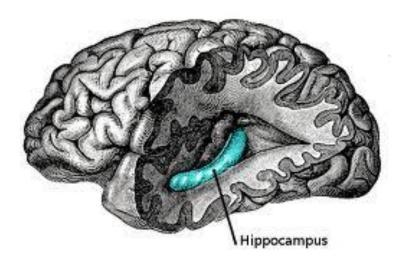
## Topology and Neuroscience: Data analysis

#### Yuri Dabaghian

Baylor College of Medicine & Rice University

dabaghian at rice.edu

## Hippocampus and space coding



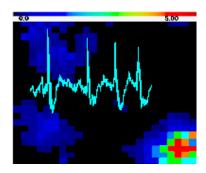
• Forming new memories

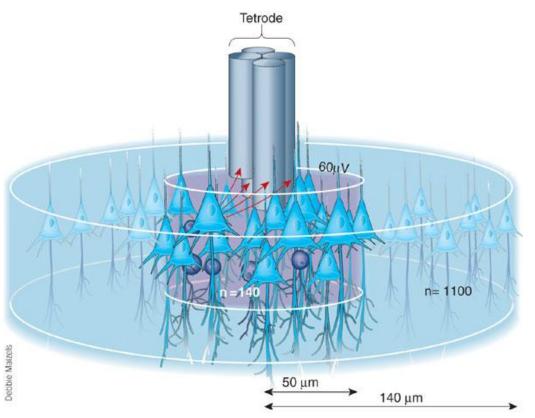


- Spatial behavior, spatial memory
  - navigation, navigation planning
  - imagining spatial scenes

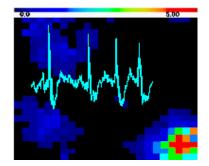
### Electrophysiology of spatial coding



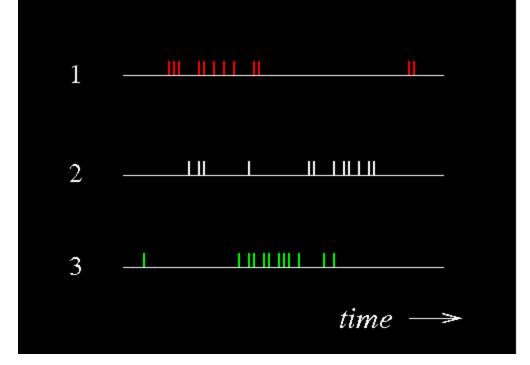




#### Electrophysiology of spatial coding



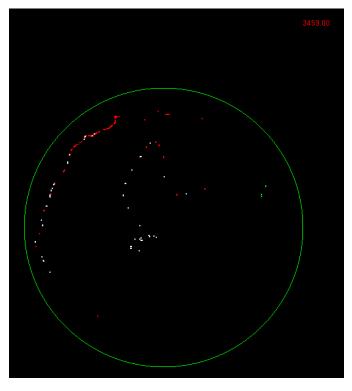


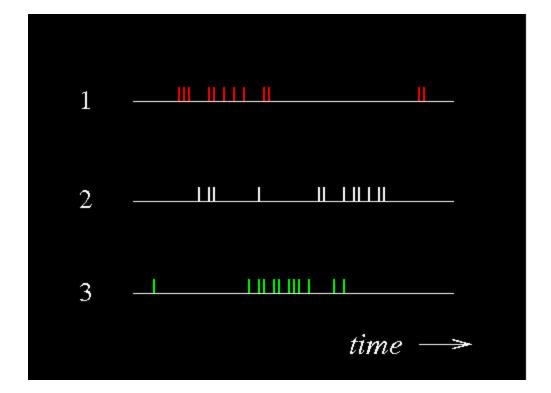


## Electrophysiology of spatial coding

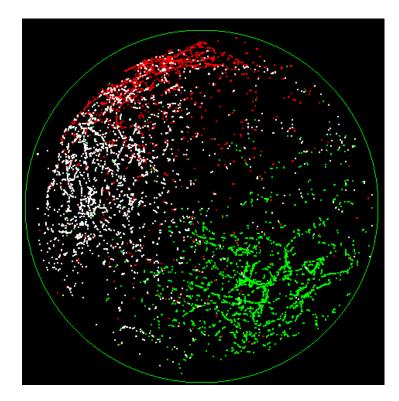


#### Firing activity is correlated with spatial location





## Hippocampal spatial map

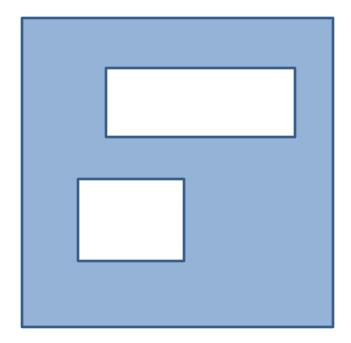


~ 4,000 cells are active simultaneously on each side

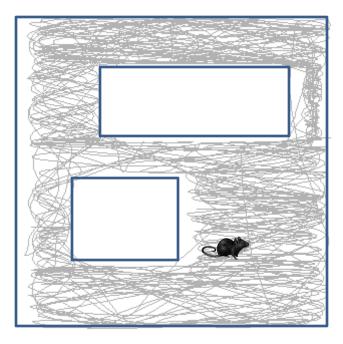
~400,000 cells total on each side

The ensemble of place cells forms a map of the environment

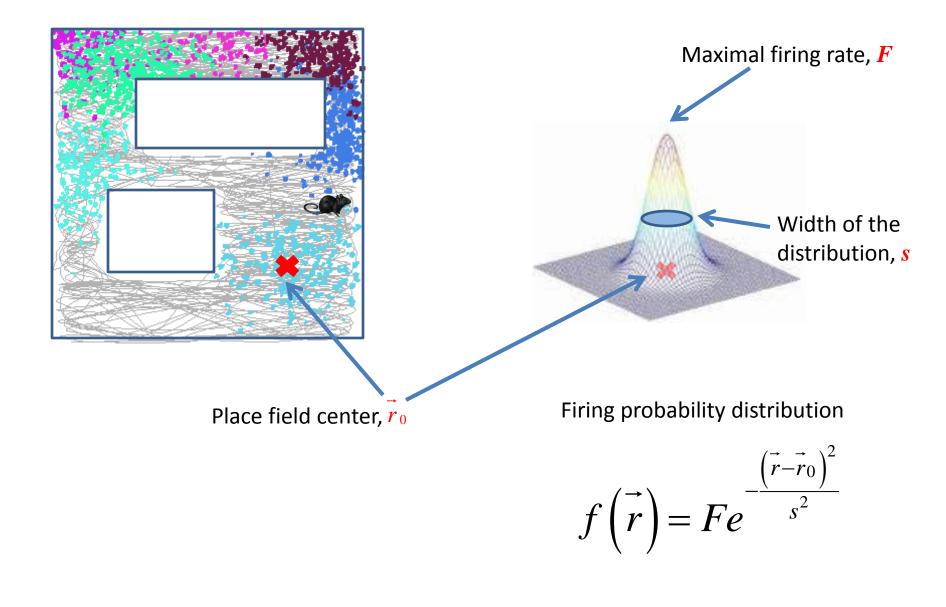
## A space with 2 holes



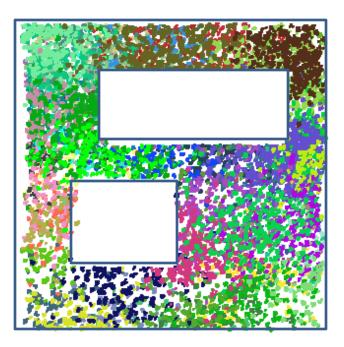
## Environment with 2 holes + trajectory



### Examples of simulated place fields



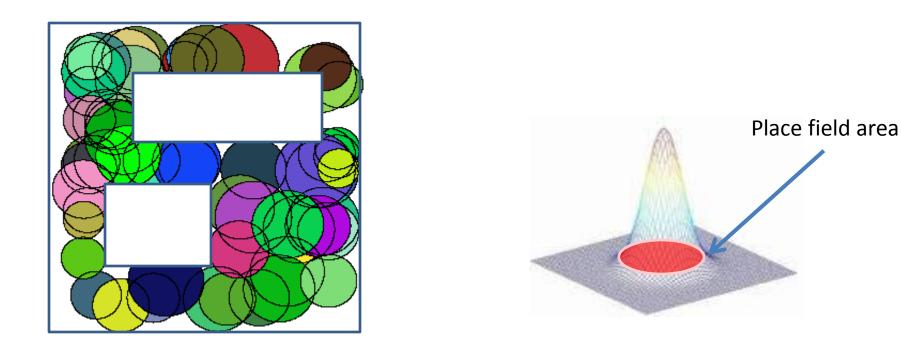
## A place field map of a 2 hole environment



