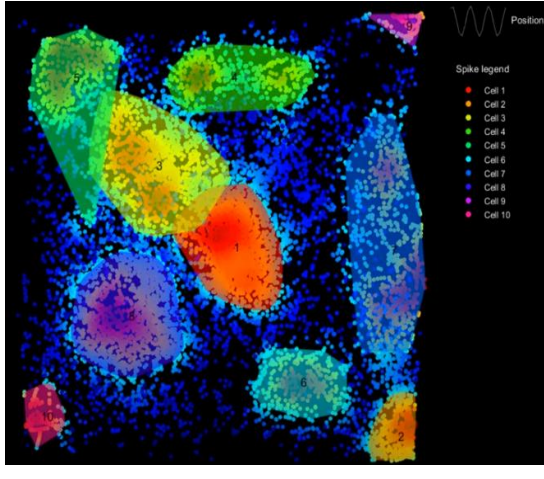




# Covering Relations in the Poset of Combinatorial Neural Codes

## MOTIVATION



Place fields of 10 recorded place cells in a rat [1] are approximately convex.

- Place cells fire in specific regions called **place fields**.
- Place fields are allocentric and believed to contribute to the **neural maps** of the environment.
- Place fields away from the walls are approximately open **convex** (e.g., circular or elliptical [2], [3]) for dorsal hippocampal place cells.

Collective cell co-firings can tell us about the convexity of their place fields [5].

## PROBLEM

How can we determine if a neural code admits an arrangement of open convex place fields?

Complexity<sup>[7]</sup>: NP-hard and  $\exists \mathbb{R}$ -hard.

## PRELIMINARIES

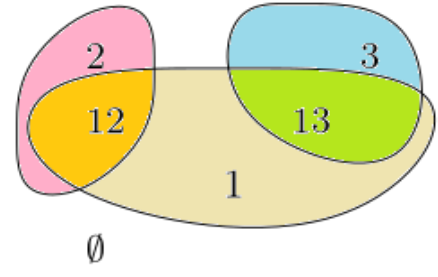
A **(combinatorial) neural code** is a collection of some subsets of  $[n] = \{1, \dots, n\}$  and  $\emptyset$ .

Elements are **codewords**, written as strings (e.g., 12 instead of  $\{1, 2\}$ ).

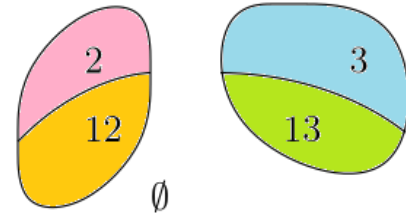
The **code** of an open cover  $\mathcal{U} = \{U_1, \dots, U_n\}$  is

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

The code  $\mathcal{C}_1 = \{\emptyset, 1, 2, 3, 12, 13\}$  is convex because it is the code  $\mathcal{C}(\mathcal{U})$  of the below open convex cover  $\mathcal{U}$ .



The code  $\mathcal{C}_2 = \{\emptyset, 2, 3, 12, 13\}$  cannot admit any arrangement of open convex cover.

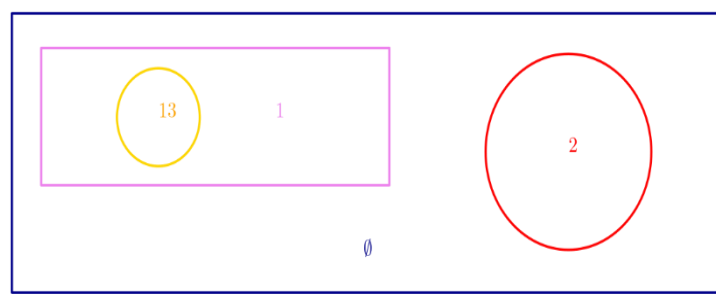


A **trunk**<sup>[6]</sup> in a neural code  $\mathcal{C}$  is either  $\emptyset$  or  $\text{Tk}_{\mathcal{C}}(\sigma)$  for some  $\sigma \subseteq [n]$  where

