

# A machine learning-powered dashboard for the exploration of high-throughput transcriptomic datasets



V. Bernu<sup>1</sup>, C. Lescure<sup>1</sup>, H. Brull Corretger<sup>1</sup>, P. Dhillon<sup>1</sup>, E. Fox<sup>1</sup>, C. Marijon<sup>1</sup>, A. Nordor<sup>1</sup>, C. Petit<sup>1</sup>, A. Behdenna<sup>1</sup> <sup>1</sup>Epigene Labs, Paris, France

# Introduction

- NCBI's GEO<sup>2</sup> database is a major repository for **high-throughput** transcriptomic datasets<sup>3</sup> referencing ~7,000,000 samples<sup>4</sup>, including a significant number of tumor biopsy samples<sup>4</sup>.
- Today, this invaluable database is underused because of technical roadblocks<sup>[1]</sup>.
- We demonstrate that recent Al developments unlock this vast potential to infer new biological understanding and shape future clinical study designs.

# Key results

- Our high-performance and state-of-the-art Al models identify the most relevant GEO datasets in oncology, representing the top 10%, which we refer to as **Epigene Labs data lake**.
- These models are integrated into a filtering and prioritization pipeline (Fig. 2), designed and developed for high scalability, modularity, and updatability.
- A GEO<sup>2</sup> dashboard (Fig. 3) enables a user-friendly interface with the Epigene Labs data lake by aggregating the results of the above-mentioned models.

