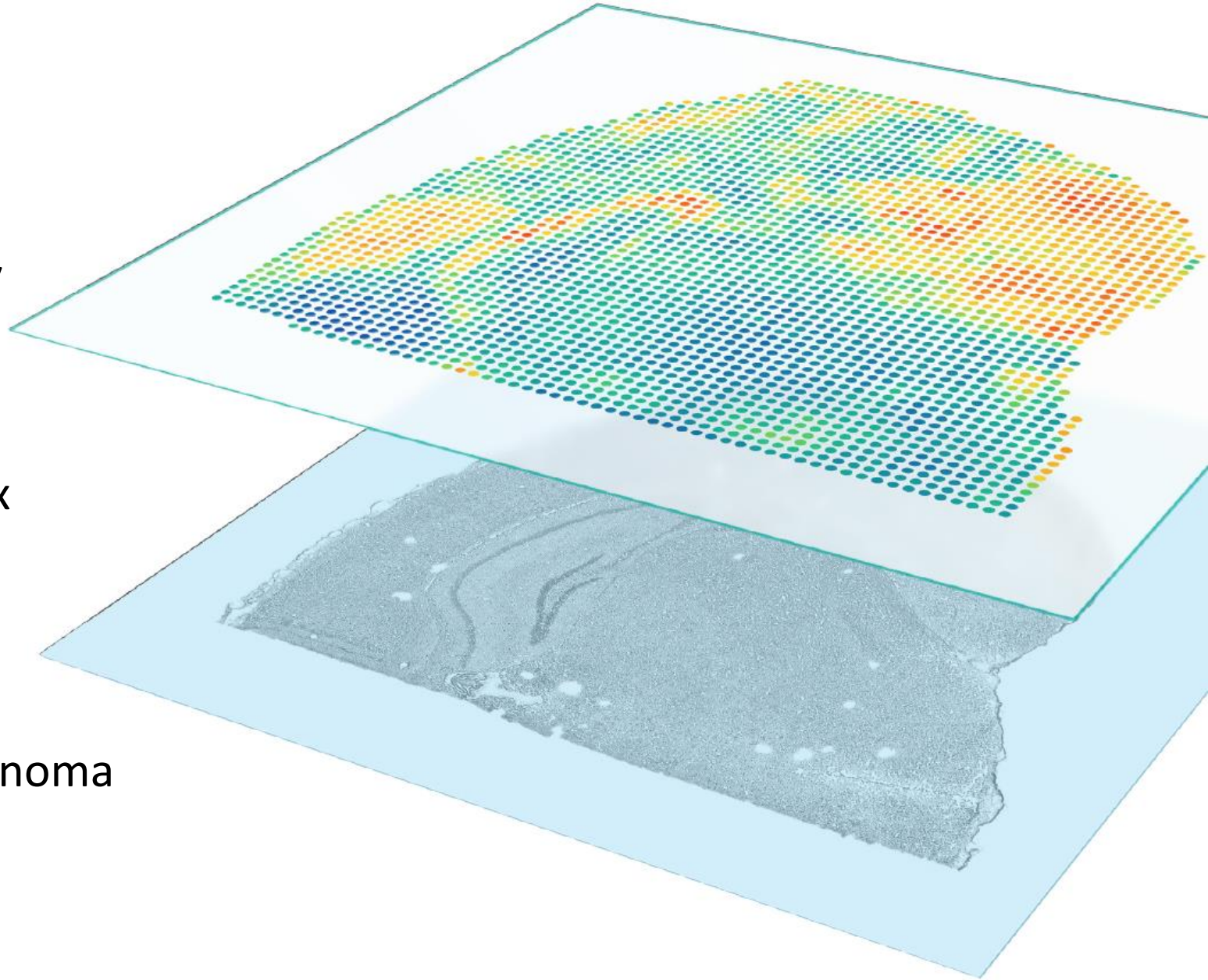


Spatial Clustering and Deconvolution

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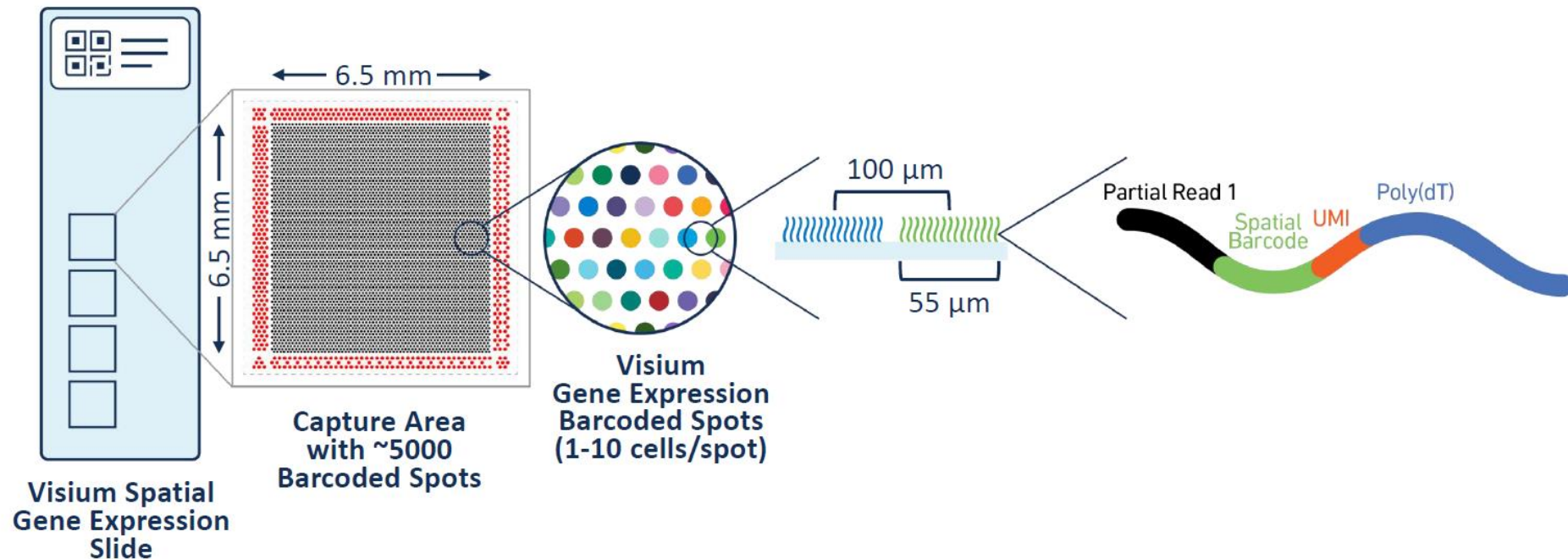
Outline

