

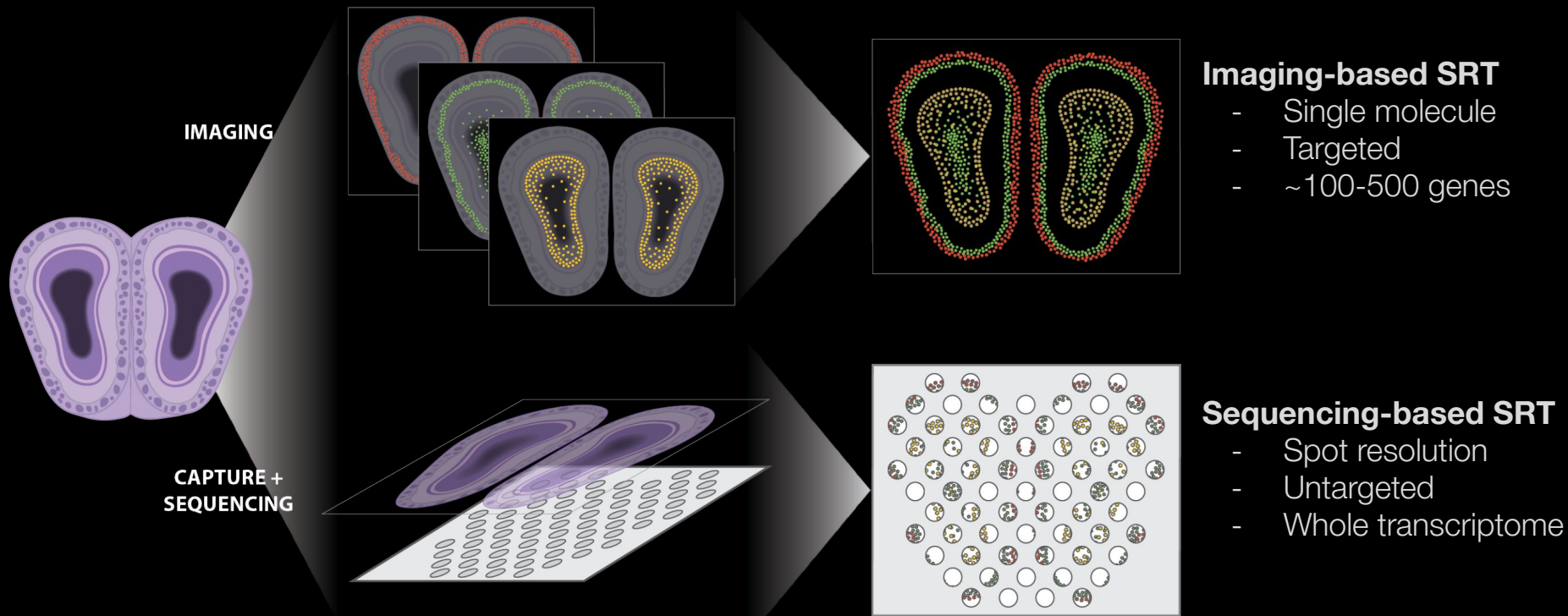
Gene count normalization in single-cell imaging-based spatially resolved transcriptomics

Lyla Atta

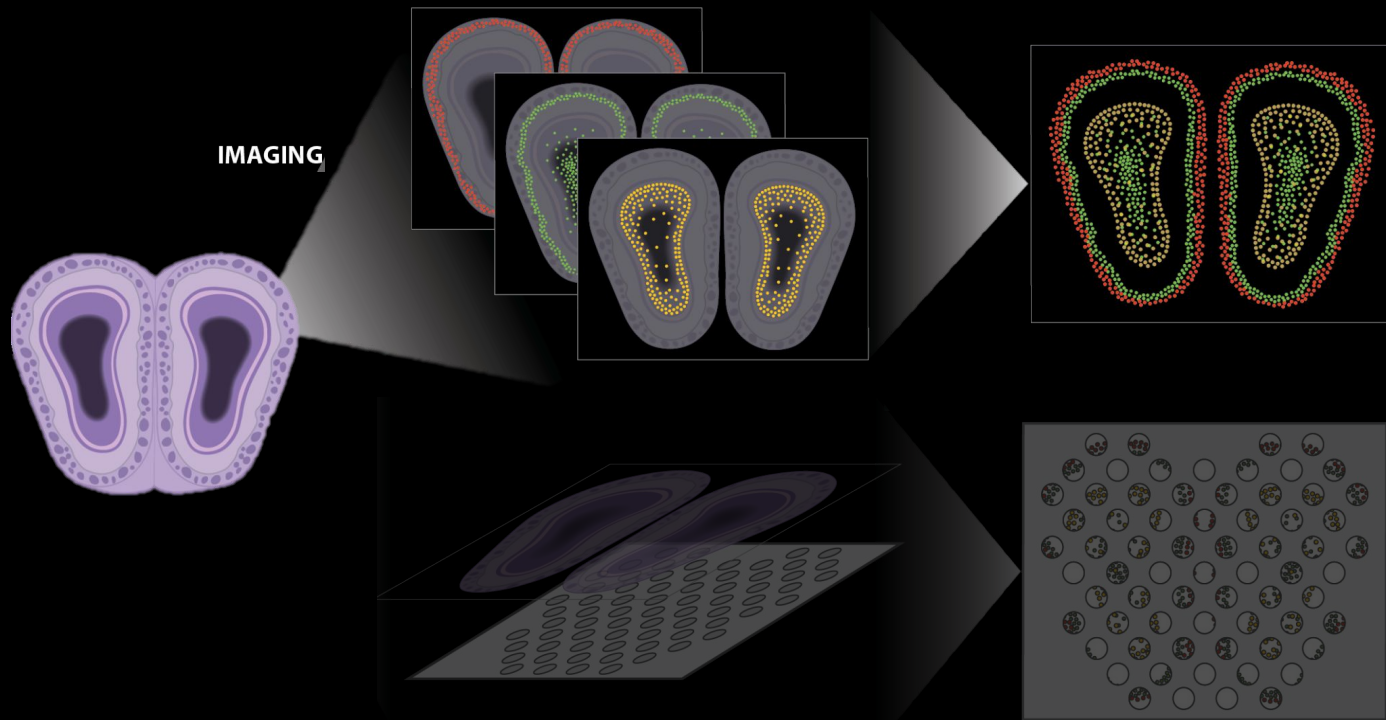
MD-PhD Candidate, JEFworks Lab
Johns Hopkins University



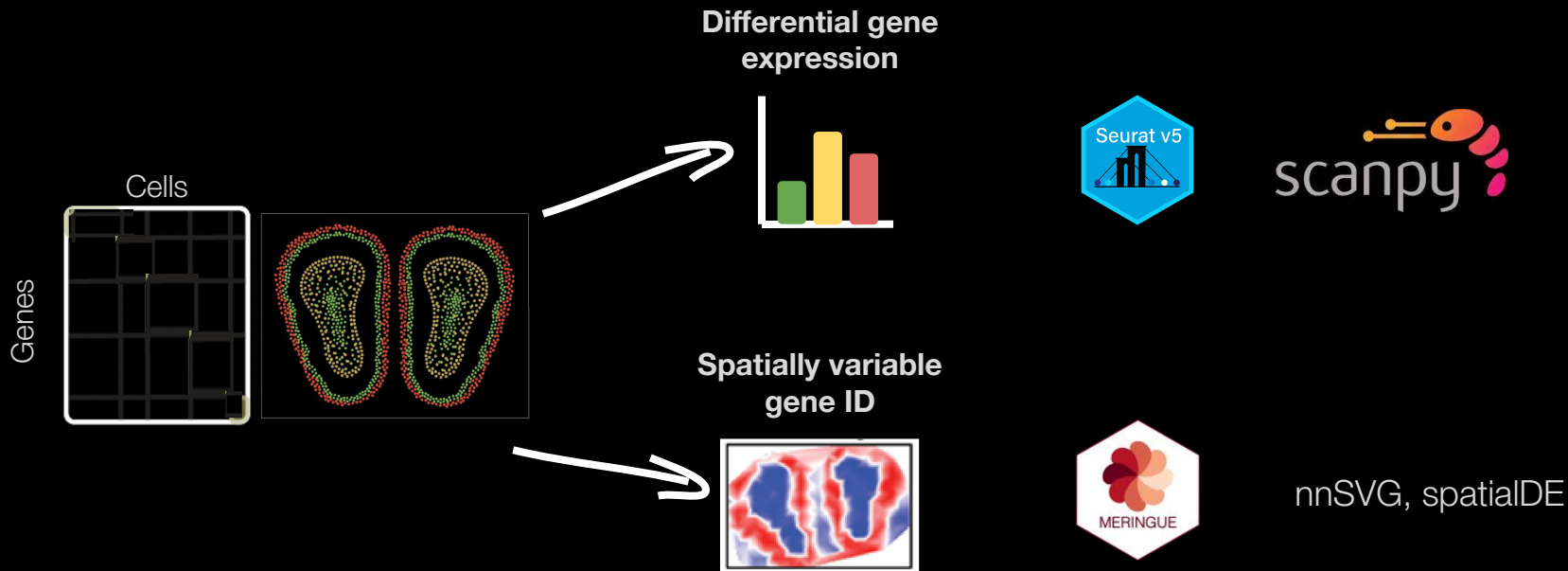
Spatially resolved transcriptomics (SRT)



