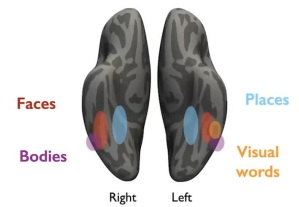


# Topological organization

Tianqin Li  
Sep 11, 2024

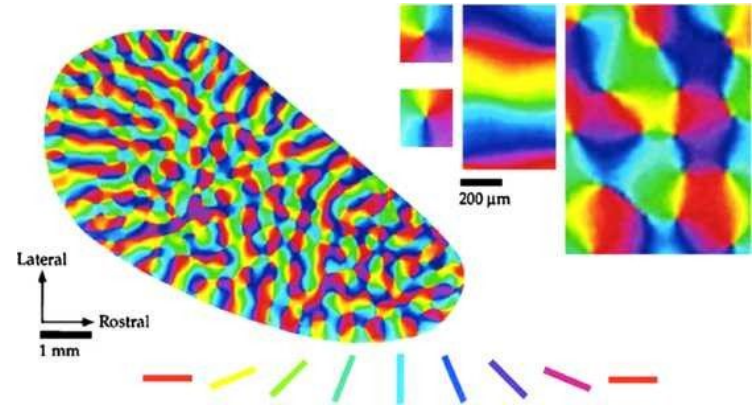
# Functional organization



Functional organization is among the most ubiquitous of neuroscience findings, appearing in the topographic maps of the visual system, and in auditory, parietal, sensorimotor, and entorhinal areas

What processes govern their emergence?

What computational function they serve?

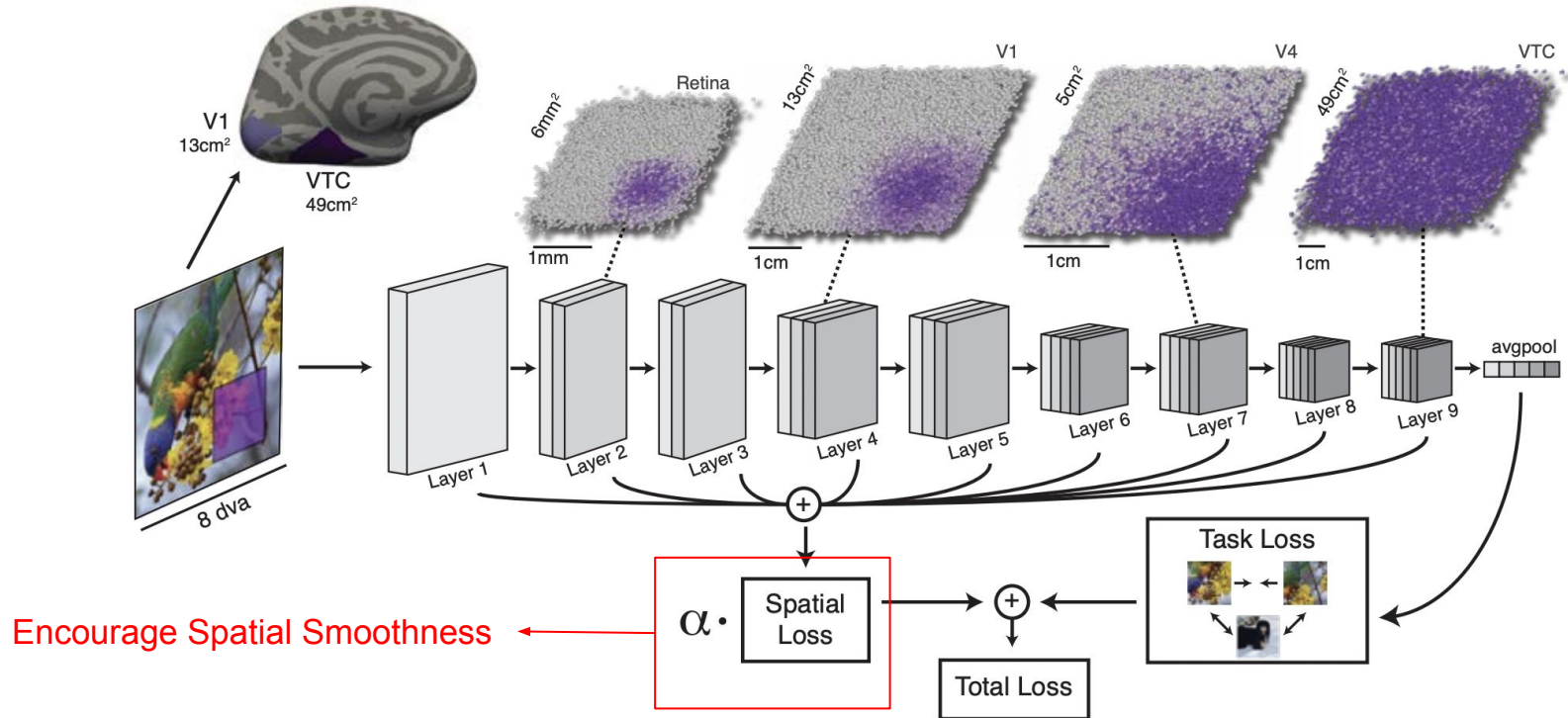


Orientation tuning in V1

# Hypothesis

- Wiring Cost Hypothesis:
  - Neuron fire together wire together.
  - To minimize total wiring cost
- Multiple Behavior Demand Hypothesis:
  - Function specialization emerges evolutionary to parallel process distinct tasks

# Topographic Deep Artificial Neural Network (TDANN)



# Method

*Spatial loss* The spatial loss function encourages nearby model units on the simulated cortical sheet to be correlated in their responses to the training stimuli. Specifically,  $SL_l$  is the spatial correlation loss computed for the  $l$ -th layer and  $SL_l$  is computed on a given batch by randomly sampling a local cortical zone and calculating for pairs of units, (1) correlation (Pearson's  $r$ ) between the response profiles,  $(\vec{R})$ , and (2) the the stabilized reciprocal Euclidean distances  $(\vec{D})$ :

$$\vec{D} = \frac{1}{(1 + \vec{d})} \quad (\text{S1})$$

where  $\vec{d}$  is the vector of pairwise cortical distances. These two terms are then related as follows:

$$SL_l = 1 - \text{Corr}(\vec{R}, \vec{D}) \quad (\text{S2})$$

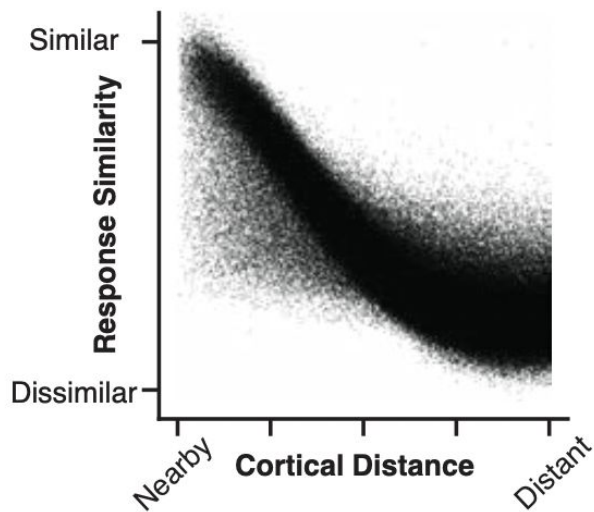
such that  $SL_l$  is minimized when nearby units have correlated responses to the training stimuli.

# TDANN hurts classification performance

Method	ImageNet classification accuracy
Baseline (ResNet-18)	48.5%
TDANN (ResNet-18 + spatial loss)	43.9%

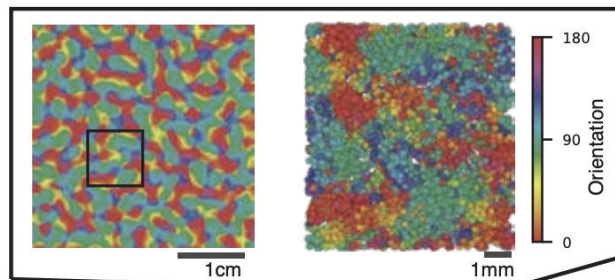
# TDANN successfully encourage spatial arrangement

## Spatial Loss Encourages Local Correlations

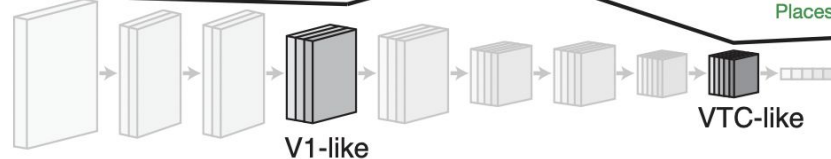
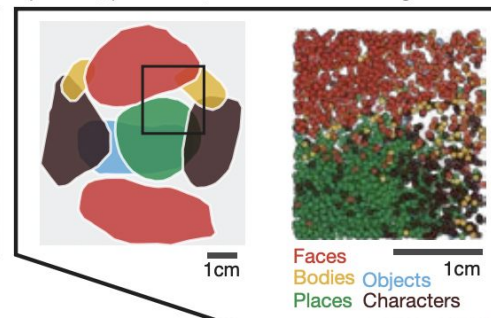


