

# A new contagion index to quantify spatial patterns of landscapes

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## Abstract

A contagion index was proposed by O'Neill *et al.* (1988) to quantify spatial patterns of landscapes. However, this index is insensitive to changes in spatial pattern. We present a new contagion index that corrects an error in the mathematical formulation of the original contagion index. The error is identified mathematically. The contagion indices (both original and new) are then evaluated against simulated landscapes.

## Introduction

A robust landscape index should quantify two distinct components of landscape diversity: composition and configuration (Li 1989). Composition refers to both the total number of patch types and their relative proportions in the landscape, whereas configuration refers to the spatial pattern of patches in the landscape. If an index only depicts composition its usefulness is limited because many existing indices do just that (e.g., Pielou 1975). And spatial configuration and its effects on landscape function are key topics of landscape ecology (Risser *et al.* 1984; Forman and Godron 1986).

A contagion index ( $D_2$ ) was proposed by O'Neill *et al.* (1988; Also Turner and Ruscher 1988; Turner 1989; Turner *et al.* 1990a; Turner 1990b; Graham *et al.* 1991) to characterize landscape pattern:

$$D_2 = EE_{\max} - EE \quad (1)$$

where

$$EE = - \sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \quad (2)$$

$$EE = - \sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \quad (3)$$

where  $n$  is the total number of patch types in a landscape mosaic,  $P_{ij}$  is the probability of patch type  $i$  being adjacent to patch type  $j$ , and  $EE_{\max}$  is the maximum of  $EE$  if there is an equal probability of any land-use type being adjacent to a randomly chosen point in the landscape (O'Neill *et al.* 1988). Note that  $EE$  is positive and has the same form as the information index (Pielou 1975). According to O'Neill *et al.* (1988), contagion measures the extent to which landscape elements are aggregated or clumped; higher values of contagion may result from landscapes with a few large, contiguous patches, whereas lower values generally characterize landscapes with many small patches.

In this paper, we show that an error exists in the definition of  $EE$  (Eq. 2), which is a critical problem because it affects the sensitivity of  $D_2$  to changes in landscape configuration. We introduce an alternative contagion index that corrects this er-

ror and tends to distinguish between landscape composition and configuration.

### Estimates of $P_{ij}$

We define a landscape mosaic as the distribution of  $n$  patch types among  $N$  pixels. Two methods can be used to estimate  $P_{ij}$  based on the definition that  $P_{ij}$  is the probability of patch type  $i$  being adjacent to patch type  $j$ . First, we assume that  $P_{ij}$  is the conditional probability that, given a pixel is of patch type  $i$ , one of its neighboring pixels belongs to patch type  $j$ , i.e.:

$$P_{ij} = P_{j/i} = N_{ij}/N_i \quad (4)$$

where  $P_{j/i}$  is the conditional probability,  $N_{ij}$  the number of adjacencies (joins) between pixels of patch types  $i$  and  $j$ , and  $N_i$  the total number of adjacencies between pixels of patch type  $i$  and all patch types (including patch  $i$  itself). Thus, we have

$$\sum_{j=1}^n N_{ij} = N_i$$

and, therefore

$$\sum_{j=1}^n P_{ij} = 1 \quad (5)$$

O'Neill *et al.* (1988) did not provide information on how to estimate  $P_{ij}$ , but an equation similar to Eq. 4 was used to estimate  $P_{ij}$  (Robert O'Neill, pers. comm.).

Second, we can assume that  $P_{ij}$  is the probability that two randomly chosen adjacent pixels belong to type  $i$  and  $j$ , respectively, i.e.:

$$p_{ij} = P_i P_{j/i} \quad (6)$$

where  $P_i$  is the probability that a randomly chosen pixel belongs to patch type  $i$  (estimated by the proportion of patch type  $i$ ), and  $P_{j/i}$  is the same as in Eq. 4. Here,

$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} = 1 \quad (7)$$

(see Appendix for details).  $P_{ij}$  is similar to the probability used by information indices, especially Shannon's conditional entropy (Shannon and Weaver 1949) and Pielou's index of mosaic's spatial diversity (Pielou 1975).

