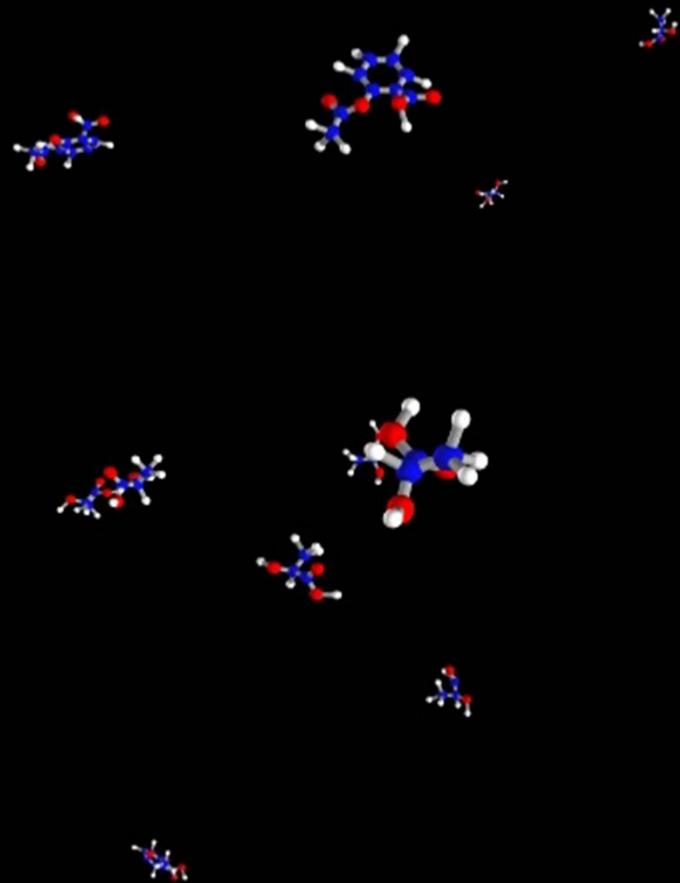


# NEURAL RELATIVITY PRINCIPLE

Alex Koulakov, *Cold Spring Harbor Laboratory*



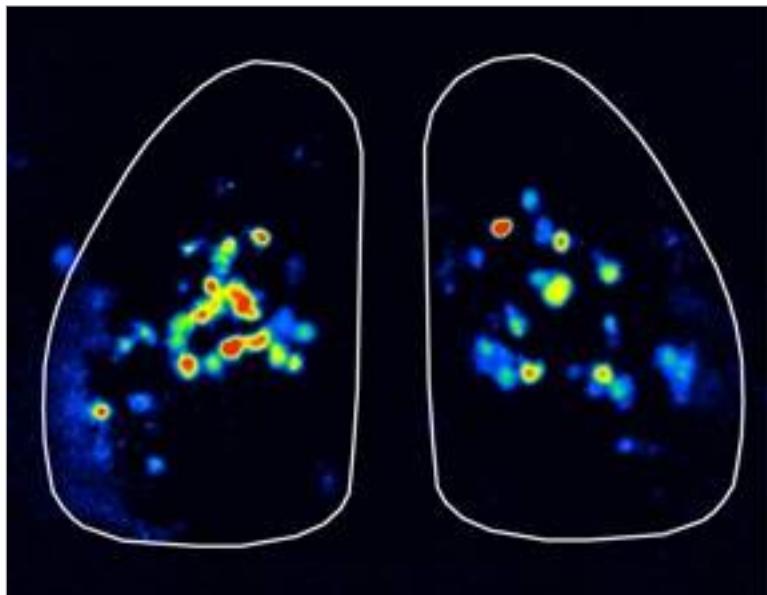
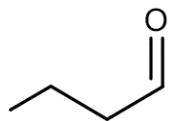


$\approx$

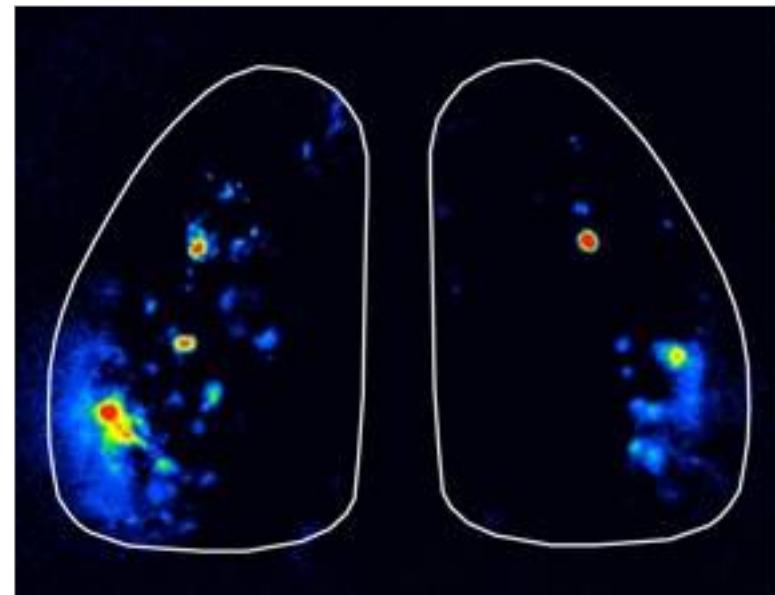
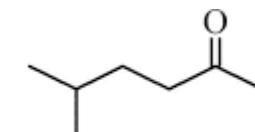
A large black mathematical symbol representing approximation or equivalence, centered between the two images.

# Odorants are represented by patterns of glomerular activation

butyraldehyde

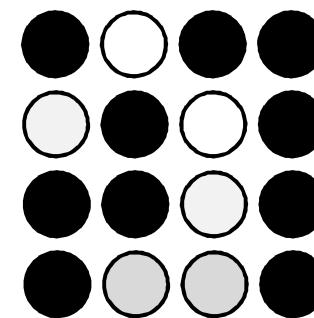
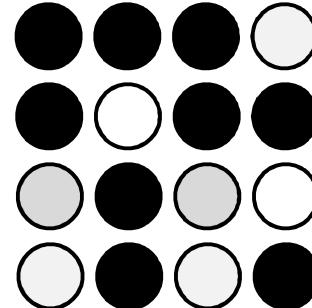
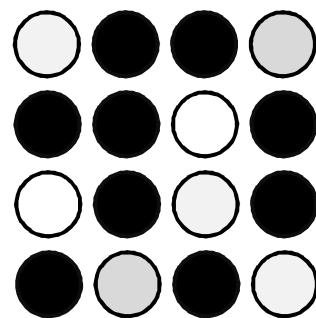


hexanone



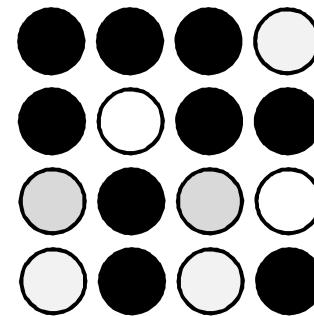
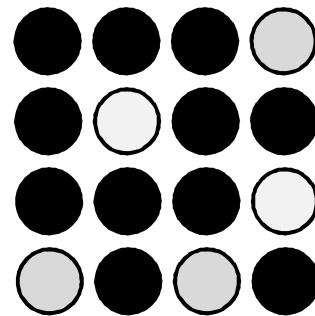
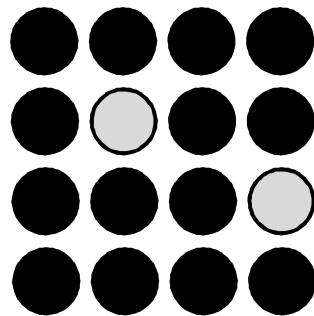
# Odorants are represented by patterns of glomerular activation

Glomeruli in the olfactory bulb (OB)



Odorant are represented by patterns of glomerular activation

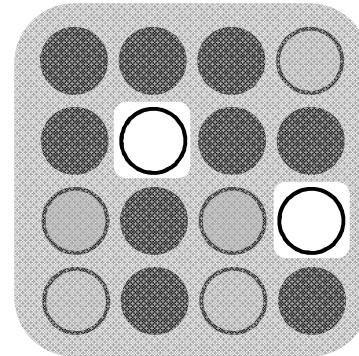
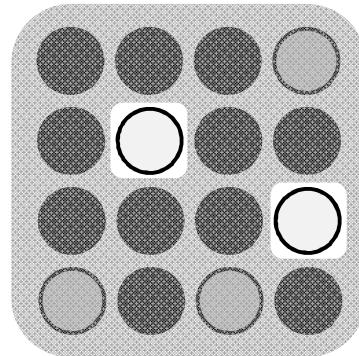
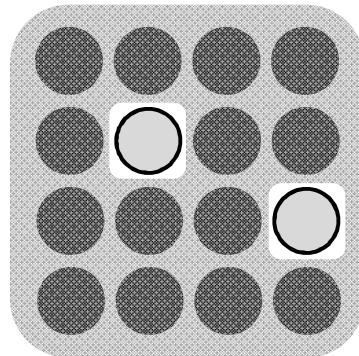
$$C_1 < C_2 < C_3$$



How can odor identity be constant?

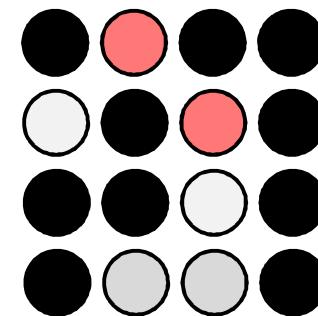
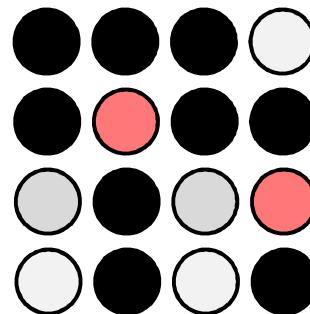
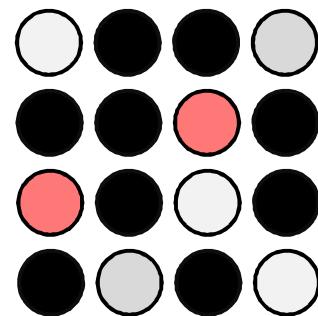
# Solution – primacy coding

Template representing an odorant includes a very small number of glomeruli of highest affinity to the odorant (activated at the smallest concentration)

 $C_1$  $<$  $C_2$  $<$  $C_3$ 

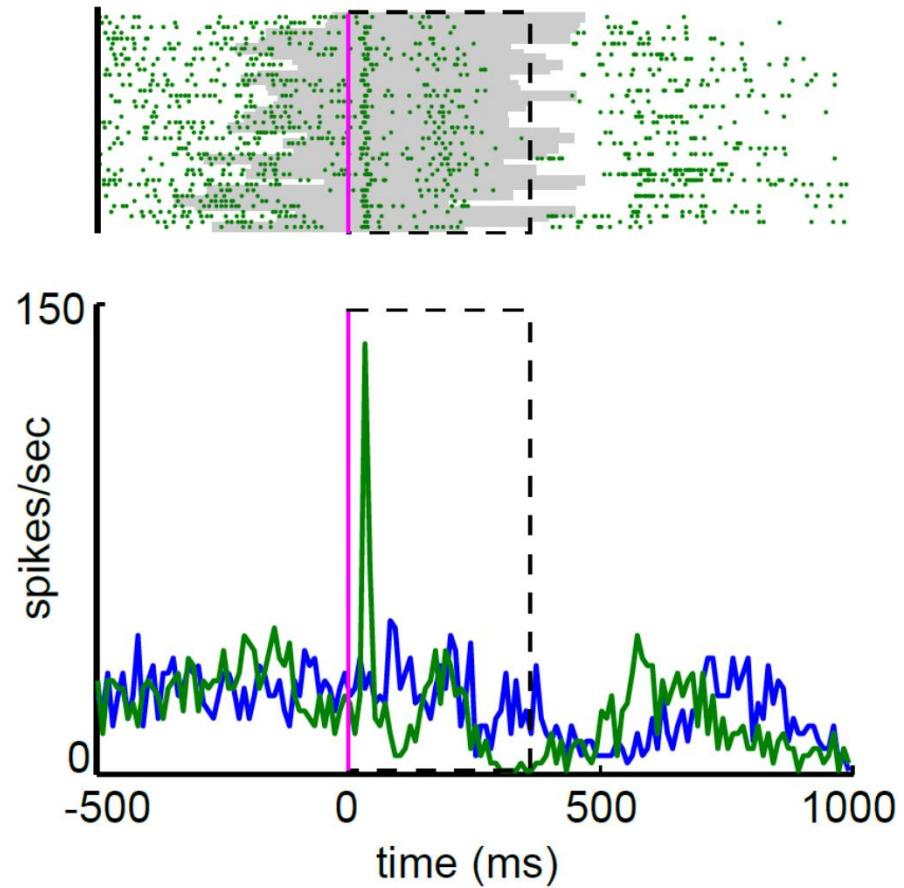
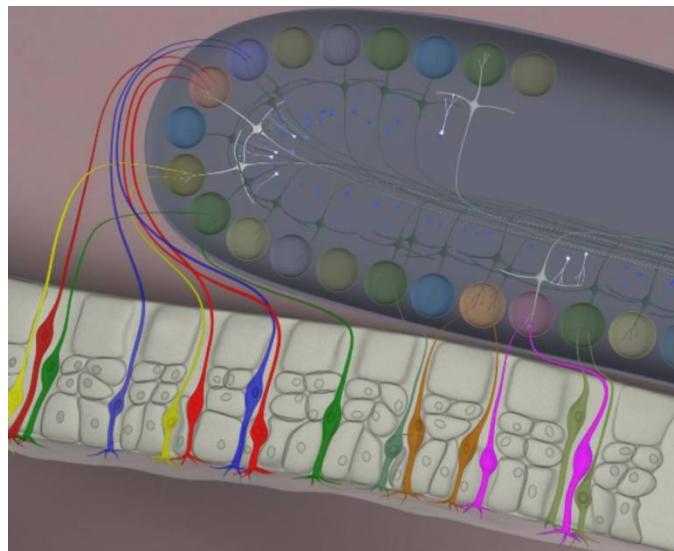
$p (= 2)$  is an important parameter of the code