## Evaluate Quantitative Topographic Benchmarks

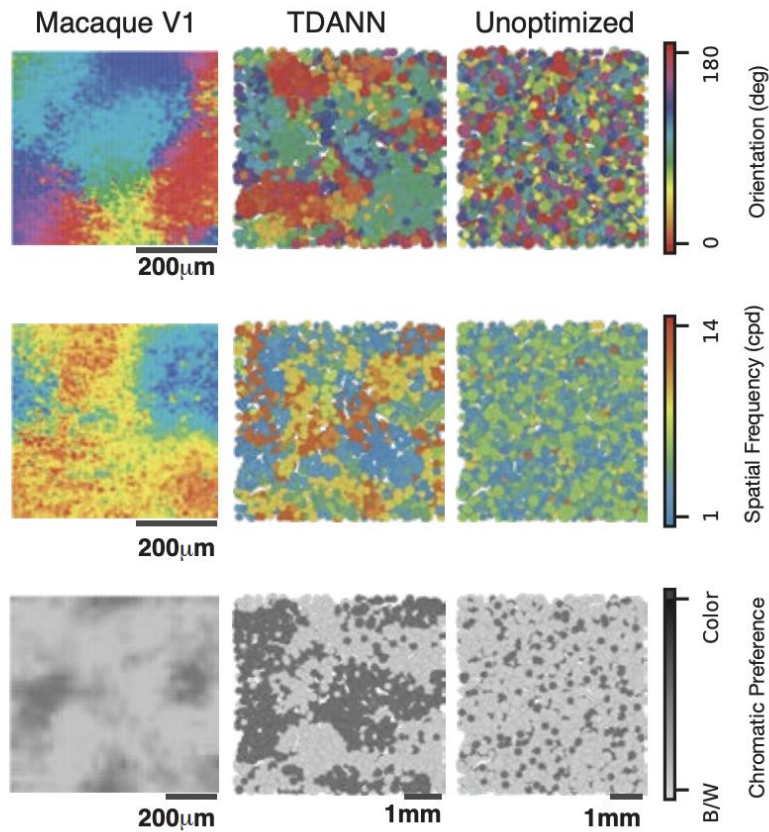
**V1:** Orientation preference map smoothness, pinwheel density, smoothness of spatial frequency and chromatic maps



