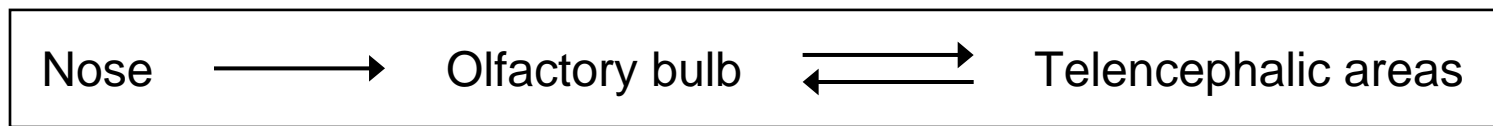
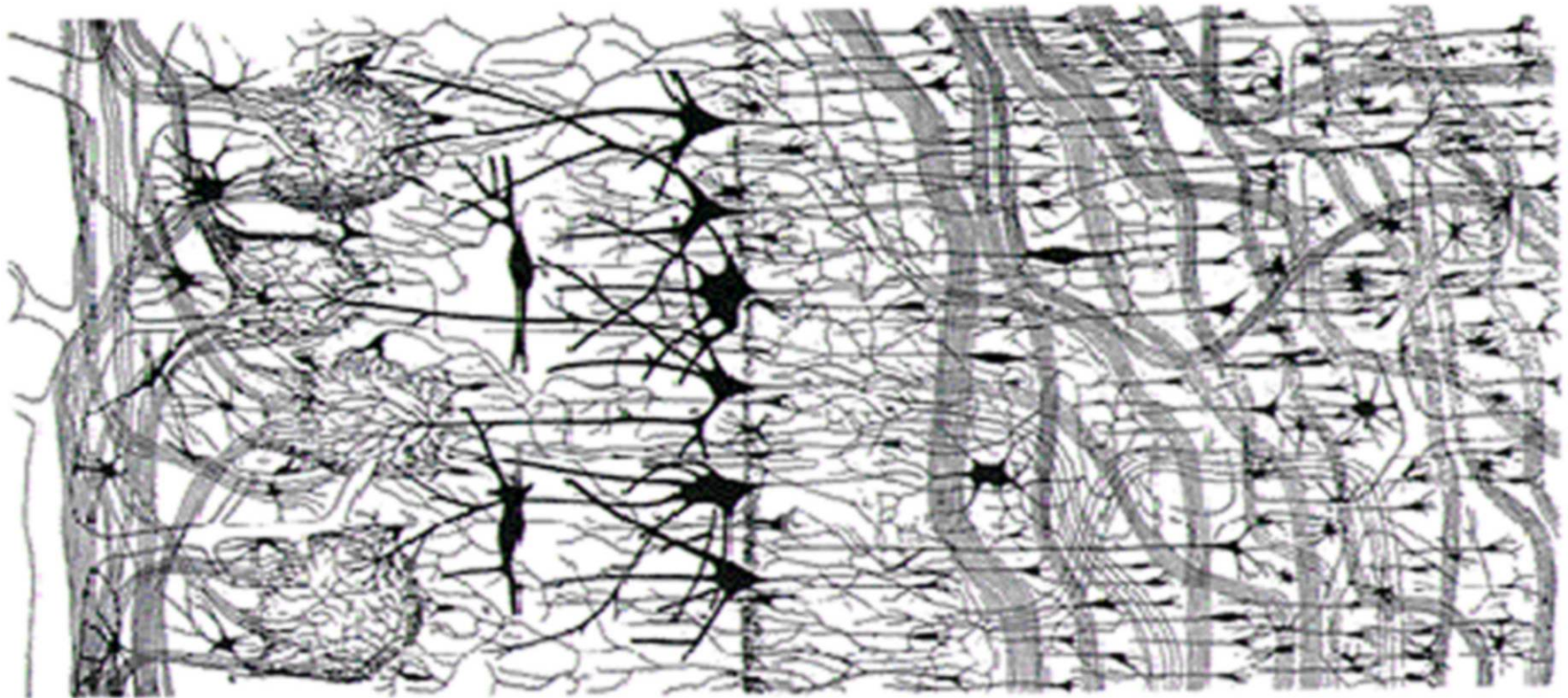


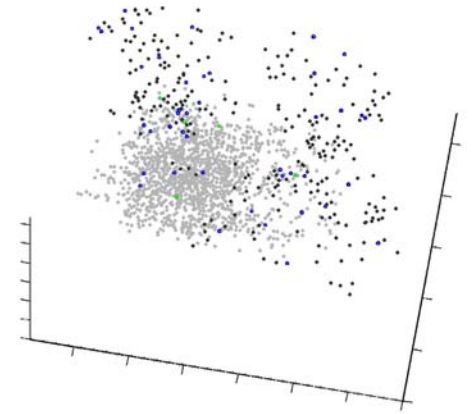
Dynamics and stable states in neuronal circuits of the olfactory system

Rainer Friedrich

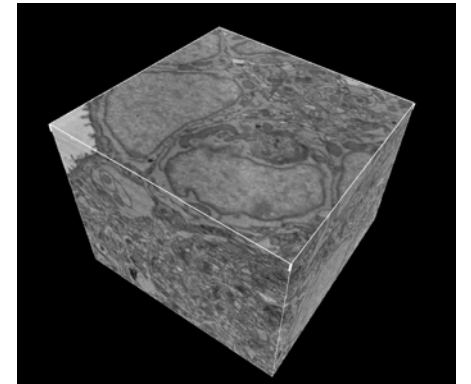


Golgi (1905)

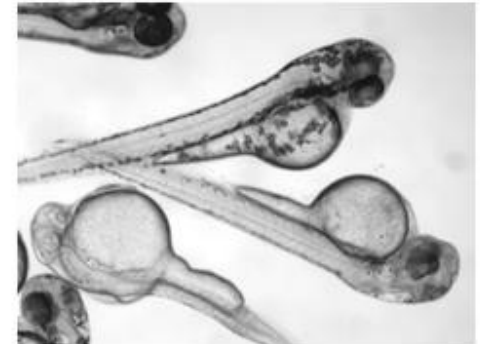
Measurement of neuronal activity patterns



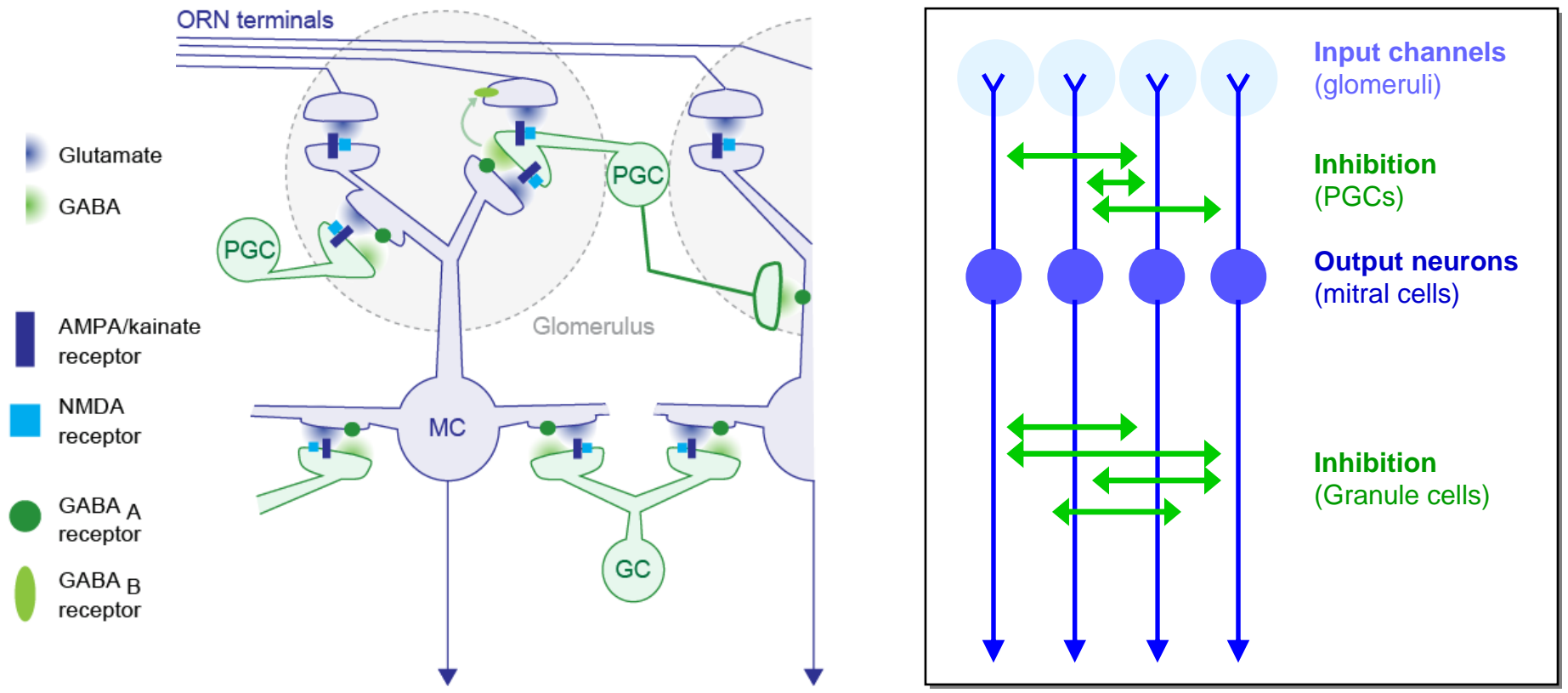
High-resolution optical manipulations



Exhaustive circuit reconstruction (EM)

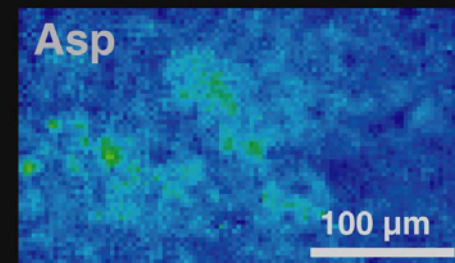
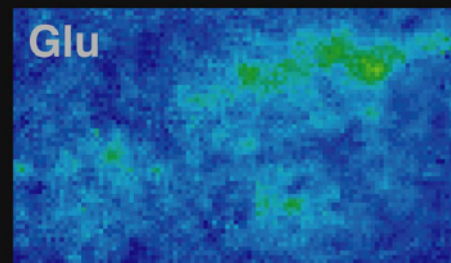
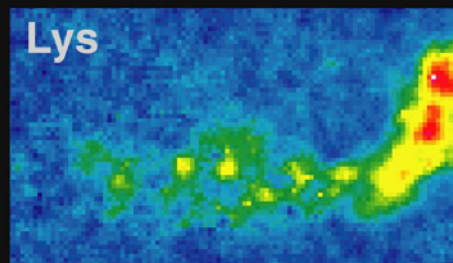
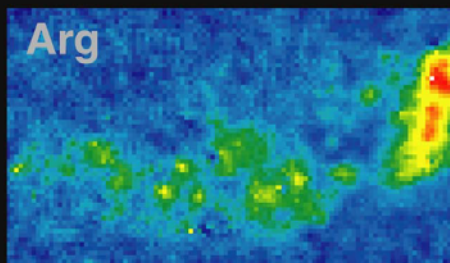
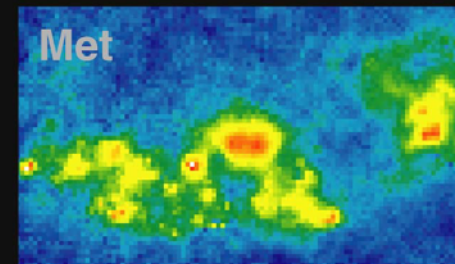
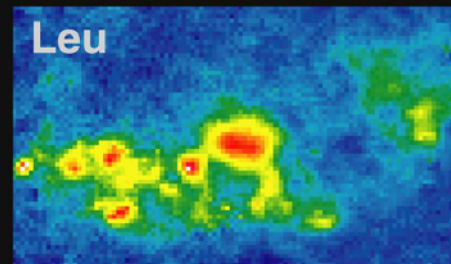
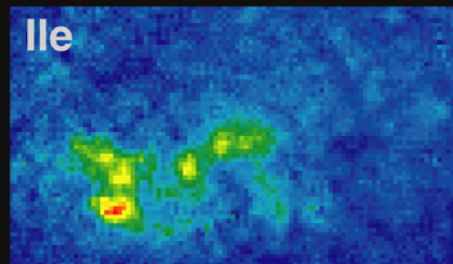
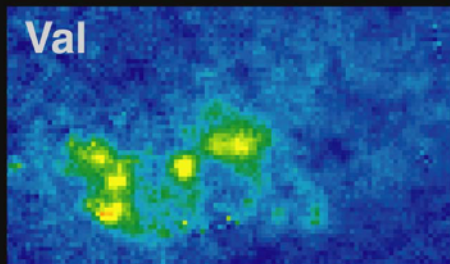
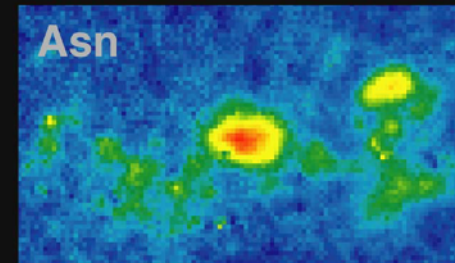
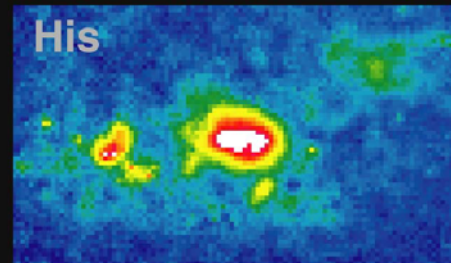
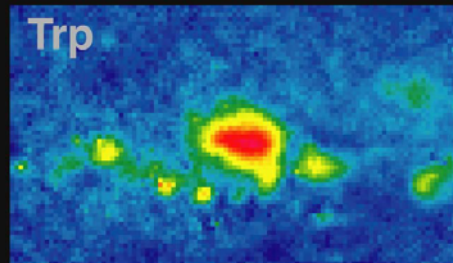
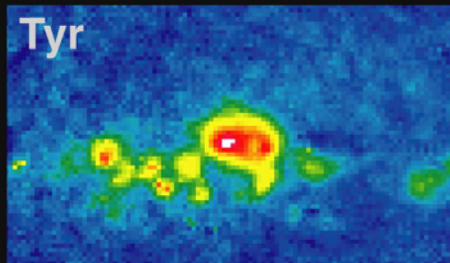
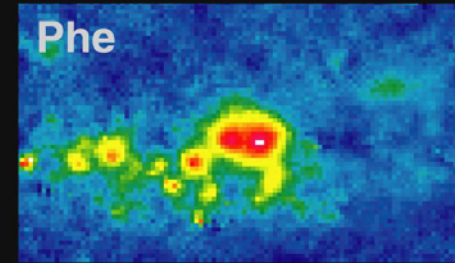
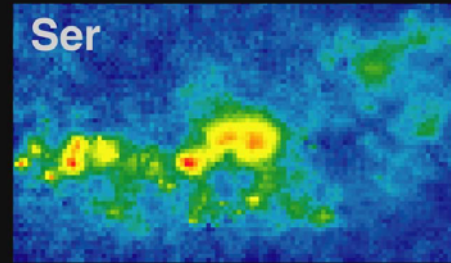
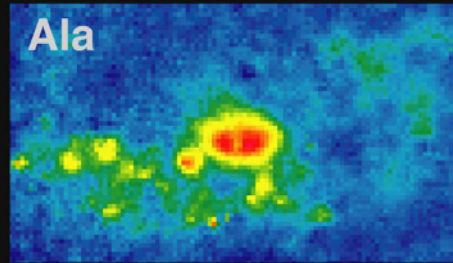
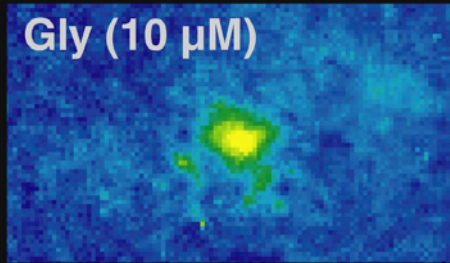
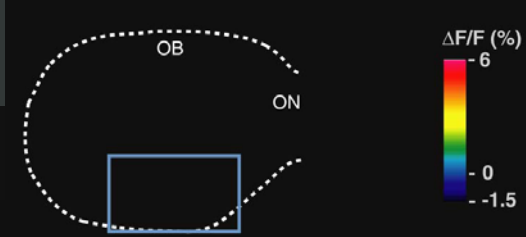