Different smells are represented by their ‘primacy’ glomeruli



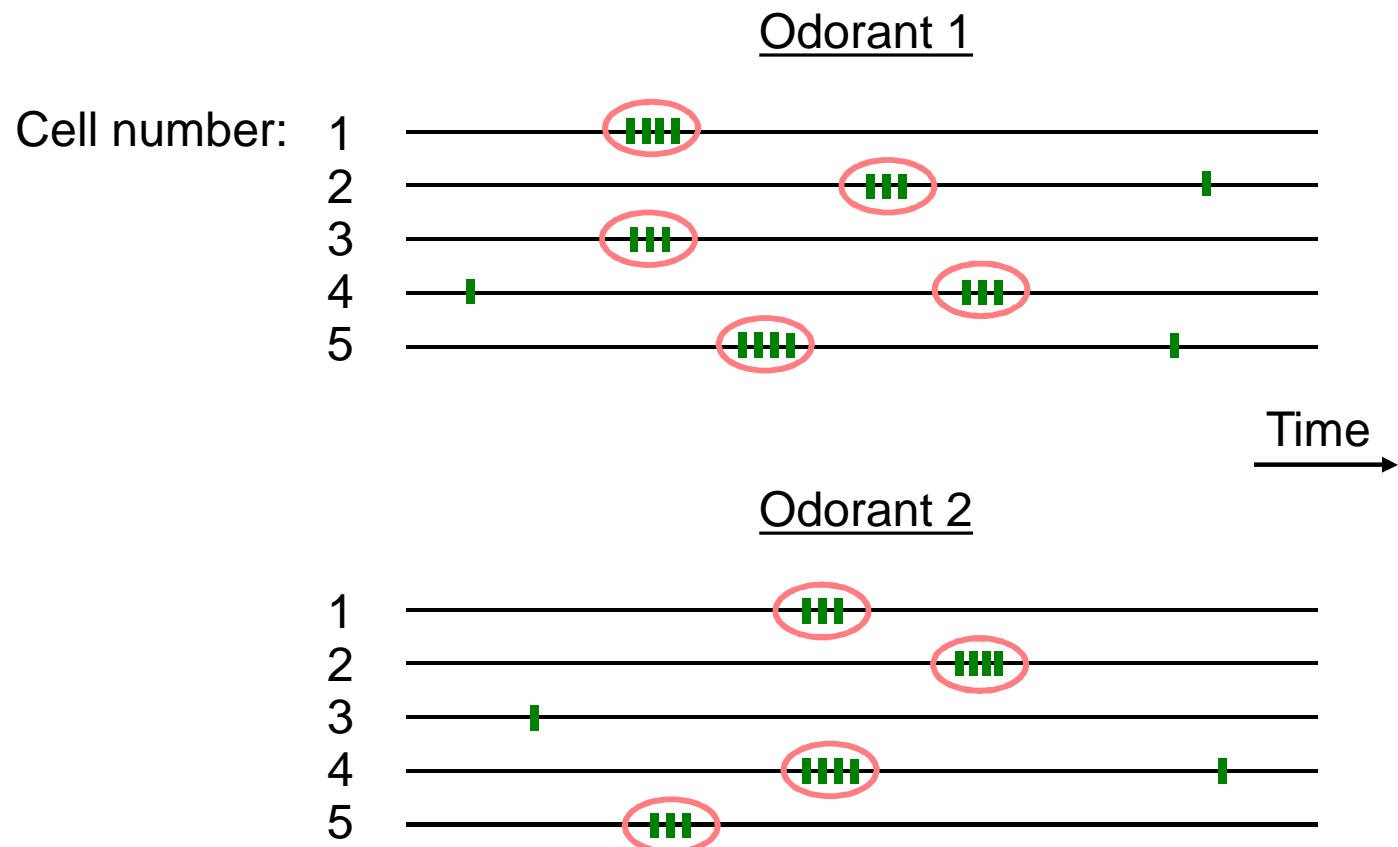
# Temporal codes

# Temporal codes in the olfactory bulb

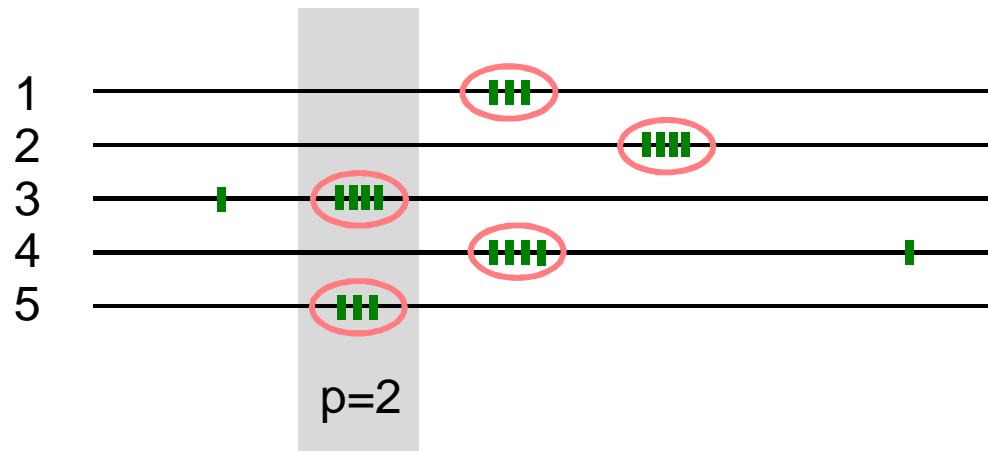


Cury and Uchida (2010)  
Shusterman, Smear, Koulakov, and Rinberg (2011)

Odor identity is represented in the sequence of sharp events generated by MCs



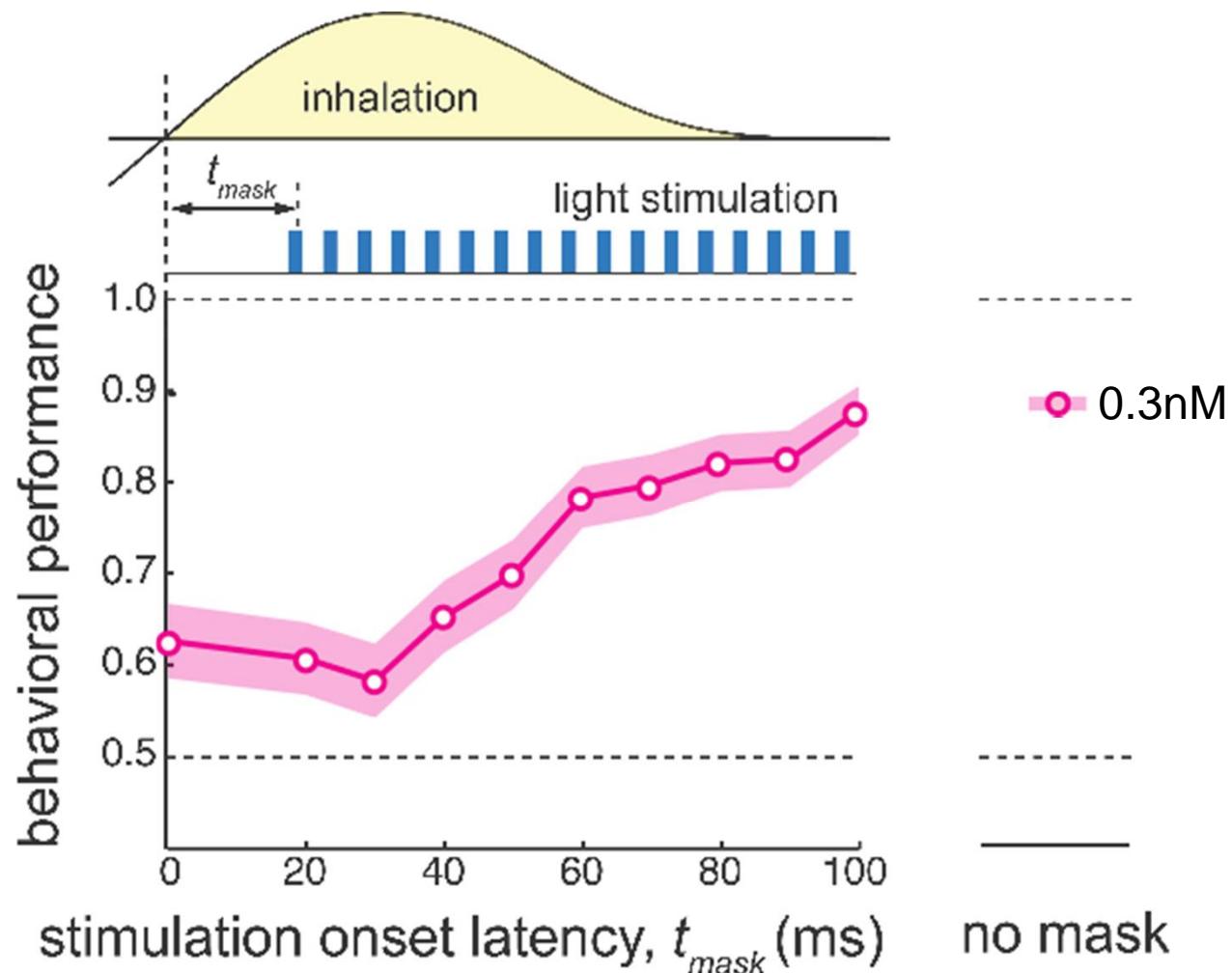
# Primacy coding



First set of  $p$  active mitral cells  
determines odorant identity

Prediction: odor identity is determined early in  
the sniff cycle

# Smell A vs smell B

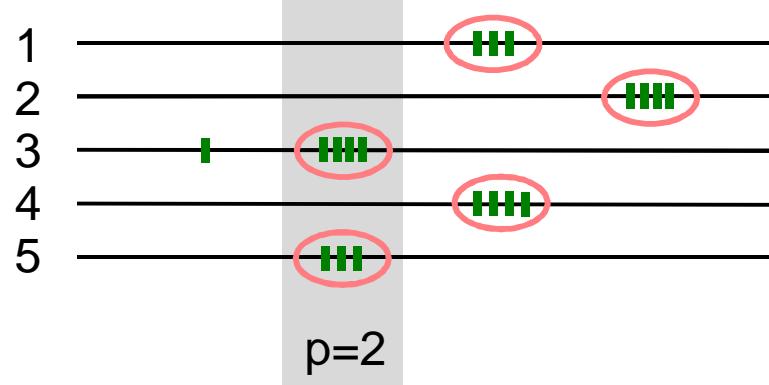


Chris Wilson (NYU)



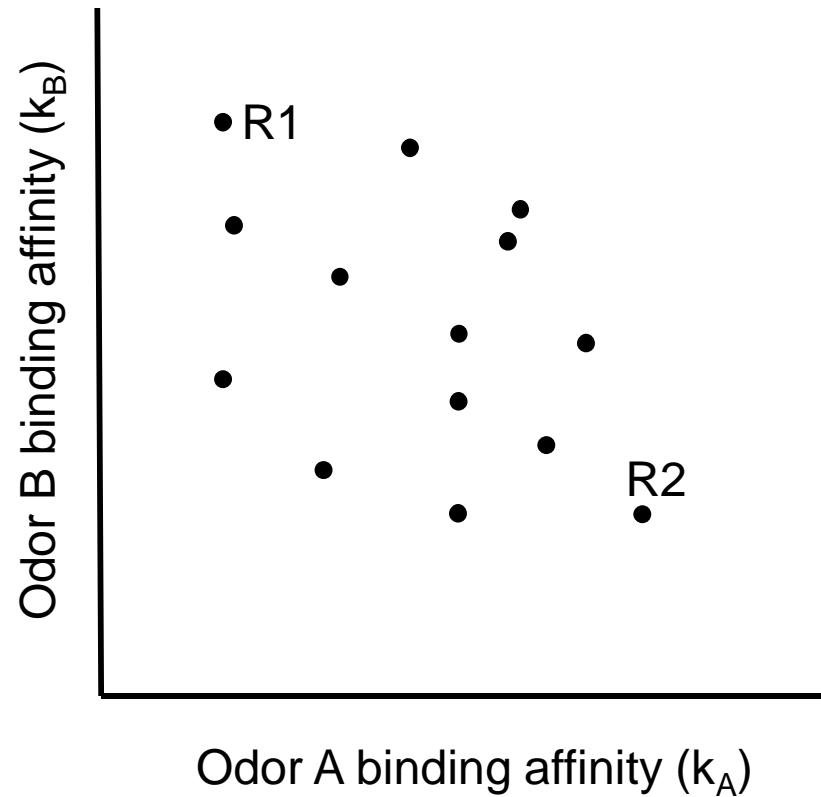
Dima Rinberg (NYU)

