



CellOntologyMapper: Consensus mapping of cell type annotation

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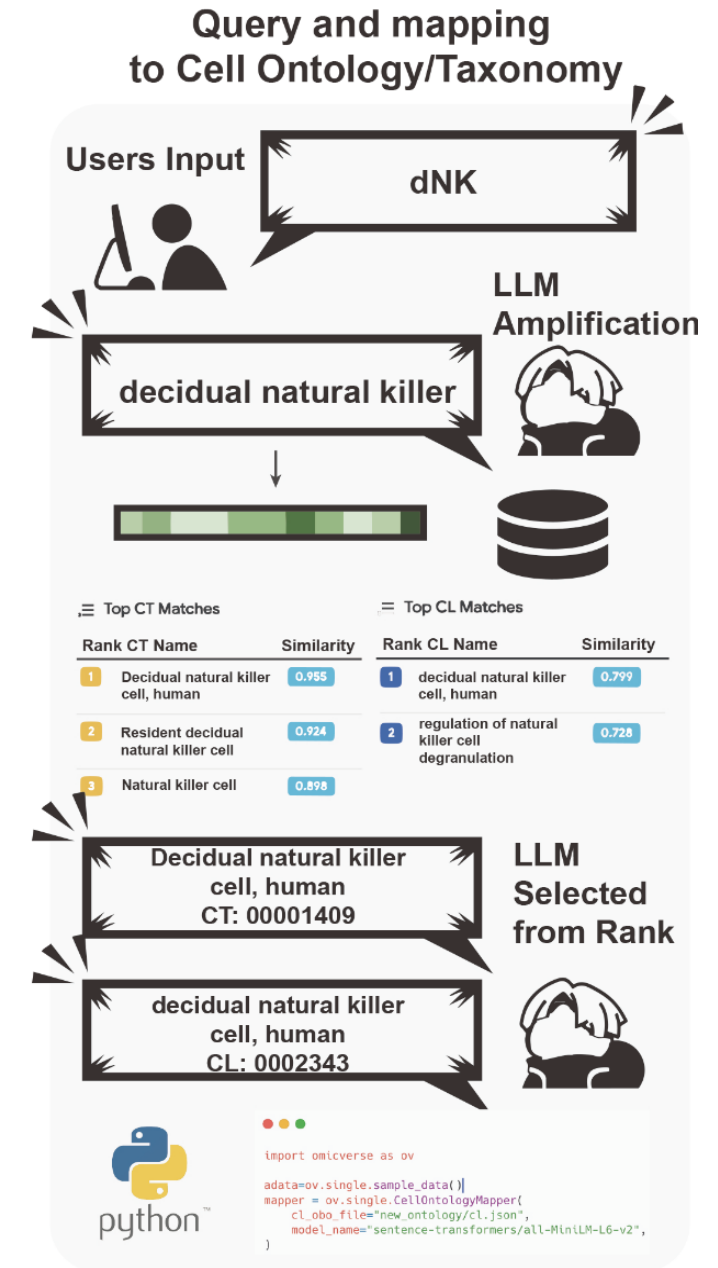
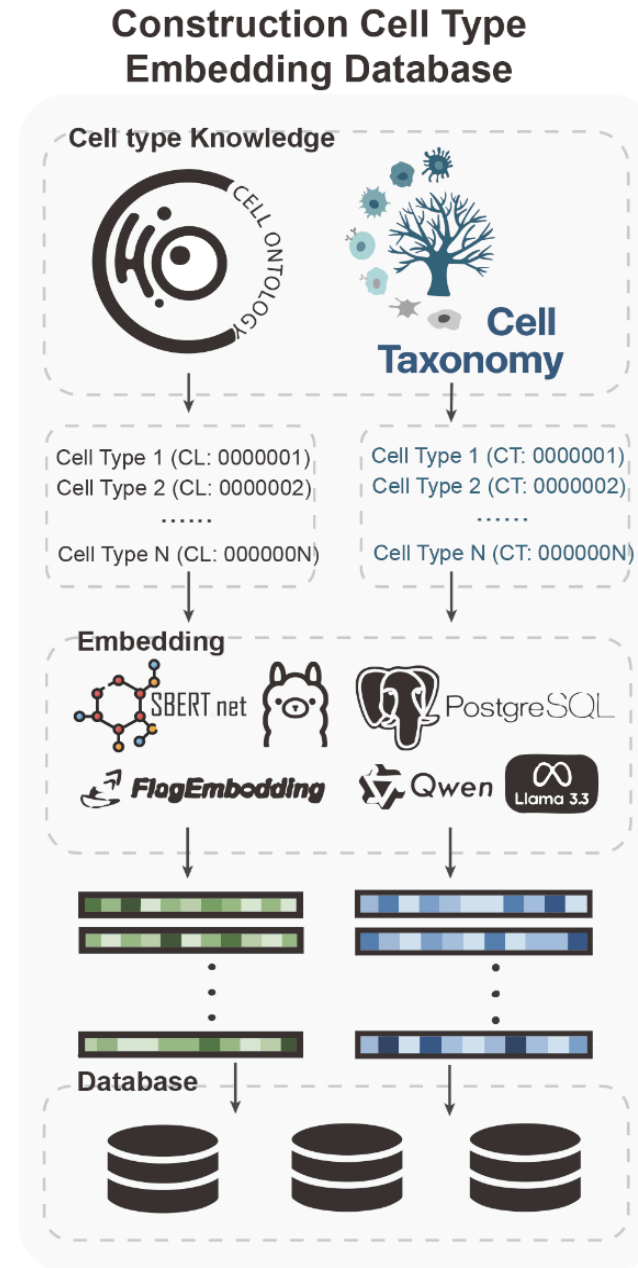


