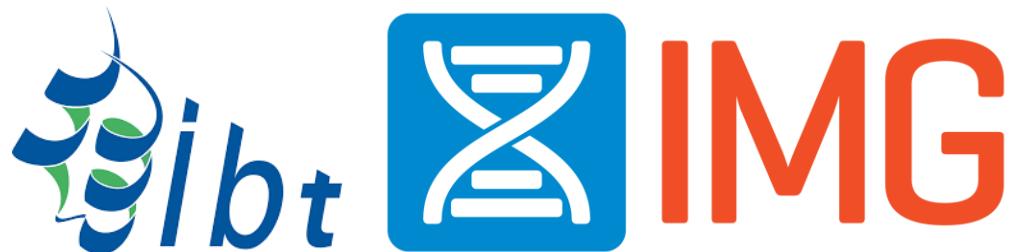
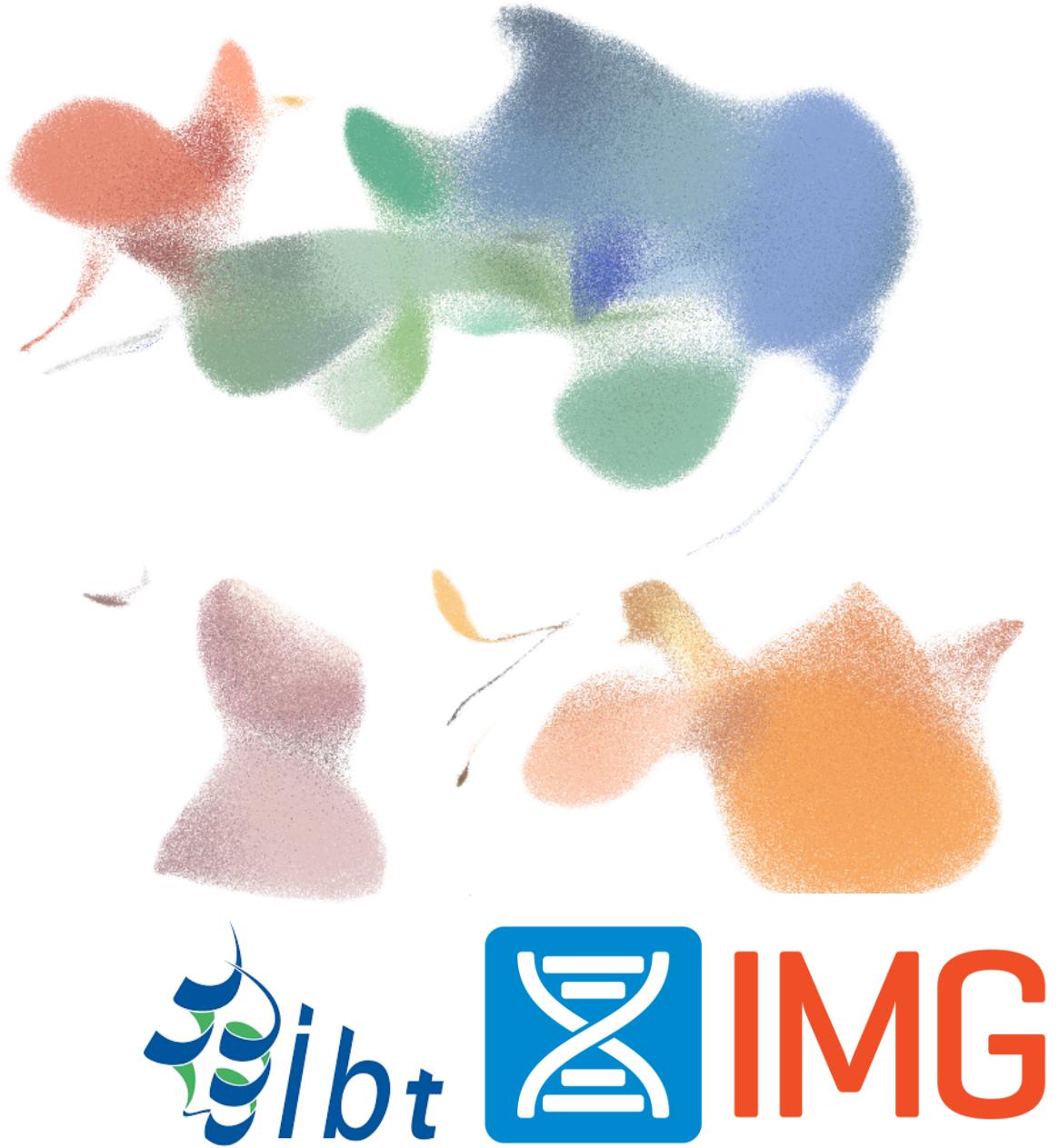


Data Integration

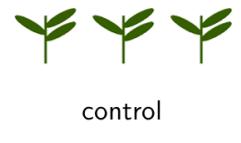
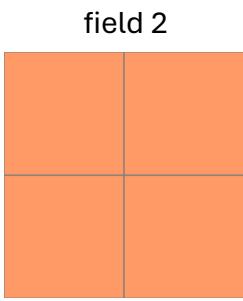
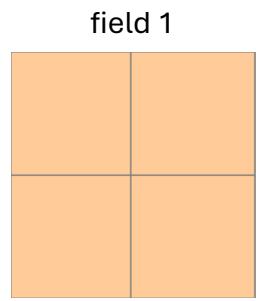
Yusuf Caglar Odabasi

December 1.-3. 2025

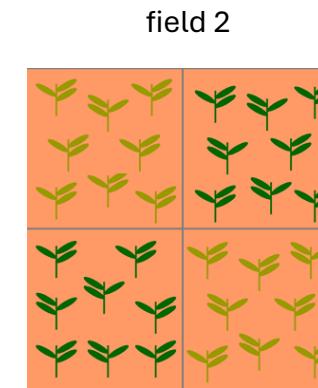
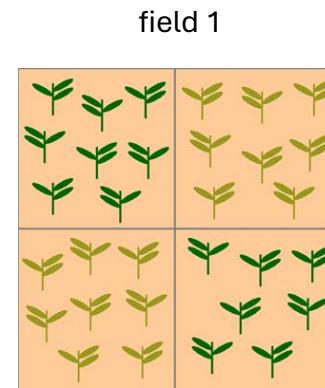
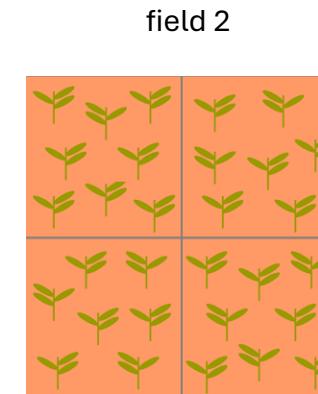
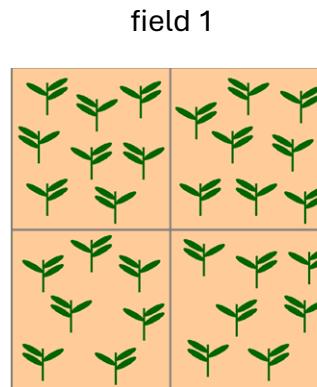
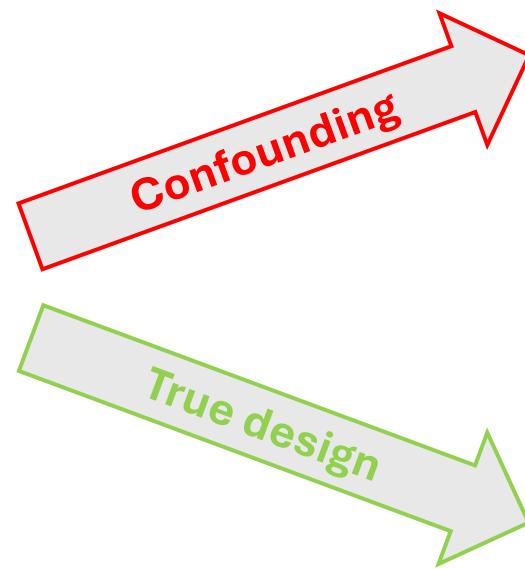
Course on scRNA-seq Data Analysis