# Primacy coding: Shazam service for the brain?

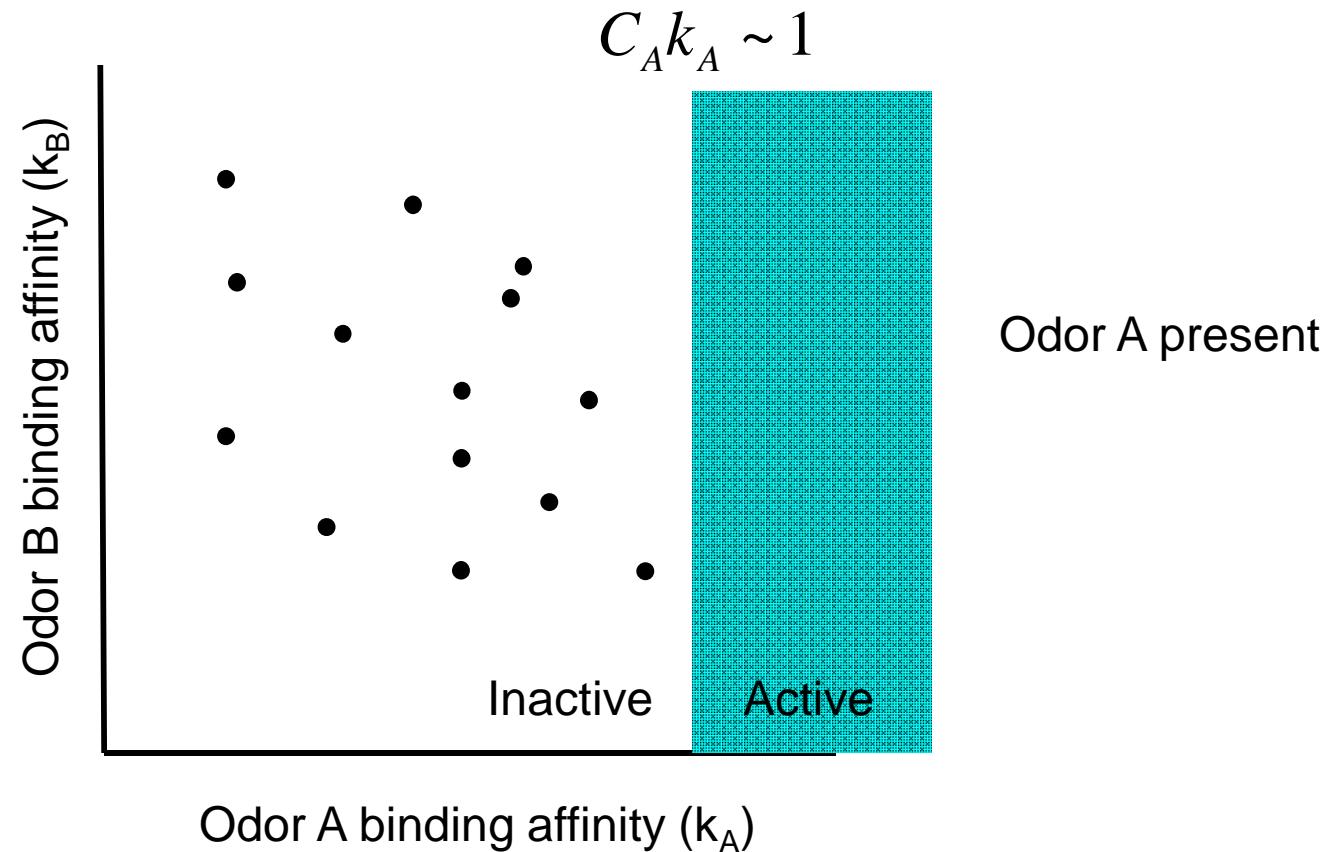


# evolution of olfactory receptors

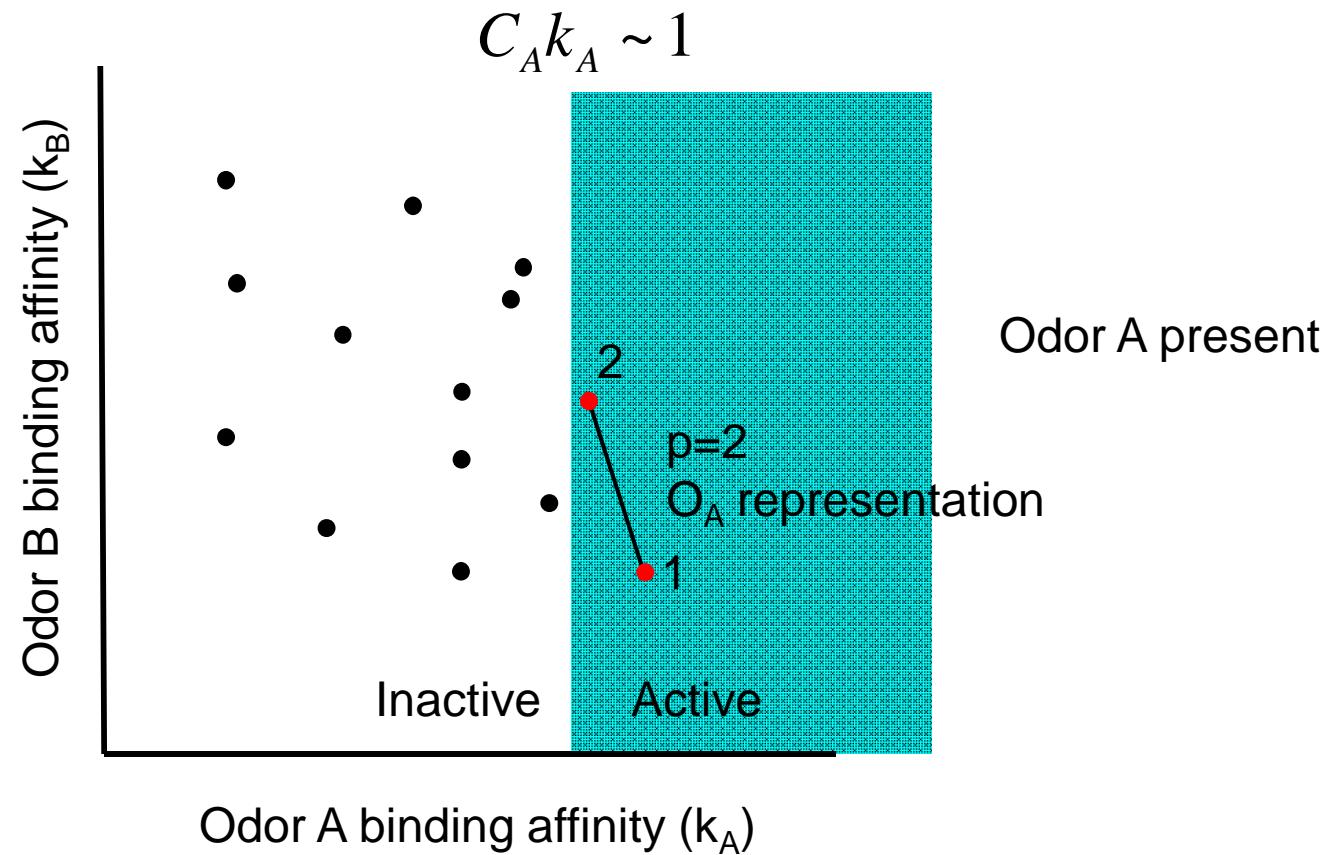
# 2D odorspace



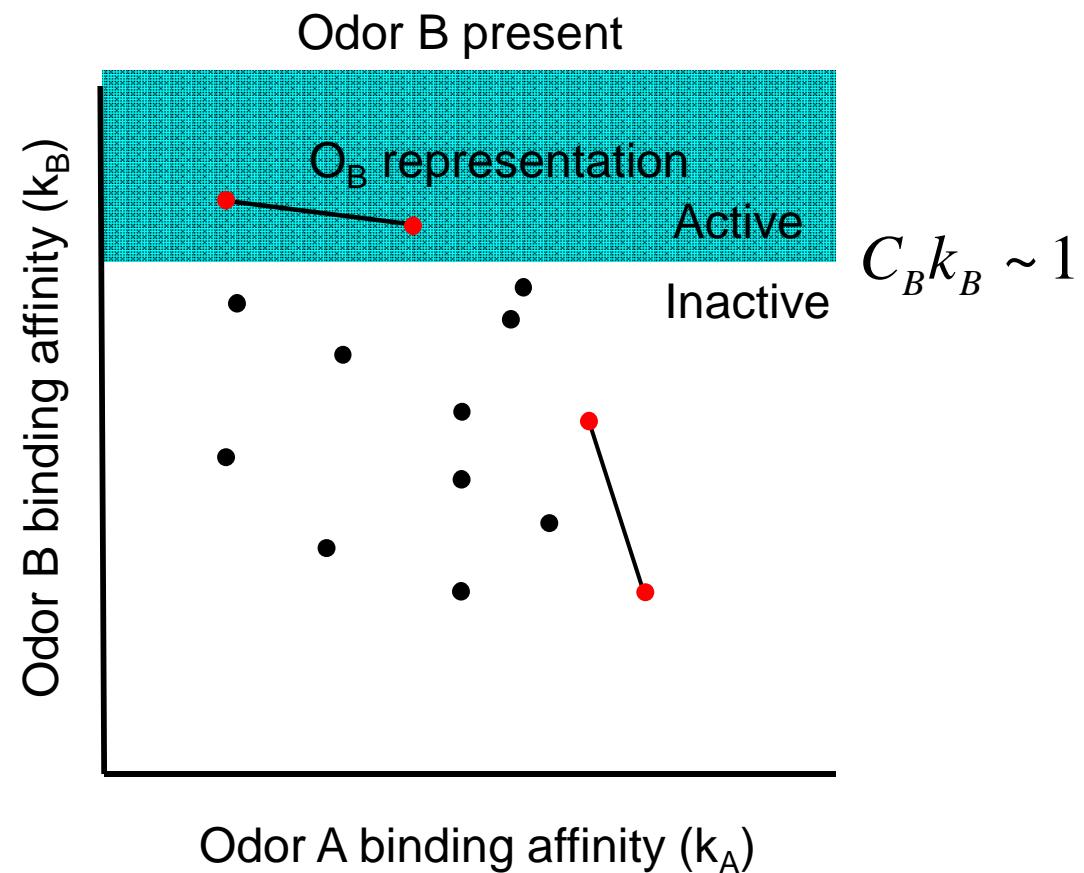
# 2D odorspace



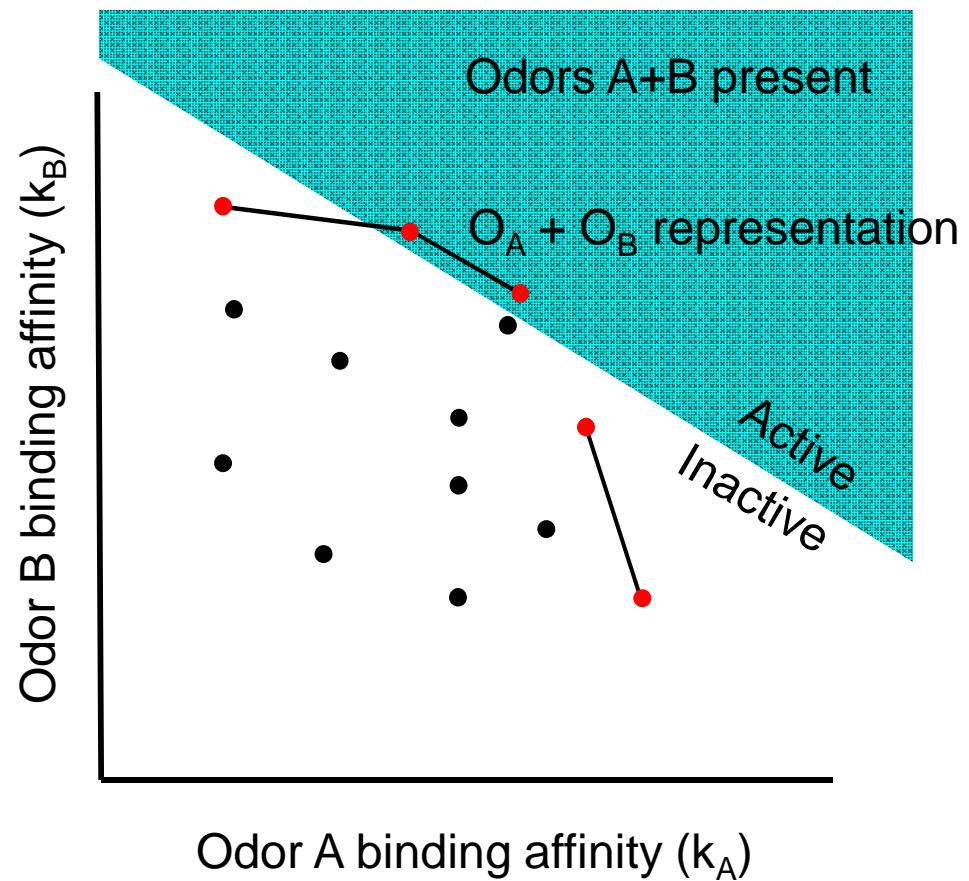
# 2D odorspace



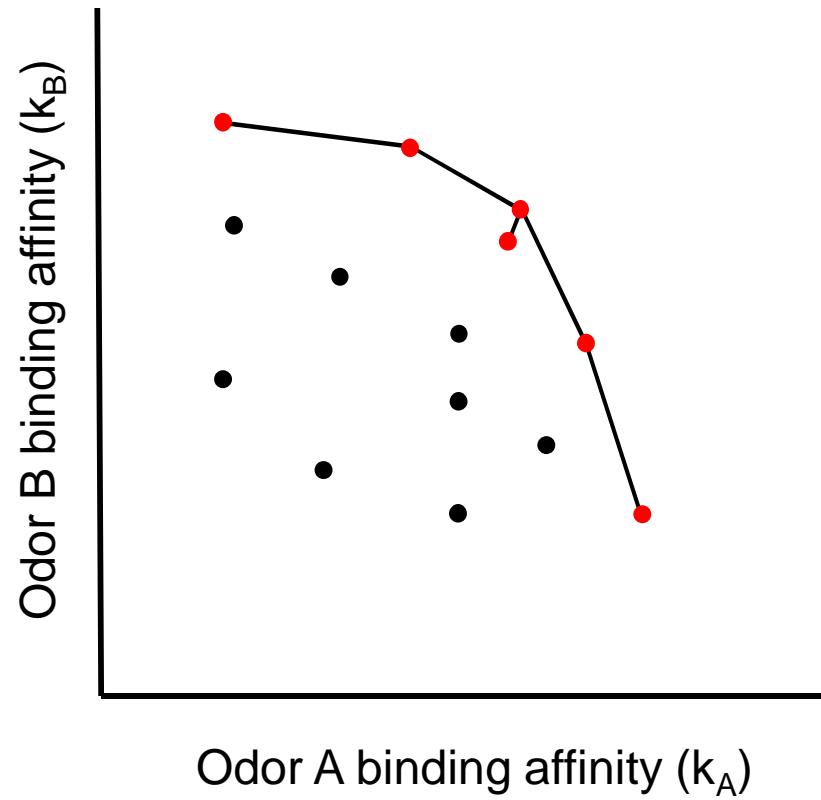
# 2D odorspace



# 2D odorspace

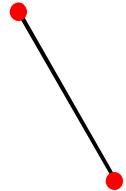


# $p=2$ primacy model

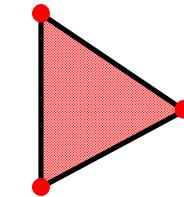


