



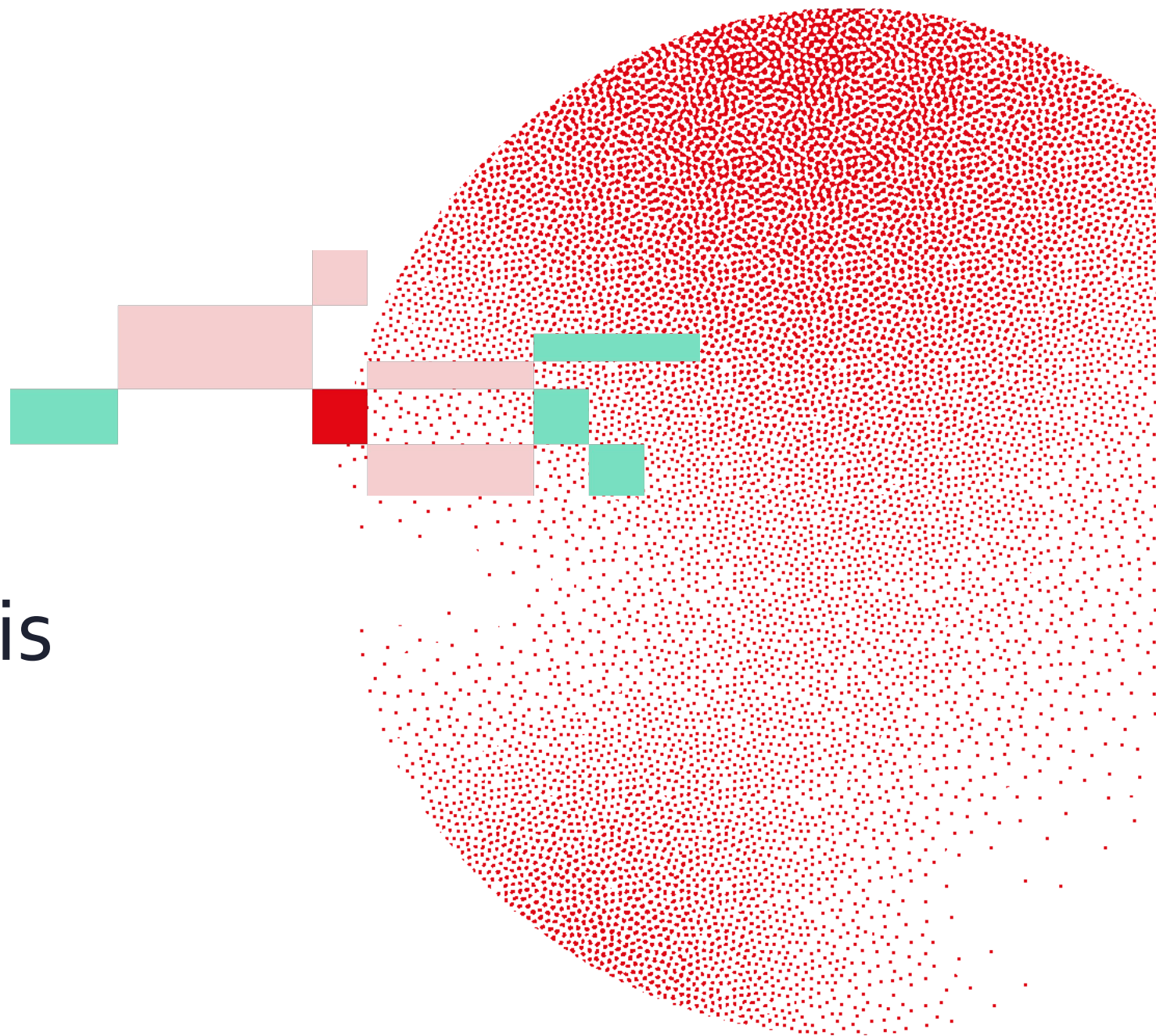
Swiss Institute of  
Bioinformatics

Introduction to Spatial Transcriptomics Data Analysis

# Multi-scale and multi-sample analysis

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05 June 2026





# Outline

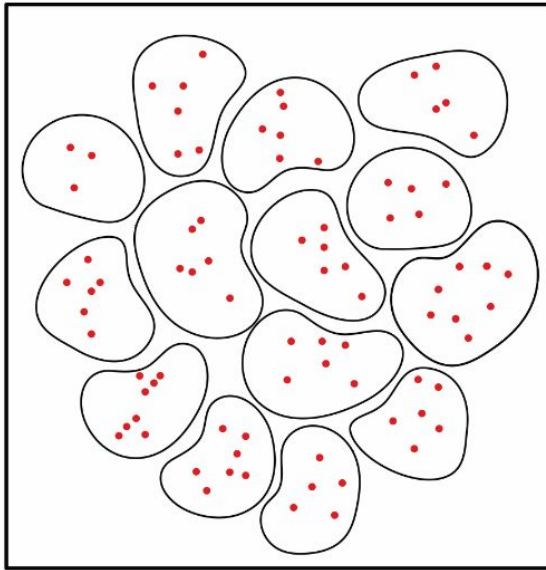
- Lecture on downstream analysis of spatial omics data
  - i) Differential co-localisation analysis with `spatialFDA`
  - ii) Analysis of multicellular anatomical structures with `sosta`
  - iii) Differential spatial pattern analysis with `DEspace`

# Differential co-localisation analysis

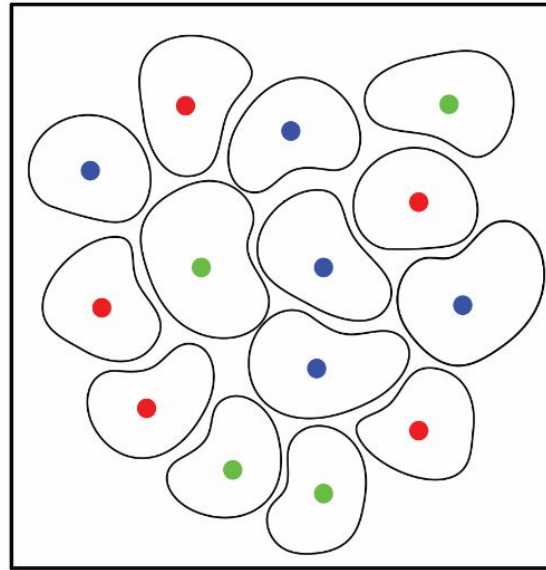


# Approximating cells as points

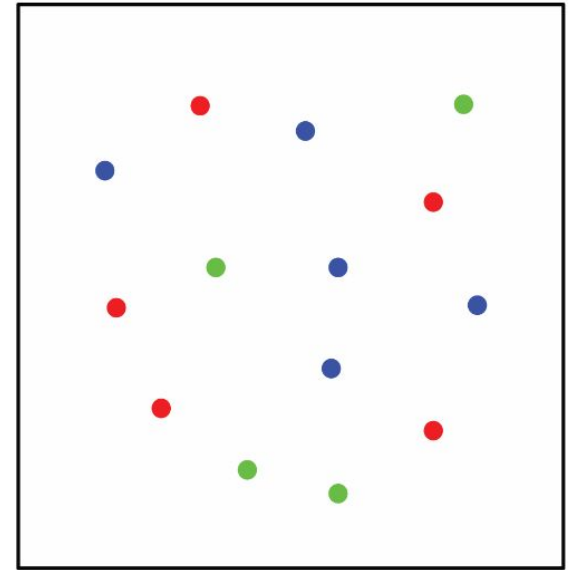
## A) Point pattern approximation of cells



