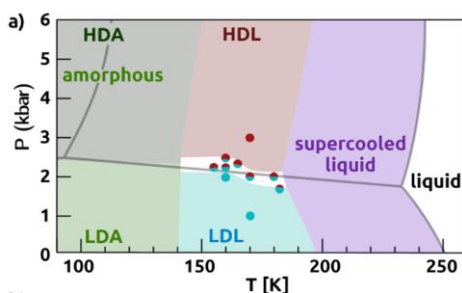


## Master 2: *International Centre for Fundamental Physics* **INTERNSHIP PROPOSAL**

Laboratory name: Institut de Minéralogie, de Physique des Matériaux et de Cosmochimie  
CNRS identification code: IMPMC – Sorbonne University  
Internship director's surname: A. Marco SAITTA – PR Classe Exceptionnelle  
e-mail: marco.saitta@sorbonne-universite.fr Phone number: 0144275236  
Internship codirector's surname: Frédéric DATCHI – IMPMC (SU)  
Internship location: IMPMC – SU – Campus Pierre et Marie Curie  
Thesis possibility after internship: YES  
Funding: Very likely (Labex/Idex), or ED's + ANR If YES, which type of funding:

### ***Machine learning methods for polymorphism and critical phenomena in exotic liquids***

First-order phase transitions are common in the solid state but for a long time judged incompatible with the nature of the liquid state. About 20 years ago, such a liquid-liquid transition (LLT) was discovered in elemental phosphorus, which has had a large impact in the scientific community as it



changed the way the liquid state was perceived. Since then, no other example of a LLT separating two thermodynamically stable liquid phases of a pure substance could be experimentally evidenced, until a recent experimental discovery of a LLT in compressed liquid sulfur by one of us [L. Henry et al, Nature 584 (2020)]. We have recently proved, by theoretical calculations, that water, instead, does not exhibit a LLT [PRX, under review, see figure], ending a 30-year long debate in the community.

Building on our recent experimental and theoretical breakthroughs, achieved thanks to state-of-the-art free-energy methods, we are strengthening our approaches through the in-house ongoing development of **quantum accuracy-level machine learning** potentials, capable to address challenges in the study of chemical reactions in more and more complex and realistic environments. In this internship project, which will lead to a PhD thesis, we will develop a machine learning (ML) description of liquid sulfur, trained on ab initio simulations; the ML description will allow us to explore the sulfur phase diagram (both in its liquid and solid phases) at an unprecedented level of both accuracy and complexity.

We look for a student willing to undertake these innovative methods and determined to carry out this project within a strong collaboration between theory and experiments. We have a consolidated expertise and a strong publication record including in the last few years, in the theoretical part only, 8 PNAS, 5 PRL, 4 SciRep, 4 J. Phys. Chem. Lett. 2 Nature Comm, 2 Phys Life Rev, 1 Nature Methods, 1 ChemSci.

**Techniques/methods in use:** statistical analysis, clustering, inference, local and global descriptors, ab initio molecular dynamics

**Applicant skills:** strong background in statistical physics, interest in computational methods, knowledge on transformations in condensed matter.

**Key words:** theory, statistical mechanics, simulations, machine learning

**The supervisor “in action”:**



Video from my General Seminar at the Physics Department of ENS

Condensed Matter Physics: YES

Macroscopic Physics and complexity: YES

Quantum Physics: YES

Theoretical Physics: YES