

# NeutralLandscapes.jl: a library for efficient generation of neutral landscapes with temporal change

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Soon to be a paper, maybe. TK authors, MKB,VB,RS,TP

# 1 Introduction

2 Neutral landscapes are increasingly used in ecological and evolutionary studies to provide a null  
3 expectation spatial variation of a given measurement. Originally developed to simulate the spatially  
4 autocorrelated data (Gardner *et al.* 1987; Milne 1992), they have seen use in a wide range of disciplines:  
5 from landscape genetics (Storfer *et al.* 2007), to landscape and spatial ecology (Tinker *et al.* 2004; Rempel  
6 & Fortin 2013), and biogeography (Albert *et al.* 2017).

7 We present `NeutralLandscapes.jl`, a package in Julia for neutral landscapes. The two primary packages  
8 used to simulate neutral landscapes are NLMR in (the R language) (Sciaini *et al.* 2018) and NLMpy (in Python;  
9 Etherington *et al.* 2015). Here we demonstrate that `NeutralLandscapes.jl`, depending on the method, is  
10 orders of magnitude faster than previous neutral landscape packages.

11 As biodiversity science becomes increasingly concerned with temporal change and its consequences, it's  
12 clear there is a gap in methodology in generating neutral landscapes that change over time. In addition we  
13 present a novel method for generating landscape change with prescribed levels of spatial and temporal  
14 autocorrelation, which is implemented in `NeutralLandscapes.jl`

## 15 Software Overview

16 This software can generate neutral landscapes using several methods, enables masking and works with  
17 other Julia packages.

18 [fig. 1](#) shows a replica of Figure 1 from Etherington *et al.* (2015), which shows the capacity of the library to  
19 generate different types of neutral landscapes, and then apply masks and categorical classification to them.

20 [Figure 1 about here.]

## 21 Interoperability

22 Ease of use with other Julia packages

23 Mask of neutral variable masked across Quebec in 3 lines.

```

24 using NeutralLandscapes
25 using SimpleSDMLayers
26
27 quebec = SimpleSDMPredictor(WorldClim, BioClim; left=-90., right=-50., top=75., bottom=40.)
28 qcmask = fill(true, size(quebec))
29 qcmask[findall(isnothing, quebec.grid)] .= false
30
31 pltsettings = (cbar=:none, frame=:box)
32
33 plot(
34     heatmap(rand(MidpointDisplacement(0.8), size(layer), mask=qcmask); pltsettings),
35     heatmap(rand(PlanarGradient(), size(layer), mask=qcmask); pltsettings),
36     heatmap(rand(PerlinNoise((4,4)), size(layer), mask=qcmask); pltsettings),
37     heatmap(rand(NearestNeighborCluster(0.5), size(layer), mask=qcmask); pltsettings),
38     dpi=400
39 )

```

[Figure 2 about here.]

## 41 **Benchmark comparison to nlmpy and NLMR**

42 It's fast. As the scale and resolution of raster data increases, neutral models must be able to scale to match  
43 those data dimensions.

[Figure 3 about here.]

## 45 **Generating dynamic neutral landscapes**

46 We implement methods for generating change that are temporally autocorrelated, spatially autocorrelated,  
47 or both.

$$48 \quad M_t = M_{t-1} + f(M(t-1))$$

## 49 **Models of change**

### 50 **Directional**

### 51 **Temporally autocorrelation**

52  $r$ : rate,  $v$ : variability,  $U$  matrix of draws from standard Normal(0, 1)

53 
$$f_T(M_{ij}) = r + vU_{ij}$$

### 54 **Spatial autocorrelation**

55  $r$ : rate,  $v$ : variability,  $[Z(\delta)]_{ij}$ : the  $(i, j)$  entry of the zscore of the  $\delta$  matrix