Cells with molecular features



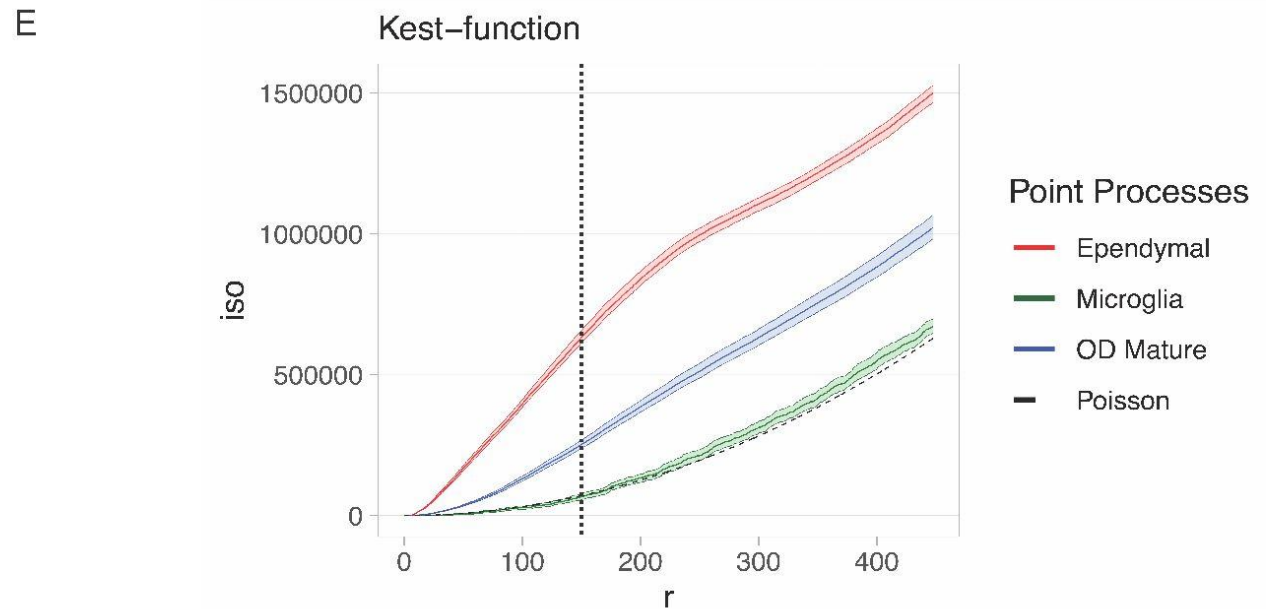
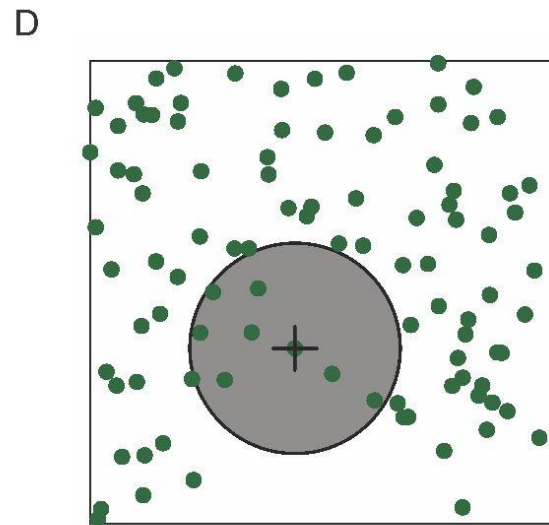
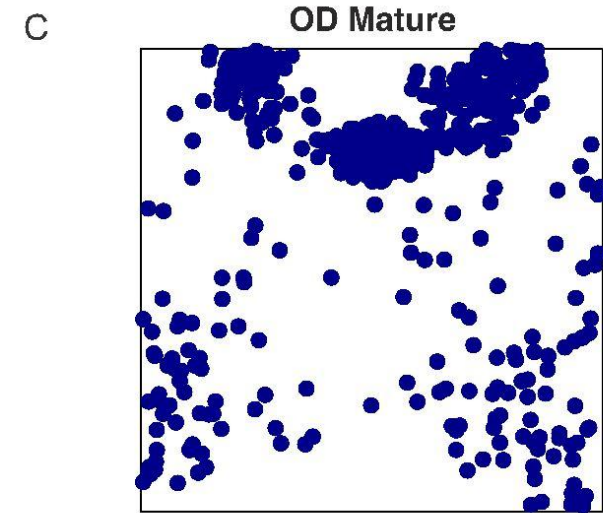
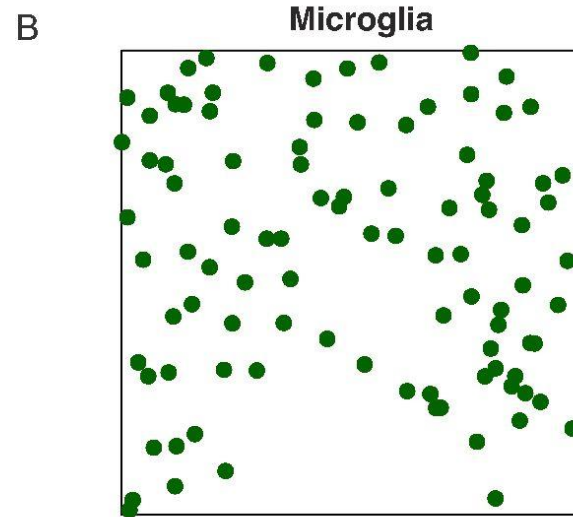
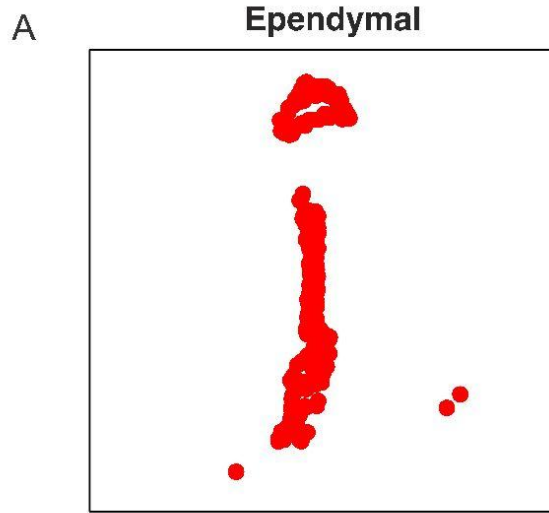
Centroids with cell type labels



Marked point pattern

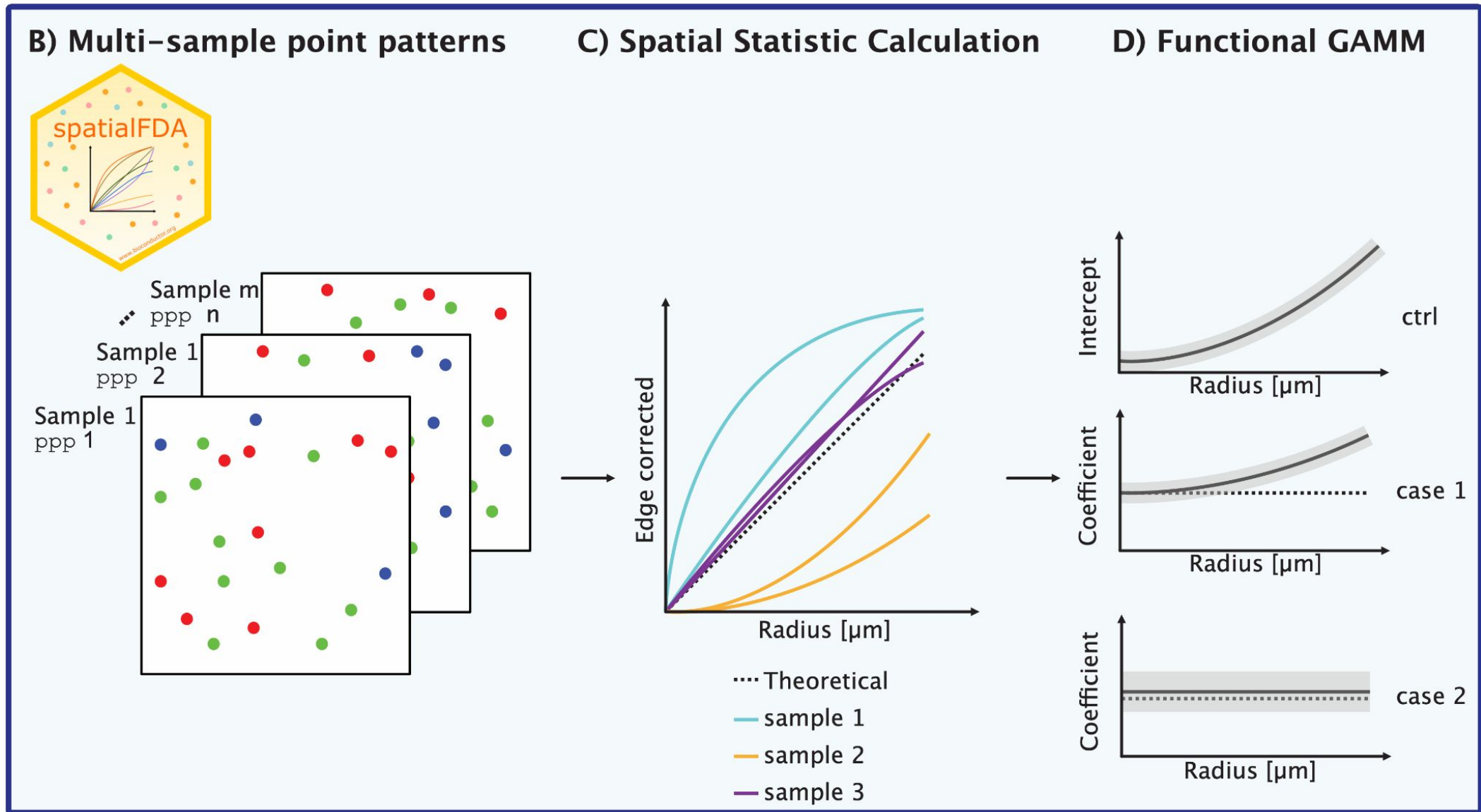


# Point pattern analysis summarises cell arrangements





# spatialFDA: Differential co-localisation across scales





# Generalised functional additive mixed models

functional GAMM

$$\underbrace{\mathbb{E}[y_i(r) | X_i, g(i)]}_{\text{functions}} = \underbrace{g^{-1}}_{\text{link}} \left( \underbrace{\alpha(r)}_{\text{intercept}} + \underbrace{\beta_{0,g(i)}(r)}_{\text{random intercept}} + \underbrace{\sum_{j=1}^J f_j(X_{ji}, r)}_{\text{predictors}} \right)$$

