

Introduction

Biologists have observed neurons in some animal brains called **place cells**, which act as position sensors. Place cells fire at high rates when the animal is inside corresponding regions of the environment called **place fields**. These fields have been observed to be approximately convex, and intersection patterns of place fields form neural codes. We investigate how the brain represents space by studying **convex neural codes** [2].

Neural Codes

A **neural code** or **code** on n neurons is a collection of firing patterns called **codewords**, $\mathcal{C} \subseteq 2^{[n]}$, where $\emptyset \in \mathcal{C}$. The **maximal codewords** of a code are its maximal elements with respect to set inclusion. We can **generate codes** using sets in Euclidean space.

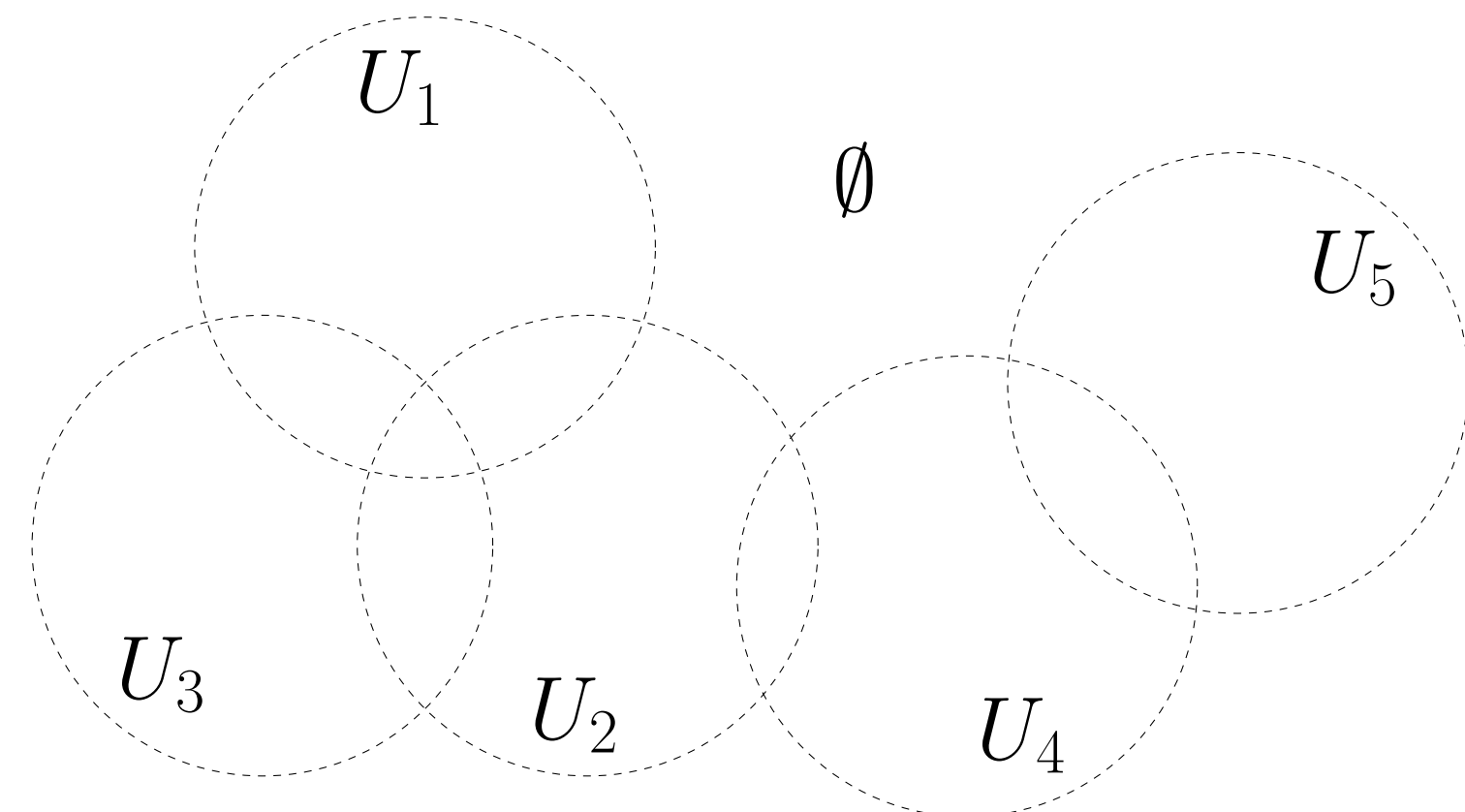


