



# Deploying User-Friendly Software:

## Six Recommendations to Make Single-Cell Foundation Models More Usable For Scientific Discovery



Link to Paper

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### TL;DR

- Problem:**
  - Most single-cell foundation models are not developed using industry standard software development practices.
  - Many are difficult to use, sometimes even not installable.
- Why we should care:**
  - Users cannot use these foundation models to further scientific discovery.
  - Leads to wasted time, computational resources, and work.
- What we (developers) should do:**
  - Aim to implement the 6 recommendations below when deploying foundation models.

### Common Pitfalls

- ✗ insufficient requirements.txt installation set up
- ✗ unresolvable dependency conflicts
- ✗ inactive maintenance
- ✗ ambiguous compute resource requirements
- ✗ lack of documentation

Table 1: Implementation of typical best practices in single-cell foundation model software (as of May 2025). This data serves as our rationale for our recommendations.

MODEL	CONTAINER	PYTHON PKG	CI	Docs
SCBERT	x	x	x	x
GENEFORMER	x	x	x	✓
UCE	x	x	✓	x
scGPT	x	✓	x	✓
SCFOUNDATION	x	x	x	x
SCLONG	x	x	x	x
CELLPLM	x	✓	x	x
SCIMILARITY	✓	✓	x	✓
TGPT	x	x	x	x
CELLLM	✓	x	x	x
GENECOMPASS	x	x	x	x
GENEPT	x	x	x	x
<b>TOTAL</b>	<b>2/12</b>	<b>3/12</b>	<b>1/12</b>	<b>3/12</b>



### Impact / Further Implementation Suggestions

- 1 • Removes need for users to resolve dependency conflicts.  
• Useful for models used in context with multiple programming languages.
- 2 • Improves user experience of installing foundation models.
- 3 • Make sure only safe changes are being deployed as the software gets updated over time.
- 4 • A low effort solution for models that are neither containerized nor wrapped as a Python package.
- 5 • List minimum requirements for memory, number of GPUs, number of CPUs allocated per task, and time allowance needed to successfully run the foundation model.
- 6 • Documentation should have high code coverage.

### Efforts Addressing Software Issues in Computational Biology

#### Current Existing Solutions

##### Grants

##### Essential Open Source Software for Science

Grant by the Chan Zuckerberg Initiative to fund initiatives to make open source software more sustainable.

##### Services

##### Helical AI

Python package that wraps biological foundation models in a uniform interface.

##### Garden

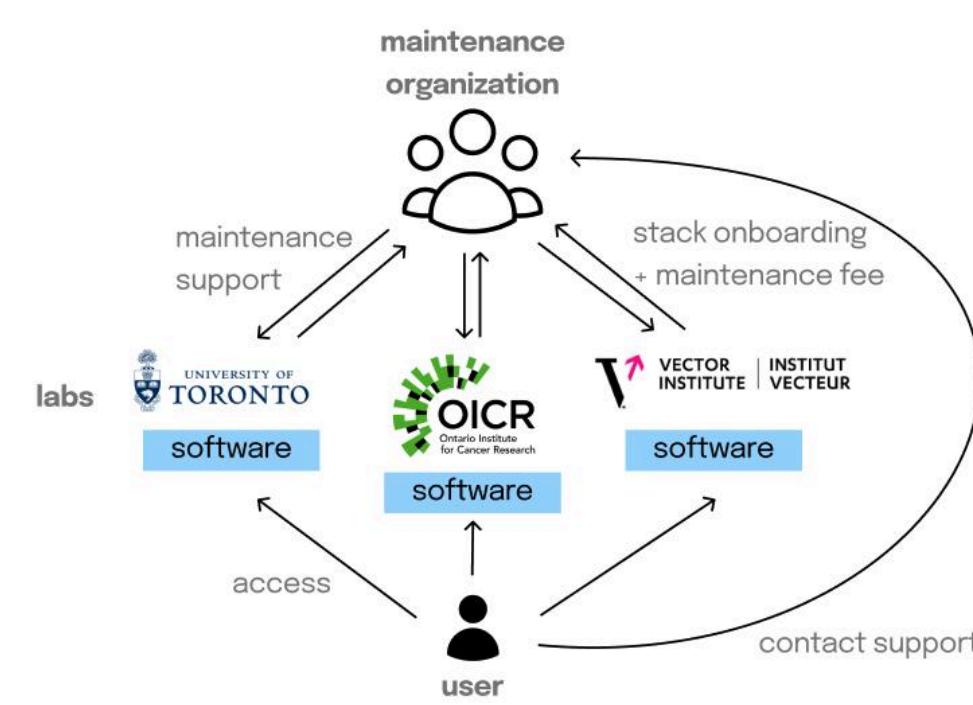
A platform that hosts AI models in a containerized environment to allow researchers to run them within a limited GPU allowance.

##### Code Ocean

Platform that hosts a wide range of bioinformatics software in a ready to use state.



#### Proposal: Centralized Software Maintenance



An organization solely dedicated to the maintenance of important software could be a sustainable solution to keep academic software usable without disrupting the established norms and incentives in academia. Potential funding sources include grants and affordable fees collected from journals/labs/users.

### Future Directions

- Automating the implementation of the 6 recommendations using LLMs.
- Collecting feedback from foundation model users and developers on the recommendations.
- Creating a website to guide new developers trying to implement the recommendations, and to showcase successful implementations.

#### References

Szalata, A., Hrovatin, K., Becker, S., Tejada-Lapuerta, A., Cui, H., Wang, B., & Theis, F. J. (2024). Transformers in single-cell omics: A review and new perspectives. *Nature Methods*, 21(8), 1430–1443. <https://doi.org/10.1038/s41592-024-0235-z>

Ziemann, M., Poulin, P., & Bora, A. (2023). The five pillars of computational reproducibility: Bioinformatics and beyond. *Briefings in Bioinformatics*, 24(6), bbad375. <https://doi.org/10.1093/bib/bbad375>