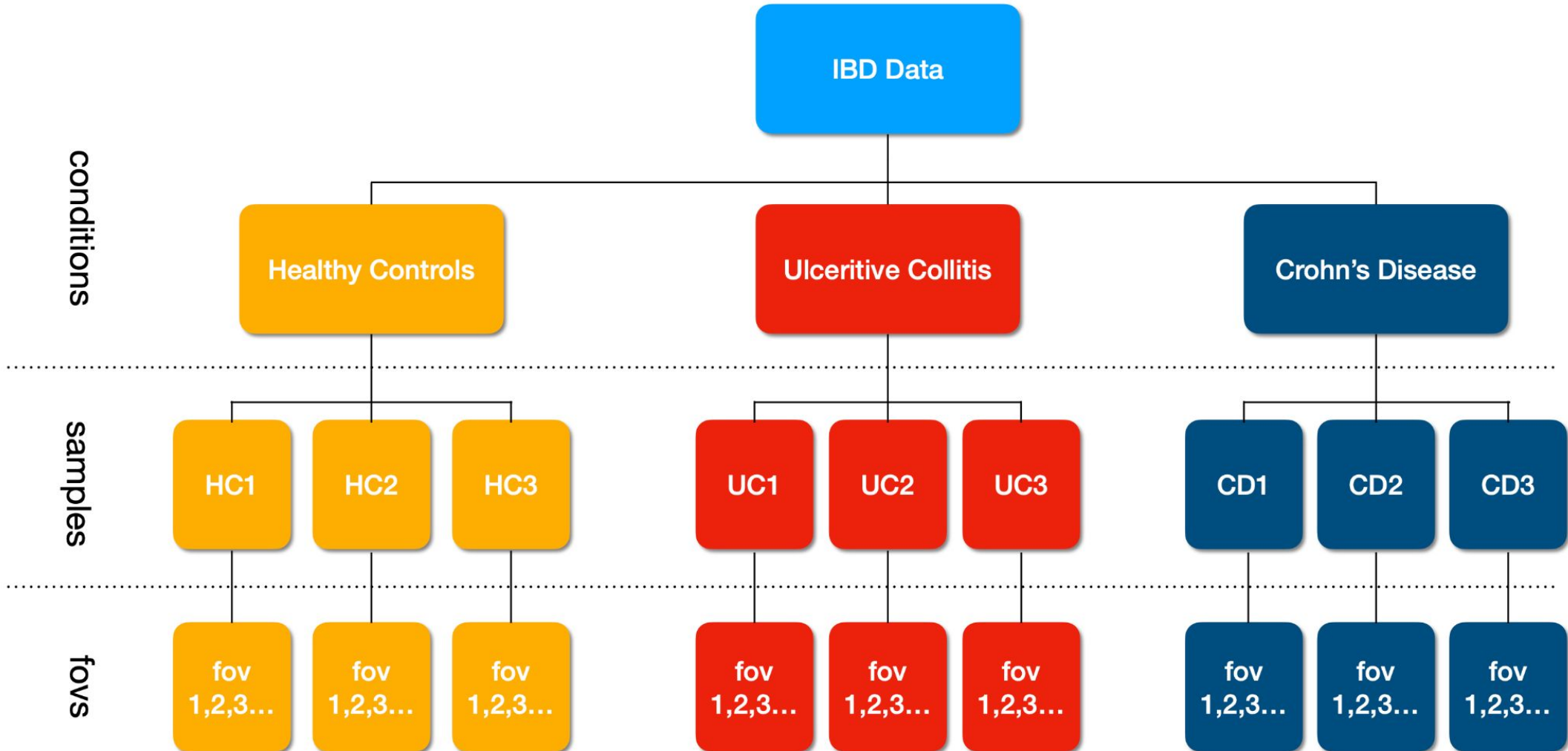
F-test of overall effect

$$H_0 : \beta(r) = 0, \forall r \in [r_{min}, r_{max}]$$

$$H_1 : \exists r \in [r_{min}, r_{max}] \text{ s.t. } \beta(r) \neq 0$$

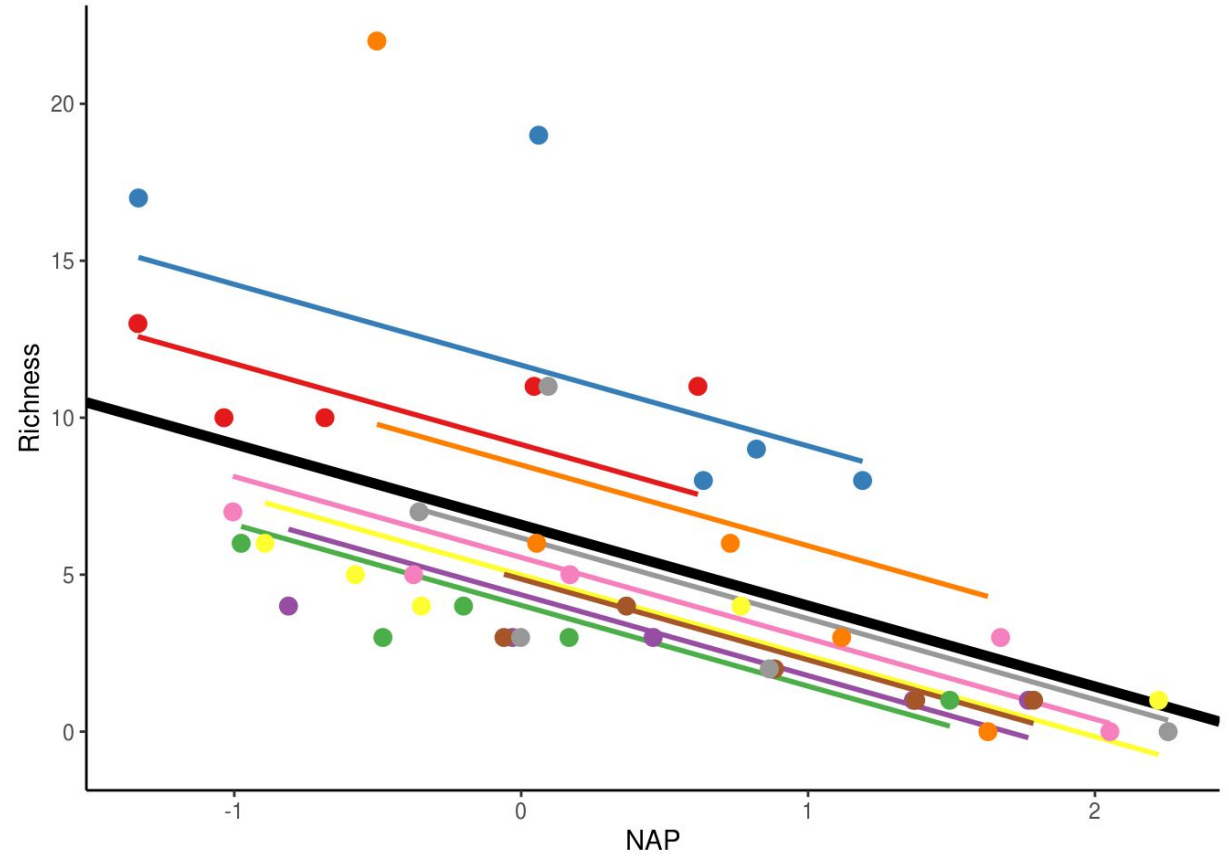
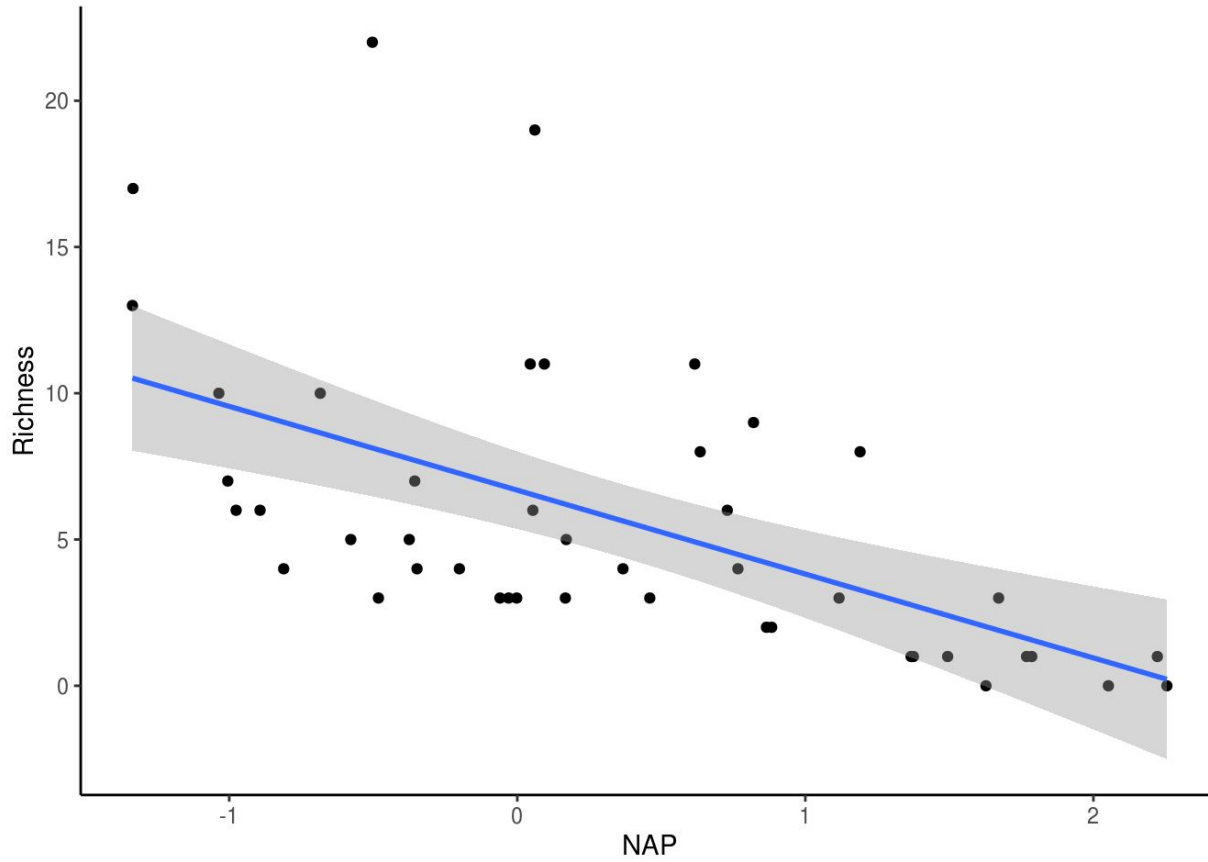


