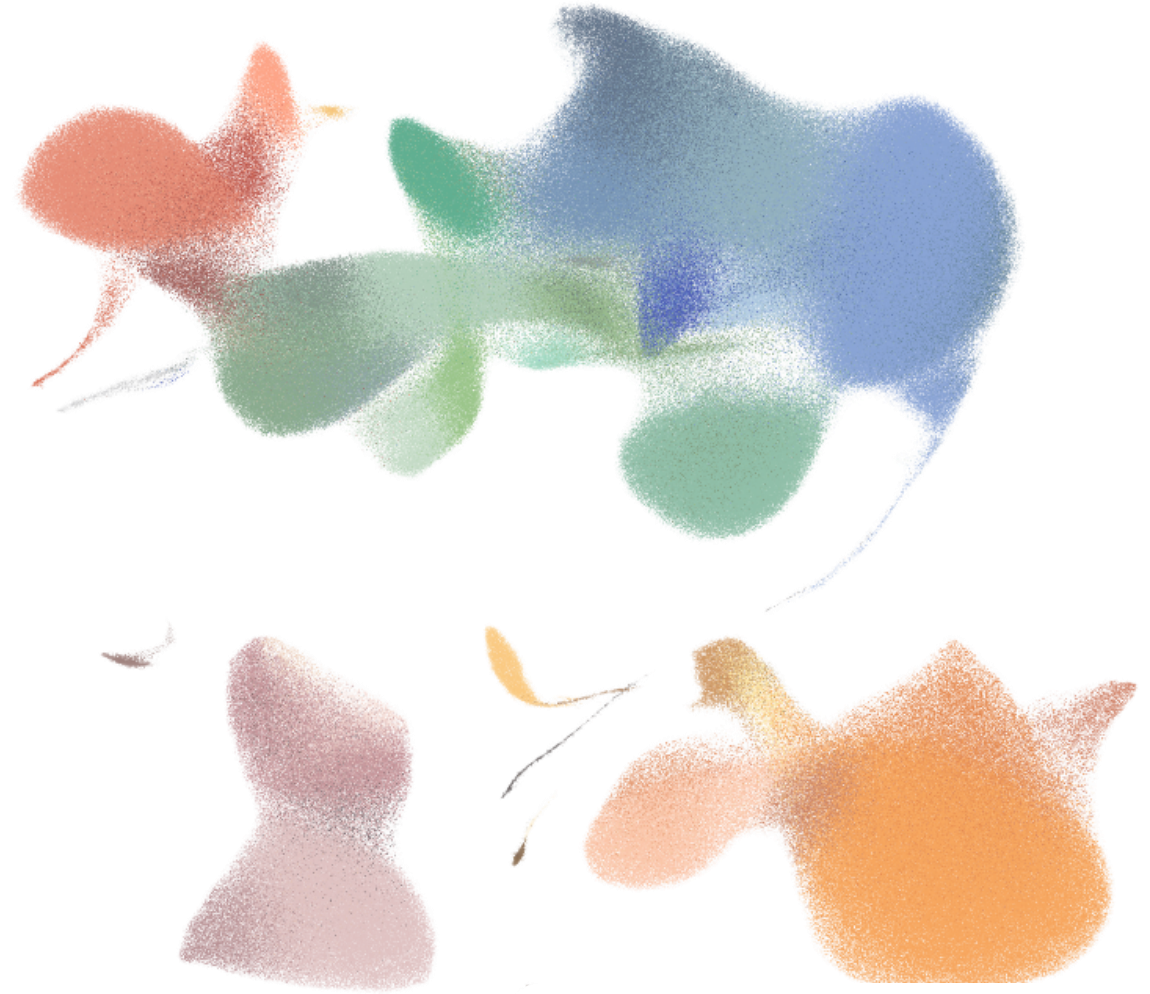


# Data Integration

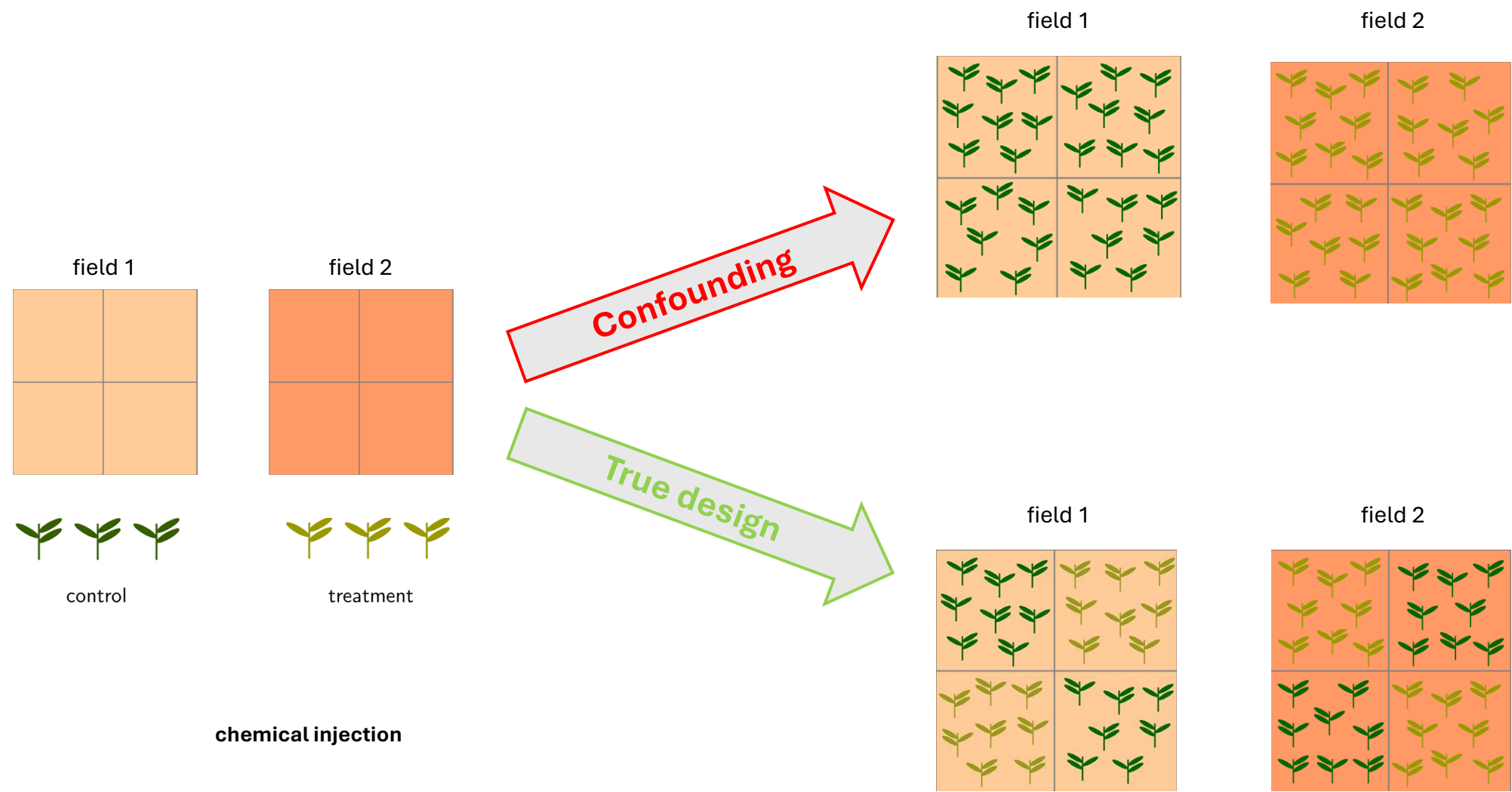
*Yusuf Caglar Odabasi*

*December 1.-3. 2025*

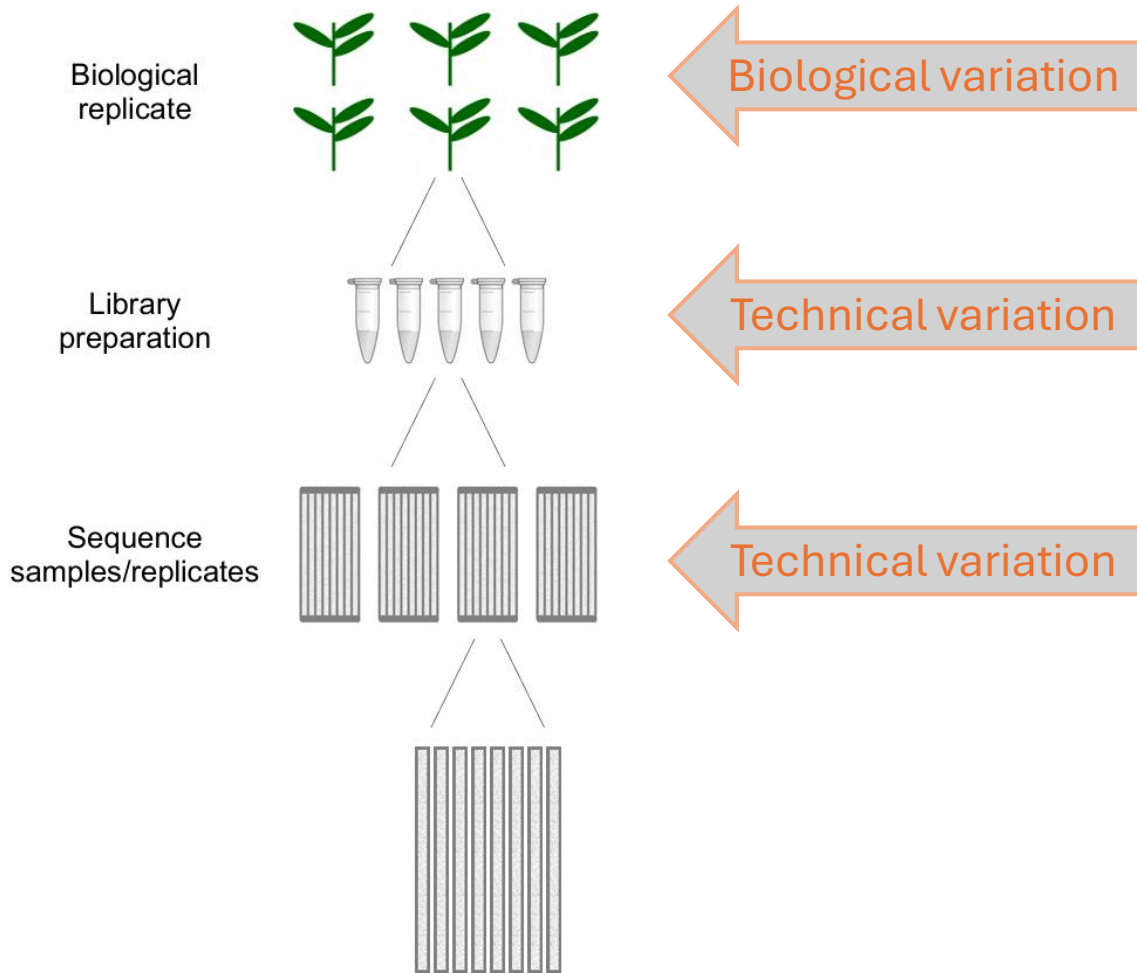
**Course on scRNA-seq Data Analysis**



# Experiment Design: Confounders and Batch effects