Figure 1: The code generated by $\{U_1, U_2, U_3, U_4, U_5\}$ is $\mathcal{C} = \{123, 24, 45, 12, 13, 23, 1, 2, 3, 4, 5\}$

If \mathcal{U} is composed by sets that are open and convex and $\mathcal{C}(\mathcal{U}) = \mathcal{C}$, then \mathcal{C} is an **(open) convex code**.

Simplicial Complexes

A code \mathcal{C} on n neurons has a **simplicial complex**

$$\Delta(\mathcal{C}) := \{\omega \subseteq [n] : \omega \subseteq \sigma \text{ for some } \sigma \in \mathcal{C}\}.$$

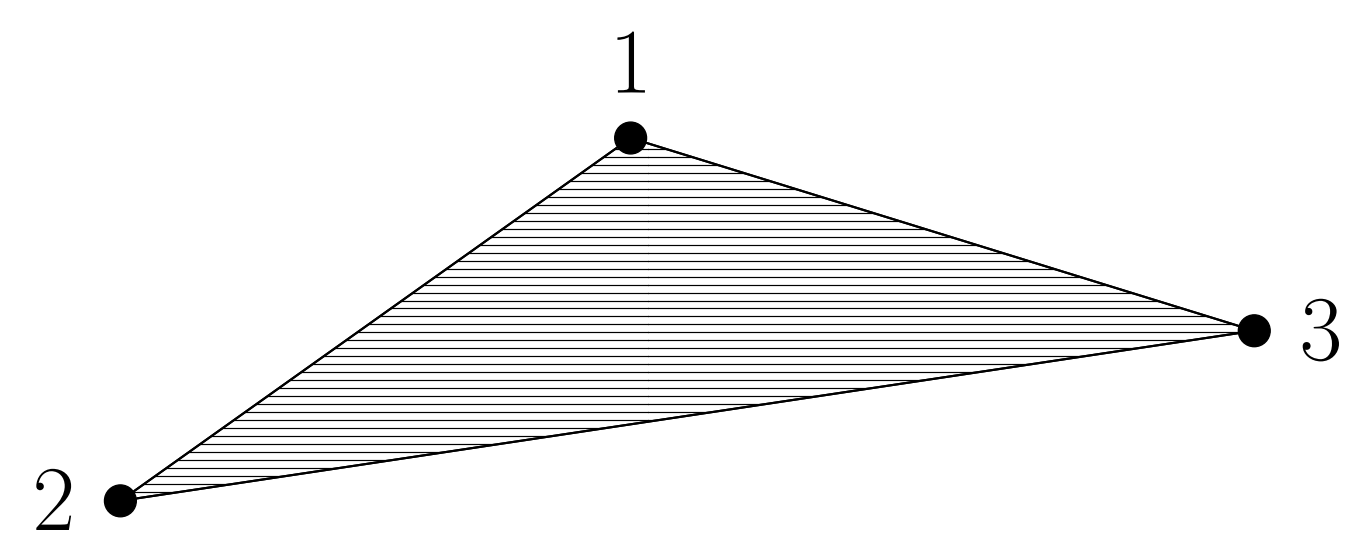


Figure 2: A geometric realization of a simplicial complex composed by one 2-simplex (a filled-in triangle)

Elements of $\Delta(\mathcal{C})$ are called **faces** and maximal faces are called **facets**. Facets correspond exactly to the maximal codewords of \mathcal{C} .

