

Neural Embeddings Rank: Aligning 3D latent dynamics with movements

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NeurIPS 2024 + NeuroAI and SSL Workshops (Oral)

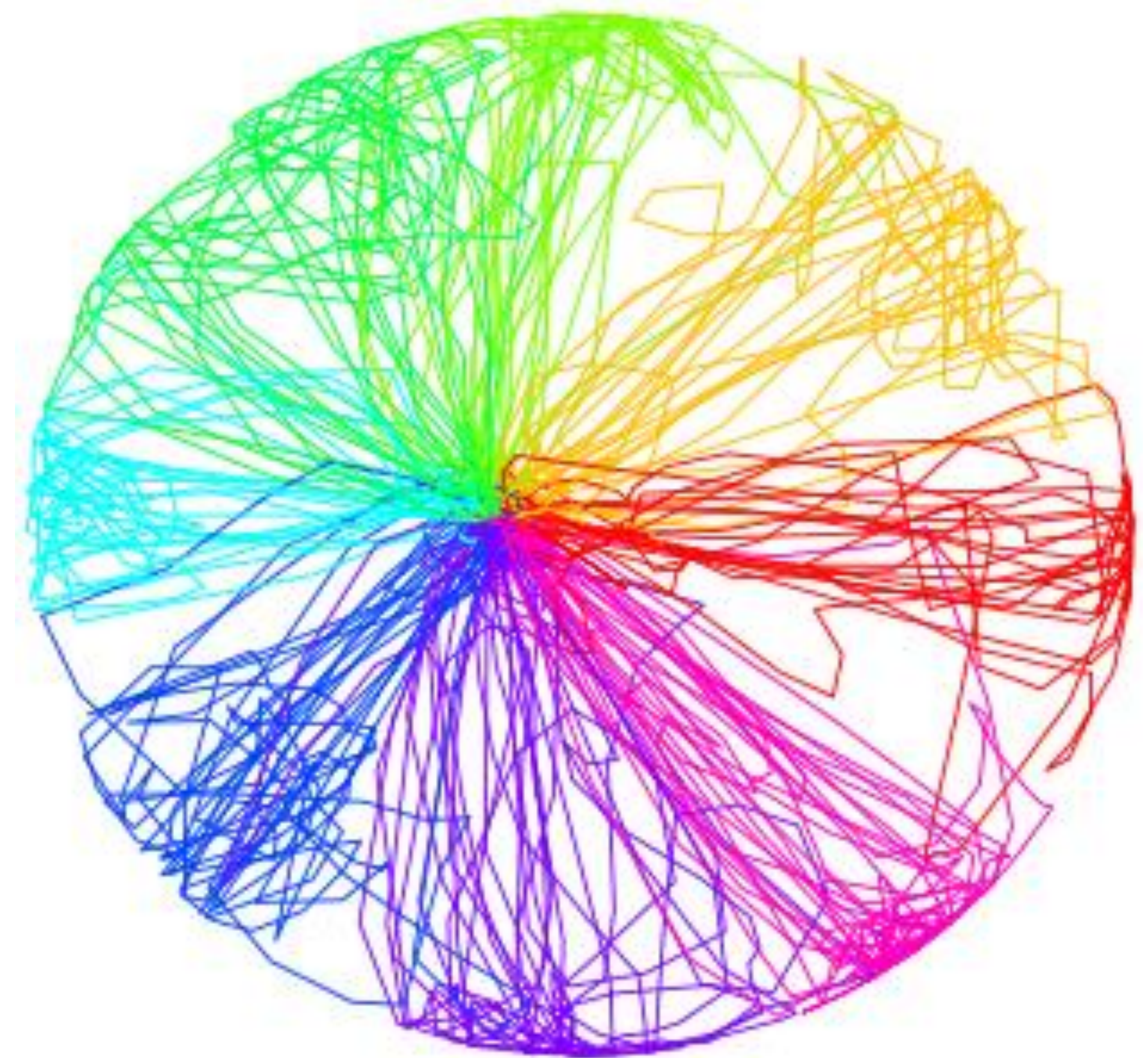
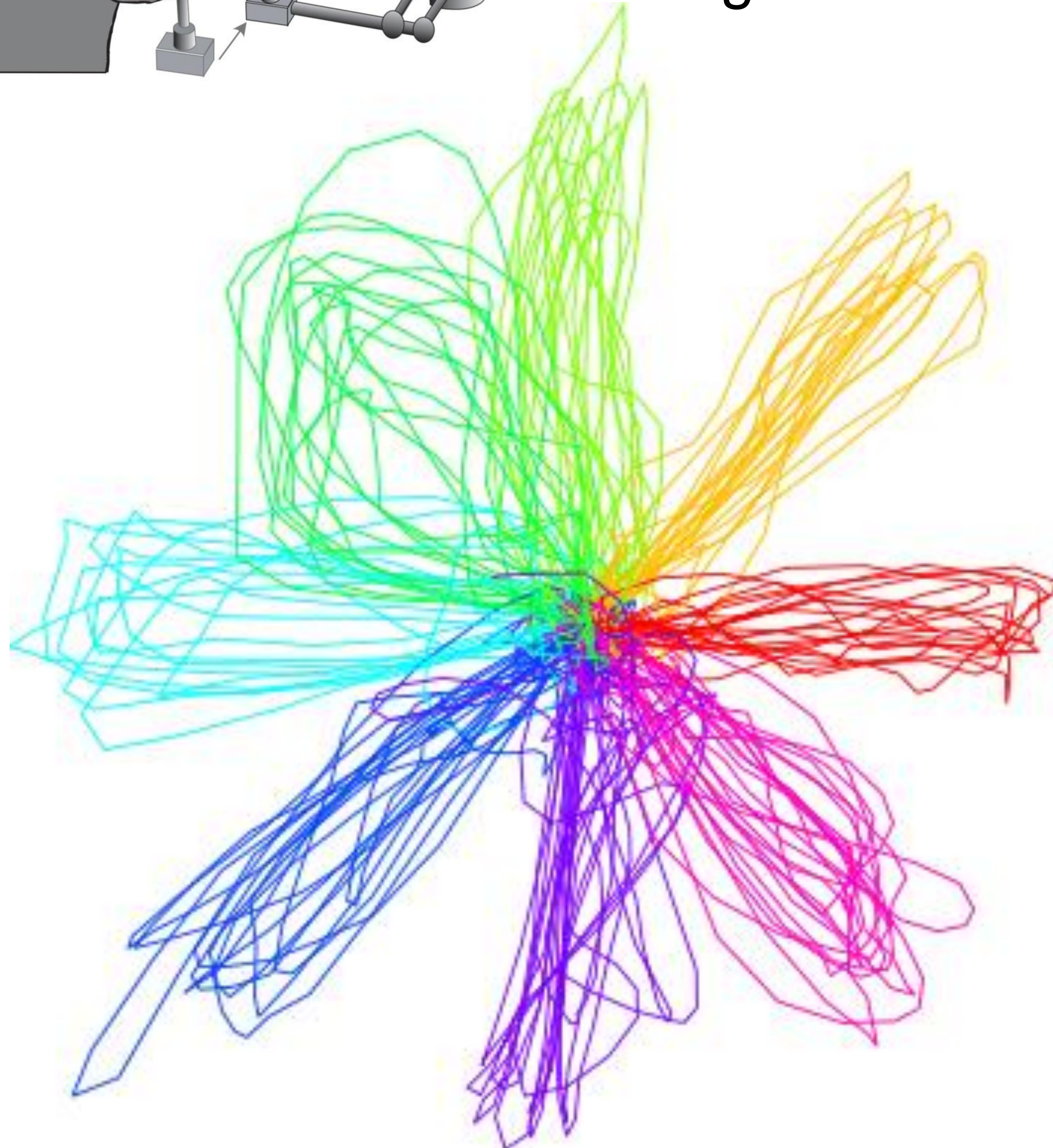