## Experiment Design: Confounders and Batch effects



### 1. Technical variability

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

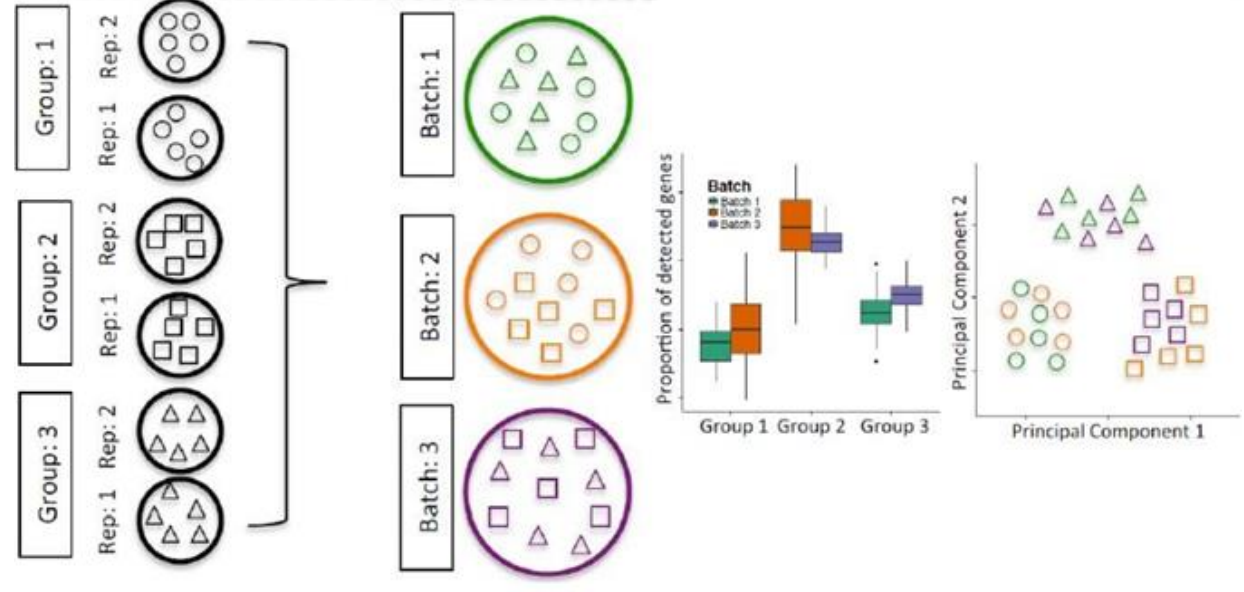
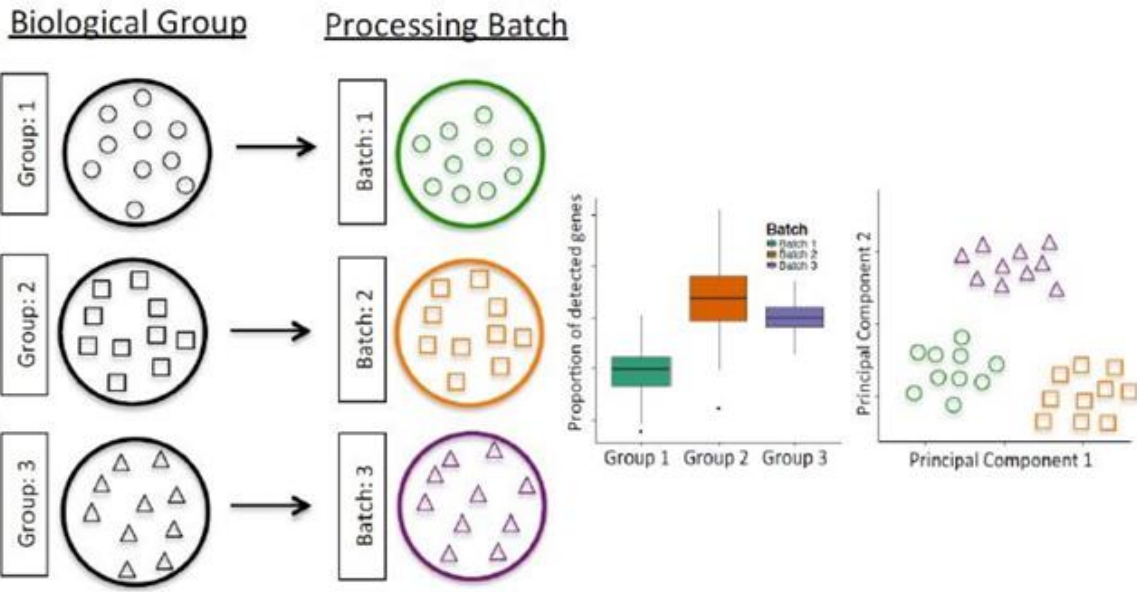
Technical 'batch effects' confound downstream analysis

