recent  
28 studies revealed that the distribution of the vegetation patch size can be described by a power  
29 law over a wide range of environmental conditions in arid ecosystems (Kéfi et al. 2007a,  
30 Scanlon et al. 2007). Furthermore, deviations from power laws were observed under high  
31 grazing pressures, leading to the hypothesis that such deviations could be used as indicators of  
32 approaching desertification in arid ecosystems (Kéfi et al. 2007a). This hypothesis now needs  
33 to be validated with additional field data, before it can be confidently used as a tool to monitor  
34 degradation in arid ecosystems.

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36 In a recent study, Maestre and Escudero (2009) (hereafter referred to as ME09) aimed to test  
37 this hypothesis with data from 29 steppes located on a rainfall gradient in southeast Spain. In  
38 all of their sites, the patch size distribution was best described by a truncated power law  
39 (TPL)<sup>1</sup>. Relating the scaling exponents of these TPLs to soil variables, the authors concluded  
40 that 1) the patch size distribution was not directly related to desertification but rather that 2)  
41 vegetation cover could be used to monitor desertification. We argue in this comment that the  
42 analyses of ME09 do not allow them to draw these conclusions, for the following two  
43 reasons. First, because all of their sites were characterized by TPLs, the authors looked only

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1 We use here the same terminology as in Kéfi et al. (2007) and Maestre and Escudero (2009), where a TPL refers to a power law with exponential cutoff, i.e. such that  $N(S)=CS^{-\gamma}\exp(-S/S_c)$  with N the number of patches of size S, C a constant,  $\gamma$  the scaling exponent (positive) and  $S_c$  the patch size above which N decreases faster than in a power law (positive).

44 at the scaling exponents  $\gamma$  of these TPLs to compare the degradation level of the sites.  
45 However,  $\gamma$  was not proven to vary with degradation in a consistent manner, and therefore  
46 the analyses of ME09 do not allow them to conclude whether vegetation cover is better  
47 related to degradation than patch size distribution. Second, although the vegetation cover is  
48 often a simple and efficient indicator of degradation, the authors do not take into account the  
49 increasing amount of literature that strongly suggests vegetation cover in arid ecosystems is  
50 likely to respond in a discontinuous way to gradual, external changes (Rietkerk et al. 1996,  
51 Lejeune et al. 1999, Scheffer et al. 2001, von Hardenberg et al. 2001, Kéfi et al. 2007b). In  
52 these cases, the vegetation cover alone simply does not provide information on the actual  
53 distance to desertification.

54

55 The categorization proposed by Kéfi et al. (2007a) is a qualitative one in that it does not  
56 provide a quantifiable distance to extinction: a shift (in time) from a pure power law to a TPL  
57 suggests that an ecosystem is degrading and approaching the desertification threshold. The  
58 sites studied by ME09 are all described by TPLs. Among sites characterized by similar patch  
59 size distributions, Kéfi et al. (2007a) do not propose any criteria to distinguish among sites of  
60 varying degradation; currently, such criteria are sorely lacking. In an attempt to compare the  
61 degradation levels of their 29 sites, ME09 investigated changes in the scaling exponent  $\gamma$  of  
62 the TPL among the different sites. This was not part of the hypothesis formulated by Kéfi et  
63 al. (2007a). It is an interesting approach, but it implicitly assumes that  $\gamma$  varies consistently  
64 with the level of stress, which has not been proven to be the case. In fact, in the data analyzed  
65 by Kéfi et al. (2007a), there does not appear to be any consistent variation of  $\gamma$  among sites  
66 characterized by different stress levels (i.e. grazing pressures). For example, with increasing  
67 grazing pressure (from medium to high) the absolute value  $\gamma$  of the TPL decreases in the data

68 from Spain but increases in the data from Morocco and Greece (see Fig. 1 in Kéfi et al.  
69 (2007a)). The lack of a clear relationship between  $\gamma$  and the stress level could very well  
70 explain why ME09 find that  $\gamma$  is not related to the perennial cover. It is noteworthy that this  
71 result is in agreement with previous studies on steppes dominated by *Stipa tenacissima* in the  
72 arid Mediterranean region. For example, it has been shown that the spatial distribution of *S.*  
73 *tenacissima* is not clearly related with its abundance (see Table 1 in Alados et al. (2006)).  
74 Furthermore, the exponent  $\gamma$  alone does not provide a complete description of the shape of  
75 the TPL; the location of the cutoff,  $S_c$ , cannot be ignored. Indeed, the latter describes where  
76 the deviation from power law behavior begins, and it is this deviation which was proposed to  
77 be linked to the level of degradation in Kéfi et al. (2007a). Thus, we doubt whether  $\gamma$  is the  
78 correct parameter to investigate. Further theoretical and empirical work is needed in order to  
79 identify the parameters which are best correlated to the stress level and which therefore  
80 should be monitored.

81

82 Another concern regarding the analysis of ME09 is that, when fitting TPLs to their data, they  
83 find negative power law exponents in the vast majority of their sites (22/29 sites listed in  
84 ME09 Table 1 and 7/8 sites illustrated in ME09 Fig. D1), in stark contrast to the positive  
85 power law exponents observed by Kéfi et al. (2007a). A TPL with a negative power law  
86 exponent can be understood as follows: the number of patches  $N(S)$  actually increases with  
87 size  $S$  until some intermediate path size is reached, at which point  $N(S)$  begins to decrease.  
88 Thus, in ME09's distributions, it is common for smaller patches to be less abundant than  
89 patches of intermediate size. For this reason, a TPL does not appear to be the most appropriate  
90 model to use to fit the data. The distributions found by ME09 actually suggest the presence of  
91 a dominant spatial scale, contrary to the scale invariance observed by Kefi et al. (2007a).