# Individual odor identities are represented by simplexes

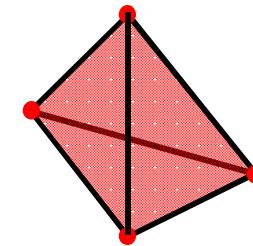
$p=2$



$p=3$

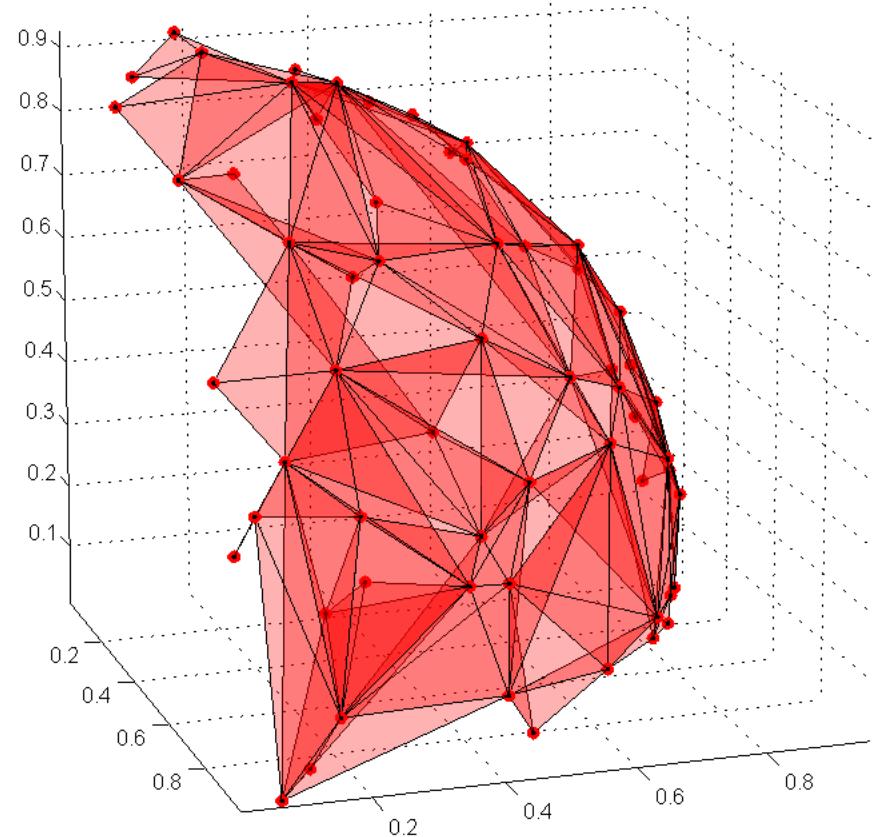


$p=4$

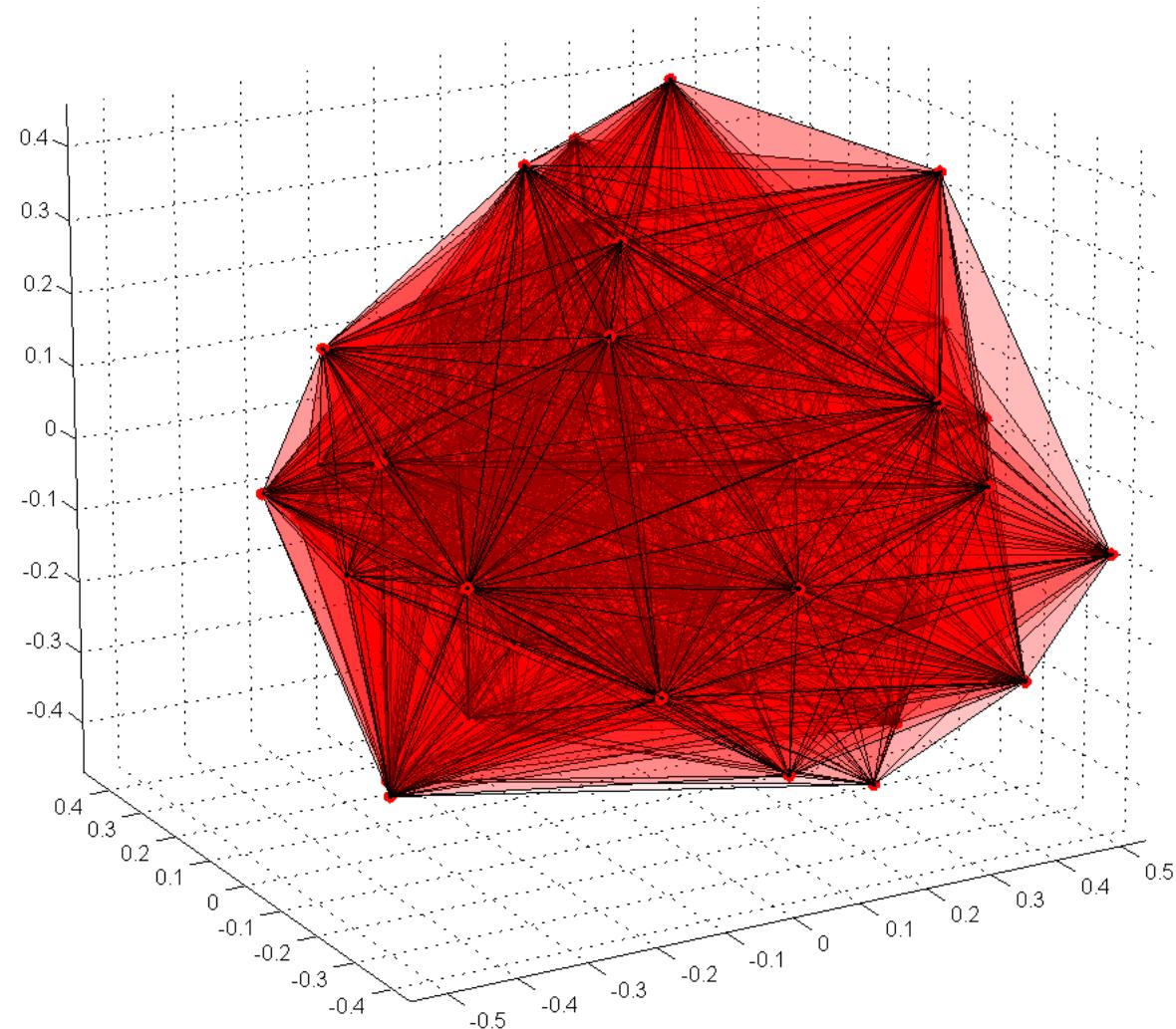


⋮ ⋮ ⋮

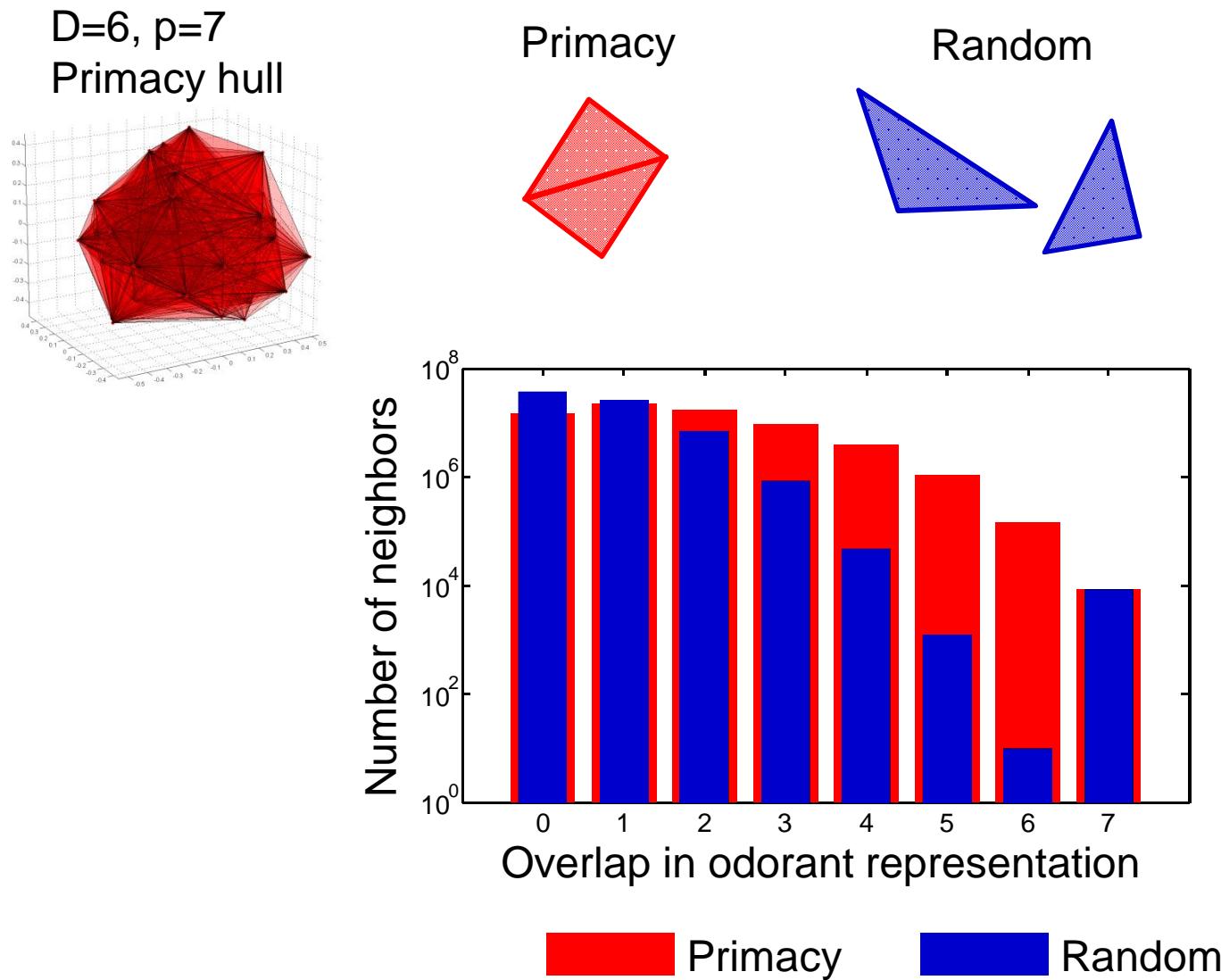
# D=3, p=3 primacy hull



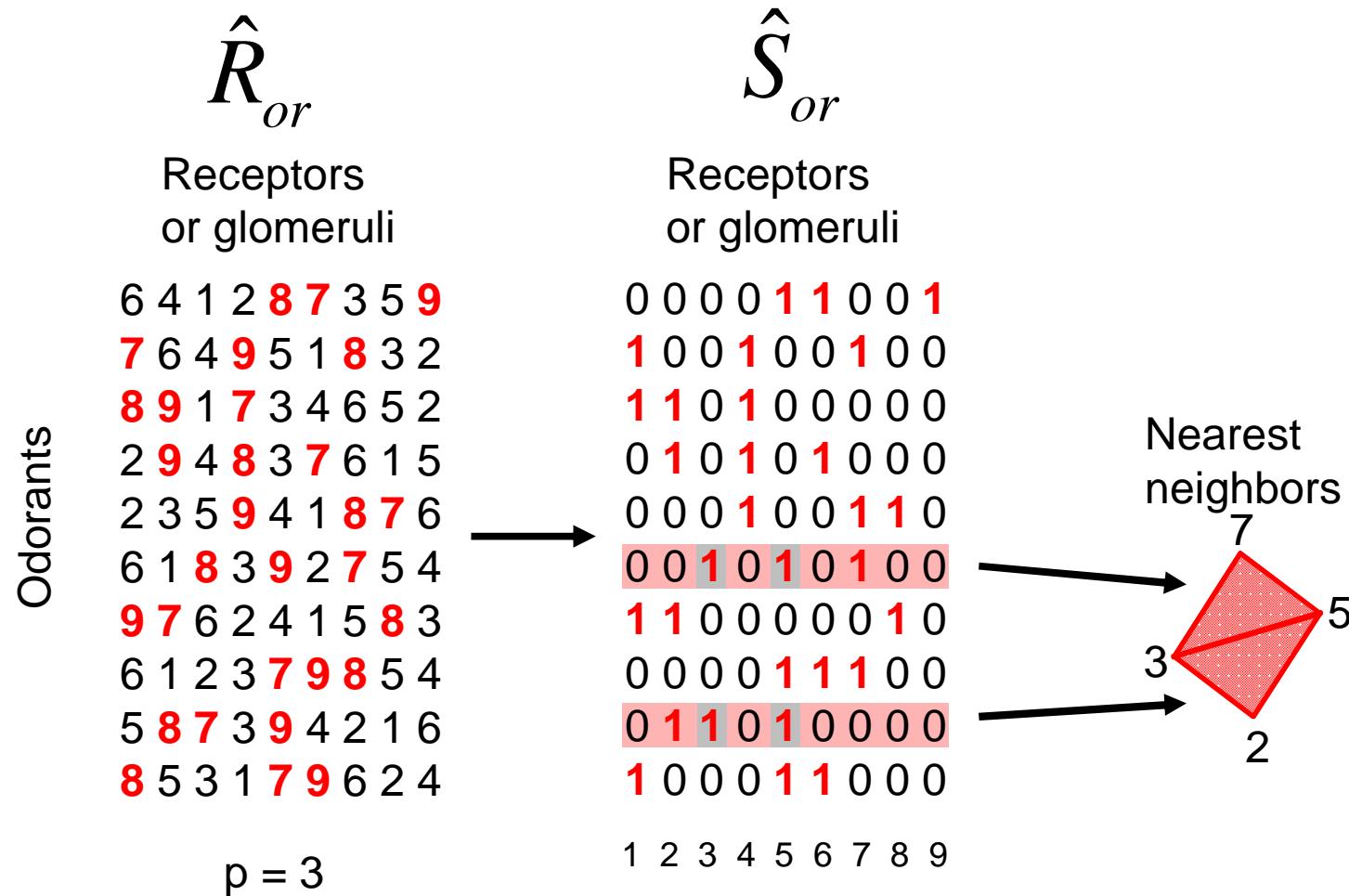
# $D=6$ , $p=7$ primacy hull



# Prediction: Higher-order correlations in the response matrix



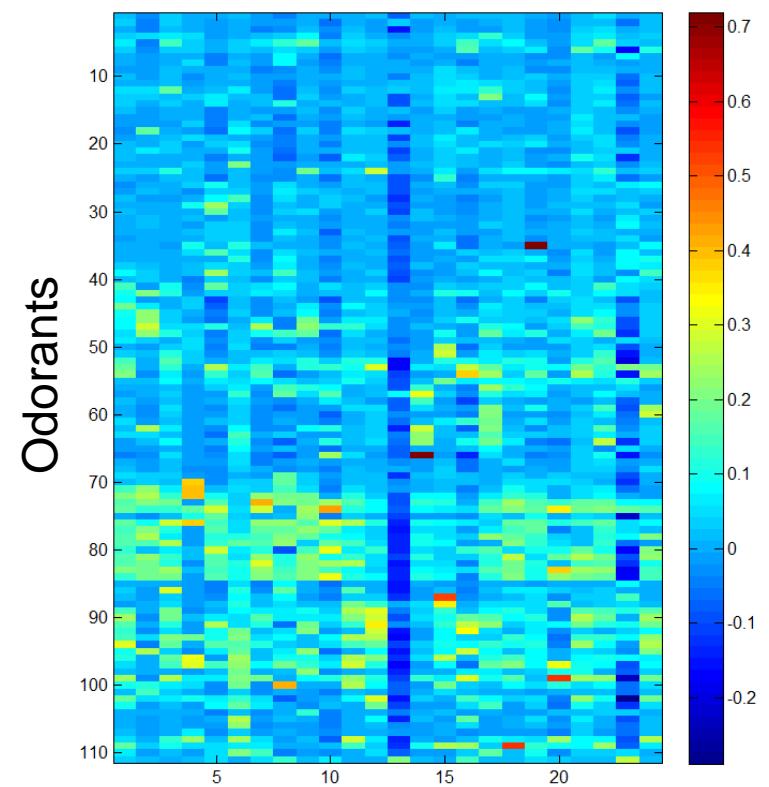
# Primacy structure in the response matrix



