Bayesian Clustering of Neural Spiking Activity Using a Mixture of Dynamic Poisson Factor Analyzers

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How to Define Neural Populations?

Brain Regions (Anatomy)



- Neurons in different groups may:
 - Interact with each other,
 - Receive common input,
 - Share the same latent structure.
- Using only anatomy or cell type may not accurately describe the data.





Can we define neural populations by activity?



Goal: Cluster neurons according to activity

- High Level Idea:
 - Mixture of latent structure models

→ Mixture of dynamic Poisson factor analyzers (mixDPFA)

- Advantages:
 - Simultaneously extract latent structure underling the activity of populations,
 - Defined populations are related to function directly,
 - Flexibly describe neural activity when the input is nonhomogeneous.

mixDPFA Bayesian inference via a MCMC algorithm



Simulation Results



Other Details in Paper

- Constraints on the model...
 - Make the model identifiable,
 - Cluster by latent trajectories, instead of baseline amplitude etc.
- Sample posteriors for latent factors...
 - Exactly by Pólya-Gamma (PG) augmentation with a Metropolis-Hasting step,
 - Compare to samples from widely used Laplace approximation.
- Fitting to large-scale Neuropixels data...
 - Bayesian clustering results are

reliable, differ from anatomical regions, and depend on stimuli.



- Cluster neurons according to activity
 - Mixture of dynamic Poisson factor analyzers (mixDPFA)
- Advantages:
 - Simultaneously extract latent structure underling the activity of populations,
 - Flexibly describe neural activity when the input is nonhomogeneous.

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