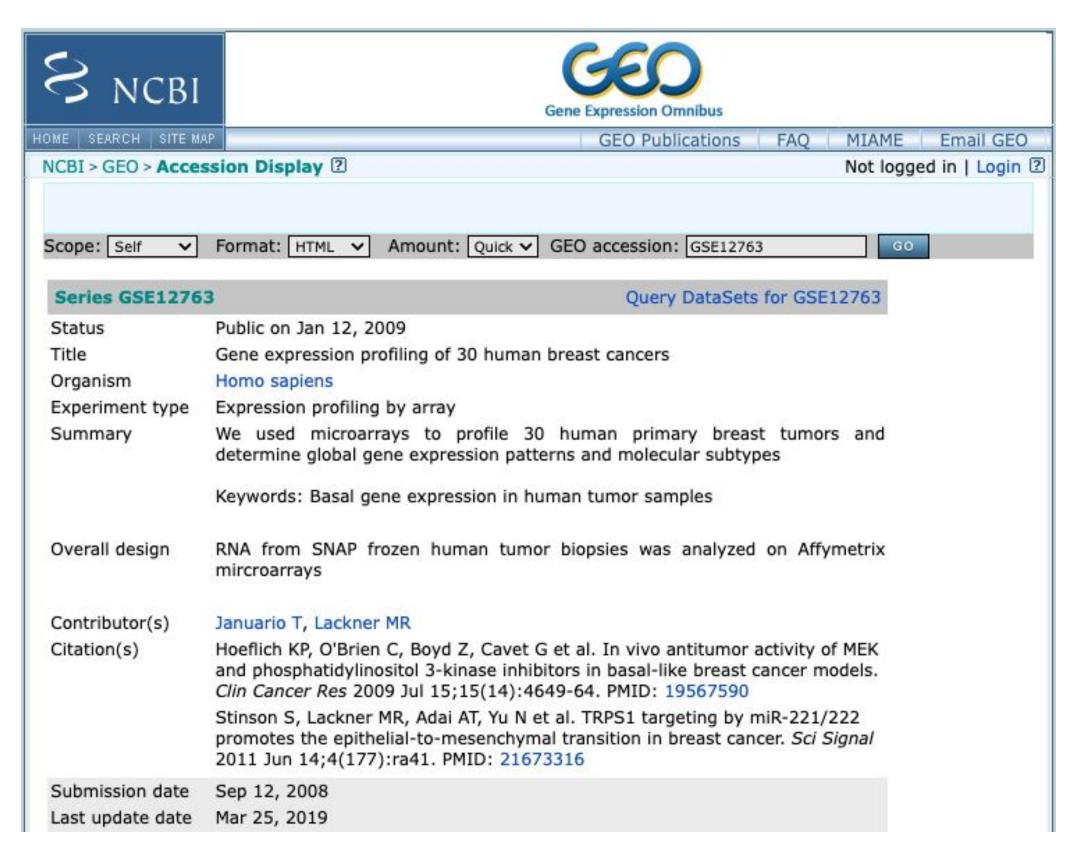


Figure 1: Screenshot of the description of a GEO<sup>2</sup> dataset

# Lexicon

- **NCBI**: National Center for Biotechnology Information<sup>[IV]</sup>.
- 2. **GEO**: The Gene Expression Omnibus is a public genomics data repository.
- 3. Dataset: Refers to a GEO<sup>2</sup> Series, (GSExxx), it is "an original submitter-supplied record that summarizes a study"[IV].
- 4. **Sample**: "A Sample record describes the conditions under which an individual Sample was handled. Each Sample record is assigned a unique and stable GEO accession number (GSMxxx)"[IV].
- 5. **Technology**: Method for the production of the gene expression profiles (e.g., Microarray or RNA-seq).
- 6. Platform: "A Platform record is composed of a summary description of the array or sequencer and, for array-based Platforms, a data table defining the array template. Each Platform record is assigned a unique and stable GEO accession number (GPLxxx)"[IV].
- 7. NER and NEN: Named Entity Recognition and Named Entity Normalization are natural language processing methods.
- 8. **LLM**: Large Language Model (e.g., GPT-4<sup>[II]</sup> or Mistral<sup>[III]</sup> models).
- 9. **AUC**: Area Under the receiver operating characteristic Curve. It is a score between 0 and 1.

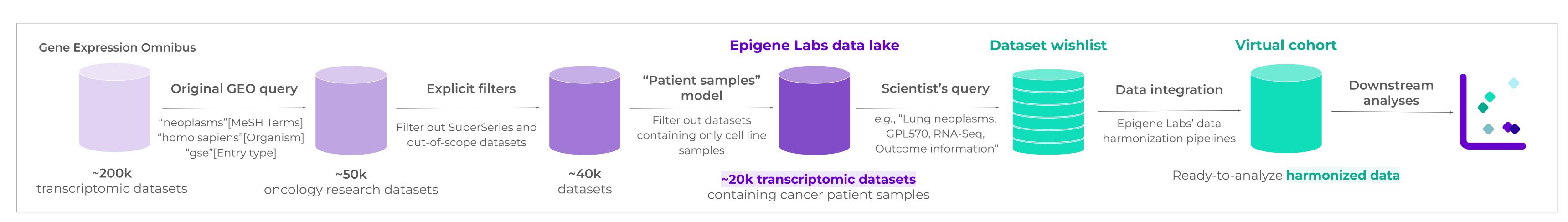


Figure 2: End-to-end pipeline, from heterogeneous datasets<sup>3</sup> to harmonized virtual cohorts

### Methods

#### Classification models

Our models (Table 1) classify datasets<sup>3</sup> based on ground truth data that were labeled by cancer scientists. They are divided into two categories:

- Dataset description-based models:
  - Utilize only dataset-level descriptions (Fig. 1).
  - Offer high scalability to encompass the entirety of Epigene Labs data lake.
- Sample description-based models:
  - Require detailed sample-level descriptions.
  - o Demand increased computational power but deliver enhanced precision in results.

#### Large Language Models (LLMs<sup>8</sup>)

**Performance metrics** 

Adaptation for treatment-related models:

- Traditional training approaches were infeasible due to limited labeled data.
- Prompt engineering strategies harness the capabilities of pre-trained **LLMs** from OpenAl<sup>[II]</sup> and Mistral<sup>[III]</sup>.

- The models are evaluated by the metrics below, depending on the format of their respective output. All three are scores between 0 and 1.
- Classic binary classification: **AUC**<sup>9</sup> (the higher the better).
- Multiclass and LLM-based binary classification: **F1-score** (the higher the better).
- Multilabel classification: Hamming loss (the lower the better).

# **GEO** dashboard

- Intuitive data mapping: Our Metabase<sup>[V]</sup> dashboard visualizes the vast landscape of the Epigene Labs data lake, integrating diverse GEO explicit attributes and model-predicted attributes.
- Dynamic filtering system: With advanced filtering capabilities, the dashboard also permits in-depth interactive exploration of the data.
- Seamless research integration: The dashboard, combined with the Al models, empowers researchers to create tailored virtual cohorts for computational oncology (Fig. 2), sourced from the GEO database.

