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

Introduction

- 2 Neutral landscapes are increasingly used in ecological and evolutionary studies to provide a null
- expectation of the variance of a given metric over space.
- 4 Wide range of disciplines: from landscape genetics [], to spatial ecology [], and biogeography [].
- 5 As biodiversity science becomes increasingly concerned with temporal change and its consequences, its
- 6 clear there is a gap generating neutral landscapes that change over time. In this ms we present how
- NeutralLandscapes.jl is orders of magnitudes faster than packages nlmpy (in python) or NLMR (in R). In
- addition we then present a novel method for generating landscape change with prescribed levels of spatial
- 9 and temporal autocorrelation.

Software Overview

- 11 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 (**nlmpycite?**), which shows the capacity of the library to generate
- different types of neutral landscapes, and then apply masks and categorical classification to them.
- Table of methods.
- 16 In

17

[Figure 1 about here.]

18 What methods have been called different things but are actually the same thing?

9 Interoperability

- 20 Ease of use with other julia packages
- 21 Mask of neutral variable masked across quebec in 3 lines.

22 Benchmark comparison to nlmpy and NLMR

- 23 It's fast. As the scale and resolution of raster data increases, neutral models must be able to scale to match
- 24 those data dimensions. Here we provide two benchmark tests. First a comparison of the speed variety of
- 25 methods from each NeutralLandscapes.jl, NLMR, and nlmpy. Second we compare these performance of
- each of these software packages as rasters become larger. We show that Julia even outperforms the NLMR
- via C++ implemention of a particularly slow neutral landscape method (midpoint displacement).
- Fig 2: Benchmark comparison of selected methods in each of the three languages
- 29 In fig 2 we should a selection of neutral landscape generators (random, edge gradient, perlin noise,
- 30 distance-gradient)

31 MPD comparison

- Why use this particular generator as the comparison? It's slow. So slow that NLMR implements it in C++.
- (NLMR implements both MPD, neighbor, randrect, and random neighboorhood in c++). Still these three
- ³⁴ algorithms, which cosinsts of 3/16 of NLMR's alg implementations, constitute 33% of its codebase.
- In this section we show our implementation of MPD is faster than NLMR's C++ MPD across all scales, up
- to the asyptotic limit imposed by the $O(n^2)$ scaling of the raster

[Figure 2 about here.]

38 Generating dynamic neutral landscapes

- ³⁹ We implement methods for generating change that are temporally autocorrelated, spatially autocorrelated,
- 40 or both.

37

41 $M_t = f(M_{t-1})$

Discussion

References

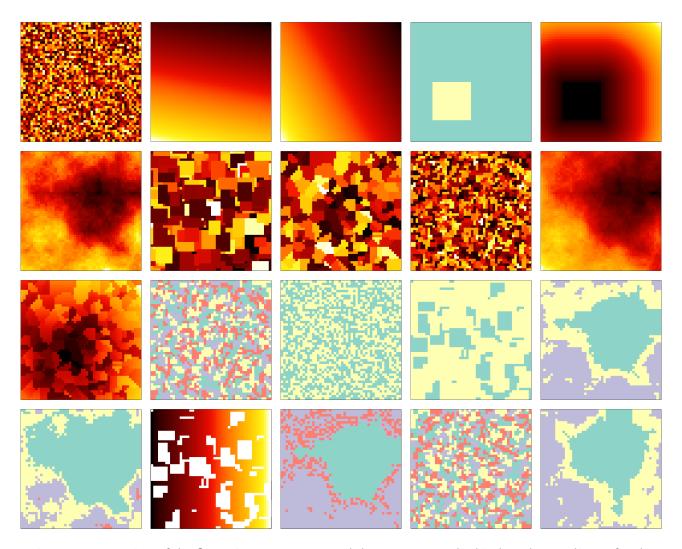


Figure 1: Recreation of the figure in nlmpy paper and the source, supplied in less than 40 lines of code.

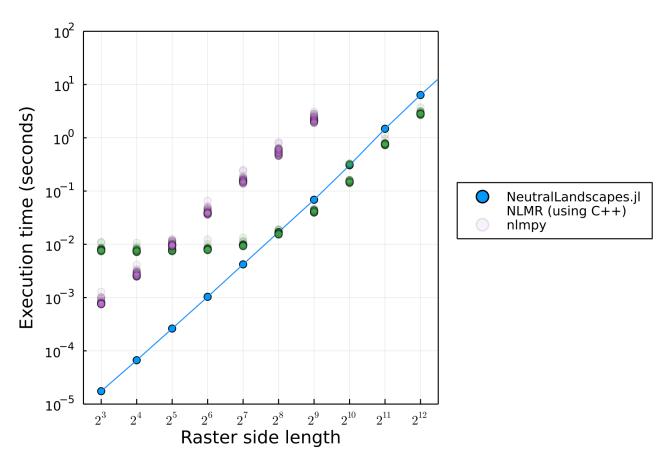


Figure 2: Comparison of speed of generating a midpoint displacement neutral landscape (y-axis) against raster size (measured as length of the size of a square raster, x-axis)