

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 of the variance of a given metric over space.

4 Wide range of disciplines: landscape genetics to 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
7 `NeutralLandscapes.jl` is orders of magnitudes faster than packages `nlmpy` (in python) or `NLMR` (in R). We
8 then present a novel method for generating landscape change with prescribed levels of spatial and
9 temporal autocorrelation, and demonstrate that it works

10 Software Overview

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

13 [fig. 1](#) shows a replica of Figure 1 from ([nlmpycite?](#))

14 Table of methods.

15 [Figure 1 about here.]

16 using `NeutralLandscapes`, `Plots`

17 `siz = 50, 50`

18

19 `Fig1a = rand(NoGradient(), siz) # Random NLM`

20 `Fig1b = rand(PlanarGradient(), siz) # Planar gradient NLM`

21 `Fig1c = rand(EdgeGradient(), siz) # Edge gradient NLM`

22 `Fig1d = falses(siz)`

23 `Fig1d[10:25, 10:25] .= true # Mask example`

24 `Fig1e = rand(DistanceGradient(findall(vec(Fig1d)))), siz) # Mask example`

25 `Fig1f = rand(MidpointDisplacement(0.75), siz) # Mask example`

```

26 Fig1g = rand(RectangularCluster(4, 8), siz)
27 Fig1h = rand(NearestNeighborElement(200), siz)
28 Fig1i = rand(NearestNeighborCluster(0.4), siz)
29 Fig1j = blend([Fig1f, Fig1c])
30 Fig1k = blend(Fig1h, Fig1e, 1.5)
31 Fig1l = classify(Fig1i, ones(4))
32 Fig1m = classify(Fig1a, [1-0.5, 0.5])
33 Fig1n = classify(Fig1g, [1-0.75, 0.75])
34 Fig1o = classify(Fig1f, ones(3))
35 Fig1p = classify(Fig1f, ones(3), Fig1d)
36 Fig1q = rand(PlanarGradient(90), siz, mask = Fig1n .== 2) #TODO mask as keyword + should mask be matrix or
37 Fig1r = ifelse.(Fig1o .== 2, Fig1m .+ 2, Fig1o)
38 Fig1s = rotr90(Fig1l)
39 Fig1t = Fig1o'
40
41 class = cgrad(:Set3_4, 4, categorical = true)
42 c2, c3, c4 = class[1:2], class[1:3], class[1:4]
43
44 gr(color = :fire, ticks = false, framestyle = :box, dpi=500, colorbar = false)
45 plot(
46     heatmap(Fig1a),      heatmap(Fig1b),      heatmap(Fig1c),      heatmap(Fig1d, c = c2), heatmap(Fig1e),
47     heatmap(Fig1f),      heatmap(Fig1g),      heatmap(Fig1h),      heatmap(Fig1i),      heatmap(Fig1j),
48     heatmap(Fig1k),      heatmap(Fig1l, c = c4), heatmap(Fig1m, c = c2), heatmap(Fig1n, c = c2), heatmap(Fig1o),
49     heatmap(Fig1p, c = c4), heatmap(Fig1q),      heatmap(Fig1r, c = c4), heatmap(Fig1s, c = c4), heatmap(Fig1t),
50     layout = (4,5), size = (1600, 1270)
51 )
52
53 savefig("./figures/figure1.png", )

```

54 **Interoperability**

55 Ease of use with other julia packages

56 Mask of neutral variable masked across quebec in 3 lines.

57 **Benchmark comparison to nlmpy and NLMR**

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

60 Here we provide two benchmark tests.

61 First a comparison of the speed variety of methods from each `NeutralLandscapes.jl`, NLMR, and nlmpy.

62 Second we compare these performance of each of these software packages as rasters become larger. We
63 show that Julia even outperforms the NLMR via C++ implementation of a particularly slow neutral
64 landscape method (midpoint displacement).

65 **Fig 2:** Benchmark comparison of selected methods in each of the three languages

66 **Fig 3** scale comparison

67 [Figure 2 about here.]