- Visium spatial technology
- Spatial clustering
 - Method
 - Example: Prefrontal cortex
- Spatial deconvolution
 - Method
 - Example: Melanoma
 - Example: Merkel cell carcinoma



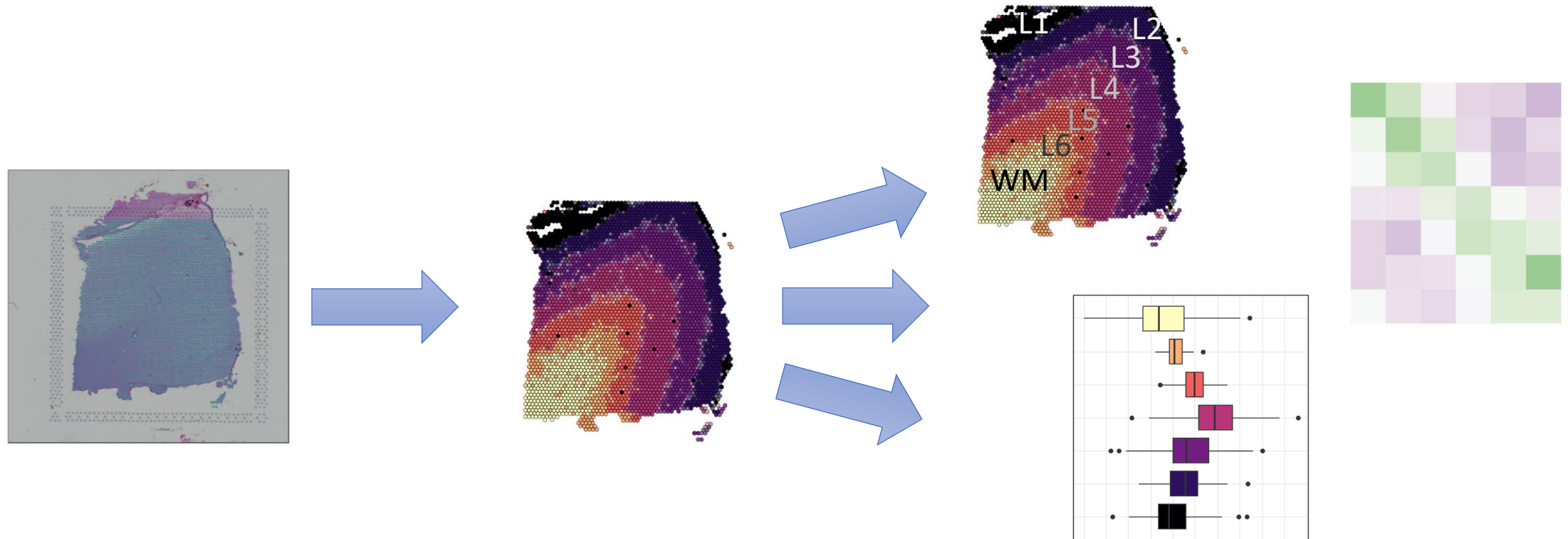
Visium spatial technology

- Spatial RNA sequencing provides high-throughput gene expression profiling while preserving the morphological context of the tissue



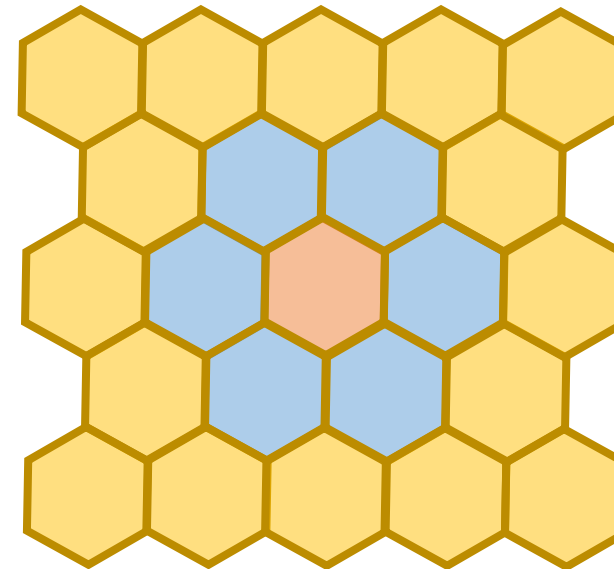
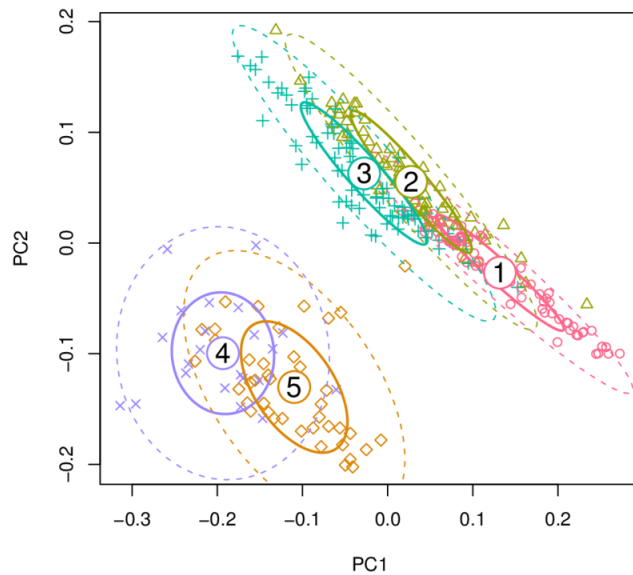
Spatial clustering