The olfactory bulb



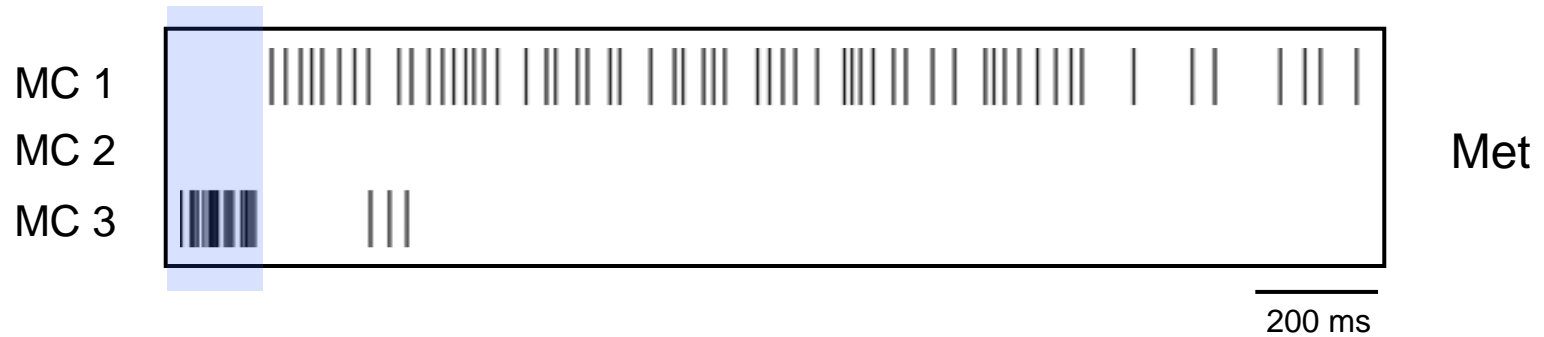
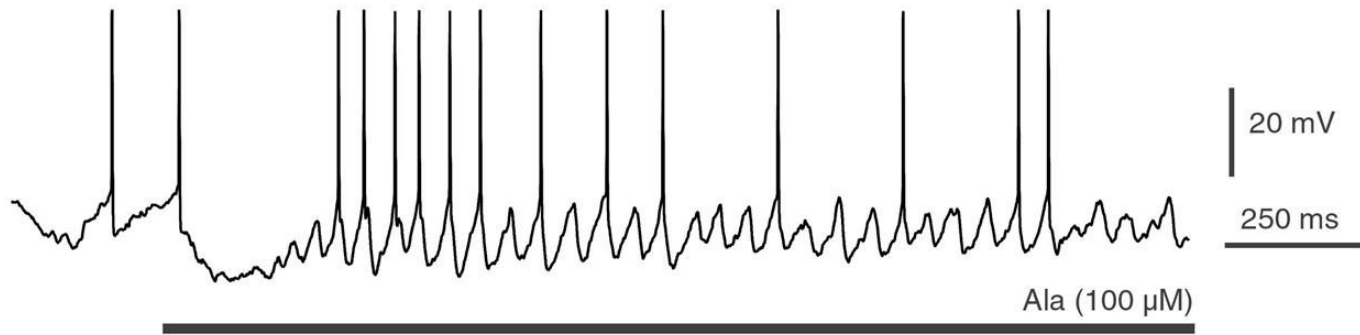
ORN: Olfactory Receptor Neuron; MC: Mitral Cell, GC: Granule Cell; PGC: Periglomerular Cell

Input activity patterns (zebrafish, amino acids)

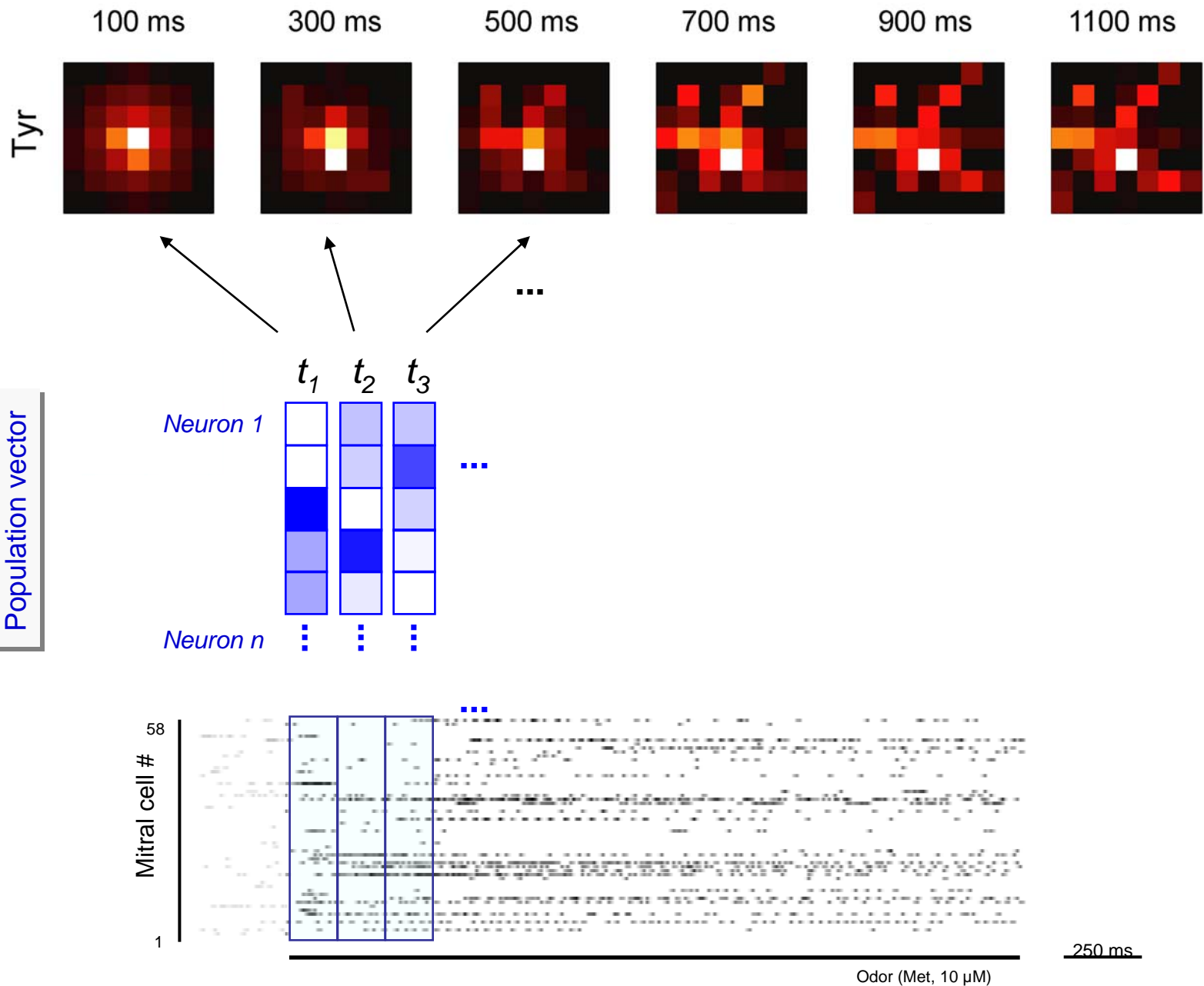


100 μ m

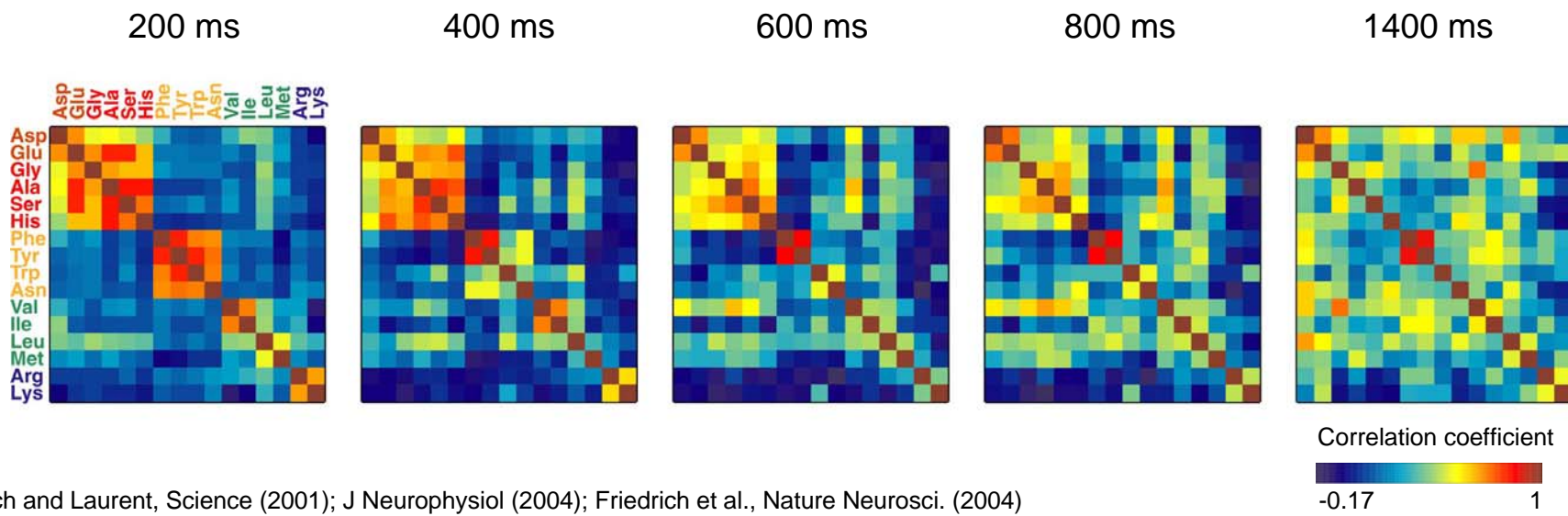
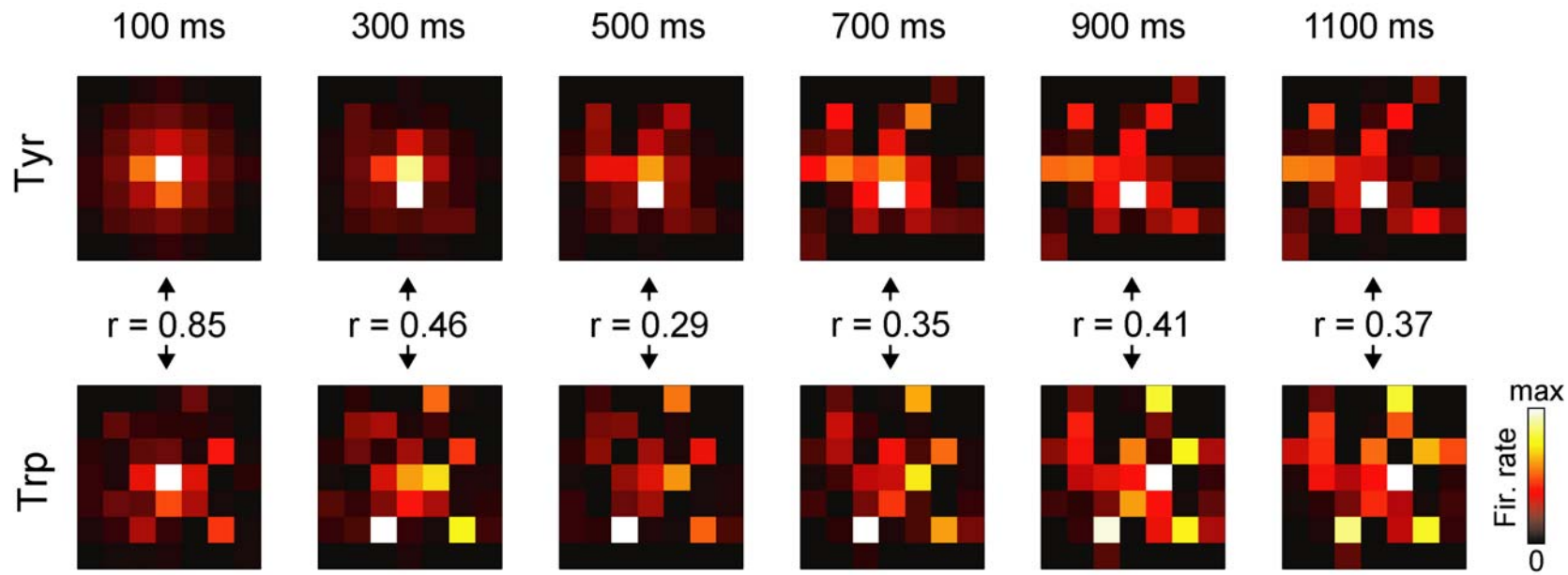
Output activity (mitral cells)