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Introduction

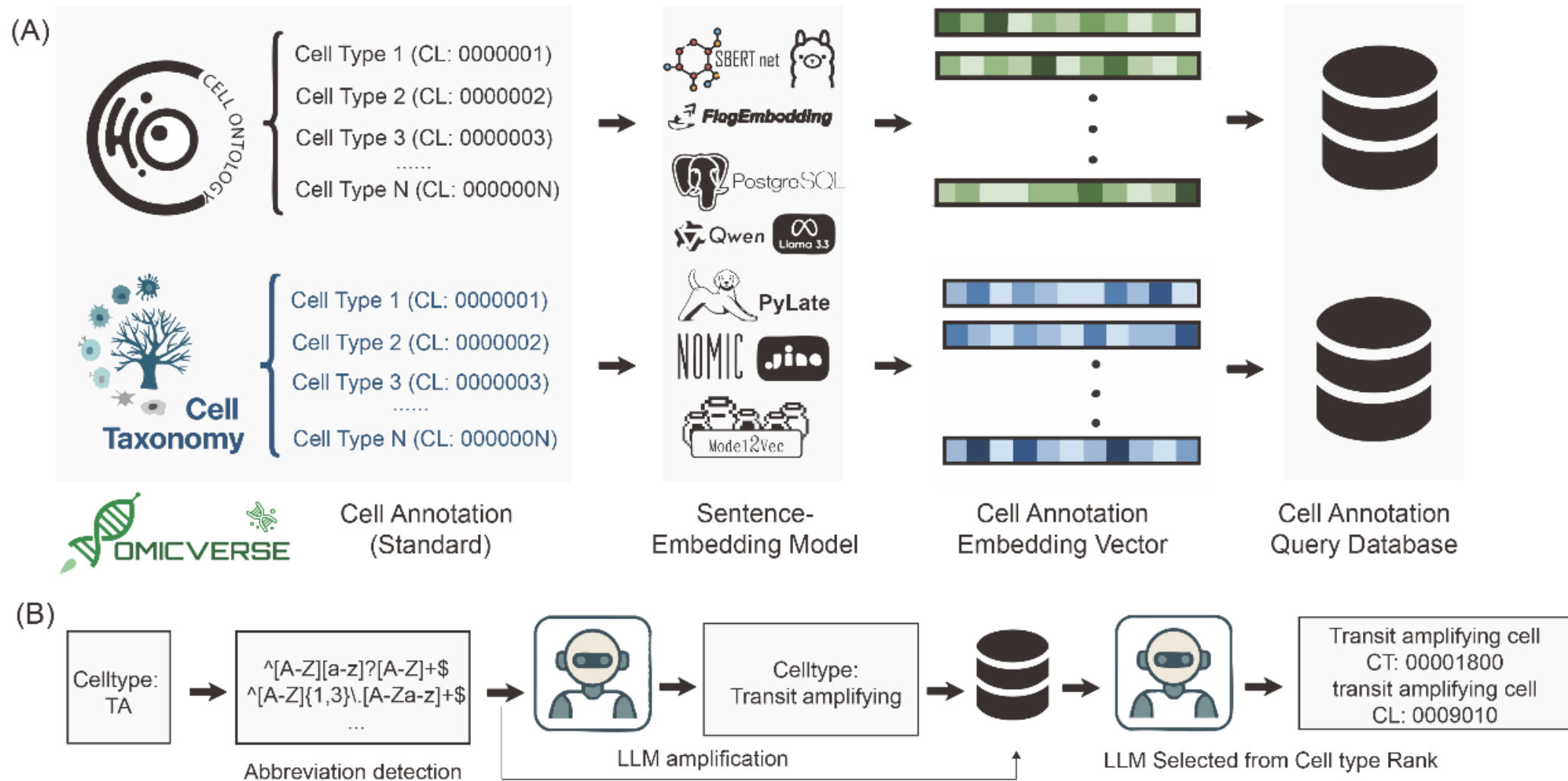
Single-cell RNA sequencing reveals cellular diversity, but inconsistent cell-type naming conventions—like "NEUT," "dM," or "Mac1"—hinder data integration and cross-study comparison, especially in cross-species research. We developed CellOntologyMapper, a computational tool that automatically maps custom cell names to standardized ontological terms using advanced natural language processing. It resolves abbreviations, synonyms, and tissue-specific contexts, and is available as a Python package in the OmicVerse, designed to integrate seamlessly into existing analysis workflows. Besides, user can also use our online website to perform the cell type name mapping (<https://omicverse.com/cellmapper/>).