- Accurate clustering is an important step that allows downstream analyses such as cell type or tissue annotation and differential expression to provide unbiased biological insights



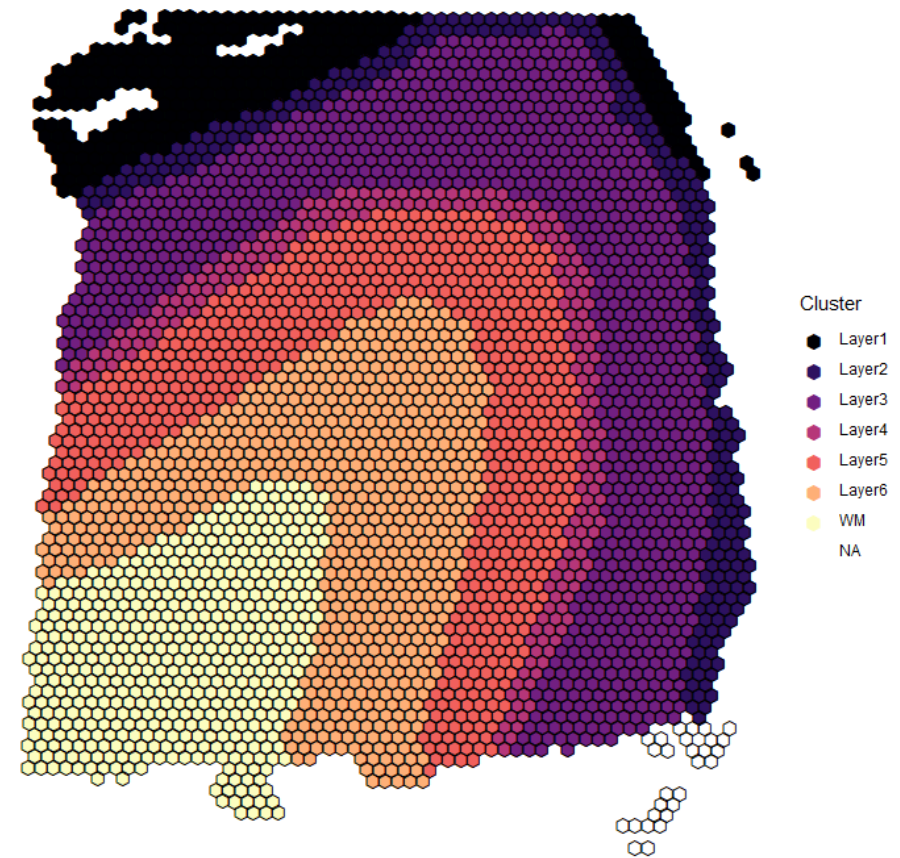
Method

- Fit a parametric model to the observed data (principal components)
 - Using MCMC, iteratively estimate the center of each cluster, variance within clusters, and cluster label of each spot
 - Spatially smooth cluster labels by encouraging neighboring spots to belong to the same cluster



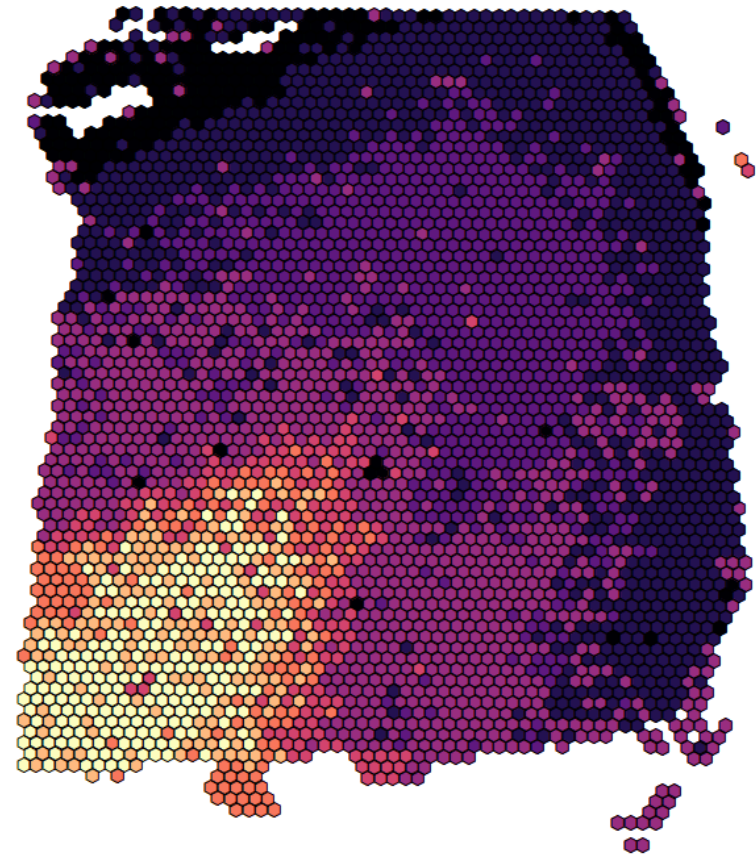
Dorsolateral prefrontal cortex (Maynard et al., 2020)