68 **Generating dynamic neutral landscapes**

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

71
$$M_t = f(M_{t-1})$$

72 **Discussion**

73 **References**

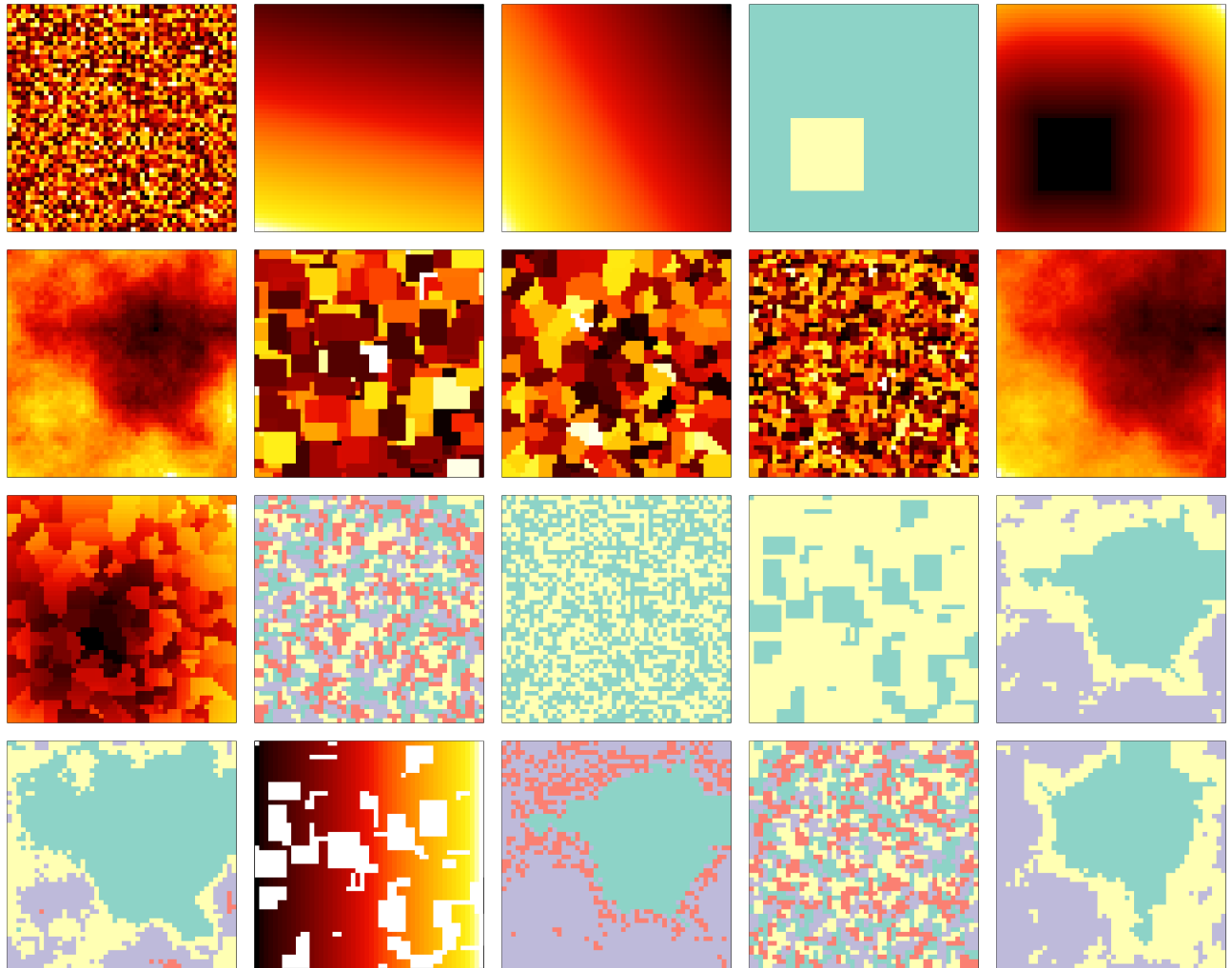


Figure 1: Recreation of the figure in nlmpy paper and the source

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)