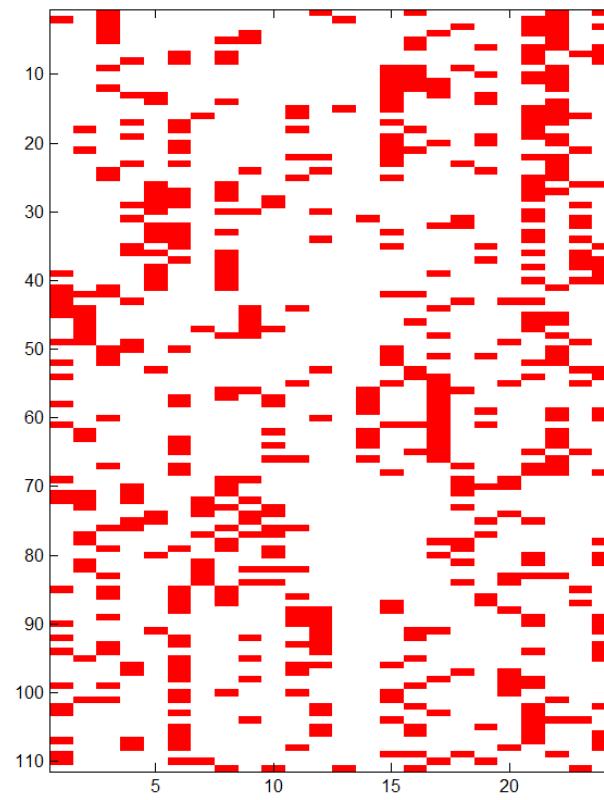
# Fly data: Hallem and Carlson (2006)



$\hat{R}_{or}$  Receptors



$\hat{S}_{or}$  Receptors



# Number of nearest neighbors, p=5

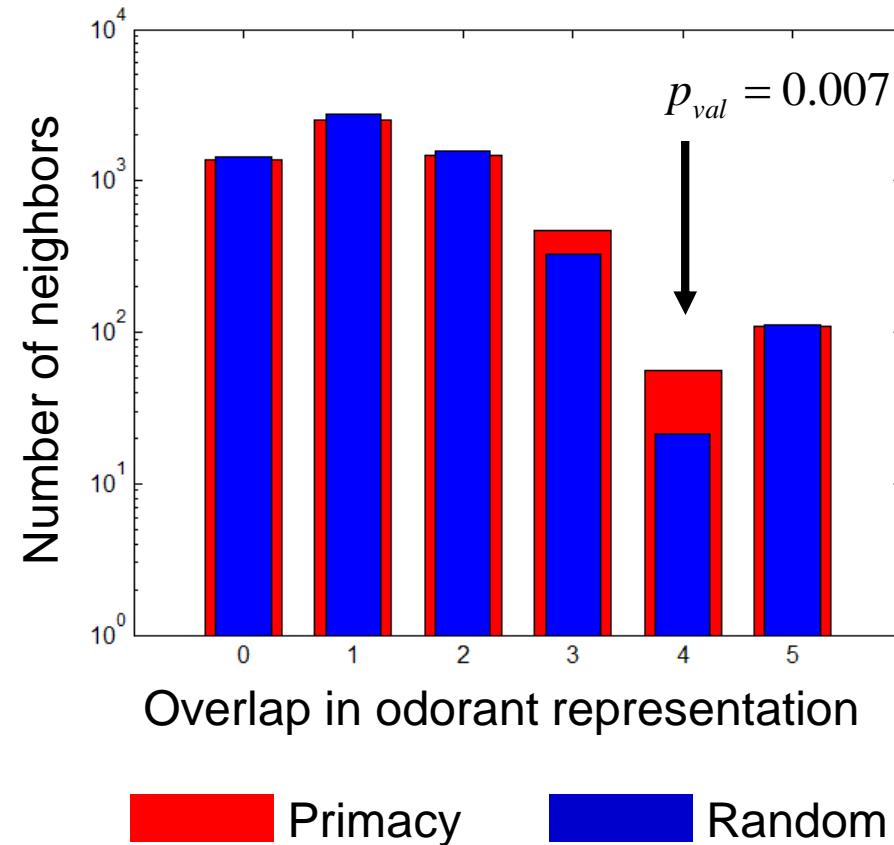


Primacy

Random



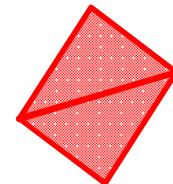
Hamza  
Giaffar



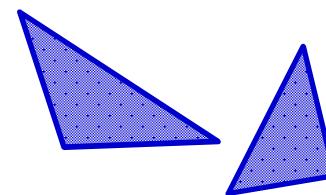
# Number of nearest neighbors, p=6



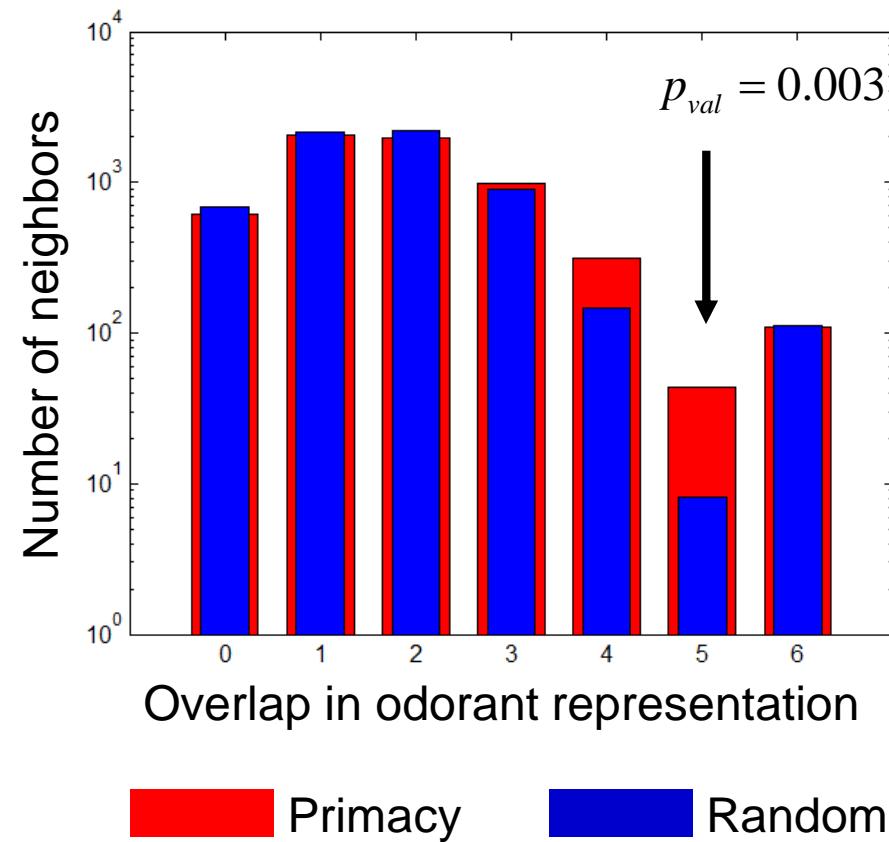
Primacy



Random



Hamza  
Giaffar

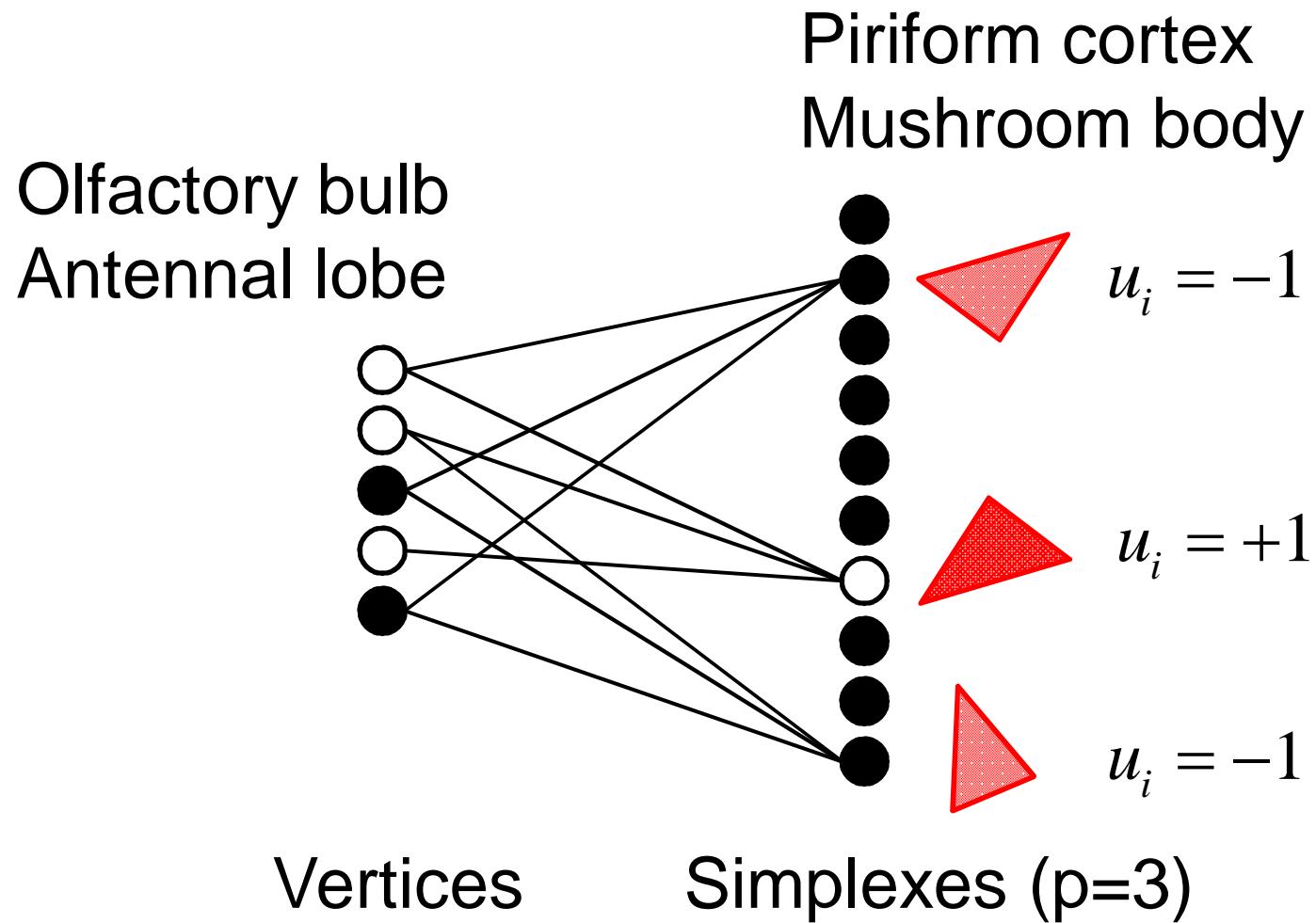


# Conclusions (OR evolution model)

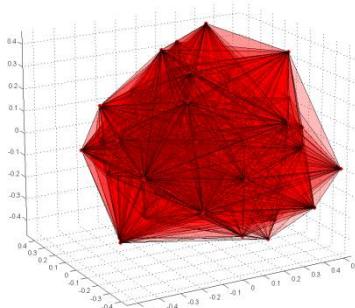
- Primacy model predicts that OR affinities occupy a narrow shell called primacy hull.
- Topmost  $p$  responses of olfactory receptors share substantial overlaps
- Increased overlaps are present in fly data ( $p=5$ )
- These findings suggest substantial higher-order correlations in OR responses