### The erroneous maximum term ( $EE_{\max}$ )

The critical error in the  $D$ , (Eq. 1) is that the maximum of  $EE$  does not equal  $EE_{\max}$  defined by Eq. 2. Here, we present two proofs: one being an intuitive argument based on the property of an information index, and the other being a more formal mathematical proof.

The *intuitive argument*. According to information theory,  $EE$  should have its greatest value when each  $P_{ij}$  is equal to a constant (e.g., Shannon and Weaver 1949; Pielou 1975). Therefore, we can assume that  $EE$  is maximum when  $P_{ij} = p$  (a constant). Substituting  $p$  for  $P_{ij}$  in Eq. 3, we have

$$EE_{\max} = - \sum_{i=1}^n \sum_{j=1}^n p \ln(p)$$

hence,

$$EE_{\max} = - n^2 [p \ln(p)] \quad (8)$$

The  $p$  value is dependent on how  $P_{ij}$  is estimated. First, we assume that  $P_{ij}$  is defined by Eq. 4. Based on the same property of information theory mentioned above, we assume that all patch types are of even proportion, that is,  $P_i = 1/n$  for all  $i$ , where  $P_i$  is the probability that a randomly chosen pixel belongs to type  $i$ . Thus, the necessary and sufficient condition for  $EE$  to be at its maximum value is

$$P_{ij} = p = 1/n \quad (9)$$

Substituting  $1/n$  for  $p$  in Eq. 8, we have

$$EE_{\max} = - n^2 [(1/n) \ln(1/n)]$$

hence,

$$\mathbb{E}E_s = n \ln(n) \quad (10)$$

Second, we assume that  $P_{ij}$  is defined by Eq. 6. Under the assumptions of the even proportion of each patch type and the completely random pattern,  $P_{j/i}$  is independent of  $P_i$  and equal to  $1/n$ . Now, the necessary and sufficient condition for  $EE$  to be at its maximum value is

$$P_{ij} = p = 1/n^2 \quad (11)$$

Substituting  $1/n^2$  for  $p$  in Eq. 8, we have

$$\mathbb{E}E_m = -n^2[(1/n^2) \ln(1/n^2)]$$

and, hence

$$\mathbb{E}E_s = 2 \ln(n) \quad (12)$$

Eq. 10 differs from Eq. 2 by a factor of 2, and Eq. 12 by a factor of  $n$ . Thus, we conclude that  $\mathbb{E}E_s$  defined in Eq. 2 is not the maximum of  $EE$ . Notice that our conclusion is not affected by the definition of  $P_{ij}$  in Eq. 3. The above reasoning is easily understandable, but lacks generality due to requirement of many assumptions.

*The mathematical proof.* Assuming that  $P_{ij}$  is expressed by Eq. 6, we give a formal mathematical proof that the maximum of  $EE$  is given by Eq. 12. Jensen's inequality is expressed as (Hardy *et al.* 1952):

$$\sum_{i=1}^n q_i f(x_i) \geq f\left(\sum_{i=1}^n q_i x_i\right) \quad (13)$$

where  $f$  is a convex, monotonic function in the range  $[a, b]$ ,  $x_i$  can be any value in the range  $[a, b]$ , and  $q_i$  is positive with the condition

$$\sum_{i=1}^n q_i = 1$$

In Eq. 13, the equality holds only if the  $x_i$ 's are equal for all  $i$ . Based upon the algebraic fact that any single summations can be re-arranged and then

expressed as double summation (Li 1966), Jensen's inequality can be expanded and rewritten as:

$$\sum_{i=1}^n \sum_{j=1}^n q_{ij} f(x_{ij}) \geq f\left(\sum_{i=1}^n \sum_{j=1}^n q_{ij} x_{ij}\right) \quad (14)$$

given

$$\sum_{i=1}^n \sum_{j=1}^n q_{ij} = 1 \quad (15)$$

Without any loss of generality, we can define

$$x_{ij} = P_{ij} \quad (16)$$

$$f(x_{ij}) = x_{ij} \ln(x_{ij}) \quad (17)$$

and

$$q_{ij} = 1/n^2 \quad (18)$$

Notice that the function in Eq. 17 is a convex, monotonic function in the range  $(0, 1]$ . Substituting Eq. 7 and Eqs. 15–18 for the corresponding terms in Eq. 14 (i.e., the expanded Jensen's inequality), we have

$$-\sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \leq 2 \ln(n) \quad (19)$$