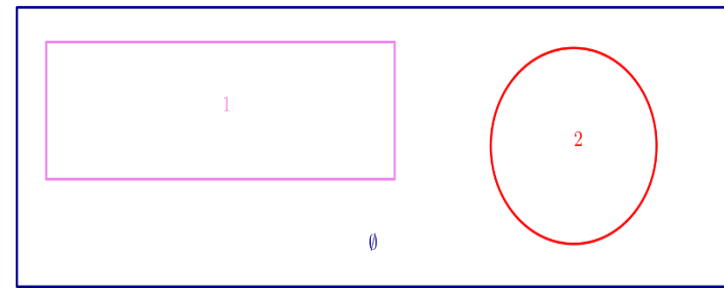
$$\text{Tk}_{\mathcal{C}}(\sigma) = \{\tau \in \mathcal{C} : \sigma \subseteq \tau\}$$

E.g.:  $\mathcal{D} = \{\emptyset, 1, 2, 13\}$

- $\text{Tk}_{\mathcal{D}}(\emptyset) = \{\emptyset, 1, 2, 13\} = \mathcal{D}$
- $\text{Tk}_{\mathcal{D}}(1) = \{1, 13\}$
- $\text{Tk}_{\mathcal{D}}(2) = \{2\}$
- $\text{Tk}_{\mathcal{D}}(3) = \{13\}$



Trunks of  $\emptyset$  (dark blue), 1 (pink), 2 (red), and 3 (orange) in  $\mathcal{D}$ .



Trunks of  $\emptyset$  (dark blue), 1 (pink), and 2 (red) in  $\mathcal{C}$ .

E.g.:  $\mathcal{C} = \{\emptyset, 1, 2\}$

- $\text{Tk}_{\mathcal{C}}(\emptyset) = \{\emptyset, 1, 2\} = \mathcal{C}$
- $\text{Tk}_{\mathcal{C}}(1) = \{1\}$
- $\text{Tk}_{\mathcal{C}}(2) = \{2\}$

A **code morphism**<sup>[7]</sup> is a function  $f: \mathcal{D} \rightarrow \mathcal{C}$  of neural codes such that preimage of every proper trunk  $T \subseteq \mathcal{C}$  is a proper trunk in  $\mathcal{D}$ .

E.g.:  $\mathcal{D} = \{\emptyset, 1, 2, 13\}$  and  $\mathcal{C} = \{\emptyset, 1, 2\}$