## Conclusion

Al techniques enable the annotation and exploration of the GEO<sup>2</sup> database, facilitating secondary analysis of oncology research omic data.

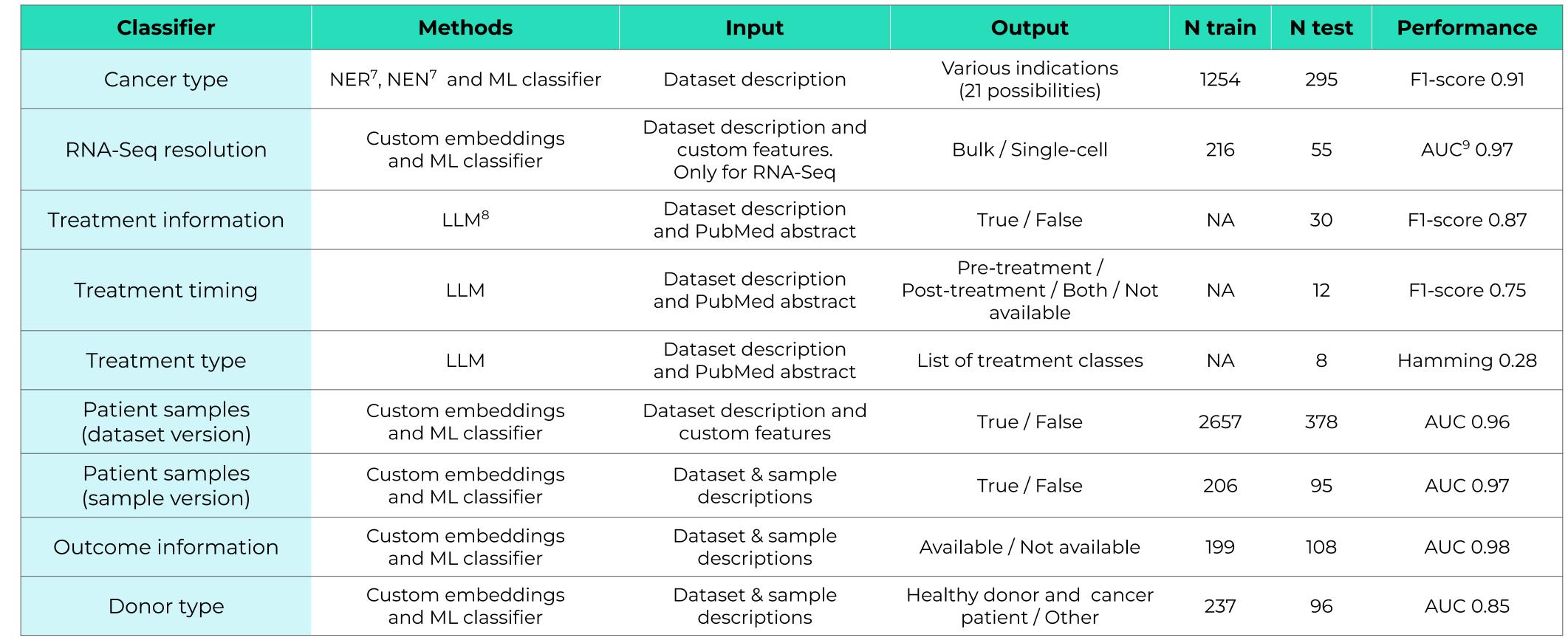


Table 1: Description, methods, and performance evaluation of the models

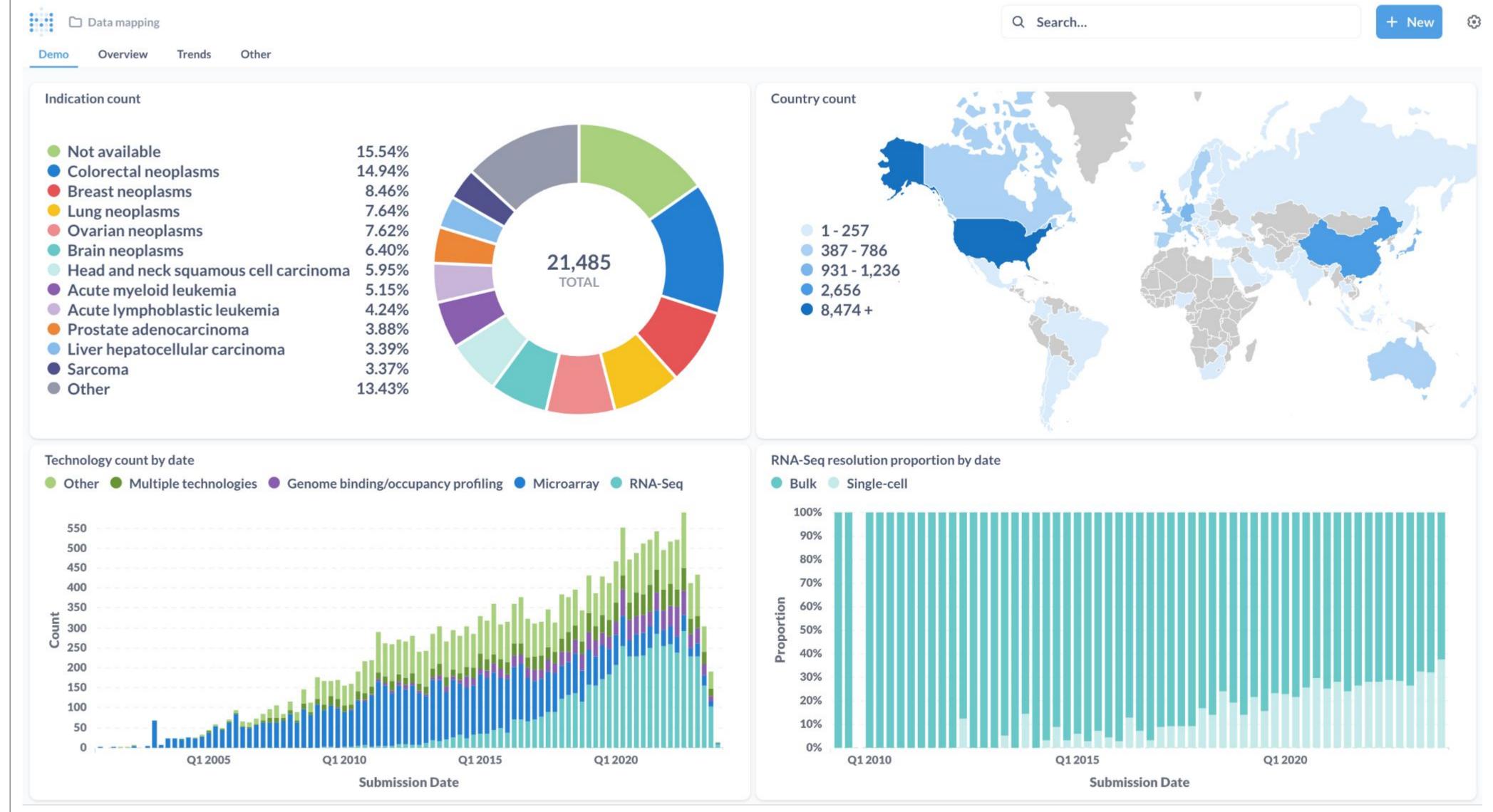


Figure 3: GEO Data Mapping Dashboard visualization example

# References

- . Hawkins, N., Maldaver, M., Yannakopoulos, A., Guare, L. & Krishnan, A. Systematic tissue annotations of genomics samples by modeling unstructured metadata. Nature Communications 13, (2022).
- II. OpenAl et al. GPT-4 Technical Report. Preprint at http://arxiv.org/abs/2303.08774 (2023).
- III. Jiang, A. Q. et al. Mistral 7B. Preprint at <a href="http://arxiv.org/abs/2310.06825">http://arxiv.org/abs/2310.06825</a> (2023).
- IV. National Center for Biotechnology Information (NCBI)[Internet]. Bethesda (MD): National Library of Medicine (US), National Center for Biotechnology Information; [1988] – [cited 26 Feb 24]. Available from: https://www.ncbi.nlm.nih.gov/.
- V. Metabase | Business Intelligence, Dashboards, and Data Visualization. https://www.metabase.com/
- Contact: Akpéli Nordor, PharmD, PhD
- (akpeli@epigenelabs.com). • Final publication number: 2P.
- The authors have no conflict of interest to declare.