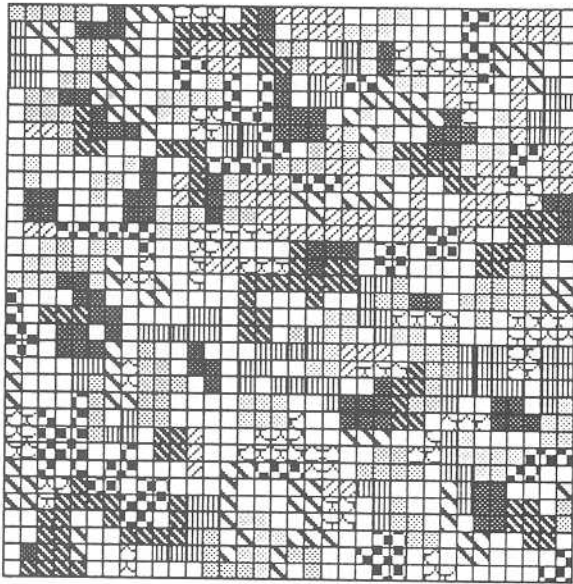
(see Appendix for details). Notice that the left side term in Eq. 19 is  $EE$ . Therefore, we come to the same conclusion that the maximum value of  $EE$  is not expressed by Eq. 2, but by Eq. 12, provided that  $P_{ij}$  is given in Eq. 6. Similar proof can be given when we assume that  $P_{ij}$  is given in Eq. 4, and the result should confirm Eq. 10.

Our conclusion is supported by the work of Gatrell (1977). He proposes a complexity (or uncertainty) index,  $H_{\text{map}}$ :

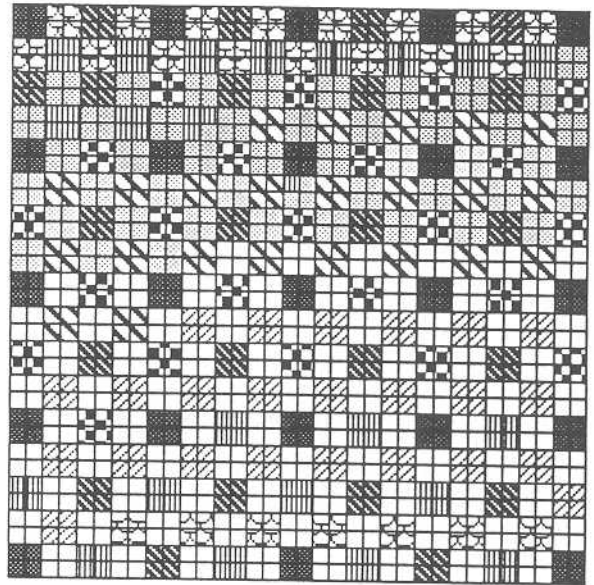
$$H_{\text{map}} = -\sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \quad (20)$$

where  $n$  is the number of colors, and  $P_{ij}$  is the joint probability of colors (or patches)  $i$  and  $j$  (Gatrell 1977). Gatrell's  $P_{ij}$  in Eq. 20 is similar to  $P_{ij}$  in