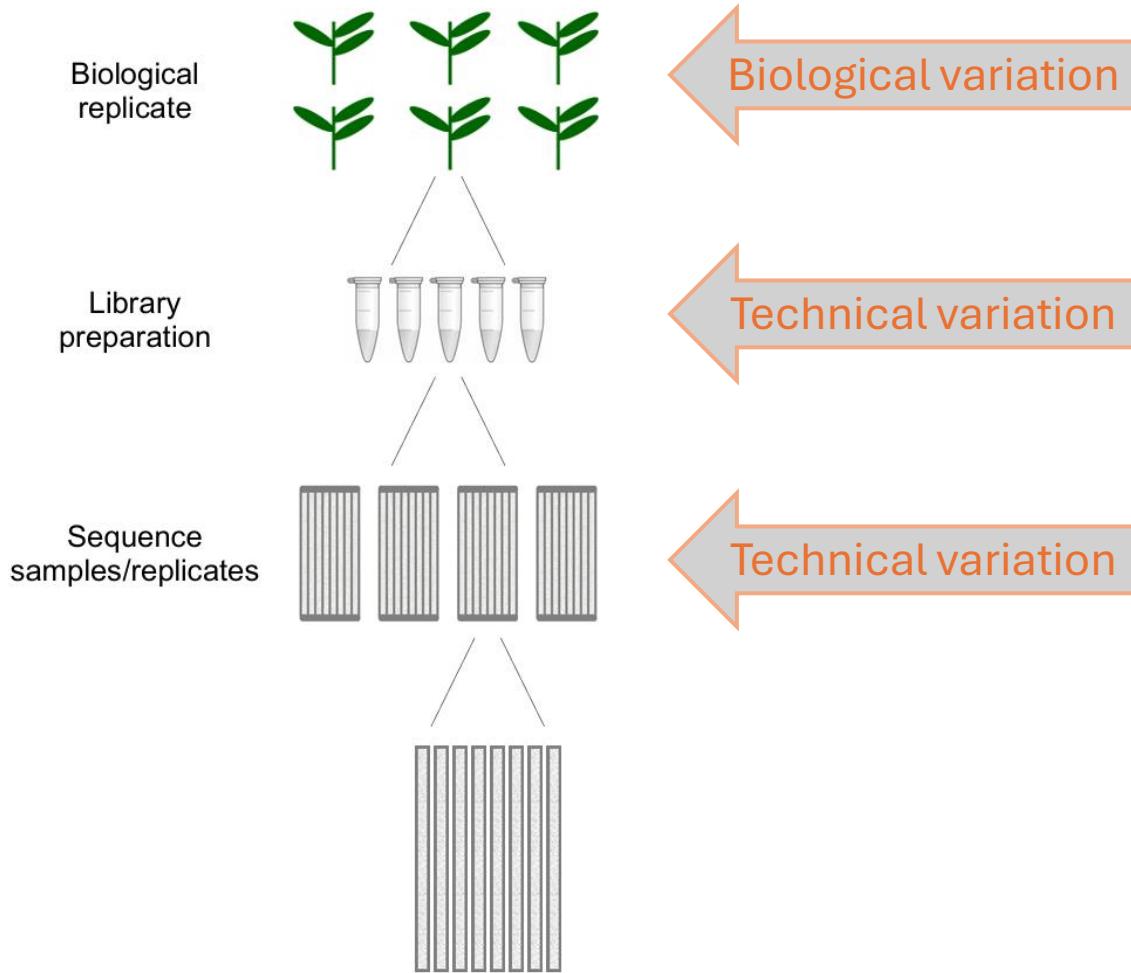
Experiment Design: Confounders and Batch effects



chemical injection



Experiment Design: Confounders and Batch effects



1. Technical variability

- Changes in sample quality/processing
- Library prep or sequencing technology

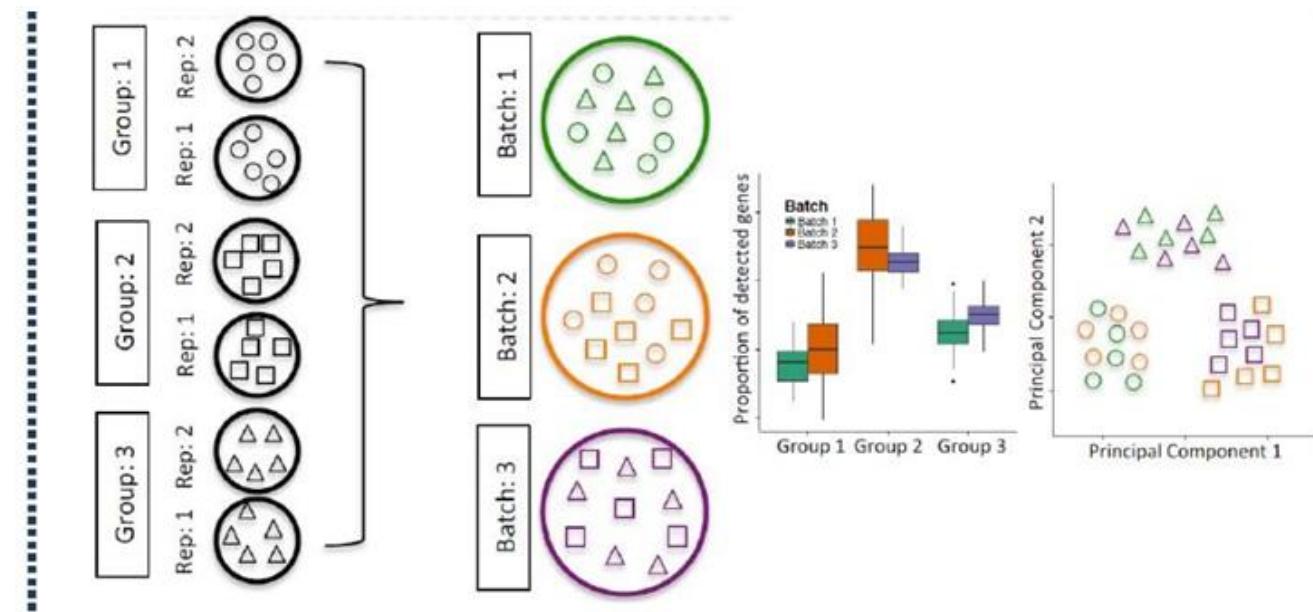
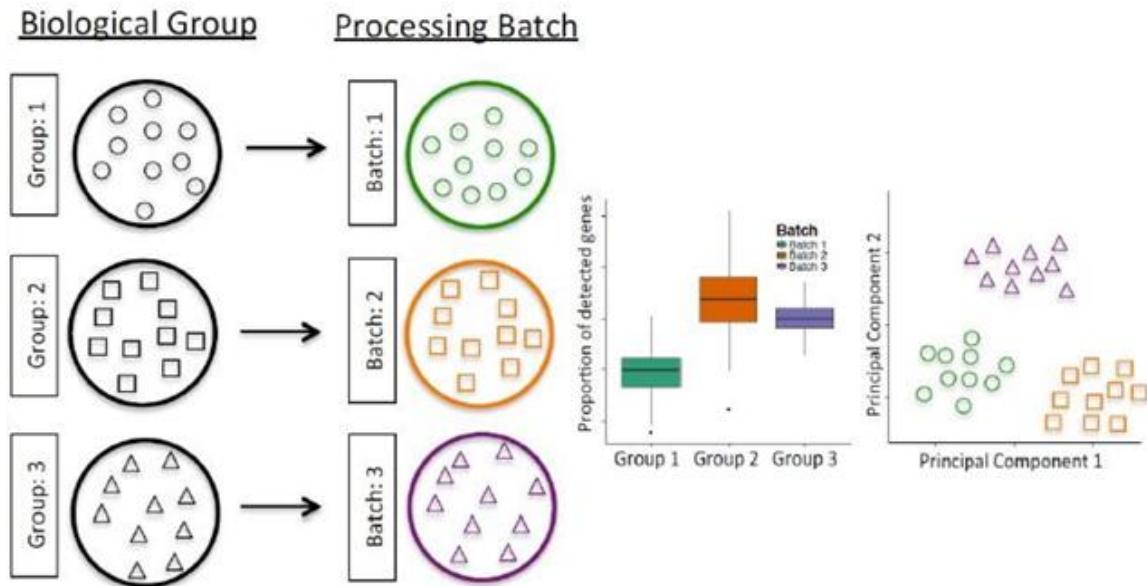
Technical 'batch effects' confound downstream analysis