- Manual annotation by expert based on cytoarchitecture and selected marker genes
 - Considered ground truth
 - Labor-intensive



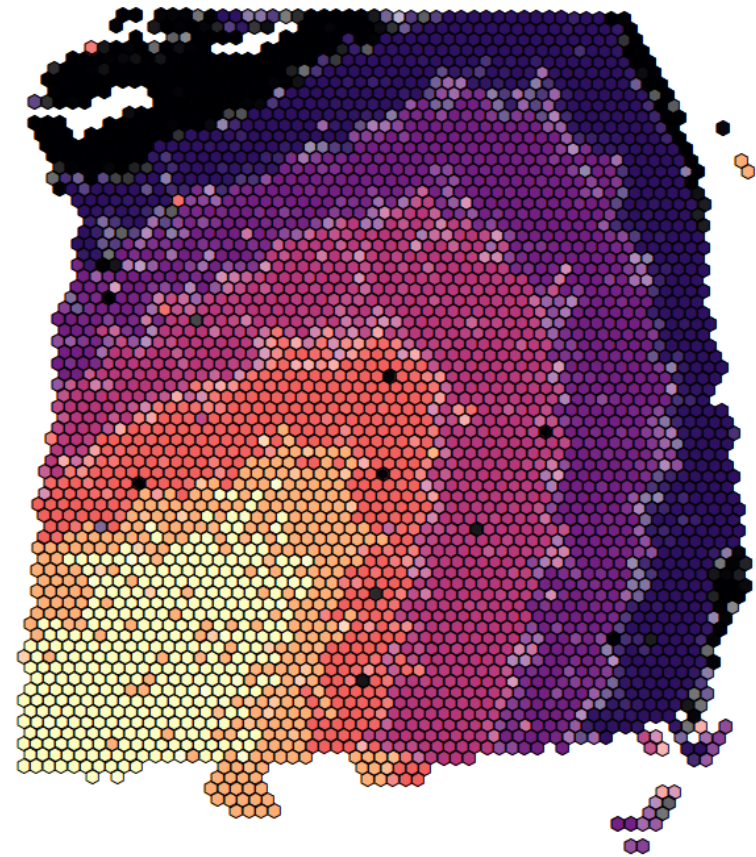
Non-spatial clustering

- Shared nearest neighbor clustering on top 50 principal components (PCs) generated from highly variable genes (HVGs) + spatial coordinates
- Set to return 8 clusters to improve clustering performance
- Adjusted Rand Index (ARI) = 0.27



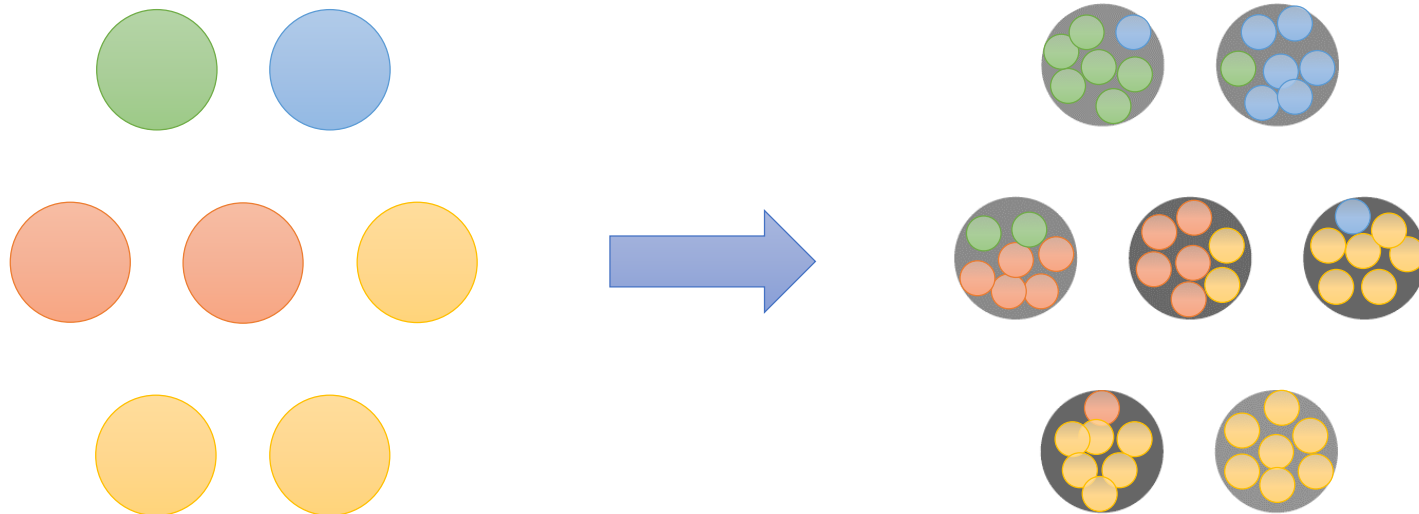
Spatial clustering

- Spatial clustering on top 9 principal components (PCs) generated from highly variable genes (HVGs)
- Set to return 7 clusters to match prior knowledge of prefrontal cortex cytoarchitecture
- ARI = 0.47