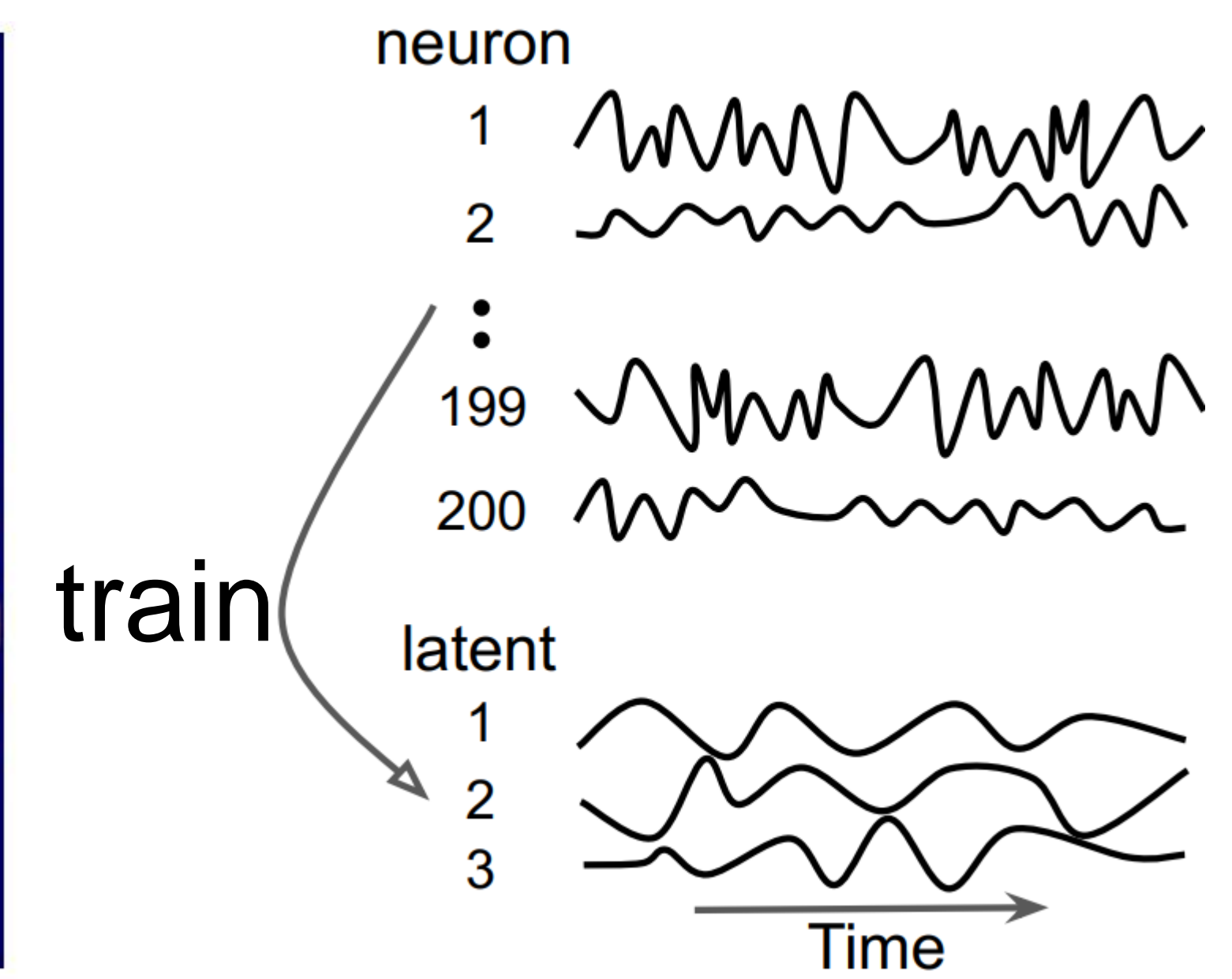
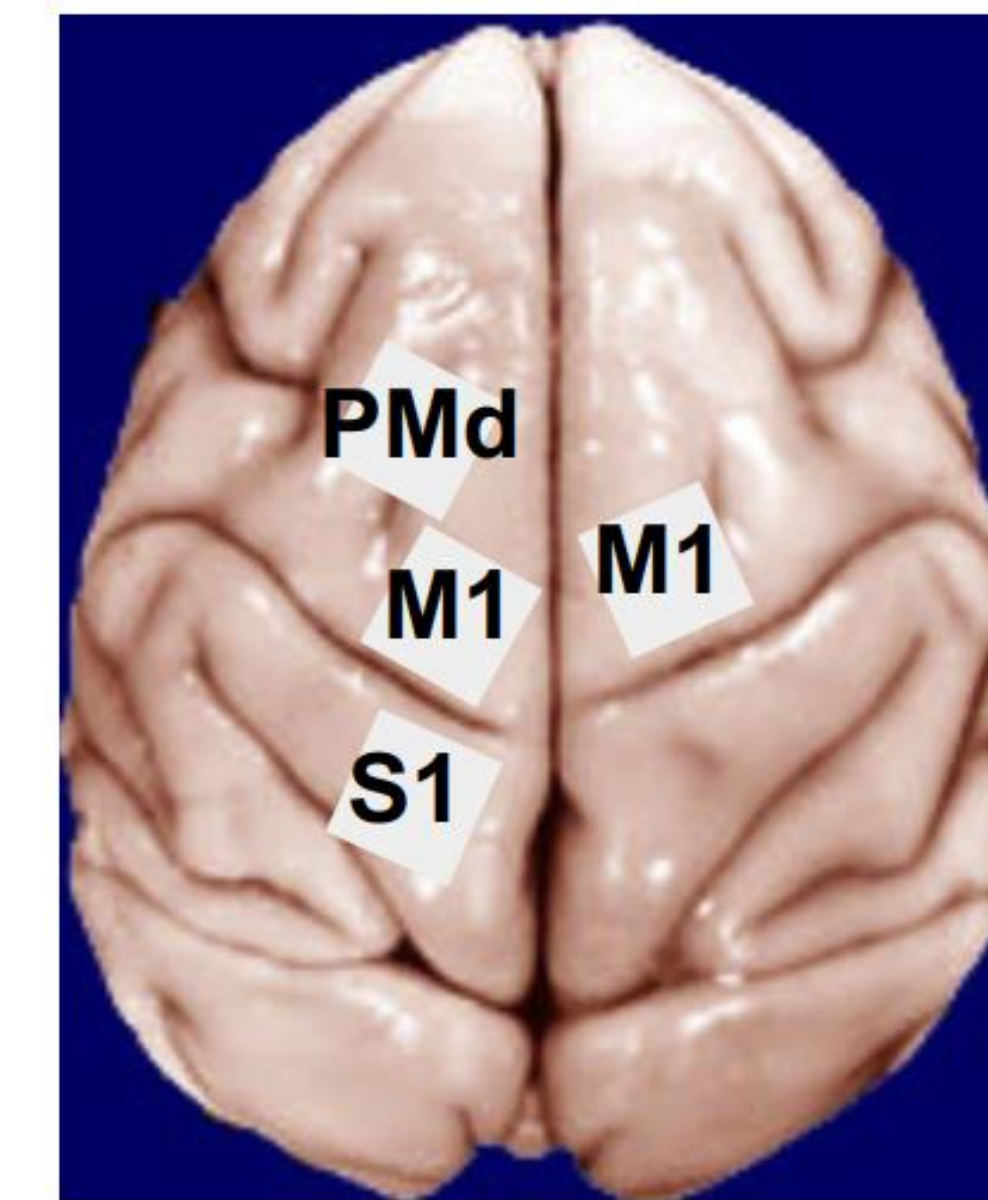
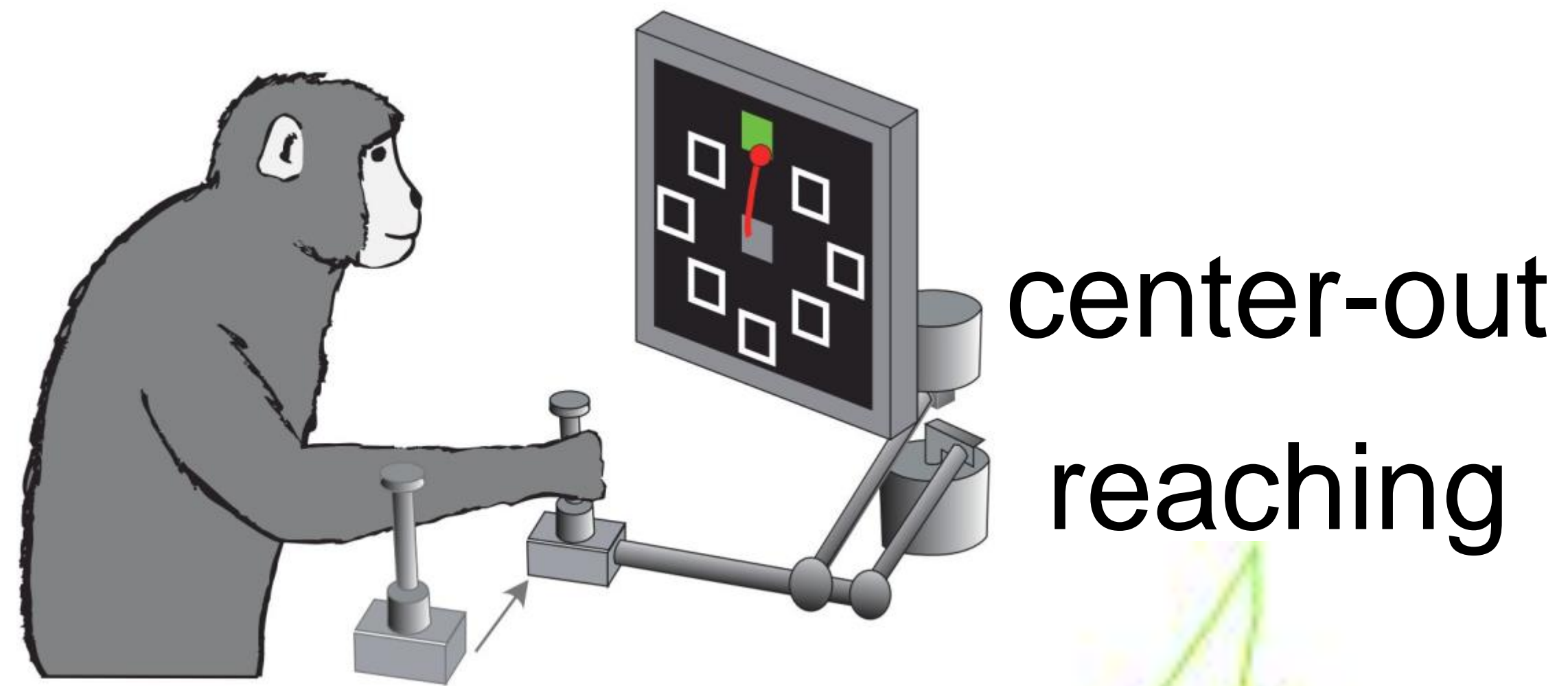