Example: In Fig. 2, 12 is a face and **123** is a facet.

Characterizing Convexity

A code \mathcal{C} is **max- \cap -complete** if it contains every intersection of at least two facets of $\Delta(\mathcal{C})$. Max- \cap -complete codes are convex, as shown in [1].

Consider a $\Delta(\mathcal{C})$ on n neurons and let $\sigma \subseteq [n]$. A **link of σ** with respect to $\Delta(\mathcal{C})$ is defined as

$$\text{Lk}_{\Delta(\mathcal{C})}(\sigma) = \{\tau \subseteq [n] \setminus \sigma : \tau \cup \sigma \in \Delta(\mathcal{C})\}.$$

Example: For $\Delta(\mathcal{C}) = \{123, 124, 45, 12, 13, 23, 14, 24, 1, 2, 3, 4, 5, \emptyset\}$, we look at the links $\text{Lk}_{\Delta(\mathcal{C})}(1)$ and $\text{Lk}_{\Delta(\mathcal{C})}(4)$:

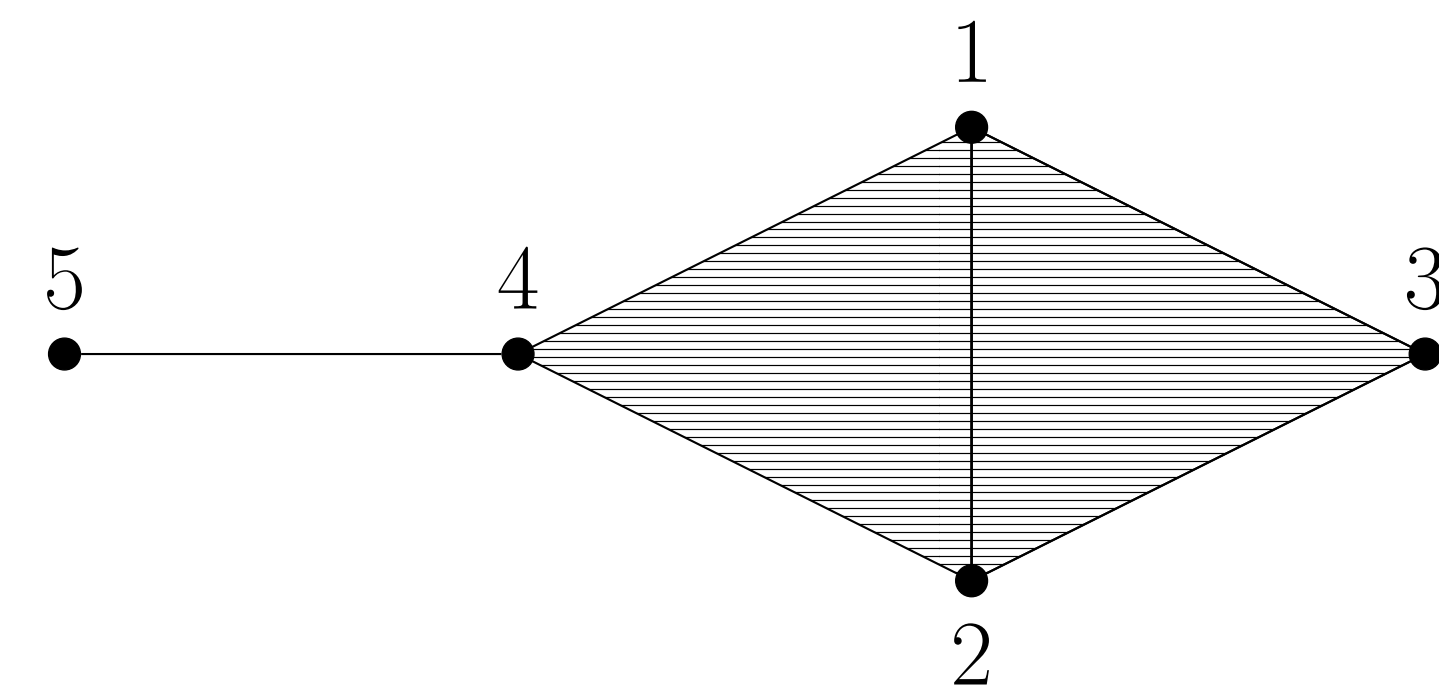


Figure 3: A geometric realization of $\Delta(\mathcal{C})$

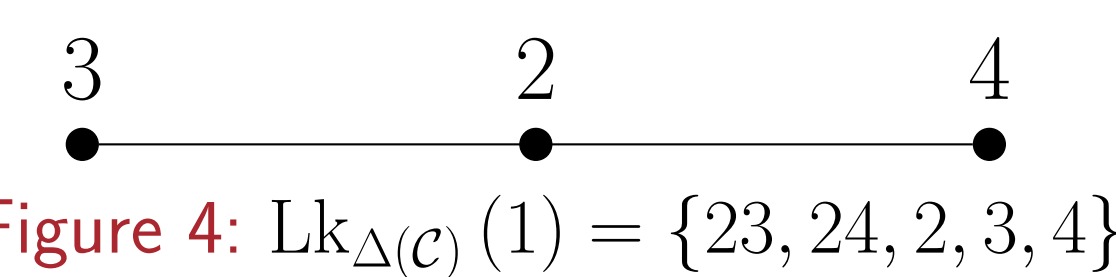


Figure 4: $\text{Lk}_{\Delta(\mathcal{C})}(1) = \{23, 24, 2, 3, 4\}$