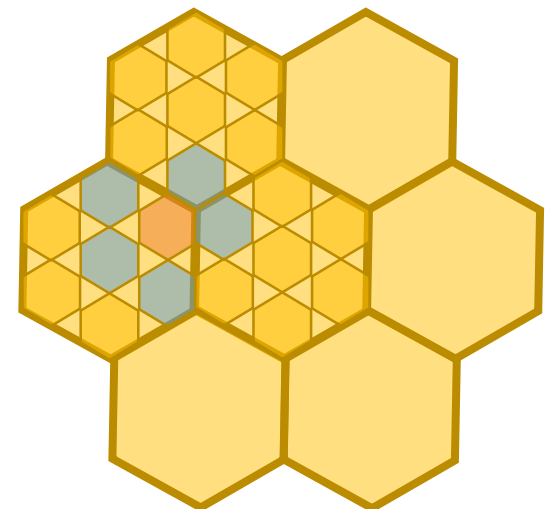
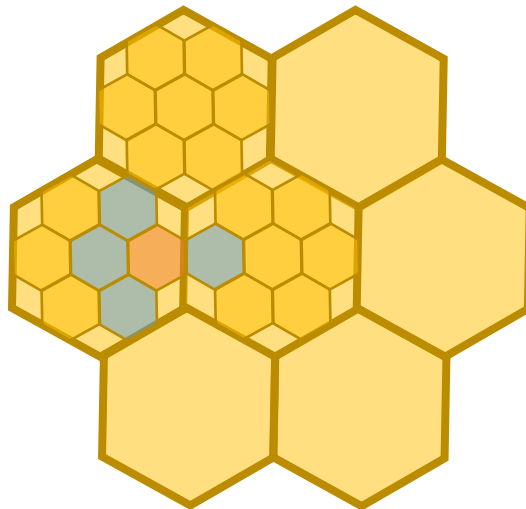
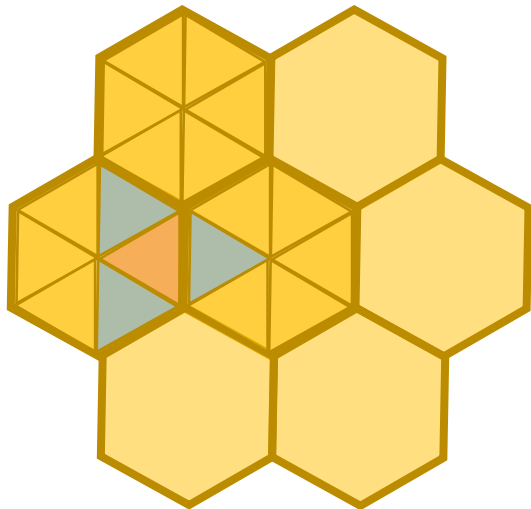
Spatial deconvolution

- Visium is not a single cell technology
- Deconvolution can:
 - Increase the resolution of the spatial map
 - Allow for better understanding of tissues composed of heterogenous mixtures of cells (e.g. tumor microenvironment)