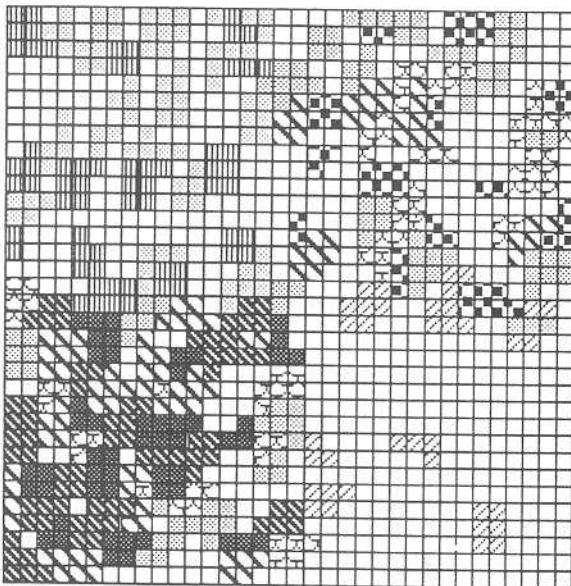
## a. Random



## b. Uniform



## c. Aggregated



## LEGENDS

-  Patch type 1 (Matrix)
-  Patch type 2
-  Patch type 3
-  Patch type 4
-  Patch type 5
-  Patch type 6
-  Patch type 7
-  Patch type 8
-  Patch type 9
-  Patch type 10

Fig. 1. Examples of the simulated landscape maps with 10 patch types for spatial configurations of (a) random (a random distribution of patches with randomly determined patch shapes, i.e. restricted random), (b) uniform (a uniform distribution of patches with strictly square patch shape), and (c) aggregated (an aggregated distribution of patches with regular patch shape). Note that a simulated map with 9 patch types can be obtained by deleting all pixels of patch type 10 and reassigning them to the matrix (i.e., patch type 1). Please refer to Li (1989) for more details about the simulation models.

Eq. 6, and  $H_{\text{map}}$  is similar to EE in Eq. 3. He has shown, using binary maps as examples, that the maximum of  $H_{\text{map}}$  is constrained by the marginal probabilities and differs with different sets of  $P_i$ 's. He suggests that  $P_{ij}$  can be calculated by assuming that  $P_i$ 's are statistically independent, that is,  $P_{ij} = P_i P_j$ . He also points out without proof that the maximum of  $H_{\text{map}}$  is "2 ln(n)", provided that  $P_i = 1/n$ .

### An alternative contagion index

We propose a new contagion index:

$$RC = 1 - EE/EE_{\text{max}} \quad (21)$$

where RC is the relative contagion, EE is defined in Eq. 3, and  $EE_{\text{max}}$  is the maximum of EE defined by Eqs. 10 or 12, depending on the definition of  $P_{ij}$ . Eq. 21 has been used by authors who originally proposed D, (Robert O'Neill, pers. comm.). If  $P_{ij}$  is given by Eq. 4, then

$$RC1 = 1 + \sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) / n \ln(n) \quad (22)$$

If  $P_{ij}$  is given by Eq. 6, then

$$RC2 = 1 + \sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) / 2 \ln(n) \quad (23)$$

All terms on the right side of Eqs. 21, 22, and 23 are the same as defined above. Note that the correct maximum term ( $EE_{\text{max}}$ ) is used in the new relative contagion index given by Eq. 22 or 23 and that Eq. 1 has no upper limit, whereas contagion as defined in Eq. 21 is a relative index, ranging from 0 to 1. RC is in essence a function of an evenness index,  $EE/EE_{\text{max}}$ ; thus, it has all the advantages and disadvantages of an evenness index (see Hurlbert 1971).