2. Biological variability

- Patient differences
- Evolution! (cross-species analysis)

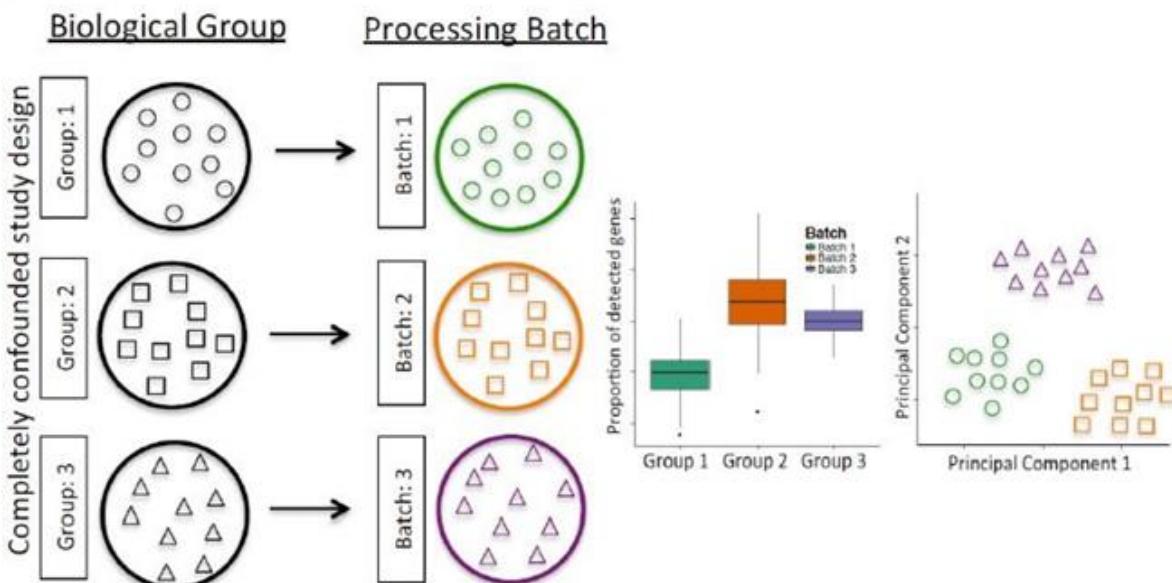
Biological 'batch effects' confound comparisons of scRNA-seq data