$$f: \{\emptyset, 1, 2, 13\} \rightarrow \{\emptyset, 1, 2\}$$

$$\emptyset \mapsto \emptyset; \quad 1, 13 \mapsto 1; \quad 2 \mapsto 2$$

is a code surjective morphism since  $f$  is onto and

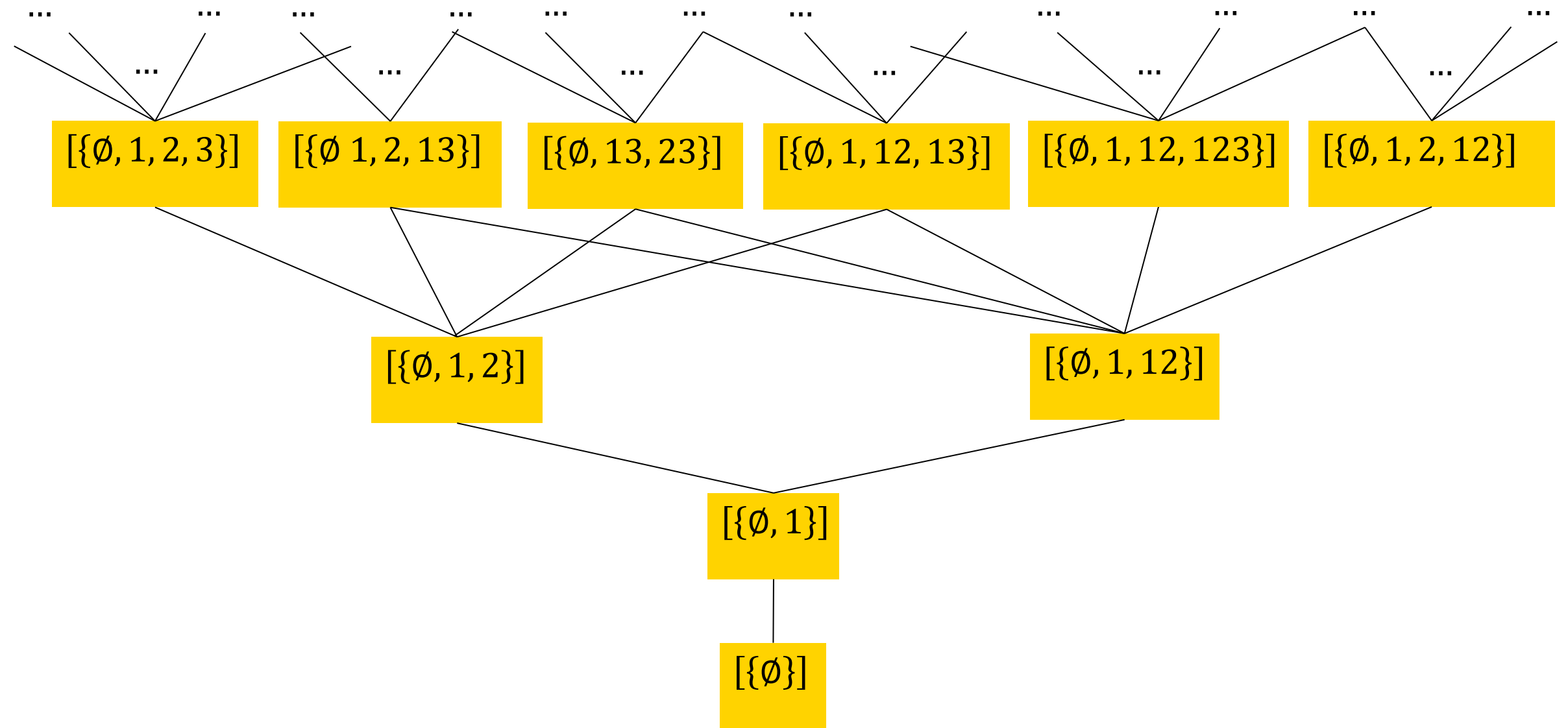
$$f^{-1}(\text{Tk}_{\mathcal{C}}(\emptyset)) = f^{-1}(\mathcal{C}) = \mathcal{D} = \text{Tk}_{\mathcal{D}}(\emptyset)$$

$$f^{-1}(\text{Tk}_{\mathcal{C}}(1)) = \{1, 13\} = \text{Tk}_{\mathcal{D}}(1)$$

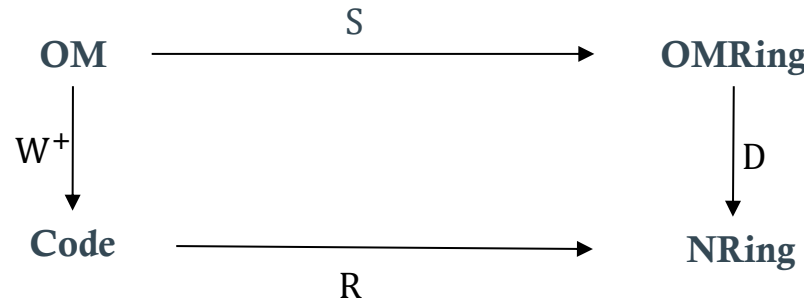
$$f^{-1}(\text{Tk}_{\mathcal{C}}(2)) = \{2\} = \text{Tk}_{\mathcal{D}}(2)$$

## THE POSET $\mathbf{P}_{\text{Code}}$

Code surjective morphisms induce a partial order named “**is a minor of**” on the set of isomorphism classes of neural codes, resulting in a poset called  $\mathbf{P}_{\text{Code}}$ <sup>[7]</sup>.