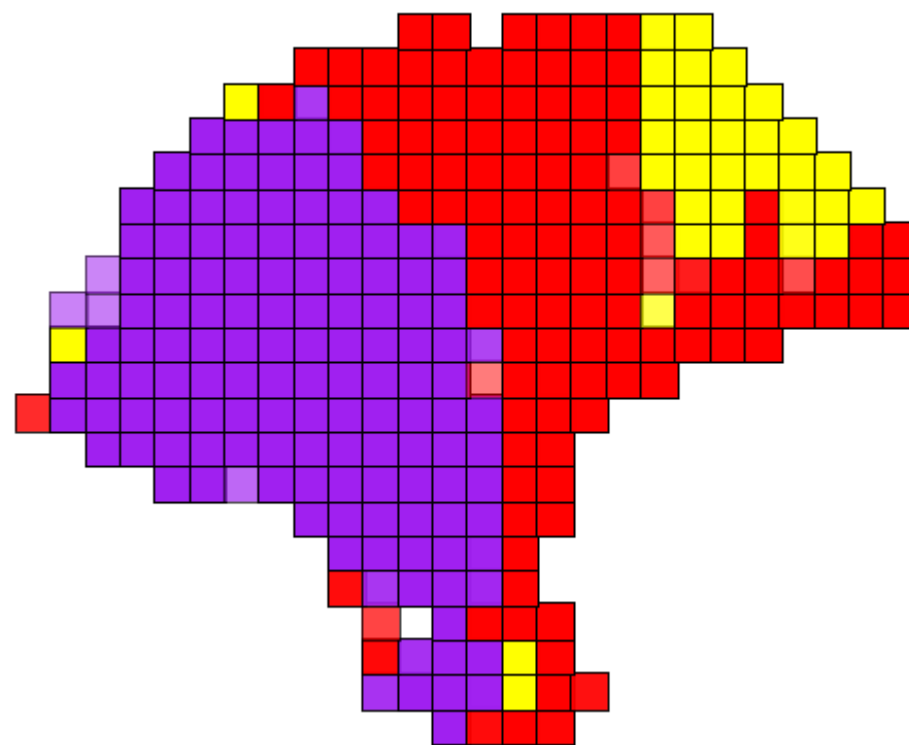
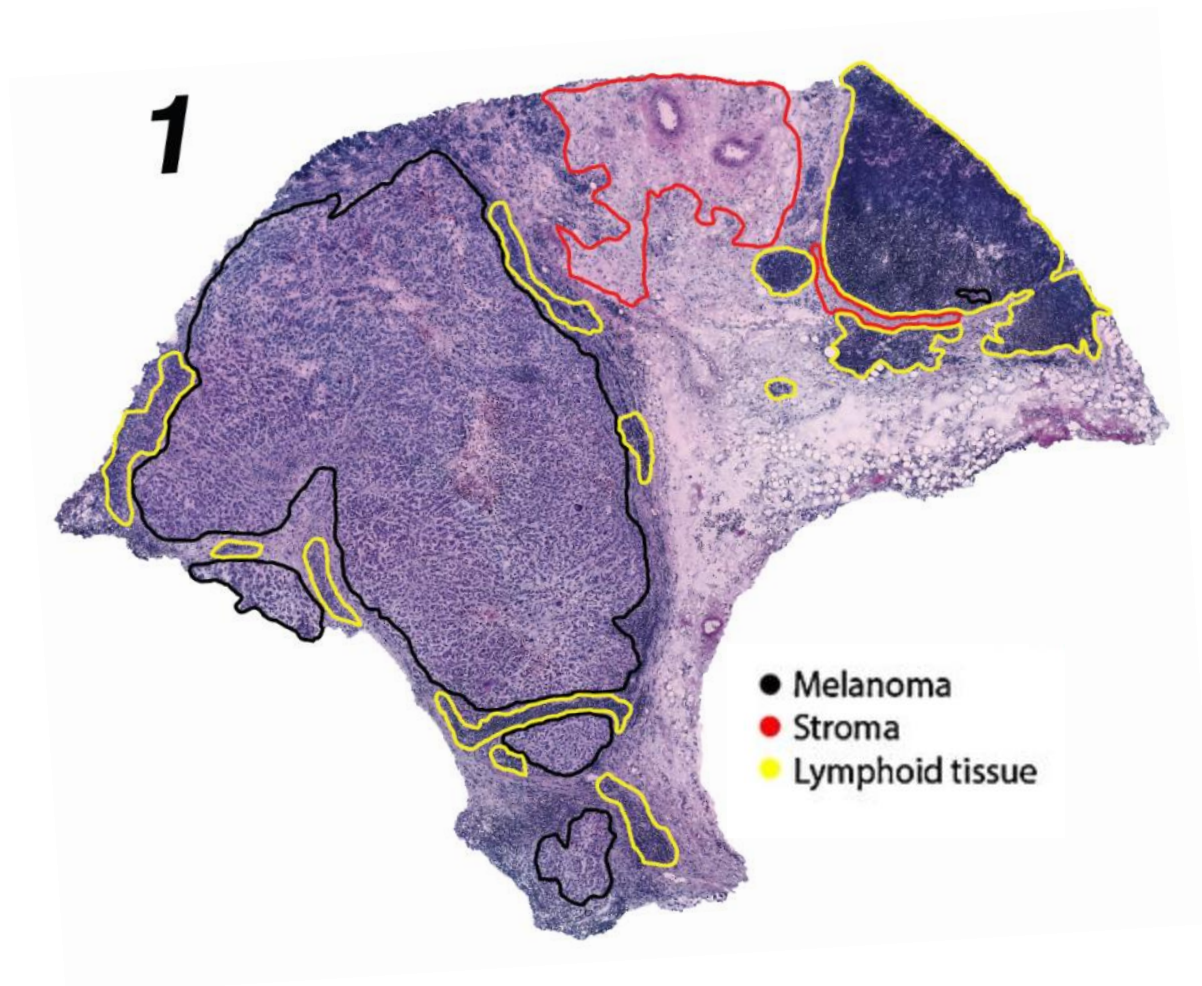