De-correlation of activity patterns

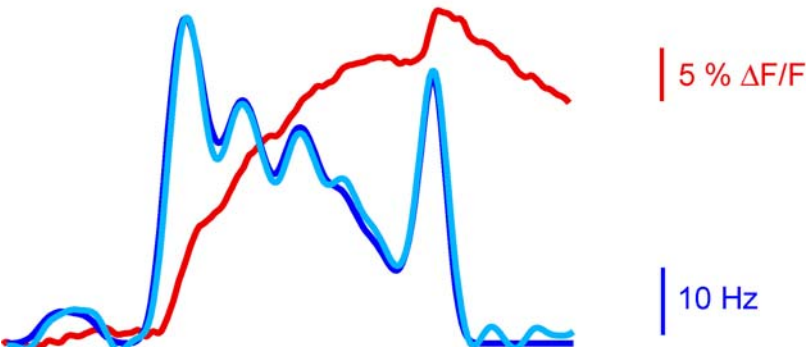
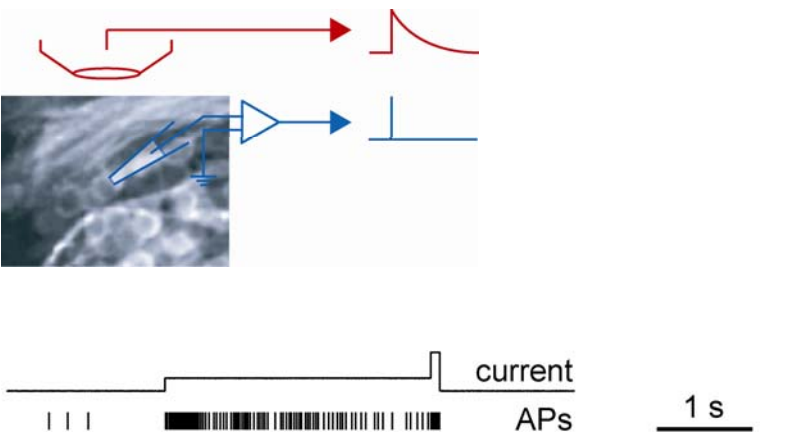


De-correlation of activity patterns

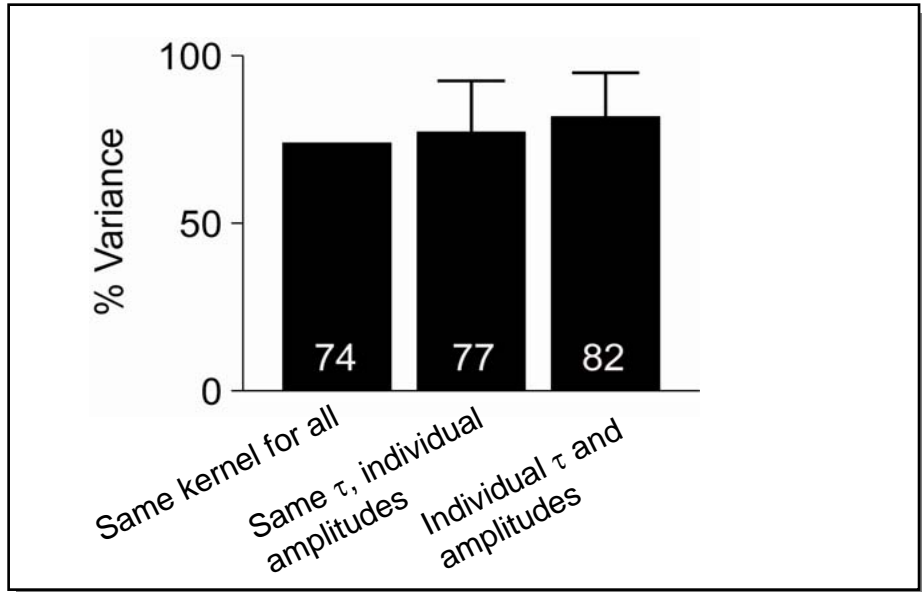
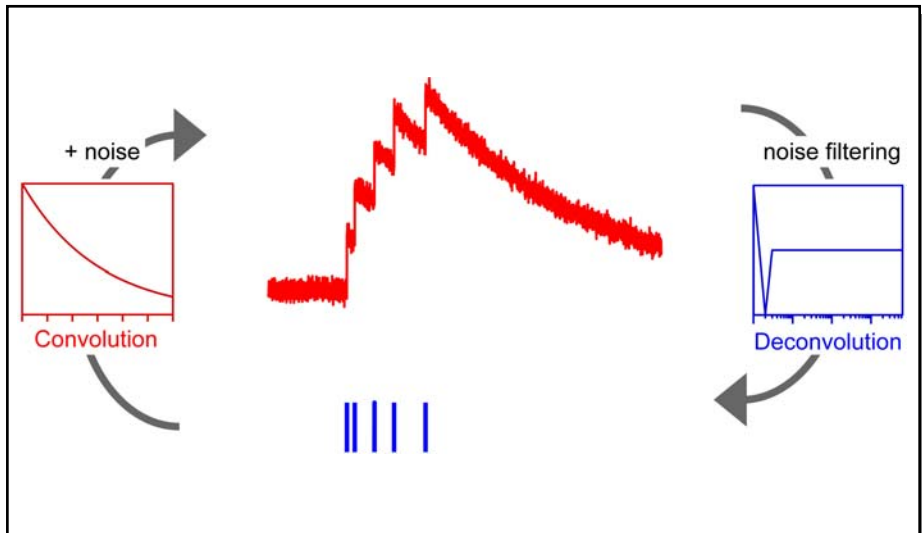


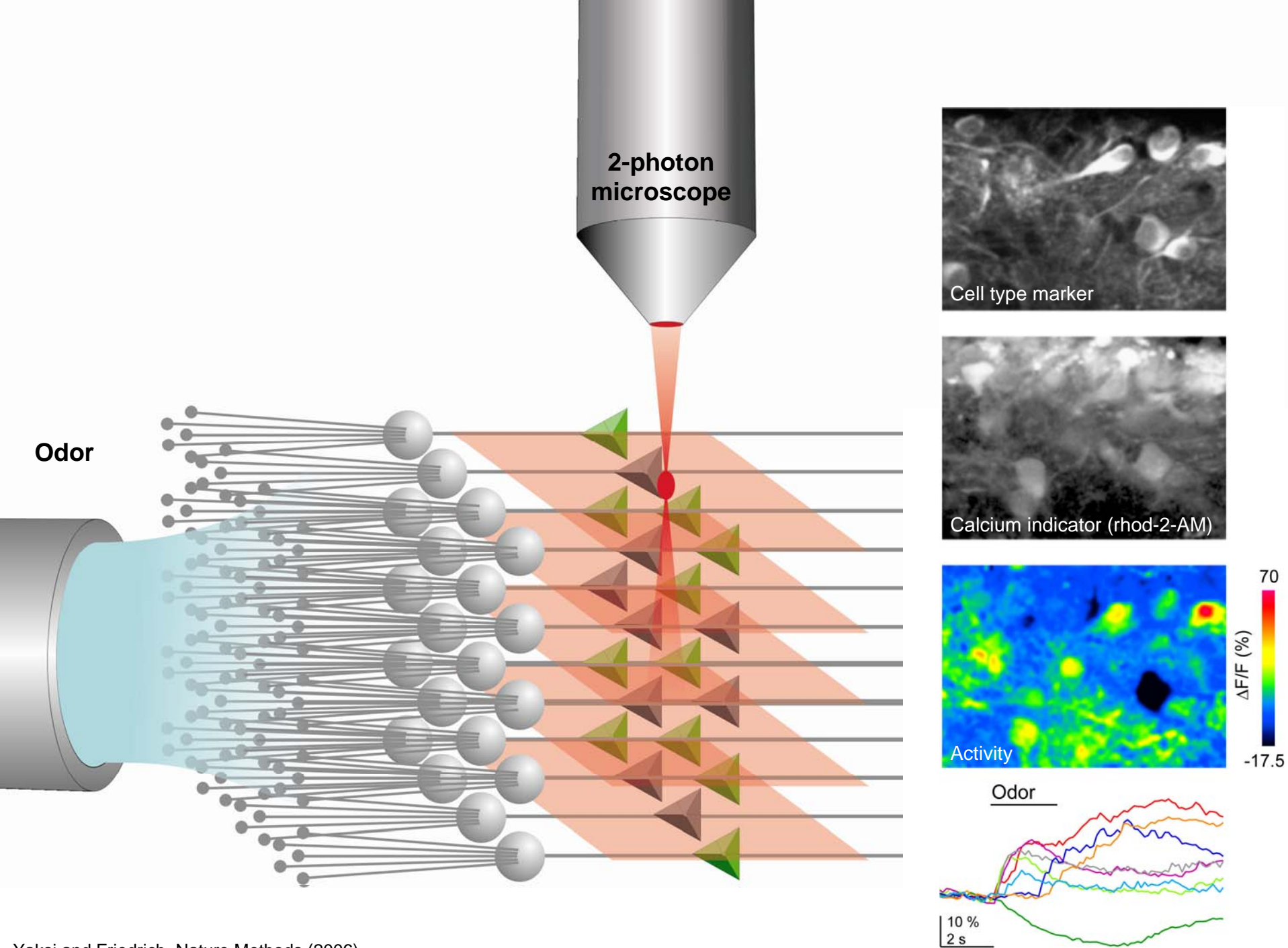
Friedrich and Laurent, Science (2001); J Neurophysiol (2004); Friedrich et al., Nature Neurosci. (2004)

