

Template

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Contents

1 This is a section	1
2 Tasks section	2
1 Research	2
2 Development	2
3 Integration	2
4 Dissemination	2
2.1 This is a subsection	3
3 Second § so that we have more things in the TOC	4
3.1 Second §§ so that we have more things in the TOC	4

This is the abstract.

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

1 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 quick brown fox jumps over the lazy dog.

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

Aligned equation:

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

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

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.

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Add cats

Table 1: Table.

Metrics	Sub	Phase 1	Phase 2
1			
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 Dependency Type	One-Dimensional			Ten-Dimensional		
	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
HHG	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	0.8	>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

Nulla malesuada porttitor diam. Donec felis erat, congue non, volutpat at, tincidunt tristique, libero. Vivamus viverra fermentum felis. Donec nonummy pellentesque ante. Phasellus adipiscing semper elit. Proin fermentum massa ac quam. Sed diam turpis, molestie vitae, placerat a, molestie nec, leo. Maecenas lacinia. Nam ipsum ligula, eleifend at, accumsan nec, suscipit a, ipsum. Morbi blandit ligula feugiat magna. Nunc eleifend consequat lorem. Sed lacinia nulla vitae enim. Pellentesque tincidunt purus vel magna. Integer non enim. Praesent euismod nunc eu purus. Donec bibendum quam in tellus. Nullam cursus pulvinar lectus. Donec et mi. Nam vulputate metus eu enim. Vestibulum pellentesque felis eu massa.

Task 4: Dissemination

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facilis. 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.

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( $A, B$ )
2:    $\mathcal{C} = \text{MGCALLLOCAL}(A, B)$  ▷ All local correlations
3:    $\tau = \text{THRESHOLDING}(\mathcal{C})$  ▷ find a threshold to determine large local correlations
4:   for  $i, j := 1, \dots, n$  do  $r_{ij} \leftarrow \mathbb{I}(c^{ij} > \tau)$  end for ▷ identify all scales with large correlation
5:    $\mathcal{R} \leftarrow \{r_{ij} : i, j = 1, \dots, n\}$  ▷ binary map encoding scales with large correlation
6:    $\mathcal{R} = \text{CONNECTED}(\mathcal{R})$  ▷ largest connected component of the binary matrix
7:    $c^* \leftarrow \mathcal{C}(n, n)$  ▷ use the global correlation by default
8:    $k \leftarrow n, l \leftarrow n$ 
9:   if  $\left(\sum_{i,j} r_{ij}\right) \geq 2n$  then ▷ proceed when the significant region is sufficiently large
10:     $[c^*, k, l] \leftarrow \max(\mathcal{C} \circ \mathcal{R})$  ▷ find the smoothed maximum and the respective scale
11:   end if
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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References and Notes

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