

# Henry W. Leung, Ph.D.

Machine Learning & Data Scientist

 [henrysky.github.io](https://github.com/henrysky)

 [henryskyleung@gmail.com](mailto:henryskyleung@gmail.com)

 [github.com/henrysky](https://github.com/henrysky)

 [henry-leung-2664b3259](https://www.linkedin.com/in/henry-leung-2664b3259)

 Bilingual in English & Chinese

 Canadian & Hong Konger

 Toronto, Canada

## SUMMARY

Machine Learning and Data Scientist with 6+ years of experience adapting ML techniques to large-scale datasets and 9+ years of software development in Python/C. Expertise in delivering data-driven solutions and insights for real-world applications.

## PROFESSIONAL EXPERIENCE

**Data Science Institute, University of Toronto**

Sept. 2023 – Oct. 2024

**Data Science Doctoral Fellow**

- Developed applications of **Transformers** architecture with a **denoising diffusion** probabilistic head as density function emulator for tabular data to improve non-Gaussian uncertainty estimation. Applicable to uncertainty quantification in finance and healthcare.
- Curated multiple datasets for a 100 TB machine learning-ready dataset hosted on **huggingface**, for the next-generation large ML models in physical science and industry.

**University of Toronto**

Sept. 2019 – Oct. 2024

**Graduate Researcher**

- Developed a **self-supervised** foundation model in **PyTorch** with a novel token-scalar embedding technique, outperforming XGBoost by 6% in multiple predictive tasks. Presented at NeurIPS and ICML for advancing structured tabular data learning.
- Designed an **unsupervised** encoder-decoder in **TensorFlow** to address a data-rich label-scarce challenge, by reducing data dimensionality by 1,000x through physics-guided techniques and label regression in the low-dimensional latent space.
- Public product release of ML-derived parameters estimation models with uncertainty quantification, achieving over a 10% accuracy improvement for **low signal-to-noise** data and a 100x speedup compared to non-ML pipelines.
- Built a Flask web app integrating an LLM for seamless natural language interaction with scientific models, demonstrating LLM-driven interaction with specialized tools in workflows.
- Implemented supervised **CNNs with dropout variational inference** for spectral data analysis, contributing to one of the **first direct mappings** of the Milky Way's inner structure.
- Developed and maintained open-source software that are **well-documented** with rich set of examples and **thoroughly tested** with more than 80% code coverages, as well as contributed to other open-source projects.

## TECHNICAL CONTRIBUTIONS

Below is a summary of key research contributions, highlighting those in machine learning conferences:

**NeurIPS (Oral in 2023, Collaboration Poster in 2024), ICML (Poster in 2024), AI for Astronomy (Oral in 2019)**

- Research in **ML adaptation** and **foundation models** for scientific and structured datasets with Transformers and diffusion models. ML-driven predictive modeling robust to noise to extract insights from large-scale datasets with applications in finance, healthcare, and physical sciences
- Discovery enabled by **data-driven insights**, including anomaly detection and ML-derived parameter estimation, enabling analyses otherwise difficult with traditional methods.

First/Second author on 9 refereed papers with 620+ citations among 16 refereed papers with 3060+ citations (h-index=11).

## EDUCATION

**Ph.D.**, Astronomy & Astrophysics, University of Toronto

Sept. 2020 – Oct. 2024

**M.Sc.**, Astronomy & Astrophysics, University of Toronto

Sept. 2019 – Aug. 2020

**H.B.Sc.**, Physics & Astronomy, University of Toronto

Sept. 2014 – Aug. 2019

## SKILLS & SOFTWARE

**Programming Languages:** Python, C (proficient), C++, SQL (intermediate), Rust (beginner).

**Frameworks/Packages:** Proficient in PyTorch, TensorFlow, Scikit-Learn, NumPy, SciPy, Pandas, PySpark, Jupyter Notebook, Matplotlib, SQLite, PostgreSQL, Docker, SSH, Git, Bash, Slurm, Hugo, SCSS, Node.js. Some experiences in ArcGIS and SolidWorks.

My open-source software packages used by the community that are well tested with continuous integration and well documented with docstrings and user guides, includes:

- **astroNN** (195 stars) – **Deep learning framework** with Keras for astronomical data, supporting TensorFlow and PyTorch.
- **MyGaiaDB** – **SQLite-based database** package for managing and querying of astronomical data locally.
- **Galaxy10** – **Benchmark dataset** for CNN-based galaxy classification, used in ML research and education.