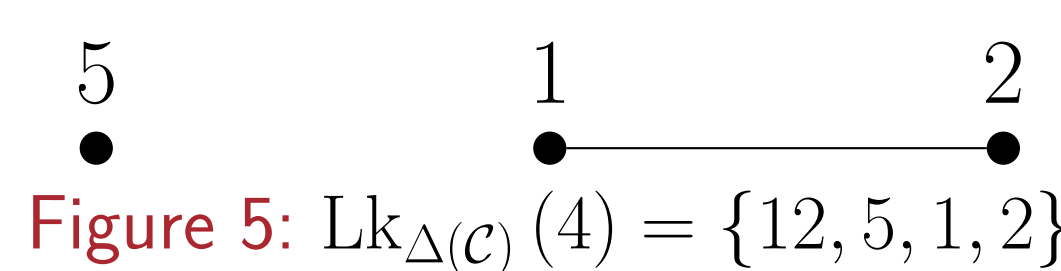


Figure 5: $\text{Lk}_{\Delta(\mathcal{C})}(4) = \{12, 5, 1, 2\}$

Let \mathcal{C} be a neural code with simplicial complex $\Delta(\mathcal{C})$. We say \mathcal{C} has a **local obstruction at σ** if there exists some nonempty face $\sigma \in \Delta(\mathcal{C})$ such that:

- σ is the intersection of facets of $\Delta(\mathcal{C})$
- $\sigma \notin \mathcal{C}$
- $\text{Lk}_{\Delta(\mathcal{C})}(\sigma)$ is not contractible.

It has been shown that codes with three maximal codewords are convex if and only if they have no local obstructions [4].

The Problem

Local obstructions do not appear in convex codes, as shown in [3]. Neither do **wheels**, a geometric and combinatorial object discussed in [6]. We investigate the converse for codes with four maximal codewords.

Conjecture

Let \mathcal{C} be a neural code with 4 maximal codewords. If \mathcal{C} has no local obstructions and no wheels, then \mathcal{C} is convex.