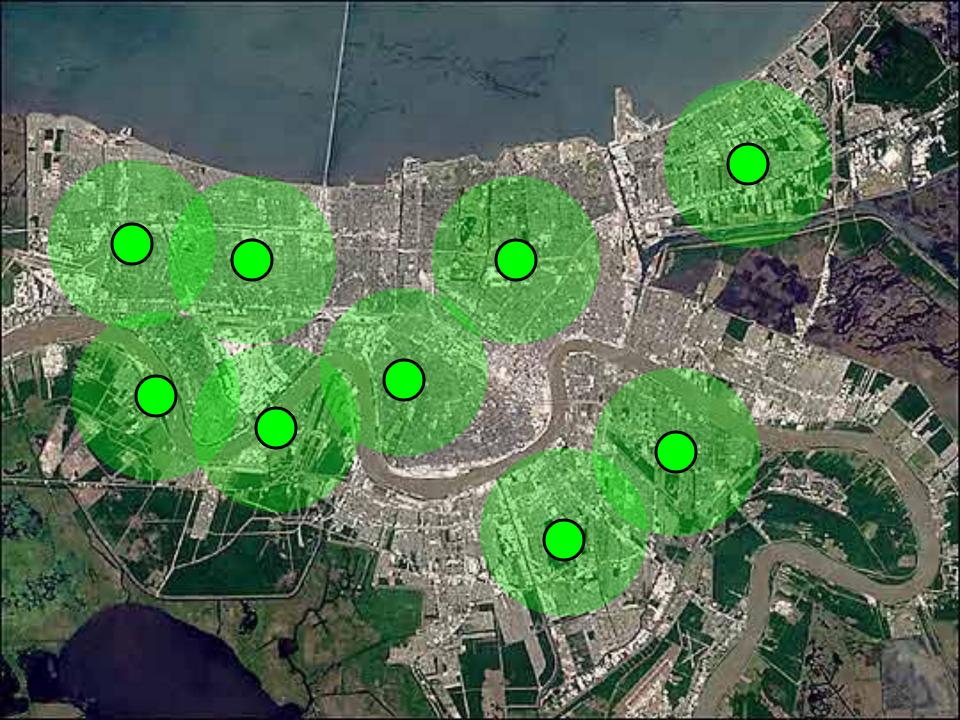
# An ensemble of *N* place cells forms a map of the environment

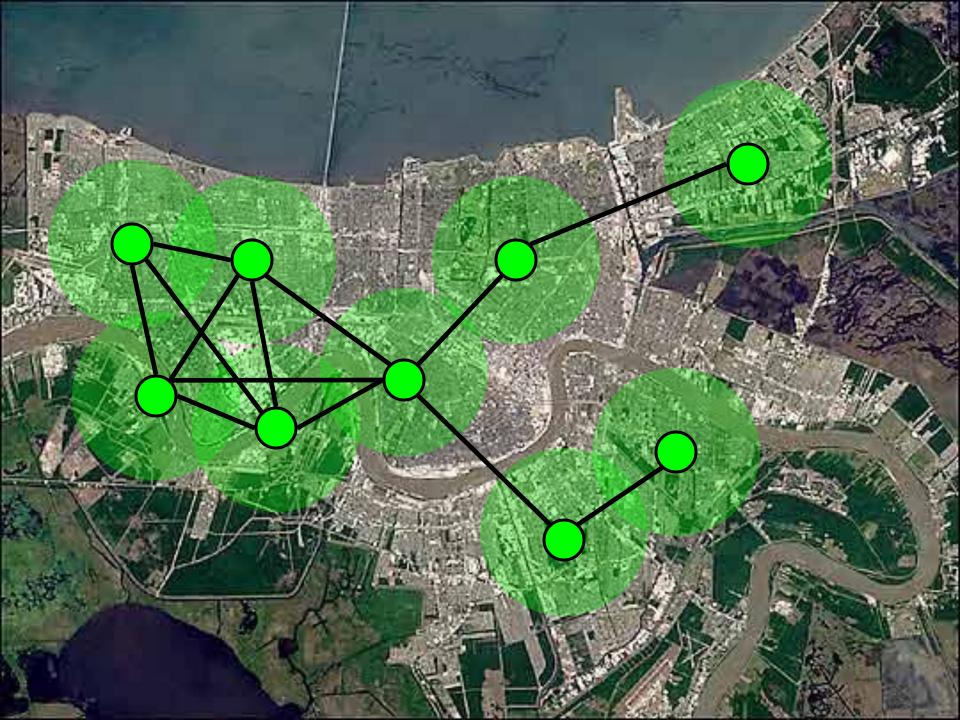
Information transmitted by the place cells to the downstream neurons