Experiment Design: Confounders and Batch effects

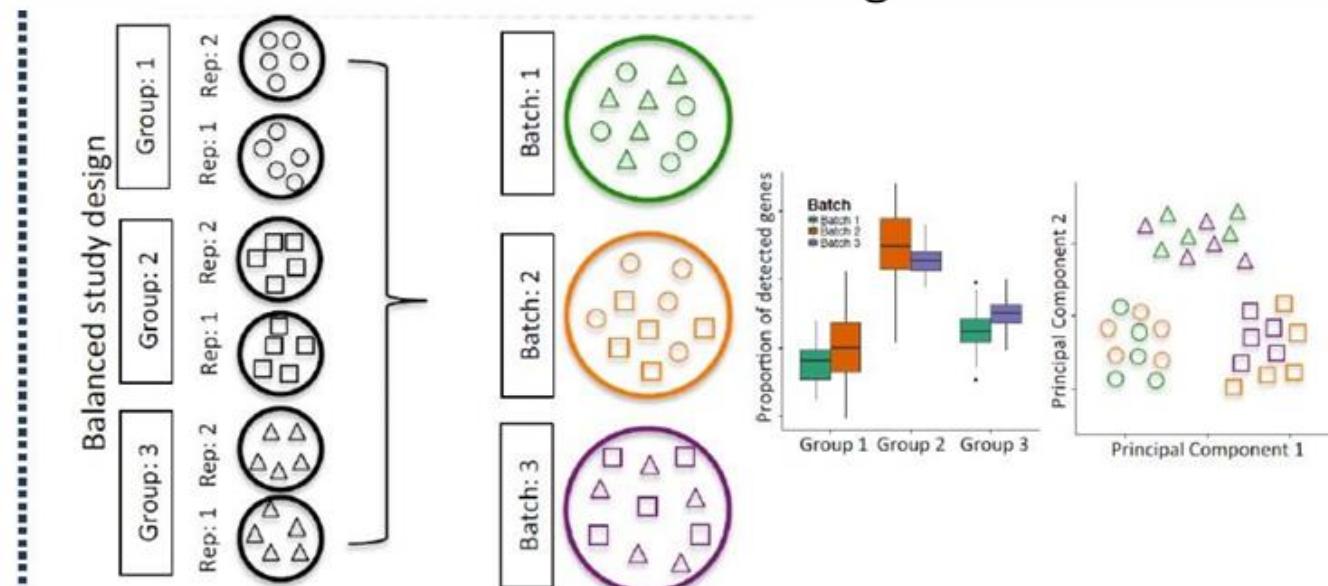


Experiment Design: Confounders and Batch effects

Confounded design

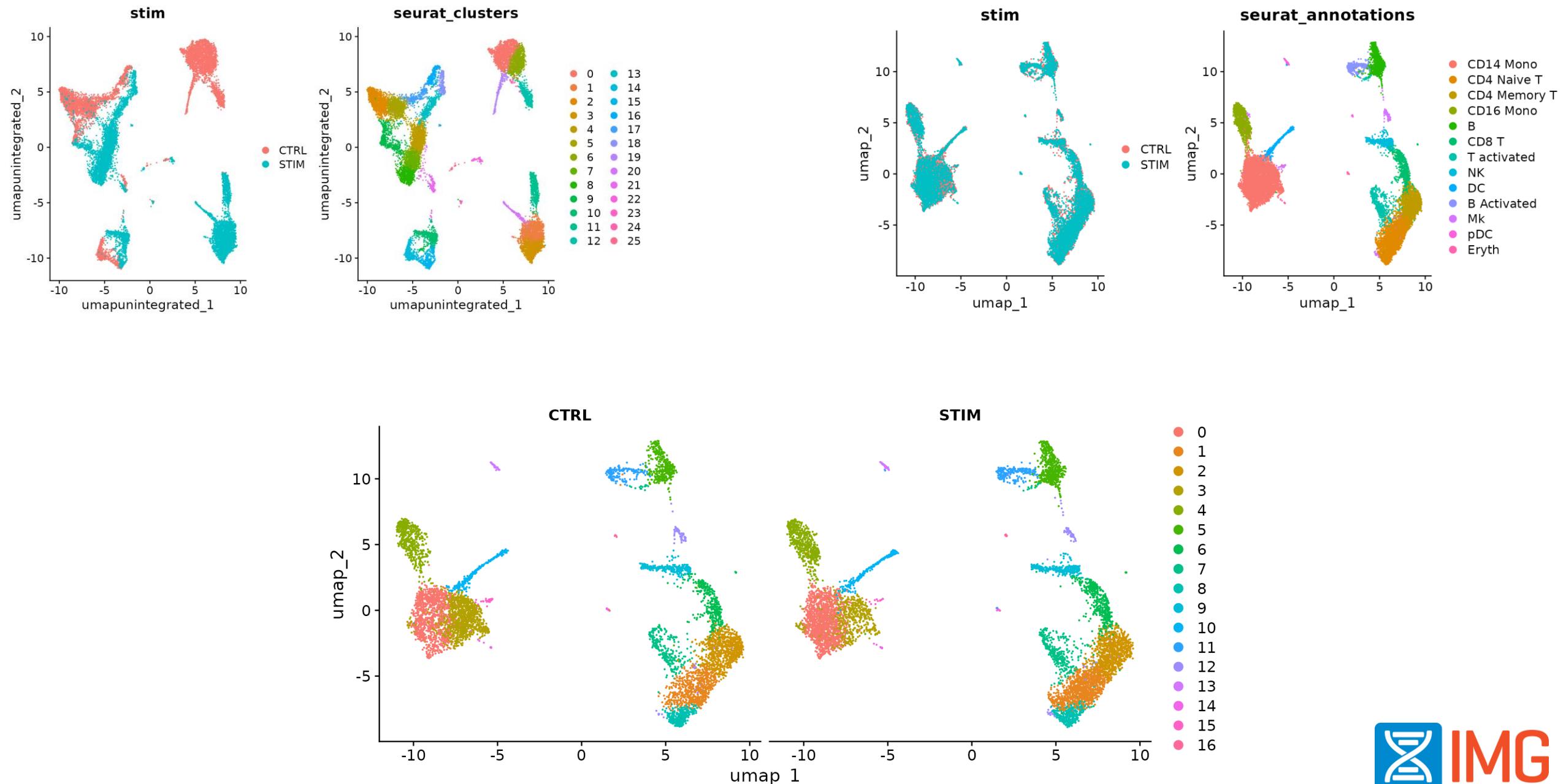


Not confounded design



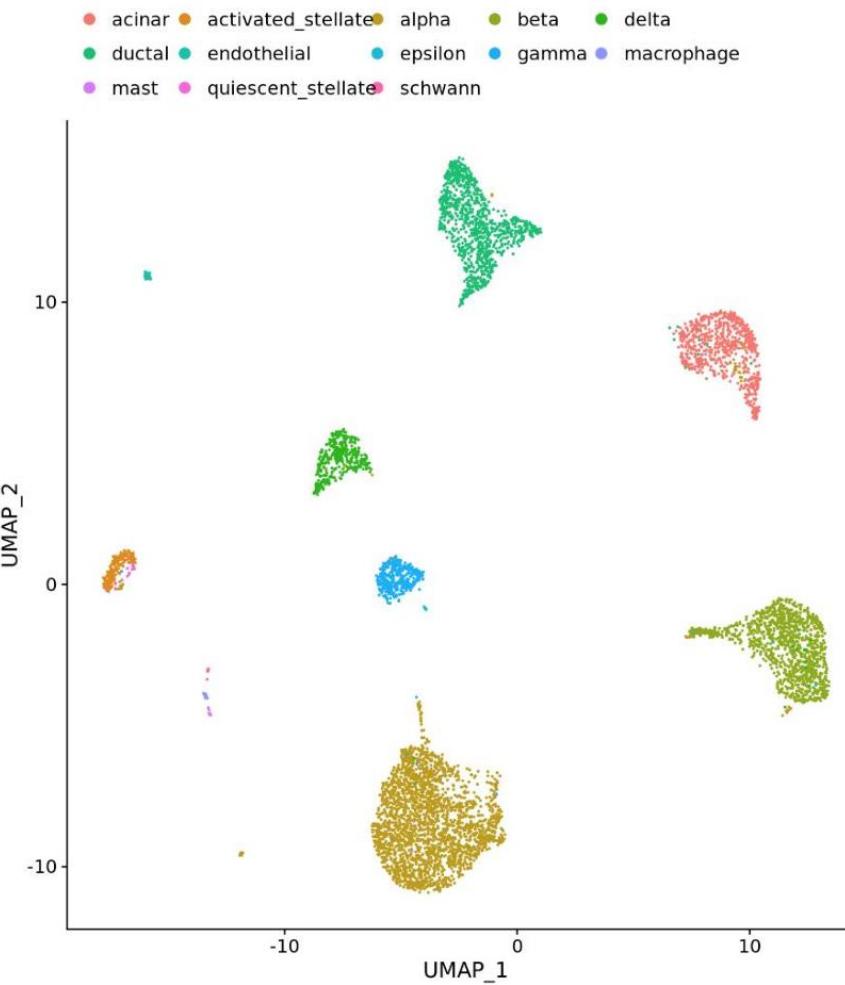
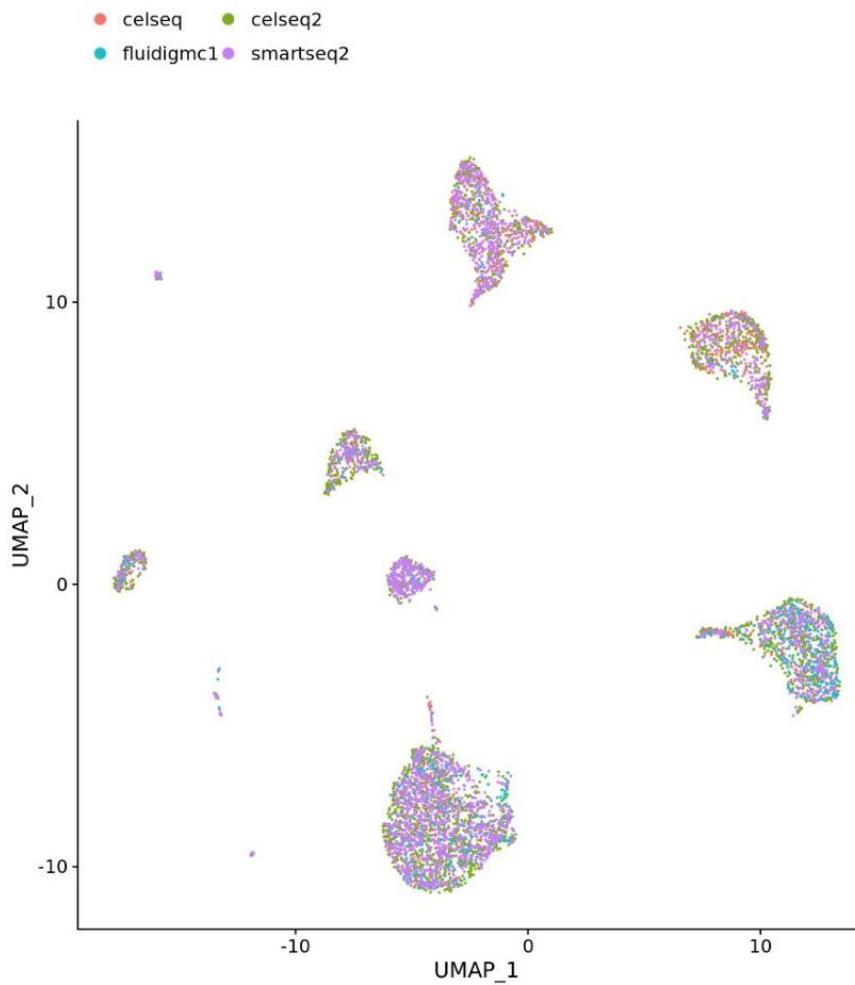
Good experimental design *does not remove batch effects*, it prevents them from biasing your results.

