# Matthieu Darcy

Website: www.matthieudarcy.com Email: mdarcy@caltech.edu GitHub: github.com/MatthieuDarcy LinkedIn: matthieu-darcy-88290a18a

#### EDUCATION

Caltech Pasadena, California

Ph.D. student in the Computing and Mathematical Sciences department.

Advisor: Prof. Houman Owhadi.

External affiliation: NASA Jet Propulsion Lab - Research Affiliate.

ENS Paris-Saclay - Institut Polytechnique de Paris

Mathématiques Vision et Apprentissage, with Highest Honors.

Imperial College London

MSc Applied Mathematics, with Distinction.

King's College London

BA Mathematics and Philosophy, first class.

2020-2021

Paris, France

2021-present

London, United Kingdom

2019-2020

London, United Kingdom

2016-2019

## EXPERIENCE

Johnson & Johnson (Janssen Pharmaceuticals) - Machine learning consultant.

05/23 - 03/24

Machine learning and mathematical modeling of disease progression.

- Developed a dynamical systems approach for the prediction and assimilation of Alzheimer's disease progression.
- Paper forthcoming.

#### French Commission for Atomic Energy (CEA) - Research intern,

05/2021-08/2021

Thesis: Deep learning for hexahedral meshing.

 Applied a convolutional neural network approach to the automatic generation of high-quality meshes for the simulation of physical systems.

#### Imperial College London - Research project.

05/2020-09/2020

Application of Kernel Flows to regression.

Developed python code for the Kernel Flows algorithm and its applications.

#### TECHNICAL SKILLS

- Programming: Python (proficient), Julia (intermediate), LaTex (proficient).
- Packages: Extensive experience in Numpy, Pytorch, Sklearn, Scipy, Jax, Pandas. Familiar with TensorFlow.
- Statistical Tools: Extensive experience in Gaussian Processes and Kernel methods, proficient in neural networks, statistics and machine learning.

#### Research Interests

I am broadly interested in scientific machine learning, specifically in the applications of Gaussian processes, kernel methods, and wavelets to the inference and predictions of stochastic (partial) differential equations and dynamical systems.

- Stochastic (Partial) Differential Equations: inference and prediction of stochastic (partial) differential equations.
- Operator Learning: learning non-linear operators using kernel methods, with applications to PDEs and integro-functional equations.
- Dynamical Systems: learning and predicting dynamical systems from data.
- Generative modeling: multiscale approaches to generative models for partial differential equations.

## Publications and Preprints

- [1] M. Darcy, E. Calvello, R. Baptista, H. Owhadi, A. M. Stuart, and X. Yang, Solving roughly forced nonlinear pdes via misspecified kernel methods and neural networks, 2025. arXiv: 2501.17110.
- [2] P. Batlle, M. Darcy, B. Hosseini, and H. Owhadi, "Kernel methods are competitive for operator learning", *Journal of Computational Physics*, vol. 496, p. 112549, 2024, ISSN: 0021-9991.
- [3] M. Darcy, B. Hamzi, G. Livieri, H. Owhadi, and P. Tavallali, "One-shot learning of stochastic differential equations with data adapted kernels", *Physica D: Nonlinear Phenomena*, vol. 444, p. 133 583, 2023.
- [4] M. Darcy, B. Hamzi, J. Susiluoto, A. Braverman, and H. Owhadi, Learning dynamical systems from data: A simple cross-validation perspective, part ii: Nonparametric kernel flows, Dec. 2021.

# TEACHING AND OUTREACH

•	SIAM chapter - Vice-President Organization of student talks and seminars.	2023
•	Teaching Assistant for graduate-level courses at Caltech ACM 118: Stochastic Processes and Regression, 2023.	2023
•	Refresher course lecturer for incoming graduate students at Caltech  Developed and taught a course reviewing linear algebra and functional analysis.	2022

### LANGUAGES

• English: native.

• French: native.