## A schematic structure of the place field map

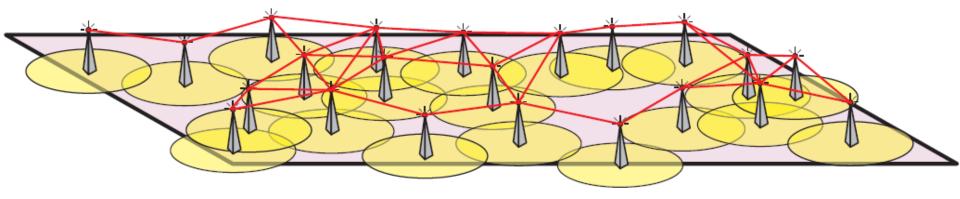


#### Information transmitted by the place cells to the downstream neurons



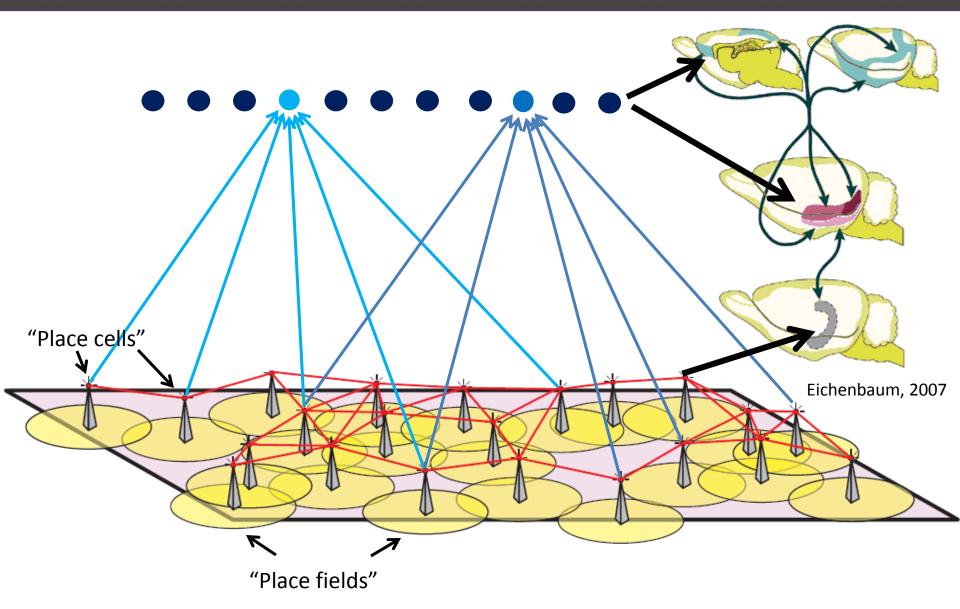


### Topology from sensor networks

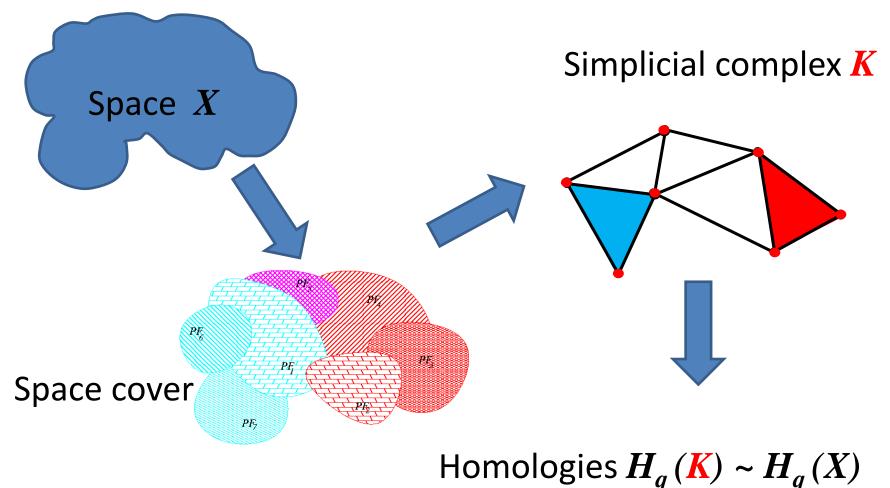


V. de Silva, Homological sensor networks, (2007)

### Topology from neural networks



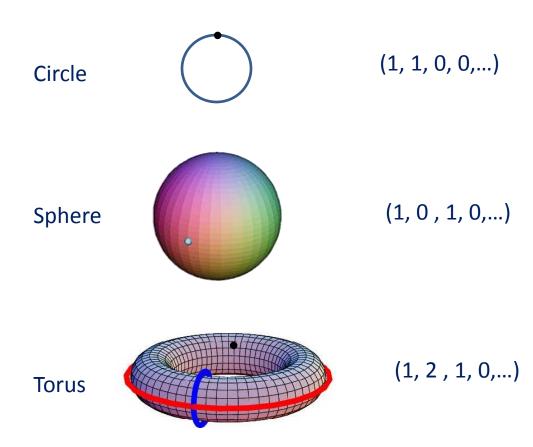
## Topological analysis - Čech complex

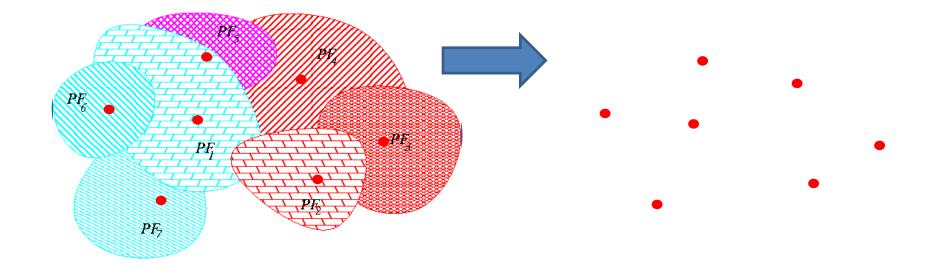


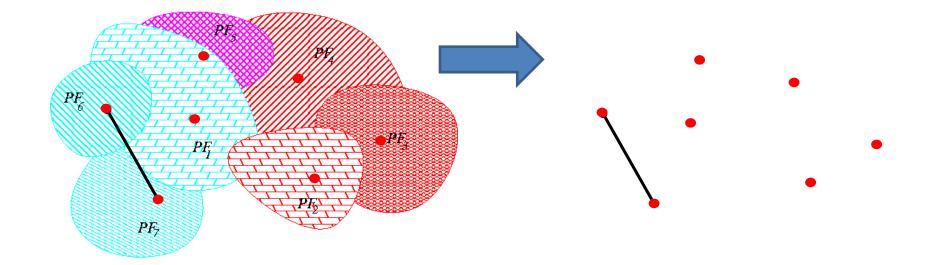
2

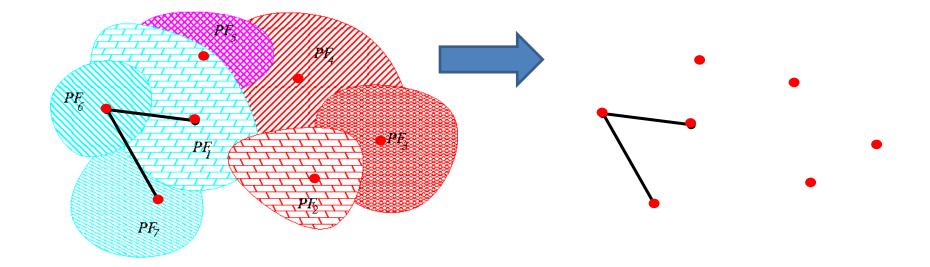
#### Betti index – base cycles in every dimension