Why do we need integration?



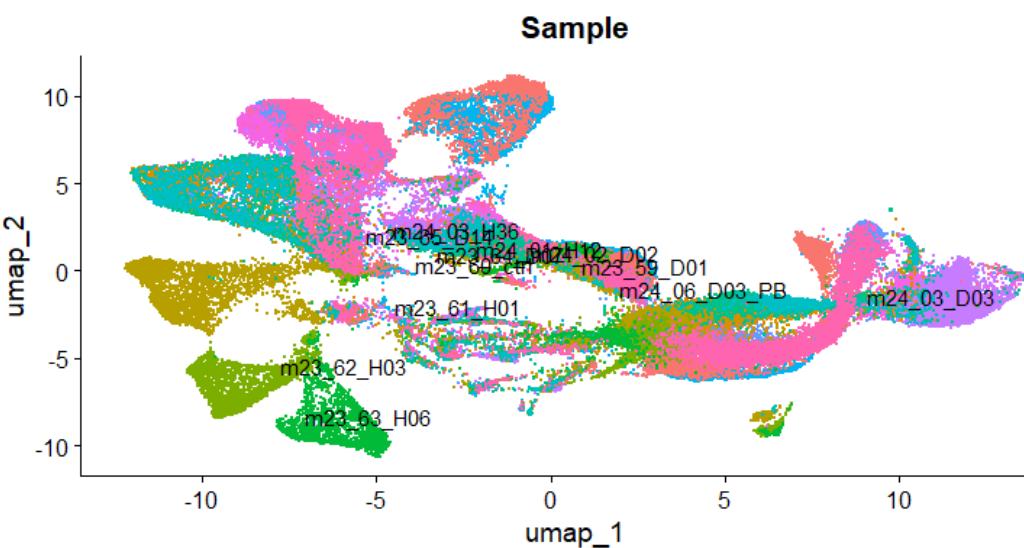
Why do we need integration?

Example scenarios for integration: datasets

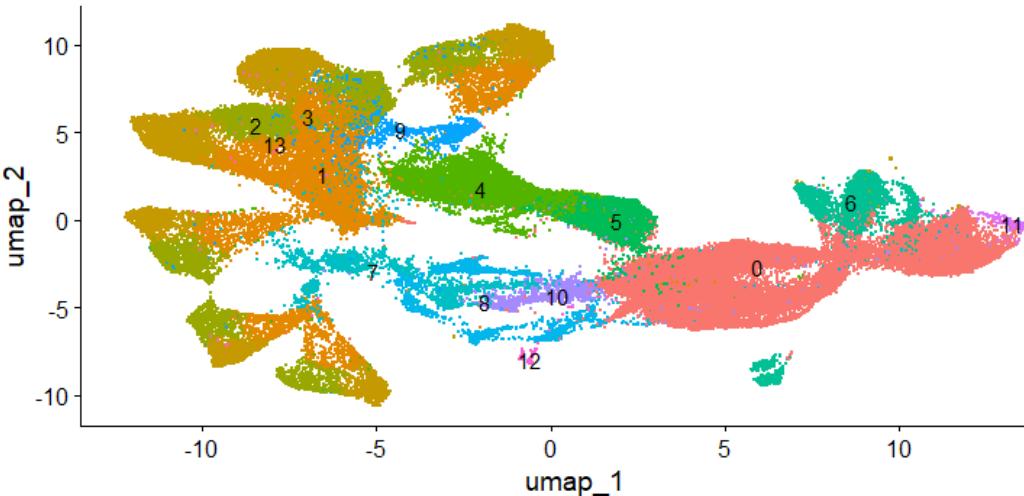


Why do we need integration?