Reconstruction of electrical activity from Ca^{2+} signal by deconvolution

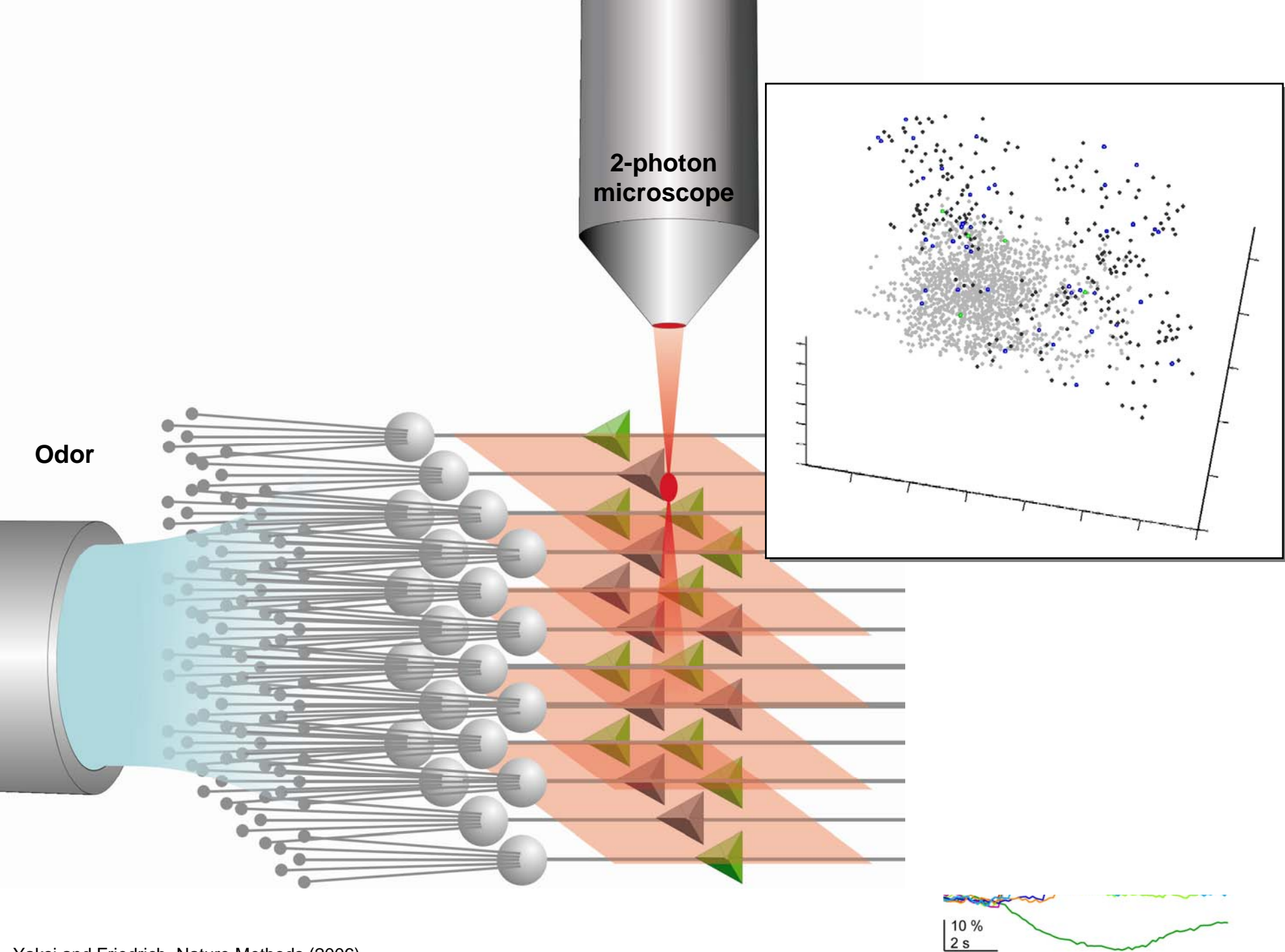


— Raw Ca^{2+} signal
— Firing rate
— Deconvolved Ca^{2+} signal



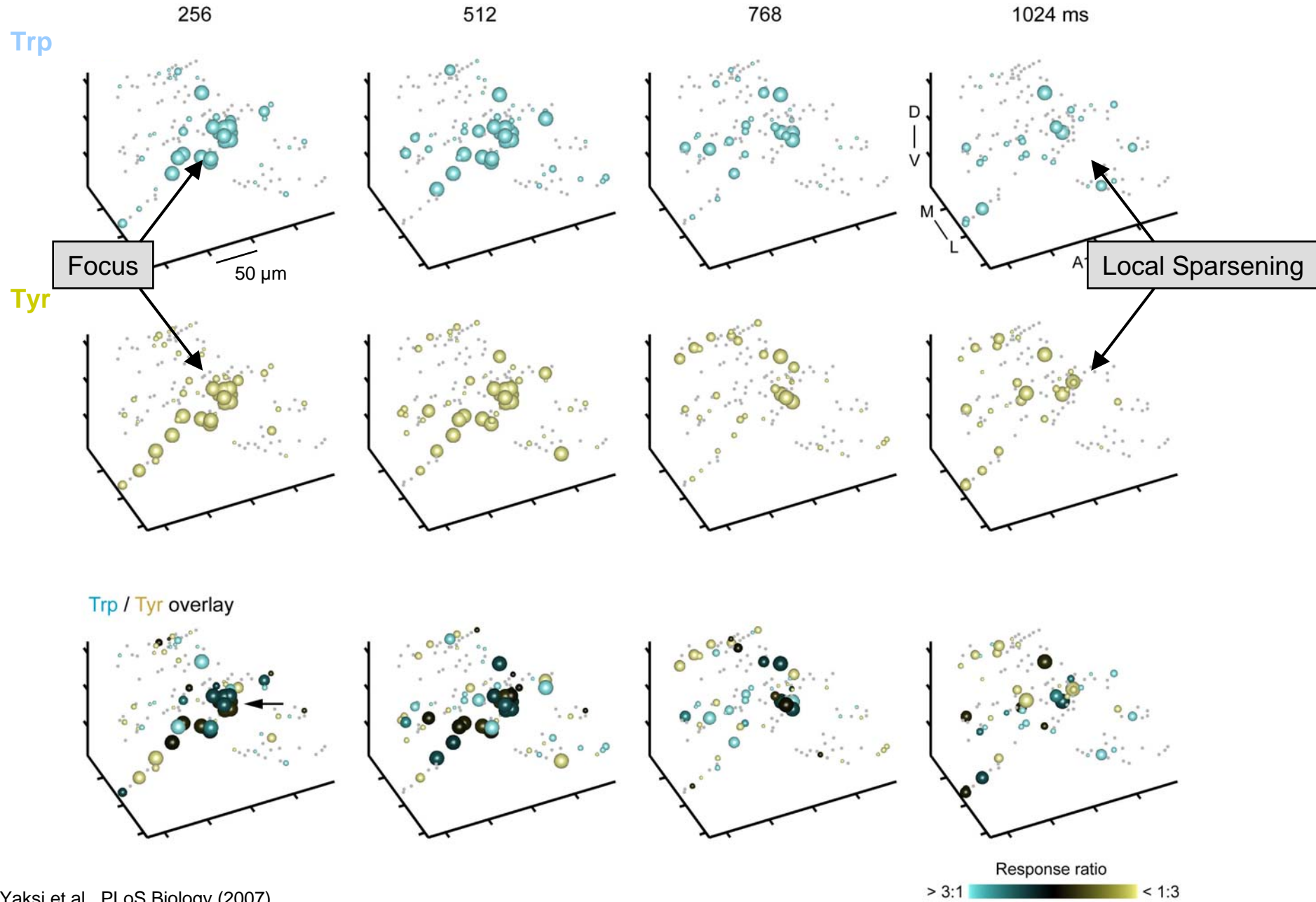


Yaksi and Friedrich, Nature Methods (2006)

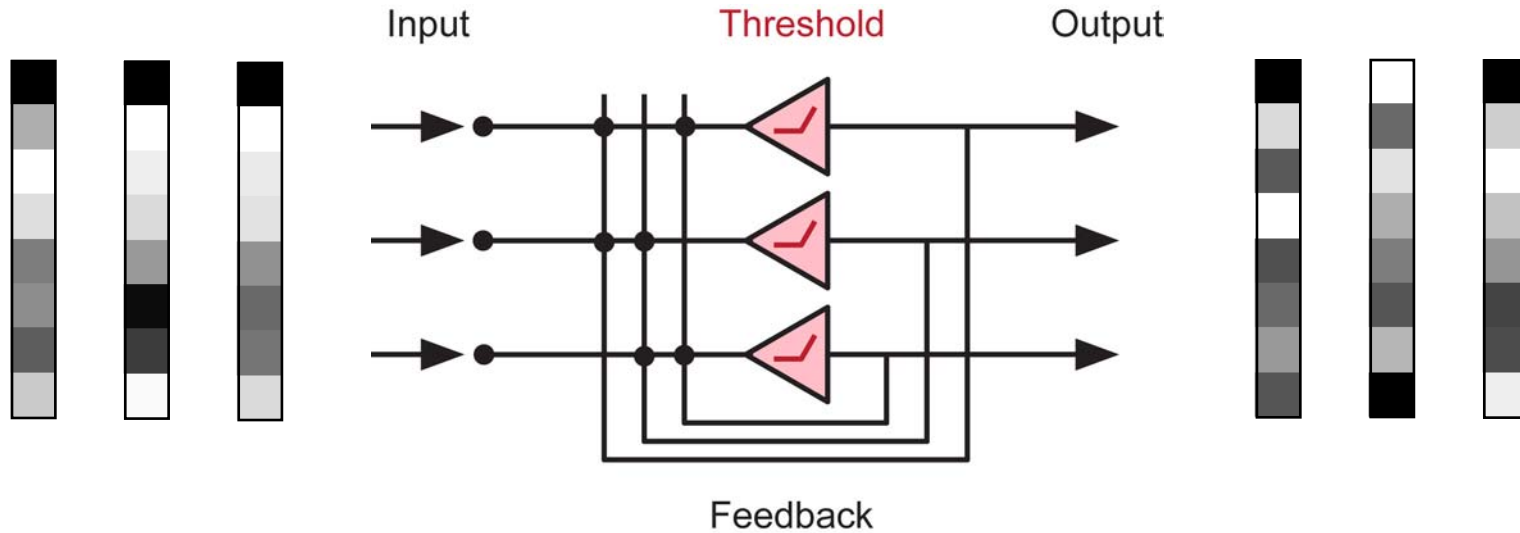


Yaksi and Friedrich, Nature Methods (2006)

Pattern decorrelation by local sparsening



Decorrelation



Channel decorrelation:
→ „Efficient“ coding

- Minimizes neuron number required for transmission

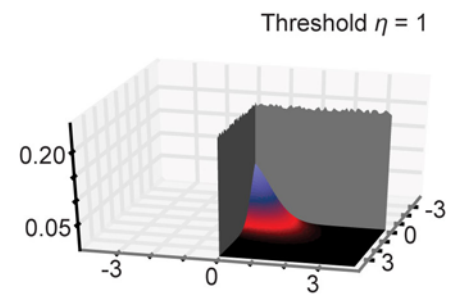
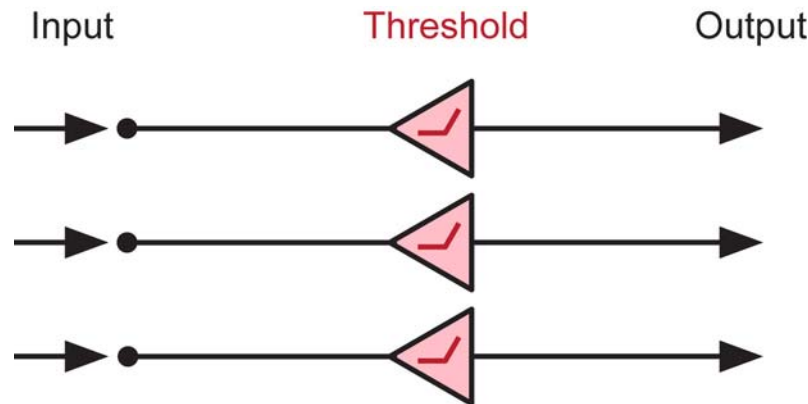
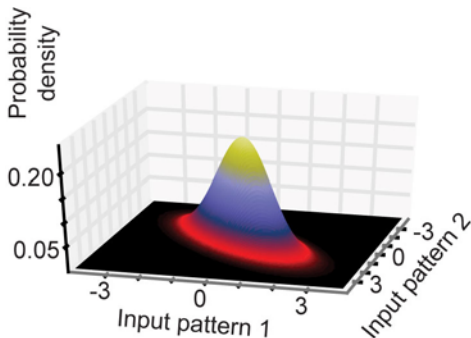
Pattern decorrelation:
→ „Informative“ coding

- Facilitates discrimination
- Important for storage by associative networks

Known strategies are:

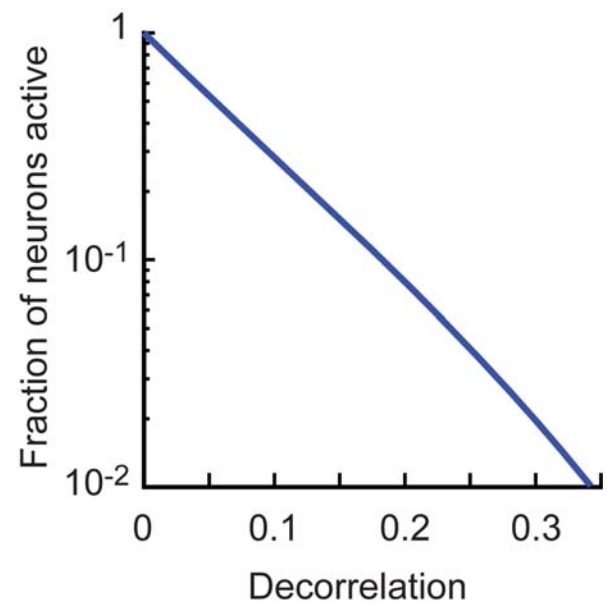
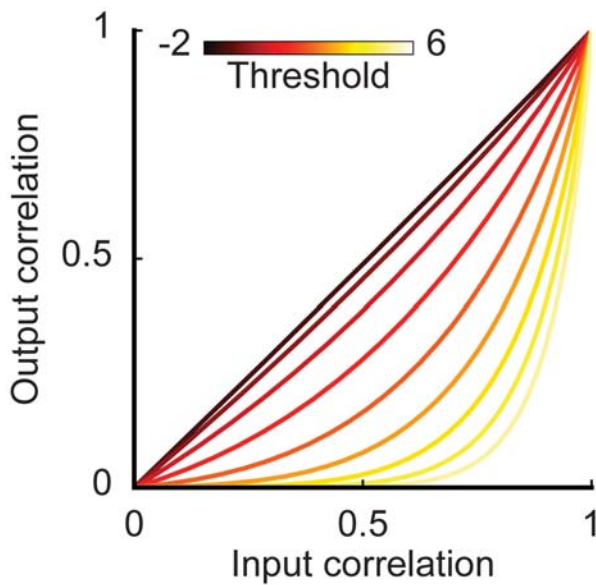
- adaptive (require prior knowledge and training)
- no obvious neuronal implementation

Decorrelation by recurrent networks: mathematical analysis

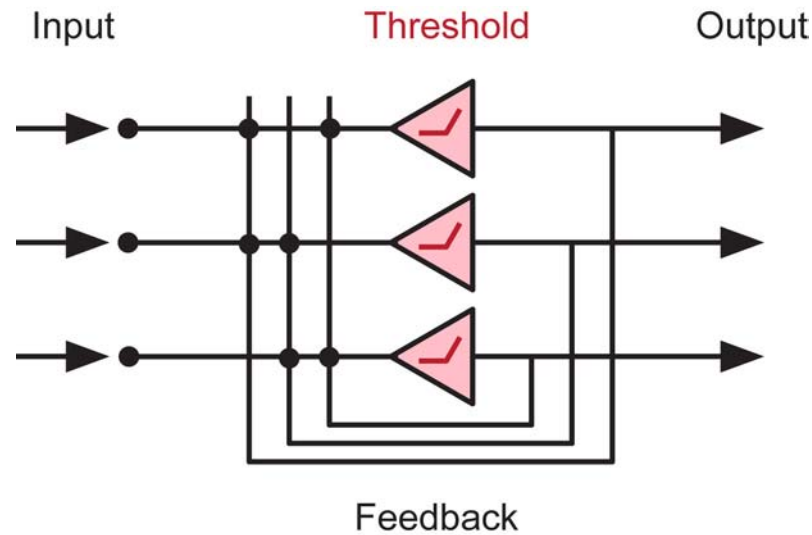


Theorem 1

Thresholding alone invariably causes decorrelation



Decorrelation by recurrent networks: mathematical analysis

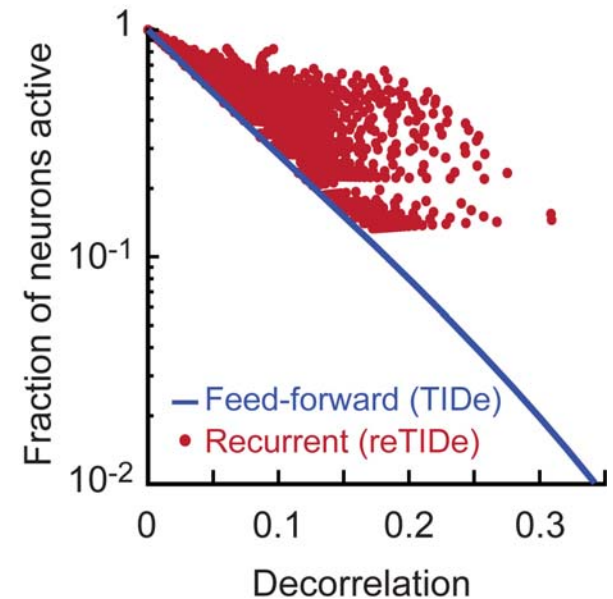


Theorem 1

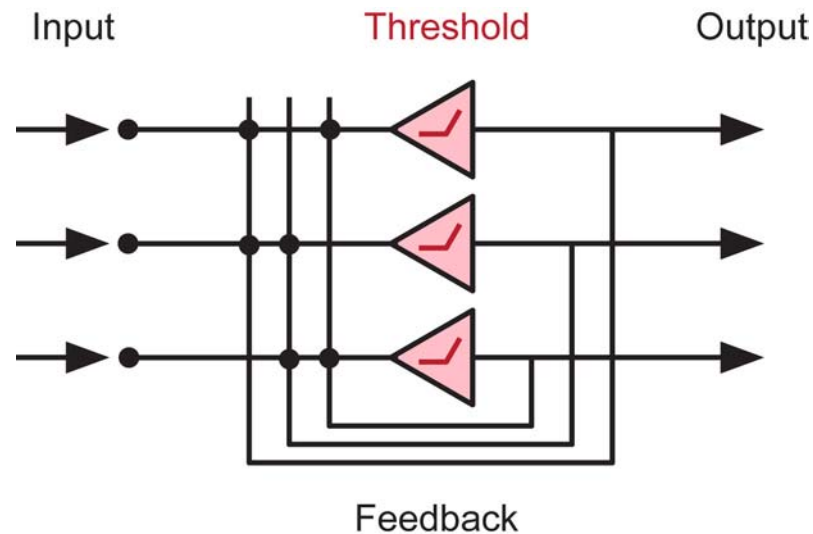
Thresholding alone invariably causes decorrelation