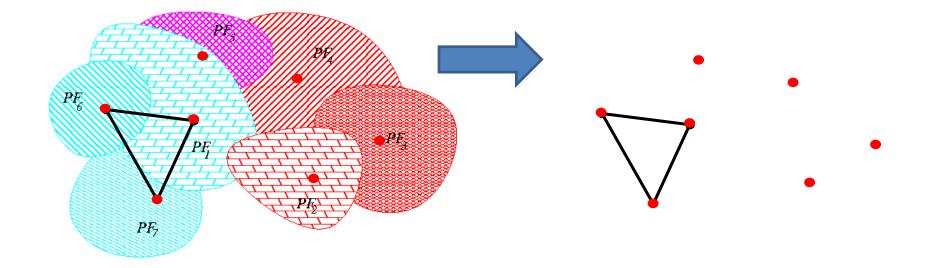


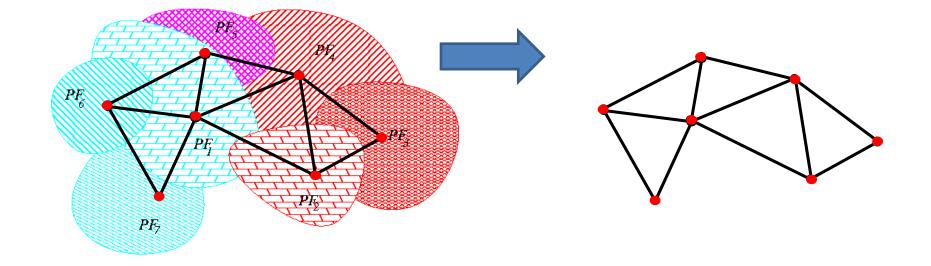


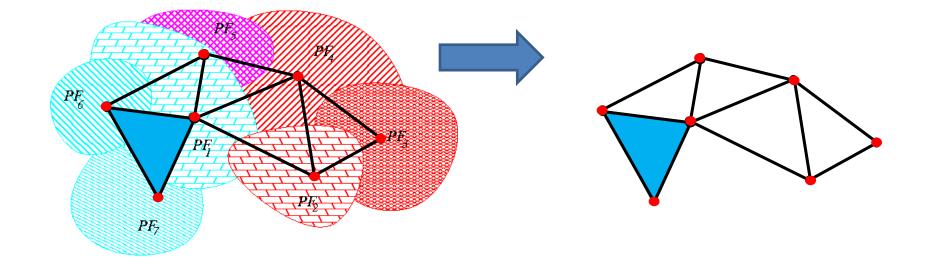


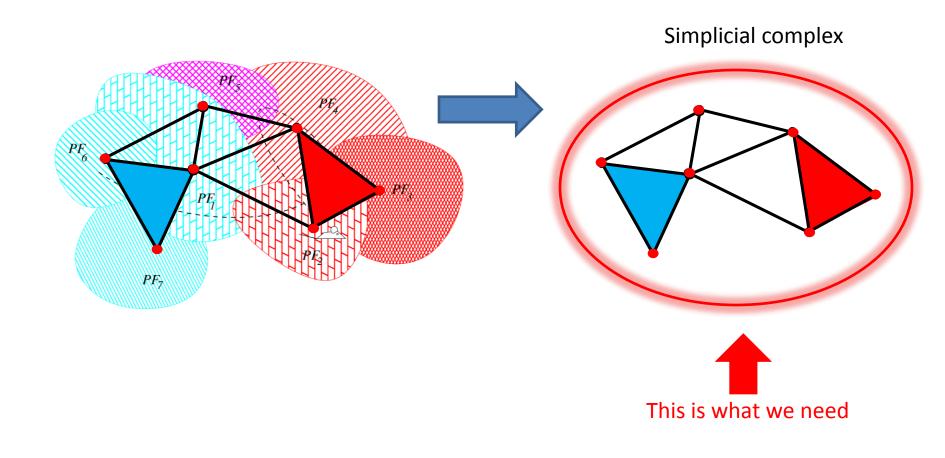




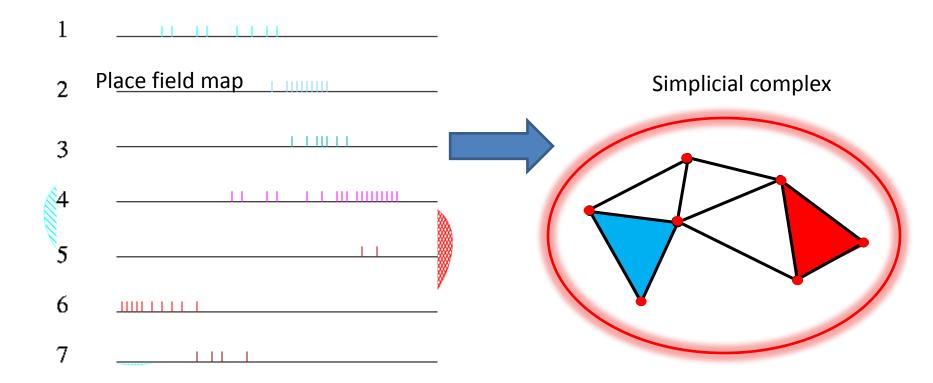


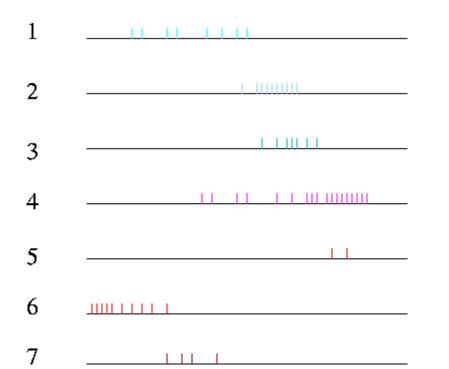


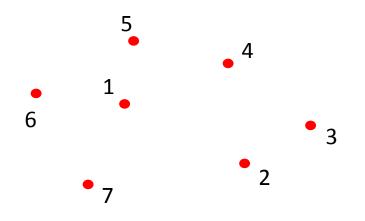


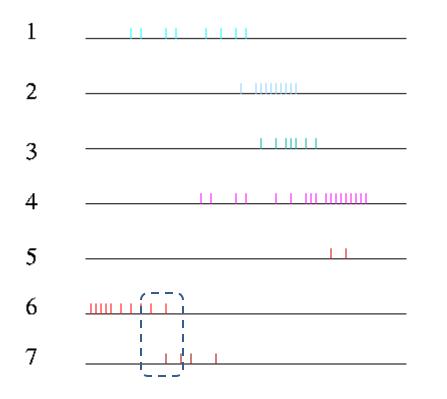


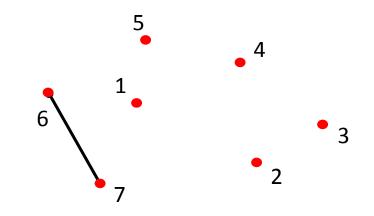
## Čech complex – between space and time

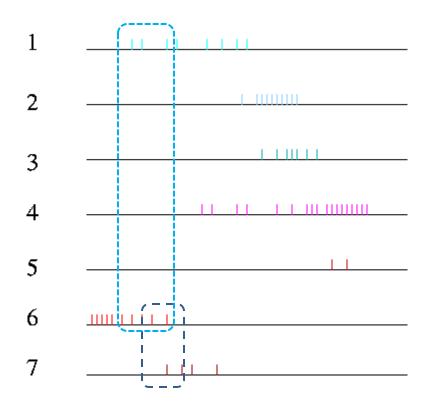


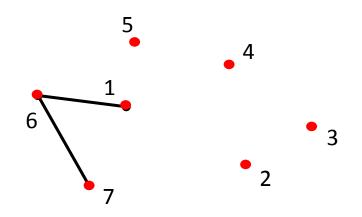


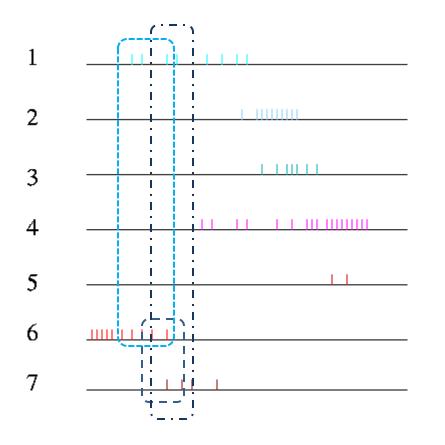


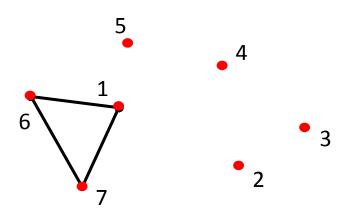


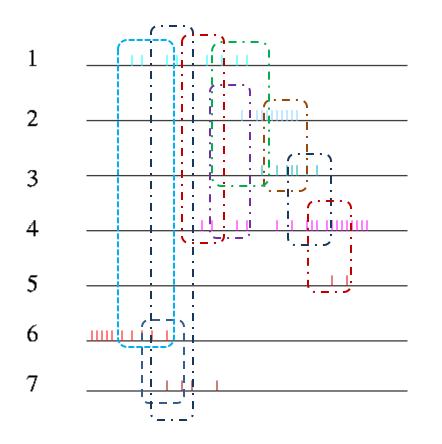


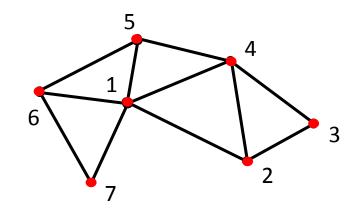


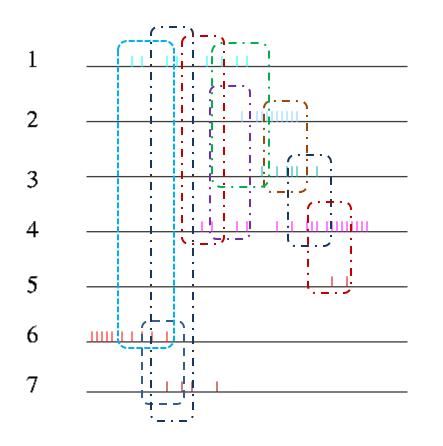


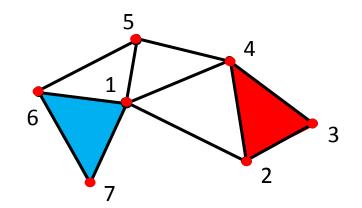




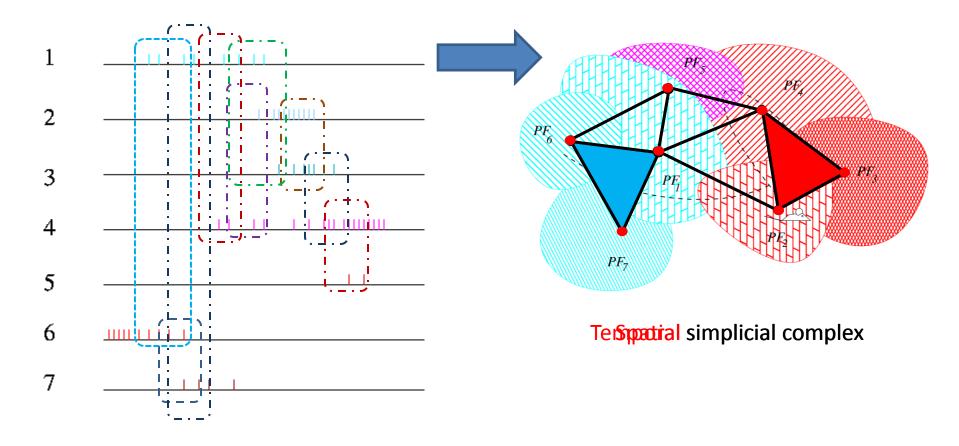




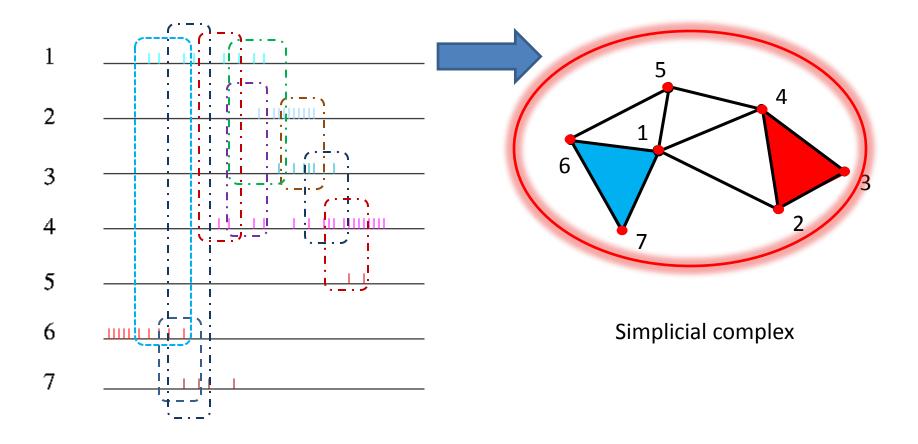




## Čech complex – between space and time

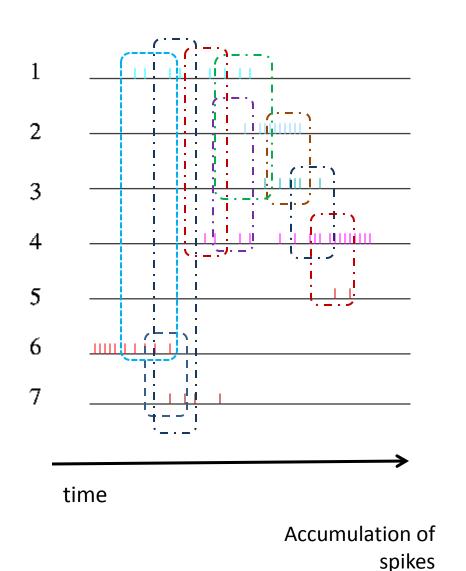


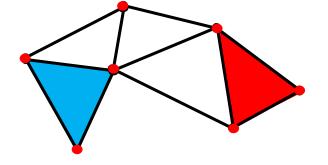
Dabaghian et al., (2007)



