



## Languages

|         |        |
|---------|--------|
| English | Fluent |
| French  | Native |
| Dutch   | Native |

## Awards

- **2025** IMCN Best Thesis Award
- **2021** F.R.S-FNRS Aspirant Renewal Fund for Scientific Research
- **2019** F.R.S-FNRS Aspirant Fund for Scientific Research
- **2018** Lhoist Berghmans MIT-UCL grant
- **2014** 'Vlaamse Fysica Olympiade' Finalist

## Technical Expertise

### ML / Deep Learning:

PyTorch, PyTorch Geometric, Lightning, TensorFlow, JAX, scikit-learn; generative models, geometric deep learning, active learning, uncertainty-aware and multi-fidelity modeling

### Scientific Computing:

Python, NumPy, pandas, matplotlib, Plotly, high-throughput workflows and scientific data pipelines

### Atomistic Simulation:

ABINIT, VASP

### Tooling:

Git, GitHub, Bash, Linux, LaTeX

## Leadership & Collaboration

- Independent research execution
- Scientific communication, community representation, and meeting facilitation
- Mentoring and team coordination
- Cross-disciplinary problem solving
- Resilience in ambiguous research settings

## Hobbies

- Videography with self-built drones
- Tennis
- Music production

# PIERRE-PAUL DE BREUCK

Research Scientist in Generative ML for  
Materials Discovery

pierrepauldb.com  
 debreuckpp@gmail.com

## EXPERIENCE

### Postdoctoral Researcher

August 2024 - present

#### ICAMS, Ruhr University Bochum

Leading generative-ML and atomistic-modelling projects for energy materials, including database infrastructure, model development, and high-throughput discovery workflows for superconductors and hydrides.

### ML Research Intern

October 2023 - February 2024

#### Mila, Quebec Artificial Intelligence Institute, Montreal, Canada

Developed generative models with GFlowNets for crystal-structure design in electrocatalysis and solid-state batteries in Yoshua Bengio's lab.

### PhD in Machine Learning for Materials Discovery 2019-2024

#### Université Catholique de Louvain, Belgium

Developed and benchmarked uncertainty-aware, multi-fidelity, and active-learning methods for materials-property prediction and accelerated discovery workflows, resulting in 3 first-author publications and 5 conference presentations, including an invited talk.

### President of the researchers association (ACIM) 2021-2023

#### IMCN institute, Université Catholique de Louvain, Belgium

Represented PhD researchers to institute leadership, led monthly feedback meetings, and organized community-building events.

### Research Intern (Graph Neural Networks) July-August 2018

#### MIT, Department of Material Science and Engineering, Cambridge USA

Worked on geometric deep-learning methods for materials applications in a collaborative research setting.

## EDUCATION

### PhD in Machine Learning for Materials Discovery 2019-2024

#### Université Catholique de Louvain, Belgium

*Dissertation: 'Small datasets, big predictions: learning methods for uncertainty-aware modelling of multi-fidelity material properties'*

### Master of Engineering Science in applied physics 2019

#### Université Catholique de Louvain

*Magna Cum Laude with honours. Master Thesis on Machine Learning in Materials Science.*

### Bachelor of Engineering Science

2014-2017

#### Université Catholique de Louvain

*Magna Cum Laude*

## TEACHING

- Teaching assistant quantum mechanics (LMAPR 1491) - 3rd year BSc Engineering** 2020-2022  
Université Catholique de Louvain
- Intro to Supervised Learning, Machine Learning for Electronic Structure Training** 2021-2023  
School  
ICTP-East African Institute for Fundamental Research under the auspices of UNESCO
- Intro to Python for 16-18 years old students** Summer 2017  
Technofutur TIC
- Mathematics, Physics and Chemistry Tutor - 1st & 2nd year Engineering** 2015-2017  
Université Catholique de Louvain

## SELECTED PRESENTATIONS

- Talk at AI4AM: AI for advanced materials - Madrid, Spain** May 2026  
*A generative material transformer using Wyckoff representation*
- Invited talk at WE-Heraeus Seminar: Machine Learning for Spectroscopy - Brussels, Belgium** May 2025  
*Generative Transformer models for the dielectric function*
- Invited talk at CECAM Machine Learning of First Principle Observables - Berlin, Germany** July 2024  
*Property predictions from limited and multi-fidelity datasets*
- Contributed talk at the APS March Meeting 2022 - Chicago, USA** March 2022  
*Bias-imbalance in data-driven materials science: a case study on MODNet*
- Contributed talk at the 17th ETSF Young Researchers' Meeting - Cagliari, Italy** September 2021  
*MODNet: property prediction for limited datasets and the bias-imbalance issue*
- Invited talk at CECAM Mixed-Gen workshop** April 2021  
*Accurate and interpretable property prediction for limited materials datasets by feature selection and joint-learning*
- Contributed talk at the APS Online March Meeting 2021** March 2021  
*MODNet: property prediction for limited materials datasets by feature selection and joint-learning*
- Poster presentation at the 2020 Virtual MRS Fall Meeting** November 2020  
*Machine Learning Materials Properties for Small Datasets. Symposium: Data Science and Automation to Accelerate Materials Development and Discovery*

## PUBLICATIONS

- Machine learning driven exploration of hydride superconductors at ambient pressure**  
P. R. Pires, T. H. B. Da Silva, K. Gao, K. H. Hiorth, T. F. T. Cerqueira, T. Cavignac, **P.-P. De Breuck**, H.-C. Wang, D. Dangić, Y.-W. Fang, A. Sanna, W. Cui, I. Errea, P. Törmä, and M. A. L. Marques  
**Computational Materials Today, 100052 (2026)**
- AI-Driven Expansion and Application of the Alexandria Database**  
T. Cavignac, J. Schmidt, **P.-P. De Breuck**, A. Loew, T. F. T. Cerqueira, H.-C. Wang, A. Bochkarev, Y. Lysogorskiy, A. H. Romero, R. Drautz, S. Botti, and M. A. L. Marques.  
**[Submitted for publication]**
- A generative material transformer using Wyckoff representation**  
**P.-P. De Breuck**, H.A. Piracha and M.A.L. Marques  
**npj. Comput. Mater. (2026)**

● **Combining feature-based approaches with graph neural networks and symbolic regression for synergistic performance and interpretability**

R. A. Gouvêa, **P.-P. De Breuck**, T. Pretto, G.-M. Rignanese, and M. J. L. Santos

**npj. Comput. Mater. (2025)**

● **Generative AI for Crystal Structures: A Review**

**P.-P. De Breuck**, H.-C. Wang, G.-M. Rignanese, S. Botti and M.A.L. Marques

**npj. Comput. Mater. (2025)**

● **High-Throughput Search for Cubic Chiral Semiconductors: Structural Insights, Database and Stability Trends**

**P.-P. De Breuck**, H.-C Wang, A. Tellez-Mora, M.A.L. Marques and A. H. Romero

**[Under preparation]**

● **Optical materials discovery and design via federated databases and machine learning**

V. Trinquet, M. Evans, C. Hargreaves, **P.-P. De Breuck**, and G.-M. Rignanese

**Faraday Discuss. (2024)**

● **Combination of ab initio descriptors and machine learning approach for the prediction of the plasticity mechanisms in  $\beta$ -metastable Ti alloys**

M. Coffigniez, **P.-P. De Breuck**, L. Choisez, M. Marteleur, M. J. Van Setten, G. Petretto, G.-M. Rignanese, and P. J. Jacques

**Materials & Design 239, 112801 (2024)**

● **Influence of roughness and coating on the rebound of droplets on fabrics**

P. J. Cruz, **P.-P. De Breuck**, G.-M. Rignanese, K. Glinel, A. M. Jonas

**Surfaces and Interfaces 36, 102524 (2023)**

● **A simple denoising approach to exploit multi-fidelity data for machine learning materials properties**

X. Liu, **P.-P. De Breuck**, L. Wang, G.-M. Rignanese

**npj Comput. Mater. 8, 233 (2022)**

● **Accurate experimental band gap predictions with multifidelity correction learning**

**P.-P. De Breuck**, G. Heymans, G.-M. Rignanese

**J Mater. Inf. 2, 10 (2022)**

● **Robust model benchmarking and bias-imbalance in data-driven materials science: a case study on MODNet**

**P.-P. De Breuck**, M. L. Evans, G.-M. Rignanese

**J. Phys.: Condens. Matter 33, 404002 (2021)**

● **Materials property prediction for limited datasets enabled by feature selection and joint learning with MODNet**

**P.-P. De Breuck**, G. Hautier, G.-M. Rignanese

**npj Comput. Mater. 7, 83 (2021)**

● **Vibrational properties of solids : a machine learning approach**

**P.-P. De Breuck**, G.-M. Rignanese

**Master Thesis (2019)**