Method

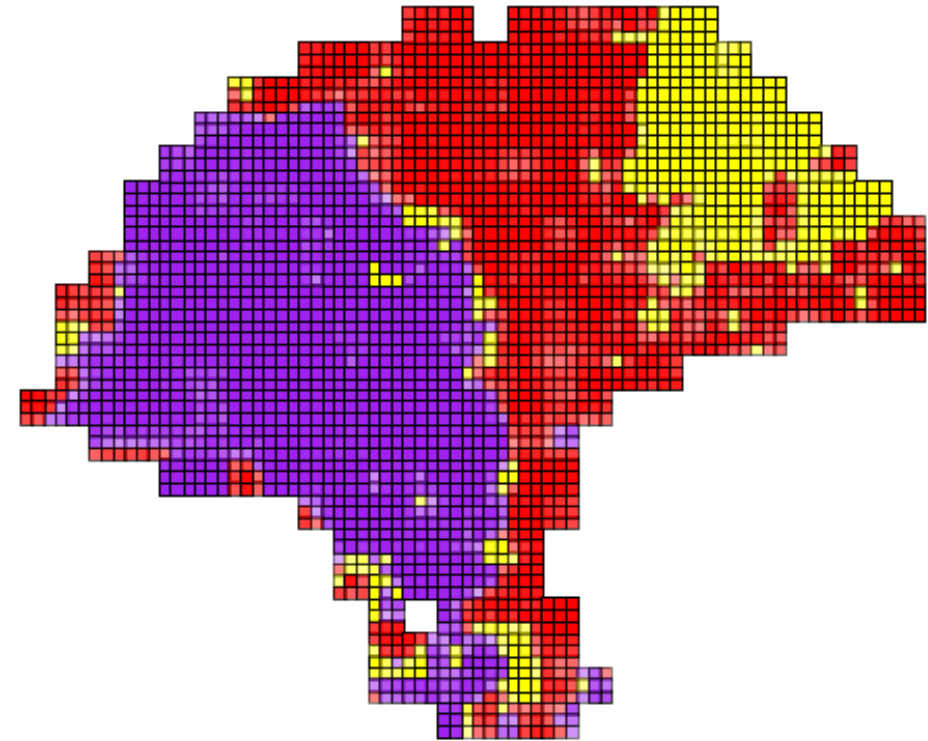
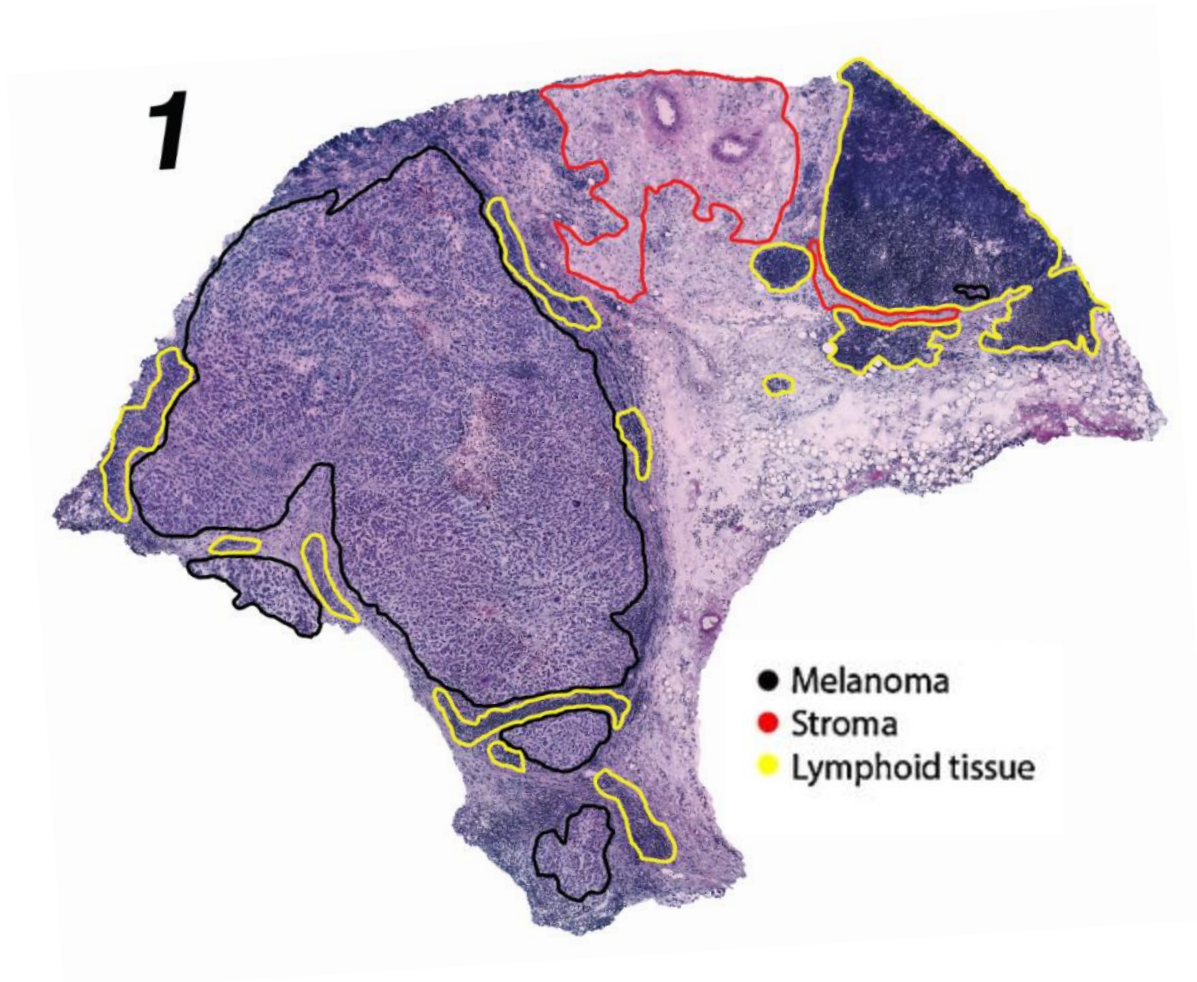
- Similar to clustering method, except now the unit of analysis is not a spot but instead a “cell”
 - Using MCMC, iteratively estimate the center of each cluster, variance within clusters, and cluster label of each “cell” while accounting for the spatial structure
 - In each iteration, jitter the features of each “cell” while keeping the sum over all “cells” within a spot fixed



Melanoma (Thrane et al., 2018)