Results: Overview of CellOntologyMapper Design

To address the challenge of inconsistent cell type naming, CellOntologyMapper standardizes cell type nomenclature through two interconnected and complementary modules: The first module constructs an extensive Cell Ontology Query Database, which serves as a foundational resource for accurate cell type mapping (1A); The second module implements an intelligent, context aware mapping pipeline aimed at converting user-provided cell type terms into standardized ontology terms (1B).

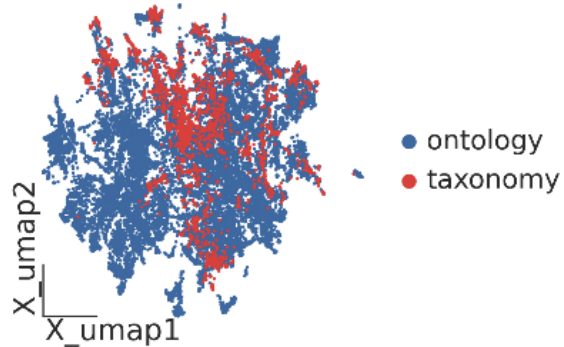




Results: Overview of CellOntologyMapper Design

The query database encompasses 19,381 curated cell type entries (1C). To investigate the semantic organization of this database, we performed dimensionality reduction using UMAP from the term embedding vector and applied unsupervised Leiden clustering, revealing 24 distinct clusters that group semantically related cell type (1D-E), validating the biological meaningfulness of our embedding approach and demonstrates the system's capacity to capture nuanced relationships within the cellular taxonomy.