Theorem 2

Recurrent connectivity amplifies threshold-induced decorrelation



Decorrelation by recurrent networks: mathematical analysis



Theorem 1

Thresholding alone invariably causes decorrelation

Theorem 2

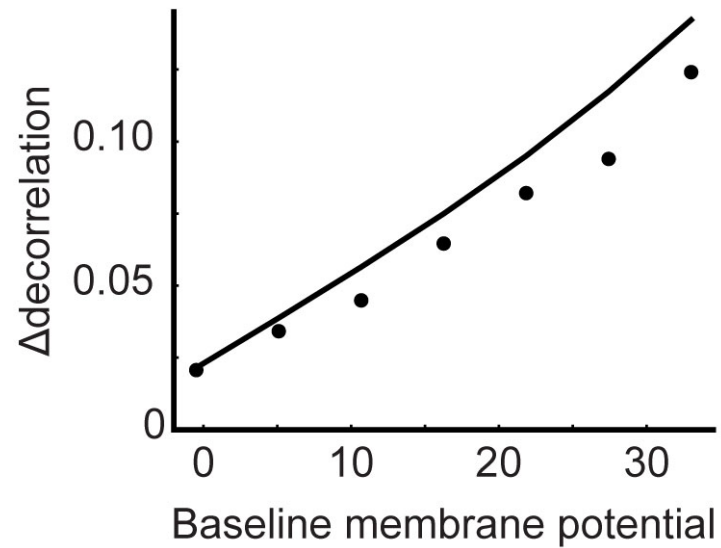
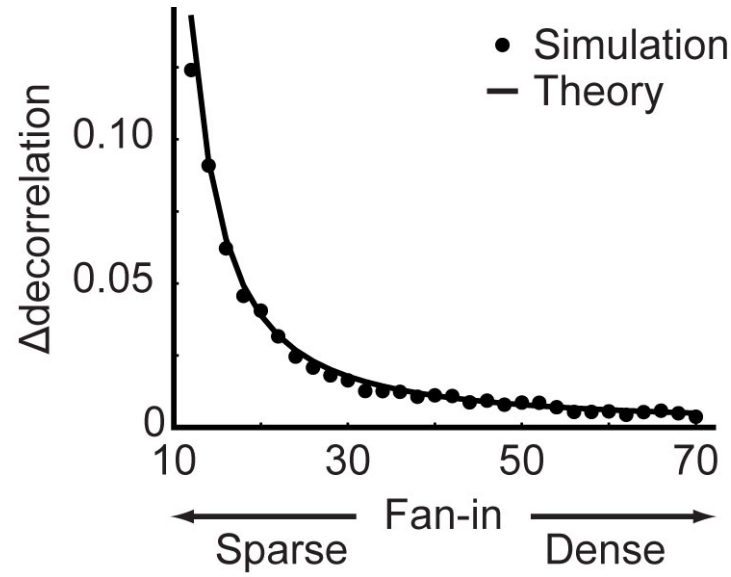
Recurrent connectivity amplifies threshold-induced decorrelation

Theorem 3

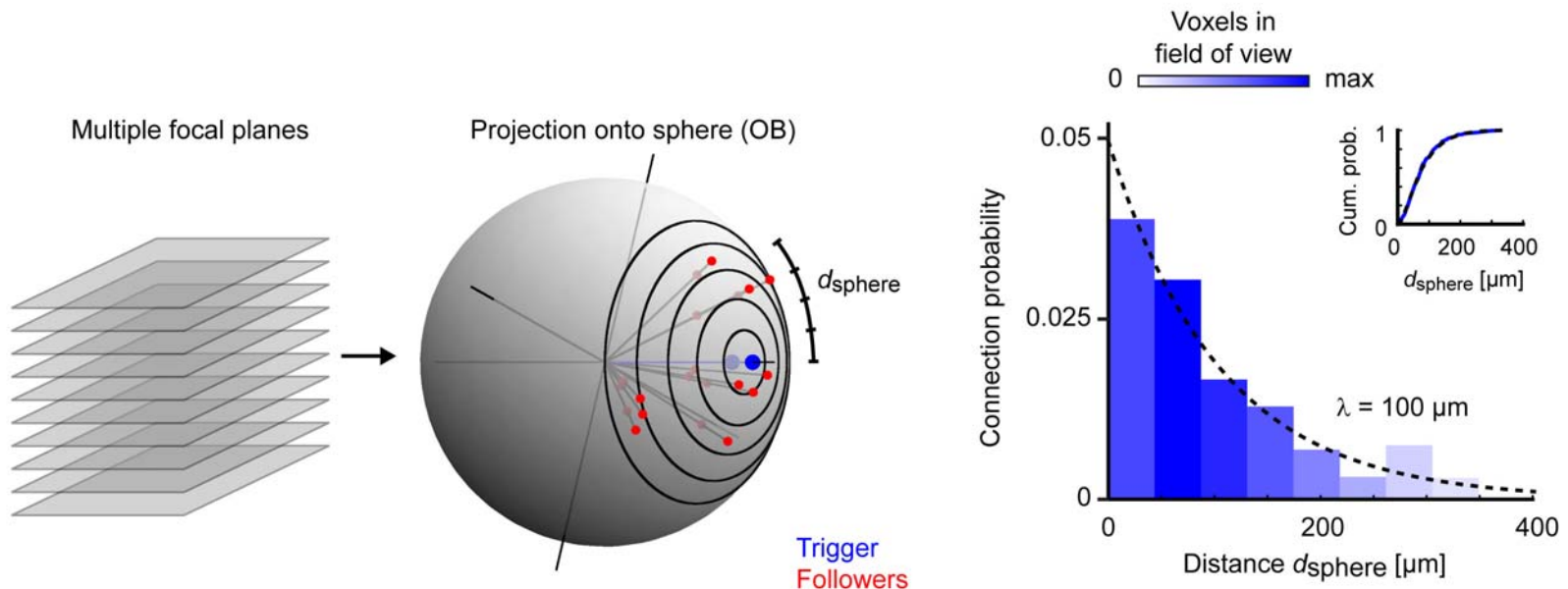
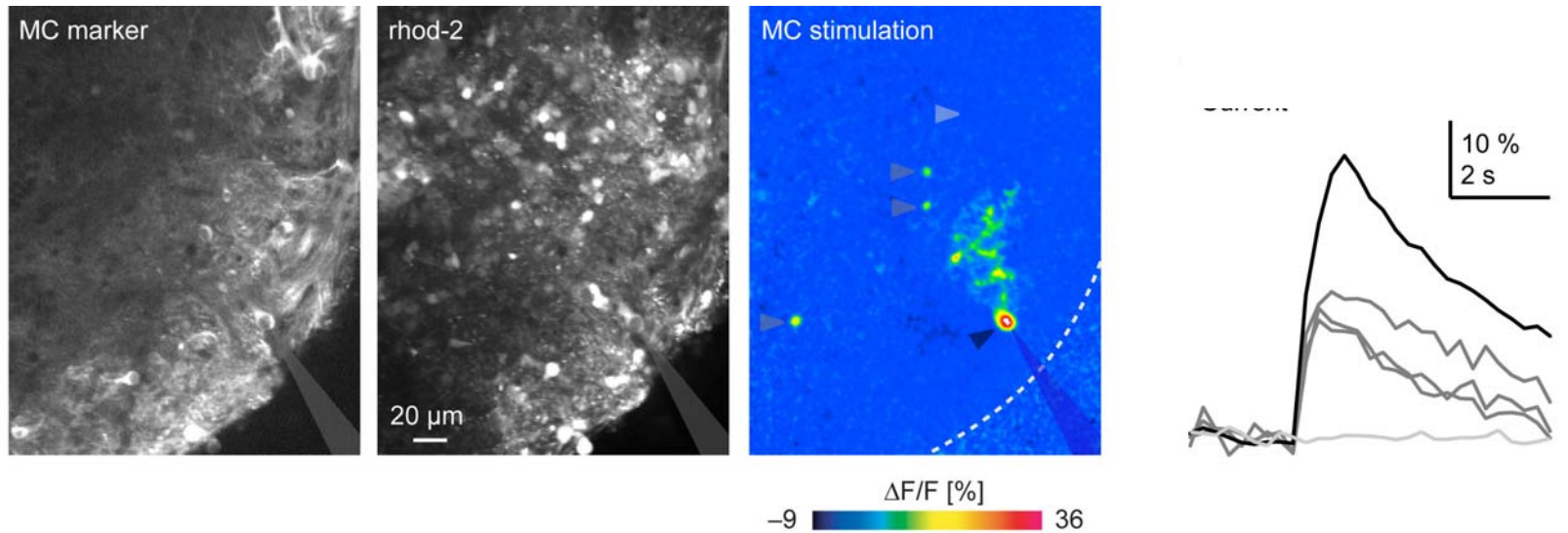
Recurrent-enhanced decorrelation increases with:

- Sparse connectivity
- High baseline activity (when „sufficiently coupled“)

Theoretical predictions and simulation results



Forward optical probing of neuronal connectivity in the olfactory bulb

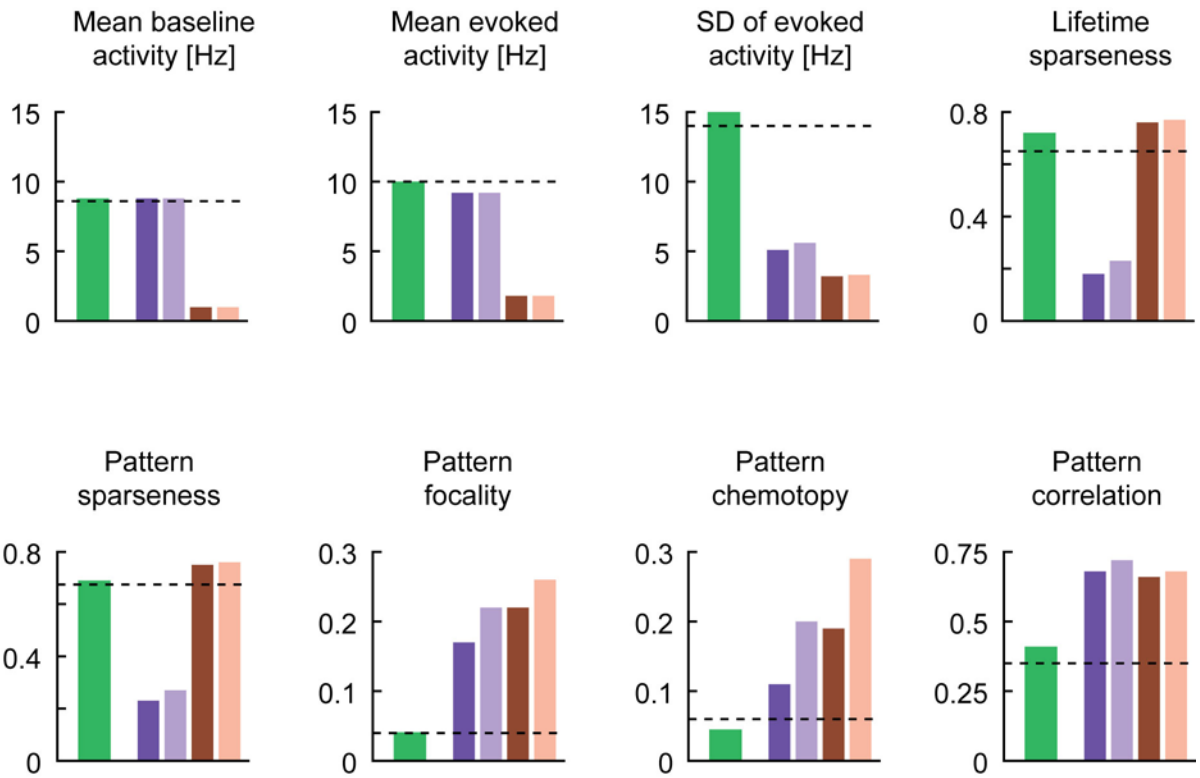
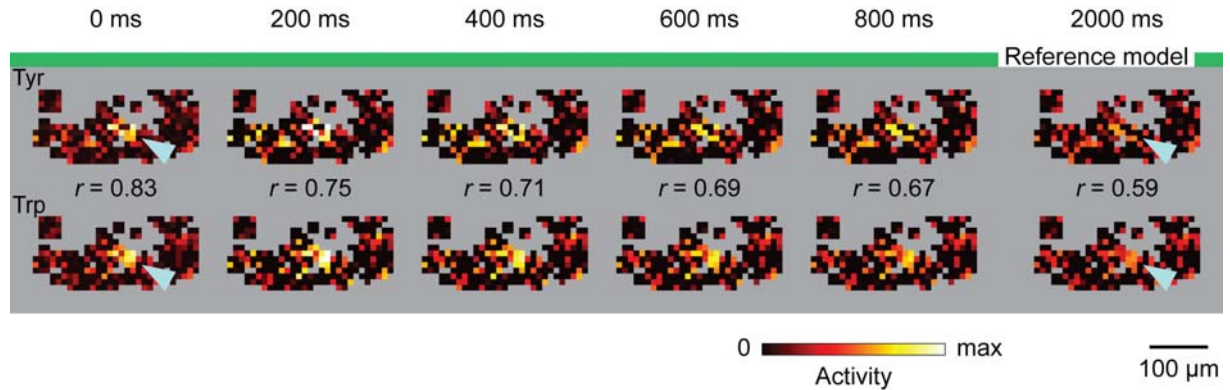
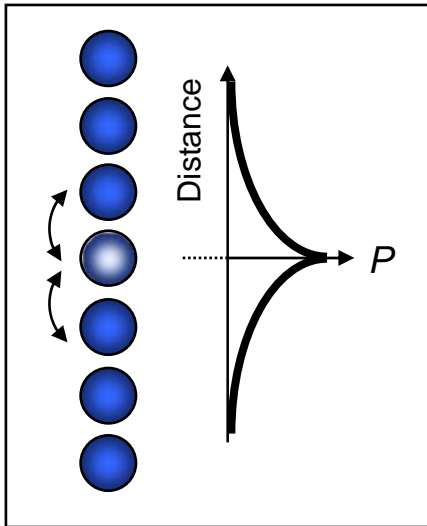