**VTC:** Category selectivity map smoothness, number of patches, patch size, colocalization of categories

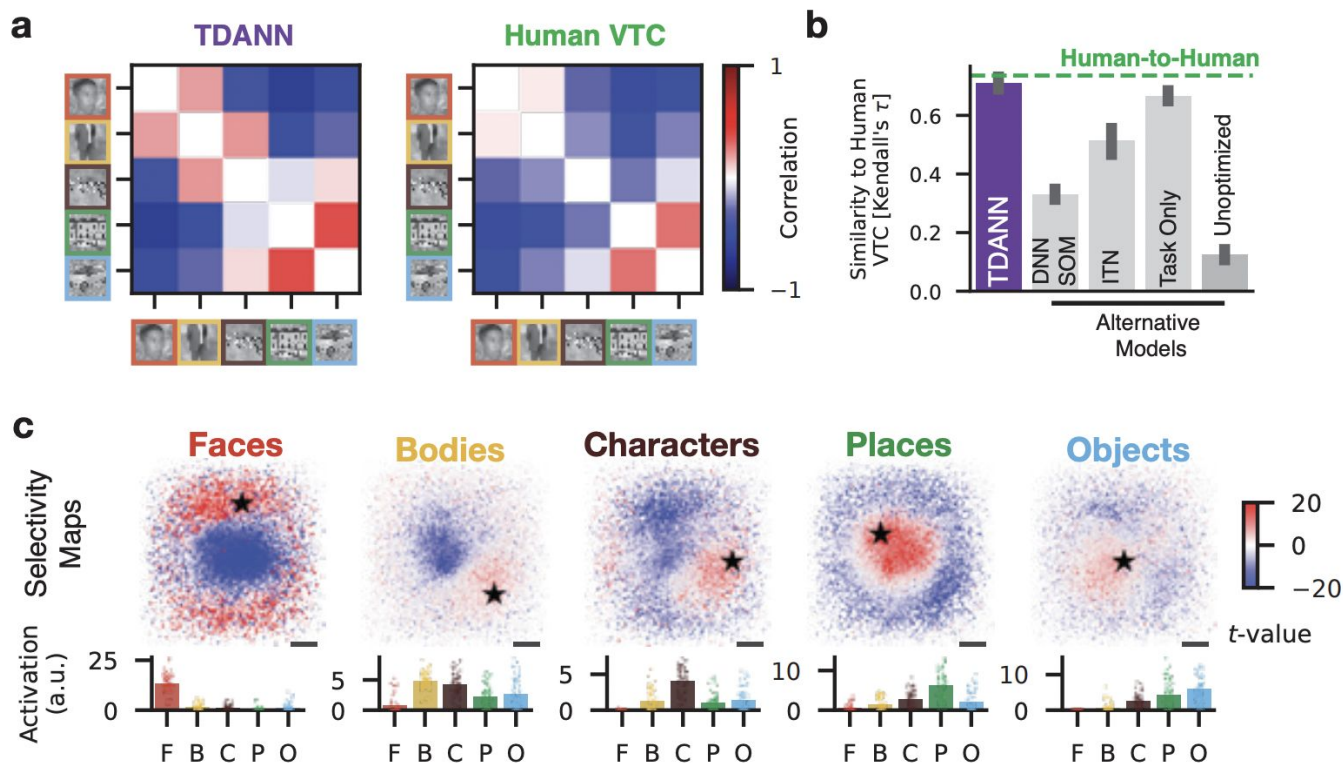


# TDANN successfully encourage spatial organization in V1

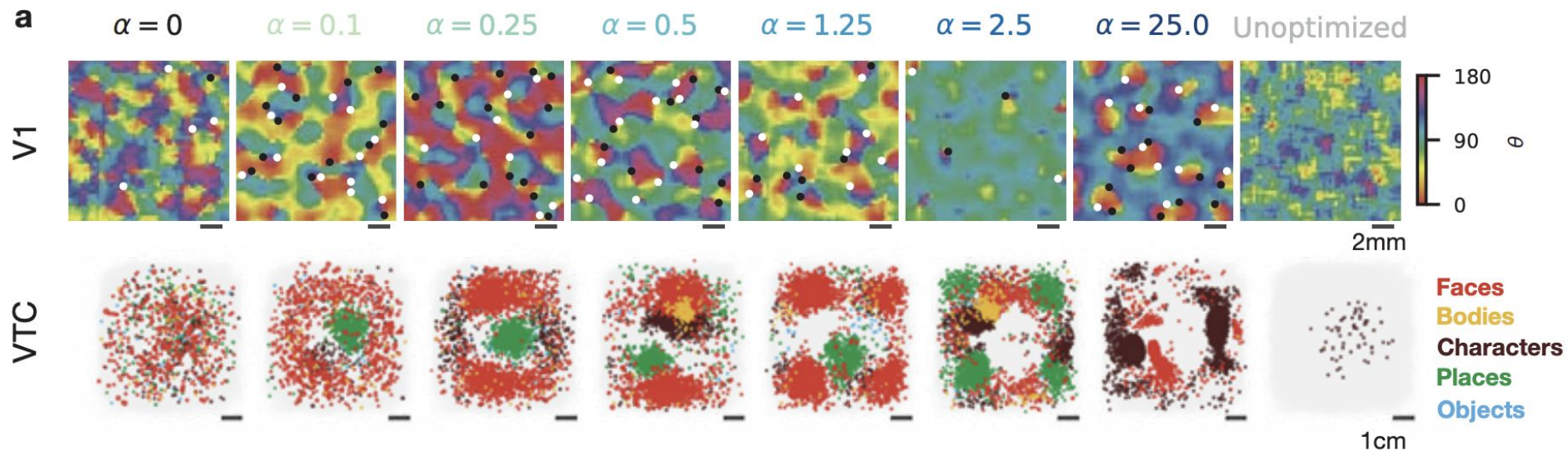




# TDANN reproduce spatial organization in ventral temporal cortex



# Topological map in TDANN

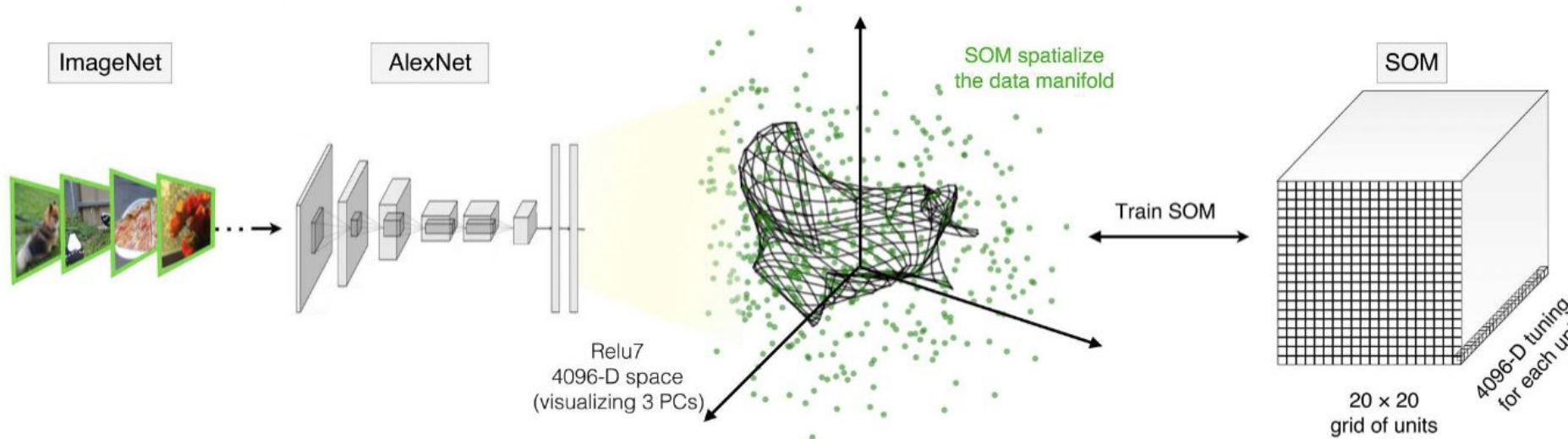


# Other Topological Organization Generation

## DNN-SOM

Doshi, Fenil R., and Talia Konkle. "Cortical topographic motifs emerge in a self-organized map of object space." *Science Advances* 9.25 (2023): eade8187.

- Train a self-organizing map



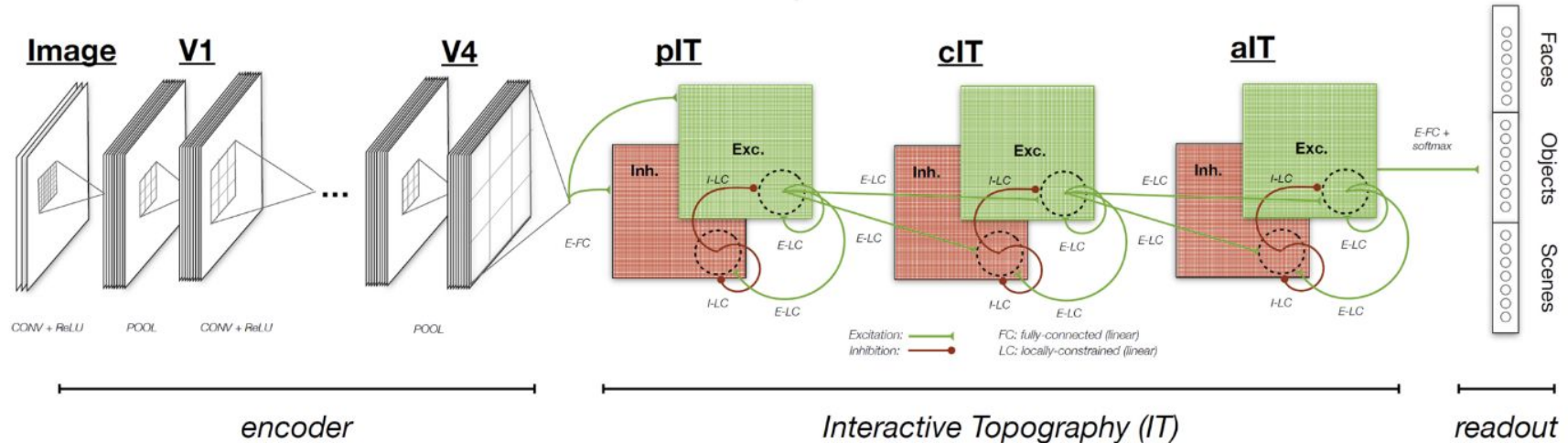
# Other Topological Organization Generation

$$L_w^{(a,b)} = \sum_{i,j} \frac{(D_{ij}^{(a,b)})^2 (W_{ij}^{(a,b)})^2}{1 + (W_{ij}^{(a,b)})^2}.$$

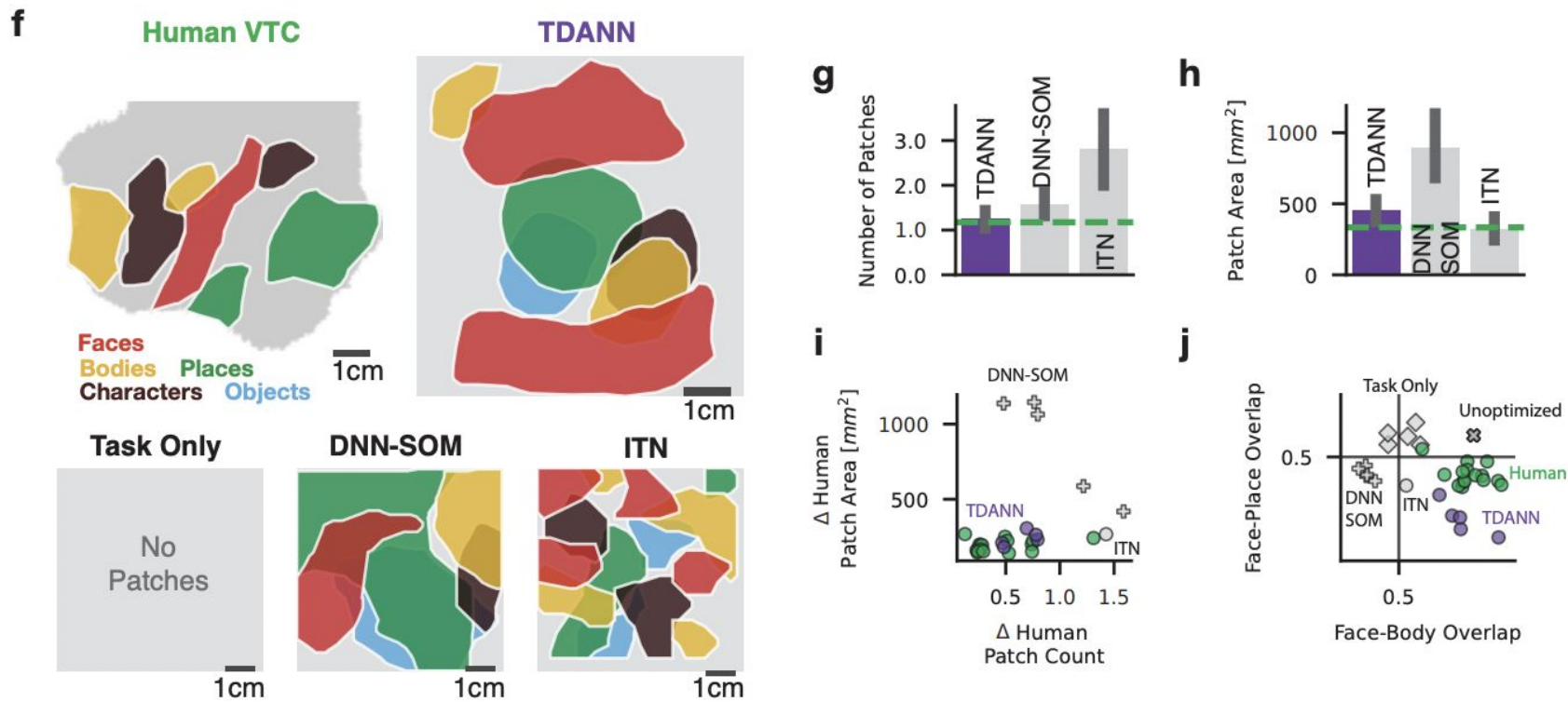
ITN

$$\tau \frac{dx_t^{(a)}}{dt} = -x_t^{(a)} + W^{(a,a)} r_t^{(a)} + W^{(a-1,a)} r_t^{(a-1)} + b^{(a)}$$

Blauch, Nicholas M., Marlene Behrmann, and David C. Plaut. "A connectivity-constrained computational account of topographic organization in primate high-level visual cortex." *Proceedings of the National Academy of Sciences* 119.3 (2022): e2112566119.