# Network models

# Expand, sparsen, and decorrelate

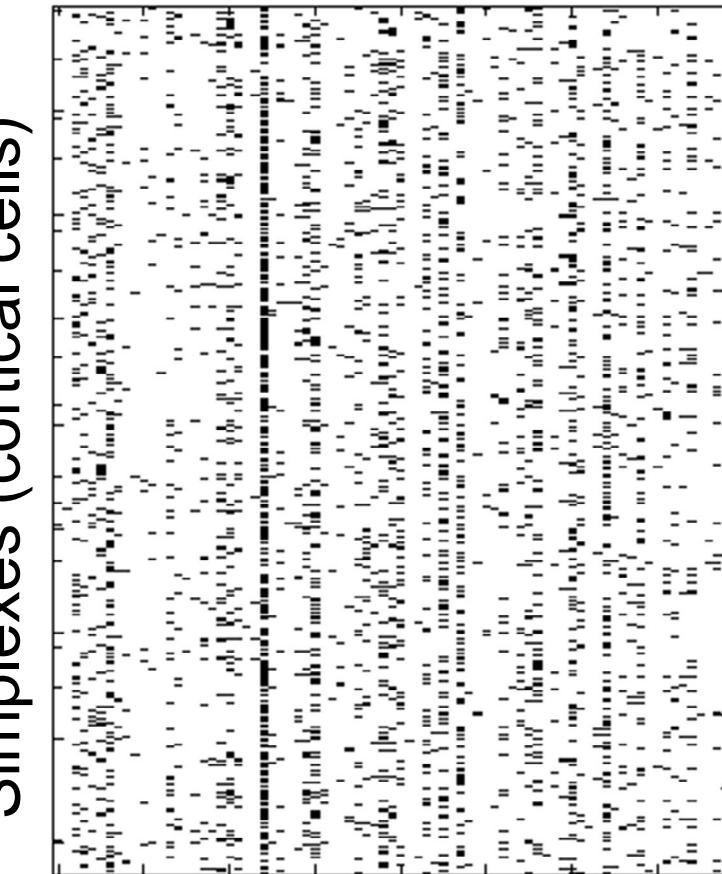


# Connectivity between glomeruli and simplexes is given by the symplectic matrix



$$\hat{S} =$$

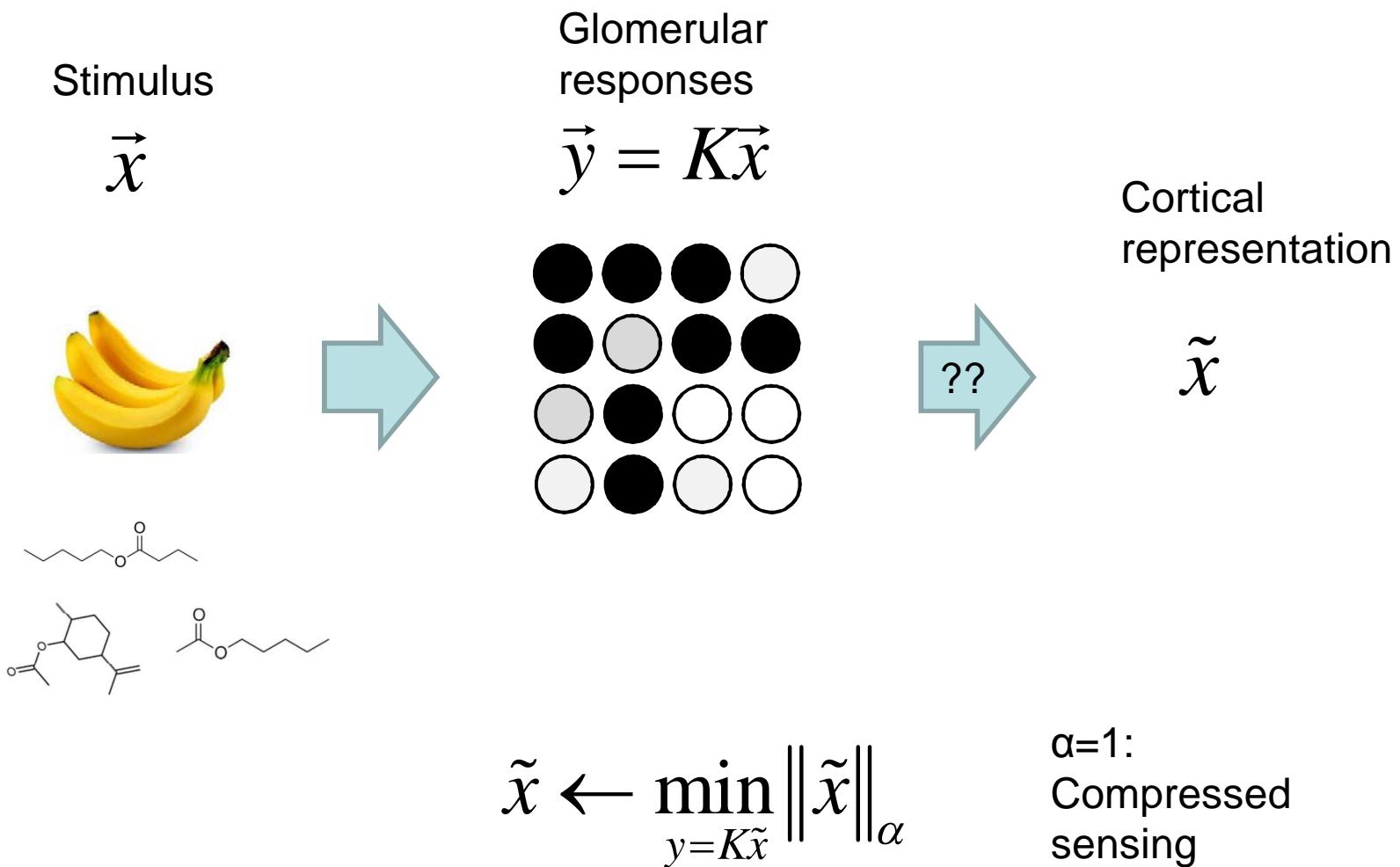
Vertices (receptors/glomeruli)



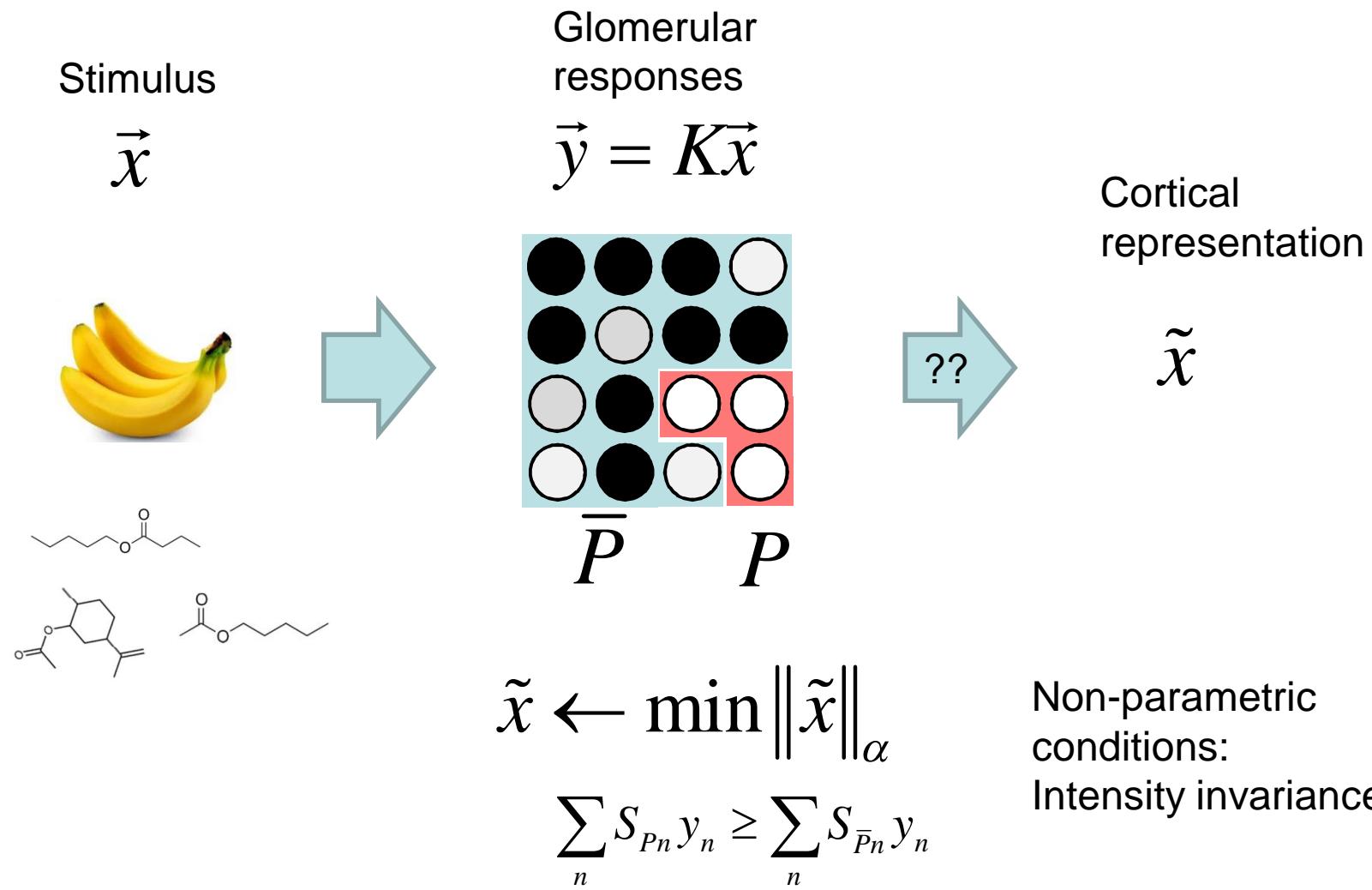
Simplexes (cortical cells)

# Dual networks

# How does decoding work?



# Primacy (relativity) principle



Non-parametric problem:

$$\tilde{x} \leftarrow \min \|\tilde{x}\|_{\alpha}, \quad \sum_n S_{Pn} y_n \geq \sum_n S_{\bar{P}n} y_n$$

Fix the scale:

$$\tilde{x} \leftarrow \min \|\tilde{x}\|_{\alpha}, \quad \sum_n S_{Pn} y_n \geq \varepsilon \geq \sum_n S_{\bar{P}n} y_n$$

Single condition:

$$\tilde{x} \leftarrow \min \|\tilde{x}\|_{\alpha}, \quad u_s \left( \sum_n S_{sn} y_n - \varepsilon \right) \geq 0$$
$$u_s = \begin{cases} +1, & s \in P \\ -1, & s \in \bar{P} \end{cases}$$

Primary problem is difficult to implement:

$$\tilde{x} \leftarrow \min \|\tilde{x}\|_\alpha, \quad u_s (\sum_n S_{sn} y_n - \varepsilon) \geq 0 \quad u_s = \begin{cases} +1, & s \in P \\ -1, & s \in \bar{P} \end{cases}$$
$$y_n = \sum_m K_{nm} \tilde{x}_m$$

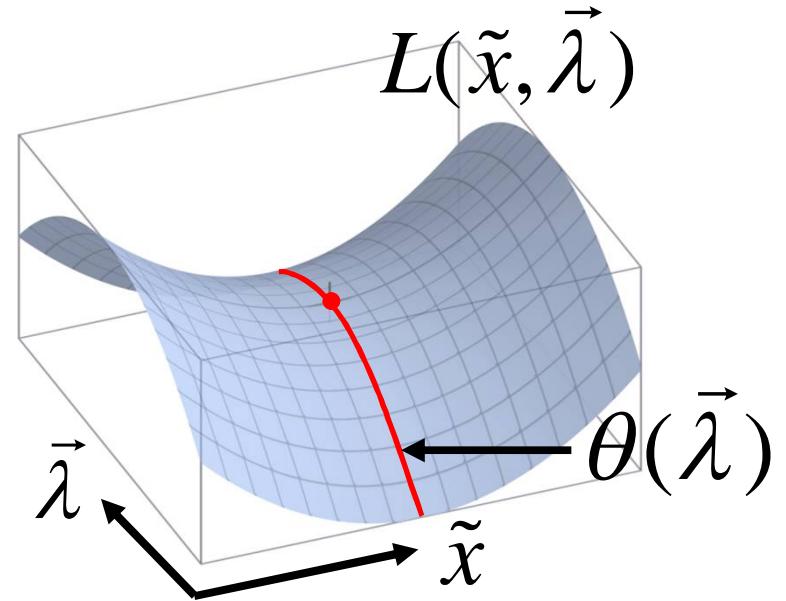
Lagrangian (Karush-Kuhn-Tucker):

$$L(\tilde{x}, \vec{\lambda}) = \|\tilde{x}\|_\alpha - \sum_s \lambda_s u_s (\sum_n S_{sn} y_n - \varepsilon) \quad \lambda_n \geq 0$$

# Duality: Lagrange coefficients become variables

Dual cost-function:

$$\theta(\vec{\lambda}) = \min L(\tilde{x}, \vec{\lambda})$$



Dual problem:

$$\vec{\lambda} \leftarrow \max \theta(\vec{\lambda})$$

$$\lambda_n \geq 0$$

$$\tilde{x} = \tilde{x}(\vec{\lambda})$$

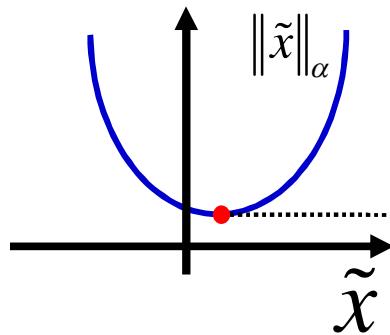
Primal problem:

$$\tilde{x} \leftarrow \min \|\tilde{x}\|_\alpha,$$

+ a bunch of very complex inequalities

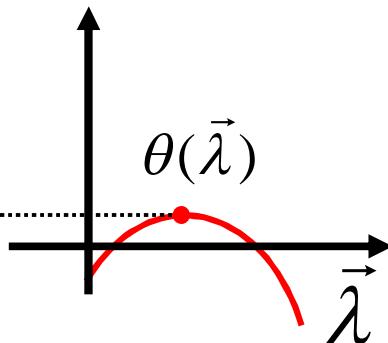
# Dual problem can be easily solved by a neural network

Primal problem:

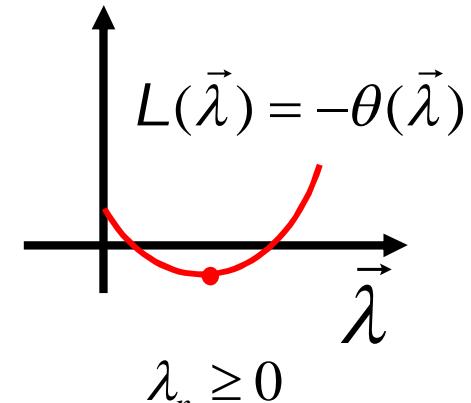


+ a bunch of very complex inequalities

Dual problem:



Network Lyapunov function:

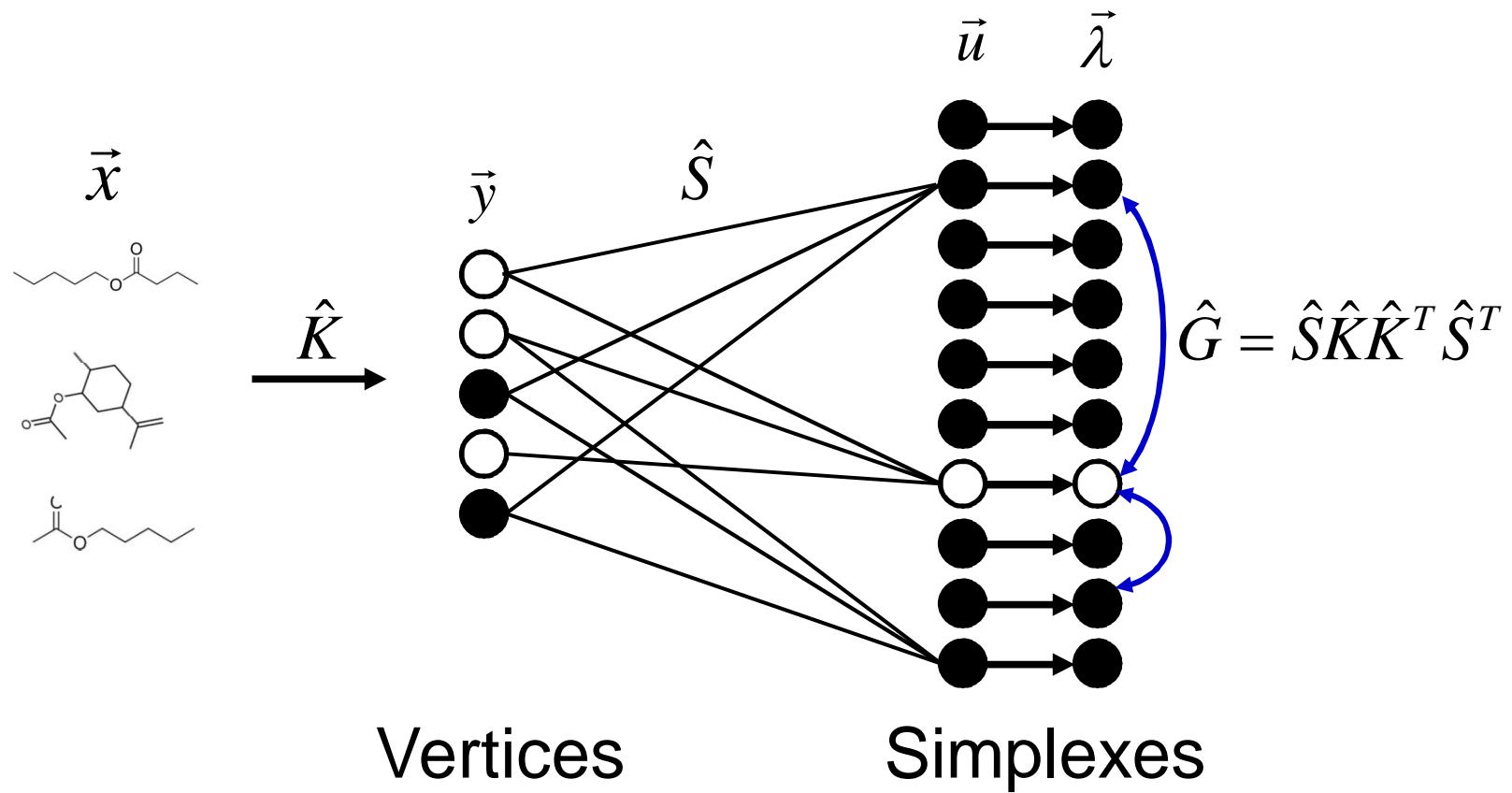


$$L(\vec{\lambda}) = -\varepsilon \sum_s u_s \lambda_s - \frac{1}{2} \sum_{sq} \lambda_s u_s G_{sq} u_q \lambda_q$$

↑  
Feedforward inputs      ↑  
Feedback

$$\hat{G} = \hat{S} \hat{K} \hat{K}^T \hat{S}^T$$

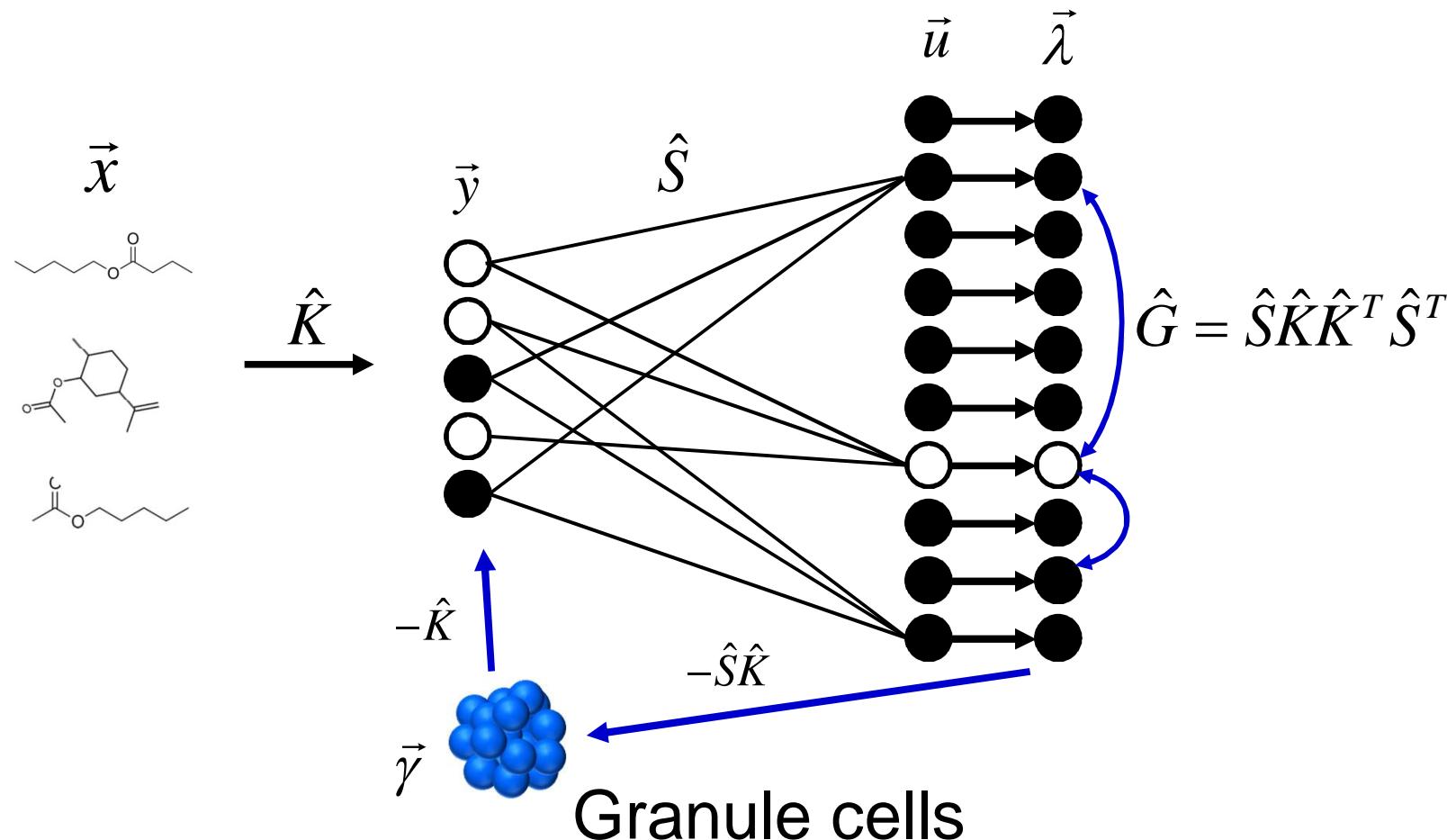
# Dual network



# Non-negativity constraint, $x_n \geq 0$

$$L(\tilde{x}, \vec{\lambda}) = \|\tilde{x}\|_\alpha - \sum_s \lambda_s u_s (\sum_n S_{sn} y_n - \varepsilon) - \sum_n \gamma_n x_n$$

$\lambda_n, \gamma_n \geq 0$



# Features of dual networks:

- Operate with Lagrange-Karush-Kuhn-Tucker multipliers rather than with original stimulus variables
- Rely on inequalities -> intensity-invariant encoding
- Implement inequalities easily ( $\lambda_n \geq 0$ )
- Sparse activity vectors
- New set of inequalities -> new set of  $\lambda$ -s -> new cell type

# Conclusions:

- Primacy model: small number of receptors activated first code for odorant identity
- More general idea – neural relativity – the implementation of invariant stimulus percepts in dual networks.

# Thanks to:

## Collaborators:

### Olfaction

- Dima Rinberg (NYU)
- Roman Shusterman (Haifa)
- Florin Albeanu (CSHL)
- Steve Shea (CSHL)
- Venki Murthy (Harvard)
- John Lisman (Brandeis)

### Lab Members

- Brian Kolterman
- Yi Wei
- Daniel Kepple
- Hamza Giaffar

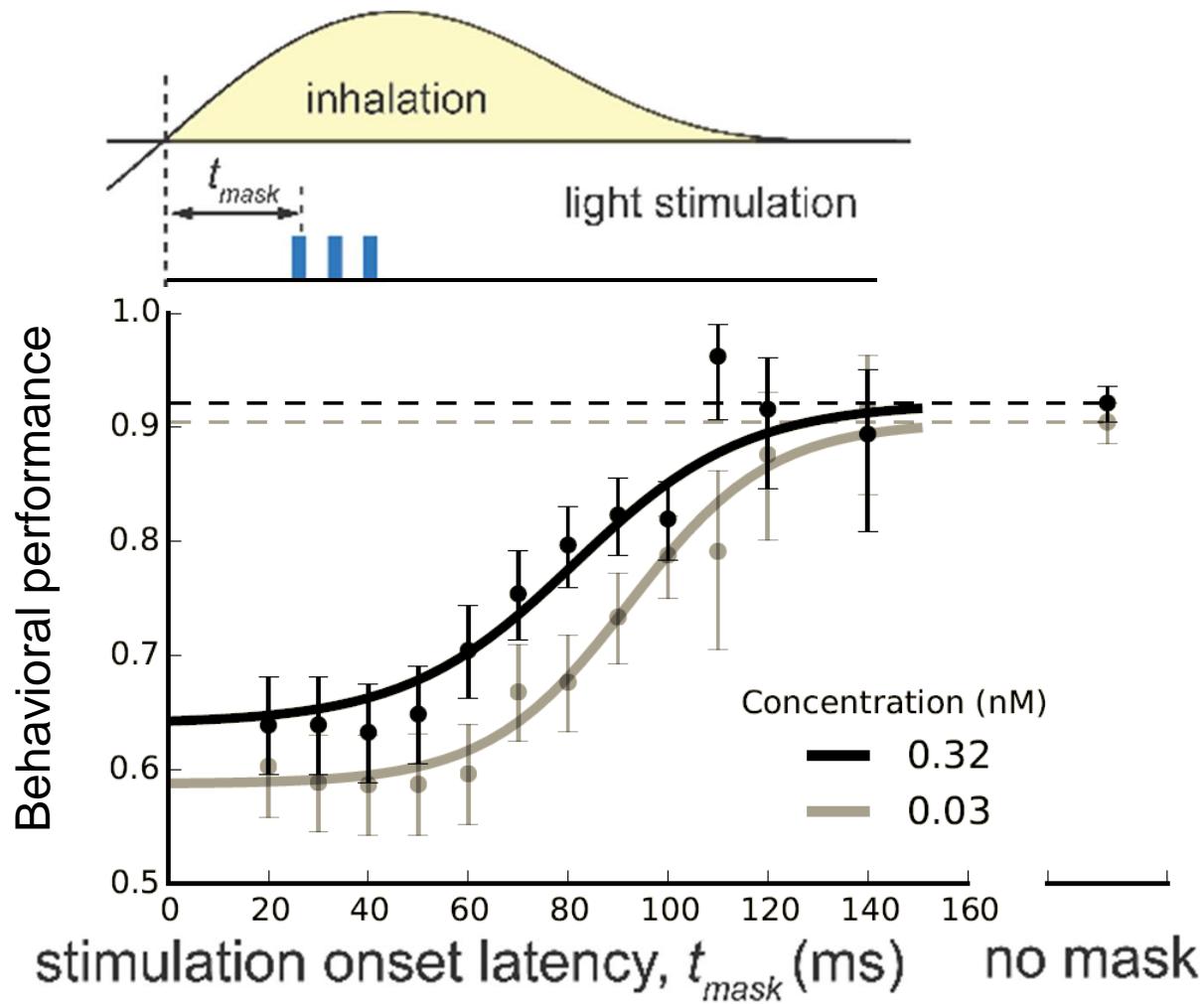
### Animation

- Dancing-lemon-studio.com

## Funding:



# Smell A vs smell B



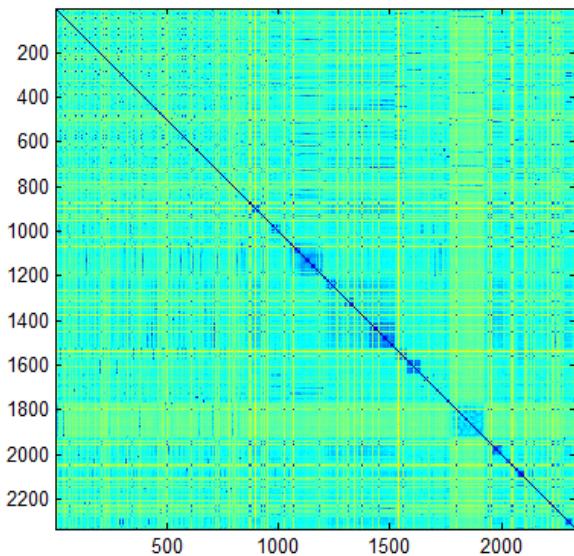
Chris Wilson (NYU)



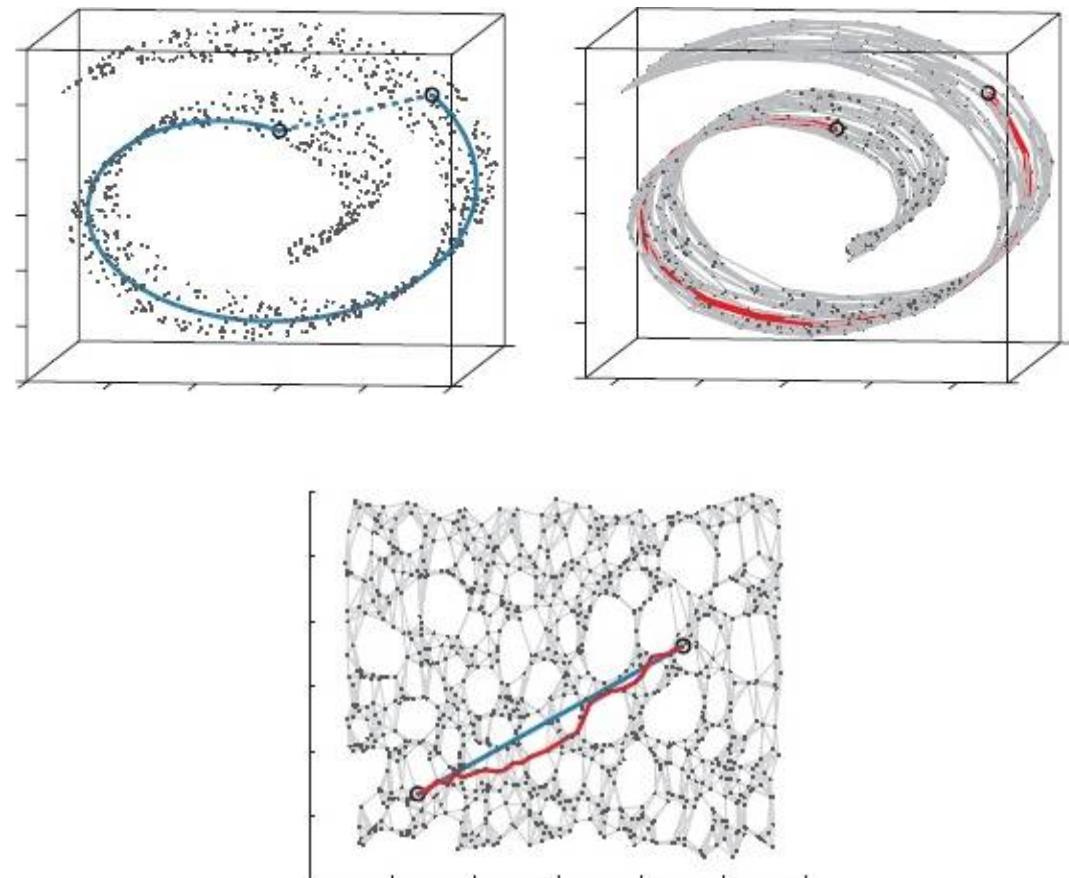
Dima Rinberg (NYU)