### 2. Biological variability

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

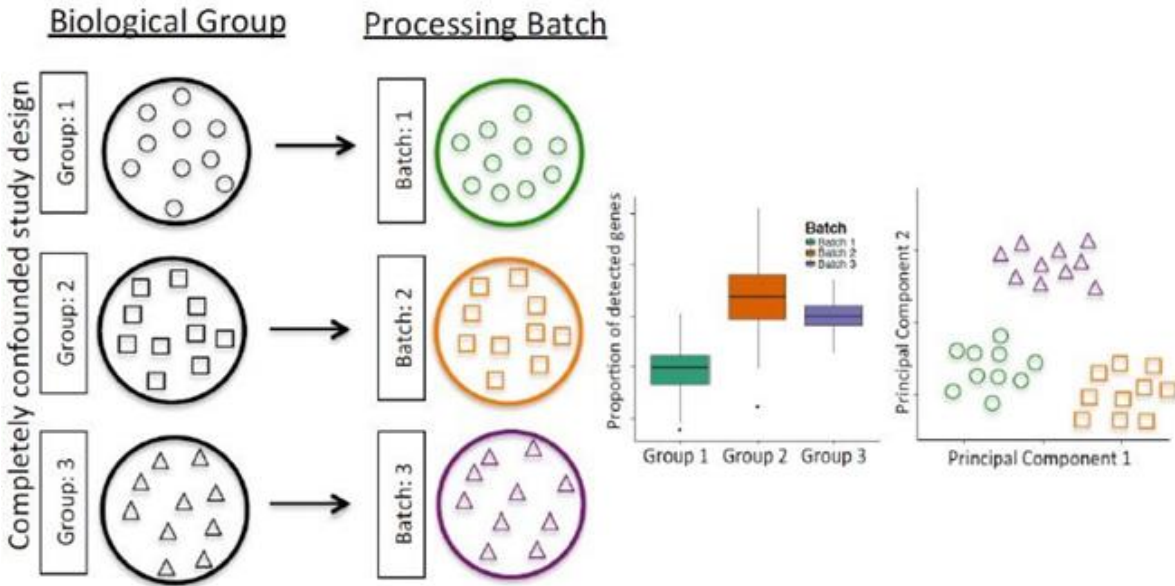
Biological 'batch effects' confound comparisons of scRNA-seq data