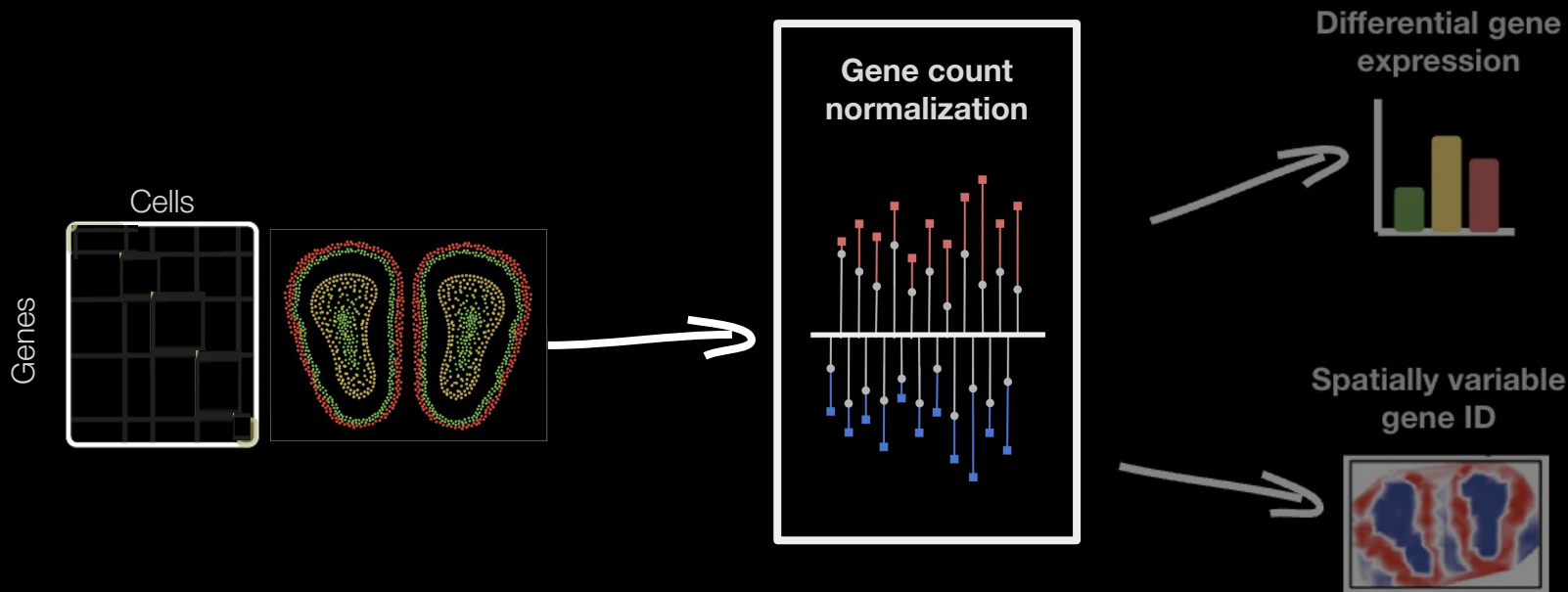
Spatially resolved transcriptomics (SRT)



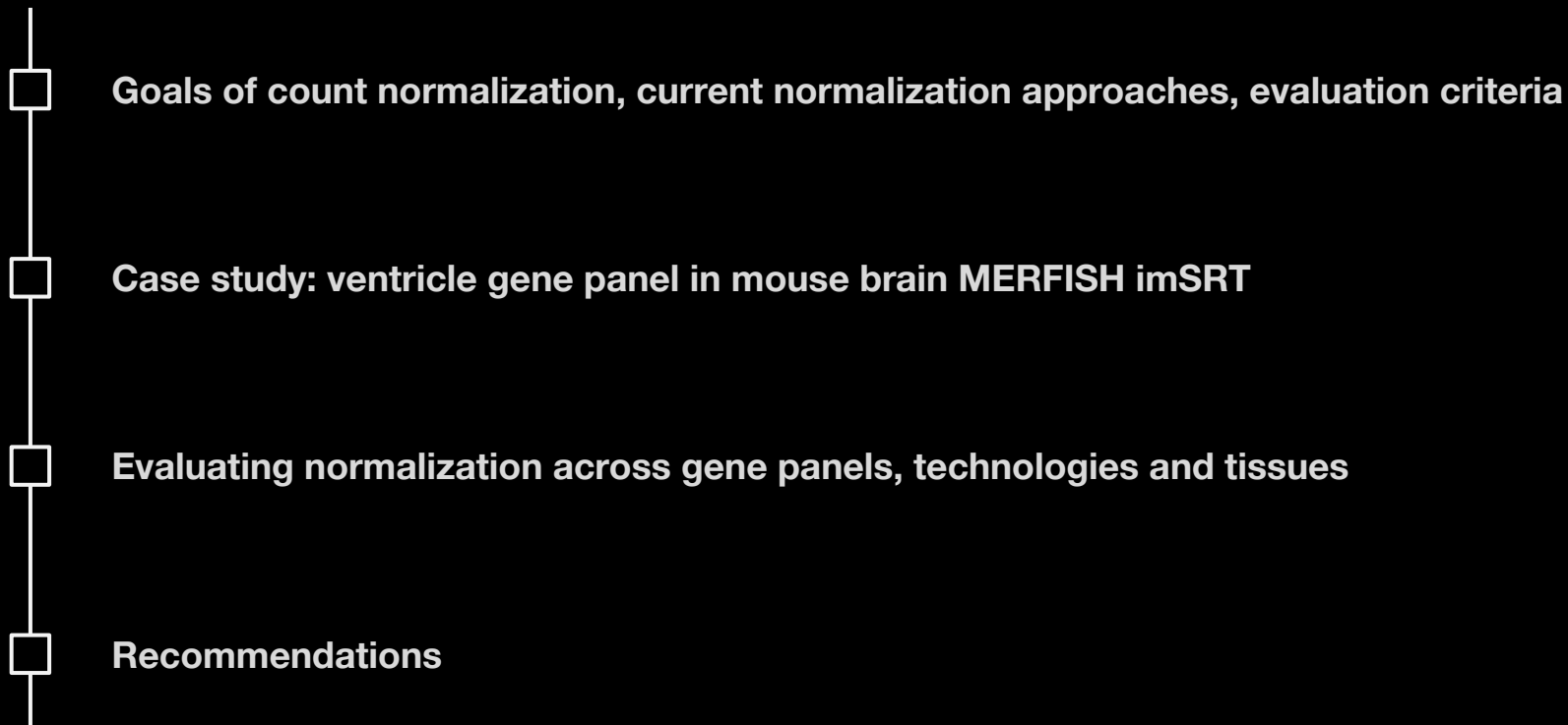
imSRT: computational methods to generate biological hypotheses



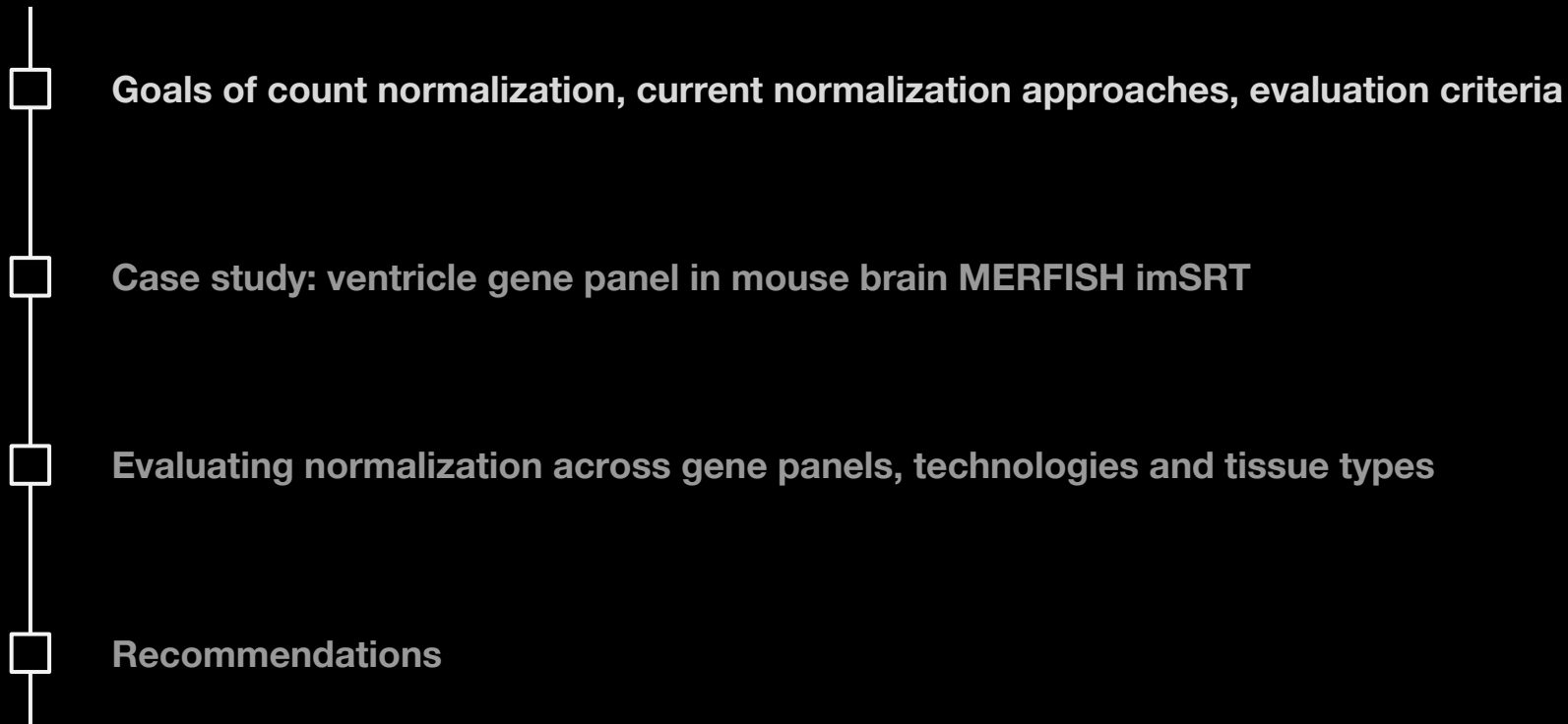
imSRT: gene count normalization is the first step to many downstream analyses



Count normalization in imaging-based SRT



Count normalization in imaging-based SRT



Gene count normalization: accounting for technical variation



- **Bulk RNA-seq:** PCR bias, reagent variation
- **scRNA-seq:** stochastic RNA capture
- **imSRT:** partial cell volume imaging

Gene count normalization: current approaches

Library size

DESeq2

TMM (edgeR)

scTransform

Cell volume/
cell area

Bulk RNA-seq
scRNA-seq
imSRT

Bulk RNA-seq

Bulk RNA-seq

scRNA-seq

imSRT

Gene count normalization: current approaches

Library size	DESeq2	TMM (edgeR)	scTransform	Cell volume/ cell area
Bulk RNA-seq scRNA-seq imSRT	Bulk RNA-seq	Bulk RNA-seq	scRNA-seq	imSRT

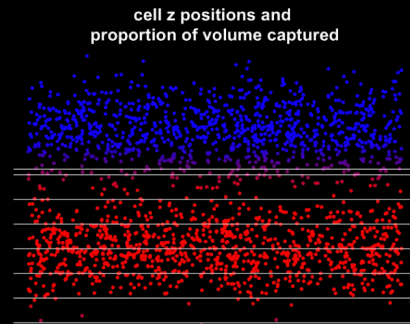
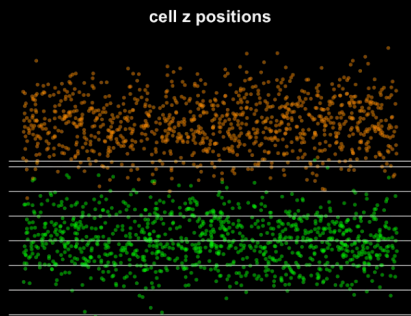
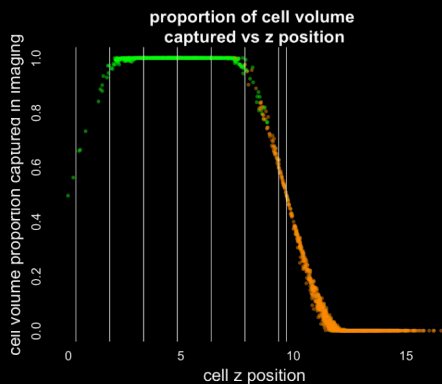
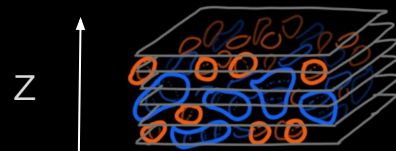


COUNT BASED NORMALIZATION

Gene count normalization: is it necessary?