JOHNS HOPKINS
BIOMEDICAL ENGINEERING



NEURAL INFORMATION
PROCESSING SYSTEMS

Movements-aligned 3D latent dynamics



State-of-the-art models are generative and contrastive self-supervised learning (SSL)

pi-VAE: Poisson identifiable variational auto-encoder

Learning identifiable and interpretable latent models of high-dimensional neural activity using pi-VAE *Ding Zhou, Xue-Xin Wei*

Part of [Advances in Neural Information Processing Systems 33 \(NeurIPS 2020\)](#)

CEBRA: Consistent **EmB**eddings of high-dimensional Recordings using **Auxiliary** variables

Learnable latent embeddings for joint behavioural and neural analysis

[Steffen Schneider](#), [Jin Hwa Lee](#) & [Mackenzie Weygandt Mathis](#) 

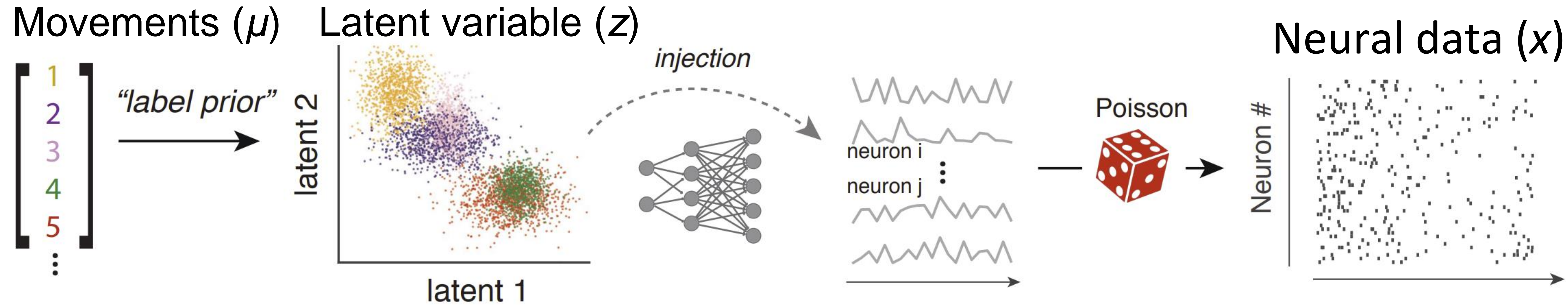
[Nature](#) **617**, 360–368 (2023) | [Cite this article](#)

143k Accesses | **1168** Altmetric | [Metrics](#)

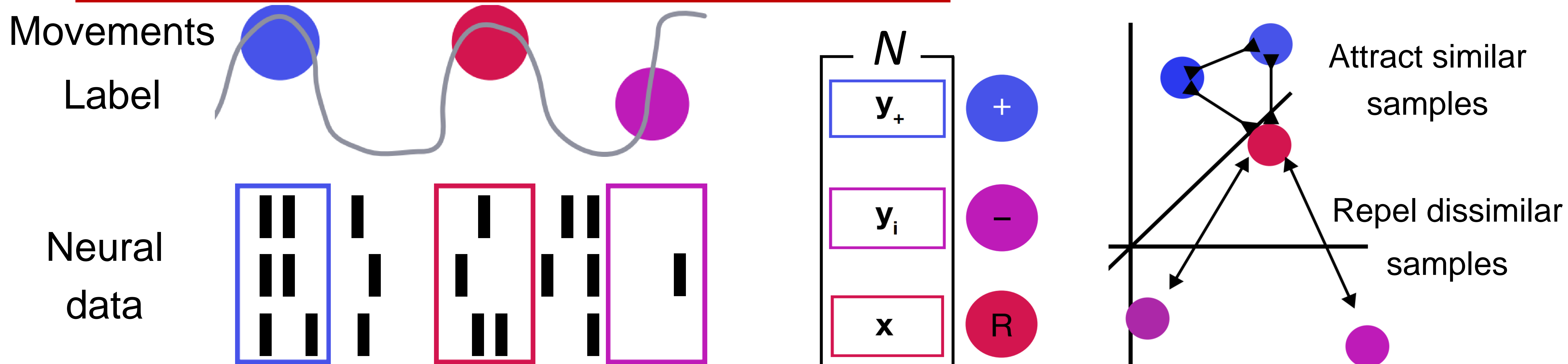
pi-VAE and CEBRA treat movements as distinct classes