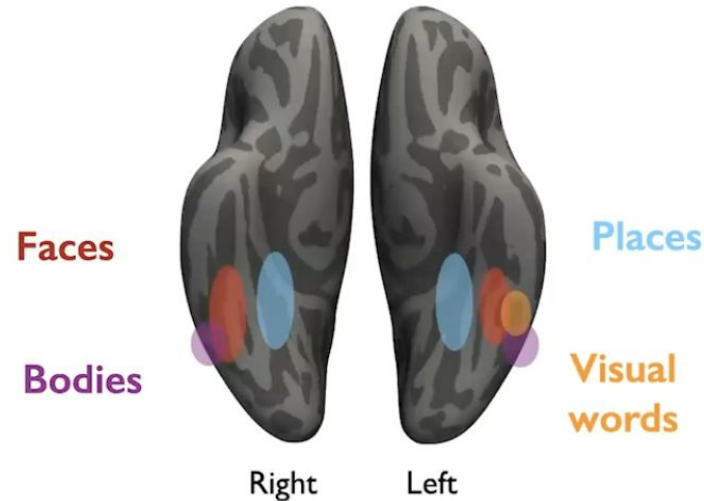


# TDANN is doing better than previous topological networks



# Another perspective of functional specialization

Topological organization is due to functional specialization of modules





# Simulate different region by separately trained CNNs

100 held-out identities



Face CNN

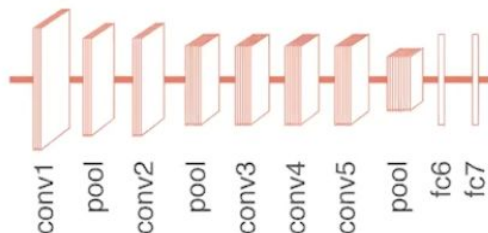
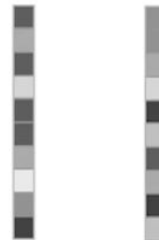
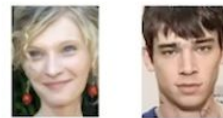


Image-wise  
feature  
extraction



100 held-out  
object categories



Object CNN



# Swap the input domain and use the feature to predict



Face CNN

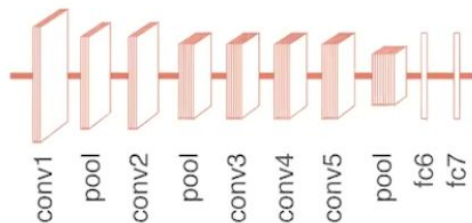


Image-wise  
feature  
extraction

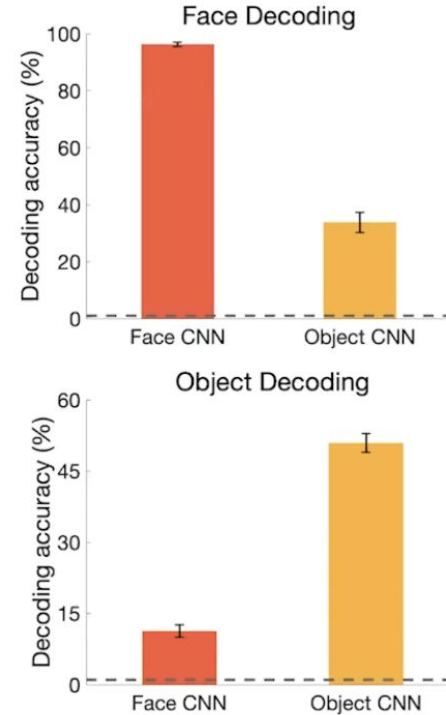
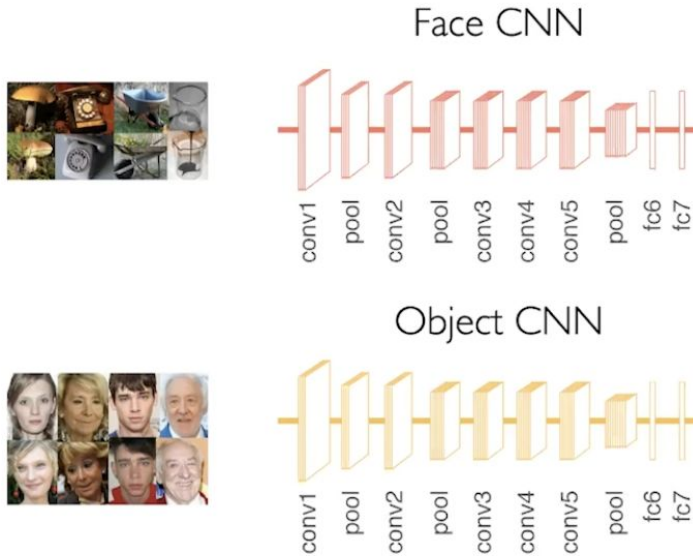


Object CNN





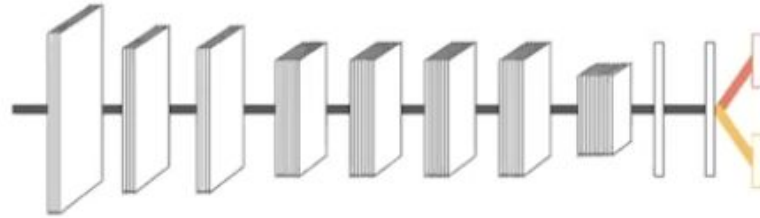
# Functional Specialization using two models



Face trained Network can't see object very well and vice versa

# What happens if we use a shared networks?

Fully-shared dual-task CNN

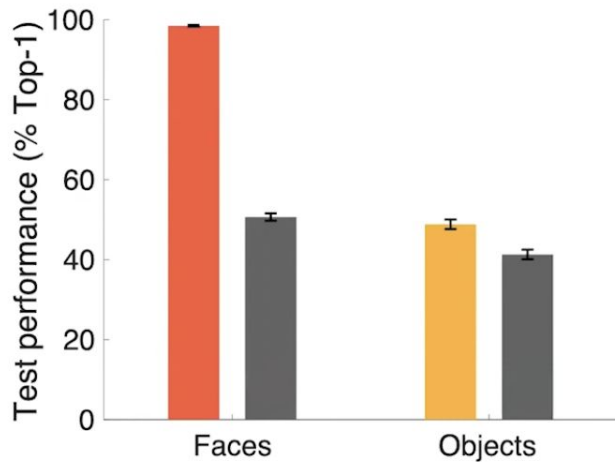
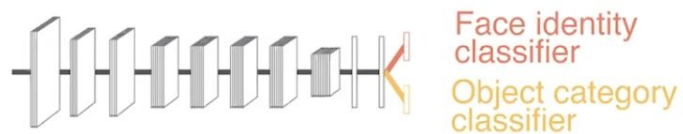