Simulation to evaluate systematic RNA capture biases

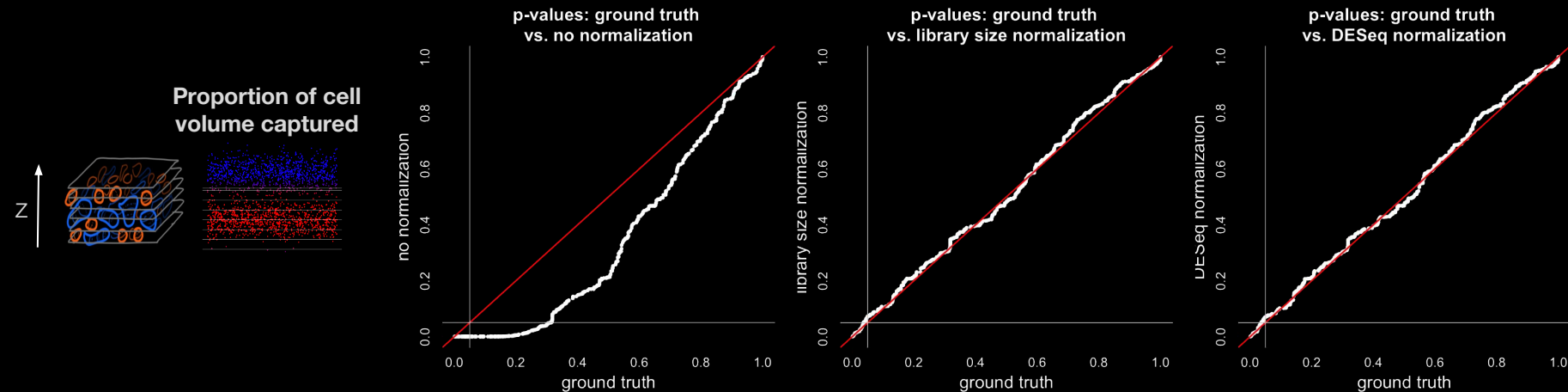
One cell type, two z locations



Gene count normalization: is it necessary?

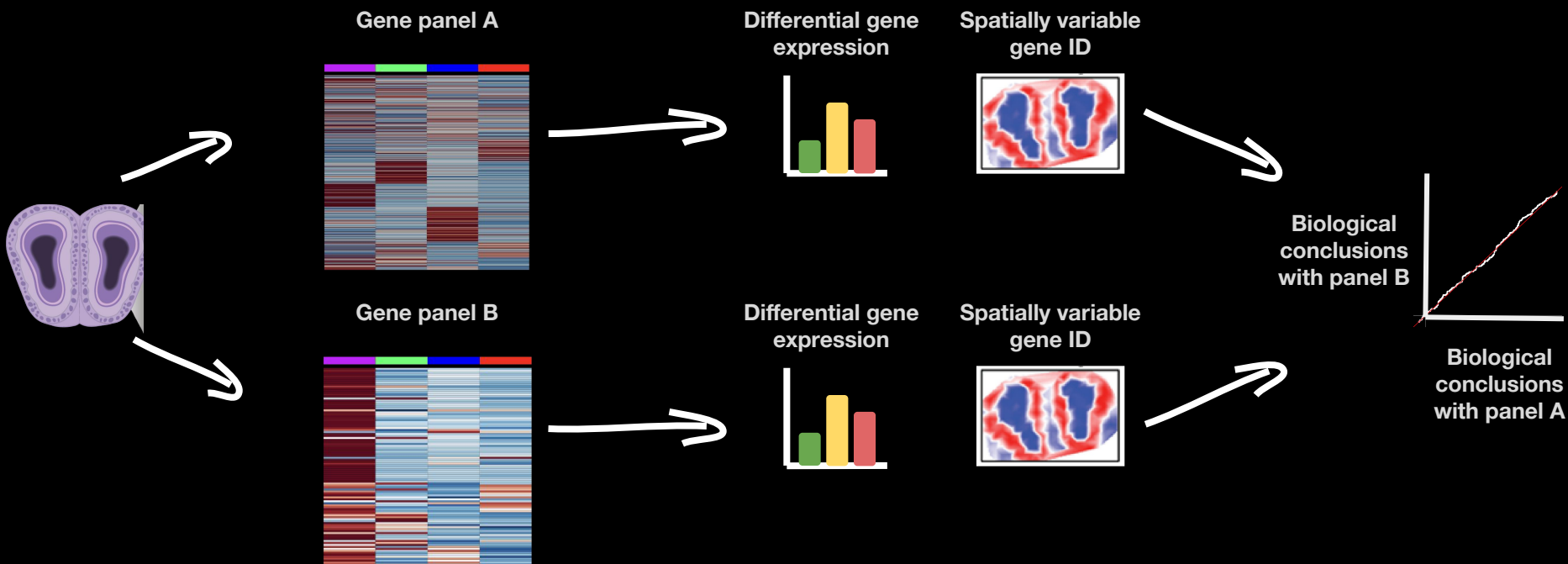
Simulation to evaluate systematic RNA capture biases

Partial cell capture in imaged volume: inflated Type I error rate without normalization

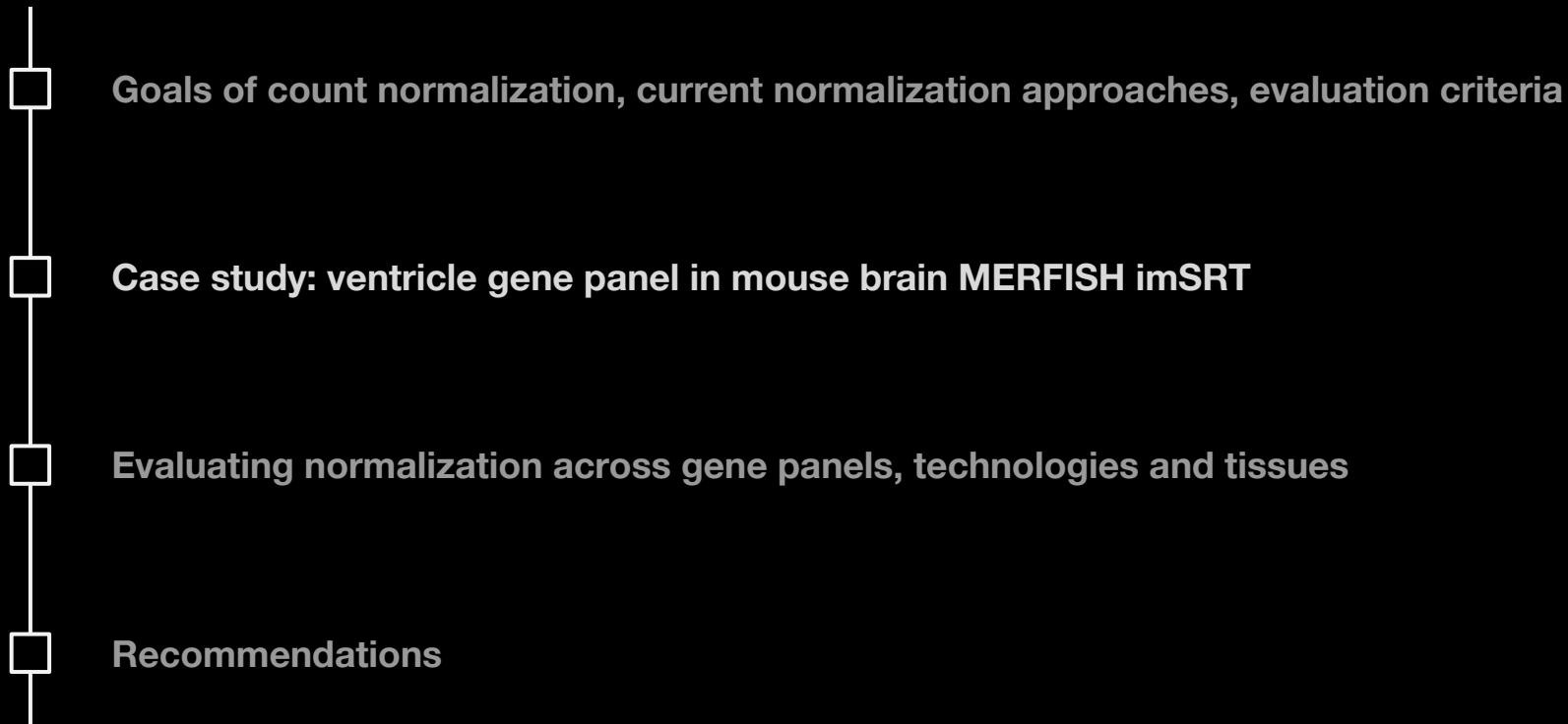


Gene count normalization: evaluation criteria

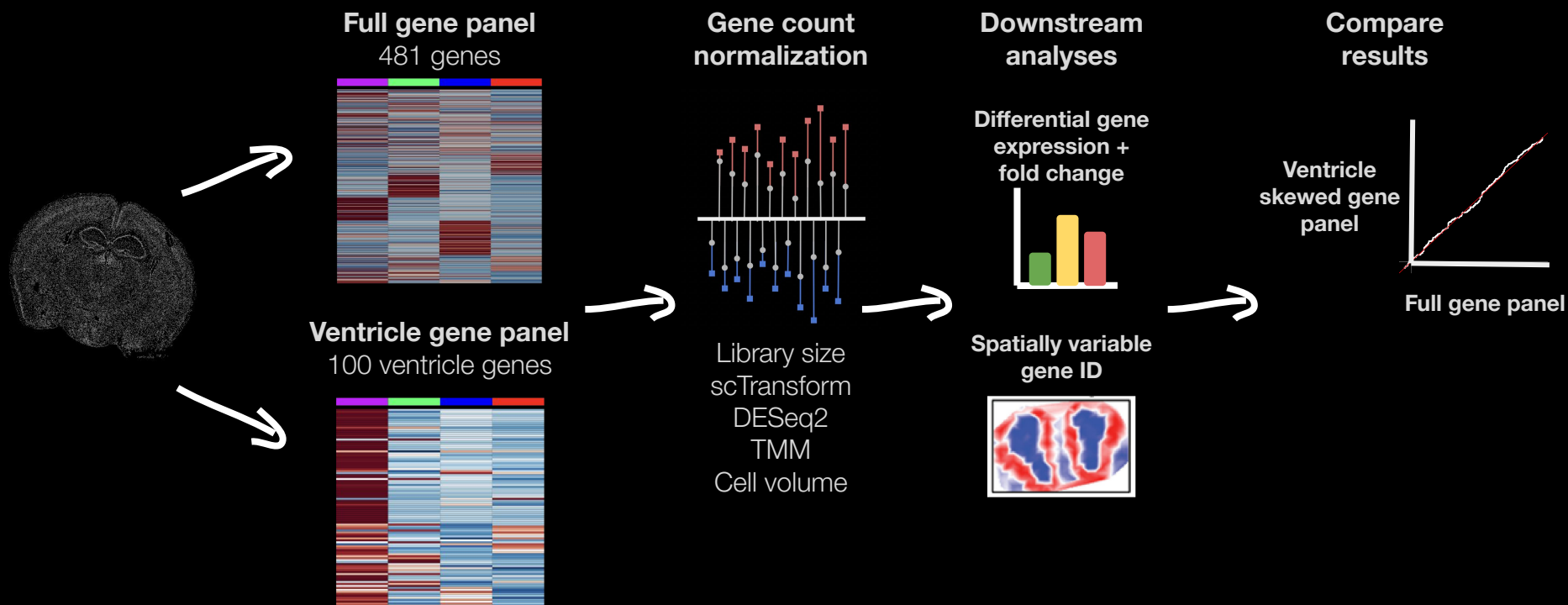
Robustness of downstream analyses with different gene panels