$$loss = -\log p(x|z) + \text{KL}(q(z|x, u) || p(z|u))$$

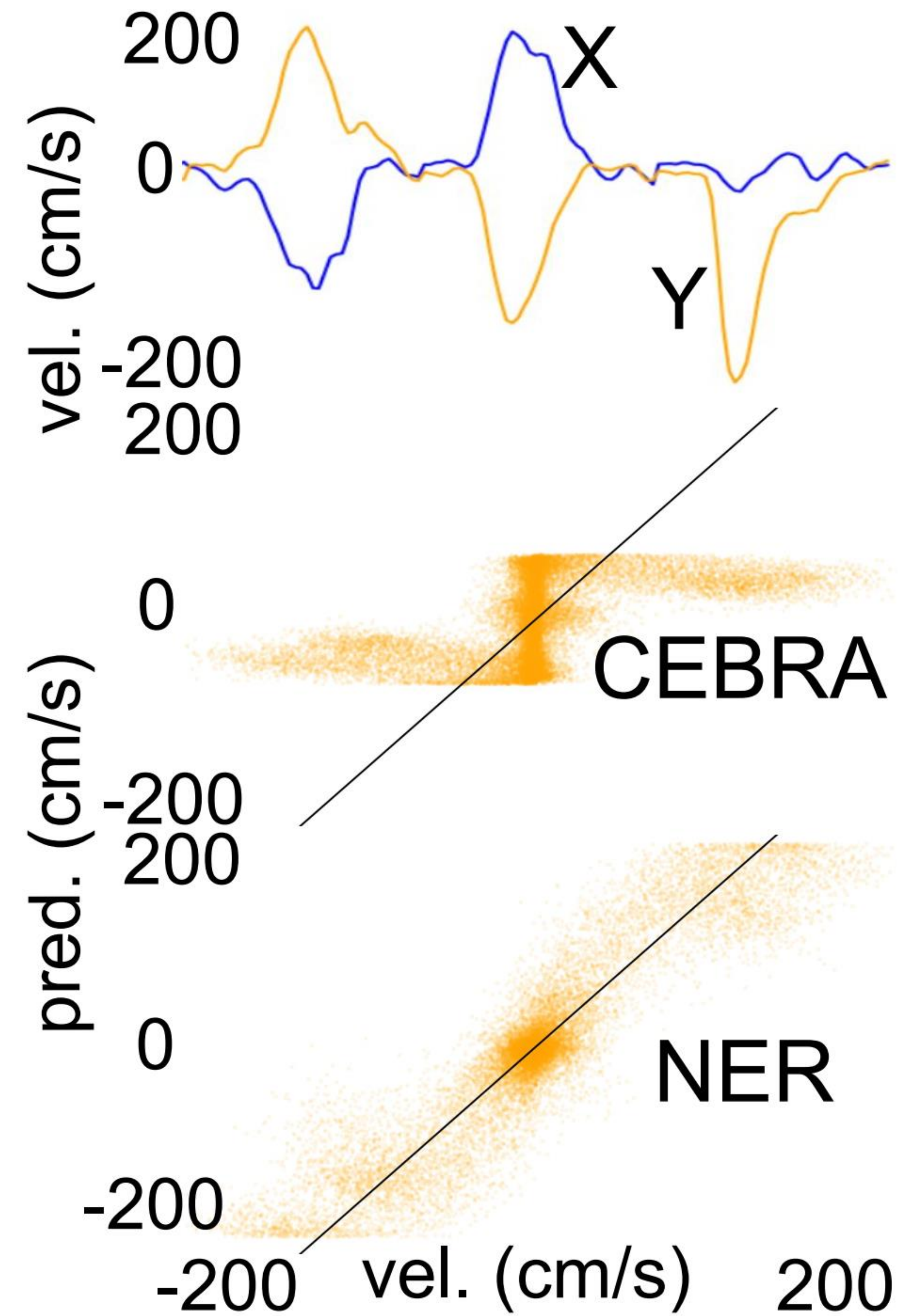
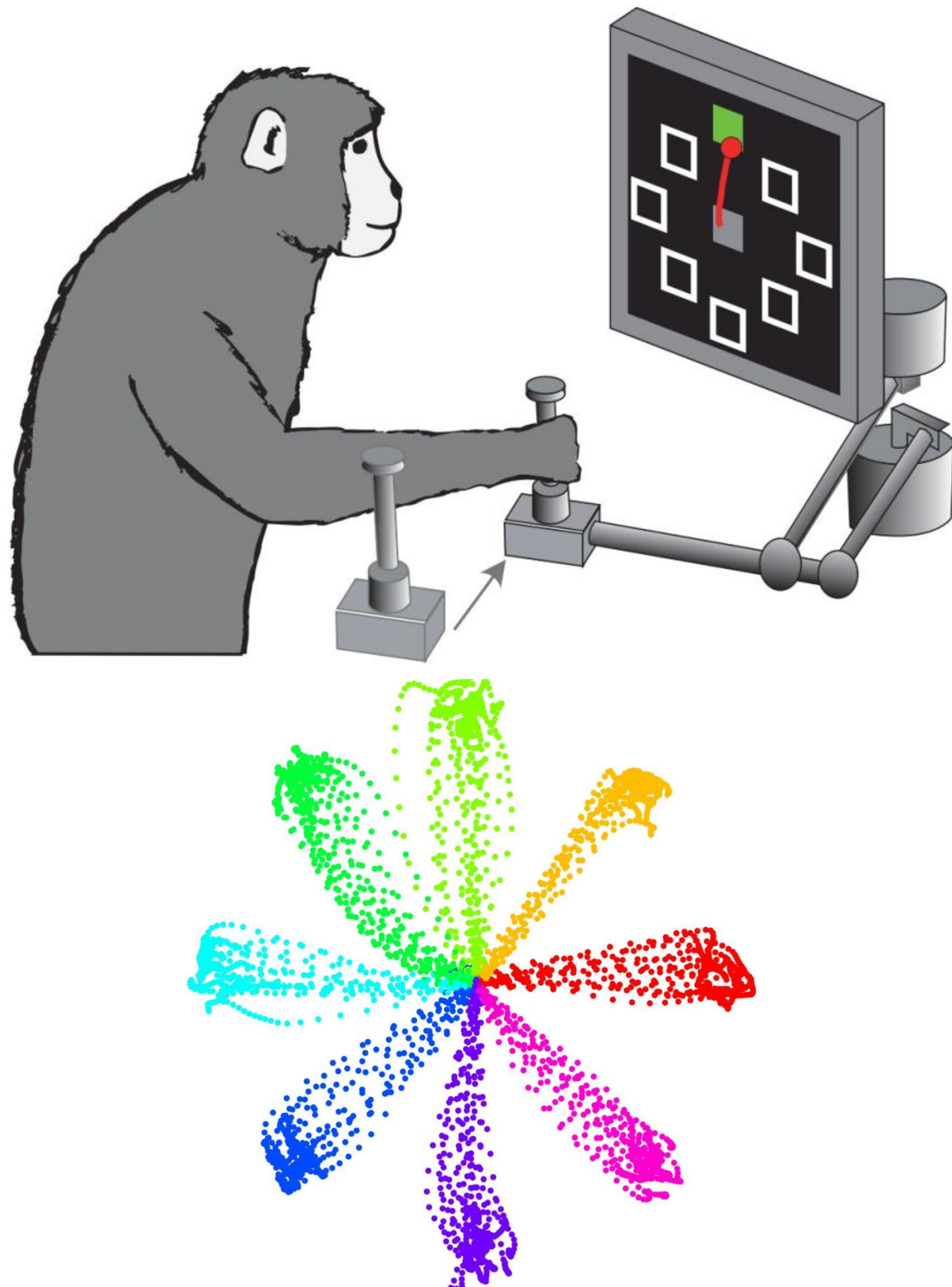
$$q(z|x, u) \propto q(z|x)p(z|u)$$