Nerves

For a collection of subsets $\mathcal{W} = \{W_1, W_2, \dots, W_n\}$ of a set X , the **nerve of \mathcal{W}** is the simplicial complex

$$\mathcal{N}(\mathcal{W}) := \{I \subseteq [n] : \bigcap_{i \in I} W_i \neq \emptyset\}.$$

A topological result called the *nerve lemma* [5] lets us investigate convexity by looking at the nerve of the set of facets \mathcal{F} of a code \mathcal{C} , denoted by $\mathcal{N}(\mathcal{F})$.

Cases

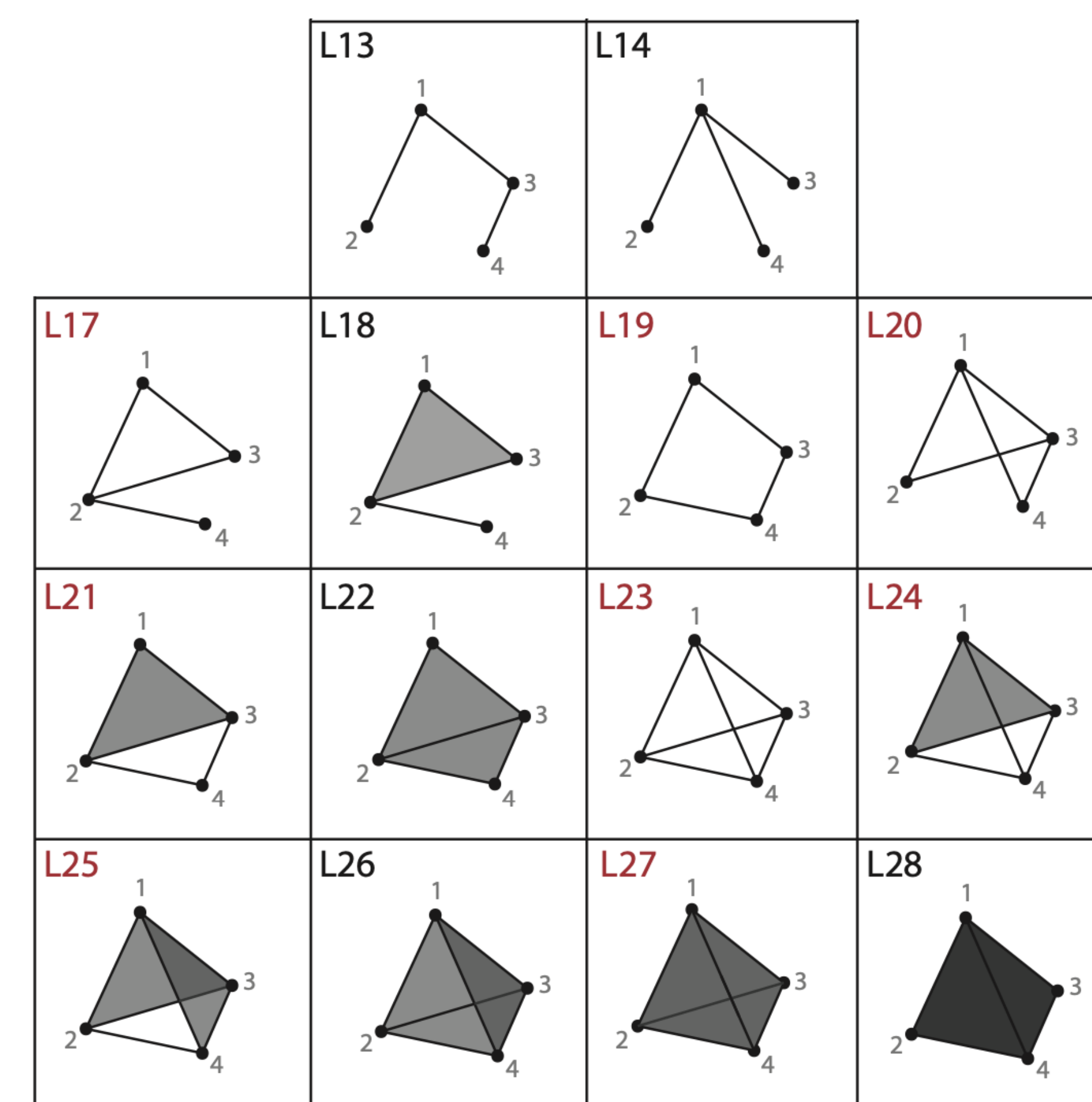


Figure 6: Connected simplicial complexes with 4 vertices [2].