56 
$$f_S(M_{ij}) = r + v \cdot [Z(\delta)]_{ij}$$

### 57 **Spatiotemporal autocorrelation**

58 
$$f_{ST}(M_{ij}) = r + v \cdot [Z(\delta)]_{ij}$$

## 59 **Rescaling to mimic real data**

## 60 **Discussion**

## 61 **References**

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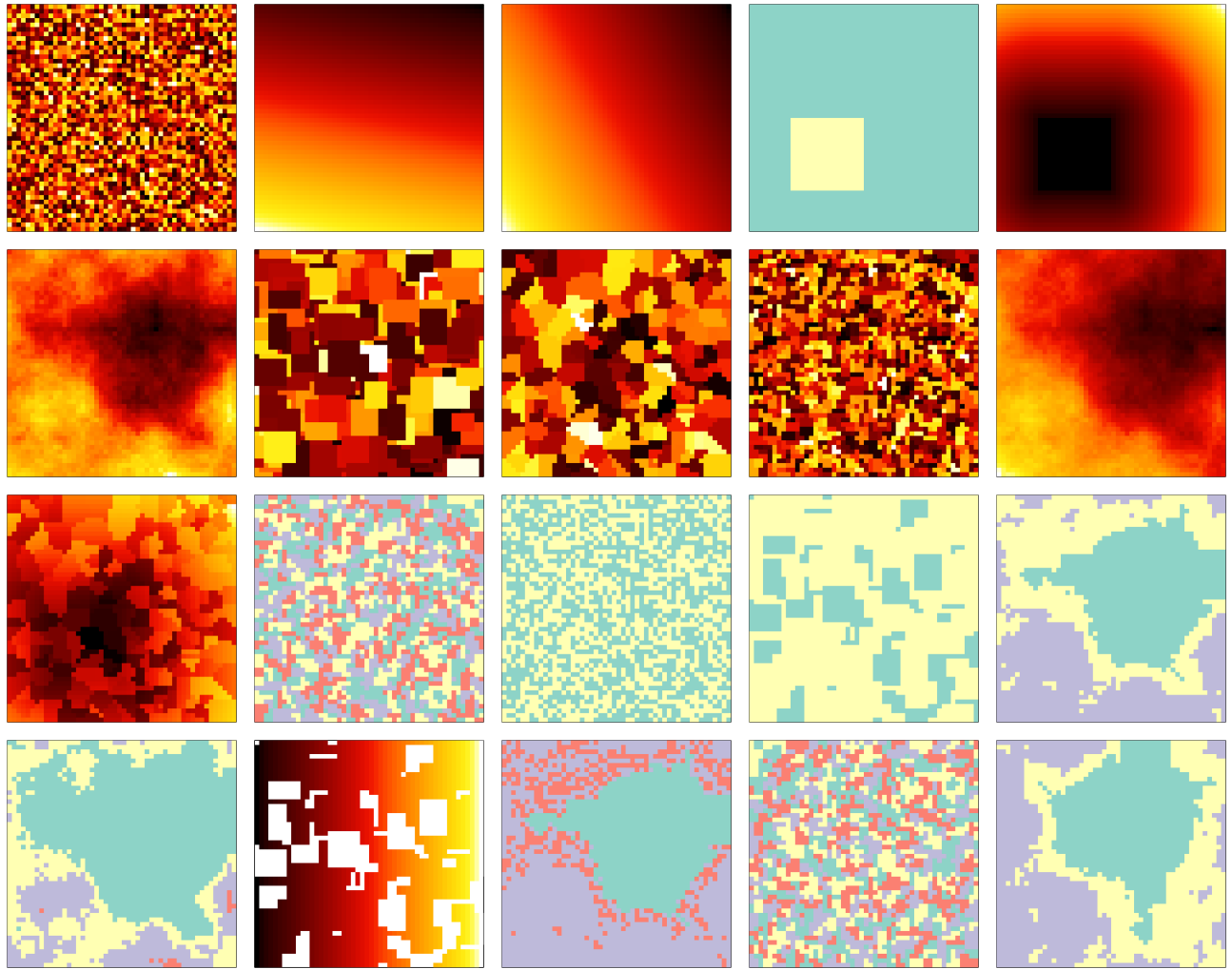


Figure 1: Recreation of the figure in n1mpy paper and the source, supplied in less than 40 lines of code.