# Experiment Design: Confounders and Batch effects

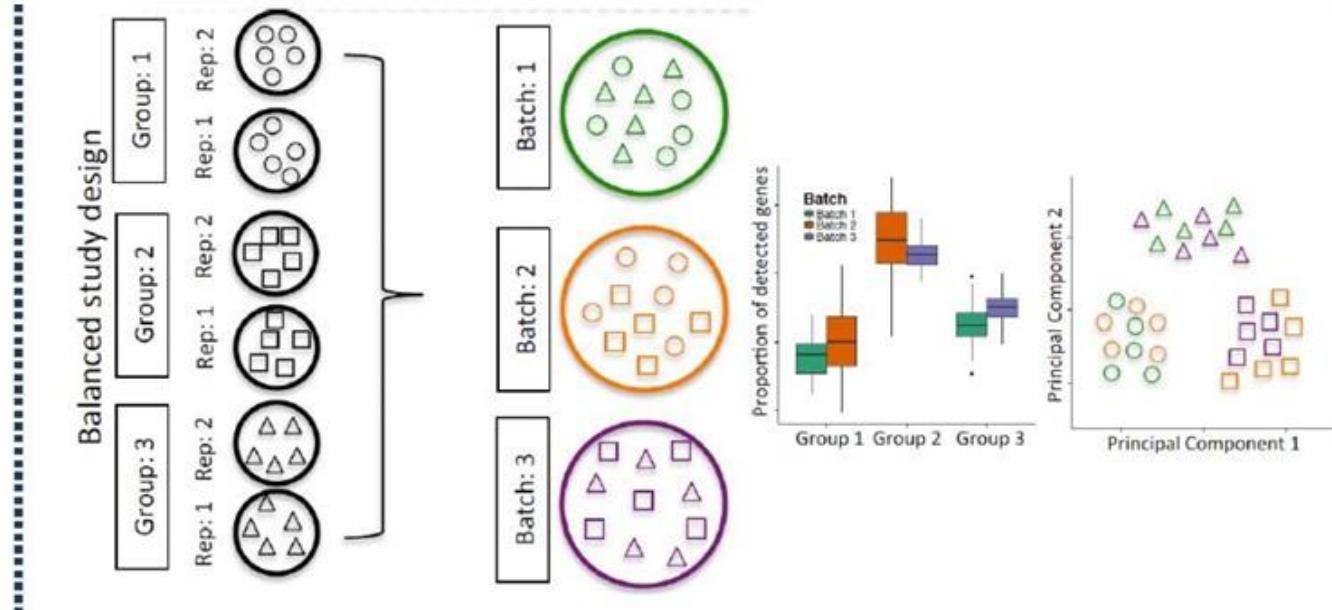


# Experiment Design: Confounders and Batch effects

## Confounded design



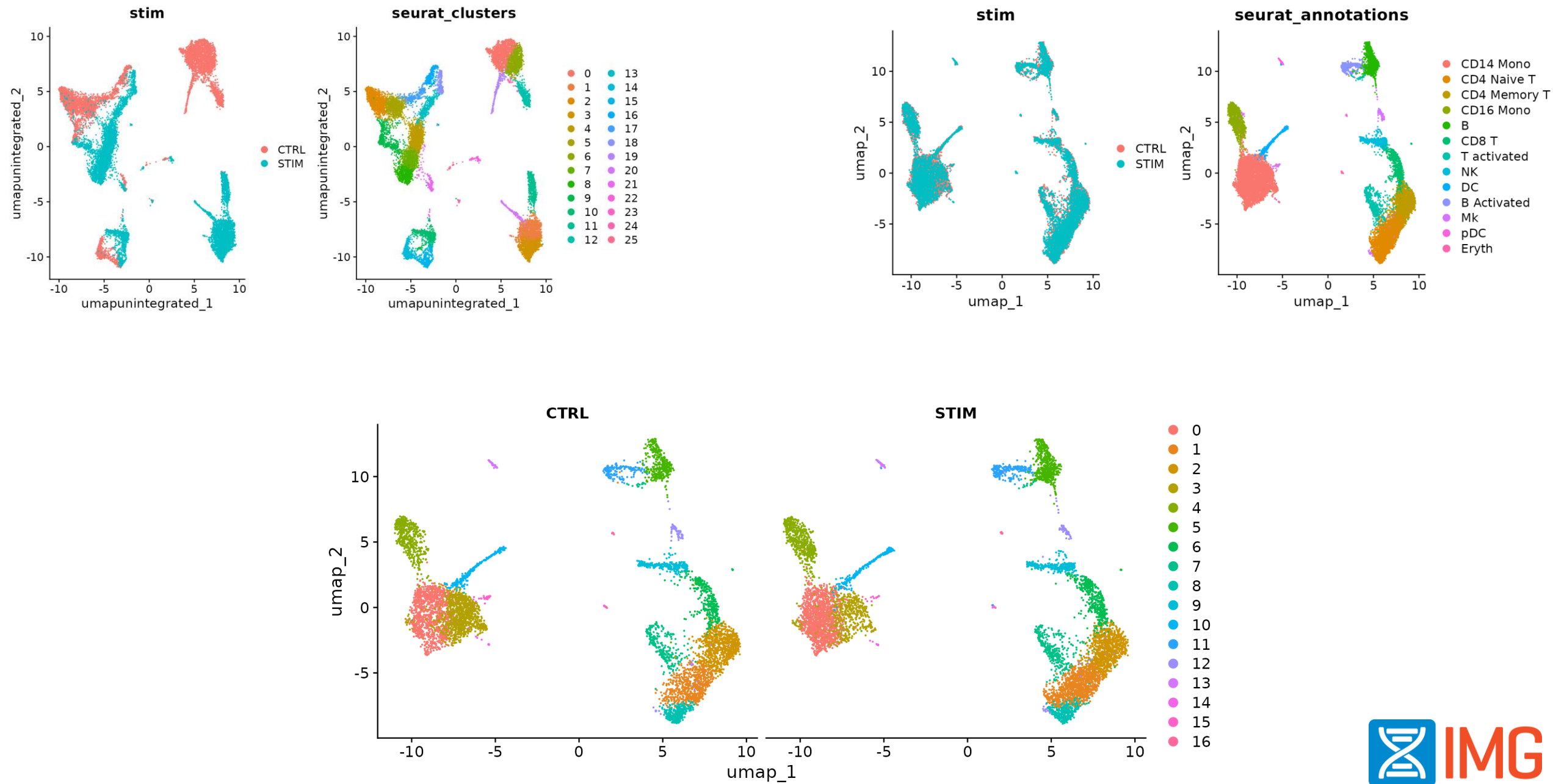