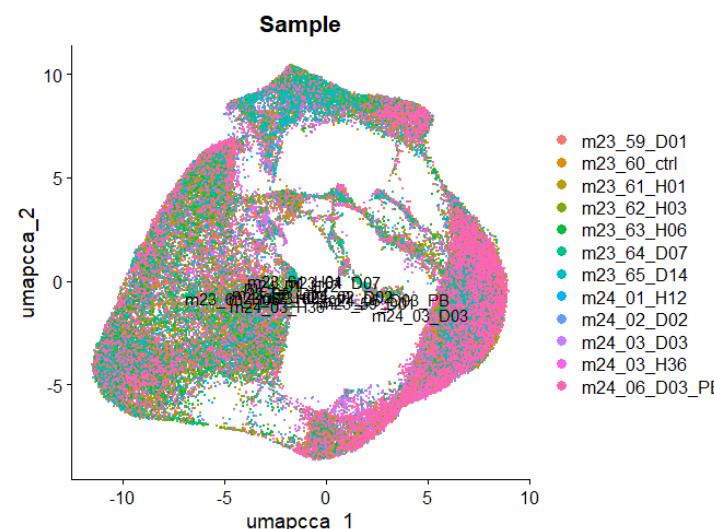
Before integration



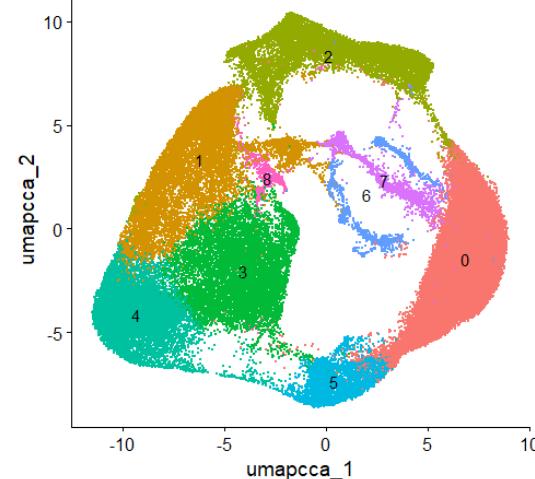
seurat_clusters



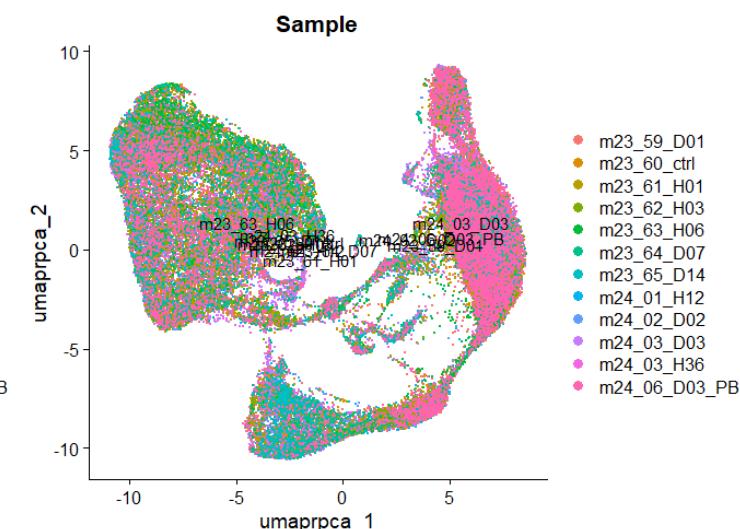
CCA integration



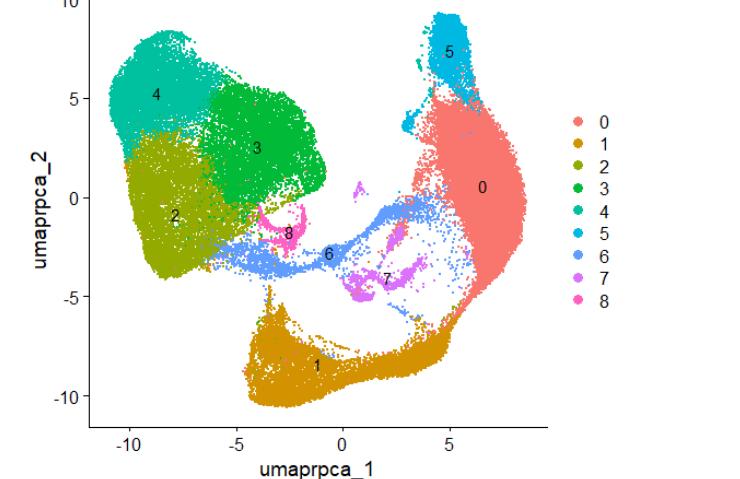
cca_clusters



RPCA integration



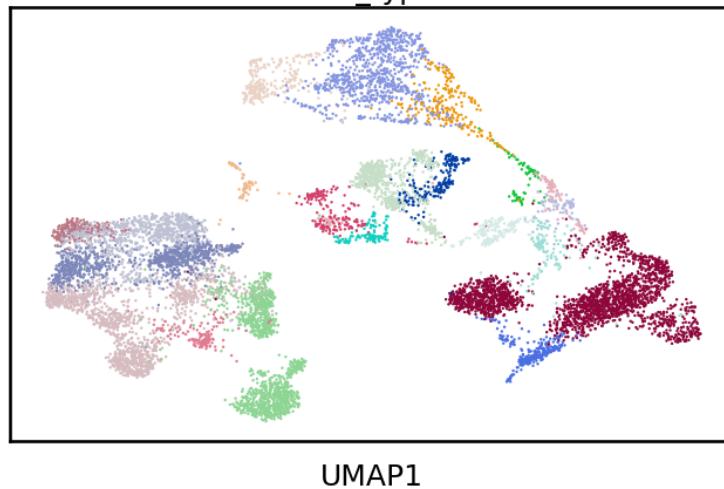
rpc_clusters



Why do we need integration?

UMAP2

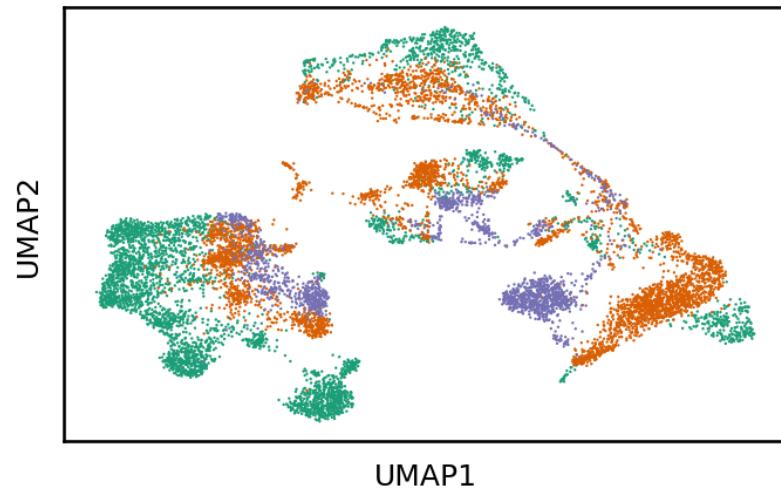
cell_type



- B1 B
- CD4+ T activated
- CD4+ T naive
- CD8+ T
- CD8+ T naive
- CD14+ Mono
- CD16+ Mono
- Erythroblast
- G/M prog
- HSC
- ILC

- Lymph prog
- MK/E prog
- NK
- Naive CD20+ B
- Normoblast
- Plasma cell
- Proerythroblast
- Transitional B
- cDC2
- pDC

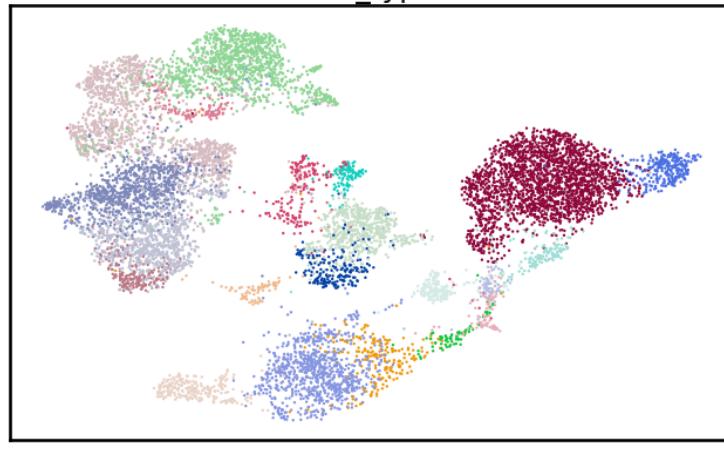
batch



- s1d3
- s2d1
- s3d7

UMAP2

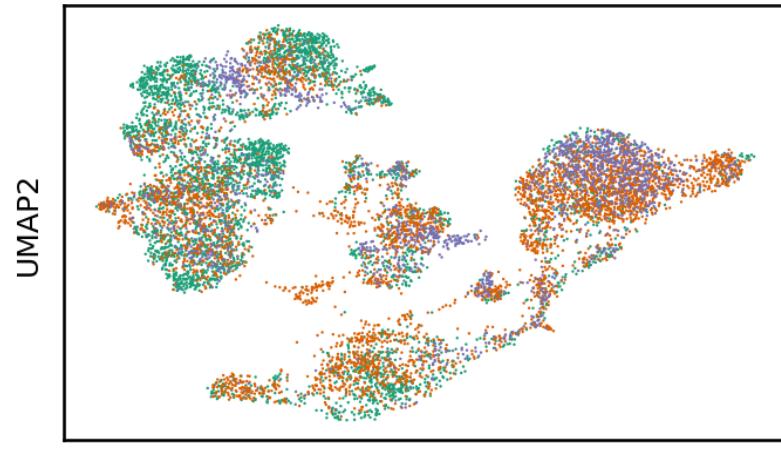
cell_type



- B1 B
- CD4+ T activated
- CD4+ T naive
- CD8+ T
- CD8+ T naive
- CD14+ Mono
- CD16+ Mono
- Erythroblast
- G/M prog
- HSC
- ILC