# Modern spatial datasets show nested covariances





# Random intercepts account for group effects

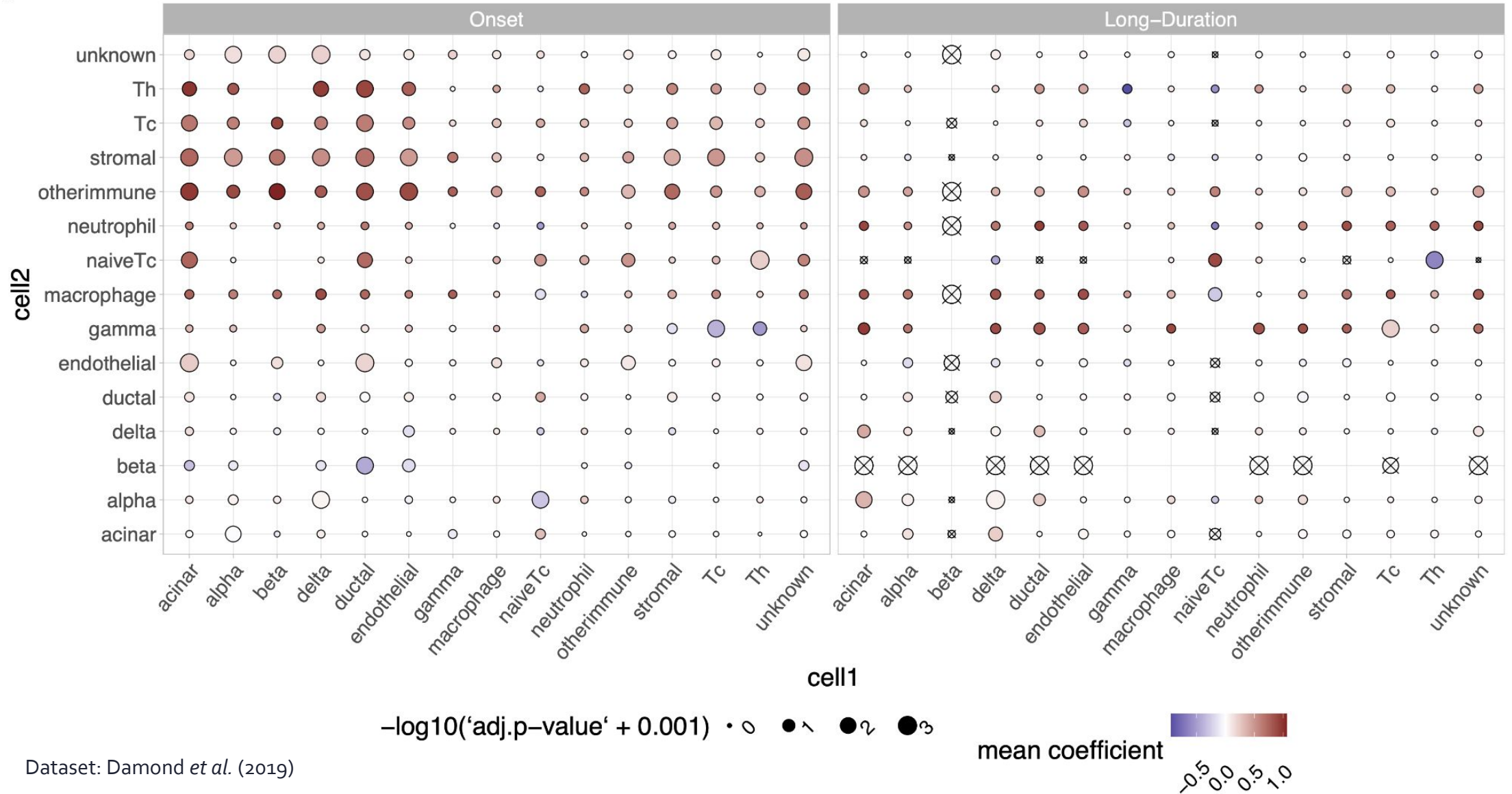


adapted from [Santangelo, 2018], licensed as CC BY 4.0



# Overall $F$ -test results of pairwise cell type combinations

A

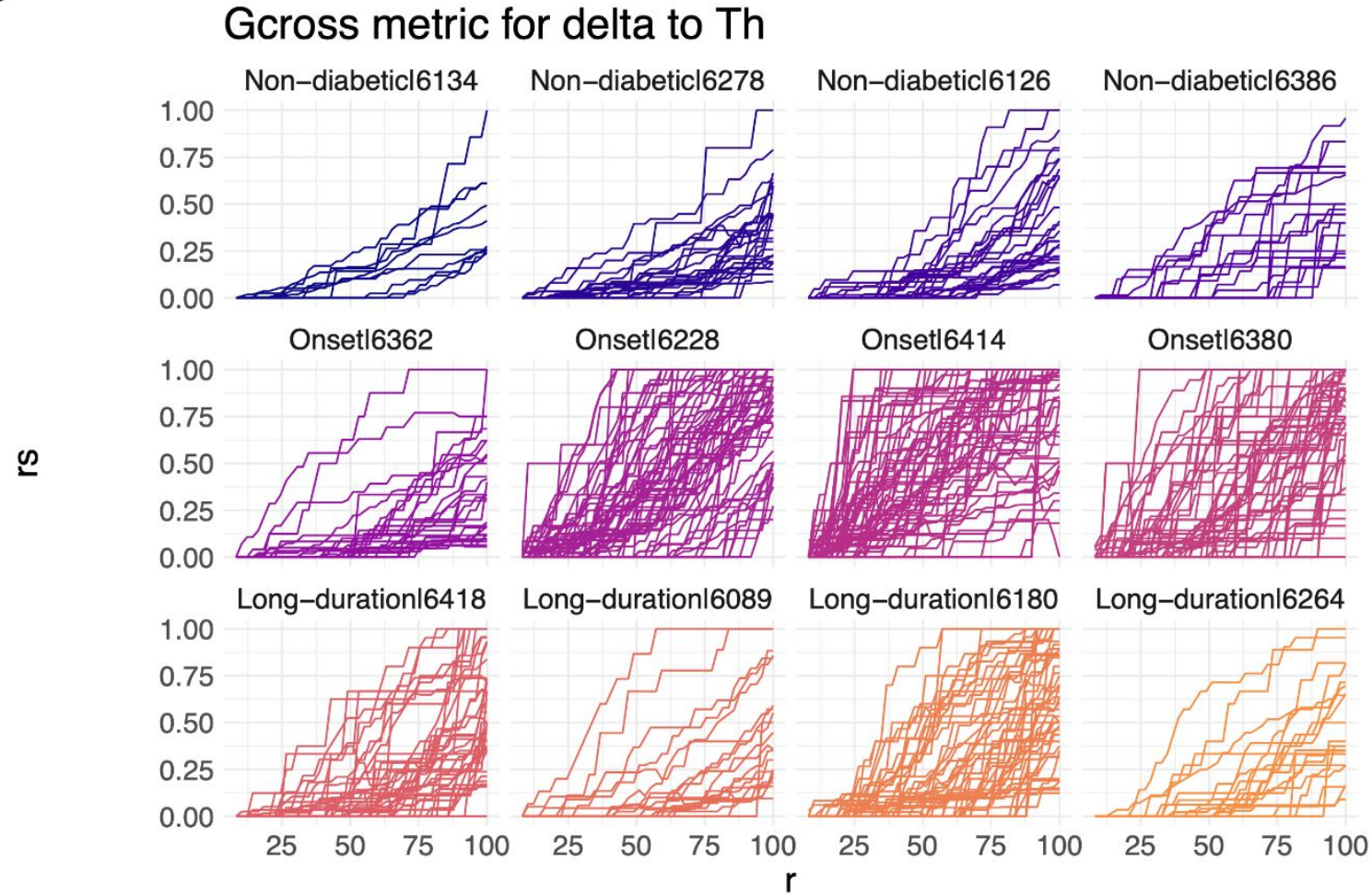


Dataset: Damond *et al.* (2019)



# fGAMM shows scale-effect of delta-Th interaction

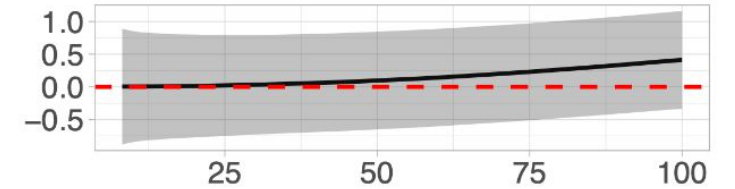
B



Dataset: Damond *et al.* (2019)