92 Indeed, some arid areas are characterized by regular vegetation patterns (Rietkerk and van de  
93 Koppel 2008), where patch size distributions do not follow power laws but instead reflect a  
94 characteristic patch size (or a range of patch sizes). Manor and Shnerb (2008) developed a  
95 promising model which can reproduce both the irregular patterns described by power law  
96 distributions and the regular patterns characterized by a dominant spatial scale. They showed  
97 how the relative strength of competition and facilitation can drive the type of pattern that  
98 emerges; strong facilitation favors irregular pattern formation while strong competition favors  
99 regular patterns. In systems characterized by regular patterns, it has been suggested that the  
100 shape of the patterns can be used to gauge the level of degradation, with spot patterns being  
101 the last to occur before desertification (Rietkerk et al. 2004). Further research is needed to  
102 determine if these findings can indeed be applied to the sites studied by ME09. More  
103 generally, what is currently lacking is a robust way of characterizing the spatial organization  
104 of ecosystems, since, depending on the type of patterns (which emerge from different  
105 underlying ecological mechanisms), the indicators that need to be monitored may vary.

106

107 Before patch size distributions can be used as a monitoring tool in systems characterized by  
108 irregular patterns, many technical issues need to be addressed and further tests need to be  
109 conducted in the field. From a practical point-of-view, the patch size distribution is indeed a  
110 more complicated tool than the vegetation cover. Among others, there are issues with the  
111 binning of the data and the fitting of the mathematical functions.

112 Traditionally, data is binned when visualizing frequency distributions (Newman 2006, Bauke  
113 2007, White et al. 2008, Clauset et al. *In press*). When the data are binned into bins of equal  
114 sizes (so-called linear binning), the right-hand side of the distribution is often noisy: the  
115 largest elements are rare, and, therefore, each bin contains only a few elements which creates

116 large variations in bin counts among bins (Newman 2005, Bauke 2007). This is a concern  
117 when dealing with patch size distributions, since we are especially interested in the behavior  
118 of the putative power law in the area around the largest, i.e. the rarest, patches. To decrease  
119 the noise in the right-hand tail of the distribution, logarithmic binning is typically employed,  
120 where the bins in the tail of distribution receive more elements than with linear binning.  
121 Various techniques have been proposed to estimate the optimum bin size (e.g. Sturges' rule,  
122 Scott's rule, and the Freedman-Diaconis rule); all strive to achieve a reasonable balance  
123 between the number of bins and the number of elements in each bin. However, these  
124 techniques do not always yield consistent results, which makes the choice of binning fairly  
125 arbitrary. A better way of plotting the data is to use the cumulative distribution function,  
126 which does not involve the binning of the data (Newman 2005, Bauke 2007, White et al.  
127 2008).

128 After binning the data, a linear fitting of the log-log transformed data is typically performed  
129 using least squares regression (Newman 2005, Bauke 2007, White et al. 2008, Clauset et al.  
130 *In press*). Fitting methods based on binning and least squares regression are widely used in  
131 ecology and in other fields to fit models to data and to estimate the scaling exponents of  
132 frequency distributions. White et al. (2008) recently demonstrated that such methods give  
133 biased results and therefore cannot be relied upon. While these biases are dangerous with  
134 regards to estimating the scaling exponent of a distribution, binned-based methods can also  
135 lead to differences in the determination of which distribution best fits the data. For example, a  
136 data set that is best described by a power law using a given bin size could be best described  
137 by a TPL when using a different bin size.

138 Independently of the way the data are plotted, a reliable alternative to least square linear  
139 regression is to use fitting methods based on maximum likelihood estimation (MLE) to

140 extract the scaling exponent of the frequency distribution (e.g. Goldstein et al. 2004). White et  
141 al. (2008) showed that MLE is the single most accurate method for estimating the scaling  
142 exponents of frequency distributions. Currently, MLE is available for the pure power law  
143 distribution (Goldstein et al. 2004, Newmann 2005, Bauke 2007) but not for the TPL  
144 distribution as defined here, which limits the application of MLE to this particular case for  
145 now, but is a promising line of future research.

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147 In conclusion, although looking at the vegetation cover is still the most obvious and practical  
148 way of assessing the "health" of an arid ecosystem, there are notable cases where the cover  
149 fails to predict desertification. Desertification can occur in sudden shifts, where ecosystems  
150 switch from an unknown vegetation cover to desert. Theoretical studies increasingly suggest  
151 that ecosystems which include facilitation may respond to gradual external changes in an  
152 abrupt, rather than gradual manner (e.g. Lejeune et al. 1999, Scheffer et al. 2001, von  
153 Hardenberg et al. 2001, Rietkerk et al. 2004, Kéfi et al. 2007b). In these cases, the vegetation  
154 cover is not an appropriate indicator of proximity to shifts and, therefore, other potential  
155 indicators (e.g. the patch size distribution) need to be further developed so that they can be  
156 used in addition to the cover. In the meantime, we advocate the continued but cautious use of  
157 the cover as a means to gauge an arid ecosystem's health, but reiterate the need to explore  
158 more robust techniques before a reliable early-warning system can be implemented.

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