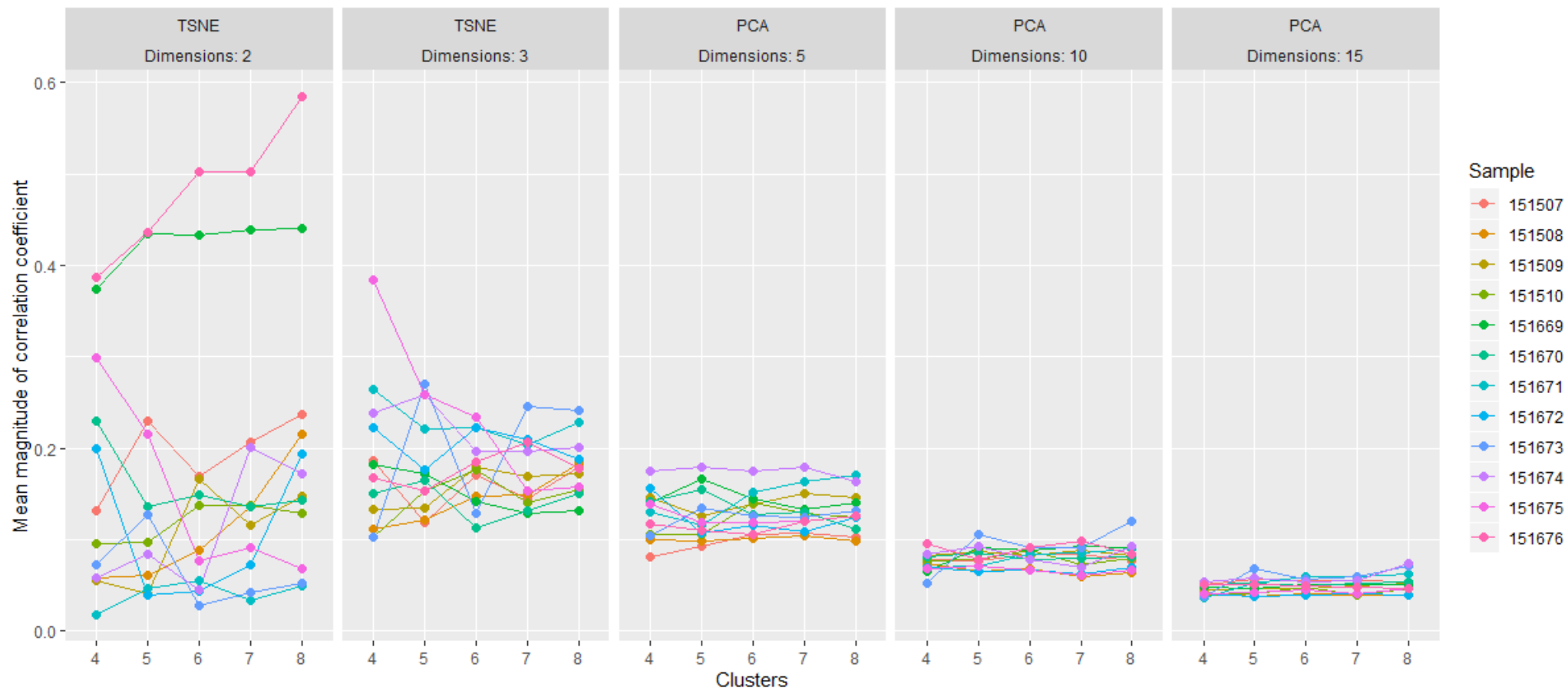


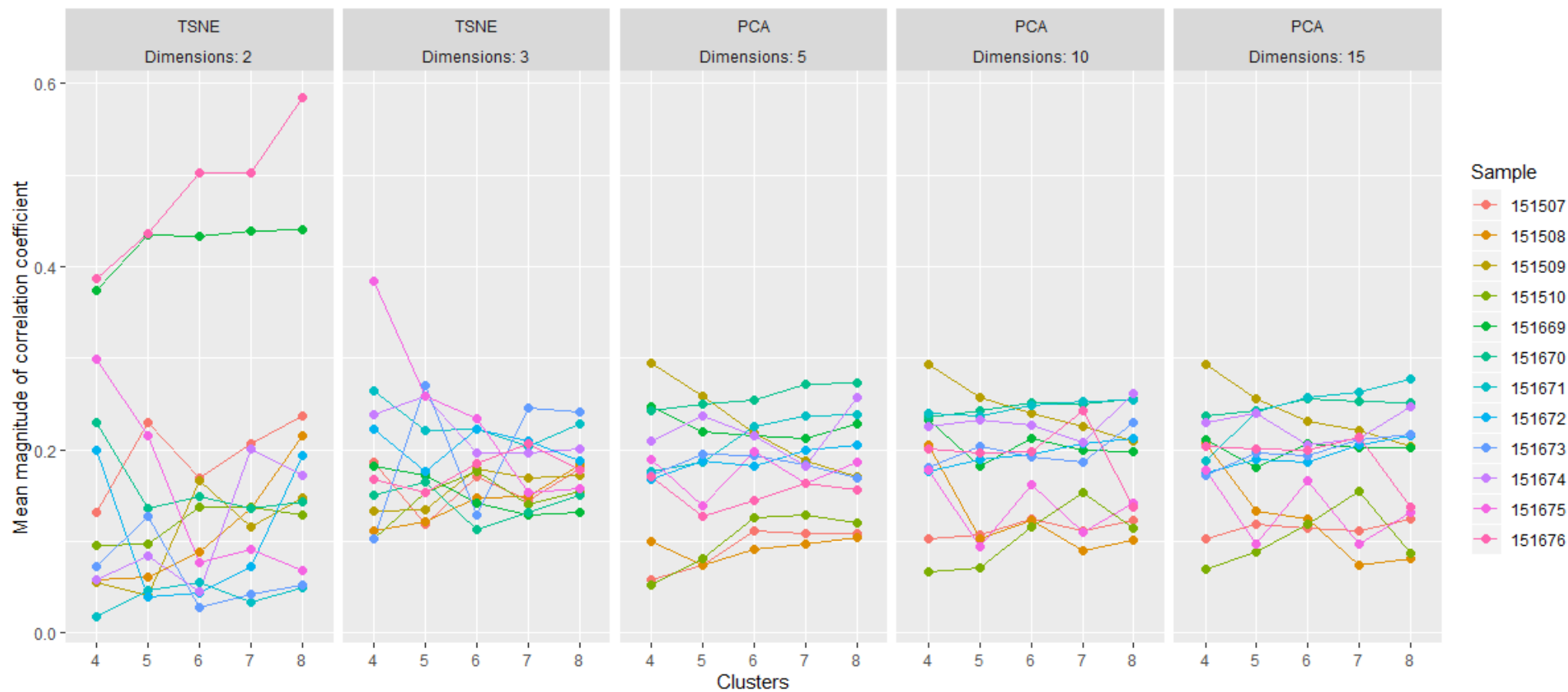
Analysis of Maynard et al., 2020 brain spatial transcriptomic data

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TSNE vs. PCA



TSNE vs. top 3 PCs



Ising Likelihood and pseudolikelihood

- Likelihood:

- $\mathcal{L}(\gamma; \mathbf{z}) = P(\mathbf{Z} = \mathbf{z}) = \frac{\exp(\gamma \sum_{\langle i,j \rangle} z_i z_j)}{\sum_{\mathbf{z}} \exp(\gamma \sum_{\langle i,j \rangle} z_i z_j)}$

- Conditional likelihood for spot i :

- $\mathcal{L}_i(\gamma; \mathbf{z}) = P(Z_i | \mathbf{Z}^{(-i)} = \mathbf{z}^{(-i)}) = \frac{P(\mathbf{Z} = \mathbf{z})}{P(\mathbf{Z}^{(-i)} = \mathbf{z}^{(-i)})} = \frac{P(\mathbf{Z} = \mathbf{z})}{\sum_{z_i} P(Z_i = z_i, \mathbf{Z}^{(-i)} = \mathbf{z}^{(-i)})} =$
 $= \frac{\exp(\gamma \sum_{\langle i,j \rangle} z_i z_j)}{\sum_{z_i} \exp(\gamma \sum_{\langle i,j \rangle} z_i z_j)}$

- Pseudolikelihood:

- $\mathcal{PL}(\gamma; \mathbf{z}) = \sum_{i=1}^n \mathcal{L}_i(\gamma; \mathbf{z})$

Maximum a posterior probability

- Following the pseudolikelihood framework, we want to find the z_i that will maximize the posterior probability for each spot:

- $$p(z_i|y_i) = \frac{p(y_i|z_i)\pi(z_i)}{p(y_i)} = \frac{p(y_i|z_i)\pi(z_i)}{\sum_{z_i} p(y_i|z_i)\pi(z_i)}$$

- If we fix the domain of z_i , the denominator does not matter
 - So we need the denominator if we want to optimize cluster number, right?

Initialization issues

- Start with weak field (Besag 1986)
 - Start with $\gamma = 0$, gradually increasing to final desired γ in successive iterations
- Estimate γ at each iteration (Besag 1986)
 - Maximize $\prod_i p_i(x_i | x^{(-i)}, \gamma)$

Next steps

- Speed up code
- Tune γ for deconvolution
- Normality of t-SNE