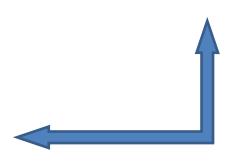
Dabaghian et al., (2007)

## Topological information unfolds in time



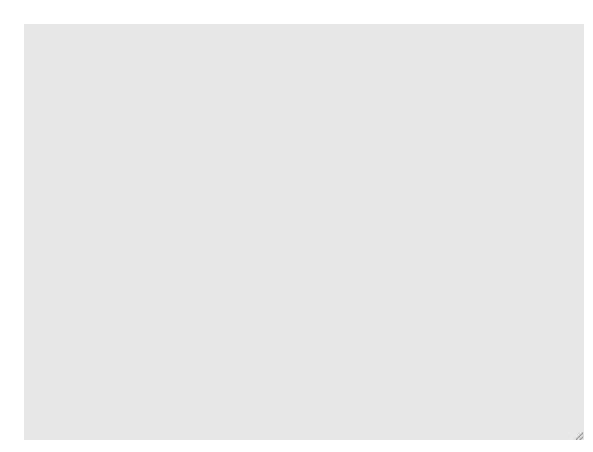


Accumulation of topological information



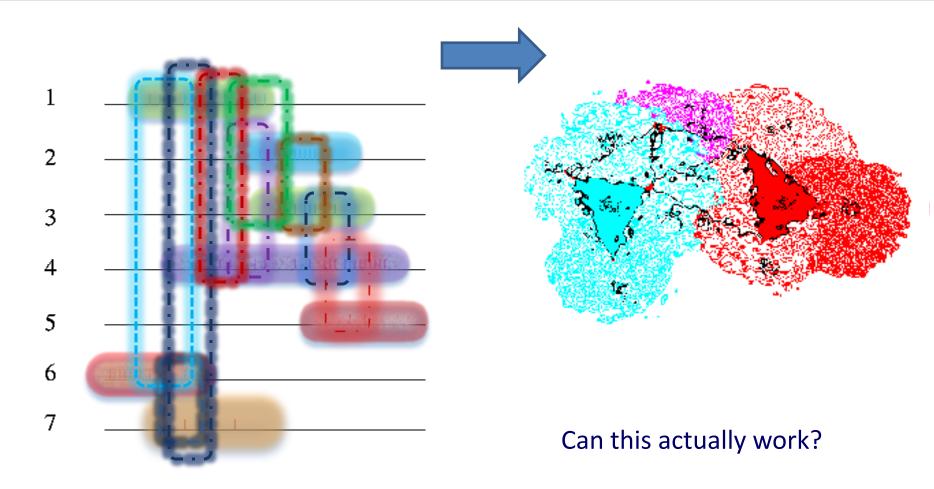
Dabaghian et al, (2008)

#### Temporal coactivity $\rightarrow$ place field connectedness

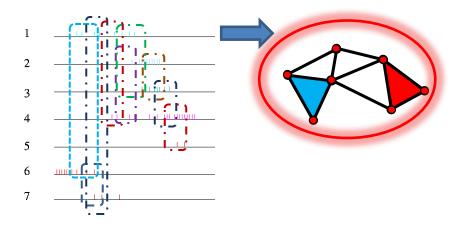


Can this actually work?

#### Can the map sustain biological variability?



#### Topics for topological analysis



- Is topology always extractible?
- Are the results robust?
- Does it take a reasonable time?

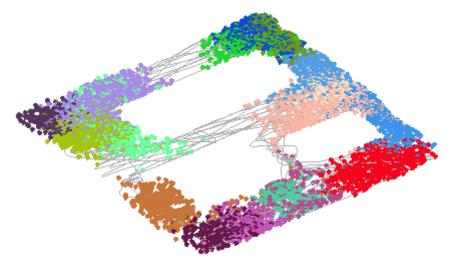
#### Answers depend on the quality of a map

#### Quality of the spatial map

#### Map characteristics:

- Firing rates of place cells
- Place field sizes
- Etc.

$$\langle F \rangle, \langle s \rangle, N$$



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#### Quality of the spatial map

#### Map characteristics:

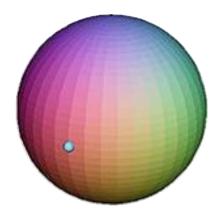
- Firing rates of place cells
- Place field sizes

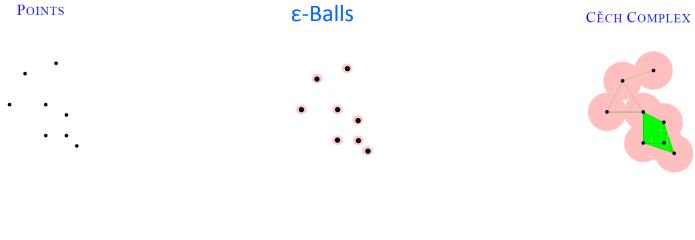
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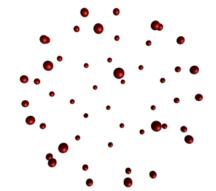
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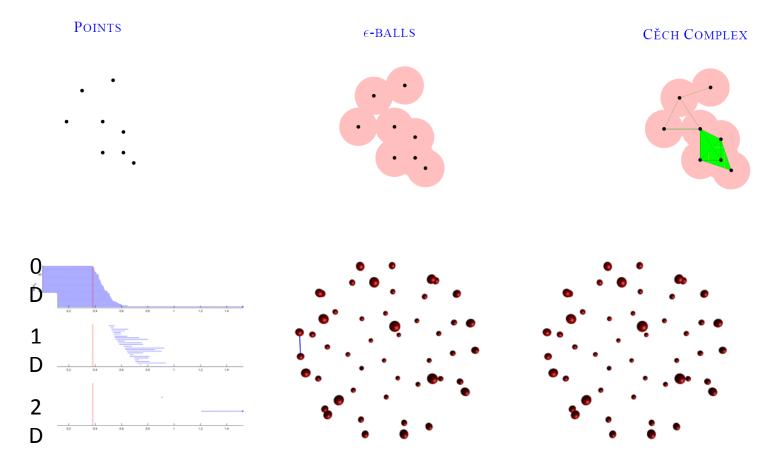
• Etc.