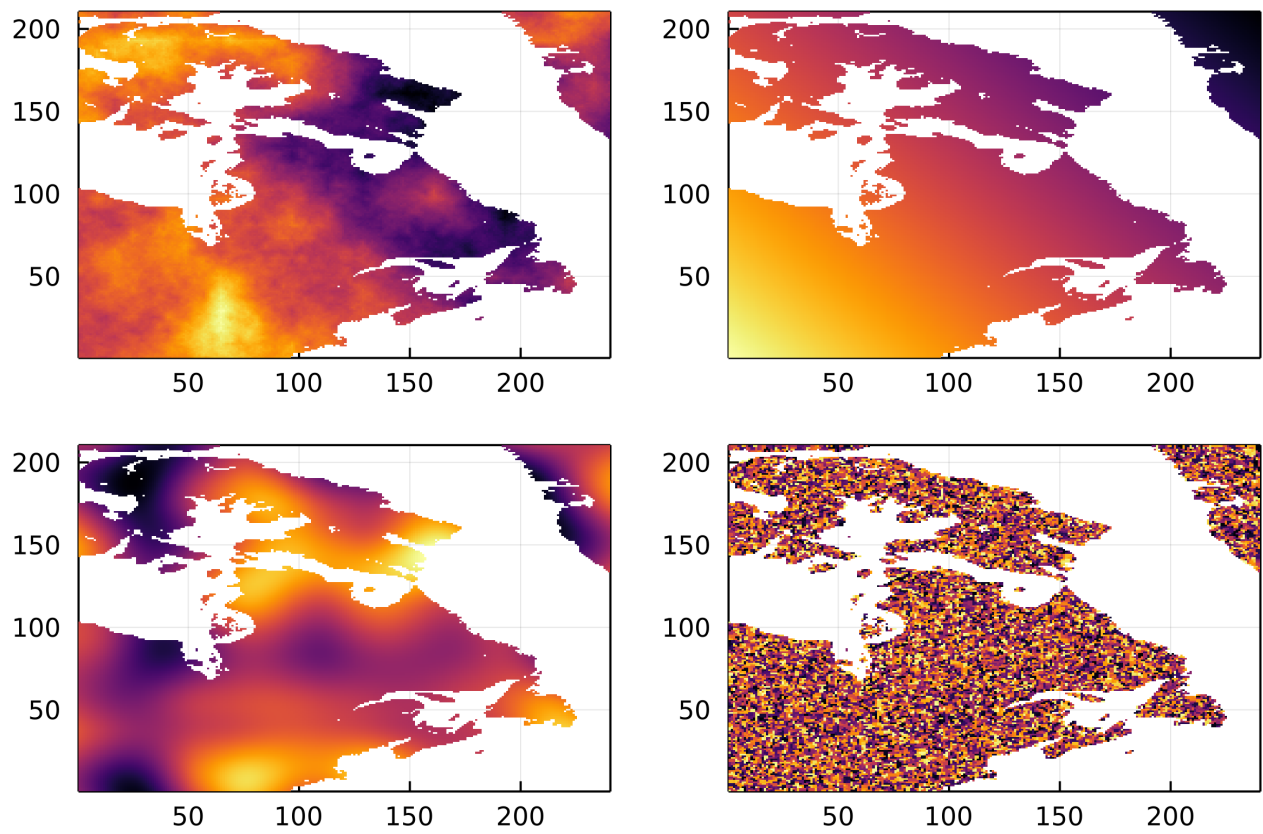


Figure 2: todo



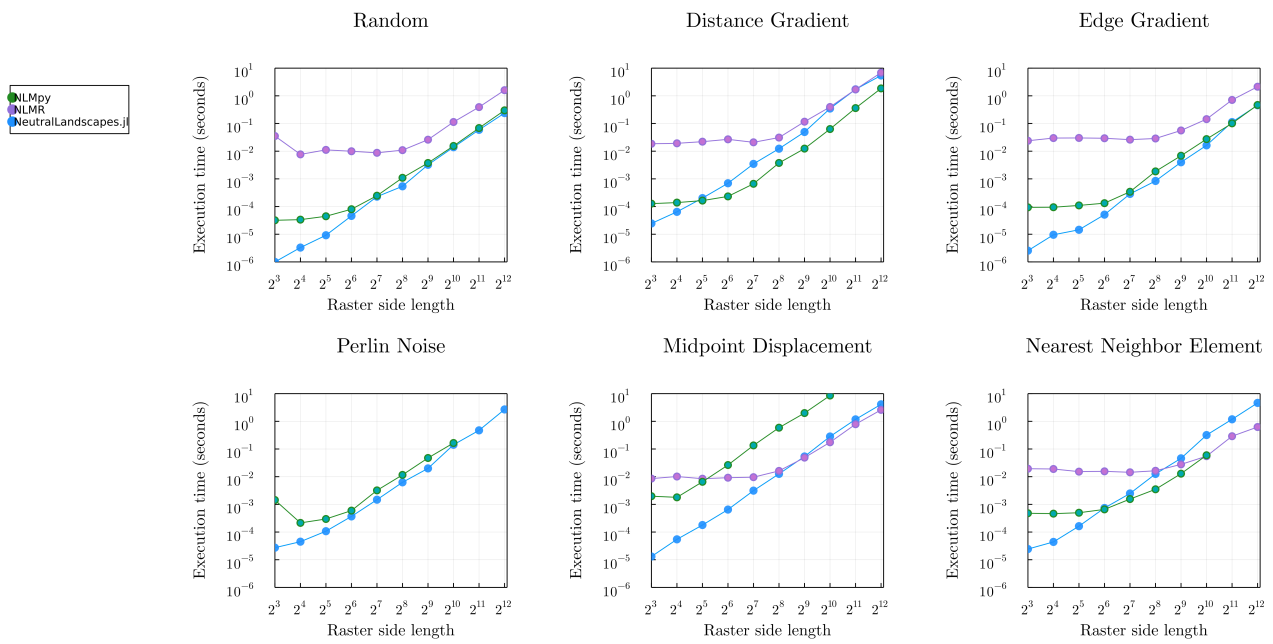


Figure 3: todo