(C) Cell Ontology: 168,41
Cell Taxonomy: 2,540



(E) Clusters: 7
epithelial
gland
regulation
proliferation

Clusters: 11
killer
immune
differentiation
alpha
regulation
thymocyte

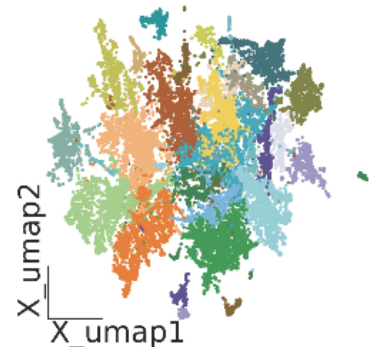
Clusters: 12
bipolar
eye
ganglion
retinal
photoreceptor

Clusters: 20
collagen
regulation
connective
tissue
fibroblast

Clusters: 21
pancreas
carcinoma
pancreatic
hepatic
duct
liver

Clusters: 22
memory
CD38
pre
B
switched

(D) Clusters



- 0
- 1
- 2
- 3
- 4
- 5
- 6
- 7
- 8
- 9
- 10
- 11
- 12
- 13
- 14
- 15
- 16
- 17
- 18
- 19
- 20
- 21
- 22
- 23

(F)

```
import omicverse as ov

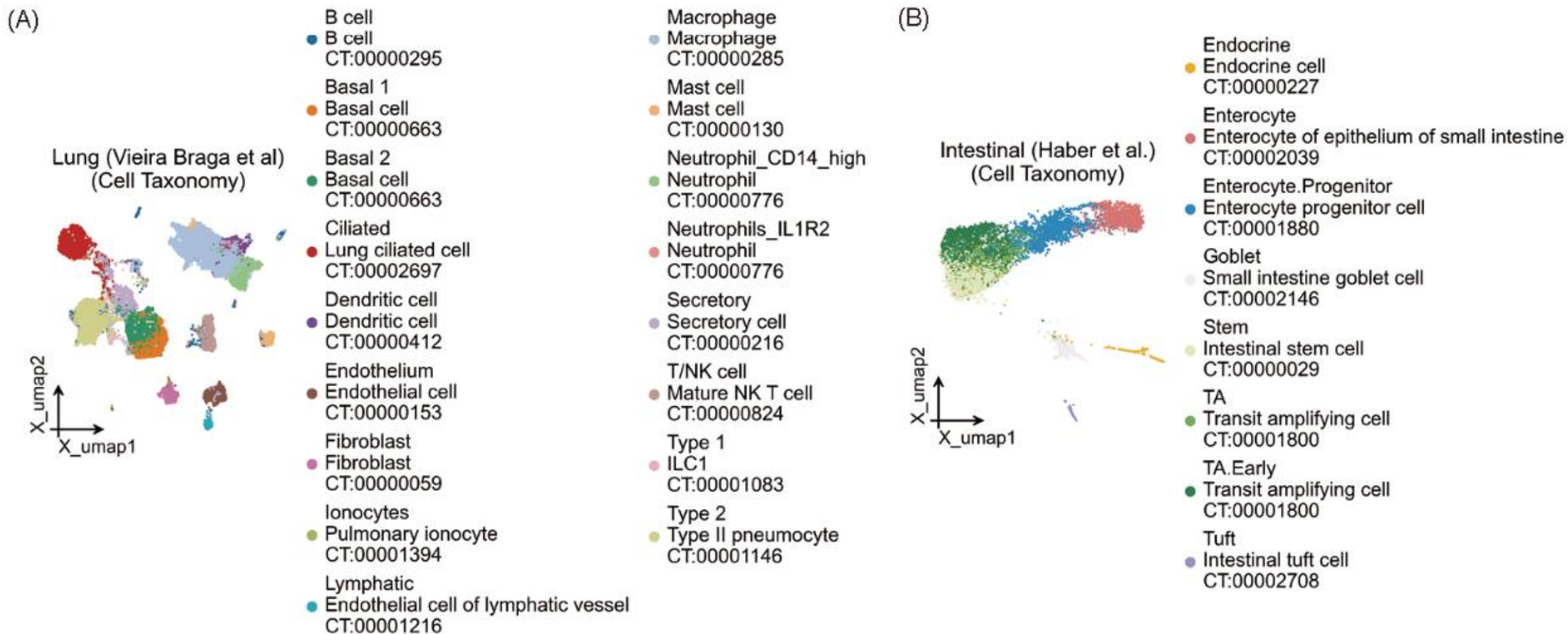
adata=ov.single.sample_data()
mapper = ov.single.CellOntologyMapper(
    cl_obo_file="new_ontology/cl.json",
    model_name="sentence-transformers/all-MiniLM-L6-v2",
)
results = mapper.map_adata_with_taxonomy(
    adata, cell_name_col="cell_label", new_col_name="enhanced_cell_ontology",
    species="Mus musculus", expand_abbreviations=True, use_taxonomy=True,
    tissue_context="Gut", threshold=0.5,
)
mapper.print_mapping_summary_taxonomy(results)
```

