

# 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

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## Introduction

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

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

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

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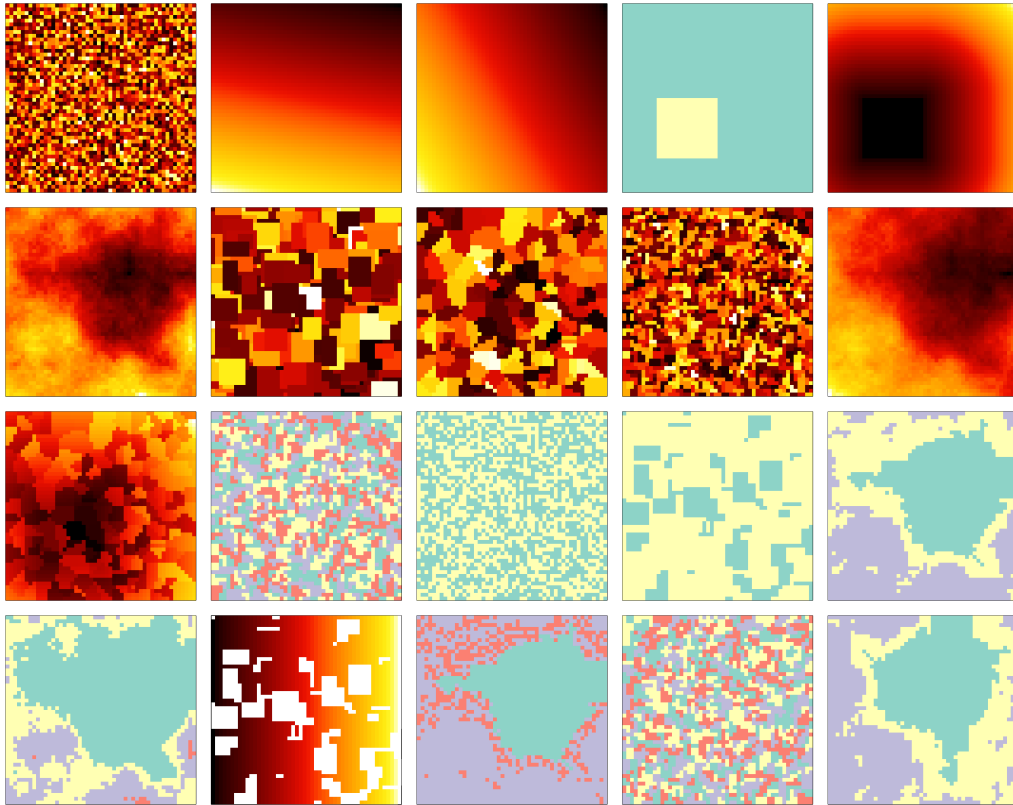
## Software Overview

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

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

### 2.1. Interoperability Ease of use with other Julia packages

Mask of neutral variable masked across Quebec in 3 lines.



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

```
using NeutralLandscapes
using SimpleSDMLayers

quebec = SimpleSDMPredictor(WorldClim, BioClim; left=-90., right=-50., top=75., bottom=40.)
qcmask = fill(true, size(quebec))
qcmask[findall(isnothing, quebec.grid)] .= false

pltsettings = (cbar=:none, frame=:box)

plot(
    heatmap(rand(MidpointDisplacement(0.8), size(layer), mask=qcmask); pltsettings),
    heatmap(rand(PlanarGradient(), size(layer), mask=qcmask); pltsettings),
    heatmap(rand(PerlinNoise((4,4)), size(layer), mask=qcmask); pltsettings),
    heatmap(rand(NearestNeighborCluster(0.5), size(layer), mask=qcmask); pltsettings),
    dpi=400
)
```