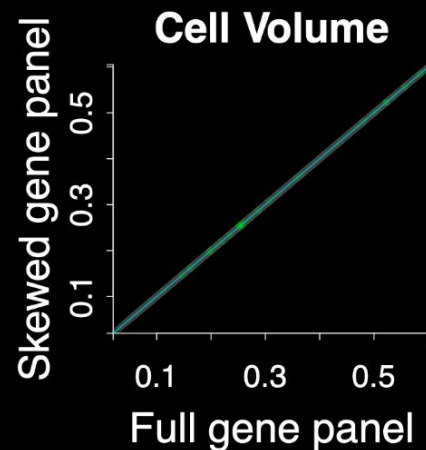
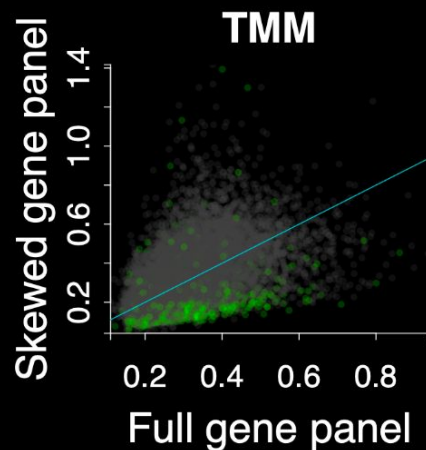
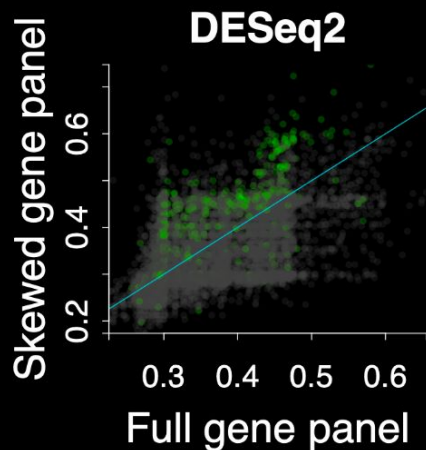
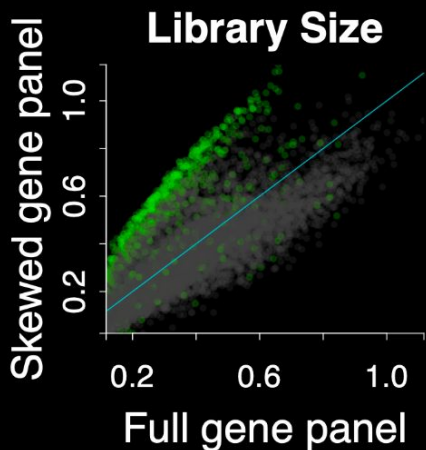
Count normalization in imaging-based SRT



Comparing normalization methods in mouse brain MERFISH with different gene panels

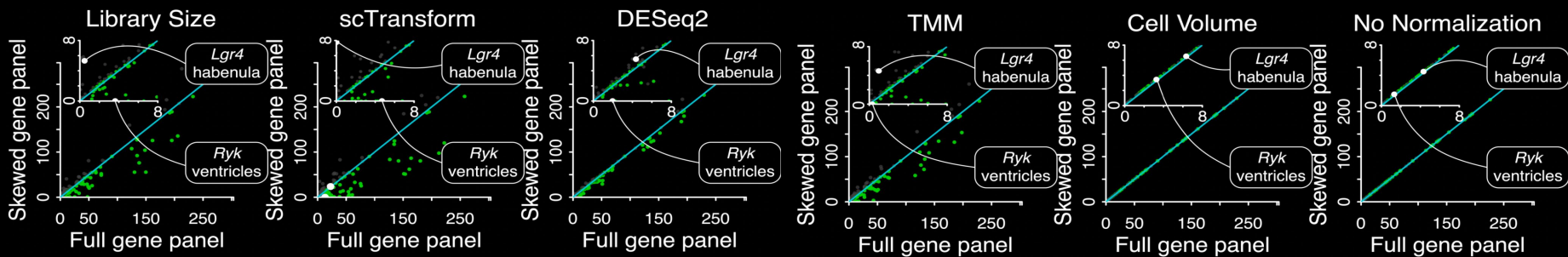


Normalizing scaling factors: tissue region specific bias with ventricle gene panel



COUNT BASED NORMALIZATION

Differential gene expression: tissue region specific bias with ventricle gene panel



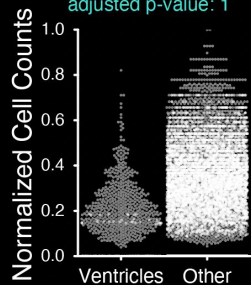
COUNT BASED NORMALIZATION

Differential gene expression: tissue region specific bias with ventricle gene panel results in false positives and negatives

Ryk: DE false negative

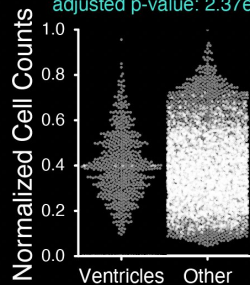
Skewed gene panel

$\log_2(\text{fold change})$: -1.03
adjusted p-value: 1



Full gene panel

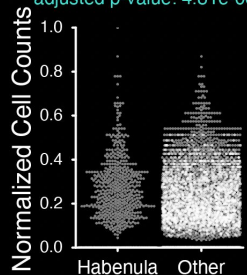
$\log_2(\text{fold change})$: 0.18
adjusted p-value: 2.37e-04



Lgr4: DE false positive

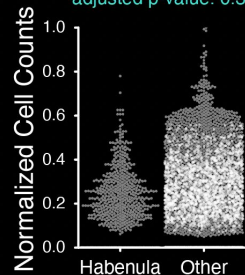
Skewed gene panel

$\log_2(\text{fold change})$: 0.48
adjusted p-value: 4.81e-06

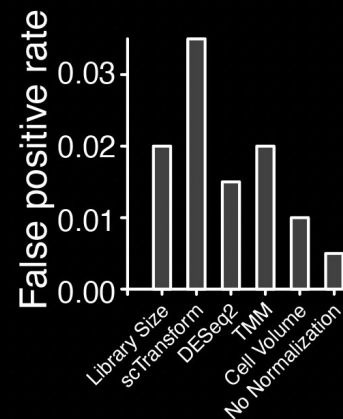


Full gene panel

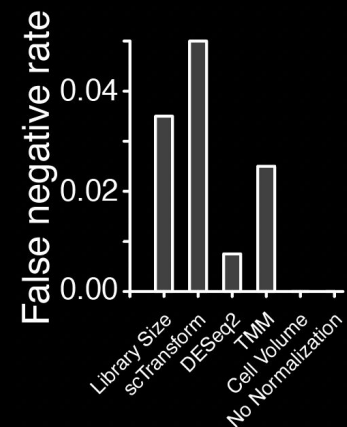
$\log_2(\text{fold change})$: -0.26
adjusted p-value: 0.33



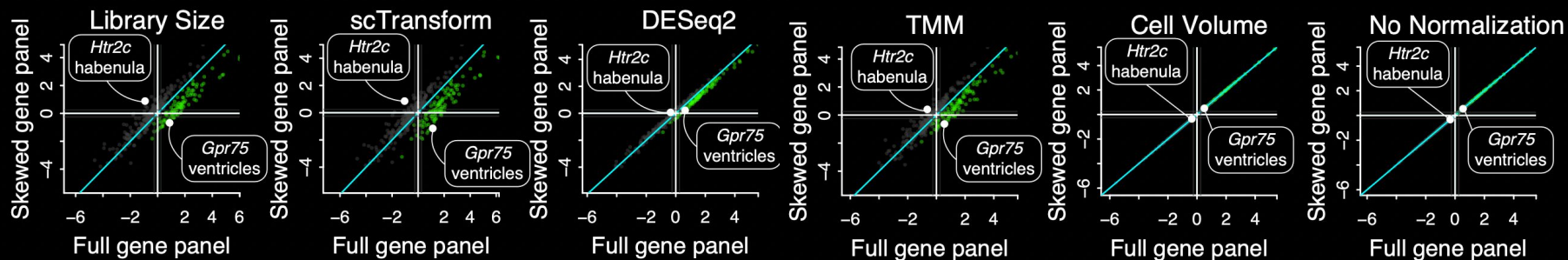
DE false negative rate



DE false positive rate

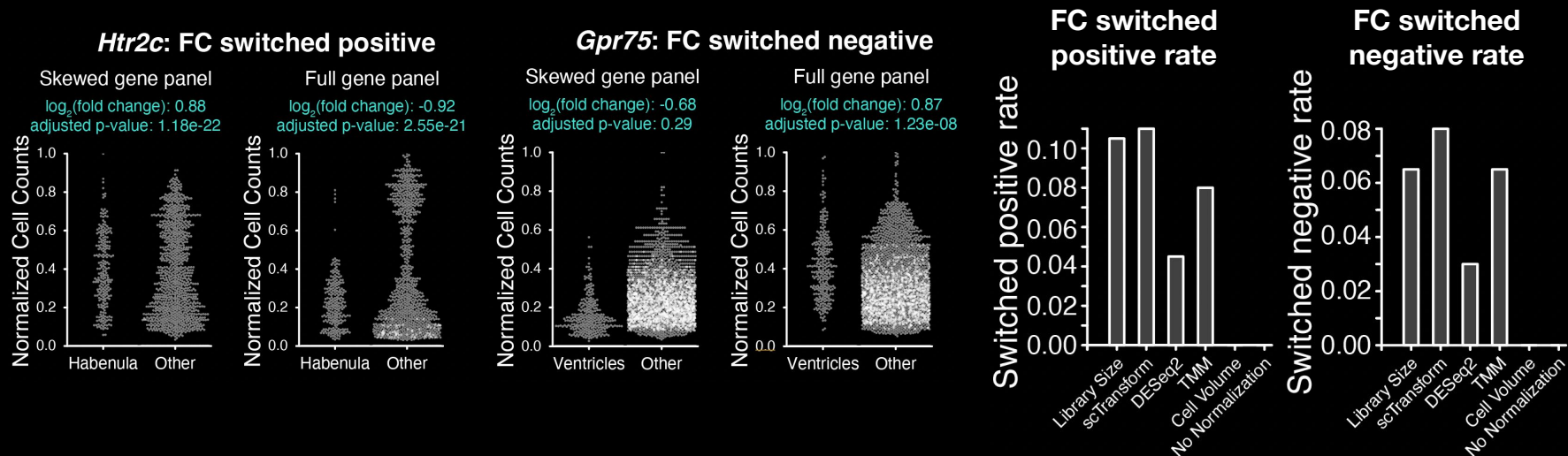


Differential gene expression: tissue region specific bias with ventricle gene panel

