Matthieu Darcy

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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.

Paris, France 2020–2021

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] R. Baptista, E. Calvello, M. Darcy, 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.