- Lymph prog
- MK/E prog
- NK
- Naive CD20+ B
- Normoblast
- Plasma cell
- Proerythroblast
- Transitional B
- cDC2
- pDC

batch



- s1d3
- s2d1
- s3d7

Types of integration models

Global models

- Fit regression model with batch effect covariate

Residuals (often using linear regression):

$$\hat{n}_{gc} = f_D(B_c, \dots)$$

$$r_{gc} = n_{gc} - \hat{n}_{gc} = n_{gc} - (\beta_0 + \beta_1 B_c)$$

in linear model case

Example:
`sc.tl.regress_out()`

Correct for fitted batch effect:

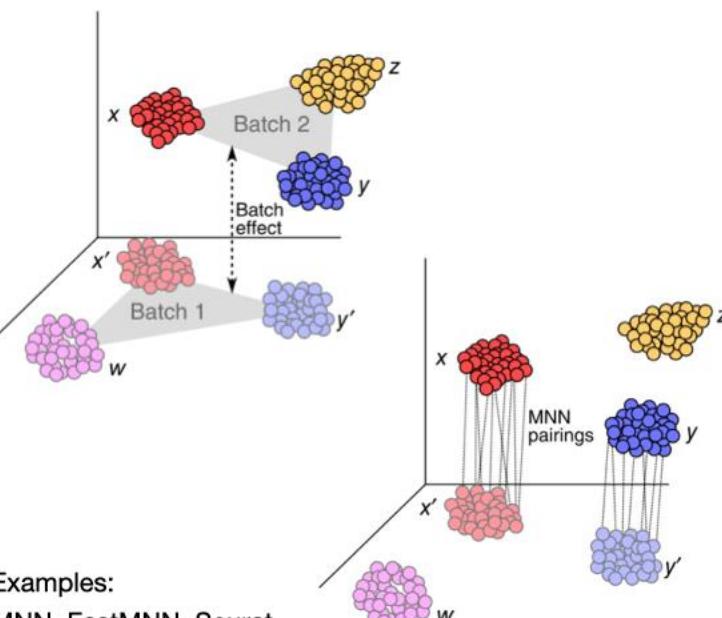
$$n_{gcb} = \alpha_g + X\beta_g + \gamma_{gb} + \delta_{gb}\epsilon_{gcb}$$

bio design matrix additive batch effect multiplicative batch effect

Example:
`ComBat - scanpy.pp.combat()`

Linear embedding models

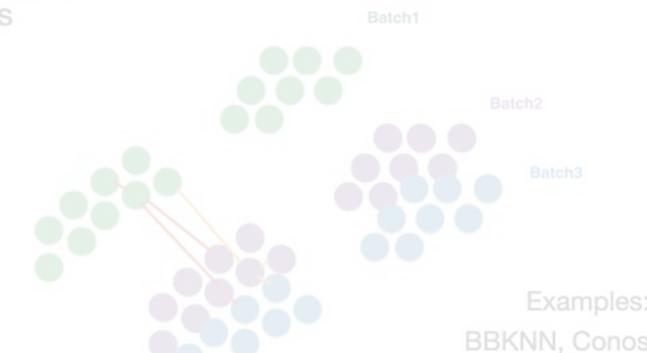
- Project cells into low dimensional embedding
- find most similar cells in other batch e.g., using mutual nearest neighbours (MNNs)
- Use MNNs as anchors to calculate a correction vector



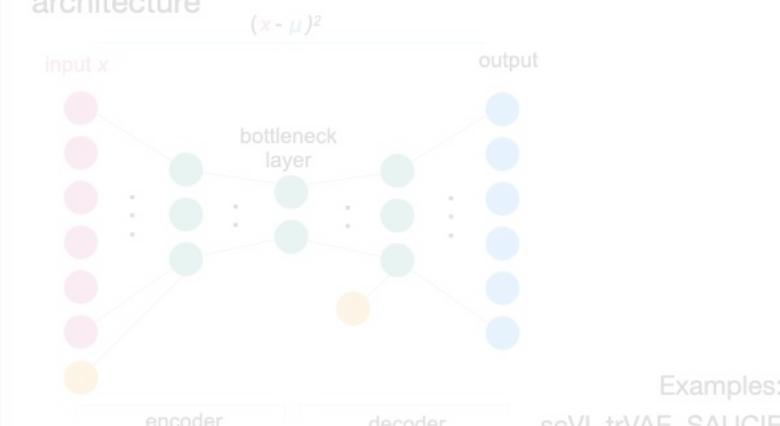
1

Graph-based methods & Deep learning

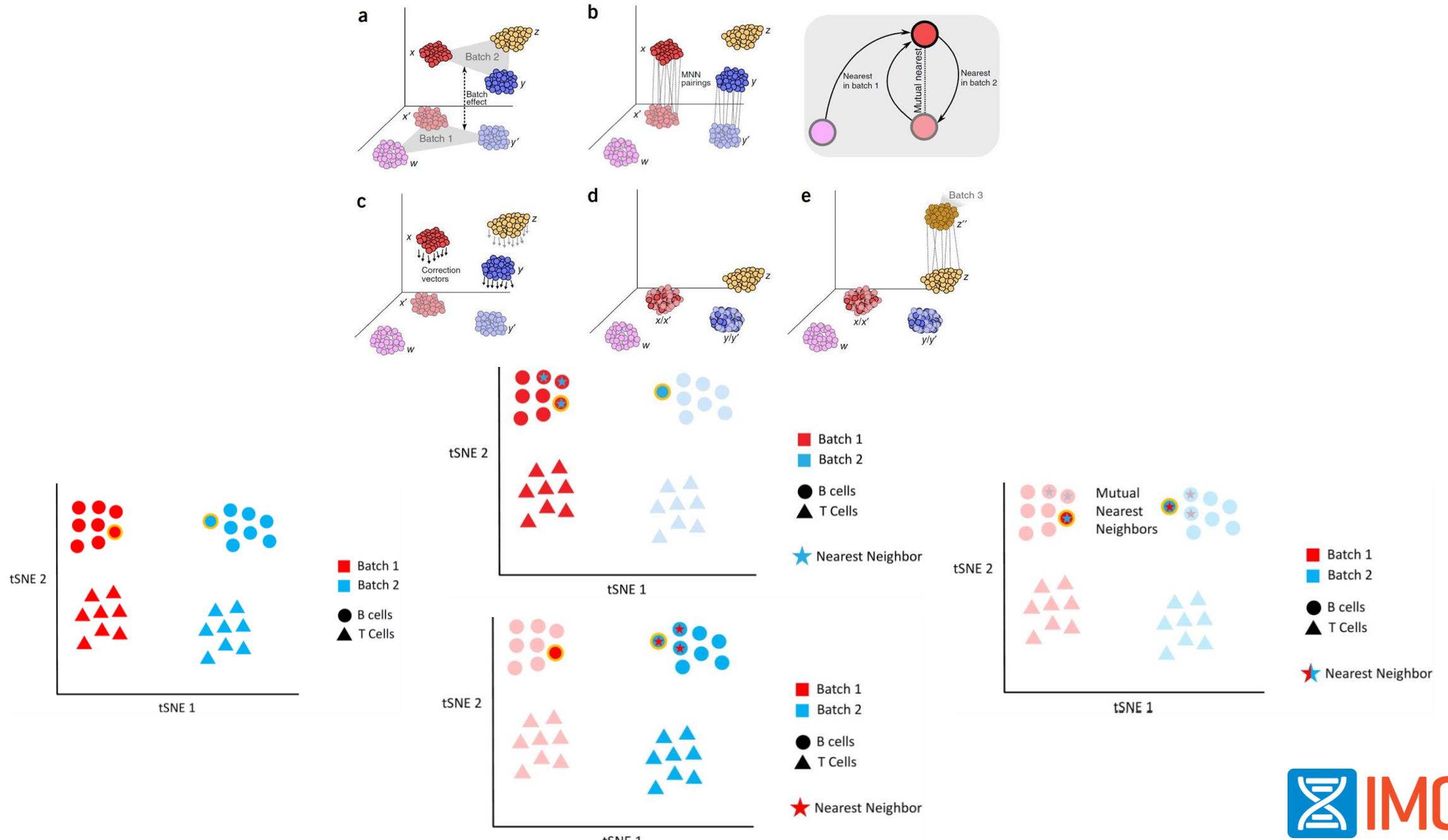
Enforce graph connections between different batches