Overall connection probability: 0.0034

Olfactory bulb circuit model

Olfactory bulb model:

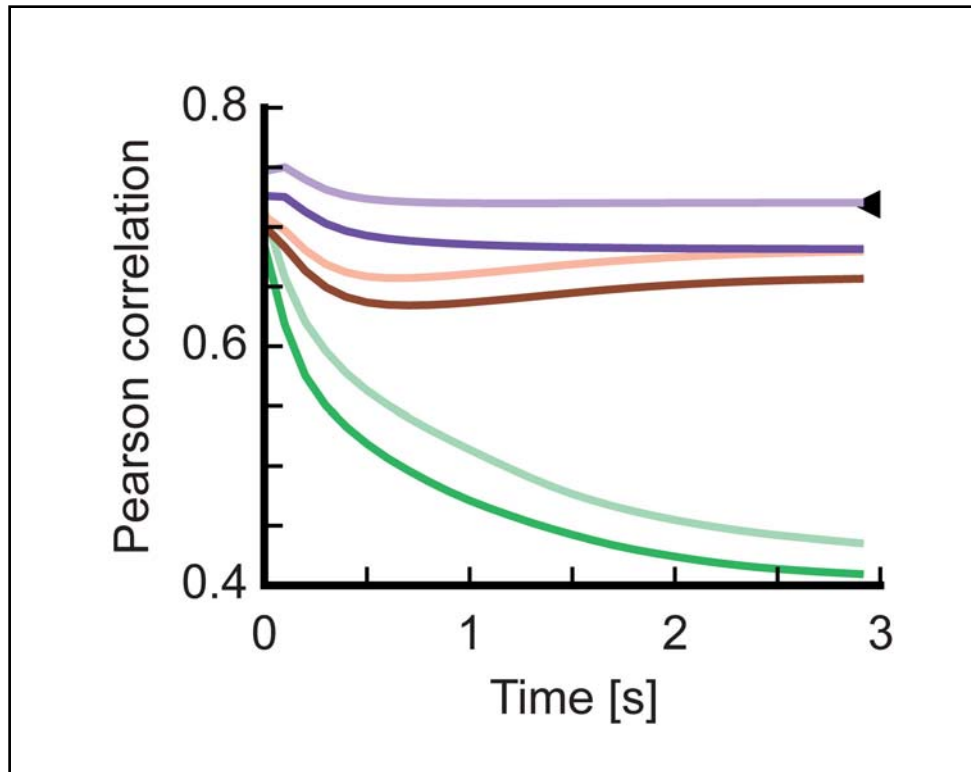
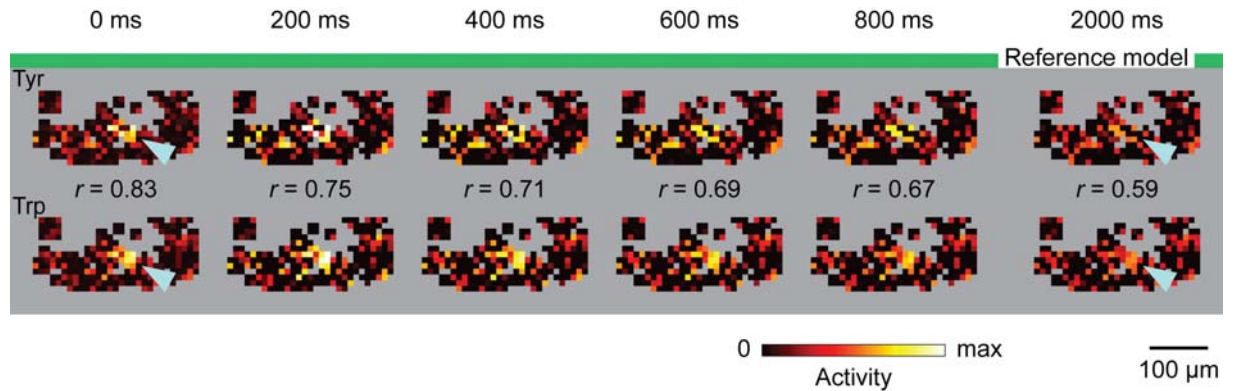
- Recurrent inhibitory connections
- Reciprocal connectivity
- Topographic connectivity
- Naturalistic inputs



Olfactory bulb circuit model

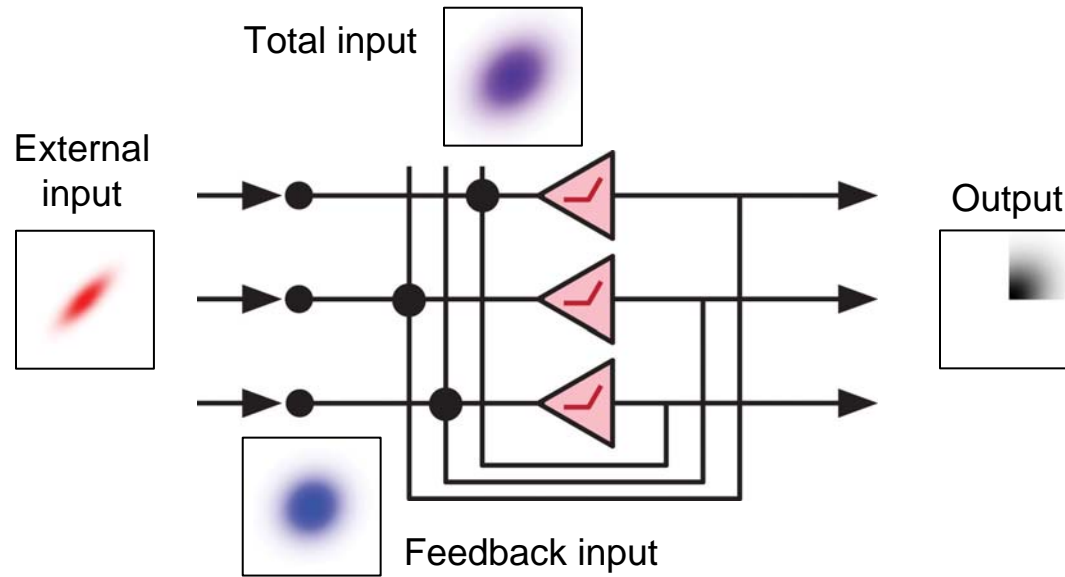
Olfactory bulb model:

- Recurrent inhibitory connections
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- Topographic connectivity
- Naturalistic inputs



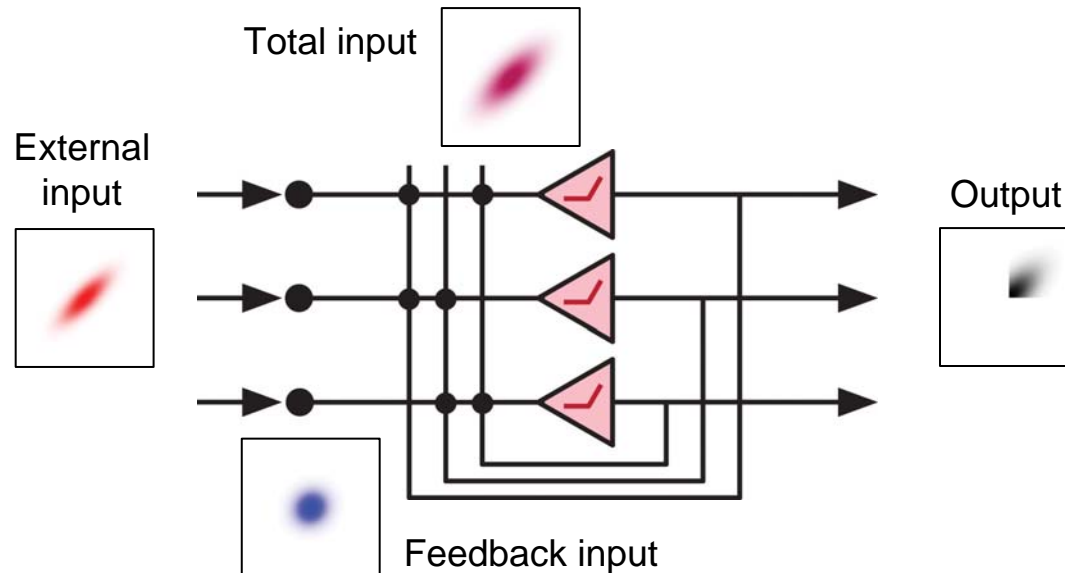
Sparse circuit:

Few but strong connections

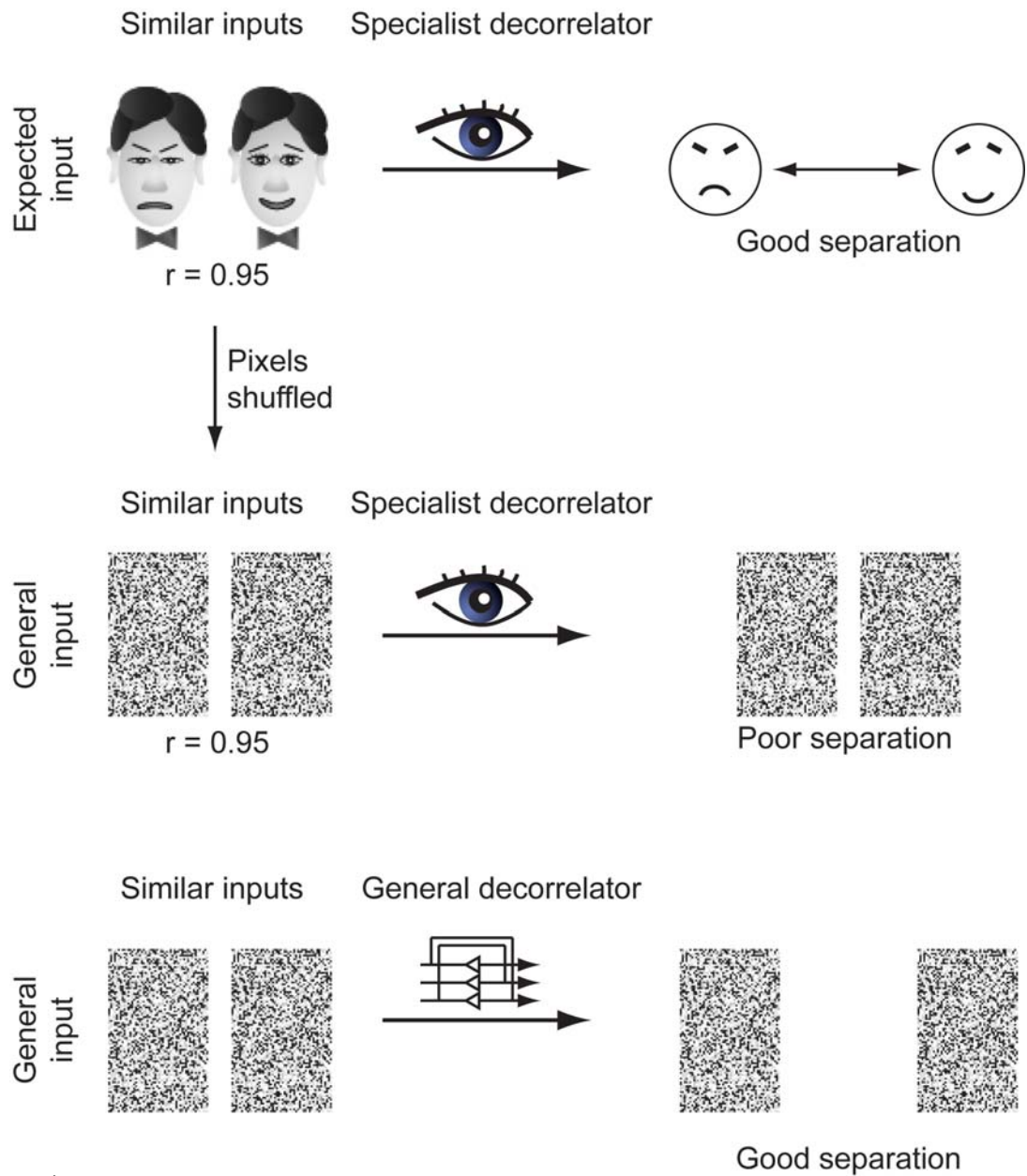


Dense circuit:

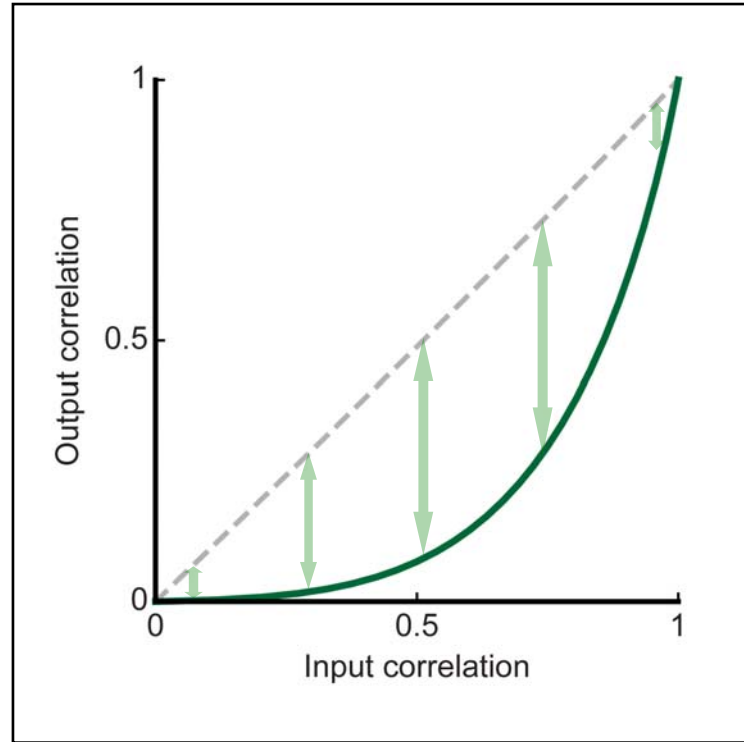
Many but weak connections



Recurrent networks act as general pattern decorrelators



Pattern decorrelation by recurrent networks is not chaotic



Neuronal representations are discontinuous:

- Decision-making: abrupt switching of neuronal output
- Sensory processing: generalization vs. separation
- ...