## Not confounded design



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

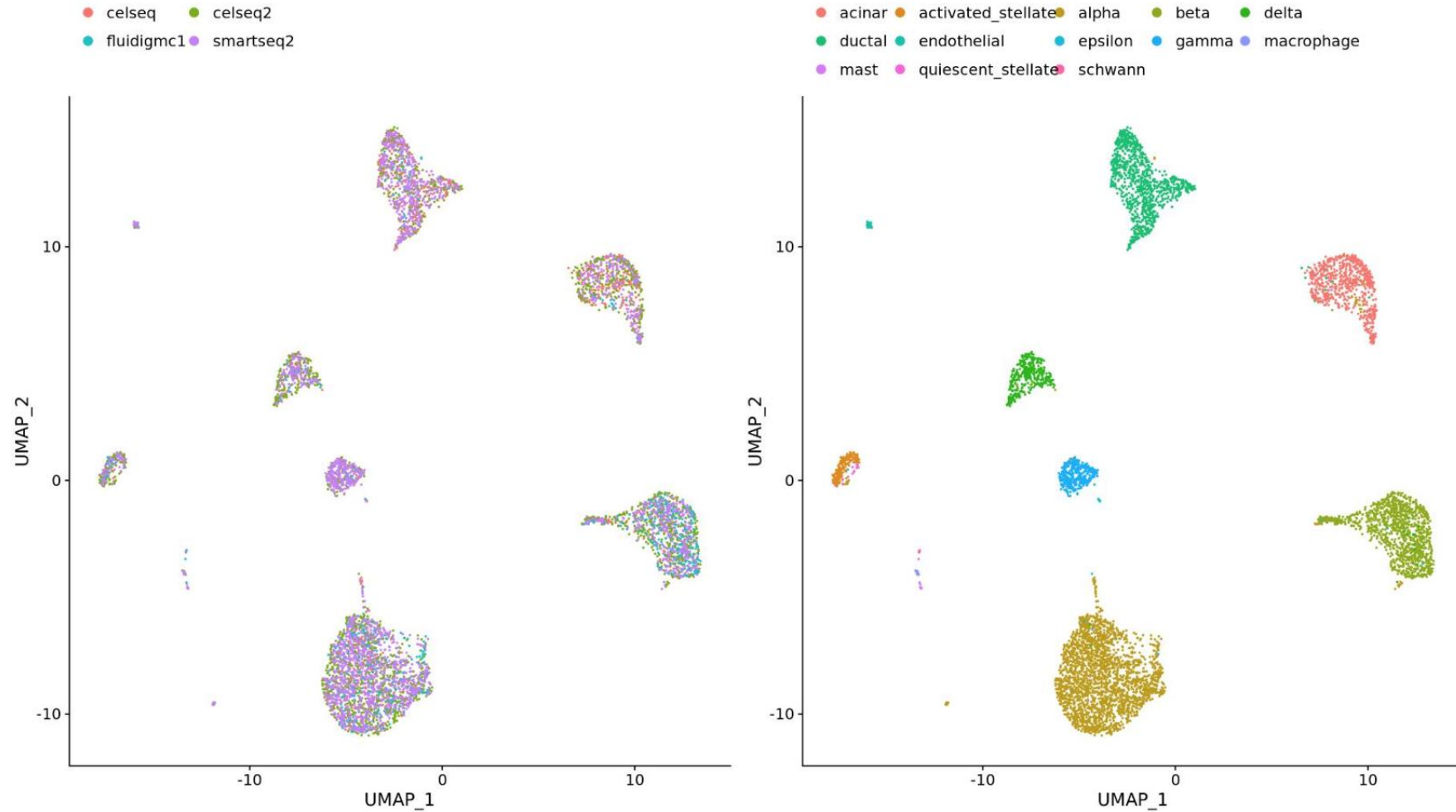


# Why do we need integration?



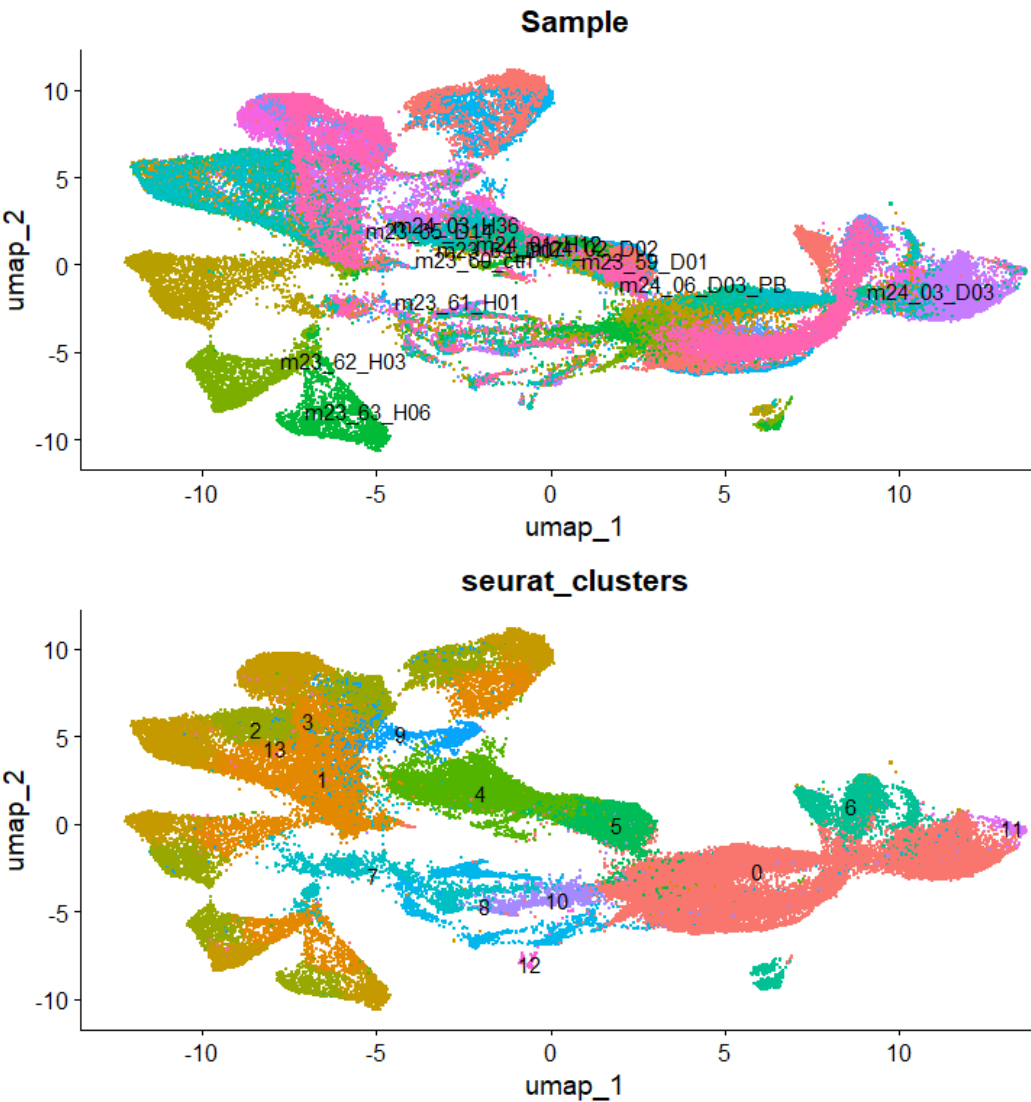
## Why do we need integration?