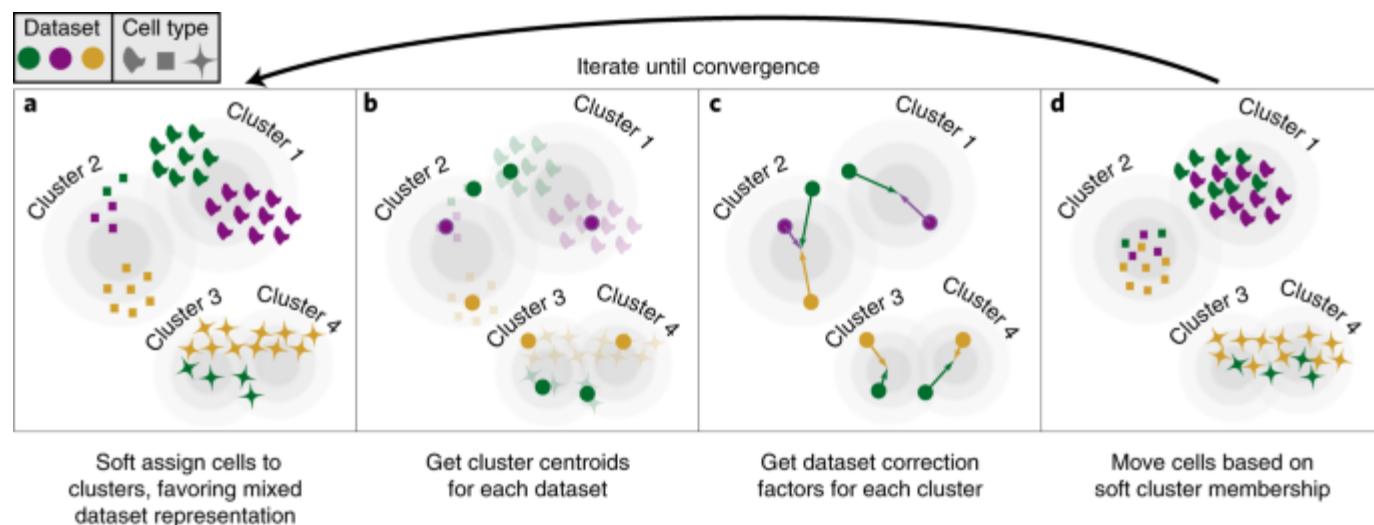
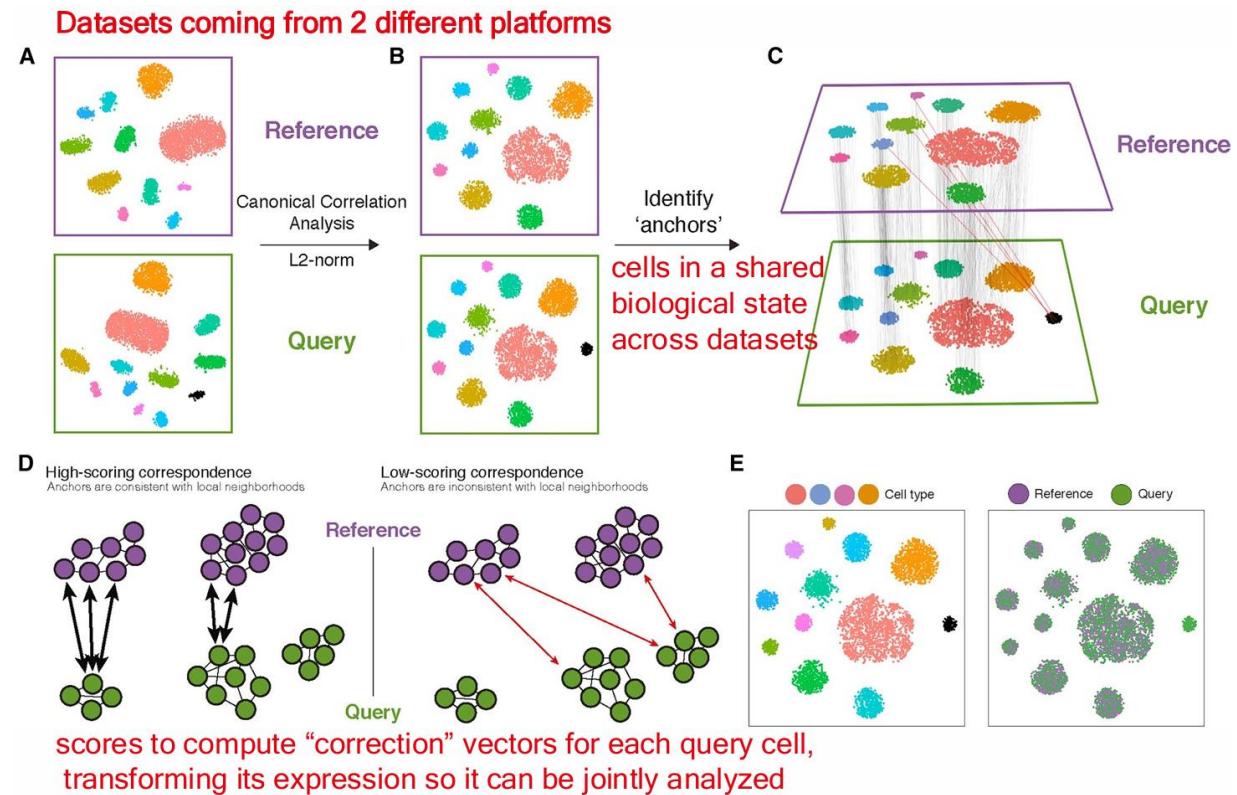
Add condition node into auto-encoder architecture



Integration using Mutual Nearest Neighbors (MNN)



Concept of integration



- https://satijalab.org/seurat/articles/integration_introduction
- https://www.sc-best-practices.org/cellular_structure/integration.html
- <https://www.singlecellcourse.org/biological-analysis.html#clustering-introduction>
- <https://bioconductor.org/books/3.12/OSCA/clustering.html#k-means-clustering>
- https://github.com/quadbio/scRNAseq_analysis_vignette/blob/master/Tutorial.md#step-2-3-data-integration-using-liger