Template

Joshua T. Vogelstein*1,2

¹Department of Biomedical Engineering, Institute for Computational Medicine, Kavli Neuroscience Discovery Institute, Johns Hopkins University

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The bibliography file has a relatively recent copy of all neurodata pubs.

This is a section

The quick brown fox jumps over the lazy dog Eq. (1).

The quick brown fox jumps over the lazy dog [1–8].

The quick brown fox jumps over the lazy dog Figure 1.

The quick brown fox jumps over the lazy dog. Amit and Geman [1]

THE QUICK BROWN FOX JUMPS OVER THE LAZY DOG.

The guick brown fox jumps over the lazy dog.

The quick [brown | chartreuse] fox jumps over the lazy [ass] dog

Aligned equation:

$$e^{i\pi} - 1 = 0, (1)$$

$$\chi = V - E + F \tag{2}$$

Add cats

Enumerate:

- 1. The quick brown fox jumps over the lazy dog
- 2. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog.

Itemize:

- The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog.
- The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog.

Description:

The The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog.

Quick brown fox jumps over the lazy dog. The quick brown fox jumps over the lazy dog.

²University of Something Else

^{*}jovo@jhu.edu

Table 1: Table.

	Metrics Sub Phase 1 Phase										
		1									
ľ	The	quick	brown	fox							

Table 2: The median sample size for each method to achieve power 85% at type 1 error level 0.05, grouped into monotone (type 1-5) and non-monotone simulations (type 6-19) for both one- and ten-dimensional settings, normalized by the number of samples required by Mgc. In other words, a 2.0 indicates that the method requires double the sample size to achieve 85% power relative to Mgc. Pearson, RV, and CCA all achieve the same performance, as do Spearman and Kendall. Mgc requires the fewest number of samples in all settings, and on average for high-dimensional settings, all other methods require about two to three times more samples than Mgc.

Dimensionality	Or	ne-Dimension	al	Ten-Dimensional						
Dependency Type	Monotone	Non-Mono	Average	Monotone	Non-Mono	Average				
MGC	1	1	1	1	1	1				
DCORR	1	2.6	2.2	1	3.2	2.6				
Mcorr	1	2.8	2.4	1	3.1	2.6				
Ннg	1.4	1	1.1	1.7	1.9	1.8				
Hsic	1.4	1.1	1.2	1.7	2.4	2.2				
MANTEL	1.4	1.8	1.7	3	1.6	1.9				
PEARSON / RV / CCA	1	>10	>10	8.0	>10	>10				
Spearman / Kendall	1	>10	>10	n/a	n/a	n/a				
MIC	2.4	2	2.1	n/a	n/a	n/a				

2 Tasks section

Task 1: Research

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Task 2: Development

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Task 3: Integration

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Task 4: Dissemination

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facilisis. Nunc elementum fermentum wisi. Aenean placerat. Ut imperdiet, enim sed gravida sollicitudin, felis odio placerat quam, ac pulvinar elit purus eget enim. Nunc vitae tortor. Proin tempus nibh sit amet nisl. Vivamus quis tortor vitae risus porta vehicula.

2.1 This is a subsection

The quick brown fox jumps over the lazy dog.

2.1.1 This is a subsubsection

The quick brown fox jumps over the lazy dog.

This is a paragraph The quick brown fox jumps over the lazy dog.

This is a subparagraph The quick brown fox jumps over the lazy dog.



Figure 1: Lion is awesome.

Algorithm 1 Mgc test statistic. This algorithm computes all local correlations, take the smoothed maximum, and reports the (k,l) pair that achieves it. For the smoothing step, it: (i) finds the largest connected region in the correlation map, such that each correlation is significant, i.e., larger than a certain threshold to avoid correlation inflation by sample noise, (ii) take the largest correlation in the region, (iii) if the region area is too small, or the smoothed maximum is no larger than the global correlation, the global correlation is used instead. The running time is $\mathcal{O}(n^2)$.

```
Input: A pair of distance matrices (A, B) \in \mathbb{R}^{n \times n} \times \mathbb{R}^{n \times n}.
Output: The Mgc statistic c^* \in \mathbb{R}, all local statistics \mathcal{C} \in \mathbb{R}^{n \times n}, and the corresponding local scale (k, l) \in \mathbb{N} \times \mathbb{N}.
 1: function MGCSAMPLESTAT(A, B)
        C = \mathsf{MGCALLLocal}(A, B)
                                                                                                                  ▷ All local correlations
         \tau = \mathsf{THRESHOLDING}(\mathcal{C})
                                                                         3:
         for i,j:=1,\ldots,n do r_{ij}\leftarrow\mathbb{I}(c^{ij}>\tau) end for
                                                                                          4:
         \mathcal{R} \leftarrow \{r_{ij} : i, j = 1, \dots, n\}
                                                                            ▷ binary map encoding scales with large correlation
 5:
         \mathcal{R} = \mathsf{CONNECTED}(\mathcal{R})
                                                                            ▷ largest connected component of the binary matrix
 6:
 7:
         c^* \leftarrow \mathcal{C}(n,n)
                                                                                              buse the global correlation by default
         k \leftarrow n, l \leftarrow n
 8:
        if \left(\sum_{i,j} r_{ij}\right) \ge 2n then [c^*, k, l] \leftarrow \max(\mathcal{C} \circ \mathcal{R})
                                                                      > proceed when the significant region is sufficiently large
                                                                       ⊳ find the smoothed maximum and the respective scale
10:
         end if
11:
12: end function
```

3 Second § so that we have more things in the TOC

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3.1 Second §§ so that we have more things in the TOC

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