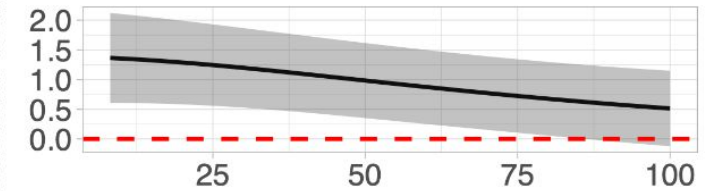
C

Intercept



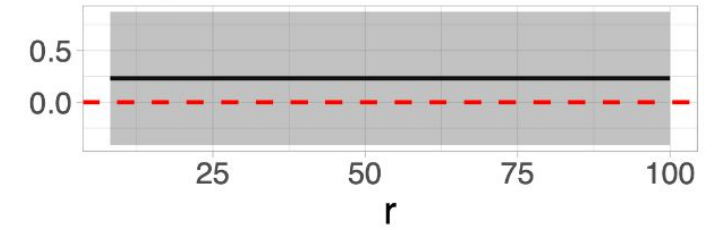
D

Onset

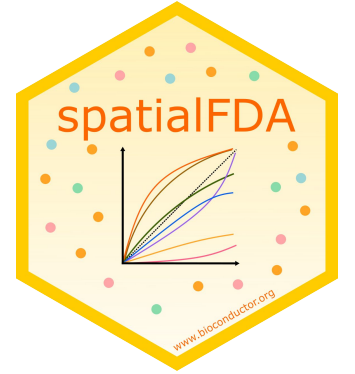


E

Long-Duration



# Summary: i) Differential co-localisation



- We approximate cells as points in space by their centroid
- The cell type is added as a mark → marked point pattern
- `spatialFDA` is a tool for differential co-localisation analysis
- mixed effects models can account for nested covariances
- The spatial scale of differential co-localisation is maintained in `spatialFDA`
- `spatialFDA` is available on R/Bioconductor

# Analysis of multicellular anatomical structures

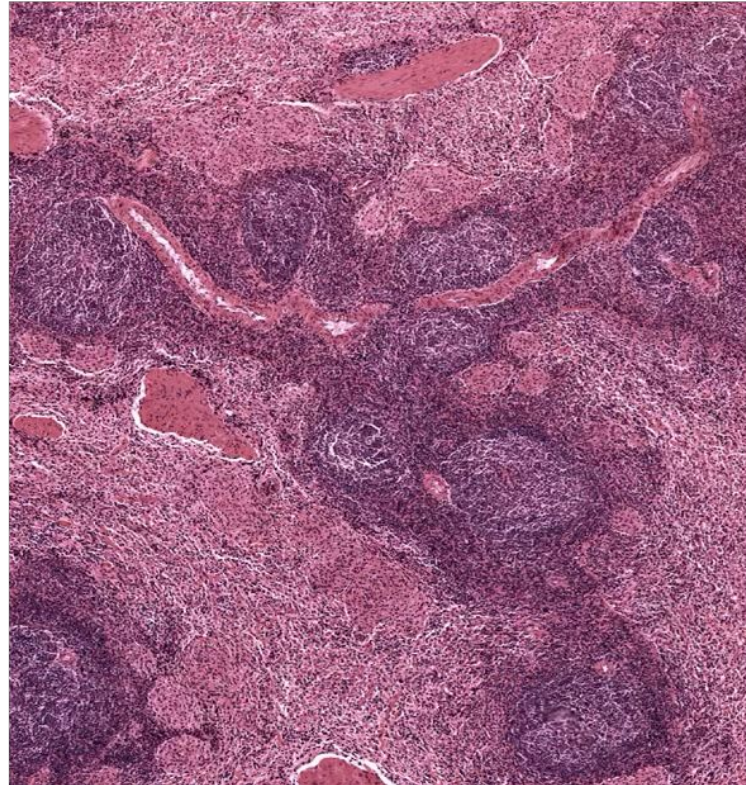


# Biological function is often reflected in morphology and location

Ileum, villi and crypt



Spleen, red and white pulp



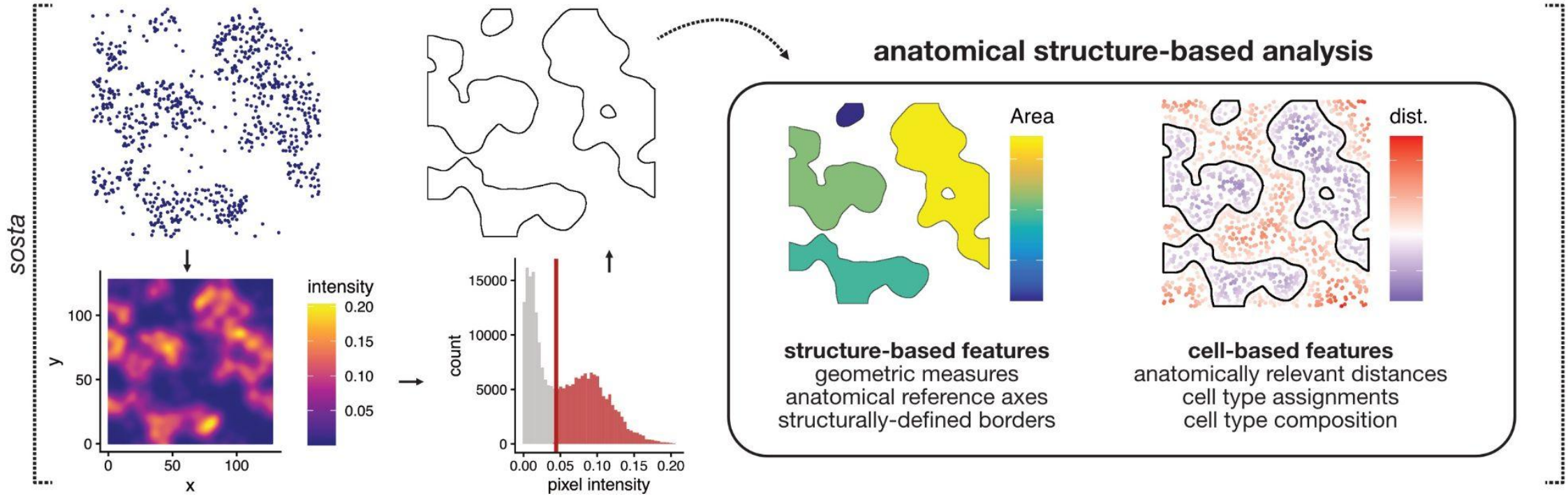
Ureter



Virginia University - Digital Histology; Pakura, Bigbee *et al.*; <https://digitalhistology.org/>; (CC BY-NC-SA 4.0)

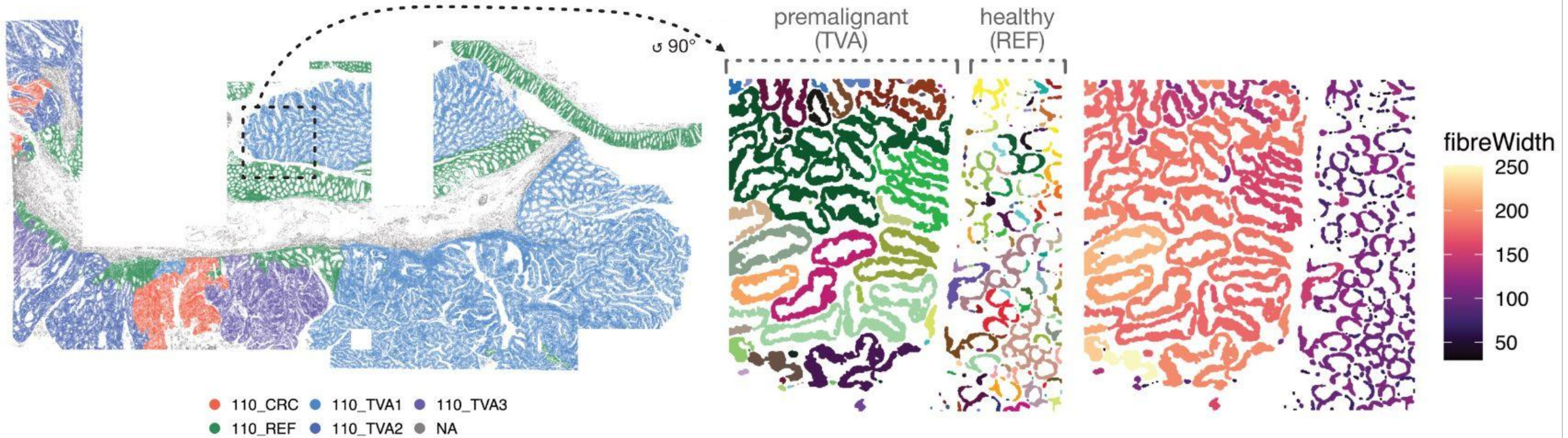


# Analysis of anatomical structures using *sosta*



Gunz et al. (2025)

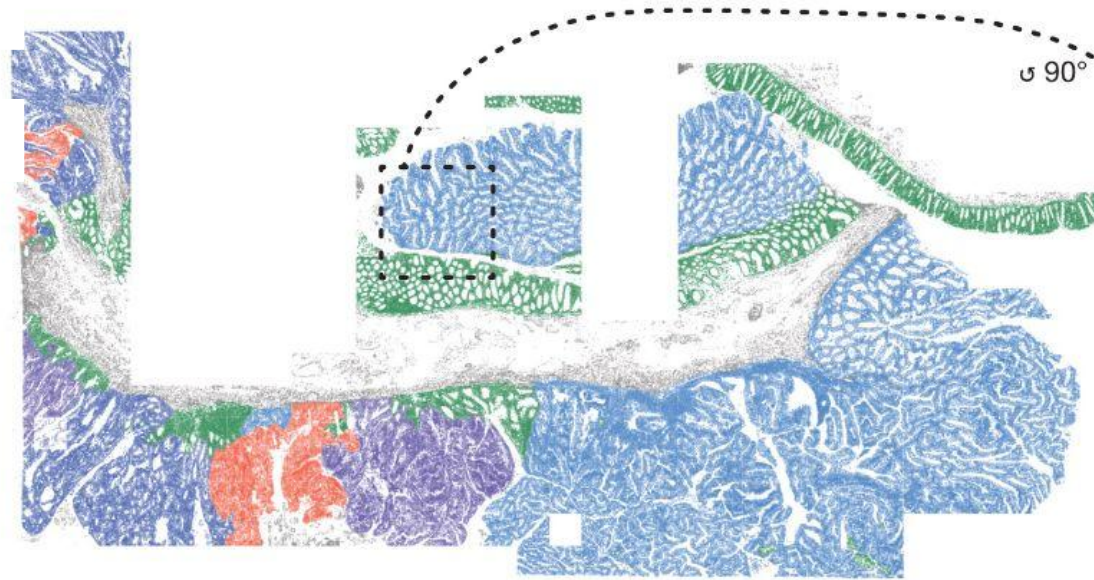
# Structure based analysis: Geometric features