### Example scenarios for integration: **datasets**

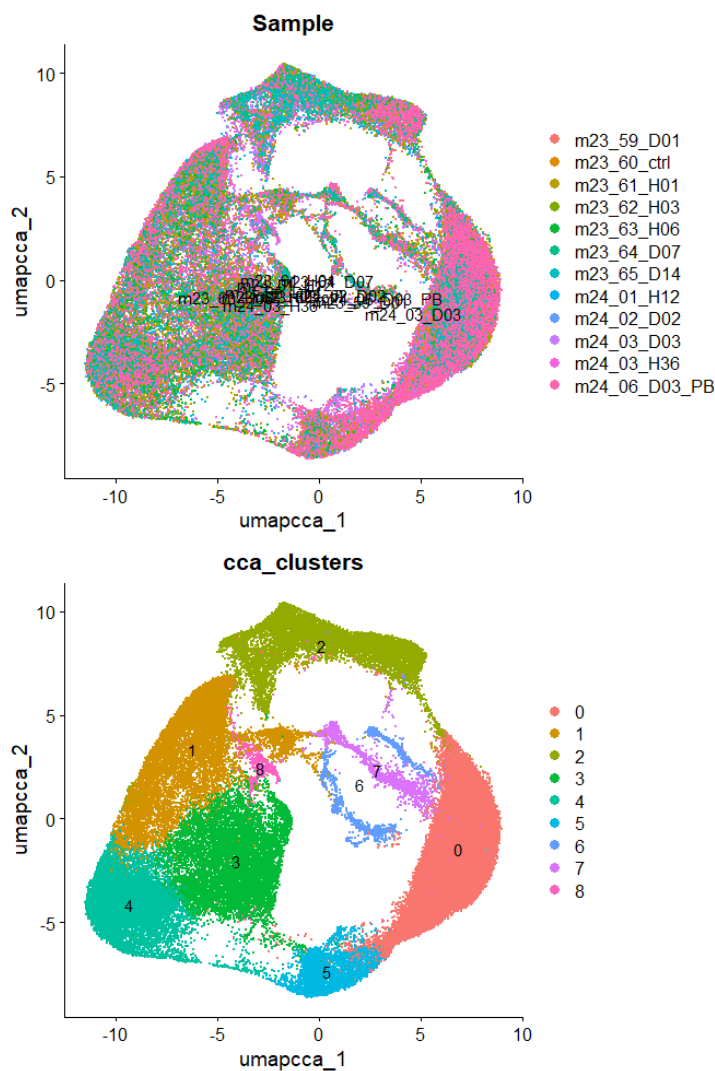


# Why do we need integration?

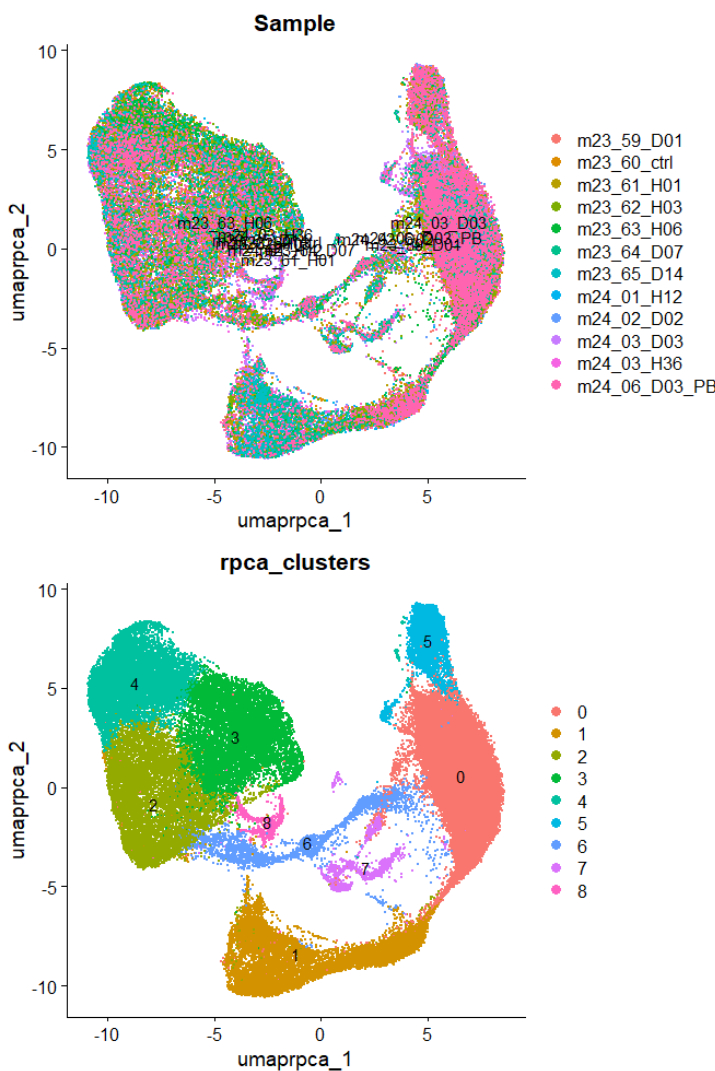
## Before integration



## CCA integration

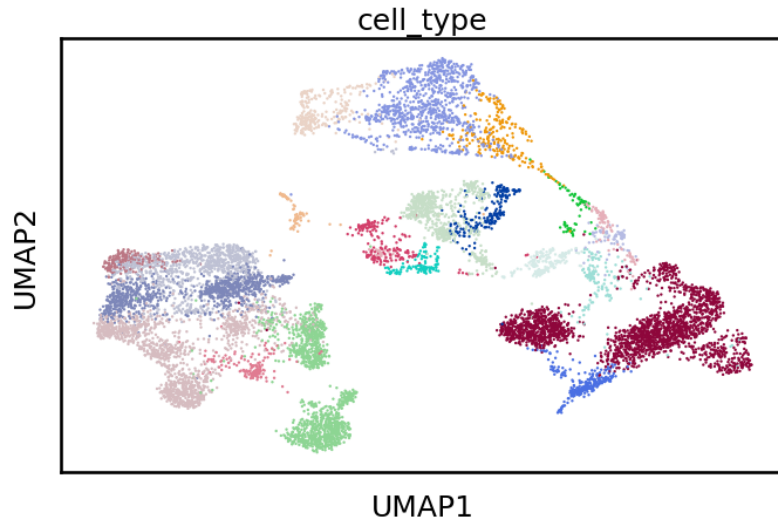


## RPCA integration

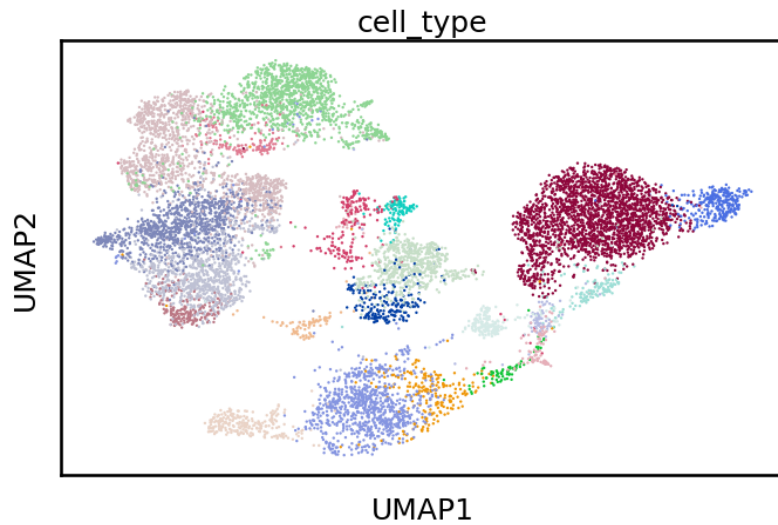
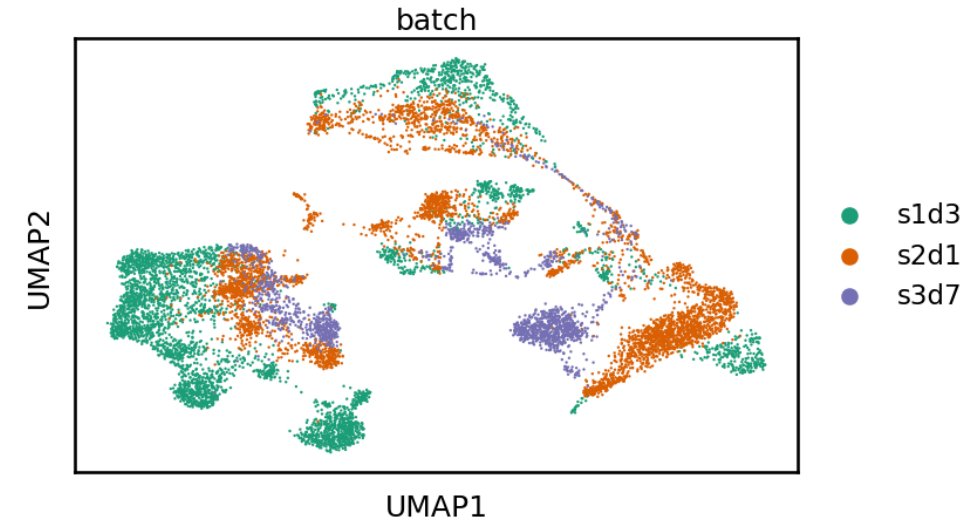




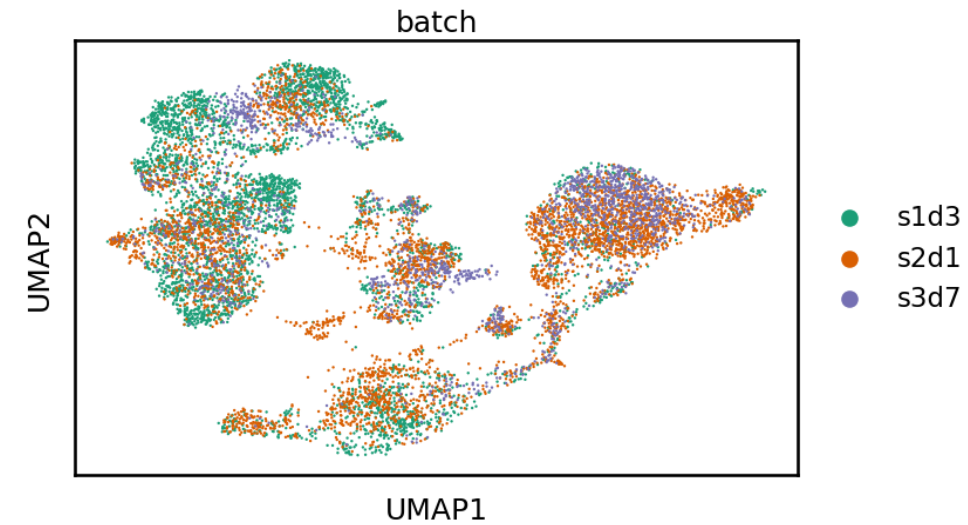
# Why do we need integration?



- 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



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- Transitional B
- cDC2
- pDC



# 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

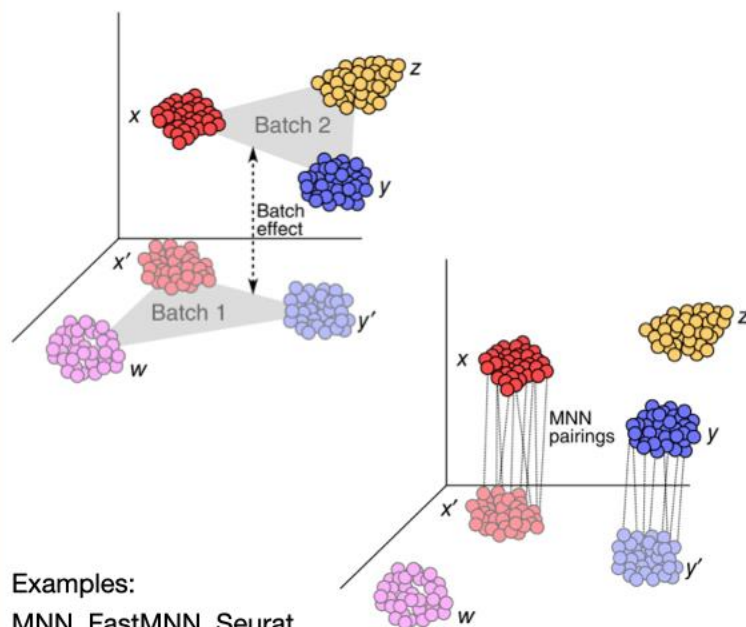
additive batch  
effect

multiplicative  
batch effect

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

## Linear embedding models

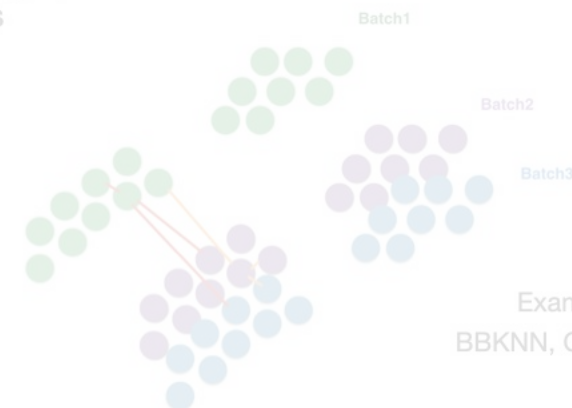
- Project cells into low dimensional embedding
- find most similar cells in other batch e.g., using mutual nearest neighbours (MNNs)
- Use MNNs as anchors to calculate a correction vector



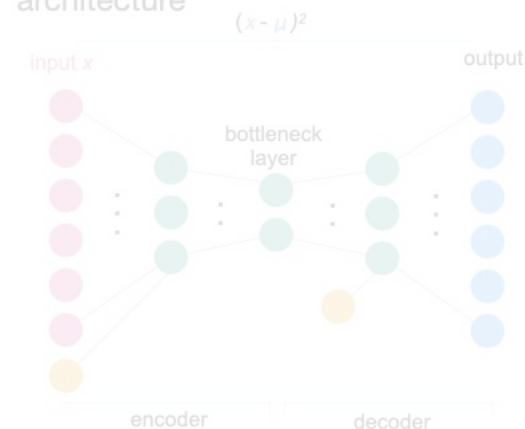
Examples:  
MNN, FastMNN, Seurat  
v3, Scanorama

## Graph-based methods & Deep learning

Enforce graph connections between different batches

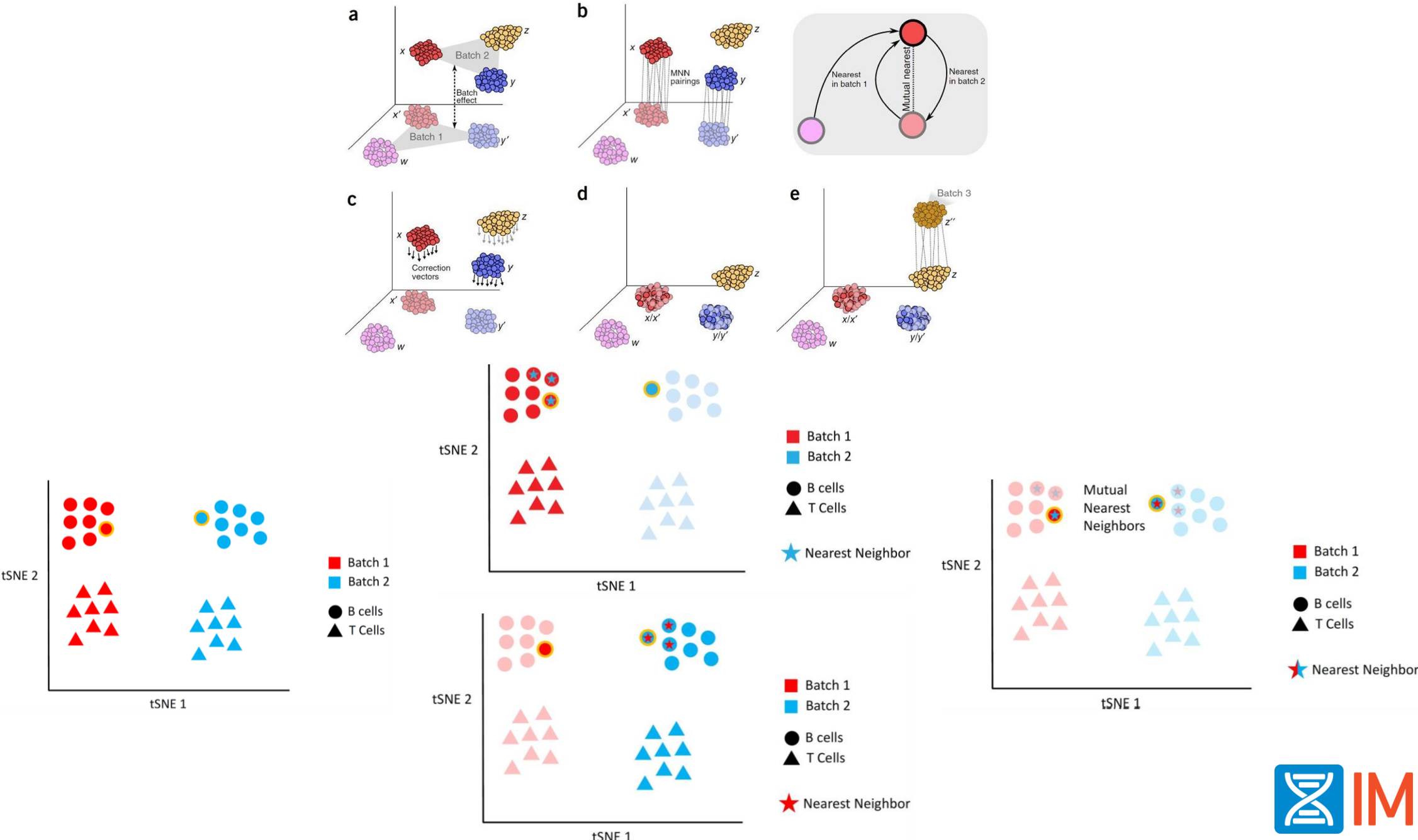


Add condition node into auto-encoder architecture

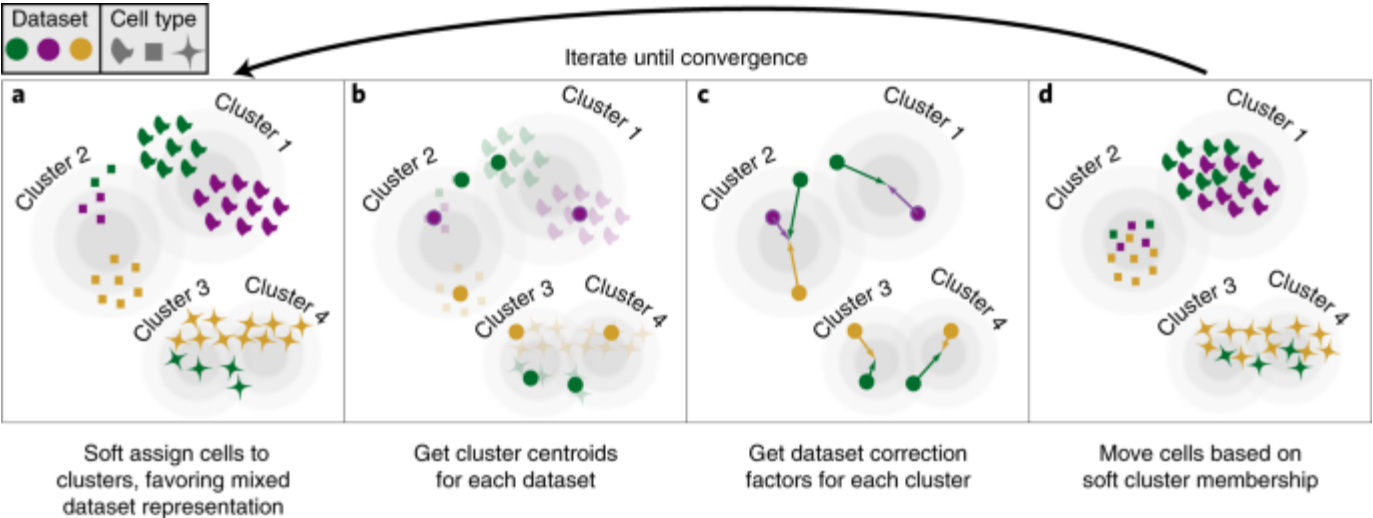
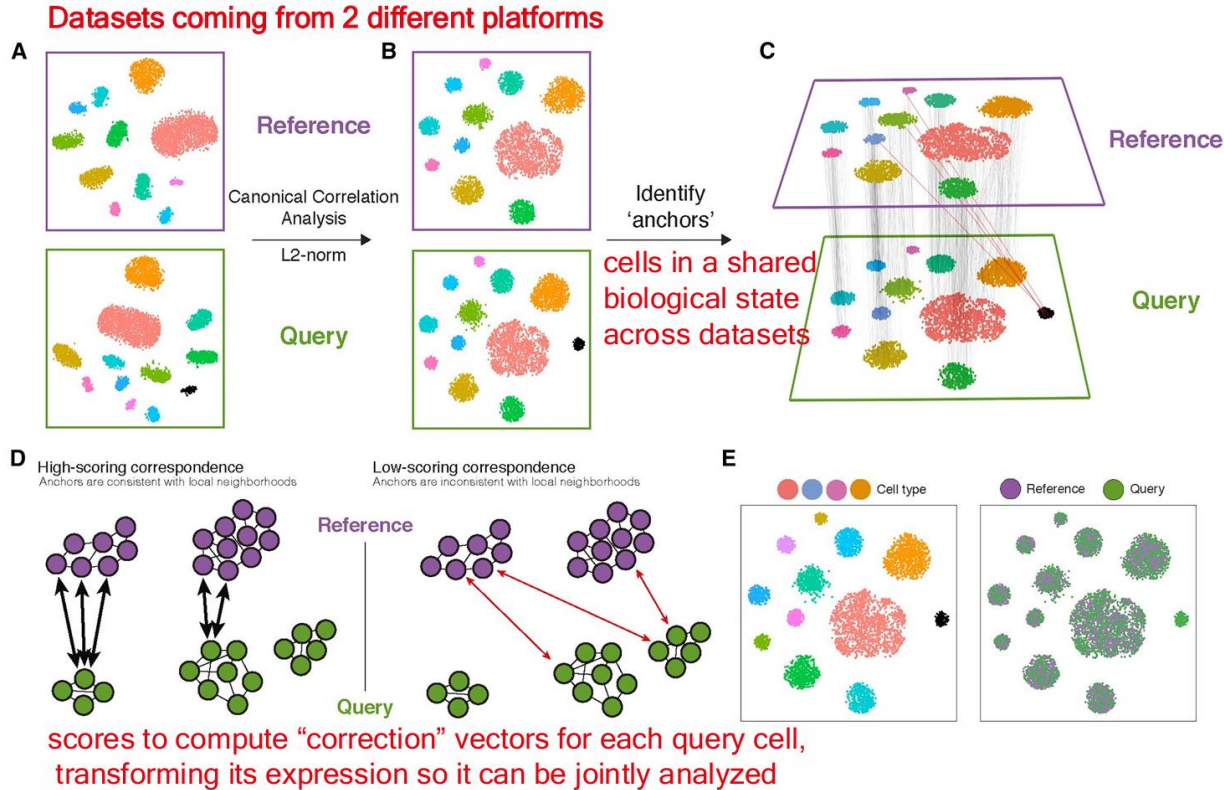


Examples:  
scVI, trVAE, SAUCIE

# Integration using Mutual Nearest Neighbors (MNN)



# Concept of integration



- [https://satijalab.org/seurat/articles/integration\\_introduction](https://satijalab.org/seurat/articles/integration_introduction)
- [https://www.sc-best-practices.org/cellular\\_structure/integration.html](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](https://github.com/quadbio/scRNAseq_analysis_vignette/blob/master/Tutorial.md#step-2-3-data-integration-using-liger)