

# Thoughts on Perturb-Seq data analysis approaches and pipelines

Kirill Tsukanov  
Senior Full Stack Developer & Data Engineer  
`ktsukanov@ebi.ac.uk`

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# Perturb-Seq data in the context of the Perturbation Catalogue

- ▶ For MAVE and CRISPR assays, we are lucky to have curated repositories (MaveDB, DepMap) with good quality, highly processed datasets.
- ▶ Perturb-Seq experiments are more complex: repositories like scPerturb aggregate dozens of studies but provide mostly raw expression counts.
- ▶ Interpretation and downstream analysis is left to the user, which can be quite complex.

# Perturb-Seq data essence

- ▶ Each observation is roughly: perturbing *gene X* in a specific cell type/tissue under set conditions yields a given gene expression profile per cell.
- ▶ Data characteristics:
  - ▶ High noise levels;
  - ▶ Pronounced batch effects;
  - ▶ Large scale: thousands of cells per perturbation, multiple conditions;
  - ▶ Raw counts require normalization, filtering, summarization, enrichment/differential expression/etc.
- ▶ Raw Perturb-Seq matrices need some systematic processing to become useful for Perturbation Catalogue users.

# Possible processing approaches for Perturb-Seq data

- ▶ Specialized pipelines exist for rigorous statistical analysis:
  - ▶ **Python:** MIMOSCA, MAESTRO (with partial AnnData compatibility).
  - ▶ **R:** SCEPTRE, Mixscape.
- ▶ Challenges:
  - ▶ Tools are highly specialized, may require steep learning curves.
  - ▶ For many, limited maintenance past the initial publication.
  - ▶ Poor compatibility with the broader Python ecosystem for single cell analysis.



# Implementation Update: Curated Studies

- ▶ **Progress:** 4 studies from scPerturb curated by Aleks - huge thanks!
- ▶ Curation process is now well established, unified, and will proceed even quicker in the future.
- ▶ Currently all studies have 1 cell type per study; in the future we'll curate more and larger studies.

Study	Size	Genes	Cells/Gene	Cell Type
adamson_2016_pilot	117M	7	500	lymphoblast
adamson_2016_upr_epistasis	479M	15	8-1500	lymphoblast
adamson_2016_upr_perturb1sg	186G	90	250-750+	lymphoblast
datlinger_2017	132M	32	50-250	T cell

# Processing approach: Pseudobulk Differential Expression

- ▶ **Input source:** scPerturb harmonised + curated to the common data schema (already implemented).
- ▶ **Strategy:** compute pseudobulk differential expression (simple and robust).
- ▶ **Workflow:**
  1. Group cells by control vs. perturbation within each cell type.
  2. Aggregate counts to pseudobulk profiles (mean expression per perturbation).
  3. Perform differential expression using t-tests with multiple testing correction.
  4. Apply filtering: adjusted p-value  $< 0.05$ ,  $-\log_2\text{FC} > 1$ .
- ▶ **Implementation:** Parallel processing with expression thresholding and robust statistical testing.





# Next Steps and Future Directions

## ▶ **Immediate:**

- ▶ Continue curating additional studies from scPerturb.
- ▶ Optimize processing pipeline for larger datasets.
- ▶ Implement user interface for browsing results.

## ▶ **Future enhancements:**

- ▶ Support for multi-cell-type studies.
- ▶ Integration with pathway enrichment analysis.
- ▶ Advanced visualization tools.
- ▶ Cell-type-specific perturbation effects.

## ▶ **Questions/Discussion:**

- ▶ Alternative analysis approaches?
- ▶ Priority datasets for curation?