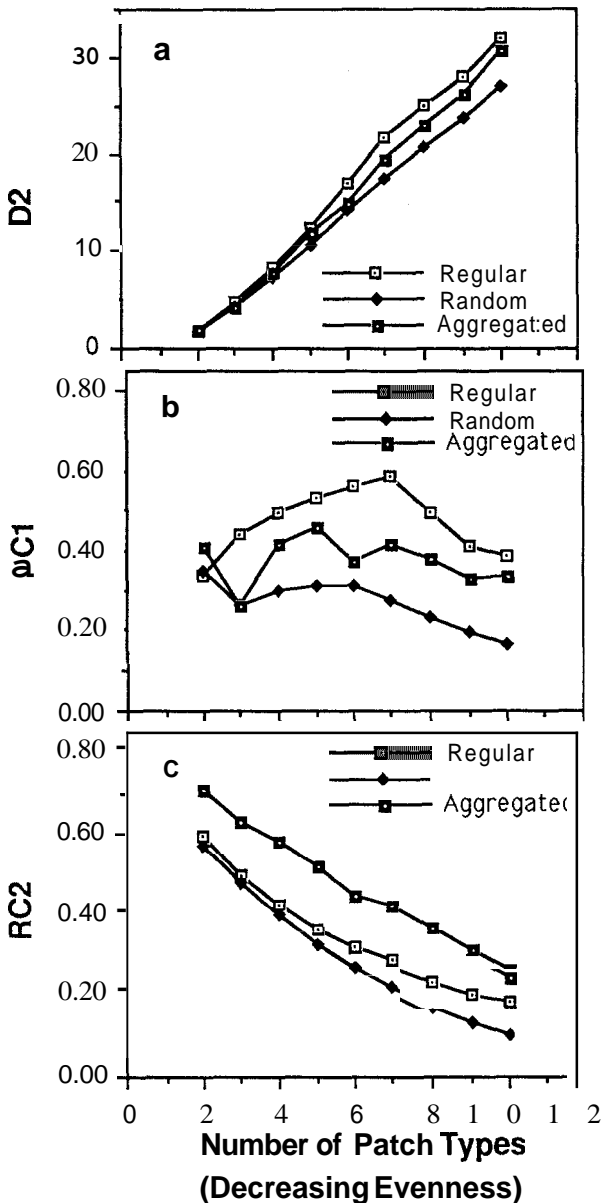
EE (Eq. 3) can be used alone as an index of spatial pattern (Robert O'Neill, pers. comm.). There is a strict, linear, inverse relationship between EE and RC; thus, EE has all the spatial information that

RC has. The only other difference between the two is that RC is an evenness index, and is scaled to compare landscapes with different values of n (Turner *et al.* 1989).

### Evaluation of contagion indices using synthetic landscapes

To evaluate and compare the performance of RC and  $D_2$ , we generated a series of landscape maps with different components of composition and configuration. The maps were composed of the landscape matrix (empty = patch type 1) and one or more other patch types (patch types 2, 3, and up to 10); see Fig. 1). Following the algorithm described in Li (1989; Li *et al.* 1992), for each of three configurations ("random", "uniform", and "aggregated") nine maps were created by introducing one patch type to the landscape matrix at a time; each patch type (except patch type 1) covered 8% of the total area (see Fig. 1). For example, the first map had two patch types, 1 and 2 (covering 92% and 8% respectively); the second three types, 1, 2, and 3 (covering 84%, 8%, and 8% respectively); etc. Values of the three contagion indices as a function of different components of landscape composition and configuration is illustrated in Fig. 2.

$D_2$  fails to distinguish differences in the three spatial configurations (Fig. 2a). In theory, contagion has a spatial component in EE (Eq. 3), thus, changes in contiguity are reflected only by changes in EE. The failure of  $D_2$  is because the inflated  $EE_{\text{max}}$  (a function of the number of patch types) greatly outweighs EE (a function of the spatial configuration). This is also evidenced by the linear increase of  $D_2$  with increase of the number of patch types. This increase of  $D_2$  contradicts the definition of the contagion index, i.e., contagion should decrease as many small patches of different types are introduced to the landscape and as the landscape matrix is broken into many smaller pieces. These results suggest that the  $D_2$  is not a good index of spatial configuration because it primarily reflects landscape composition (e.g., number of patch types) and is insensitive to spatial configuration.



**Fig. 2.** Relationships between the contagion indices and the two controlled variables: spatial pattern and number of patch types.  $D_2$ , RCI, and RC1 are the original contagion index (Eq. 1) and the new, relative contagion indices (Eqs. 22 and 23), respectively. See the text for details. Notice that there is a decreasing gradient of evenness of the proportions of patch types (from 0.29 to 0.09) along the x axis.