$$loss = -sim(x, y_+) + \log \sum_{i=1}^N e^{sim(x, y_i)}$$




But movements are continuous and highly imbalanced



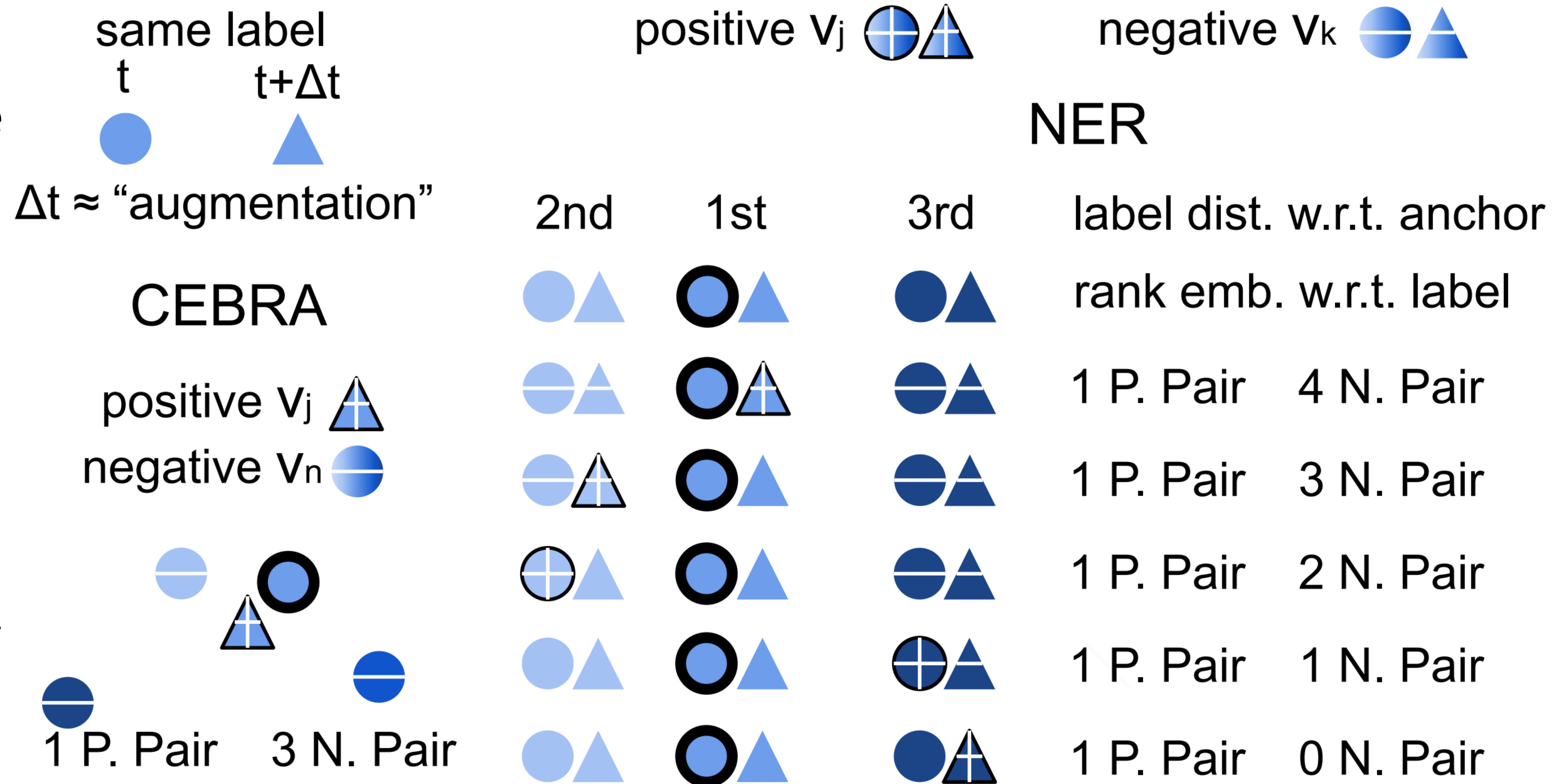
Loss functions of CEBRA and NER

$$l_{CEBRA}^{(i)} = -\log \frac{\exp(\text{sim}(v_i, v_j)/\tau)}{\sum_{n=1}^N \exp(\text{sim}(v_i, v_n)/\tau)}$$

$$l_{NER}^{(i)} = \frac{1}{2N-1} \sum_{j=1, j \neq i}^{2N} -\log \frac{\exp(\text{sim}(v_i, v_j)/\tau)}{\sum_{v_k \in S_{i,j}} \exp(\text{sim}(v_i, v_k)/\tau)}$$

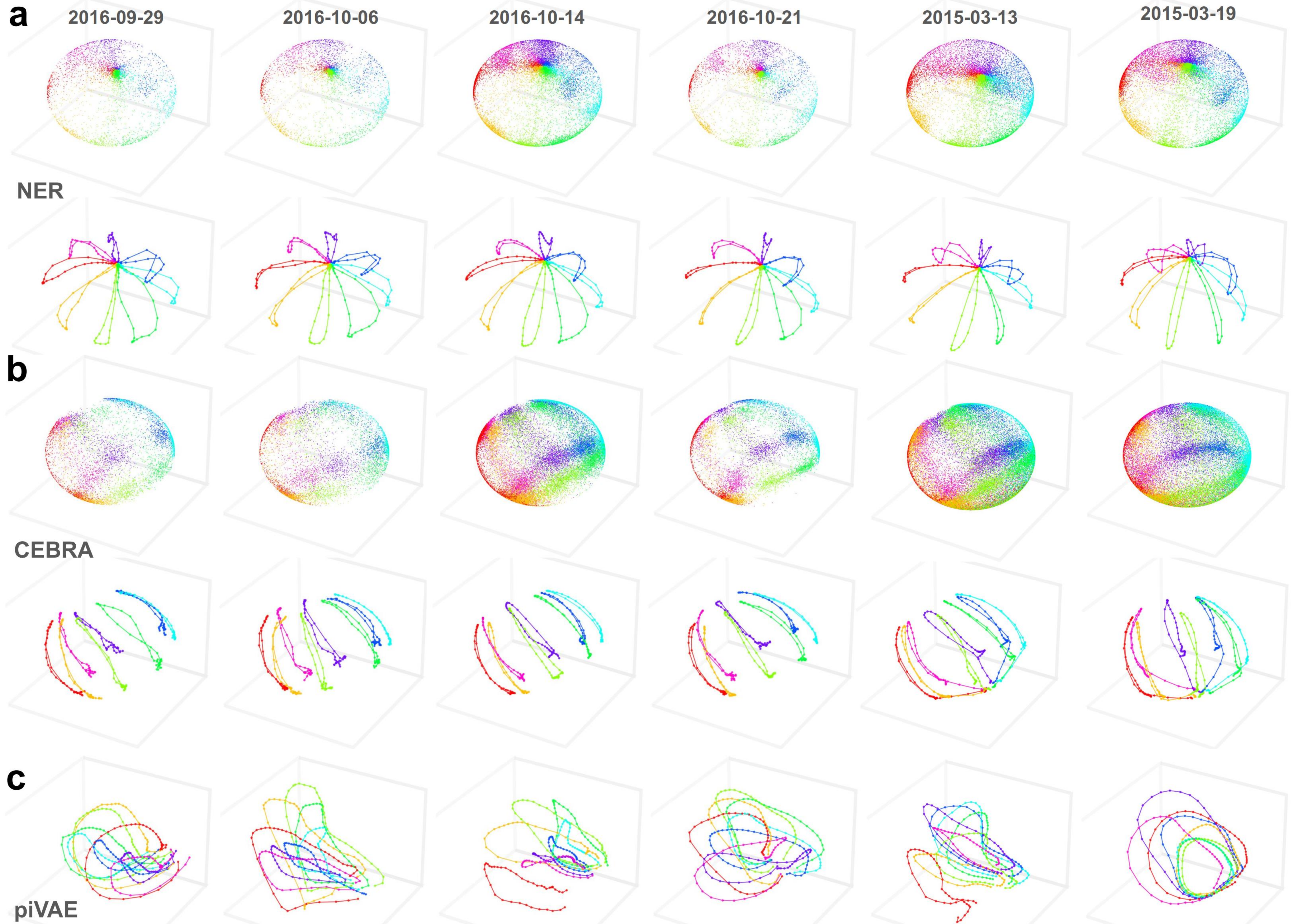
$S_{i,j} := \{v_k \mid k \neq i, d(y_i, y_k) \geq d(y_i, y_j)\}$ anchor v_i  continuous label