### 3 ---

#### Benchmark comparison to `n1mpy` and NLMR

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

### 4 ---

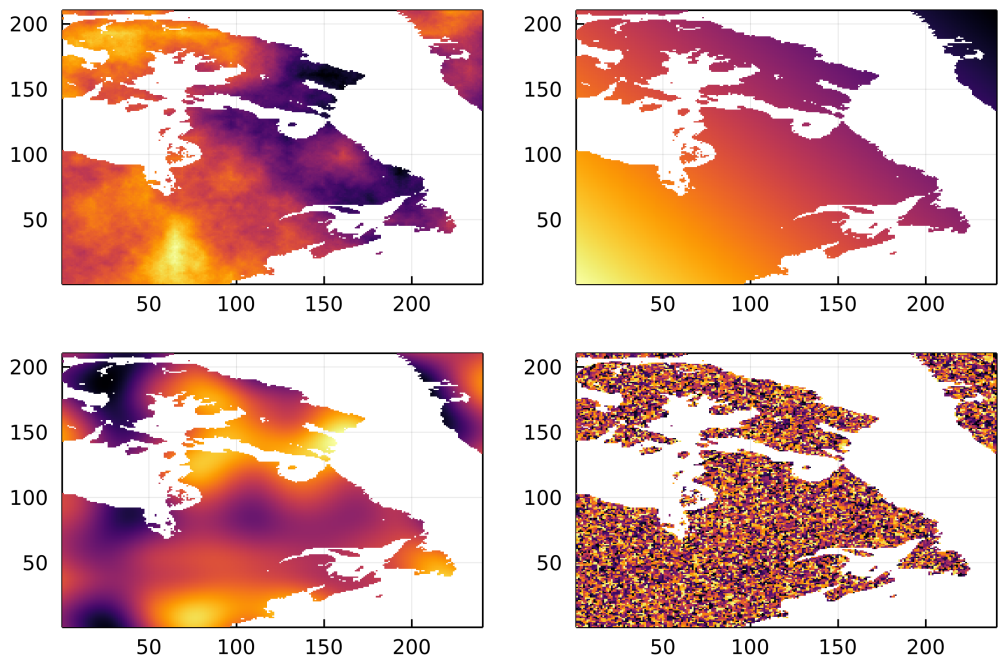


Figure 2 todo

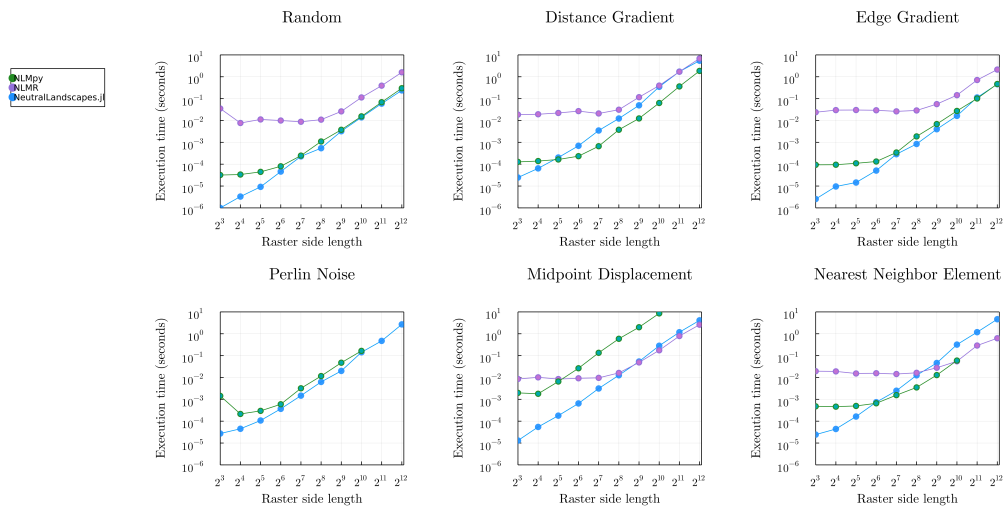


Figure 3 todo

## Generating dynamic neutral landscapes

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

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

### 4.1. Models of change

#### 4.1.1 Directional

**4.1.2 Temporally autocorrelation**  $r$ : rate,  $v$ : variability,  $U$  matrix of draws from standard Normal(0, 1)

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

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

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

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

### 4.2. Rescaling to mimic real data

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## Discussion

## References

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