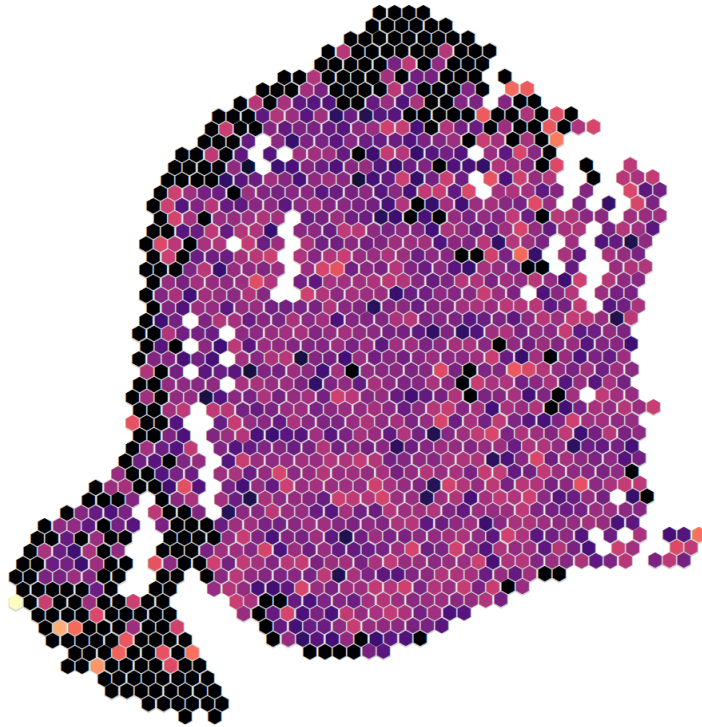


Deconvolution reveals heterogeneity in the tumor immune microenvironment

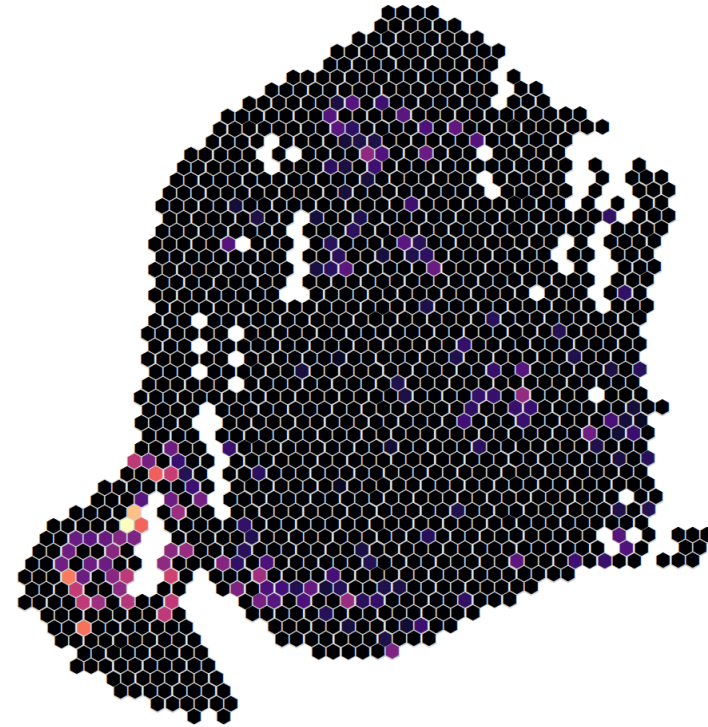


Merkle cell carcinoma (MCC)

CHGA (cancer marker) expression

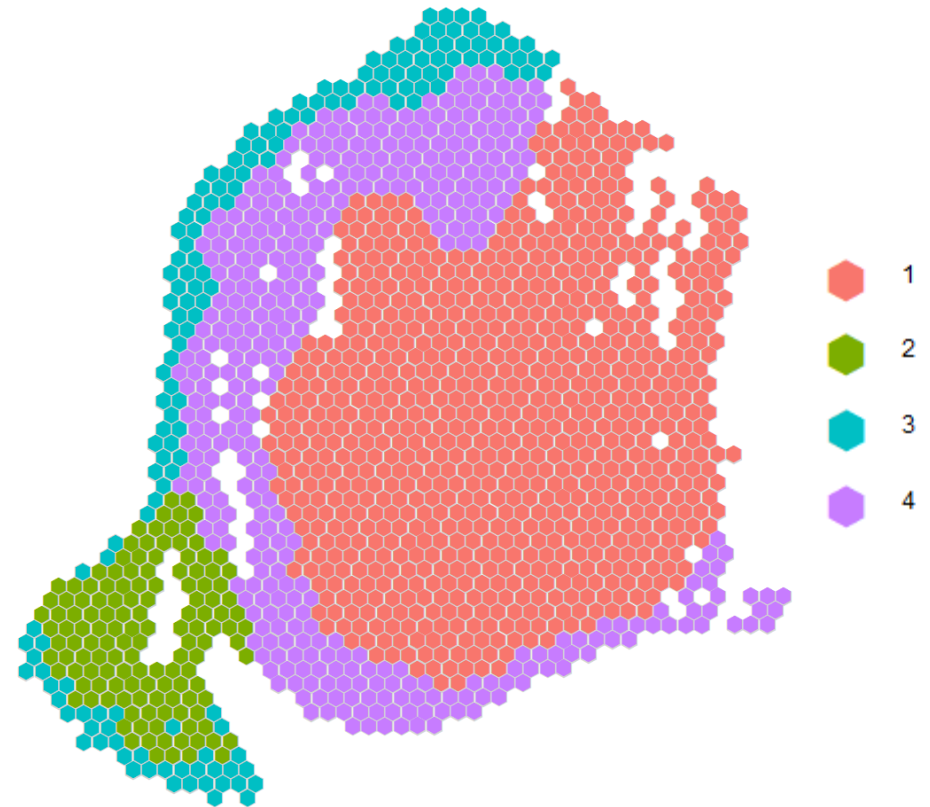


CD3 (*T cell marker*) expression



Spatial clustering

- Clusters 1 and 4 from spatial clustering correspond to the tumor
- Cluster 2 has the highest immune expression



Spatial deconvolution

- Most of the immune cells (green) are at the periphery of the tumor
- However, there is some evidence of immune infiltration inside the tumor

