Mathematical Science Research Institute Morphisms and Neural Codes

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Figure: Red areas indicate the place cells firing at a higher rate compared to the blue areas. Place fields are of four neurons in rat hippocampus. Data was provided by the Pastalkova lab. Giusti et al. 2015.

- Morphisms and open convexity
- Explore other properties

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Example

For the code $C = \{\emptyset, 1, 2\}$

1 is a codeword

123 is not a codeword

Neural Codes

Definition

Let $\mathcal{U} = \{U_1, \ldots, U_n\}$ where $U_1, \ldots, U_n \subseteq \mathbb{R}^d$. The *neural code* generated by \mathcal{U} is given by

$$C(\mathcal{U}) = \left\{ \sigma \subseteq [n] \mid \left(\bigcap_{i \in \sigma} U_i\right) \setminus \left(\bigcup_{j \in [n] \setminus \sigma} U_j\right) \neq \emptyset \right\}.$$

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Example

Consider $C = \{\emptyset, 12, 23, \underline{123}, \underline{234}\}.$

- Maximal codewords are 123 and 234.
- 12 is not maximal since it is contained in 123.

Let *C* be a neural code on *n* neurons and let $\sigma \subseteq [n]$. The *trunk* of σ in *C* is the set of all codewords of *C* that contain σ :

 $Tk_{\mathcal{C}}(\sigma)\coloneqq \{c\in \mathcal{C}\mid \sigma\subseteq c\}.$

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Definition

Let C and D be neural codes.

- A function $f: C \to \mathcal{D}$ is a *morphism* if for each trunk $Tk_{\mathcal{D}}(\sigma)$ in \mathcal{D} the preimage $f^{-1}(Tk_{\mathcal{D}}(\sigma))$ is also a trunk in C for all $\sigma \in \mathcal{D}$.
- A morphism f is an *isomorphism* if f is a bijection and f^{-1} is also a morphism.

$$Tk_{\mathcal{C}}(\sigma)\coloneqq \{c\in \mathcal{C}\mid \sigma\subseteq c\}.$$

Example

Consider $C = \{\emptyset, 1, 3, 12\}$ and $\mathcal{D} = \{\emptyset, 1, 2\}$. Define the function $f : C \to \mathcal{D}$ by:

 $f(\emptyset) = \emptyset$ f(1) = 1 f(3) = 2 f(12) = 1.

The trunks of $\mathcal D$ are

 $Tk_{\mathcal{D}}(\varnothing) = \mathcal{D} \qquad Tk_{\mathcal{D}}(1) = \{1\} \qquad Tk_{\mathcal{D}}(2) = \{2\} \qquad Tk_{\mathcal{D}}(12) = \varnothing.$

The pre-images of the trunks of \mathcal{D} are as follows: $f^{-1}(\mathcal{D}) = \mathcal{C}$ $f^{-1}(\{1\}) = \{1, 12\}$ $f^{-1}(\{2\}) = \{3\}$ $f^{-1}(\emptyset) = \emptyset$.

Closed Convexity

Definition

A code C, where $C = C(\mathcal{U})$, is closed convex if there is a realization \mathcal{U} such that the U_i 's can be written as closed convex sets in some \mathbb{R}^d .

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Example



Results

The following is a generalization of a result in Jeffs 2021.

Theorem (Maldonado, Morales, Shaw 2022)

Let *P* be a property of a code *C* such that *P* is closed under intersection. The image under a morphism of a code with property *P* also has property *P*.

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Example

Consider $C = \{\emptyset, 1, \underline{3}, \underline{12}\}$ and $\mathcal{D} = \{\emptyset, \underline{1}, \underline{2}\}$. Consider the surjective morphism $f : C \to \mathcal{D}$ given by:

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V₁ Closed Convex Realization of D

Theorem (Maldonado, Morales, Shaw 2022)

The image under a morphism of a code with *k* maximal codewords has up to *k* maximal codewords.

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Both C and D have 2 maximal codewords.

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Both C and D have 2 maximal codewords.

• Isomorphism invariant

A code \mathcal{D} is a *minor* of a code C, denoted by $\mathcal{D} \leq C$, if there exists a surjective morphism $f : C \to \mathcal{D}$. A proper minor of C is a minor that is not isomorphic to C.

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Example

Let $C = \{\emptyset, 1, 3, 12\}$ and $\mathcal{D} = \{\emptyset, 0, 1, 2\}$, and let $f : C \to \mathcal{D}$ given by

 $f(\emptyset) = \emptyset$ f(1) = 1 f(3) = 2f(12) = 1.

Note, f is a surjective morphism, yet f fails to be injective. Therefore D is a proper minor of C.

A neural code *C* is *minimally non-closed convex* if *C* is not a closed convex code, but every proper minor of *C* is closed convex.

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Example/Proposition (Maldonado, Morales, Shaw 2022)

The following code is not minimally non-closed convex.

$$\begin{split} C = \{ \underline{2345}, \underline{123}, \underline{124}, \underline{145}, 12, 14, 23, 24, 45, 2, 4, \varnothing, \underline{237}, \underline{238}, \underline{367}, \underline{678}, \underline{26}, \\ & 37, 67, 6, 8 \} \end{split}$$

Results

Remark

No code was known to be minimally non-closed, but open convex.

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Theorem (Maldonado, Morales, Shaw 2022)

The code

 $C_6 = \{\underline{125}, \underline{234}, \underline{145}, \underline{123}, 4, 23, 15, 12, \emptyset\}$

is minimally non-closed convex.



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Lemma (Maldonado, Morales, Shaw 2022)

Let $f : C \to D$ be a surjective morphism and $d \in D$ be a maximal codeword. Then there exists a maximal codeword $c \in C$ such that $c \in f^{-1}(\{d\})$.

Lemma (Maldonado, Morales, Shaw 2022)

Let $f : C \to D$ be a surjective morphism such that C and D have the same amount of maximal codewords. Then for maximal $d \in D$, f(c) = d for a unique maximal codeword $c \in C$.

Lemma (Maldonado, Morales, Shaw 2022)

Let $f: C_6 \to D$ be a surjective morphism, where D has four maximal codewords and $C_6 = \{\underline{125}, \underline{234}, \underline{145}, \underline{123}, 4, 23, 15, 12, \emptyset\}$. Then, for non-maximal $\sigma \in C_6$, $f(\sigma) \in \mathcal{D}$ is non-maximal.