

# Chris Pedersen

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## RESEARCH INTERESTS

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I am a research scientist in deep learning. My expertise is in the construction of bespoke machine learning models to solve problems across science, including cosmology, bioinformatics, and turbulence modelling in fluid dynamics.

## EXPERIENCE

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**Courant Institute of Mathematical Sciences, New York University** September 2022 - present  
Postdoctoral Associate

*Topics:* Developing new ML approaches to improve the accuracy and robustness of climate models

**Center for Cosmology and Particle Physics, New York University** September 2021 - August 2022  
Postdoctoral Associate

*Topics:* Applications of statistical techniques and deep learning to challenges in astrophysics.

**Center for Computational Astrophysics, Flatiron Institute** September 2021 - present  
Guest Researcher

## EDUCATION

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**University College London** September 2017 - August 2021  
Ph.D. in Astrophysics

- Primarily responsible for a 4.5M CPU hour allocation on a Tier 1 machine (code performance evaluation, running simulations, data management, postprocessing, scientific analysis and publications.)
- Development of simulation code in C/C++, construction of Bayesian inference and Gaussian process modelling pipelines in Python.

**Cardiff University** September 2012 - July 2017  
MPhys in Physics with Astronomy, First class honours

- Bayesian inference using Markov chain Monte Carlo simulations in Python, in the context of gravitational wave astronomy.

## ML PROJECTS LED

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**Diffusion modelling for turbulent flows (2023 - present):**

- Novel application of diffusion model framework to stabilise the neural emulation of turbulent fluid flows (project ongoing)

**Neural emulation for improved turbulence modelling (2023):**

- Responsible for model design and implementation, construction of hybrid MLops and simulation analysis pipeline (PyTorch, Weights and Biases).
- Led to a [workshop paper at ICML 2023](#) improving the stability and computational cost of turbulence simulations.

**Cancer-net (2022 - 2023):**

- Using graph neural networks to model the progression of prostate and brain cancer based on the genetic mutations within the tumor.
- Responsible for model design and implementation in PyTorch, model optimisation, and performance validation (submitted to [Nature Machine Intelligence](#), and [publicly available code](#)).

## Wavelet scattering for cosmological inference (2022):

- Created a new wavelet based CNN to extract information from cosmological data, significantly outperforming a standard CNN in the low-data regime.
- Led to a [workshop paper at ICML 2022](#) and [publicly available code](#).

## Cosmology emulator (2020-2021):

- Built a surrogate model (or *emulator*) using Gaussian processes, enabling accurate Bayesian parameter inference of observational data from just 30 training simulations.
- Led to a [publication](#) and [publicly available code](#).

## TECHNICAL SKILLS

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Computational skills	Python, Linux/Bash, git, LaTeX, High Performance Computing, PyTorch, TensorFlow, Jax, SciPy, scikit-learn, numerical methods
Statistical & ML techniques	Bayesian inference, Markov chain Monte Carlo simulations, linear regression, logistic regression, SVM, PCA, Gaussian processes, <a href="#">convolutional neural networks</a> , graph neural networks, autoencoders, wavelet scattering networks diffusion models, <a href="#">data parallelism (DDP)</a> , <a href="#">model parallelism (Deepspeed)</a>
Software development	<a href="#">cancer-net</a> , <a href="#">torch-qg</a> , <a href="#">LearnableWavelets</a> , <a href="#">LaCE</a> , <a href="#">cup1d</a> , <a href="#">MP-Gadget</a> (Contributor), <a href="#">fake_spectra</a> (Contributor)

## SELECTED PUBLICATIONS

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Complete list available at [available here](#)

- **C. Pedersen**, T. Tesileanu, T. Wu, S. Golkar, M. Cranmer, Z. Zhang, S. Ho  
*Reusability report: Prostate cancer stratification with diverse biologically-informed neural architectures*,  
[Submitted to Nature Machine Intelligence](#)
- **C. Pedersen**, L. Zanna, J. Bruna, P. Perezhogin  
*Reliable coarse-grained turbulent simulations through combined offline learning and neural emulation*,  
[ICML 2023 Synergy of Scientific and Machine Learning Modeling Workshop](#)
- **C. Pedersen**, A. Font-Ribera, N. Y. Gnedin  
*Compressing the cosmological information in one-dimensional correlations of the Lyman- $\alpha$  forest*,  
[Astrophysical Journal \(2023\)](#)
- **C. Pedersen**, M. Eickenberg, S. Ho  
*Learnable wavelet neural networks for cosmological inference*,  
[ICML 2022 Machine Learning for Astrophysics Workshop](#)
- **C. Pedersen**, A. Font-Ribera, K. K. Rogers, P. McDonald, H. V. Peiris, A. Pontzen, A. Slosar  
*An emulator for the Lyman- $\alpha$  forest in beyond- $\Lambda$ CDM cosmologies*, [JCAP \(2021\)](#)
- **C. Pedersen**, A. Font-Ribera, T. D. Kitching, P. McDonald, S. Bird, A. Slosar, K. K. Rogers, A. Pontzen  
*Massive neutrinos and degeneracies in Lyman-alpha forest simulations*, [JCAP \(2020\)](#)