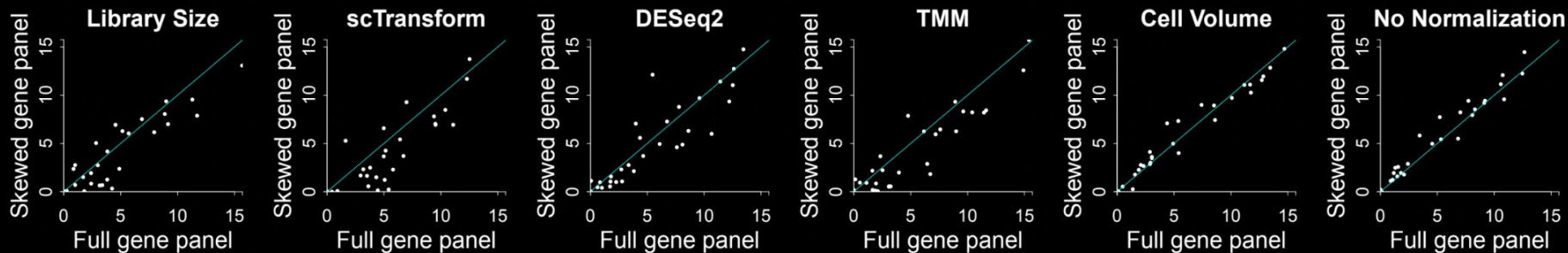


COUNT BASED NORMALIZATION

Differential gene expression: tissue region specific bias with ventricle gene panel results in switched gene fold changes



Spatially variable gene identification with ventricle gene panel results in false positives and negatives

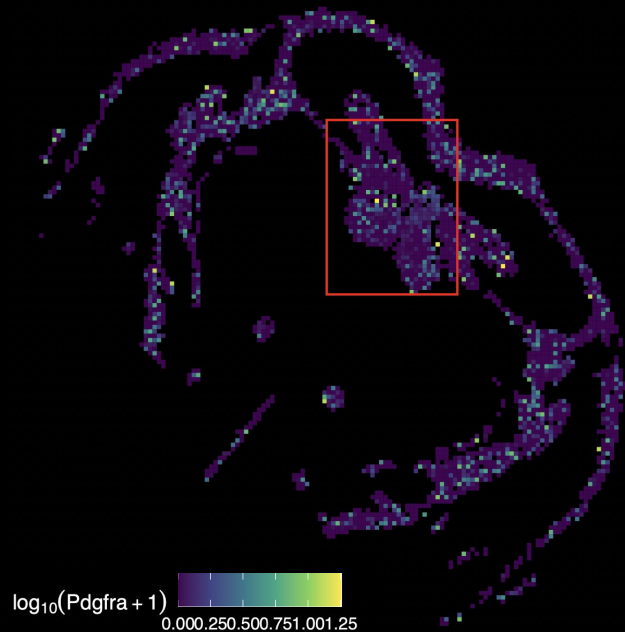


COUNT BASED NORMALIZATION

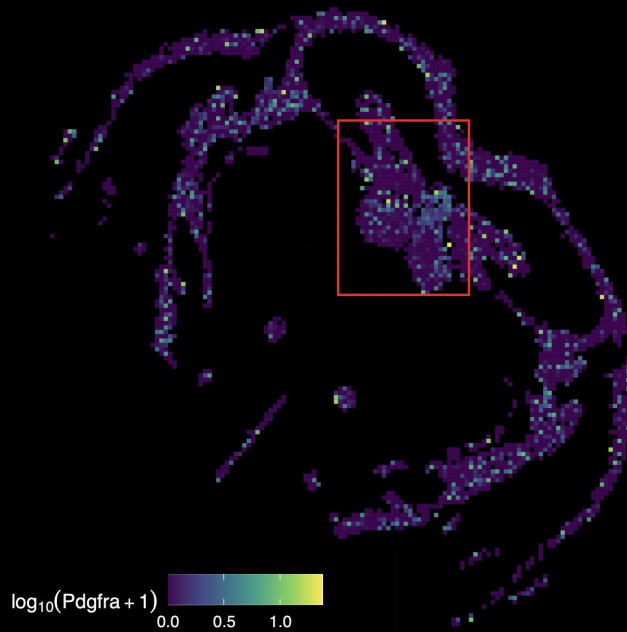
Spatially variable gene identification with ventricle gene panel results in false positives and negatives

Pdgfra, library size normalization

Ventricle gene panel



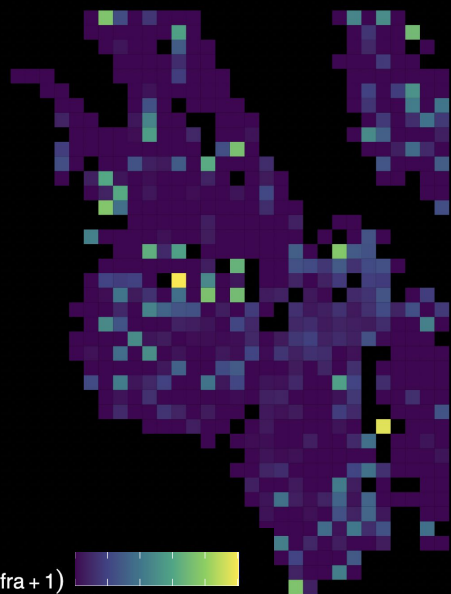
Full gene panel



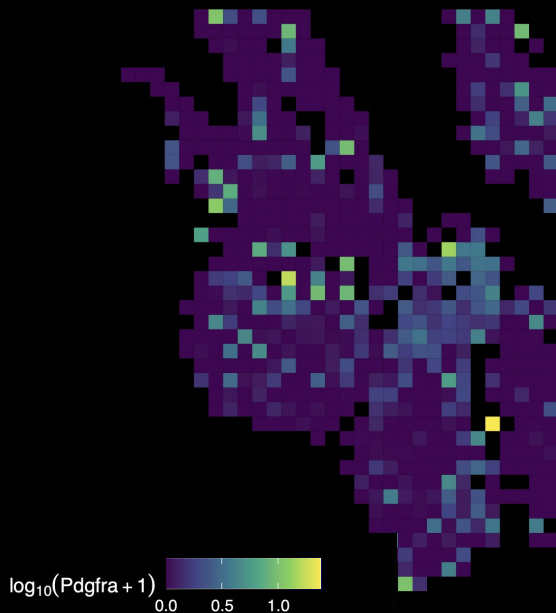
Spatially variable gene identification with ventricle gene panel results in false positives and negatives

Pdgfra, library size normalization

Ventricle gene panel



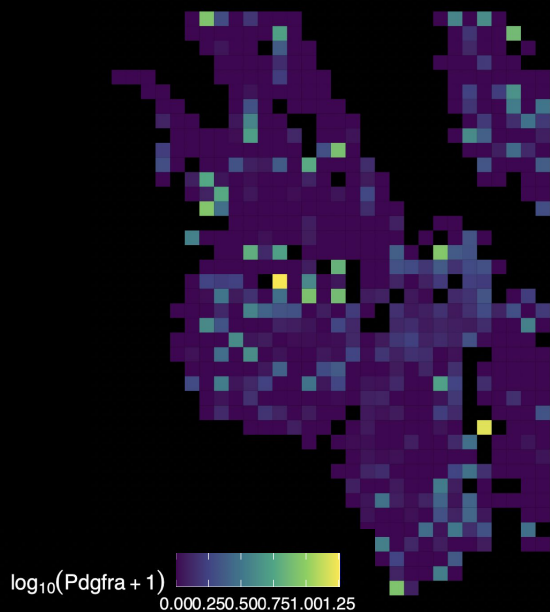
Full gene panel



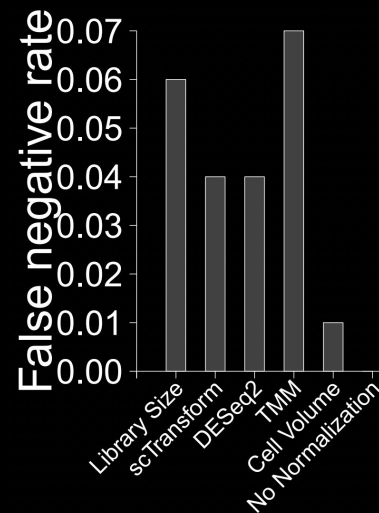
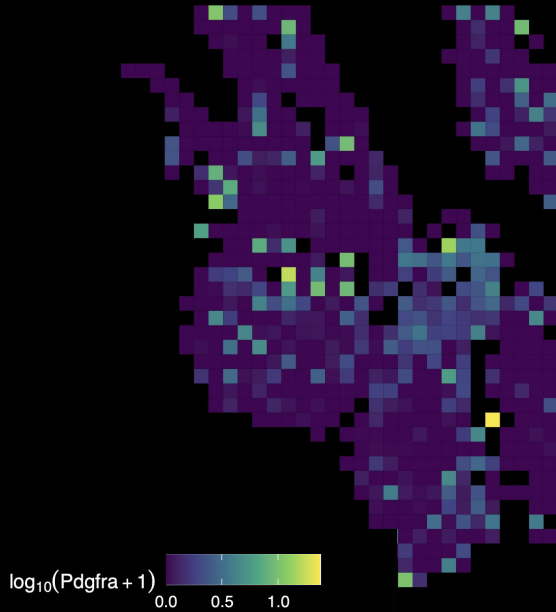
Spatially variable gene identification with ventricle gene panel results in false positives and negatives