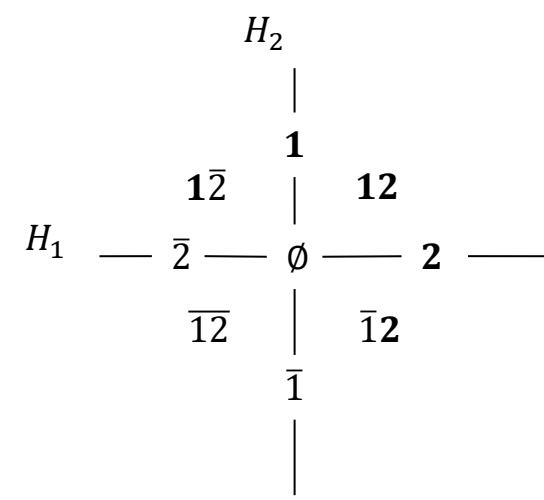


The category of acyclic oriented matroids **OM**, that of oriented matroid rings **OMRing**, that of neural codes **Code**, and that of neural rings **NRing** commute in the following diagram<sup>[8]</sup>:

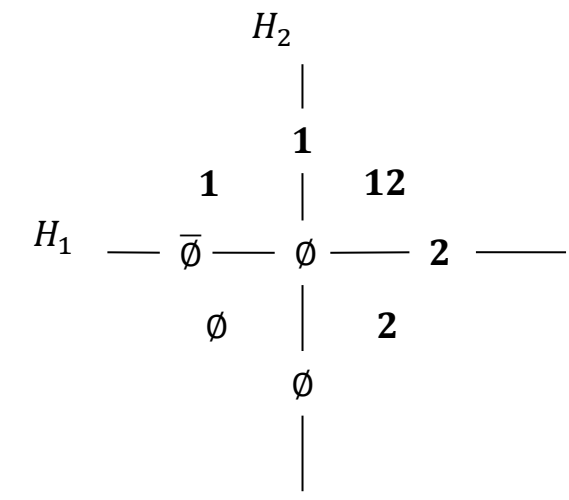


## CONJECTURE

Every open convex neural code is the image of some representable oriented matroid under some code surjective morphism<sup>[8]</sup>.



The covectors of an oriented matroid arising from an arrangement of  $H_1$  and  $H_2$ .



The convex code  $\mathcal{C} = \{\emptyset, 1, 2, 12\}$  given by the positive open half-spaces of  $H_1$  and  $H_2$ .

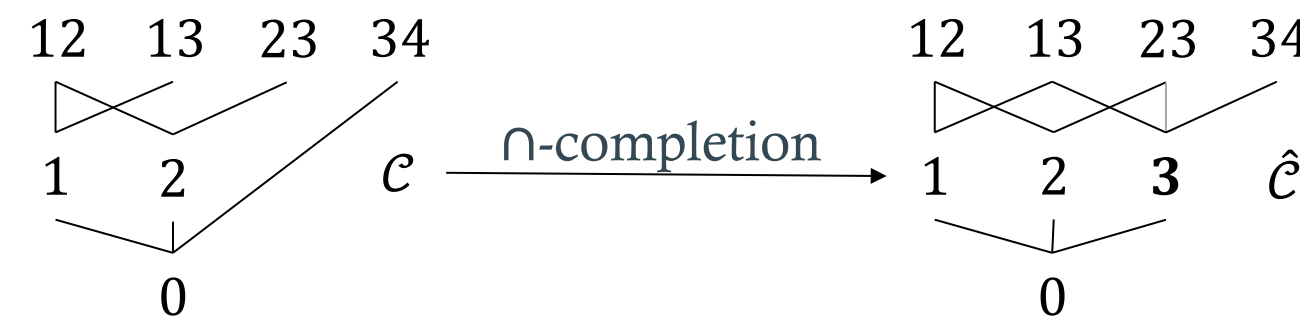
## OUR CONTRIBUTIONS

### ISOLATED SUBSETS

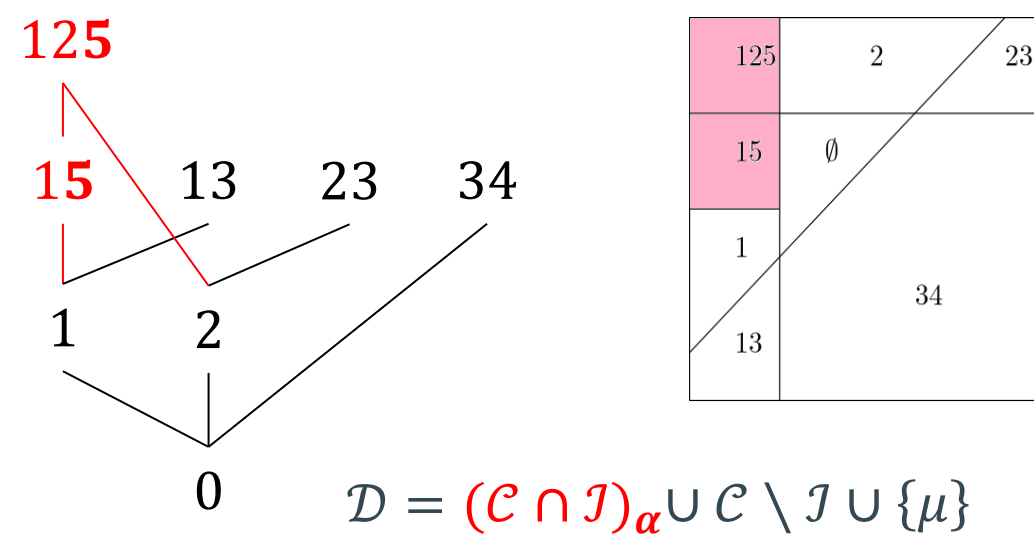
Given an  $\cap$ -complete neural code  $\mathcal{C}$ . An **isolated subset**  $\mathcal{I}$  of  $\mathcal{C}$  is an  $\cap$ -complete subset that satisfies the condition that  $\sigma \not\supseteq \tau, \forall \sigma \in \mathcal{C} \setminus \mathcal{I}, \forall \tau \in \mathcal{I} \setminus \{\mu\}$  where  $\mu$  is the minimal codeword in  $\mathcal{I}$ .

### COVERING CODES

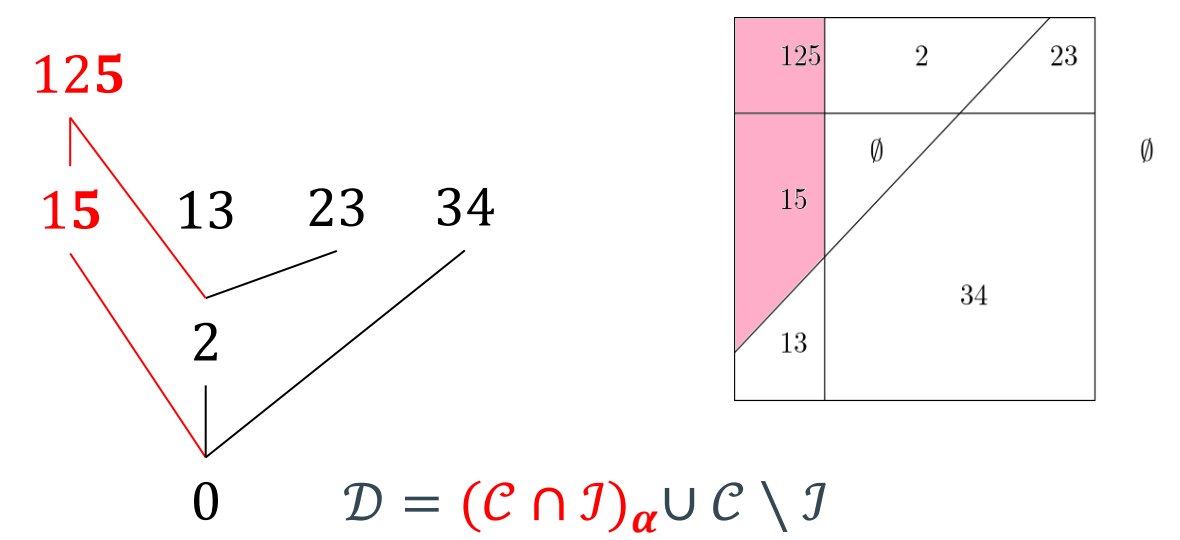
For every isolated subset  $\mathcal{I}$  of the  $\cap$ -completion  $\hat{\mathcal{C}}$  of  $\mathcal{C}$ , depending on the conditions on  $\mathcal{I}$ , one can construct a neural code  $\mathcal{D}$  covering  $\mathcal{C}$  in  $\mathbf{P}_{\text{code}}$  by one of the following ways (with examples).



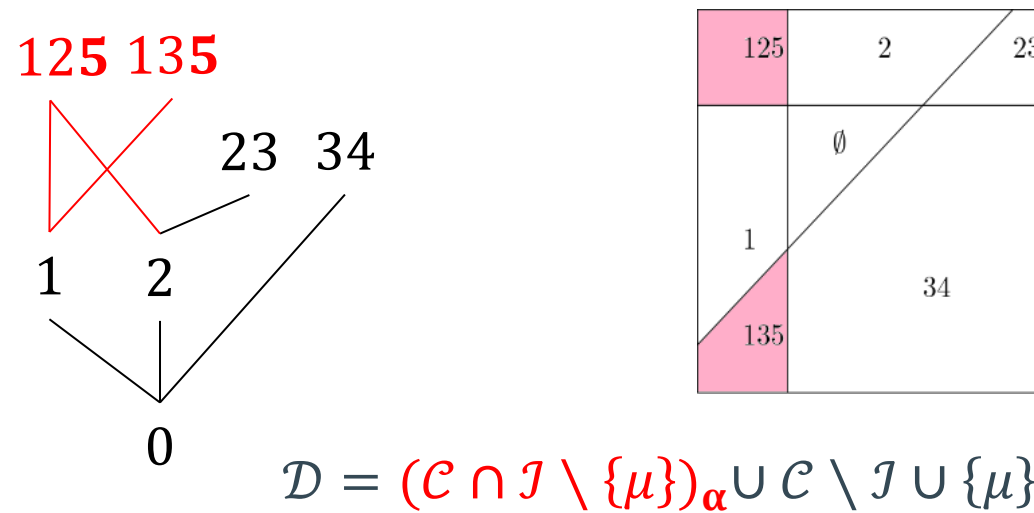
**Type 1:**  $\mu \in \mathcal{C}$ .  
E.g.:  $\mathcal{I} = \{1, 12\}$