Intuitively, for an anchor v_i , any other embedding v_j in the batch is positively contrasted with it, enforcing the similarity between v_i and v_j to be greater than that between v_i and any other embedding v_k in the batch if the label distance between y_i and y_k is larger than that of y_i and y_j .

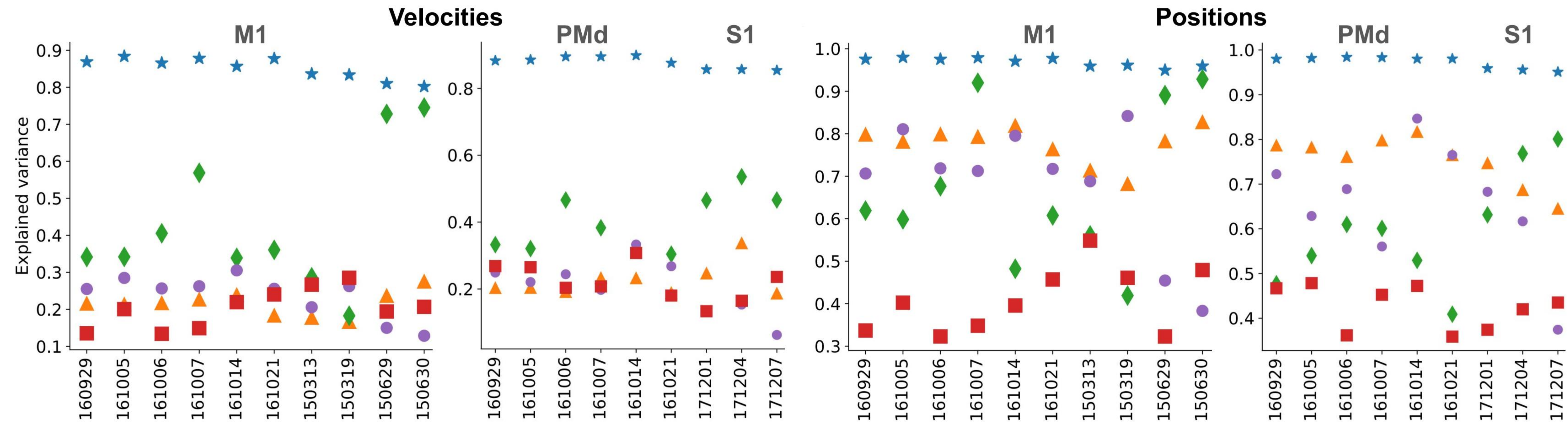


M1 in Left Hemisphere

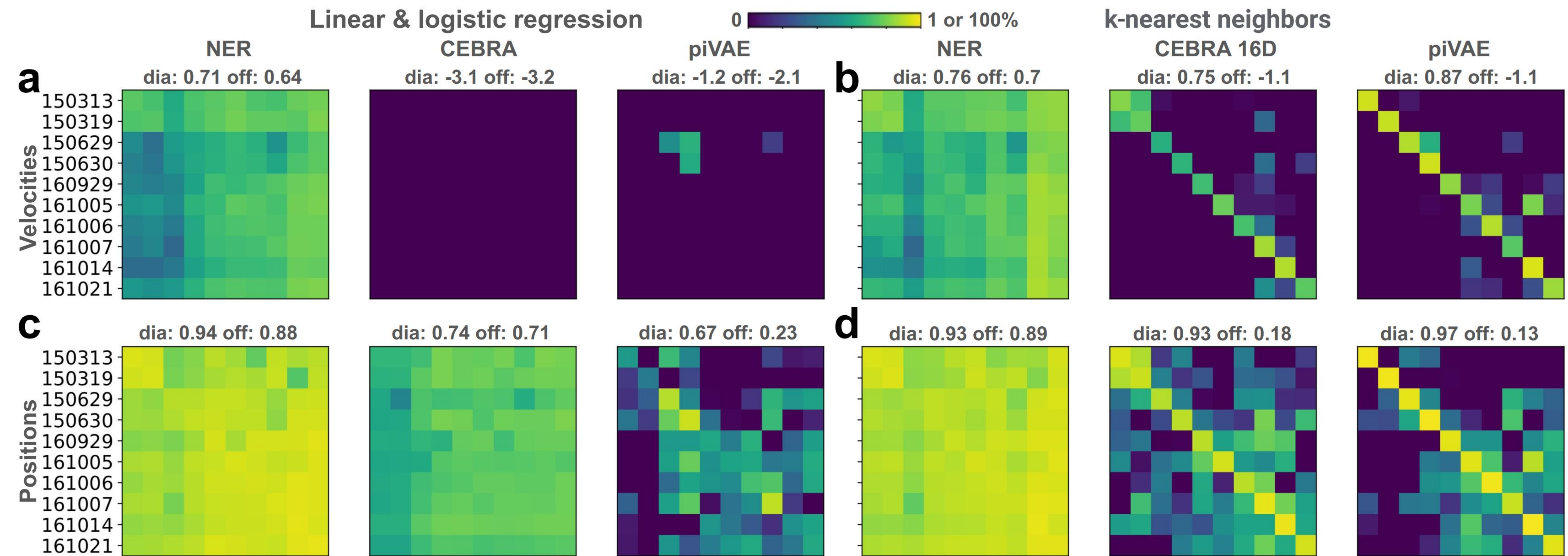
M1 in Right Hemisphere



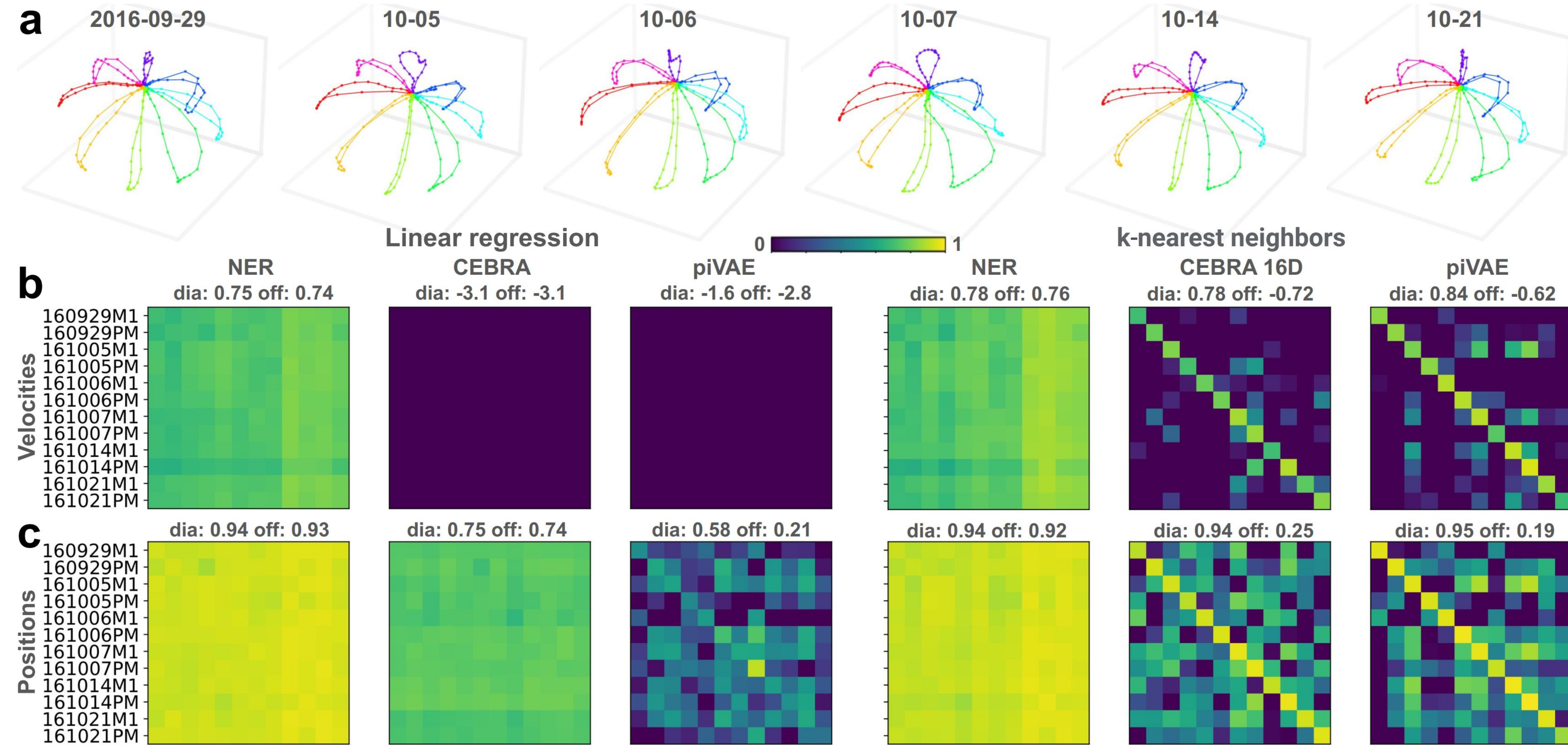
NER explains the largest variance in hand velocities and positions in M1, PMd and S1



Long-term and cross-hemisphere decoding in M1



Latent dynamics in PMd & decoding between M1 & PMd



Code, data, and figures are available

<https://github.com/NeuroscienceAI/NER>

NeuroscienceAI / NER Public

[Code](#) [Issues](#) [Pull requests](#) [Actions](#) [Projects](#) [Security](#) [Insights](#)

main 1 Branch Tags [Code](#)

NeuroscienceAI	Update README.md	7c3e528 · 2 days ago	45 Commits
NER_Code_1021	Add NER code to NER_Code_1021 folder		last week
NER_Figs_ipynb	Add Jupyter Notebook to NER_Figs_ipynb folder		2 weeks ago