Face identity  
classifier

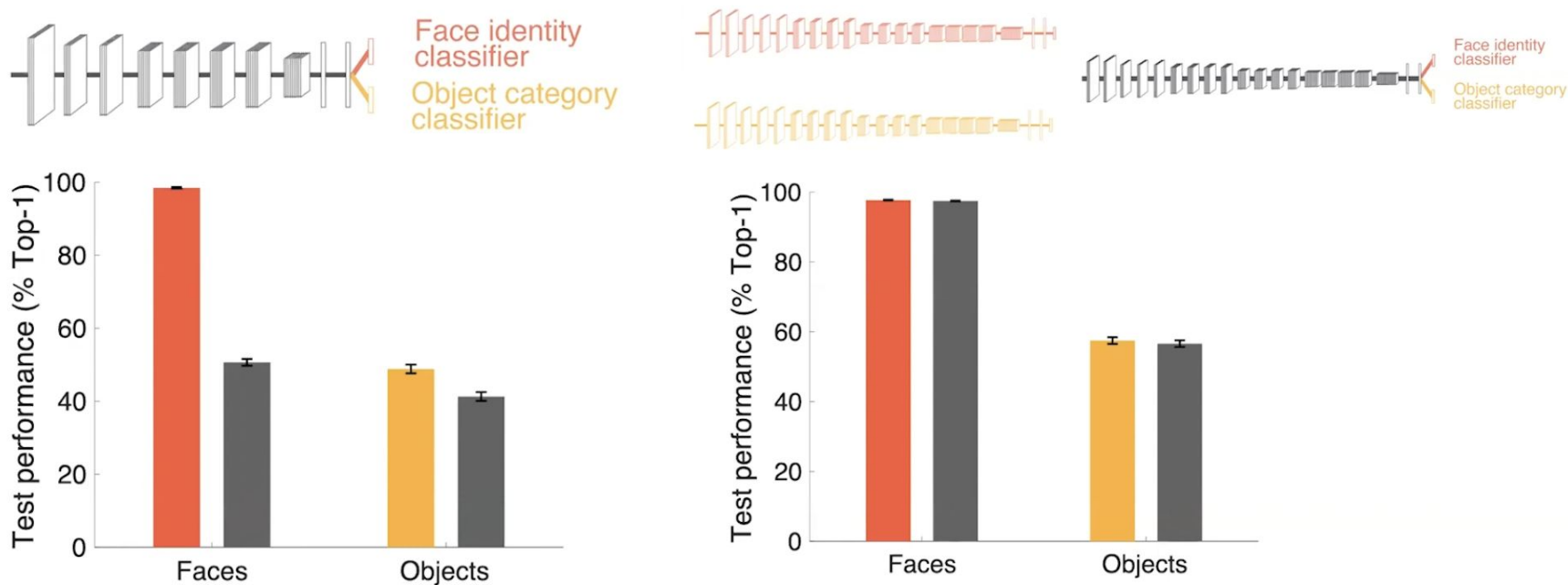
Object category  
classifier

This is expected to perform worse than the separation specialists

# Experiments on shared networks

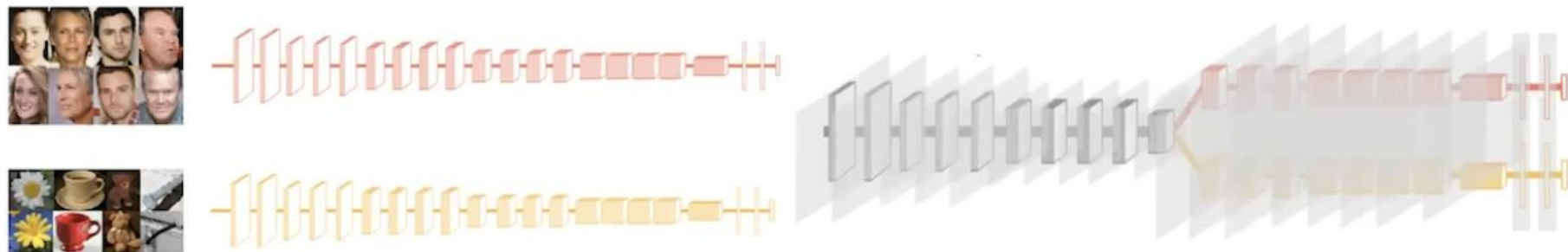


# Experiments on shared networks



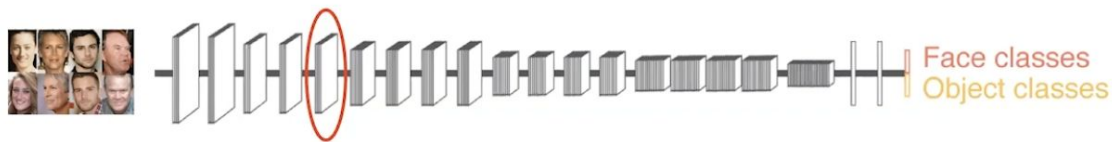
Larger networks can have better capability of incorporating both tasks

# Can functional separation achieved automatically?

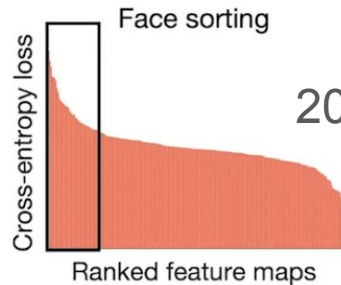


Did the network discover the segregation automatically?

# Lesion test to drop specialist neurons

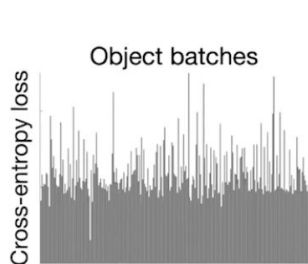
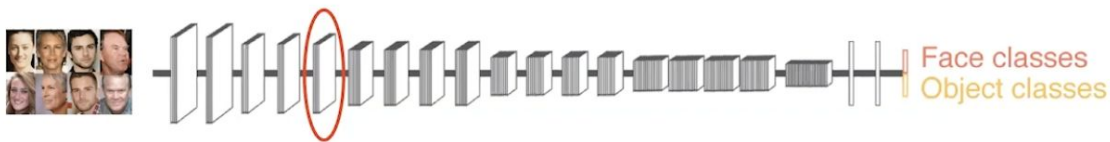


Feature maps



20% feature maps that are specialized to face

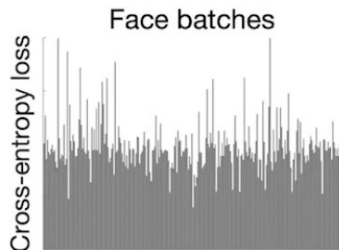
# Lesion test to drop specialist neurons