Hypothesis: abrupt switching between network states?

George



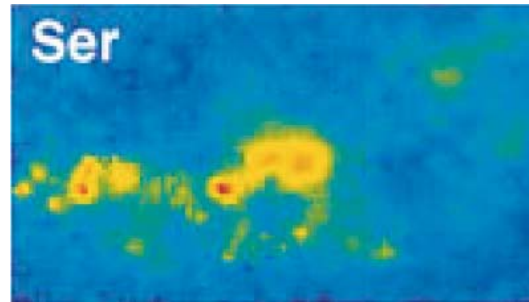
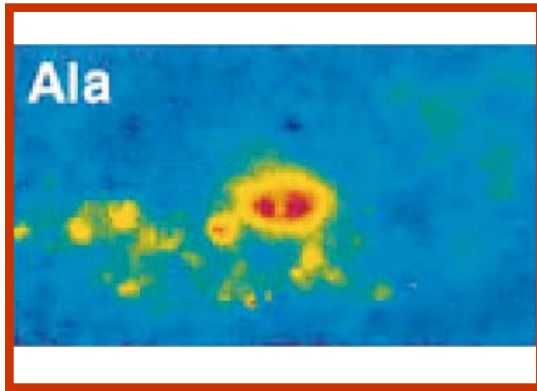
Pattern generalization

Almost George



Pattern separation

Pattern classification: generalization vs. separation

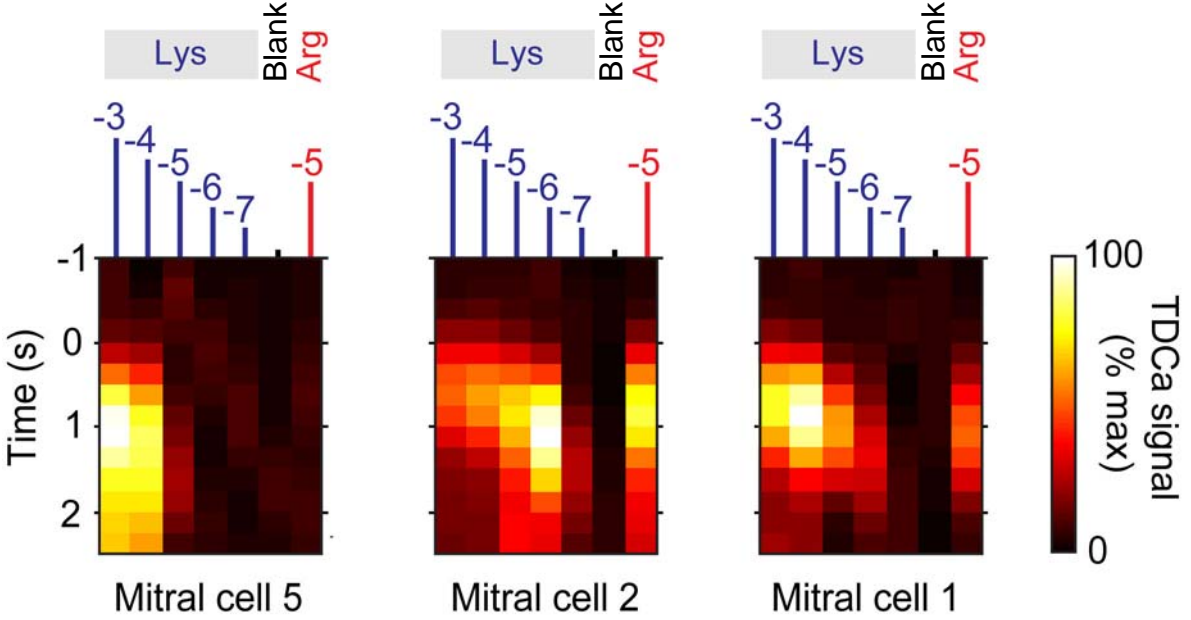
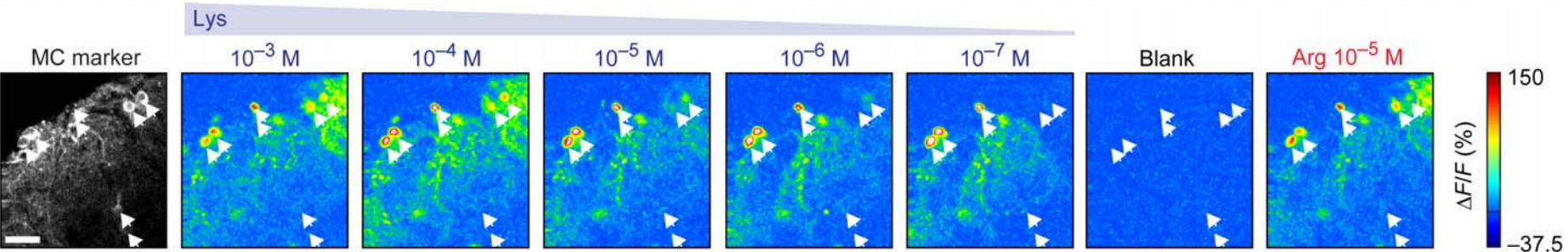


Pattern generalization

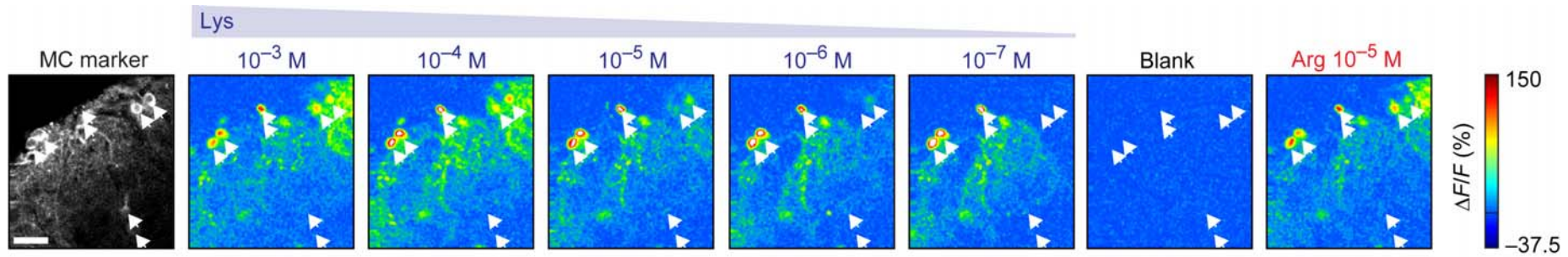


Pattern separation

Concentration-invariance of mitral cell response patterns

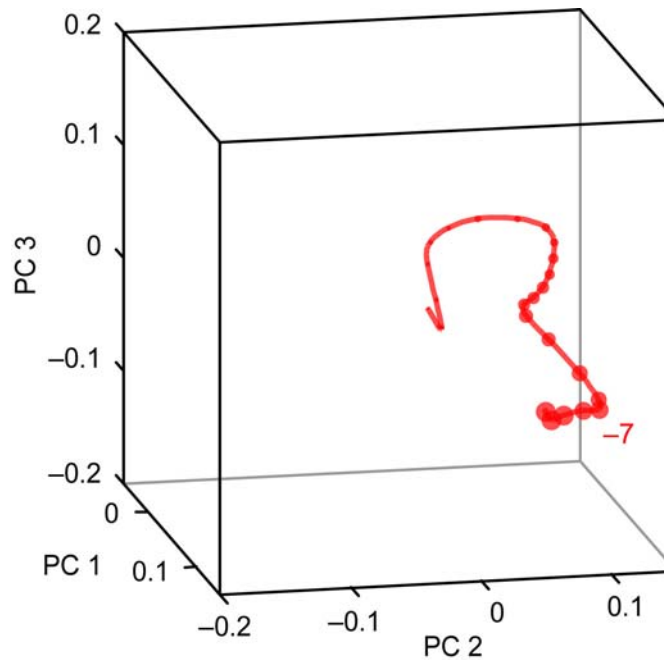
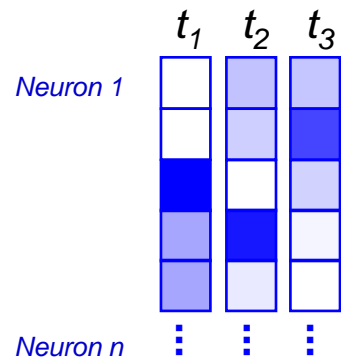


Concentration-invariance of mitral cell response patterns

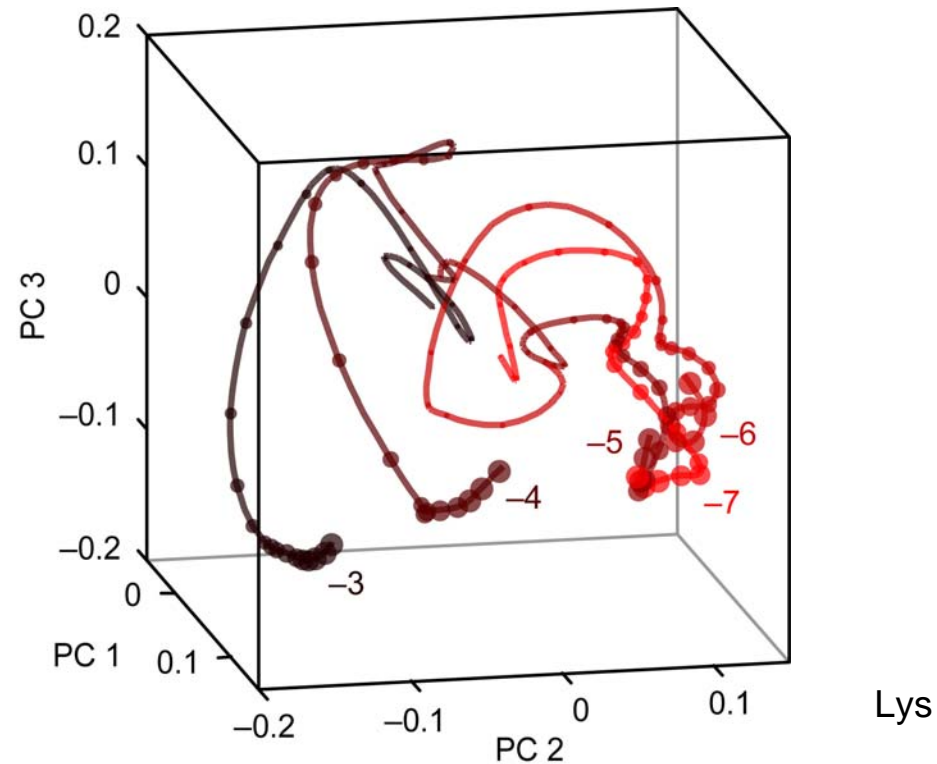


Principal component analysis

Population vectors

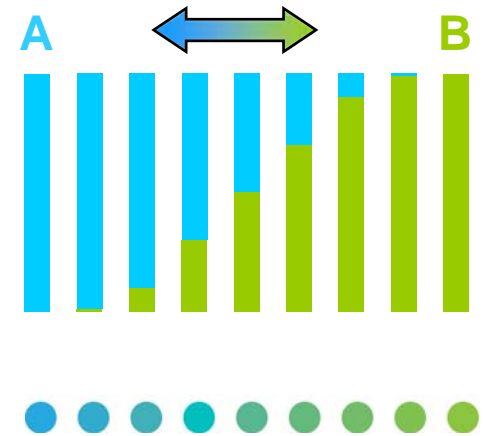


Concentration-invariance of mitral cell response patterns

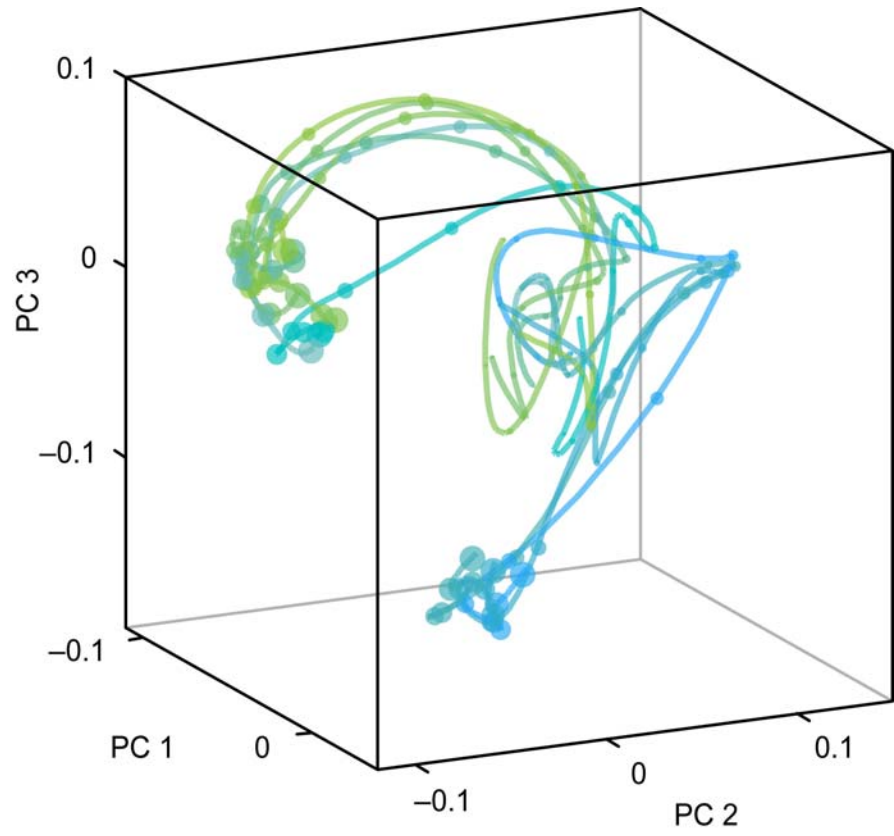


Output patterns are largely invariant within a range of concentrations

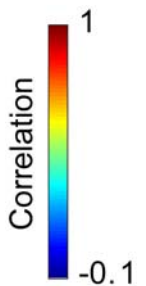
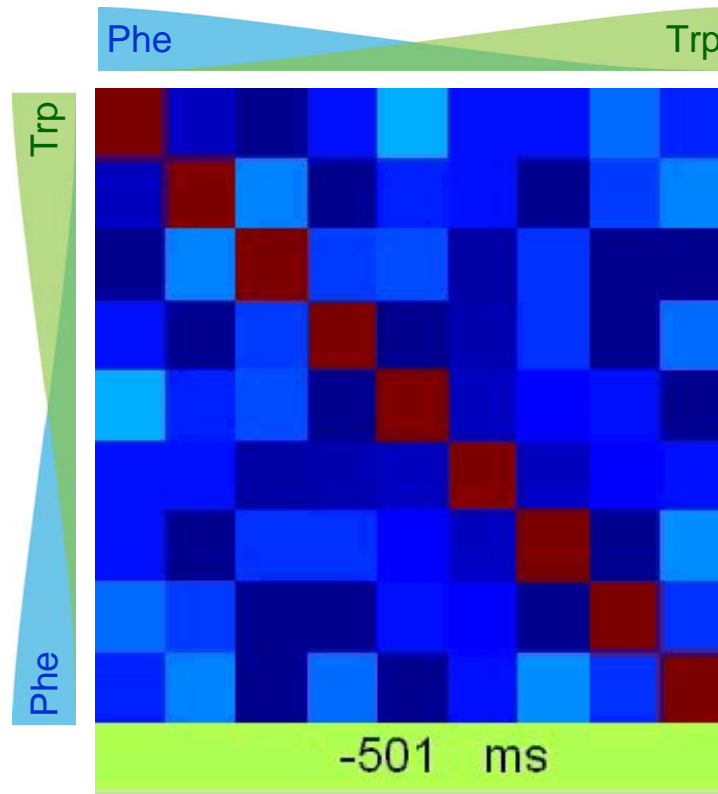
Odor morphing: similar odors



Odor morphing: similar odors

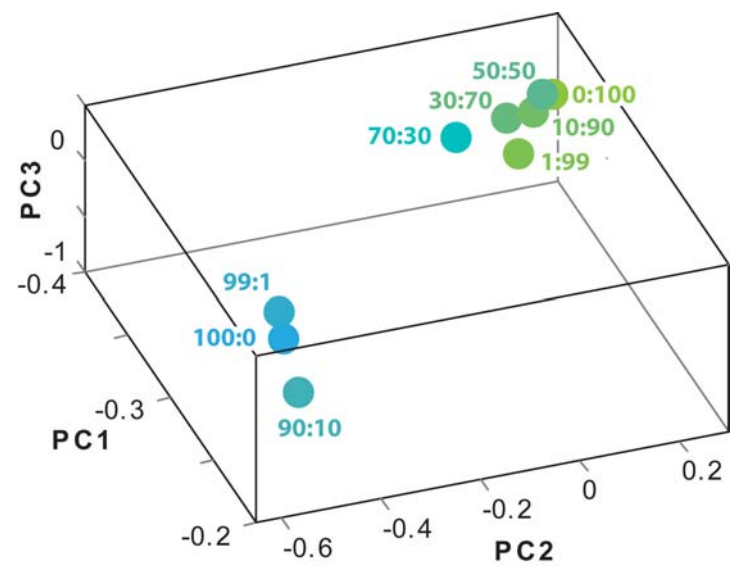
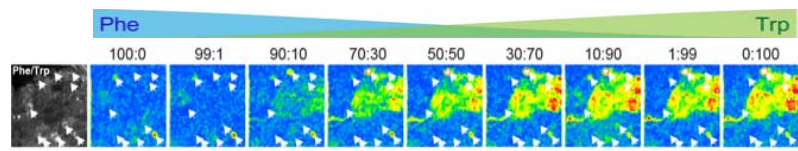


Odor morphing: similar odors

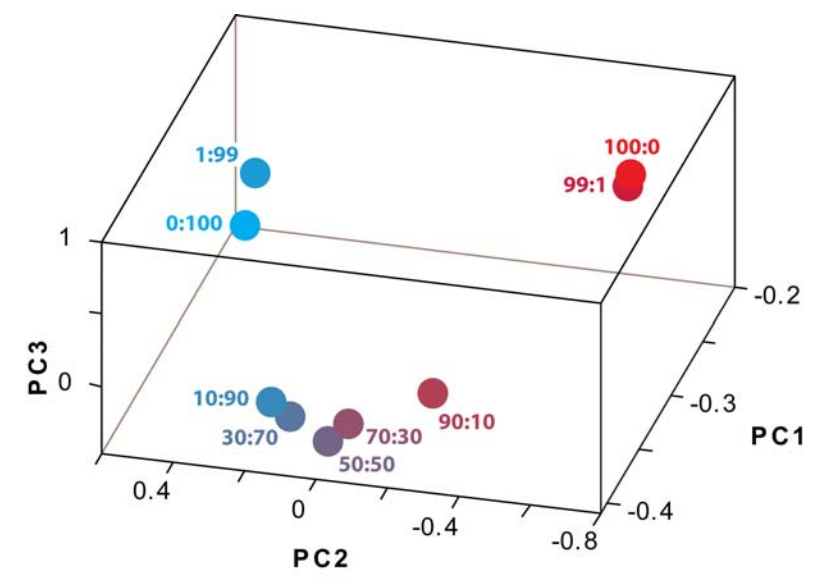
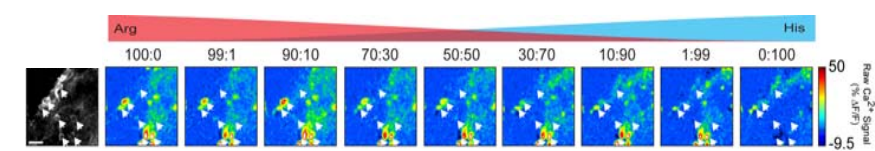


Odor morphing: discrete transitions

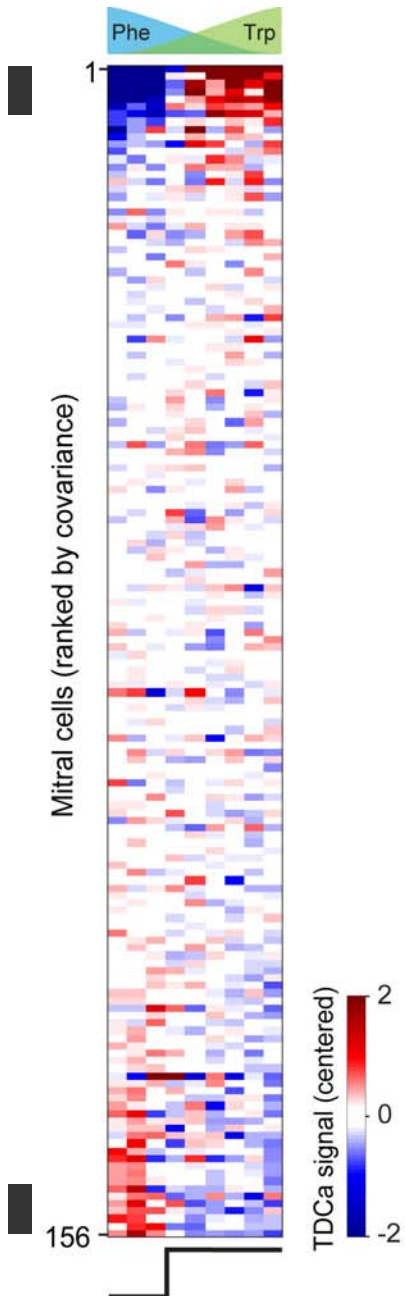
Similar odors



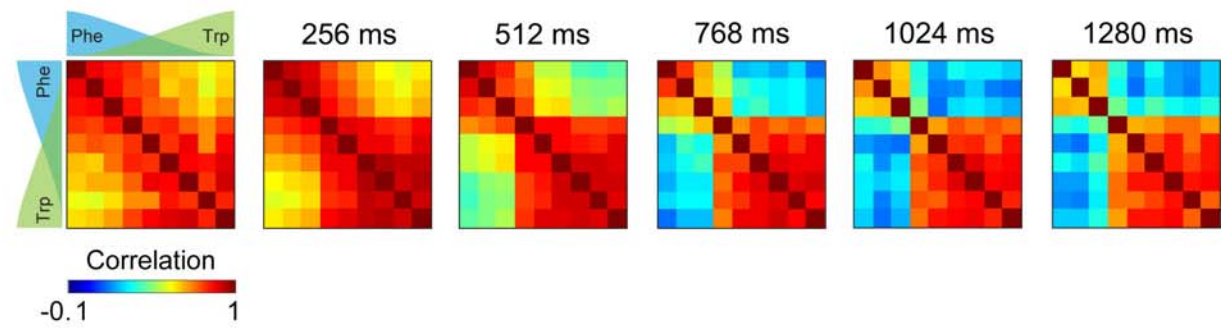
Dissimilar odors



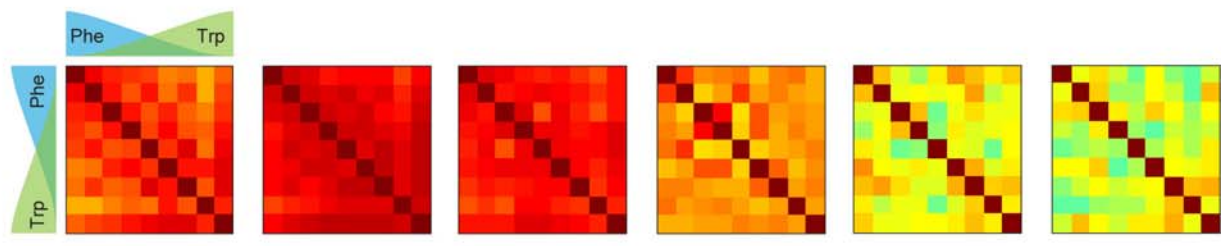
Pattern transitions are mediated by small, coordinated ensembles



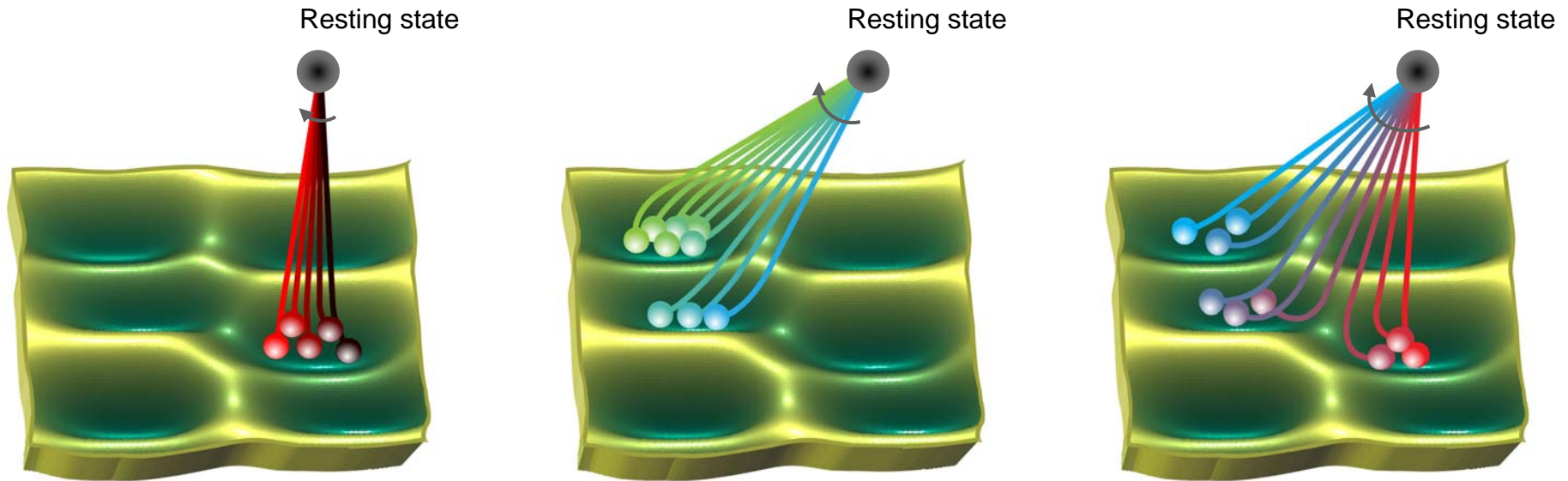
All mitral cells:



“Switching” cells (10 %) removed:



Discrete pattern classification: summary



- Discretization of coding space
- Noise-tolerance vs resolution
- Sensory filter \rightarrow invariances.
- Coordinated switching of responses in small ensembles.

- Consistent with psychophysics.
- Consistent with attractor networks.