RC1 varies with different spatial configurations (Fig. 2b) but does not change much with the increase in the number of patch types, due to the method used to estimate  $P_{ij}$  (Eq. 4). RC1 is a good

index if the objective is to compare spatial configuration of landscapes regardless of how many patch types they have; RC1 may not be appropriate to characterize the overall contiguity of landscapes because contagion should decrease with increase in the number of patch types, as discussed above. In contrast, RC2 distinguishes the three different spatial configurations and is sensitive to the increase in the number of patch types (Fig. 2c). Hence, RC2 is the only index that quantifies both components of landscape diversity.

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**Appendix**

1. Derivation of Eq. 7: 
$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} = 1$$

Assume that  $A_i$  is the total area (or total number of pixels) of patch type  $i$  and  $A$  the total area of the landscape. Then, the probability that a randomly chosen pixel belongs to patch type  $i$  can be estimated by the proportion of patch type  $i$ :

$$P_i = A_i / A \tag{A.1}$$

Substituting Eqs. A.1 and 4 for  $P_i$  and  $P_{j/i}$  in Eq. 6, we have

$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} = \sum_{i=1}^n \sum_{j=1}^n P_i P_{j/i}$$

hence,

$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} = \sum_{i=1}^n \sum_{j=1}^n (A_i / A) (N_{ij} / N_i)$$

and therefore

$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} = \sum_{i=1}^n (A_i / A) \sum_{j=1}^n (N_{ij} / N_i) \tag{A.2}$$

Note that, according to our definitions above, we have

$$\sum_{i=1}^n A_i = A \tag{A.3}$$

and

$$\sum_{j=1}^n N_{ij} = N_i \tag{A.4}$$

Substituting Eqs. A.3 and A.4 in Eq. A.2, we have

$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} = 1.$$

2. Derivation of Eq. 19: 
$$-\sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \leq 2 \ln(n)$$

We start with reconstructing the left and right sides of Jensen's inequality (Eq. 14) separately. Substituting Eqs. 16, 17, and 18 for the left side terms in Eq. 14, we obtain

$$\sum_{i=1}^n \sum_{j=1}^n q_{ij} f(x_{ij}) = \sum_{i=1}^n \sum_{j=1}^n (1/n^2) P_{ij} \ln(P_{ij})$$

hence,

$$\sum_{i=1}^n \sum_{j=1}^n q_{ij} f(x_{ij}) = (1/n^2) \sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \tag{A.5}$$

Substituting Eqs. 11, 15 and 17 for the right side terms in Eq. 14, we have

$$f\left(\sum_{i=1}^n \sum_{j=1}^n q_{ij} x_{ij}\right) = \left\{ \sum_{i=1}^n \sum_{j=1}^n (1/n^2) P_{ij} \right\} \left\{ \ln \left[ \sum_{i=1}^n \sum_{j=1}^n (1/n^2) P_{ij} \right] \right\}$$

hence,

$$f\left(\sum_{i=1}^n \sum_{j=1}^n q_{ij} x_{ij}\right) = \left\{ (1/n^2) \sum_{i=1}^n \sum_{j=1}^n P_{ij} \right\} \left\{ \ln \left[ (1/n^2) \sum_{i=1}^n \sum_{j=1}^n P_{ij} \right] \right\}$$

and therefore

$$f\left(\sum_{i=1}^n \sum_{j=1}^n q_{ij} x_{ij}\right) = (1/n^2) \ln(1/n^2) \quad (\text{A.6})$$

Replacing both sides in Jensen's inequality (i.e., Eq. 14) with the results from Eqs. A.5 and A.6, we obtain

$$(1/n^2) \sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \geq (1/n^2) \ln(1/n^2)$$

hence,

$$\sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \geq \ln(1/n^2)$$

and therefore

$$-\sum_{i=1}^n \sum_{j=1}^n P_{ij} \ln(P_{ij}) \leq 2 \ln(n).$$