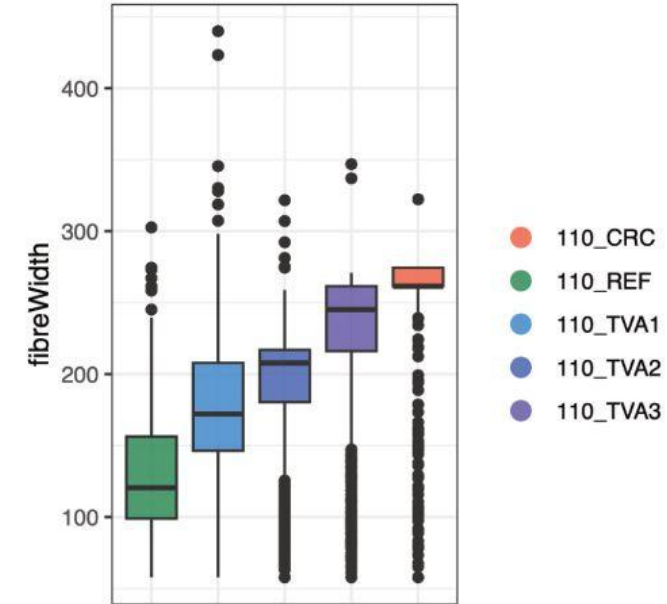
Dataset: Crowell *et al.* (2025)



# Structure based analysis: Geometric features



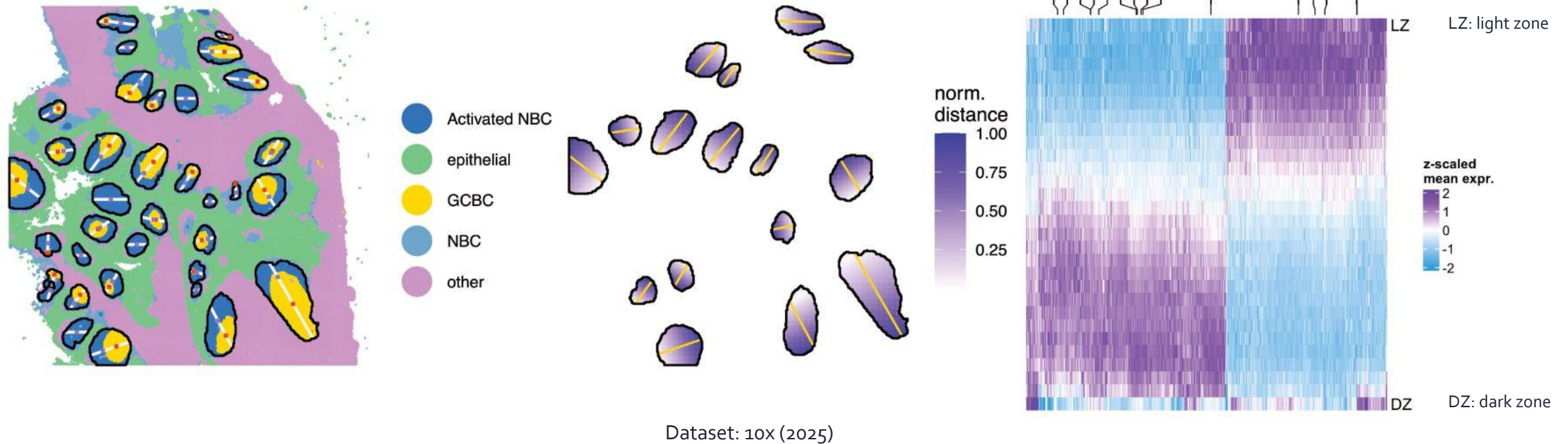
● 110\_CRC ● 110\_TVA1 ● 110\_TVA3  
● 110\_REF ● 110\_TVA2 ● NA



- Fibre width increases along the pathological progression: healthy (HC) to tubulovillous adenoma (TVA) to colorectal cancer (CRC)
- Reflects gradual morphological changes related to increase in crypt thickness.



# Structure based analysis: Gene expression



- Germinal centers are well defined anatomical structures with clear orientation
- Structure-based gene expression analysis to recover DZ / LZ markers

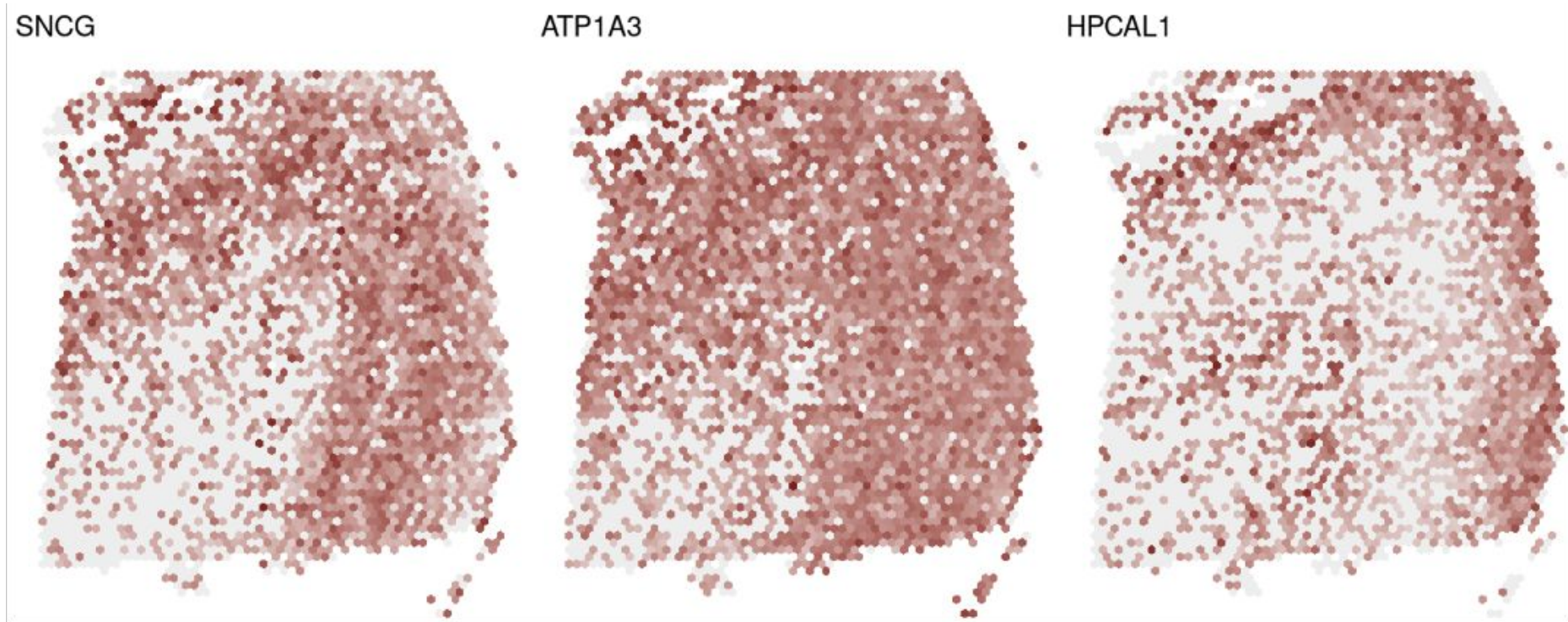


## Summary: ii) Analysis of multicellular anatomical structures

- Analysis of anatomical structures in spatial omics data inspired by “form reflects function”
- Geometrical features can reflect disease progression
- Anatomical information can be used as a reference system for spatial gene expression analysis
- Common challenge: what are relevant anatomical structures and how to identify them
- `sosta` is an R/Bioconductor package for analysis of anatomical structures in spatial omics data