Feature maps



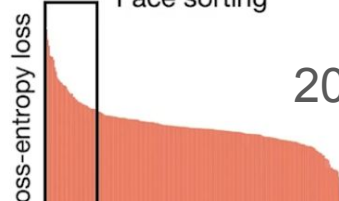
Object sorting



Feature maps

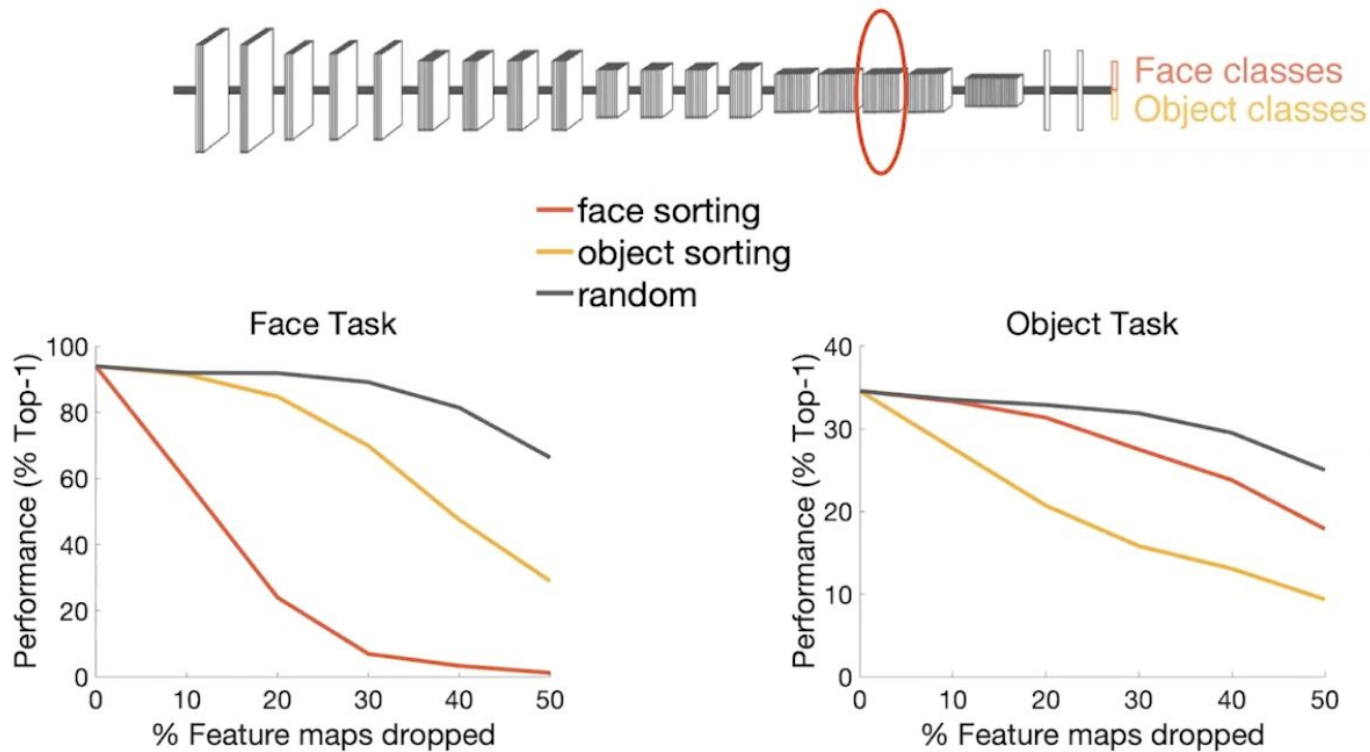


Face sorting



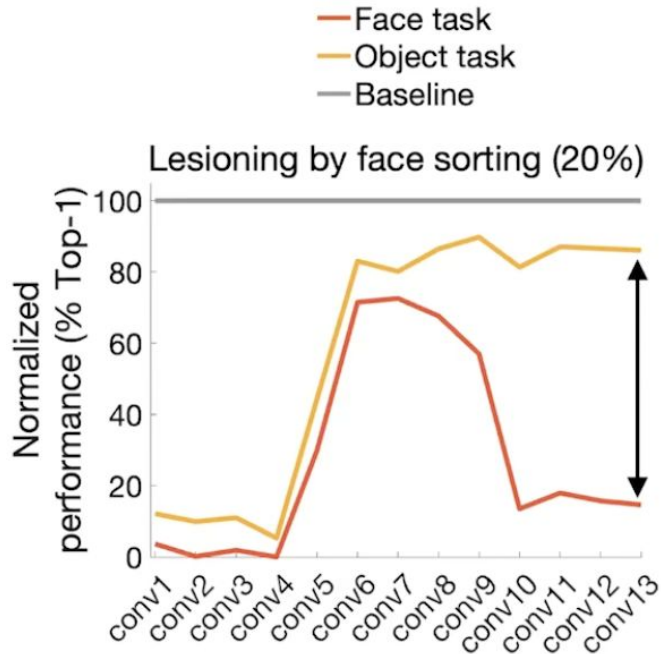
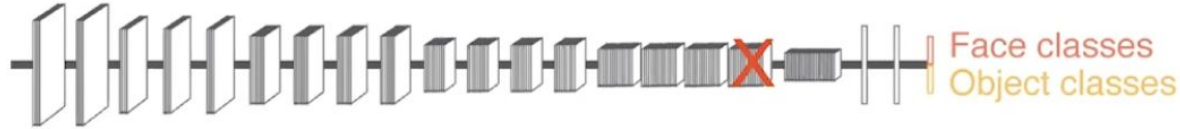
20% feature maps that are specialized to face

# Functional segregation in the large networks

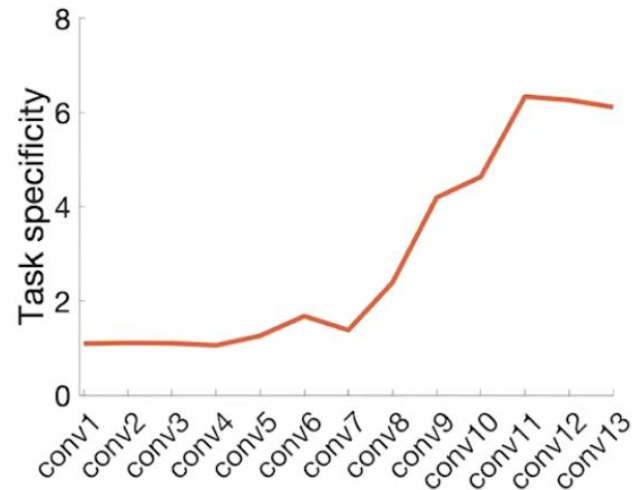




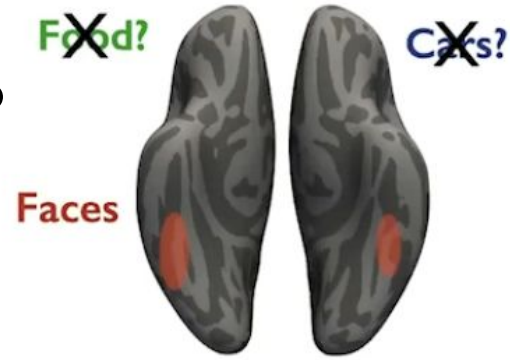
# At which stage the system become specialized?



**Control:** replicates with matched performance!



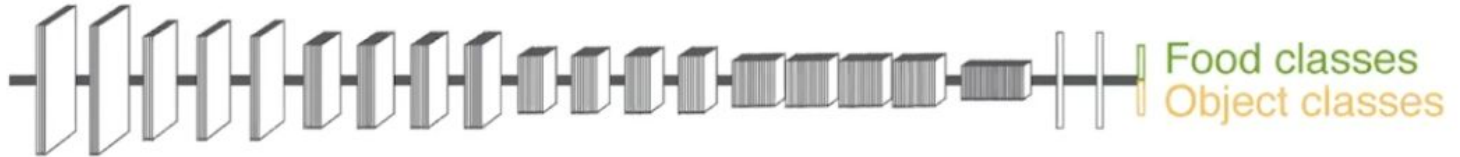
# Do any two tasks requires specialization?



Food-101 dataset:  
101 categories



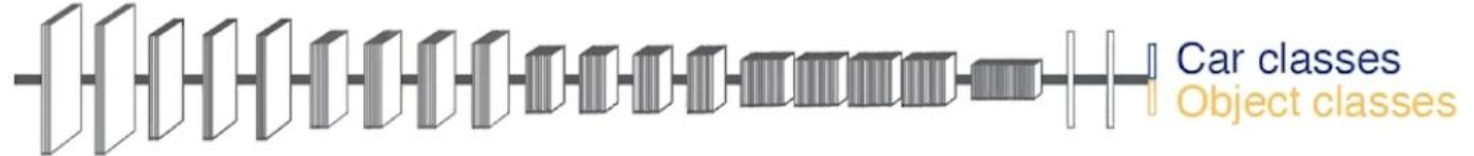
Food - Object CNN



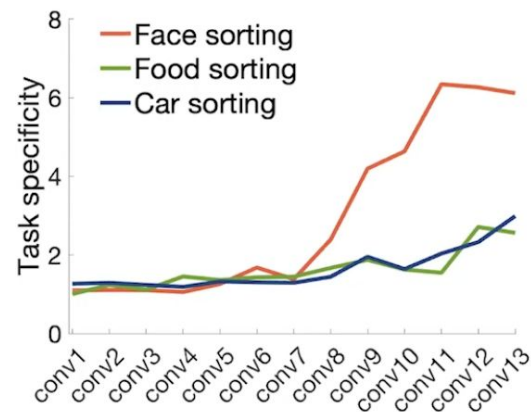
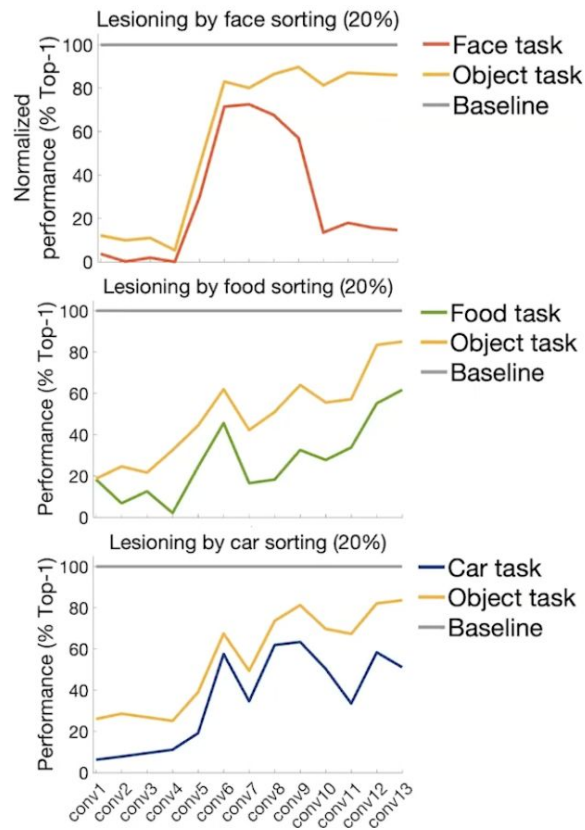
Car dataset:  
1109 categories