$$\langle F \rangle, \langle s \rangle, N$$

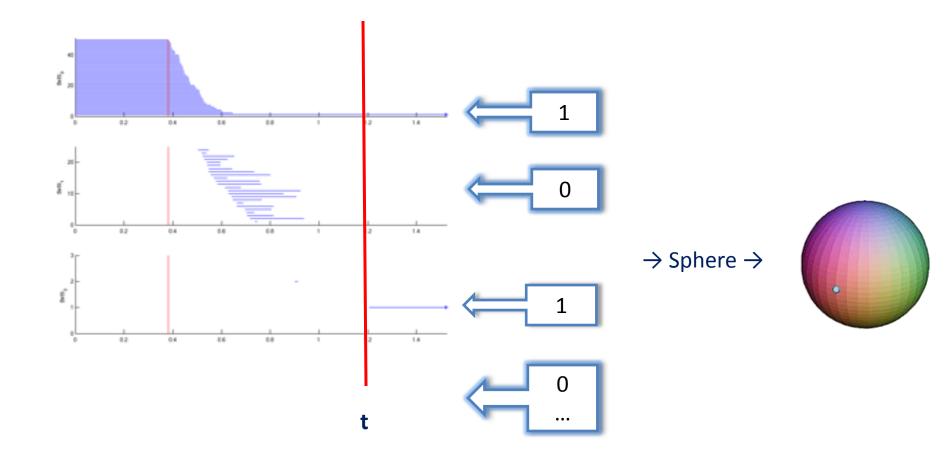






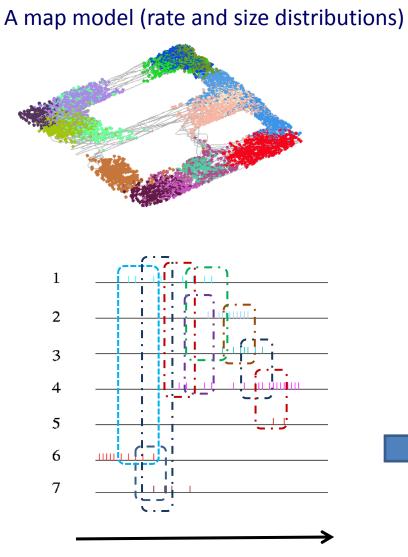


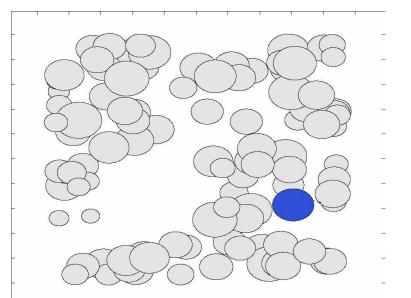
#### Topological barcode of a sphere

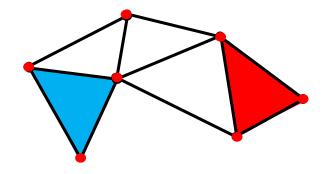


Topological barcode (1, 0, 1, 0, 0 ...)

### Computational test of a map's





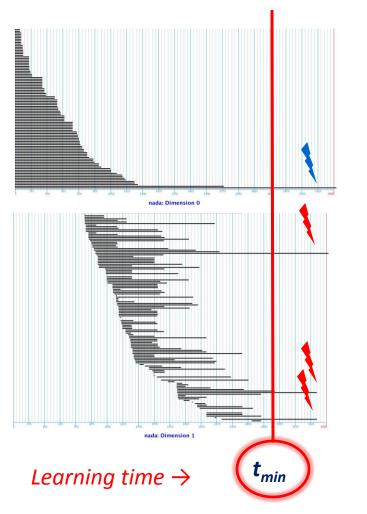


time

### Computational test of a map's

# A map model (rate and size distributions) time

#### Persistent homology, loop dynamics



D

1D

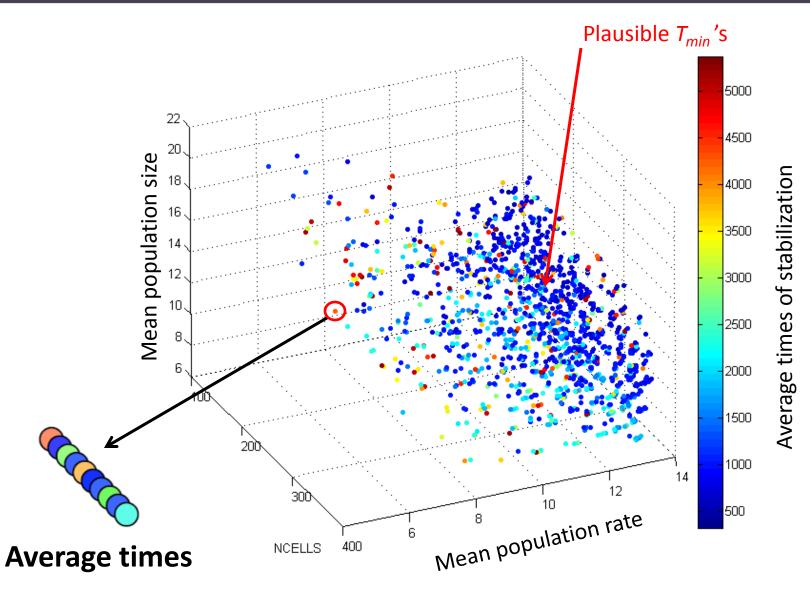
#### Learning time via map parameters

# $T_{\min} = T_{\min}\left(\langle F \rangle, \langle s \rangle, N\right)$

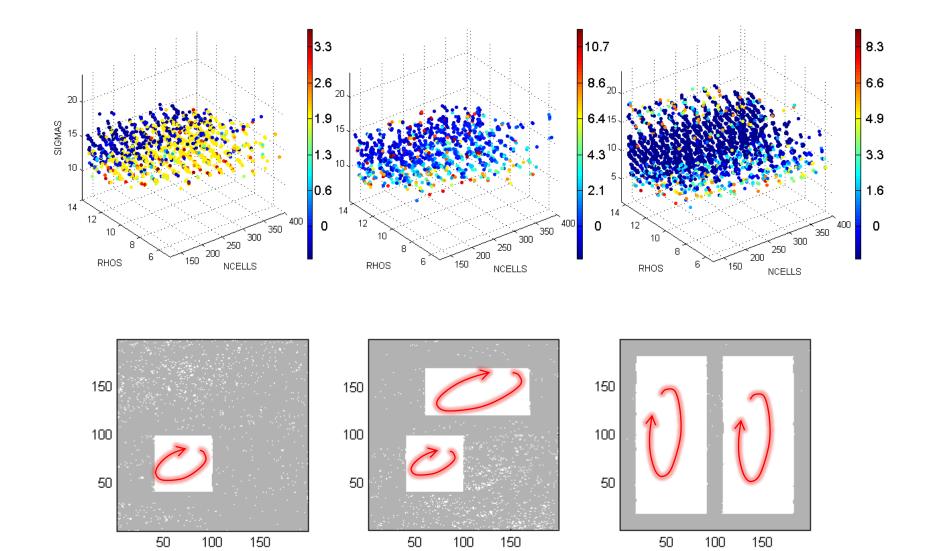
Learning time depends on map parameters

a. firing rate,  $\langle F \rangle$ b. place field size,  $\langle S \rangle$ c. number of cells, *N* 

# Minimal learning times $T_{min} = T_{min} \left( \langle F \rangle, \langle s \rangle, N \right)$

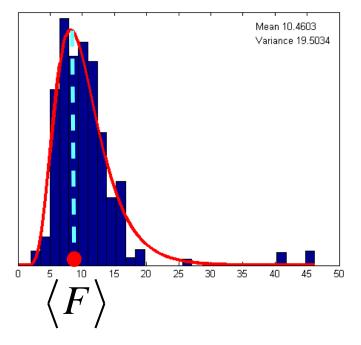


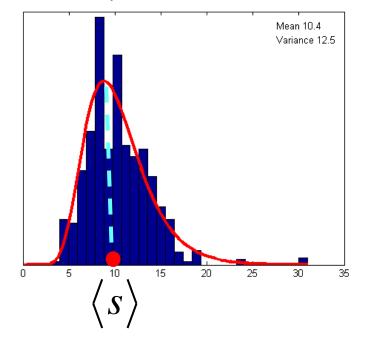
#### Minimal learning times $T_{min}$ , different arenas



#### Place cell ensemble activity parameters

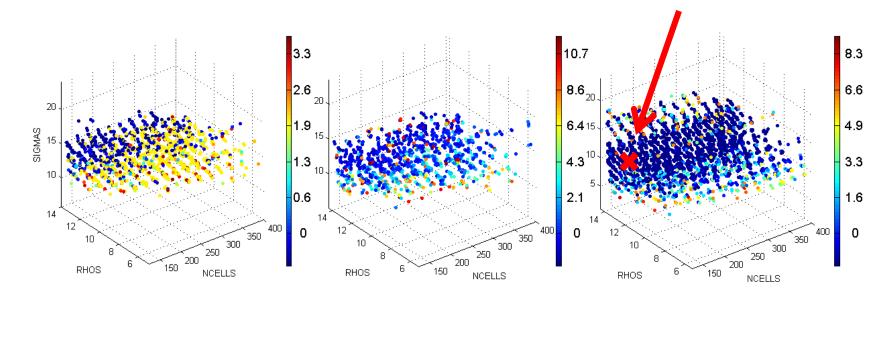


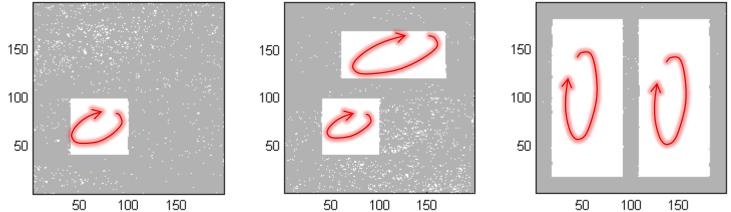




#### place field sizes

# Minimal learning times, T<sub>min</sub>

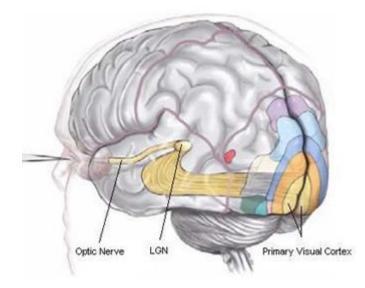


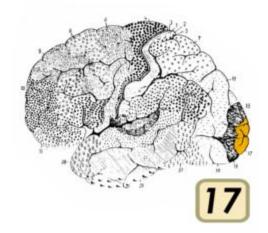


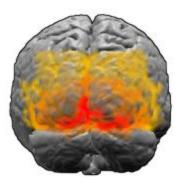
#### Predictive power of neuronal topology

- 1. Which parameters affect the map?
- 2. How does  $t_{min}$  change with the biological parameters?
- 3. How does  $t_{min}$  change with the geometry of the environment?
- 4. What is the "stability range" of the firing activity?
- 5. etc ...