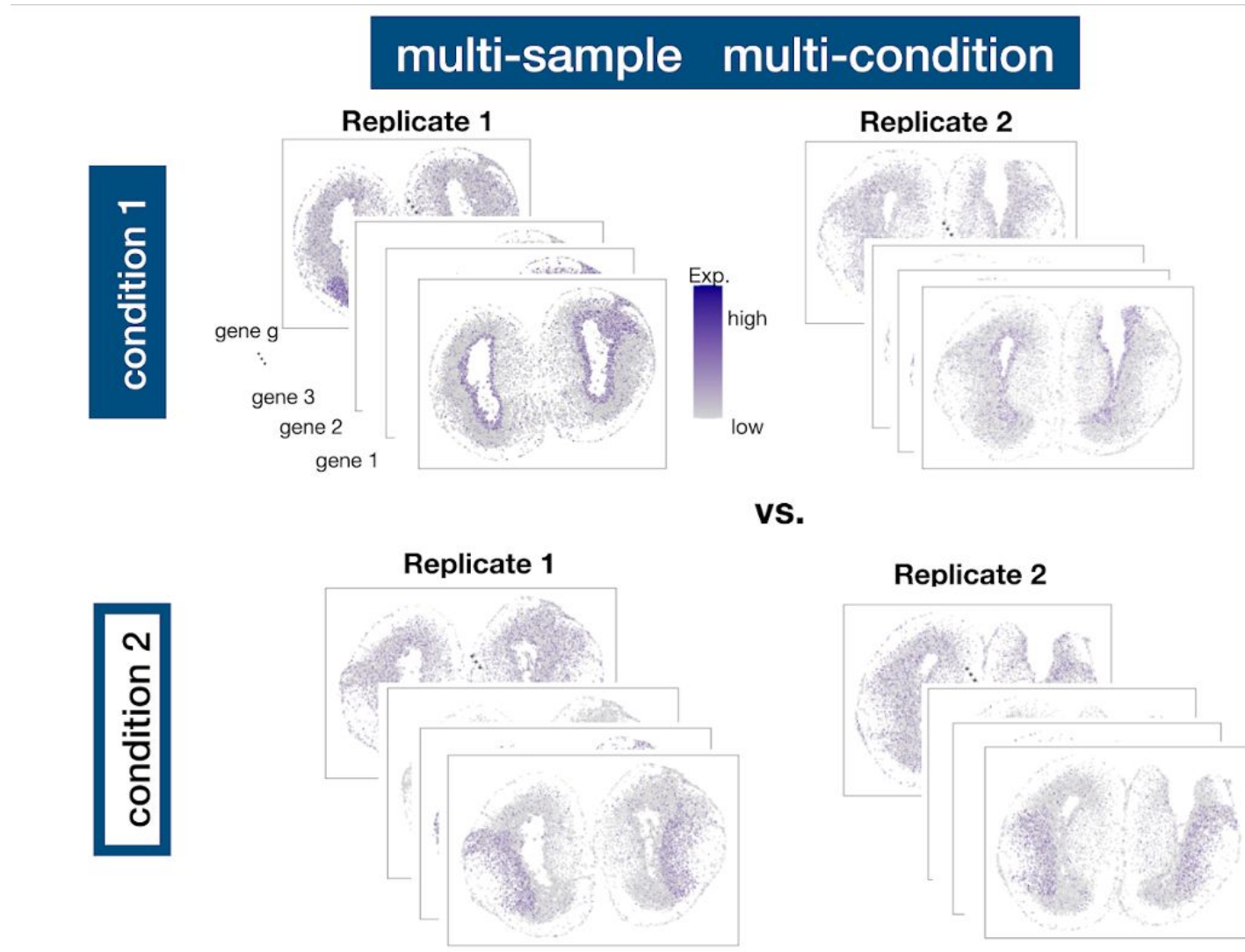
# Differential spatial pattern analysis

# Recap: spatially variable genes



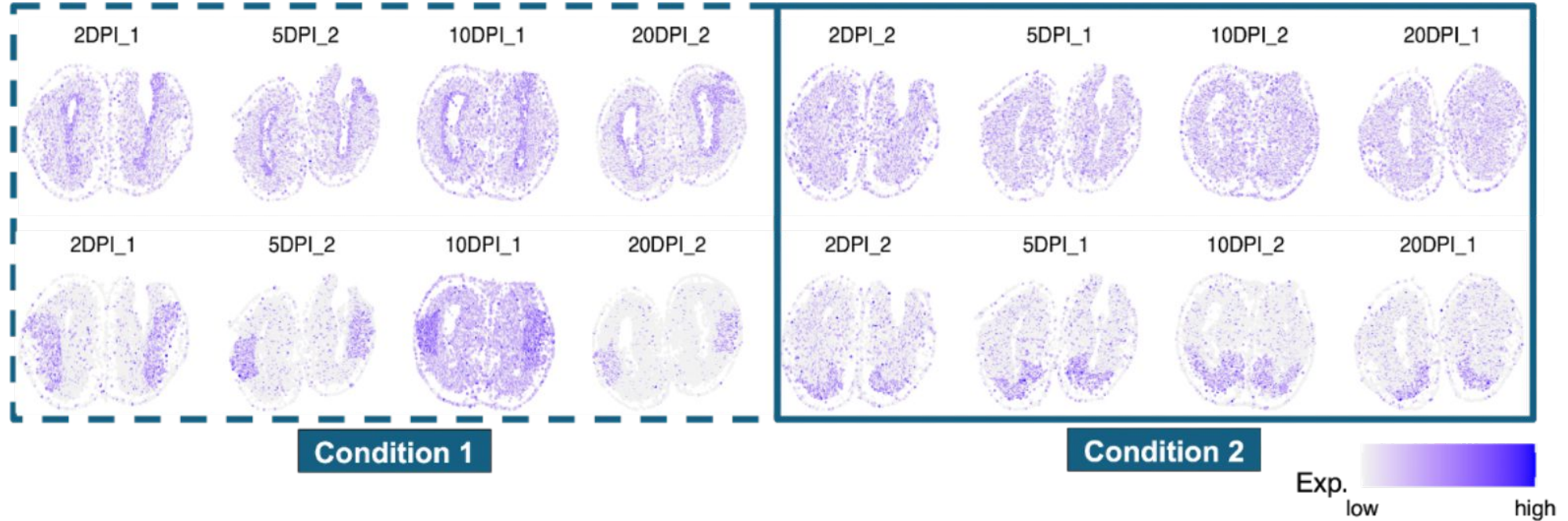


# From one tissue section to multiple samples and conditions





# Differential spatial patterns (DSP)

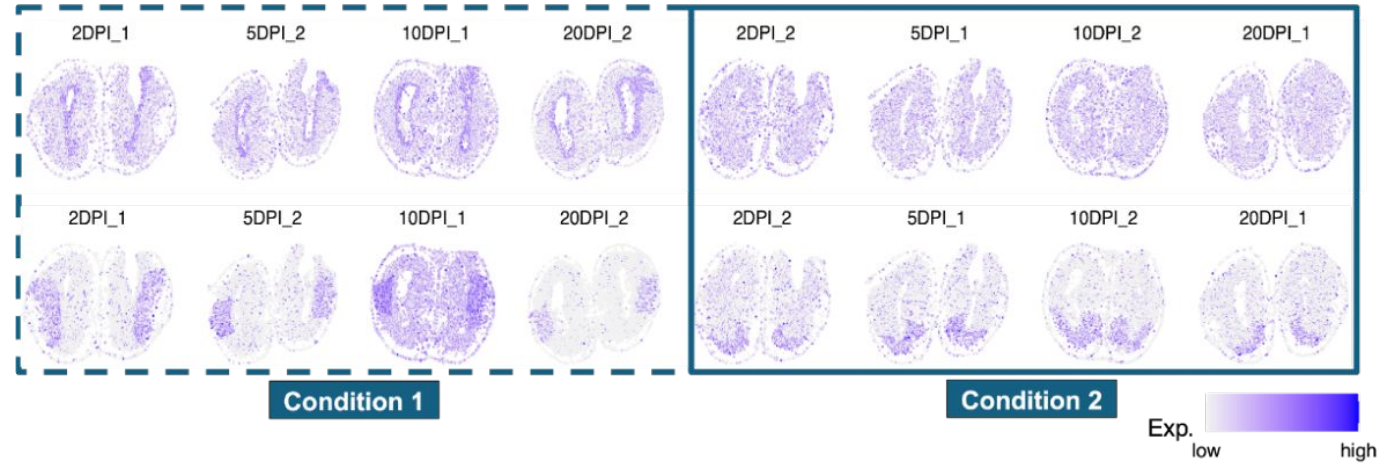


DSP genes are those whose spatial expression patterns change across groups, such as different treatment conditions or time phases.