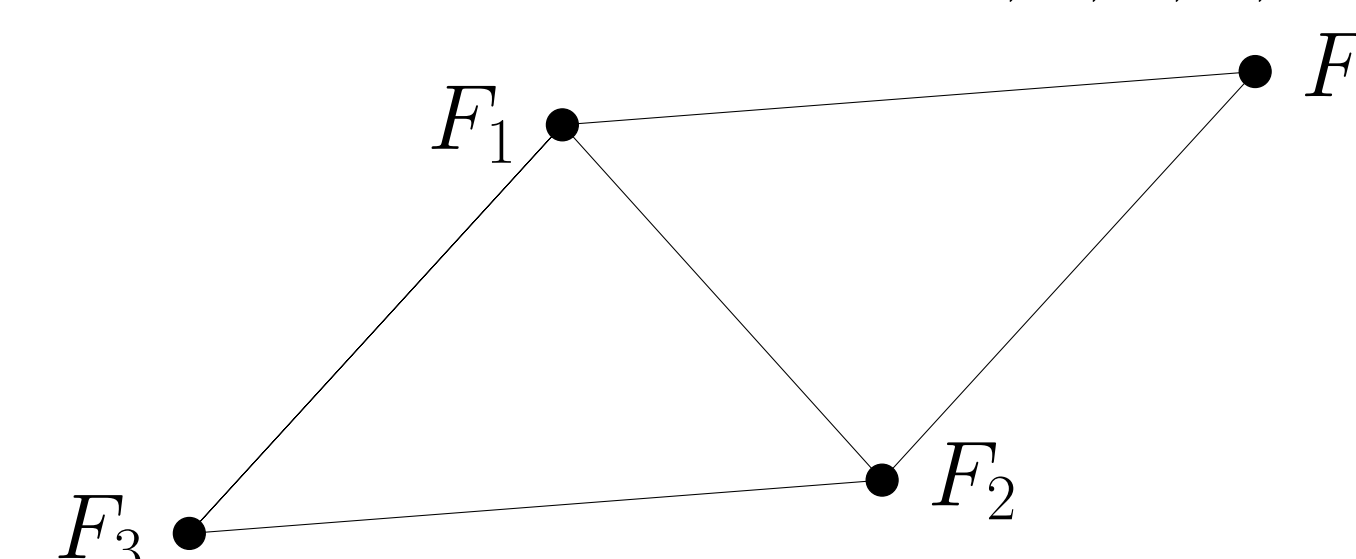
Results

Theorem

Let \mathcal{C} be a 4-maximal code. If $\mathcal{N}(\mathcal{F})$ contains no 2-simplices, then the following are equivalent:

- \mathcal{C} has no local obstructions
- \mathcal{C} is max- \cap -complete
- \mathcal{C} is convex.

Example: If $\mathcal{C} = \{123, 145, 24, 35, \sigma_1, \dots, \sigma_k\}$, $\mathcal{N}(\mathcal{F})$ does not contain 2-simplices. Thus \mathcal{C} is convex exactly when facet intersections 1, 2, 3, 4, 5 are in \mathcal{C} .



Results Continued

Theorem

Let \mathcal{C} be a 4-maximal code. If $\mathcal{N}(\mathcal{F})$ is the simplicial complex L18 or L21, then \mathcal{C} is convex if and only if it contains no local obstructions.

Future Work

We believe our current results can be extended to the L22 case. However, we expect wheels to play a role in cases L24 through L28.

Acknowledgements

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References

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