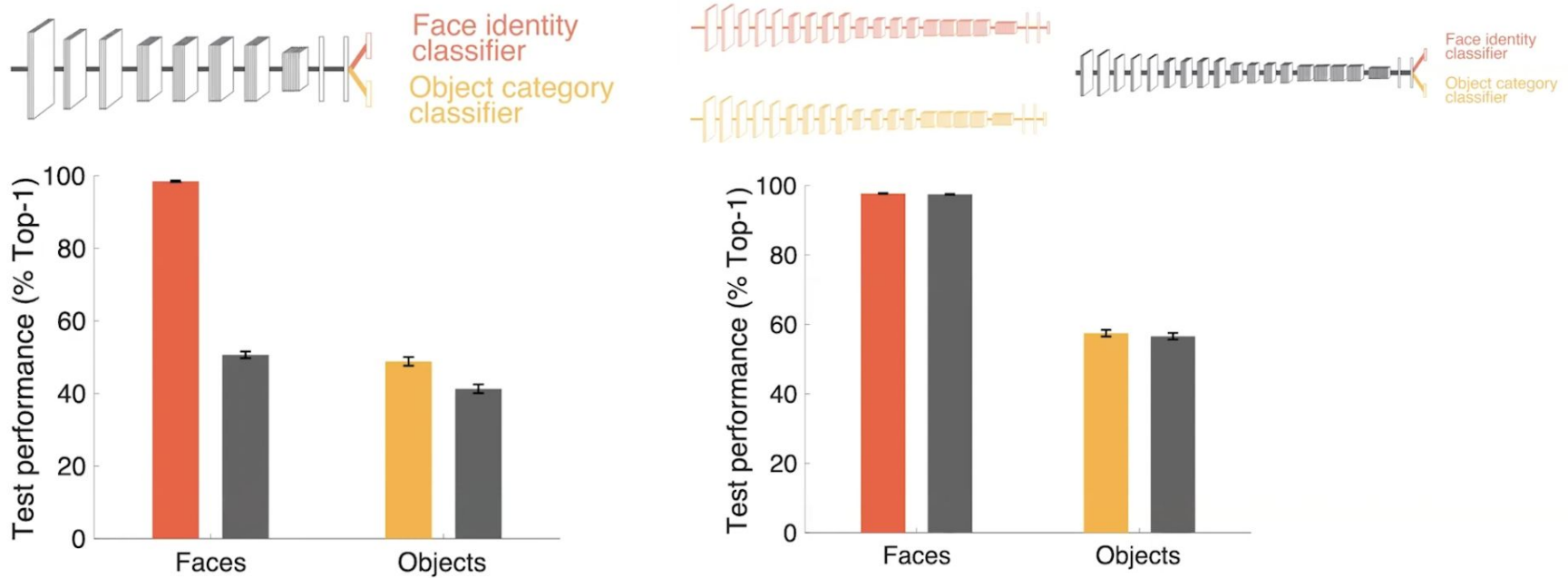
Car - Object CNN



# Food and Car shows less segregation than Face / Objects



# Benefits of having such segregation



Having such segregation capability will boost the shared network performance

# Topological Organization

- The above experiments suggest another possible explanation for why we observe the topological organization

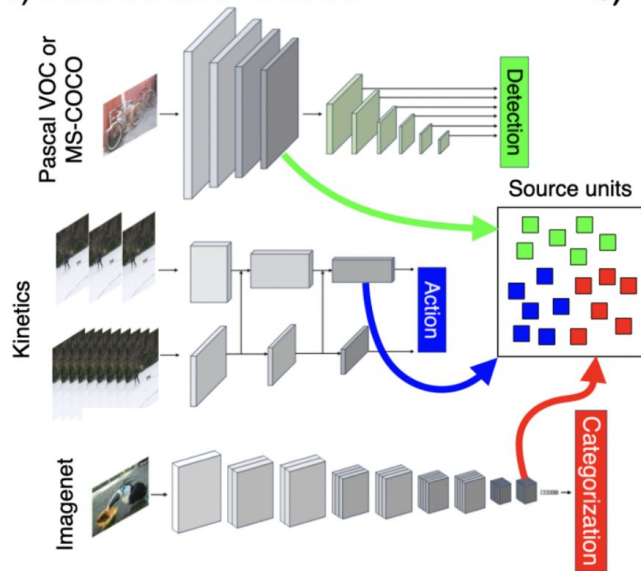
They might emerge as functional segregation due to the nature of dataset

Each region can be a specialized module for prototypical tasks

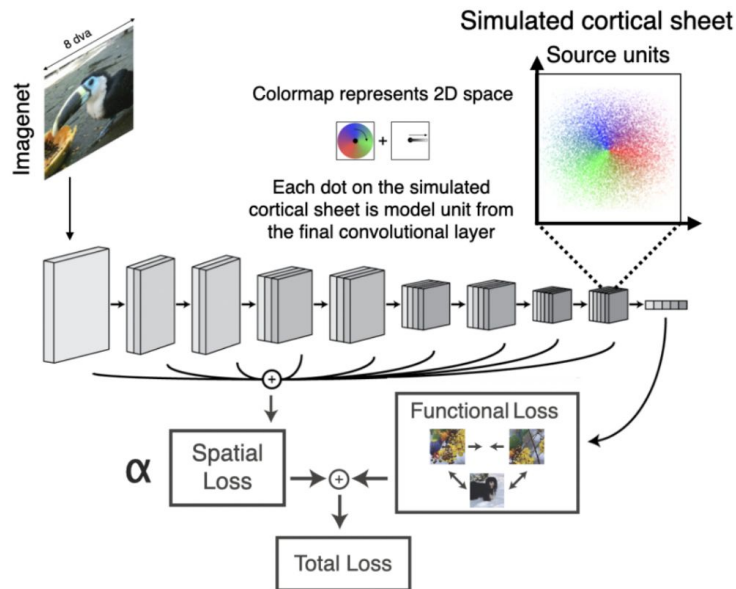
- Brings the question of the necessity of having explicit loss to enforce smooth topological structures.
- Such segregation has clear benefit of performance improvement

# Brain data mapping: TDANN v.s. Functional Specialists

a) Multi-behavior models



b) Topographic Deep Artificial Neural Network (TDANN)



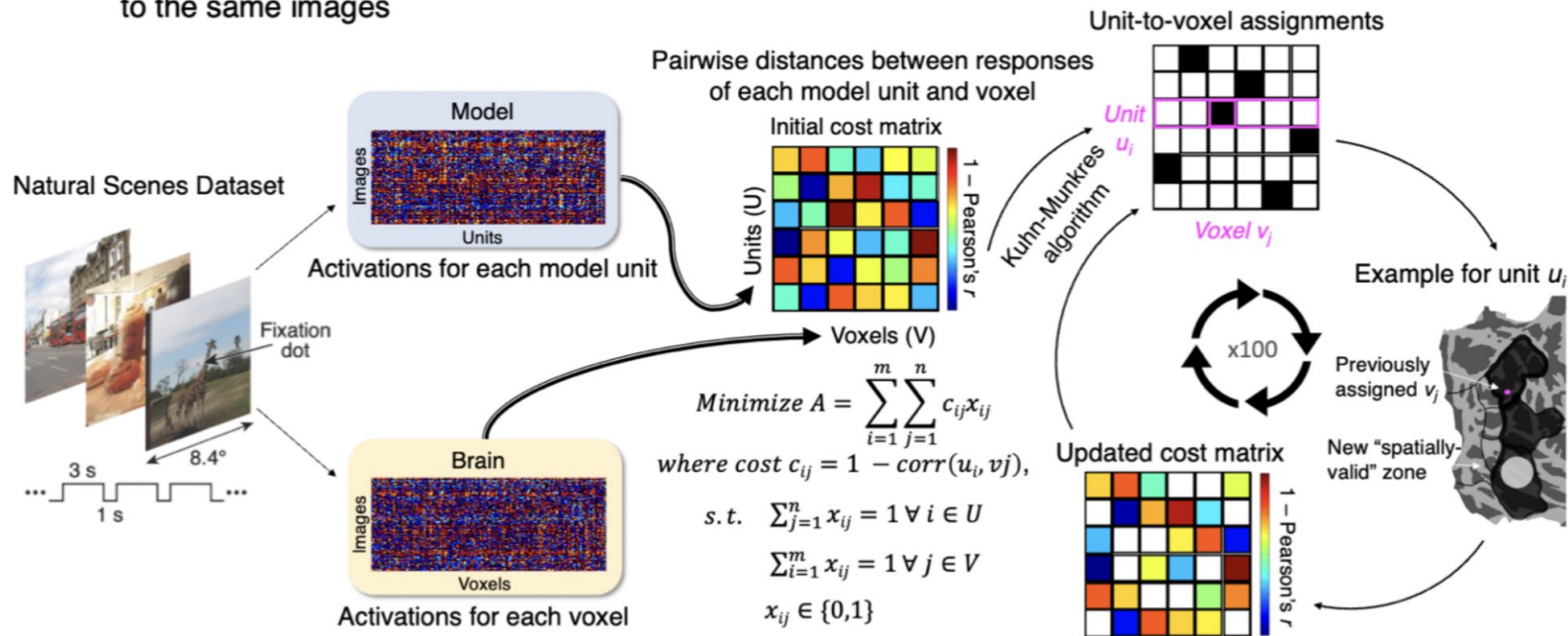
Who is better match for the brain data? (Finzi, Dawn, et al, 2023)