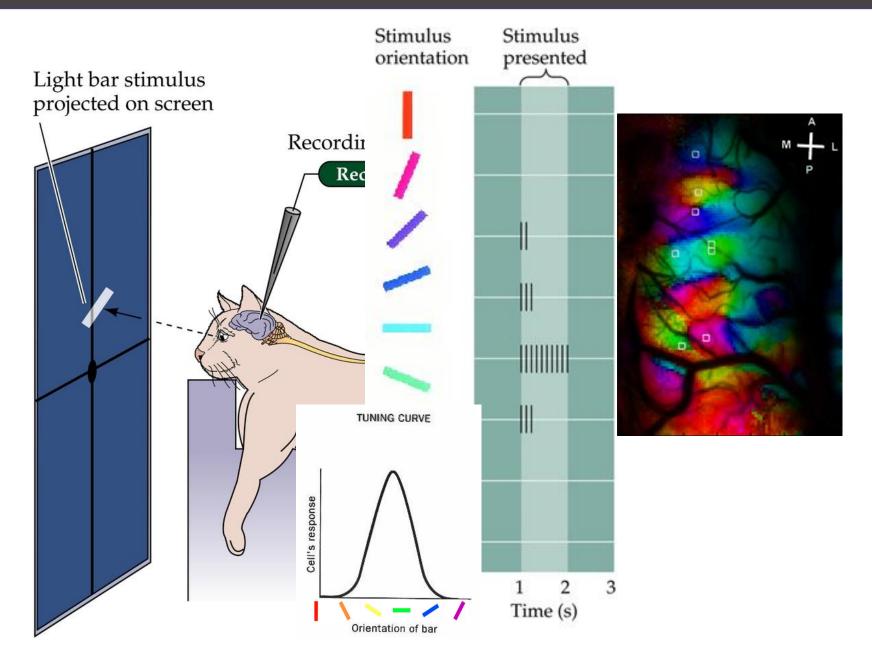
#### Visual cortex

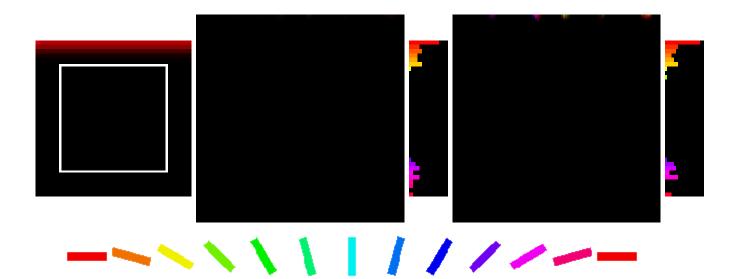




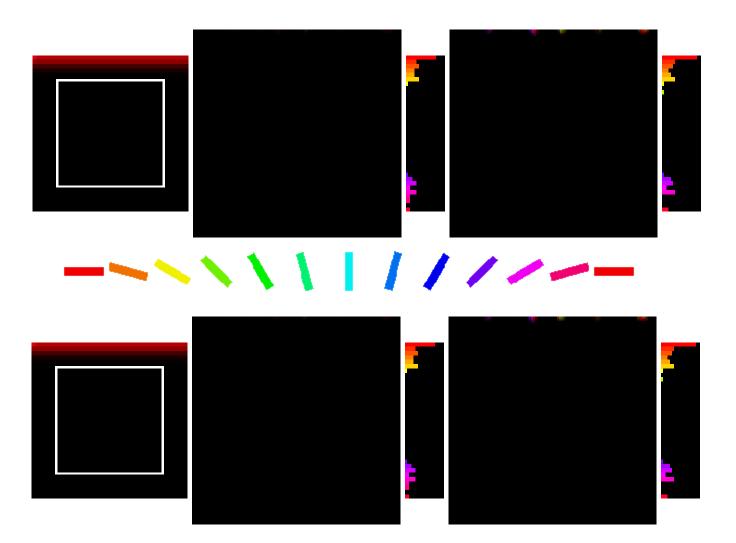


#### Visual cortex – orientation selectivity



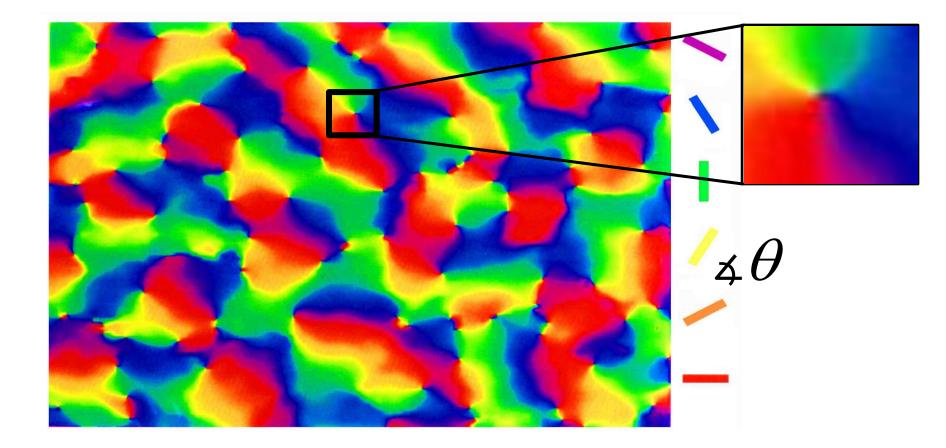


http://homepages.inf.ed.ac.uk/jbednar/index.html



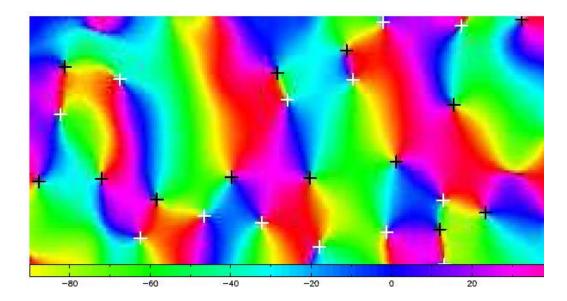
http://homepages.inf.ed.ac.uk/jbednar/index.html

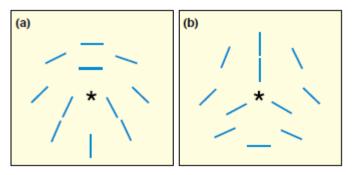
#### Visual cortex – orientation selectivity



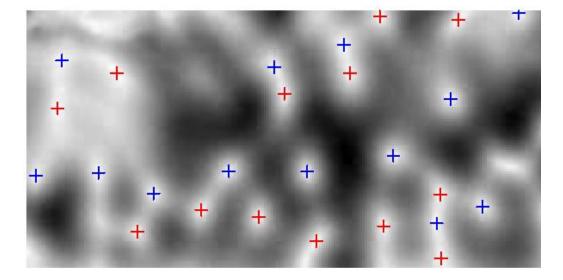
 $\theta = f(x, y)$ 

#### Director field with topological singularities





N. Swindale, 1996



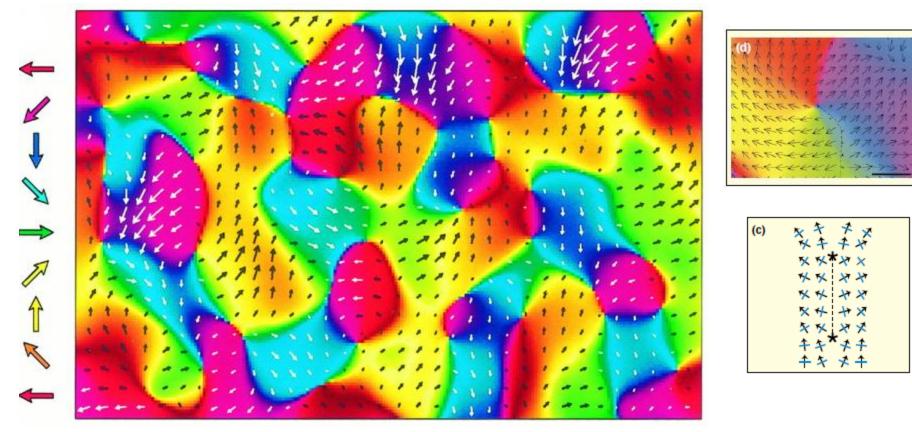
Topological charges of the director field singularities

Courtesy of V. Kalatsky, UH



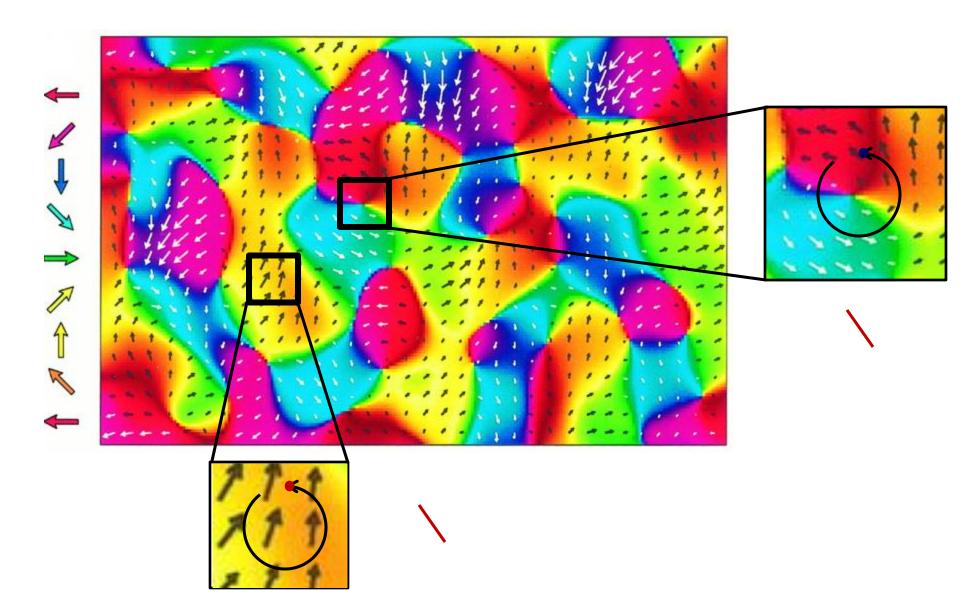
http://homepages.inf.ed.ac.uk/jbednar/index.html

#### Vector field with topological singularities

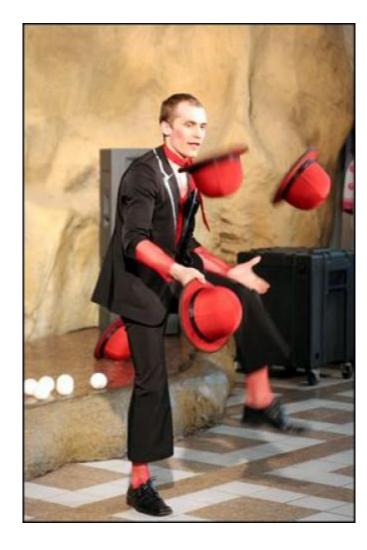


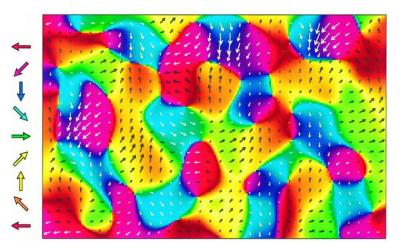
N. Swindale, 1996

#### Vector field with topological singularities



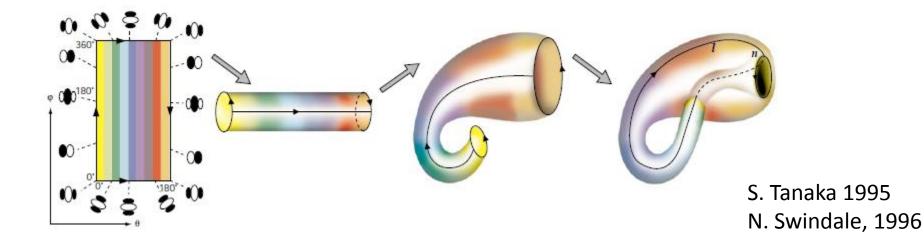
#### Dynamics of the vector field





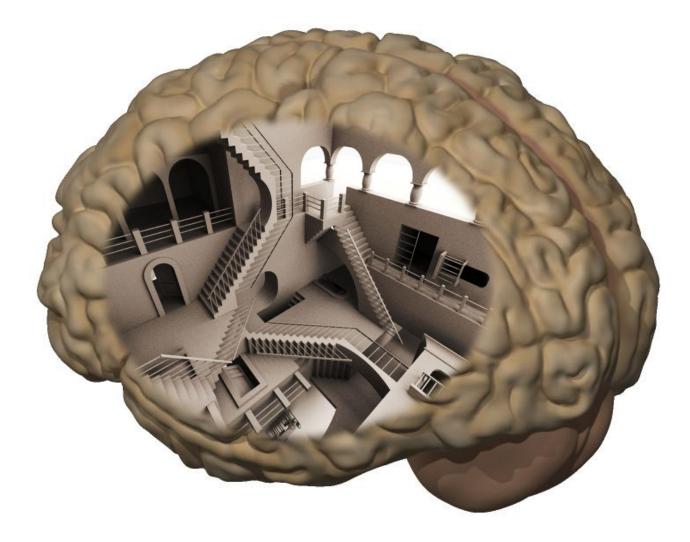
# Why so much geometry? How is it used???

### Topological structure of the receptive fields

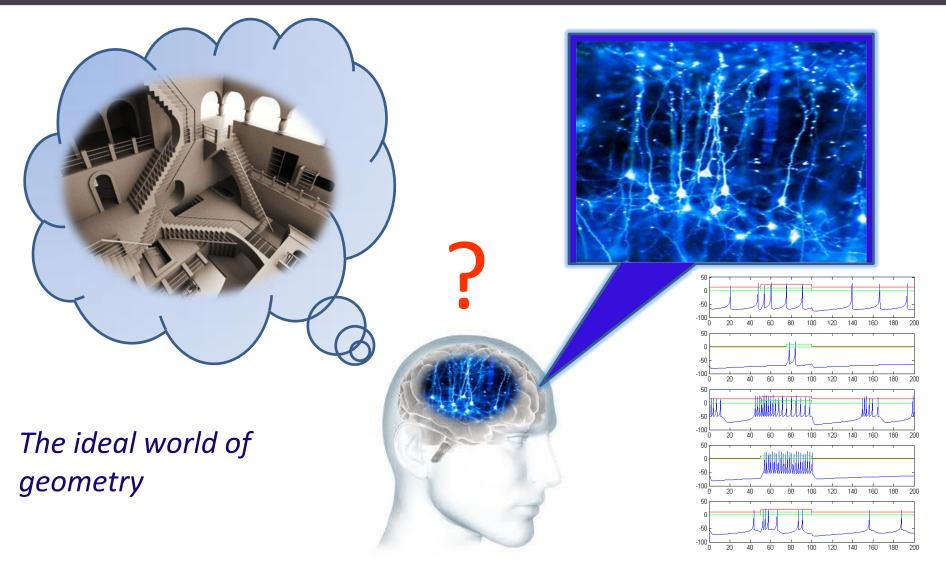


Why so much topology? How is it used???

# Neuroscience and geometry



# Neuroscience and geometry



Time code machine

# Neuroscience and geometry

...Quapropter bono christiano, sive mathematici, sive quilibet impie divinantium, maxime dicentes vera, cavendi sunt, ne consortio daemoniorum animam deceptam, pacto quodam societatis irretiant.

St. Augustine, Confessions II, xvii, 37

...Hence, a devout Christian must avoid mathematicians and all impious soothsayers, especially when they tell the truth, for fear of leading his soul into error by consorting with demons and entangling himself with the bonds of such association.

# Acknowledgements

- Steve Cox, Rice University
- Harel Shouval, University of Texas
- Facundo Memoli, Stanford University
- Gunnar Carlsson, Stanford University
- Loren Frank, UCSF
- The Audience