Pdgfra, library size normalization

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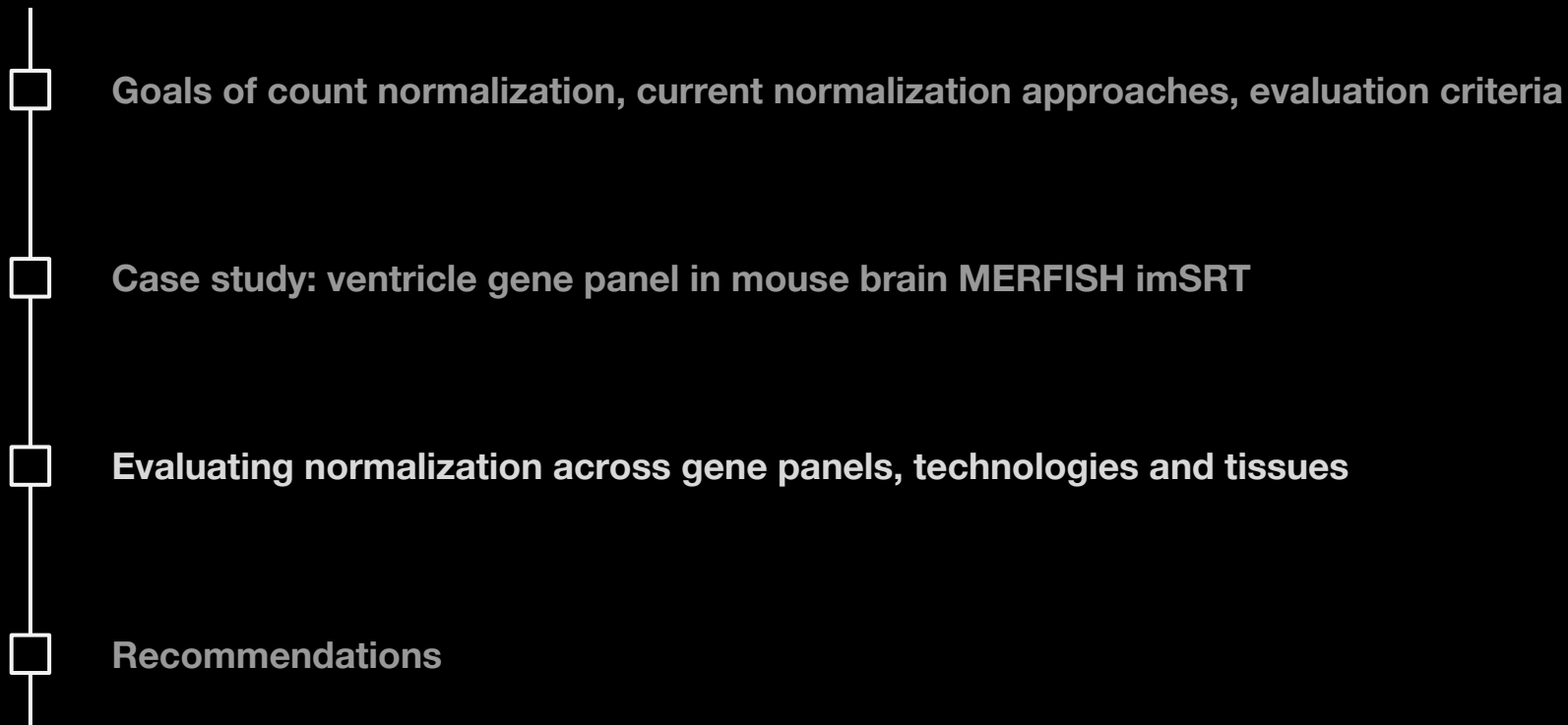


Full gene panel

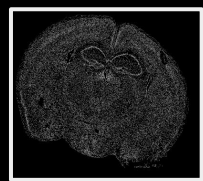


Normalization

Count normalization in imaging-based SRT



Region-specific biases in normalized gene expression generalize across gene panels



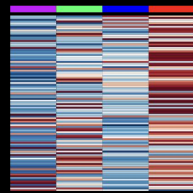
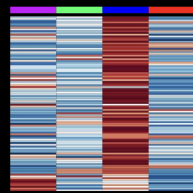
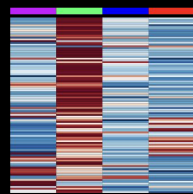
Habenula panel