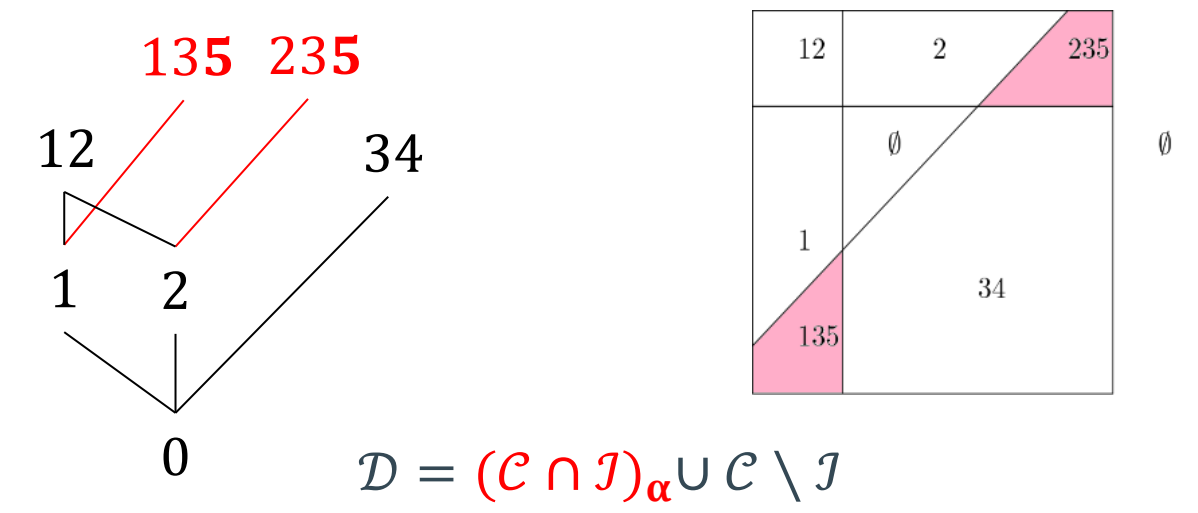
**Type 2:**  $\mu \in \mathcal{C}$  and  $\text{Tk}_{\mathcal{C}}(\mu) \setminus \mathcal{I} \neq \emptyset$ .  
E.g.:  $\mathcal{I} = \{1, 12\}$



**Type 3:**  $\mu \in \mathcal{C}$  and  $\mu = \bigcap_{\sigma \in \mathcal{C} \setminus \mathcal{I}} \sigma$ .  
E.g.:  $\mathcal{I} = \{1, 12, 13\}$



**Type 4:**  $\mu \notin \mathcal{C}, \mu = \bigcap_{\sigma \in \mathcal{C} \setminus \mathcal{I}} \sigma$ , and  $\text{Tk}_{\mathcal{C}}(\mu) \setminus \mathcal{I} \neq \emptyset$ .  
E.g.:  $\mathcal{I} = \{3, 13, 23\}$



## References

- [1] R. Grieves, E. Duvelle, and K. Jeffery (2019). “10 place cells (rat hippocampus CA1) recorded simultaneously over 50 minutes of foraging.”
- [2] R.U. Muller, J.L. Kubie, J.B. Ranck, Jr. (1987). “Spatial firing patterns of hippocampal complex-spike cells in a fixed environment.” In: J. Neurosci. 7, pp.1935-1950.
- [3] J. O’Keefe and N. Burgess (1996). “Geometric determinants of the place fields of hippocampal neurons.” In: Nature 381, pp. 425-428.
- [4] C. Curto and V. Itskov (2008). “Cell Groups Reveal Structure of Stimulus Space.” In: PLOS Comput. Biol. 4,

- [5] C. Curto et al. (2013). “The neural ring: an algebraic tool for analyzing the intrinsic structure of neural codes.” In: Bulletin of mathematical biology 75 (9), pp. 1571-1611.

- [6] R. A. Jeffs (2020). “Morphisms of neural codes.” In: SIAM Journal on Applied Algebra and Geometry.

- [7] A. Kunin, C. Lienkaemper, and Z. Rosen (2020). “Oriented matroids and combinatorial neural codes.” In: Combinatorial Theory.

\*Note:  $(S)_{\alpha}$  denotes the operation of adding a new neuron  $\alpha$  to all codewords in  $S$ .