NER_Figs_pdf	Add files via upload		2 days ago
data	Add raw data to data folder		last week
data_NER	Add NER processed data to data_NER folder		last week
README.md	Update README.md		2 days ago

- Fig1_Method_CEBRA_Lin_Reg_1021.ipynb
- Fig1_Method_NER_Lin_Reg_1021.ipynb
- Fig2_Example_emb.ipynb
- Fig3_Batch_M1_NER_CEBRA.ipynb
- Fig3_Batch_M1_UMAP_PCA_dPCA.ipynb
- Fig3_Batch_M1_piVAE.ipynb
- Fig3_CEBRA_embeddings_rotate.ipynb
- Fig3_NER_embeddings_rotate.ipynb
- Fig3_PCA_embeddings.ipynb
- Fig3_UMAP_embeddings_rotate.ipynb
- Fig3_UMAP_label_embeddings_rotate.ipynb
- Fig3_dPCA_embeddings.ipynb
- Fig3_piVAE_embeddings_rotate.ipynb
- Fig4_Batch_explained_var_kNN.ipynb
- Fig4_Batch_explained_var_linear_reg.ipynb
- Fig4_M1_explained_var.ipynb
- Fig4_PMd+S1_explained_var.ipynb
- Fig5_Batch_M1_self+cross_decode_kNN_1022.ipynb
- Fig5_Batch_M1_self+cross_decode_linear_1022.ipynb
- Fig5_M1_cross_decode.ipynb
- Fig6_Batch_PMd_NER_CEBRA.ipynb
- Fig6_Batch_PMd_piVAE.ipynb
- Fig6_Batch_self+cross_decode_kNN_1023.ipynb
- Fig6_Batch_self+cross_decode_linear_1023.ipynb