We provide a user-friendly API through the OmicVerse Python package to perform cell type standardization (1F), and also provide an website <https://omicverse.com/cellmapper/>. This implementation facilitates seam-less integration into existing single-cell analysis work-flows while maintaining the sophisticated mapping cap-abilities of the underlying framework.



Results: The robust cross-scale performance

To rigorously evaluate the versatility, precision and robustness, we conducted comprehensive validation using multiple independent single cell datasets spanning diverse scales, biological contexts and annotation complexities. CellOntologyMapper can accurately annotate transitional and diverse populations, and correctly identified rare populations like Transit amplifying and goblet cell subtypes, demonstrating its ability to handle nuanced biological contexts in which cell identity spans continuous developmental and differentiation spectra.



Results: The robust cross-scale performance

We subjected CellOntologyMapper to an even more stringent test using a highly complex trophoblast dataset, incorporated various embryonic cell populations and often represented by acronyms and short forms common. Despite these substantial challenges, CellOntologyMapper exhibited outstanding performance, successfully resolving ambiguous abbreviations and accurately mapping each annotated population to validated Cell Taxonomy terms. Critically, the system preserved biological coherence across distinct developmental trajectories, effectively differentiating closely related progenitor and differentiated trophoblast states, demonstrating sophisticated semantic understanding and contextual annotation accuracy.





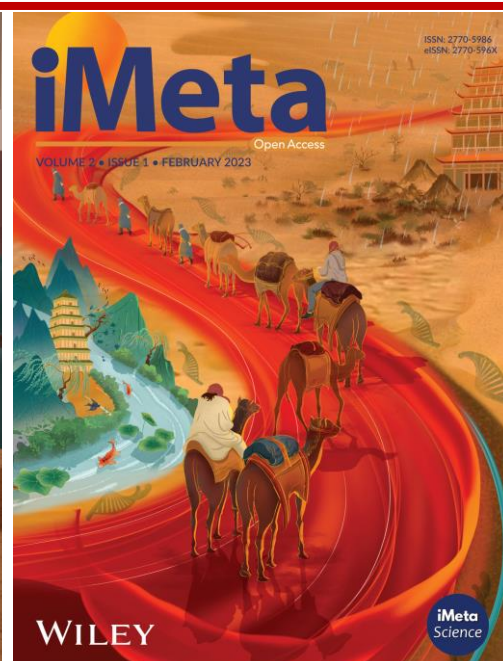
Summary

- ❑ CellOntologyMapper automatically maps diverse and often ambiguous user-defined cell type names to standardized terms from the Cell Ontology and Cell Taxonomy using advanced NLP techniques, including sentence transformers and large language models.
- ❑ The tool achieves high mapping accuracy (0.835 for CL, 0.911 for CT) across datasets of varying scales and complexities, including challenging cases with extensive abbreviations and developmental lineages.
- ❑ Available as both a Python package within the OmicVerse ecosystem and a user-friendly web interface, facilitates seamless integration into existing single-cell analysis workflows, promoting reproducible and large-scale meta-analyses.
- ❑ Website: <https://omicverse.com/cellmapper/> and <https://github.com/Starlitnightly/CellOntologyMapper>

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