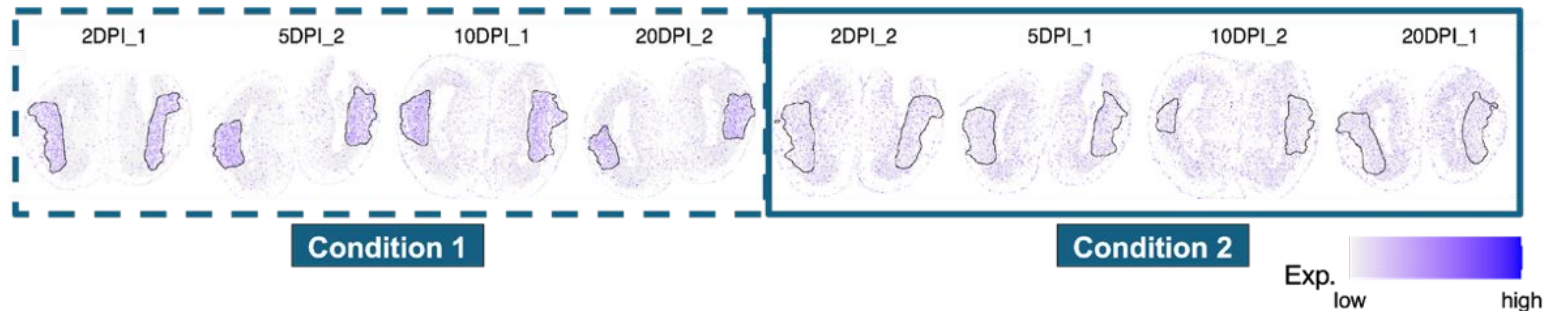


# DESpace:

1. Global test: detect DSP genes.



2. Individual cluster test: identify the spatial clusters where gene abundance varies across conditions





# Obtaining shared domains: BANKSY + Harmony

Harmony correction on Banksy PCA embedding

1. Compute BANKSY representation per sample

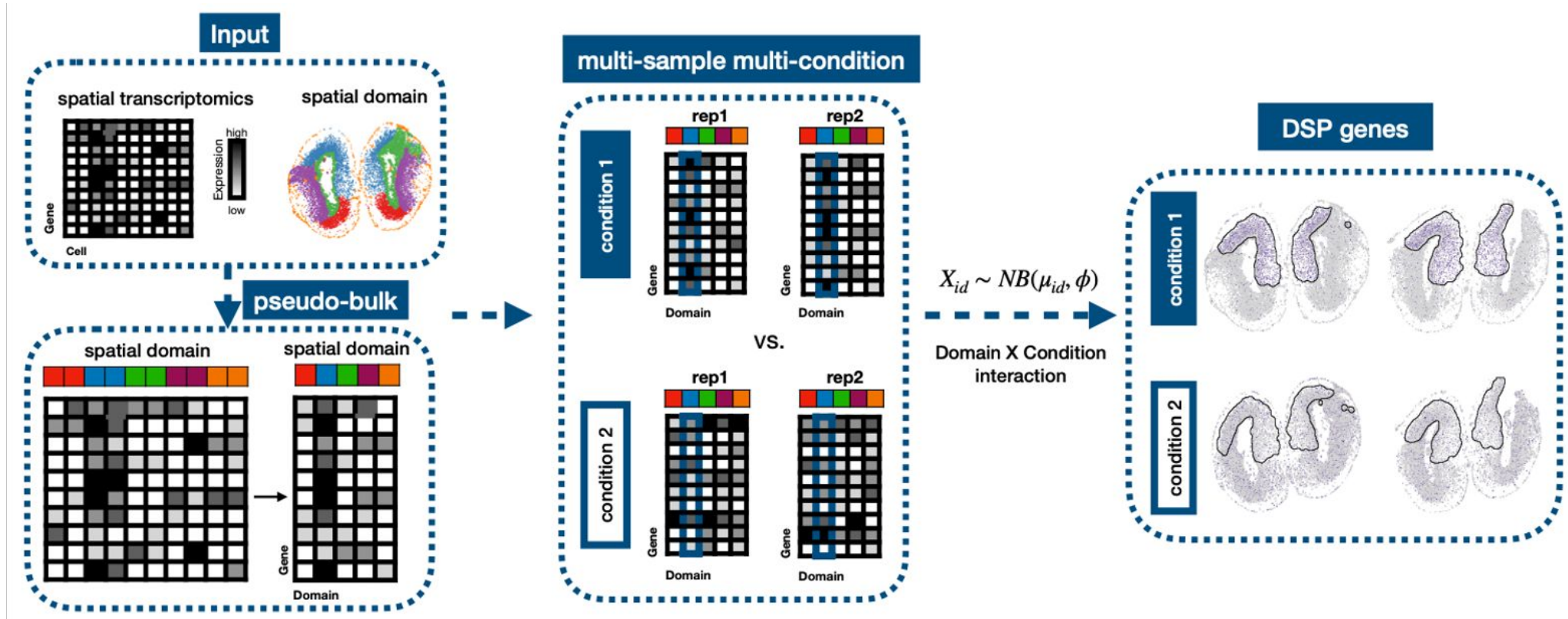
2. PCA on BANKSY representation

3. Use Harmony to correct sample/batch effects in PCA space

4. Run UMAP/clustering on Harmony-corrected embedding



# Methodological details

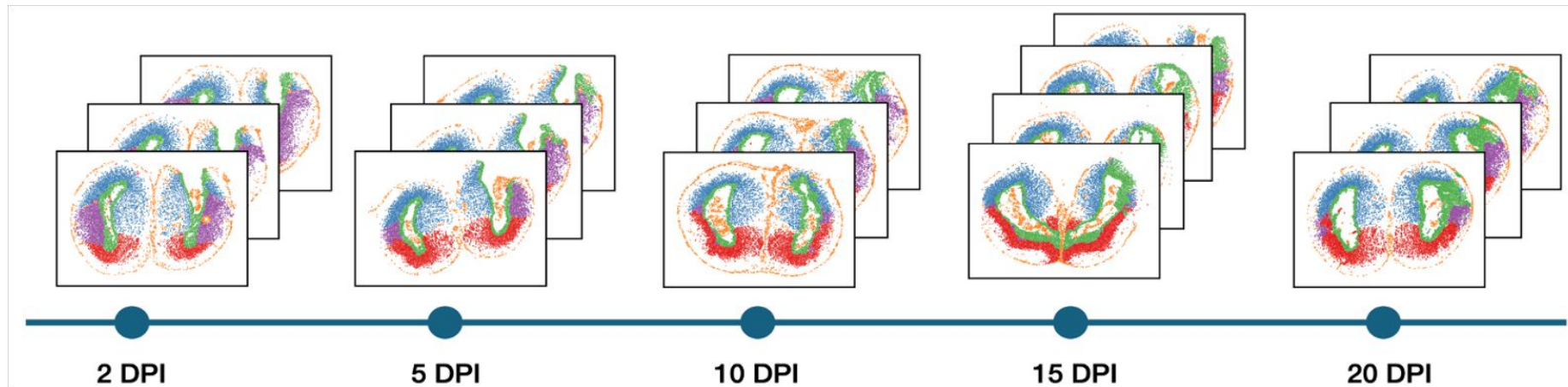


$$X_{id} \sim NB(\mu_{id}, \phi),$$
$$\log(\mu_{id}) = \log(M_{id}) + \beta_d + \beta_{c_i} + \beta_{c_i d},$$

for  $i = 1, \dots, N$ ,  $d = 1, \dots, D$ ,  
and  $c_i = 1, \dots, N_c$ ,

# Biological example

- ARTISTA (Stereo-seq) dataset captures axolotl brain regeneration at single-cell resolution
- 16 samples from 5 time points, i.e., days post-injury (DPI)
- 5 consistent spatial clusters



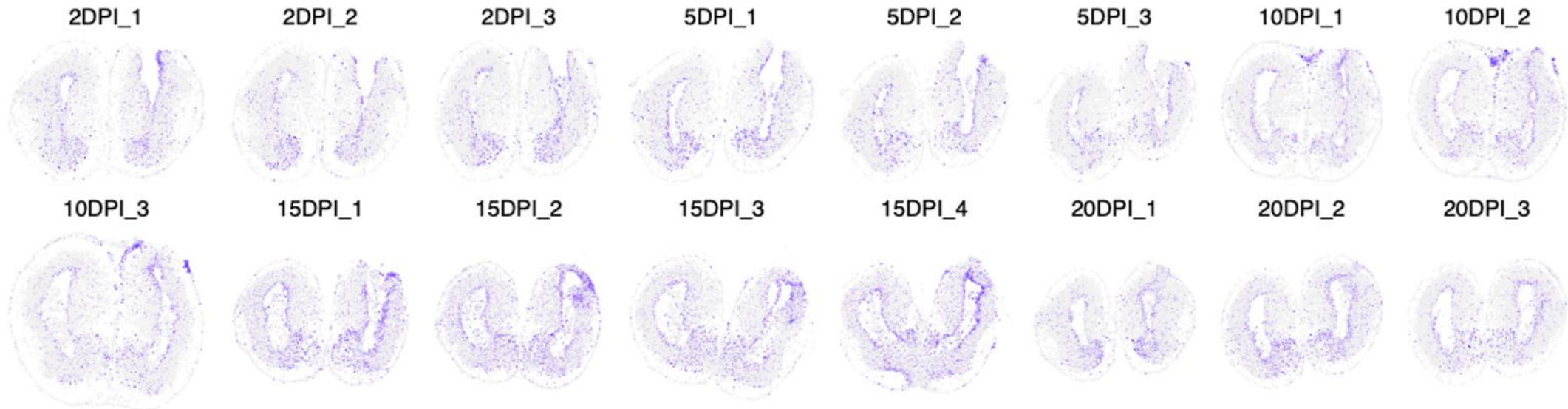
Wei, Xiaoyu, et al. "Single-cell Stereo-seq reveals induced progenitor cells involved in axolotl brain regeneration." *Science* 377.6610 (2022): eabp9444.



# Biological example

DSP gene example

TNC: a glycoprotein in adult neurogenic niches, involved in tissue repair and regeneration





## Summary: iii) Differential spatial pattern

- SVGs capture expression patterns that vary across tissue structure within sample.
- DSP analysis asks whether spatial expression patterns change across conditions.



# References

Cai, Peiying, Mark D. Robinson, and Simone Tiberi. "DESpace2: detection of differential spatial patterns in spatial omics data." *bioRxiv* (2025): 2025-06.

Crowell, Helena L., et al. "Tracing colorectal malignancy transformation from cell to tissue scale" *bioRxiv* (2025): 2025-06.

Damond, Nicolas, et al. "A map of human type 1 diabetes progression by imaging mass cytometry." *Cell metabolism* 29.3 (2019): 755-768.

Emons, Martin, et al. "Differential co-localisation analysis of multi-sample and multi-condition experiments with spatialFDA." *bioRxiv* (2026): 2026-04.

Gunz, Samuel, Helena L. Crowell, and Mark D. Robinson "Analysis of multicellular anatomical structures from spatial omics data using sosta" *bioRxiv* (2025): 2025-10.

Santangelo, James S. "Linear Mixed-Effects Models." *UofT Coders*, 2018, [uoftcoders.github.io/rcourse/leco8-linear-mixed-effects-models.html](https://uoftcoders.github.io/rcourse/leco8-linear-mixed-effects-models.html). Accessed 8 Dec. 2025.

Wei, Xiaoyu, et al. "Single-cell Stereo-seq reveals induced progenitor cells involved in axolotl brain regeneration." *Science* 377.6610 (2022): eabp9444.