Fiber tracts panel

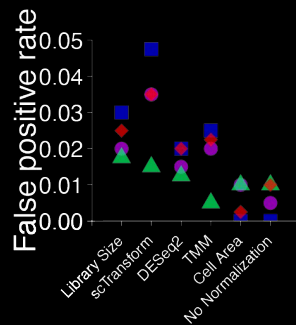


Dentate gyrus panel

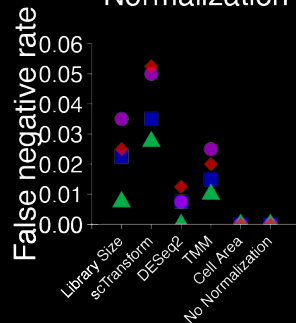


Ventricle panel

DE error rates

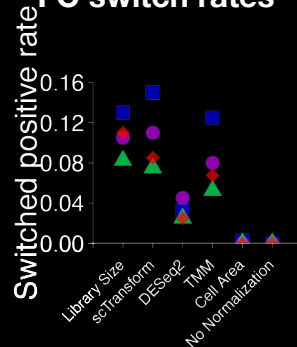


Normalization

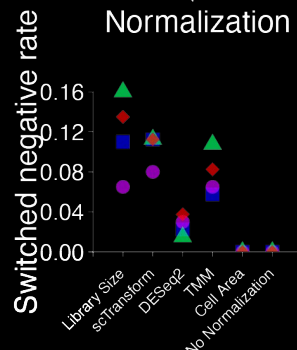


Normalization

FC switch rates

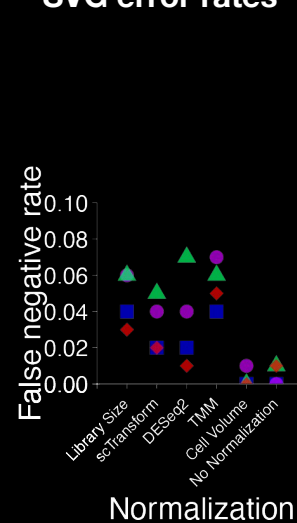


Normalization



Normalization

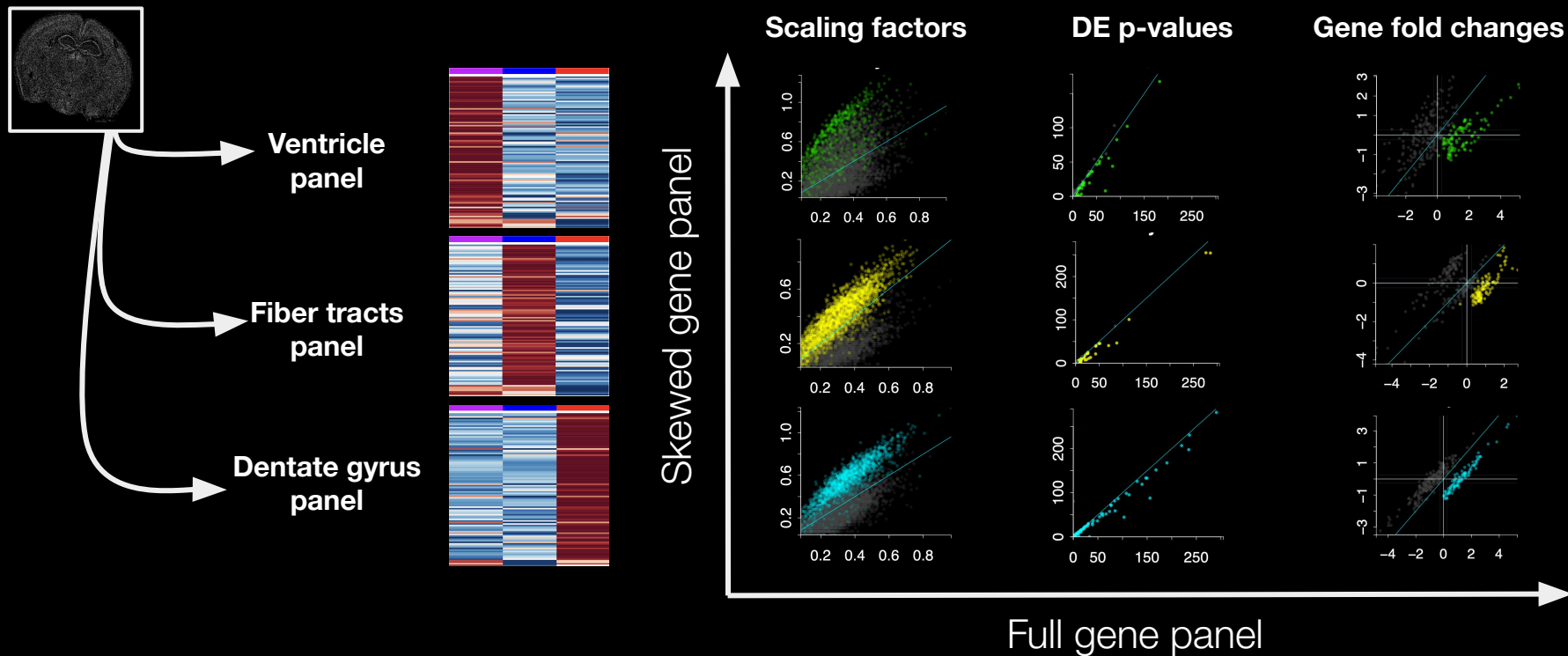
SVG error rates



Normalization

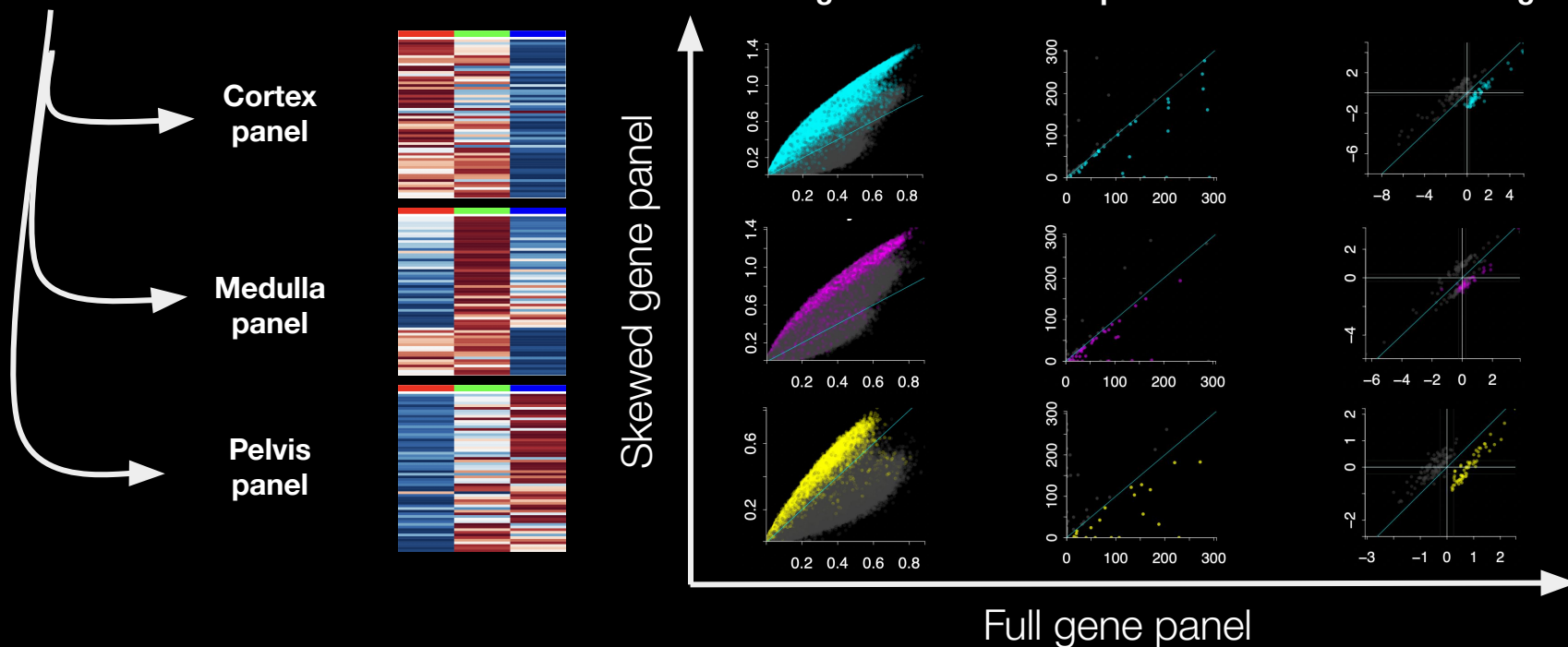
Region-specific biases in normalized gene expression generalize across imSRT technologies

Library size normalization with STARmapPLUS in mouse brain



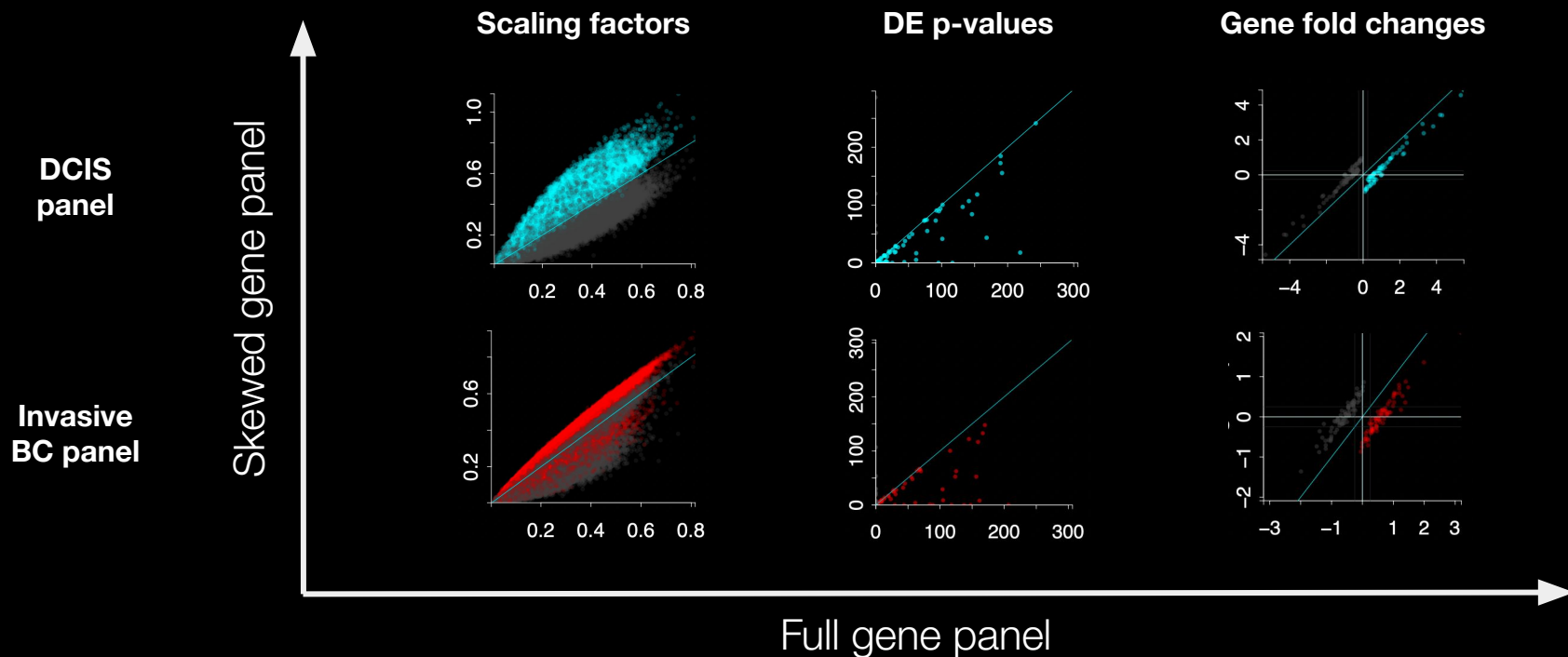
Region-specific biases in normalized gene expression generalize across tissue types

Library size normalization with seqFISH in mouse kidney



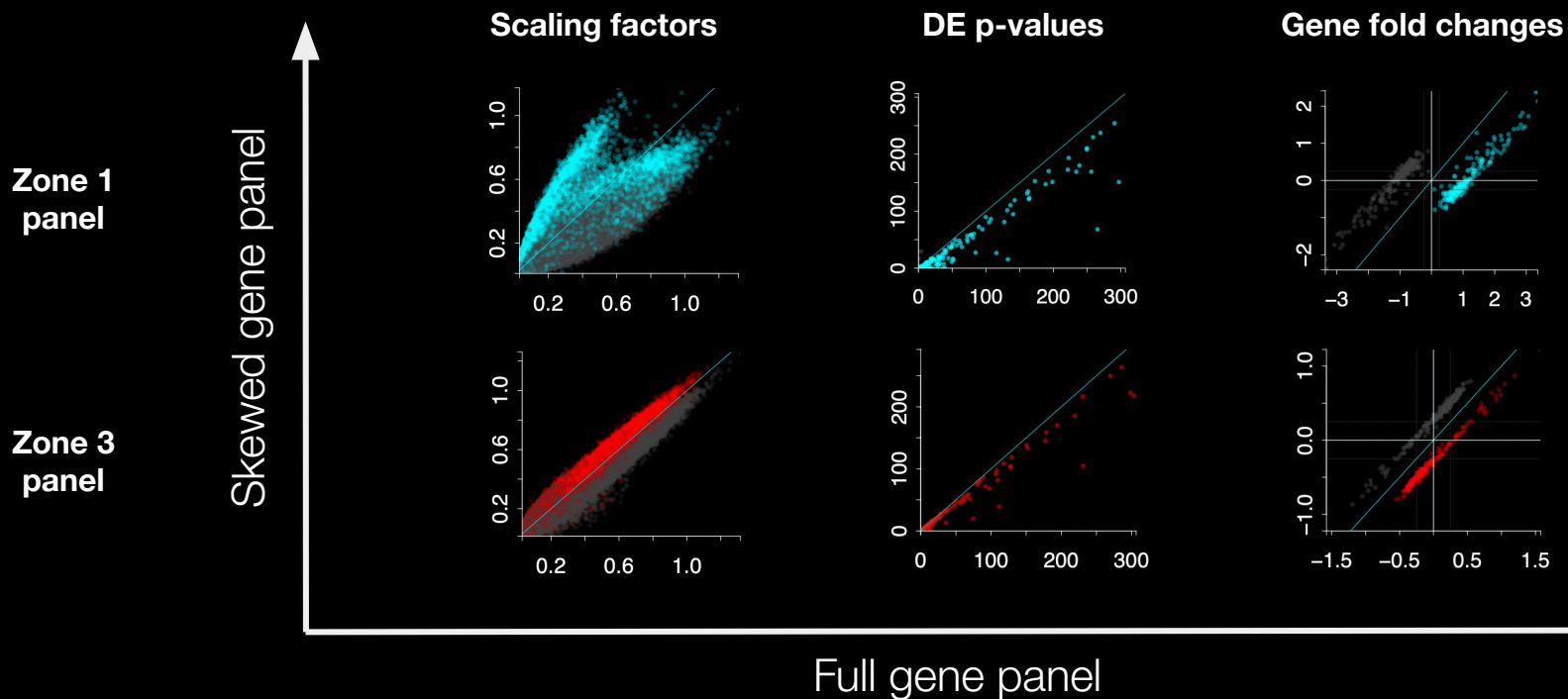
Region-specific biases in normalized gene expression generalize across tissue types

Library size normalization with 10X Xenium in human breast cancer

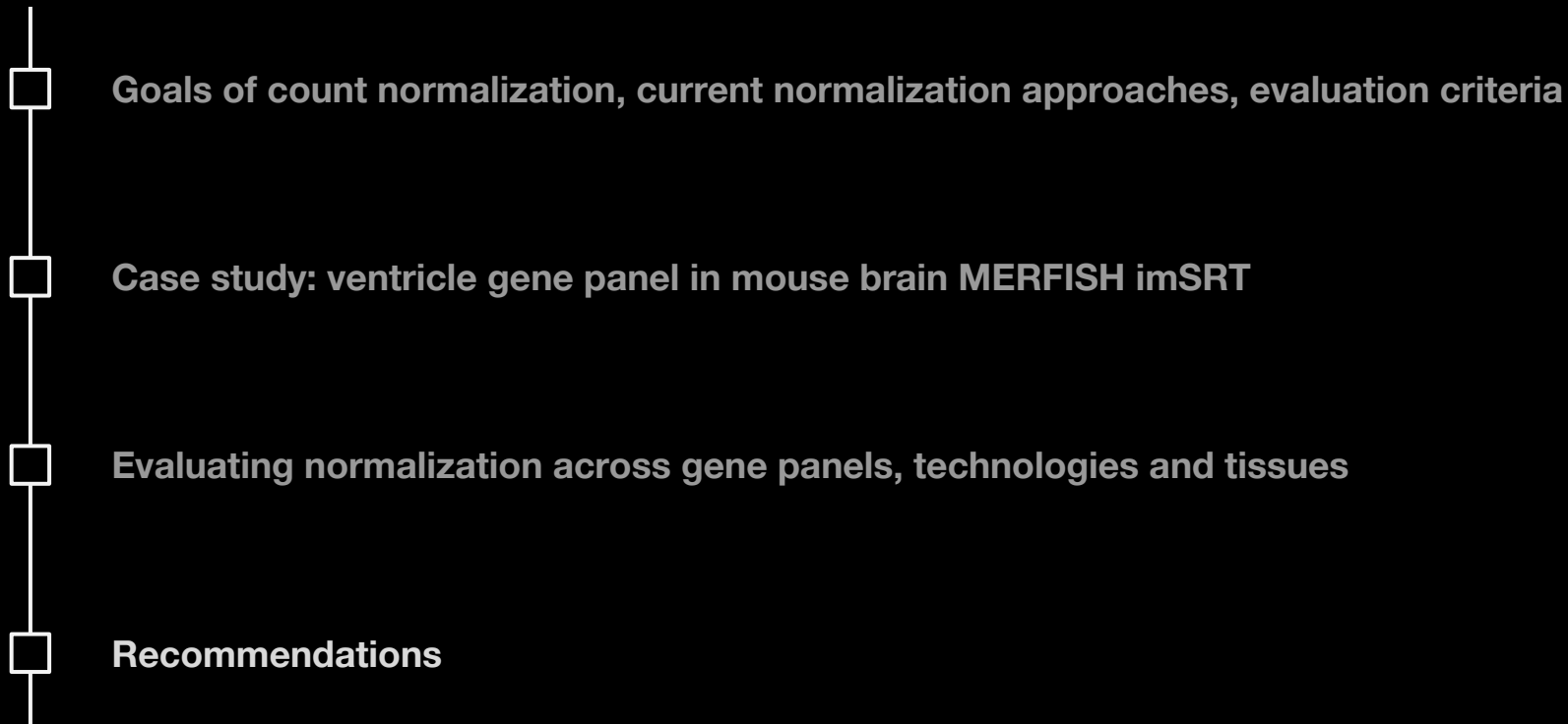


Region-specific biases in normalized gene expression generalize across tissue types

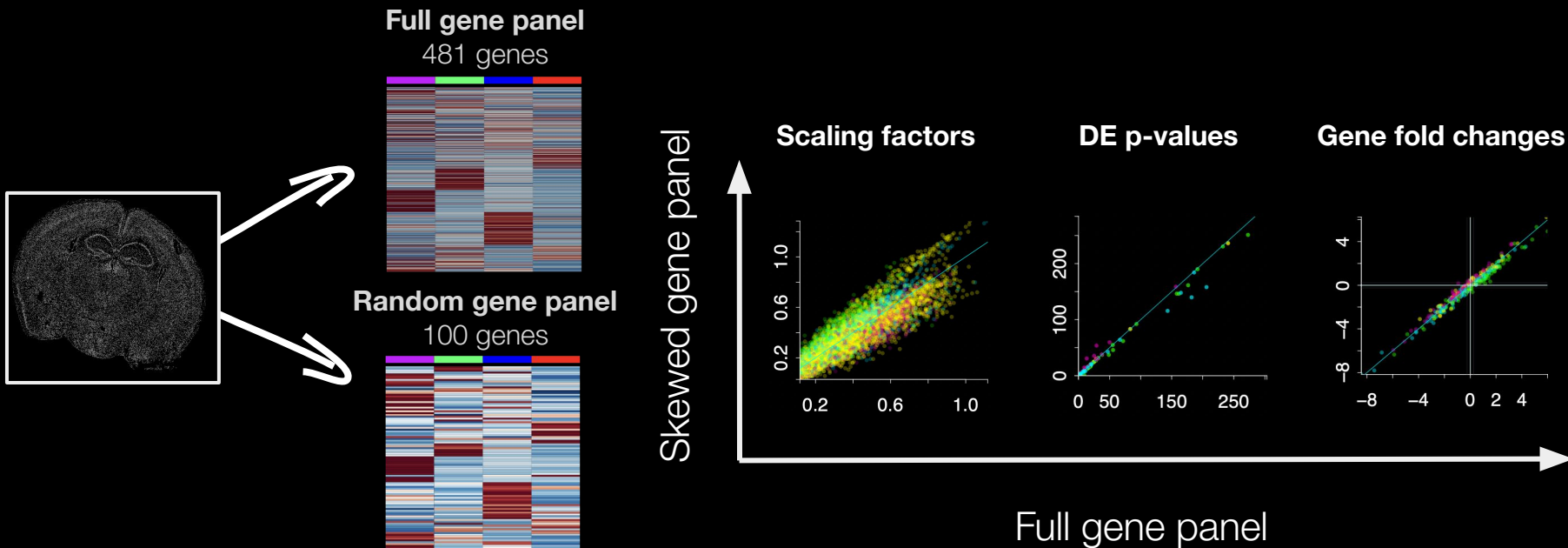
Library size normalization with CosMx in human liver



Count normalization in imaging-based SRT

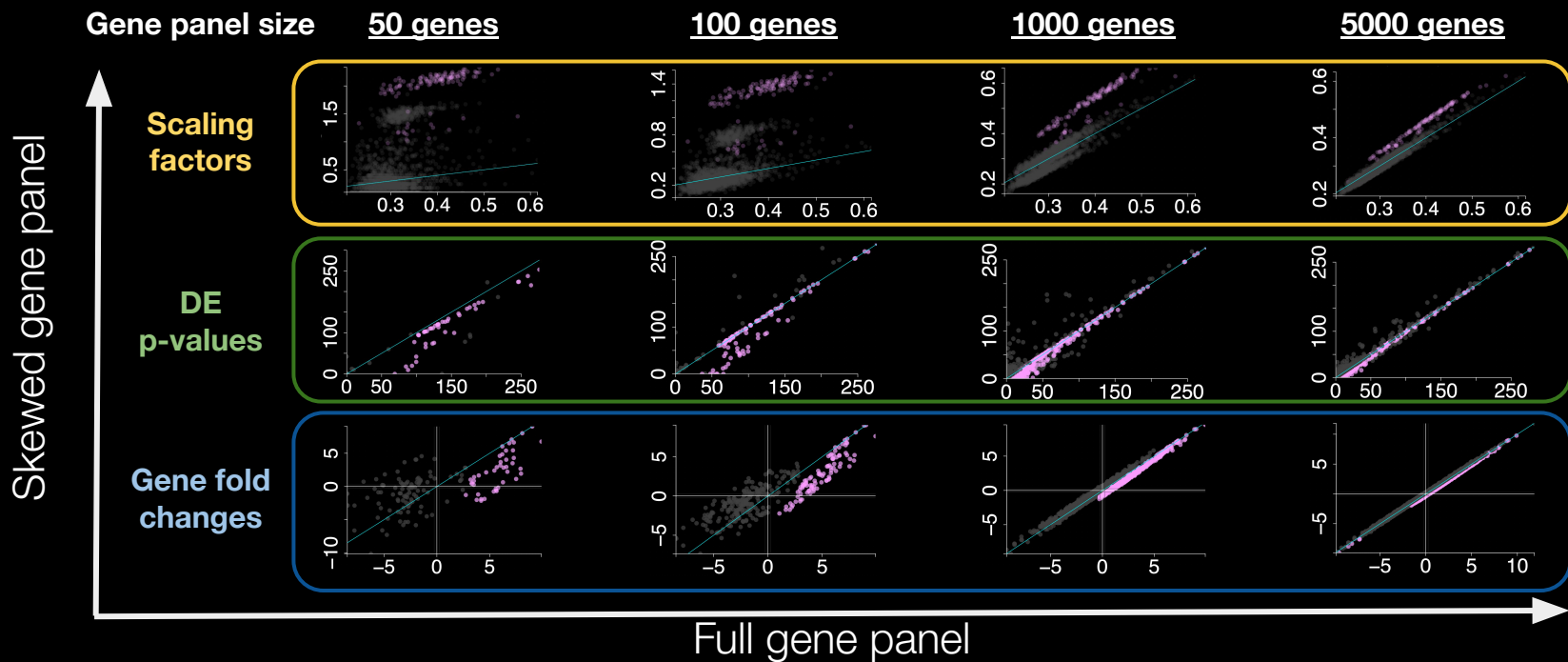


Region-specific biases can be mitigated with more representative gene panels

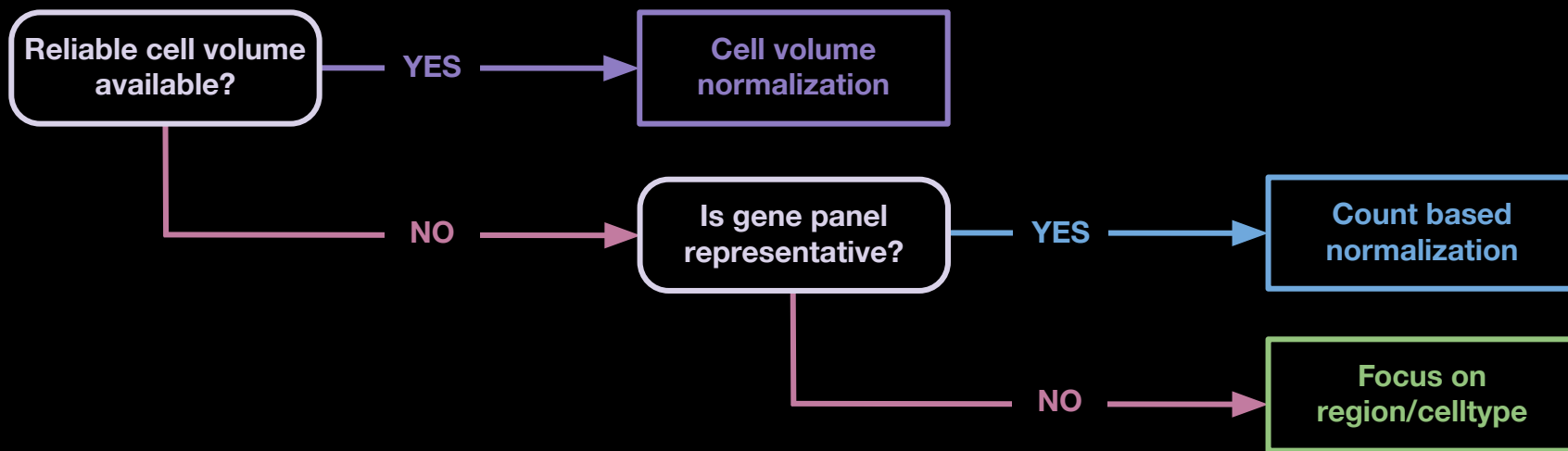


Region-specific biases can be mitigated with larger gene panels

Library size normalization with monocyte-skewed gene panels simulated from sorted PBMS scRNA-seq



Recommendations



Gene count normalization in single-cell imaging-based spatially resolved transcriptomics



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Dee Velazquez

Vivien Jiang

Gene count normalization in single-cell imaging-based spatially resolved transcriptomics

 Lyla Atta,  Kalen Clifton,  Manjari Anant,  Jean Fan

doi: <https://doi.org/10.1101/2023.08.30.555624>

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bit.ly/jefworks-imSRT-norm