# Mapping model neurons to brain voxels

## c) Linking models to brains

Measure brain & model responses to the same images

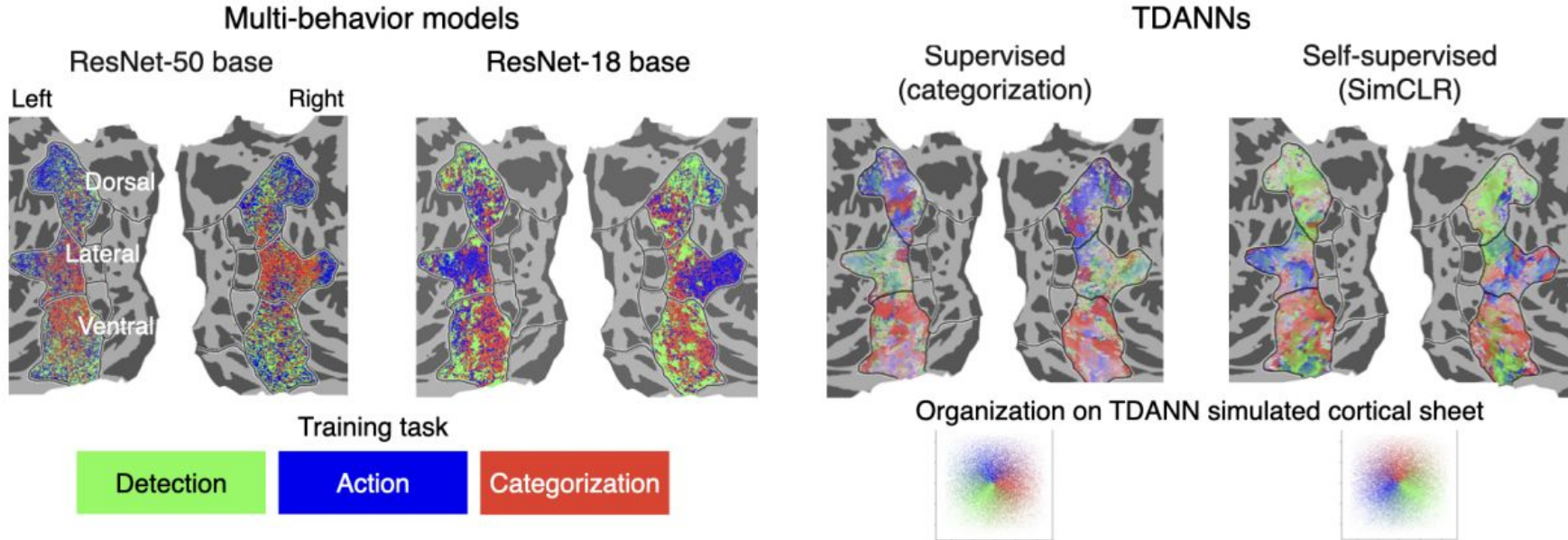
Find optimal 1-to-1 mapping that minimizes cost using iterative algorithm





# TDANN matches the brain topology better

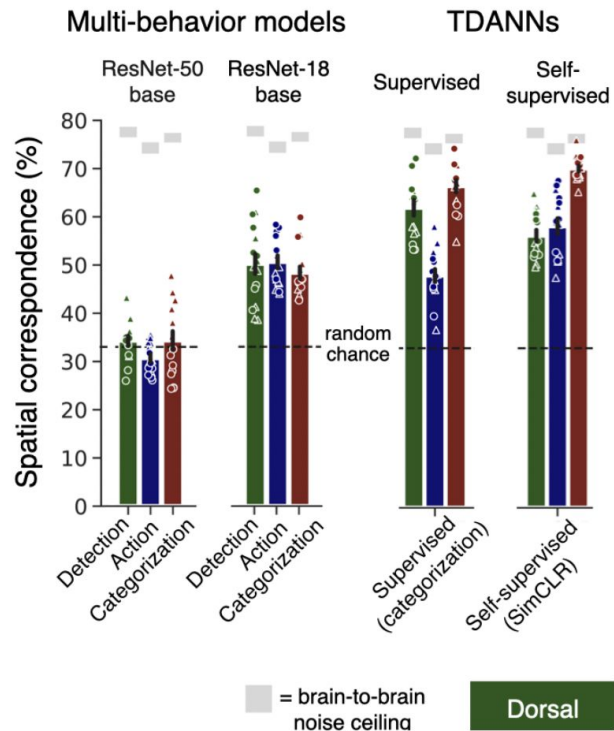
## a) Model to brain topography



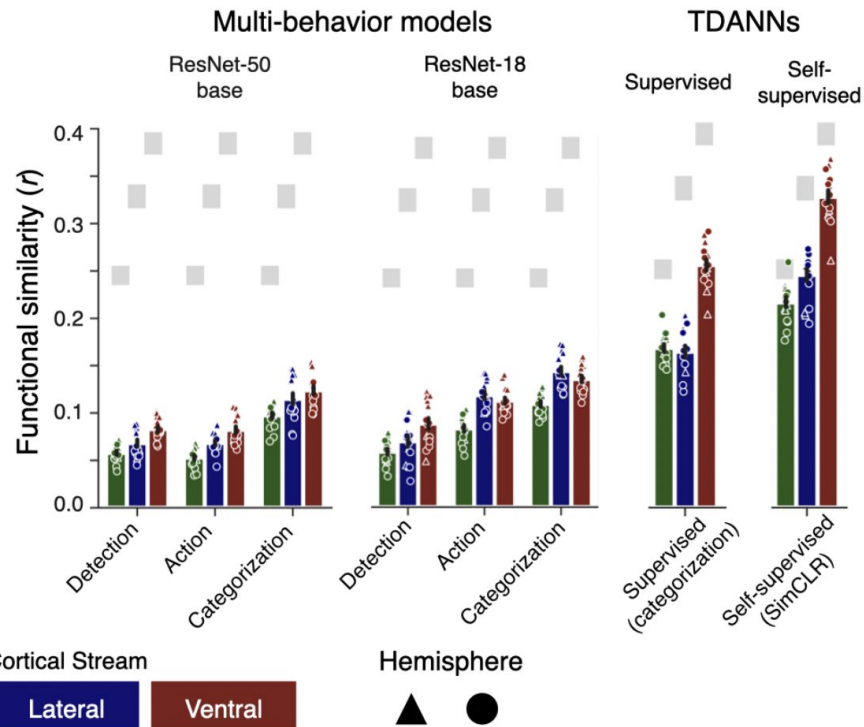


# Spatial and functional correspondence

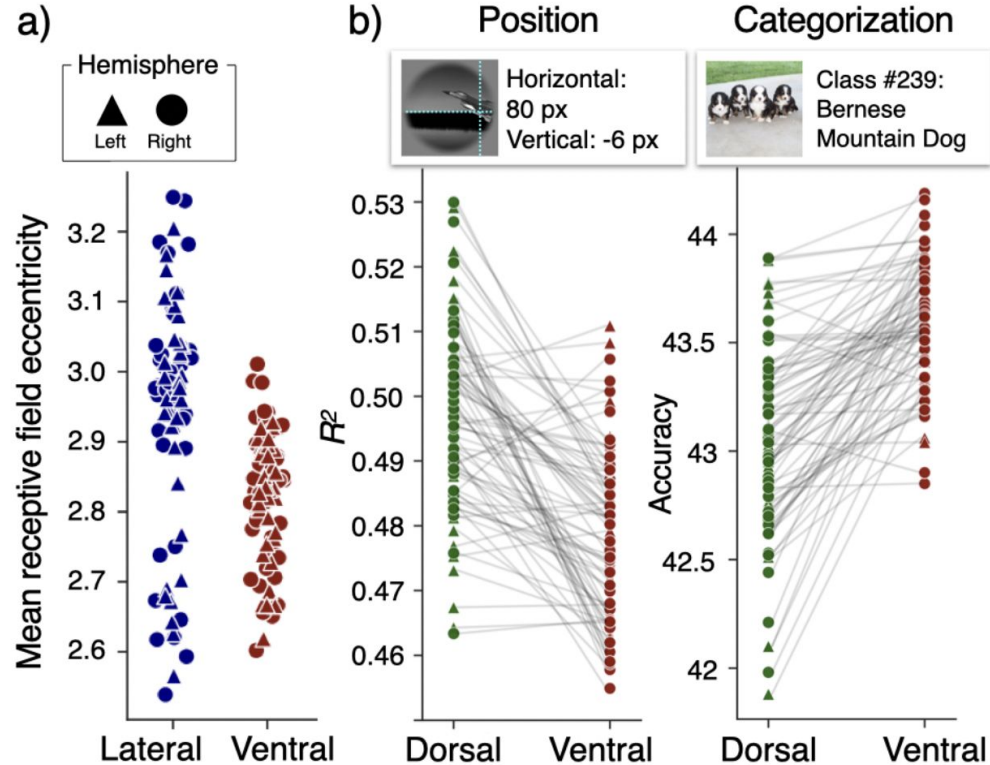
b) Model to brain spatial correspondence



c) Model to brain functional correspondence



# Functional segregation emerges from the TDANN



# Summary

- Two hypothesis on the emergence of topological structure.
- TDANN is most brain like
- Most of the work is not focused on the computational advantages of topological organization

## Potential advantages:

- Learn faster (Convergence)
- Caching system (inference faster)

# Future direction

- Closer look at the recurrent connections