# Mouse+human OR Isomap algorithm

OR – OR distance matrix

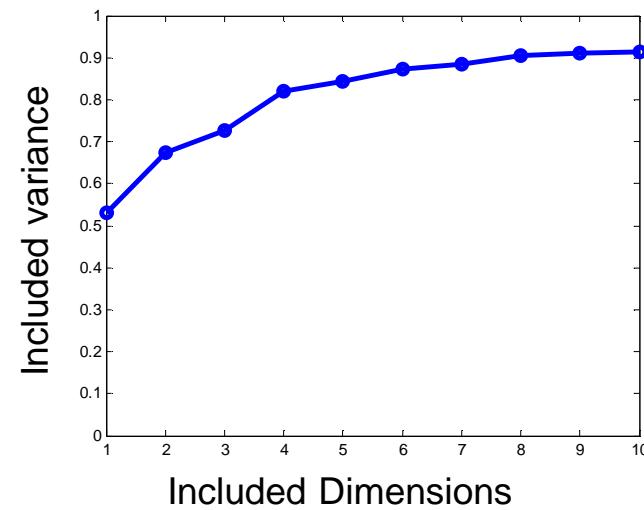
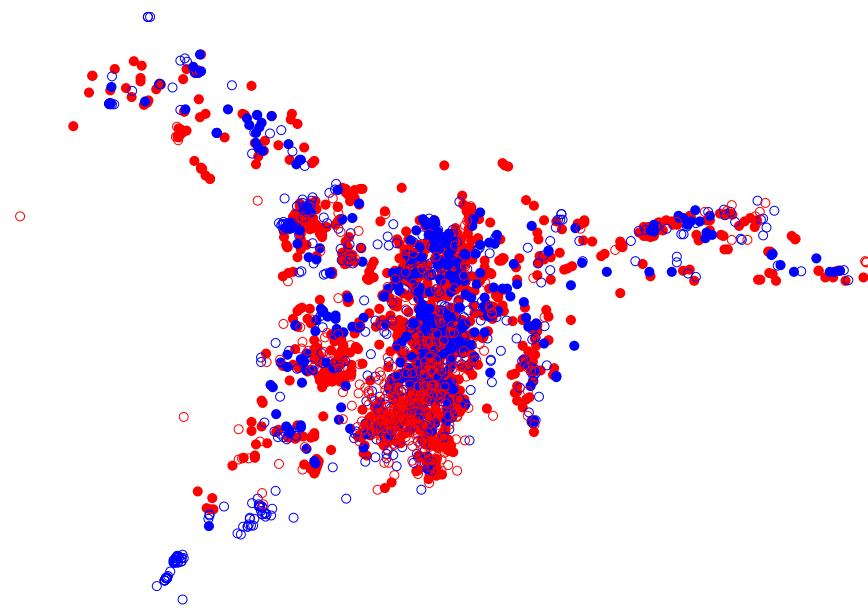


Isomap algorithm



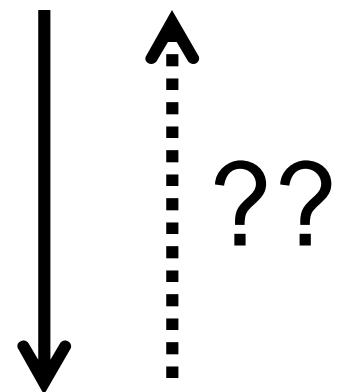
Tenenbaum, de Silva and Langford (2000)

# Mouse+human OR Isomap space



Can primacy model provide insight into the emergence of low-dimensional olfactory manifolds?

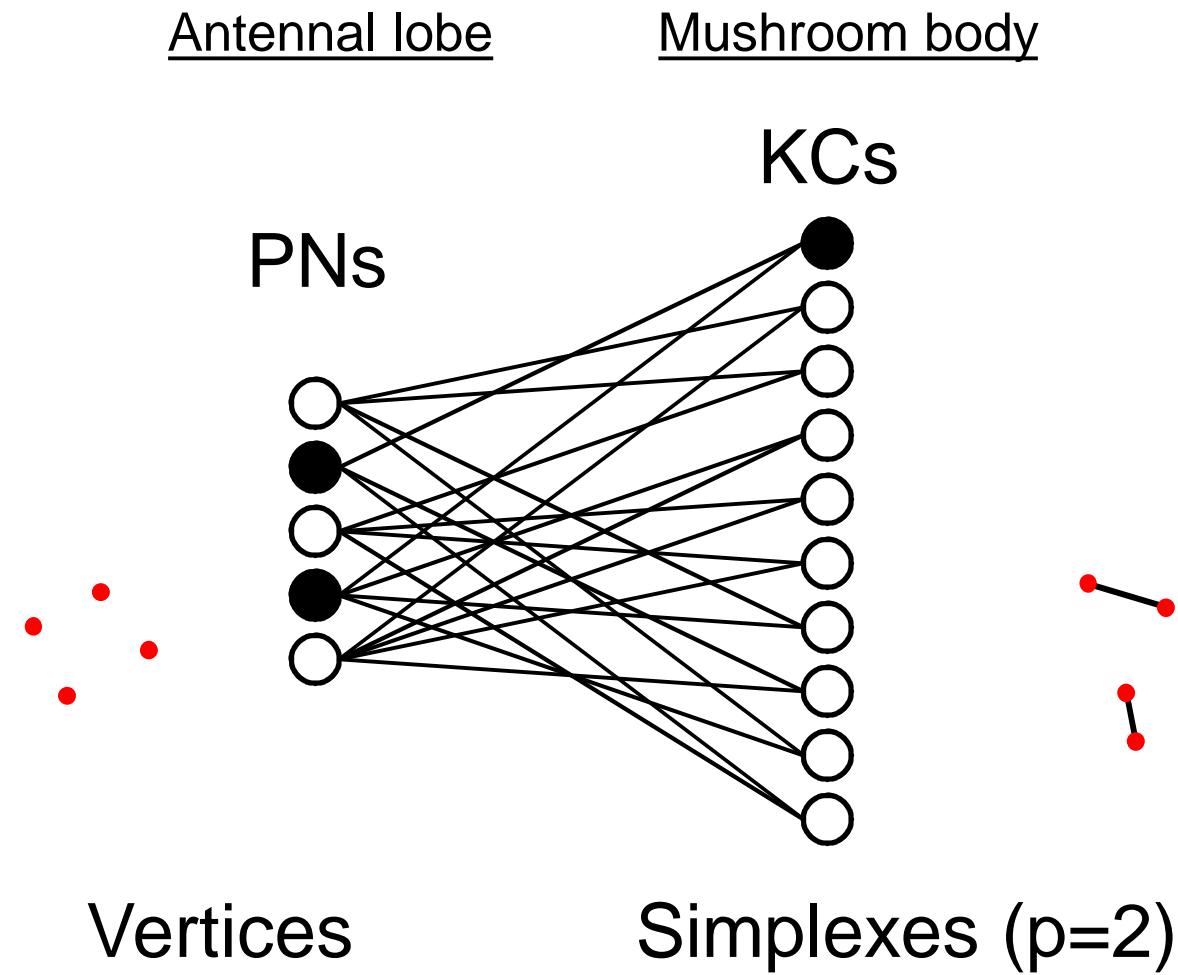
# Dimension of olfactory space, D



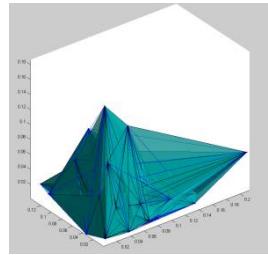
Primacy model ( $p \sim D$ )

# Primacy and connectivity

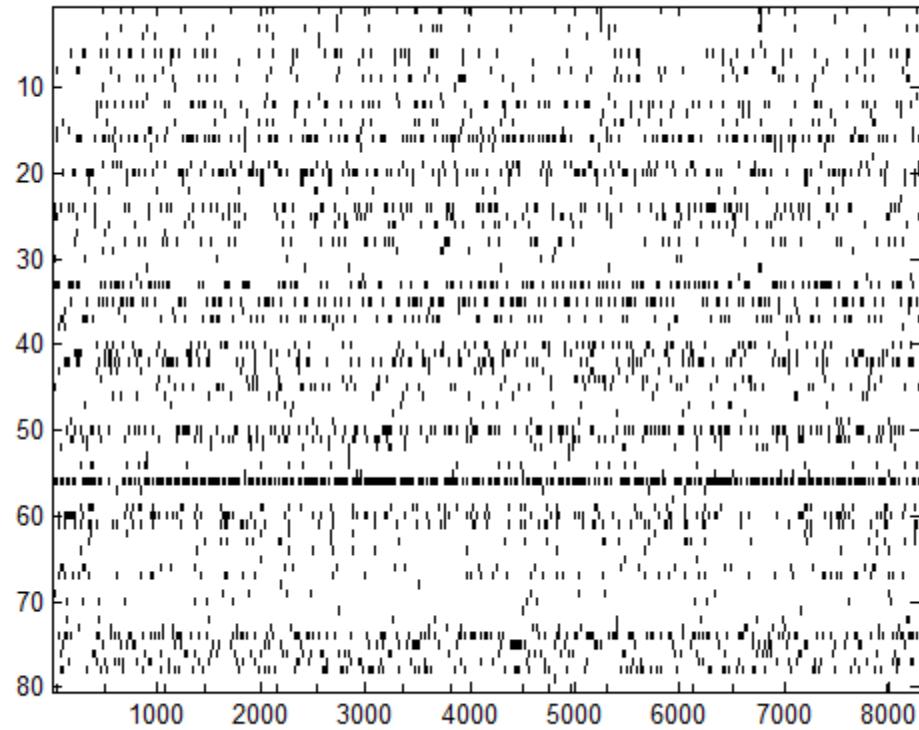
# Fly olfactory system



# Connectivity between glomeruli (PN, vertices) and KC (simplexes) looks random

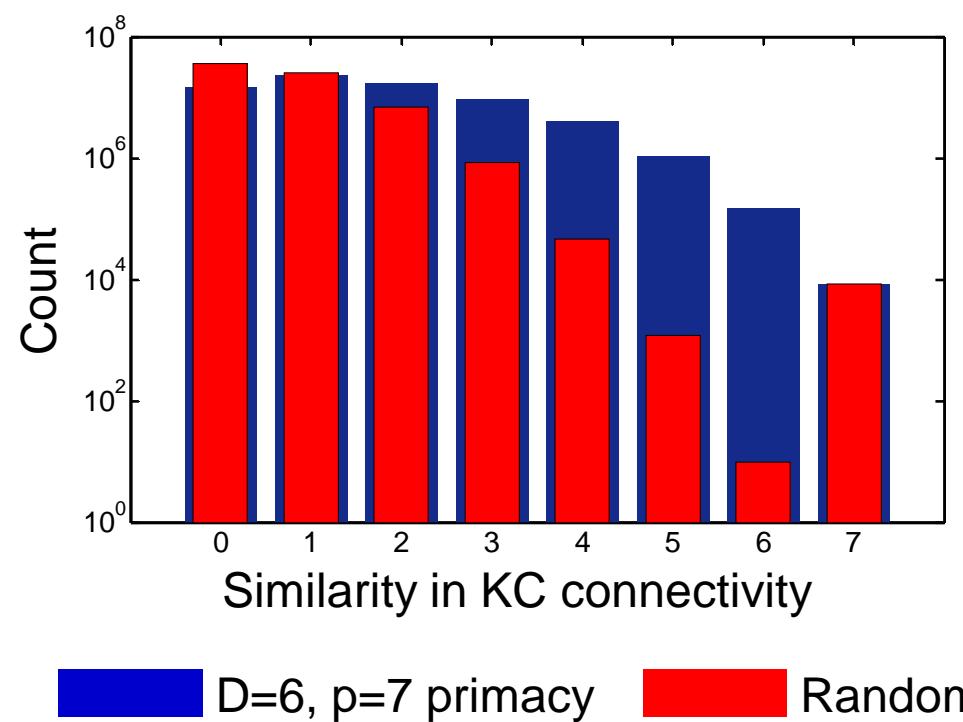
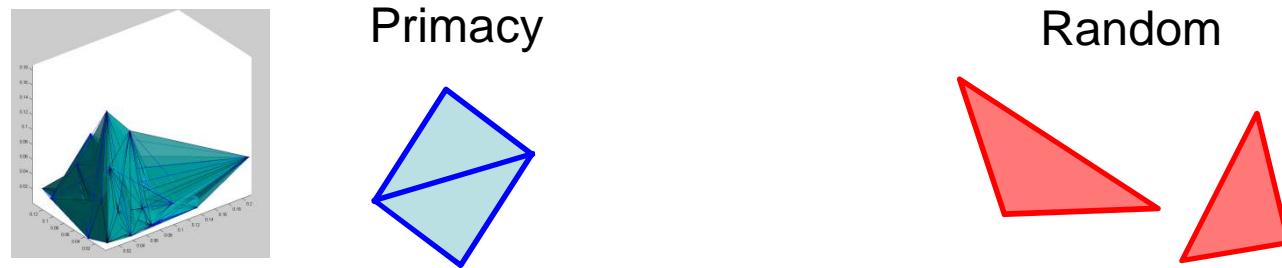


Vertices  
(glomeruli)



Simplexes  
(kenyon cells)

# Higher-order correlations in connectivity



# Conclusions:

- Primacy model: small number of receptors activated first code for odorant identity
- Primacy code favors the representation of low-dimensional olfactory manifolds
- The network architecture based on sparse connectivity between AL and MB is ideal for implementing primacy coding

# Thanks to:

## Collaborators:

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- Andreas Schaefer (MPI)
- John Lisman (Brandeis)
- Ivan Iossifov (CSHL)

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- Toma Marinov
- Daniel Kepple

### Animation

- Dancing-lemon-studio.com

## Funding:

