

# PhD interview presentation

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Master's thesis midway presentation  
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# Academic Background

Summary and interest?

## Education

- Bachelor in physics
- Master in Computational Science: Materials Science

## Interest

- Computational approach
- Design and creation
- Optimization problems
- Machine learning

# Programming experience

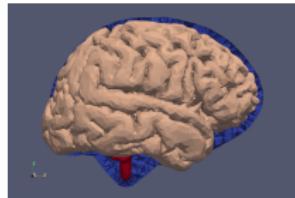
## High level languages

### Python

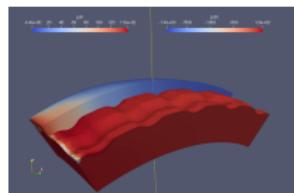
- Workhorse of all my work.
- Object oriented code.
- Machine learning through PyTorch and TensorFlow.

### Julia

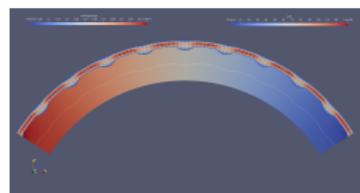
- Alternative to Python
- Finite element analysis using Gridap
- Mesh geometris using Gmsh



(a)



(b)



(c)

# Programming experience

## Lower level languages

### C++

- Object oriented code.
- Project work in computational physics.
  - Eigenvalue problems and matrix operations.
  - Ordinary and partial differential equations
  - Integration
  - Simulation of stochastic systems.
- Two years teaching assistant.

### C

- High performance computing
  - General code optimization.
  - Data traffic and cache management.
  - Parallel programming (MPI and OpenMP).
  - Modern data architecture.

# Programming experience

## Molecular dynamics

### LAMMPS

Large-scale Atomic/Molecular Massively Parallel Simulator

- Statistical mechanics and thermodynamics
- Structure recognition
- Porous systems
- Contact forces
- Friction properties

# Programming experience

## Machine learning (ML)

### Pytorch (/ TensorFlow)

- Basic ML concepts:
  - Optimizers
  - Loss / cost functions
  - General learning strategies (learning rates, drop out, normalization etc.)
- Data handling and augmentation.
- Fundamental architecture types
  - Feed Forward Networks
  - CNN
  - RNN
  - GAN
- Network performance analysis
- Basic basic techniques for AI predictions explanations
  - Shapley values
  - Gradient linearization methods

# Master's thesis

## Scope

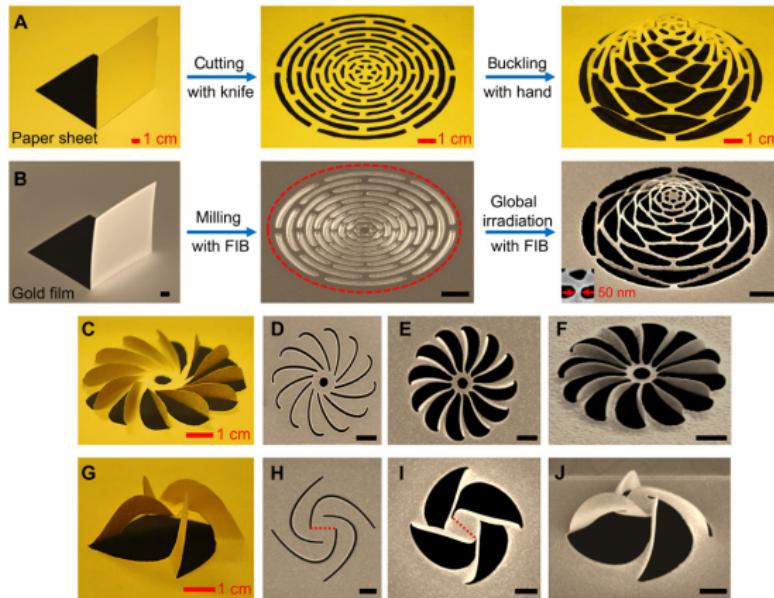
Tuning frictional properties of graphene sheets using kirigami inspired cuts and inverse design

- ① Sheet kirigami: Alter graphene sheet using atomic scale cuts.
- ② Forward simulation: Calculate frictional properties of the sheet using MD simulations.
- ③ Inverse design: Predict cut patterns based on frictional properties and optimize for desired properties using machine learning.
  - Low / high friction coefficient.
  - Coupling between stretch and friction.

# Master's thesis

## Motivation

Kirigami: Variation of origami with cuts permitted.

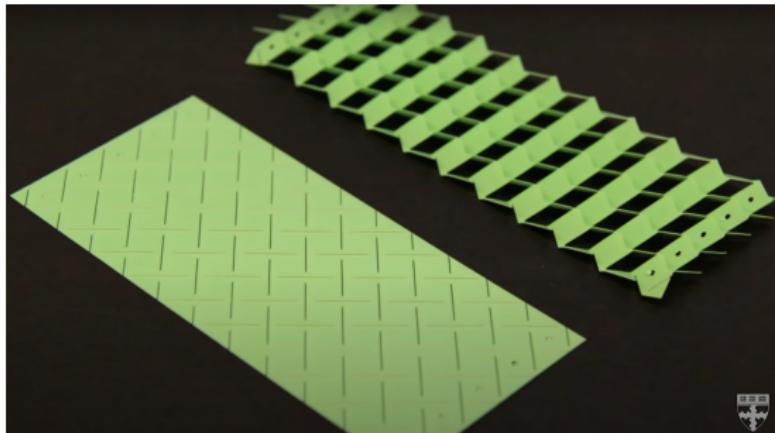


**Figure:** Example of transition from macro- to nano-kirigami using a focused ion-beam (FIB) (Nano-kirigami with giant optical chirality, ZHIGUANG LIU, 2018).

# Master's thesis

## Choosing a cut pattern

- Kirigami design on macroscale.



**Figure:** New pop-up strategy inspired by cuts, not folds - Leah Burrows, Harvard John A. Paulson School of Engineering and Applied Sciences.

# Master's thesis

Translating patterns to MD

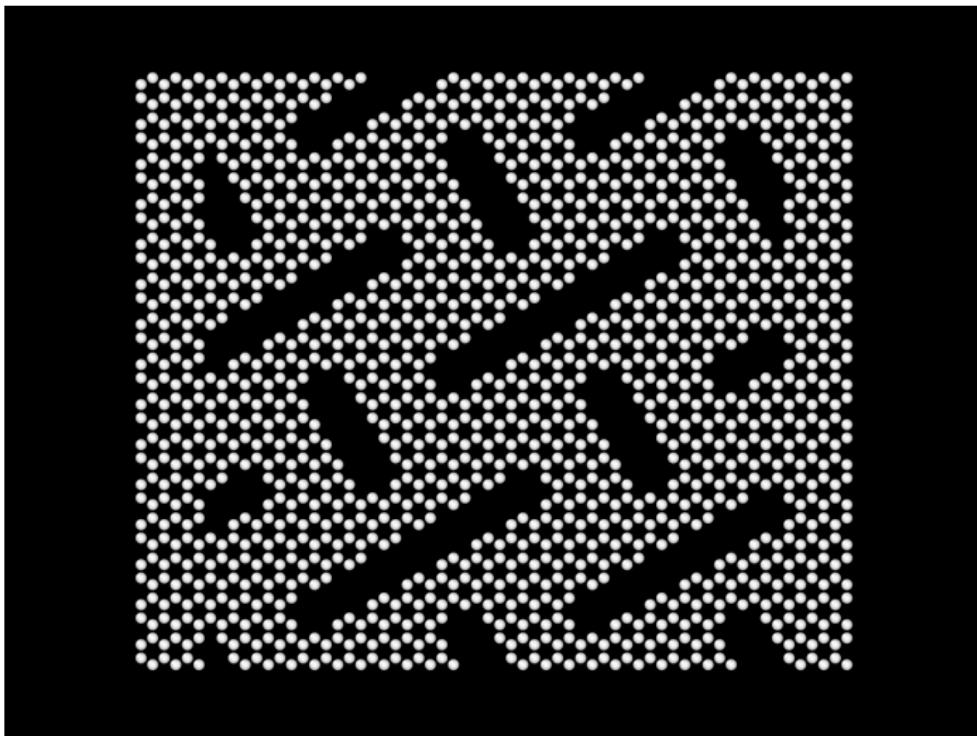


Figure: Kirigami sheet stretch in vaccuum.

# Master's thesis

## Investigating 3D buckling

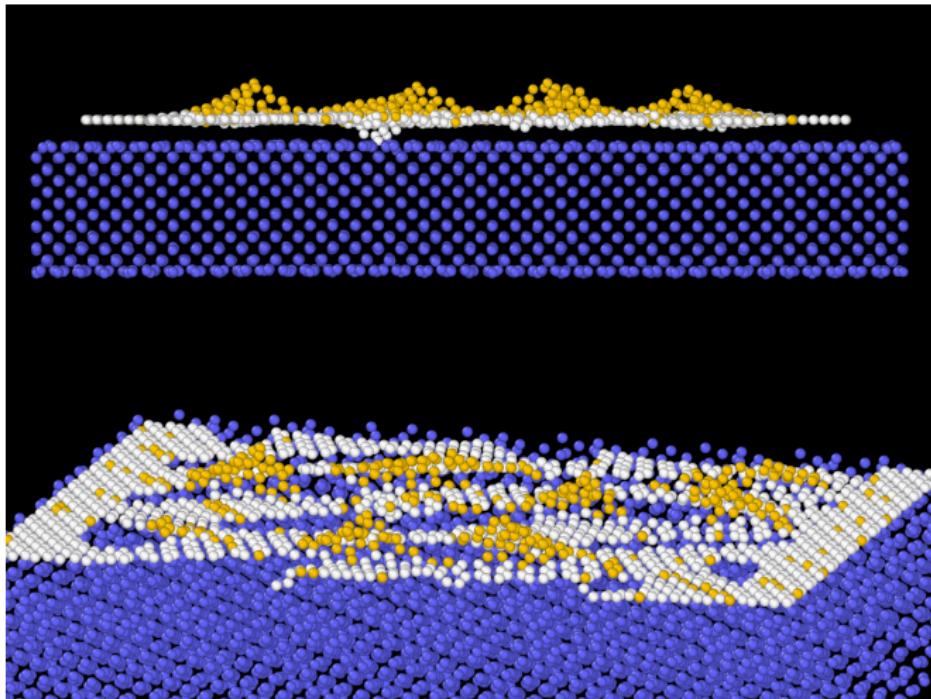
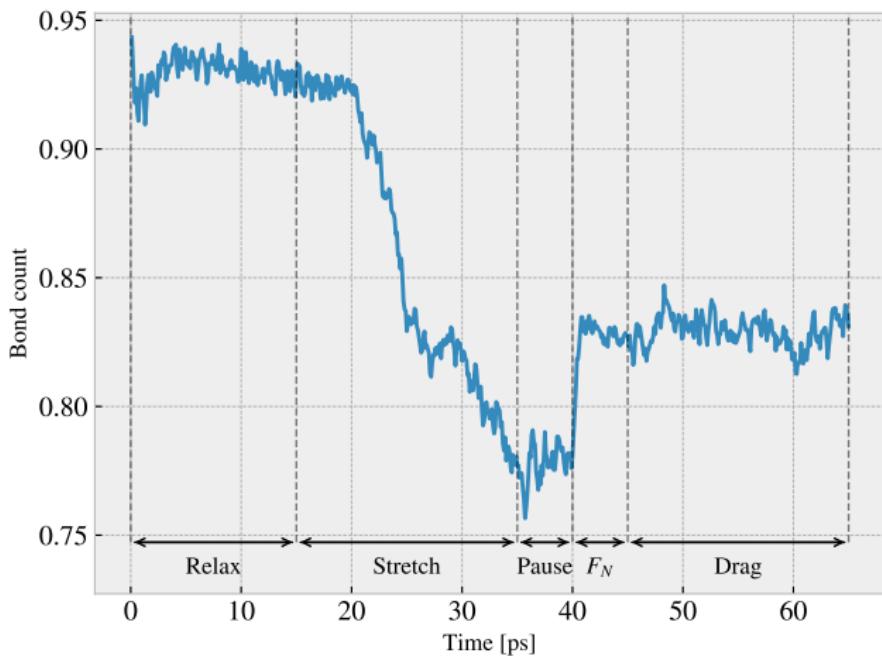


Figure: Kirigami stretch in contact with Si-substrate.

# Master's thesis

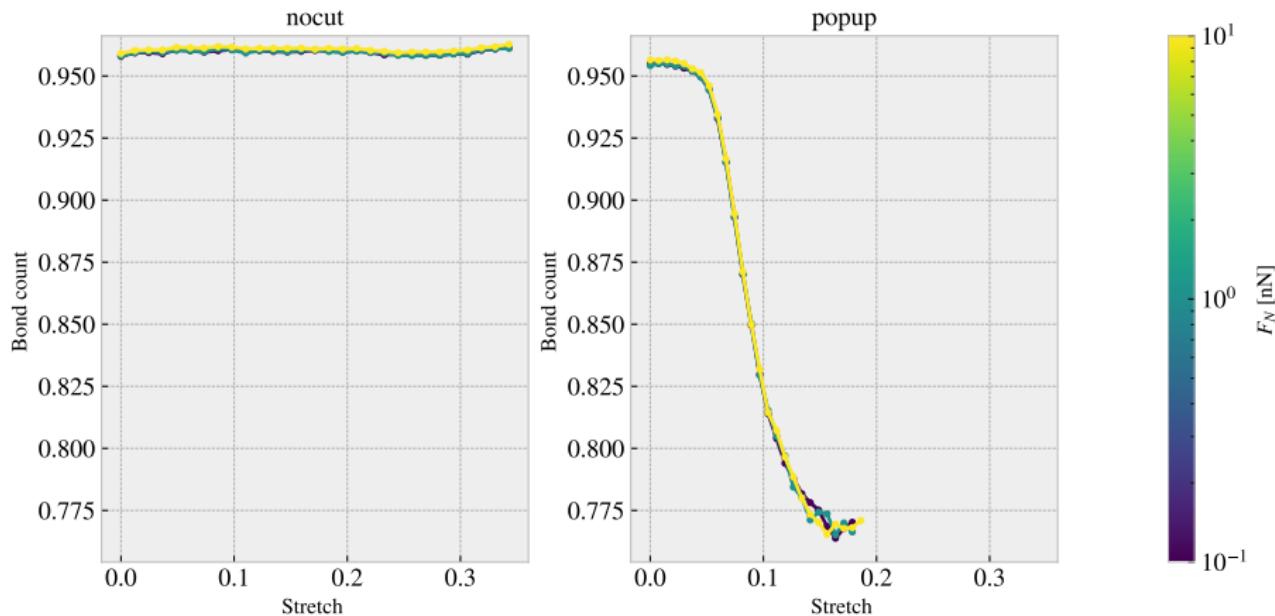
## Investigating 3D buckling



**Figure:** Contact area approximation: Number of C-Si bonds within a threshold distance of 110% the LJ interaction equilibrium distance.

# Master's thesis

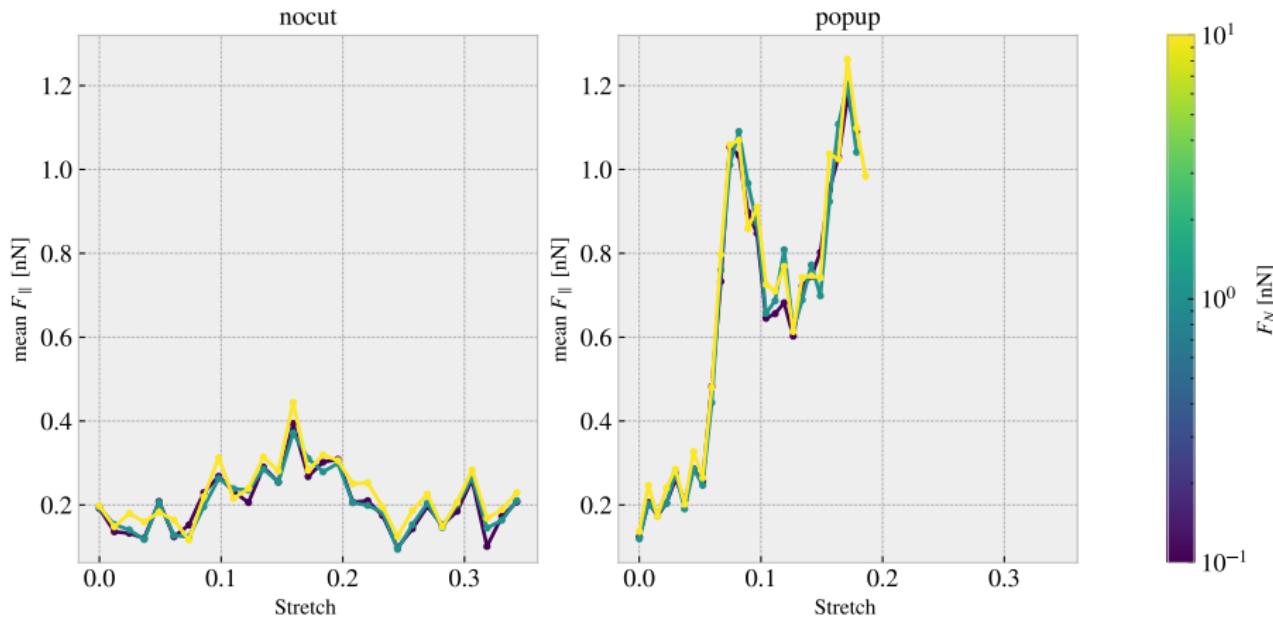
## Friction stretch dependency



**Figure:** Relative number of bonds between sheet and substrate as a function of stretch of the sheet with and without cuts.

# Master's thesis

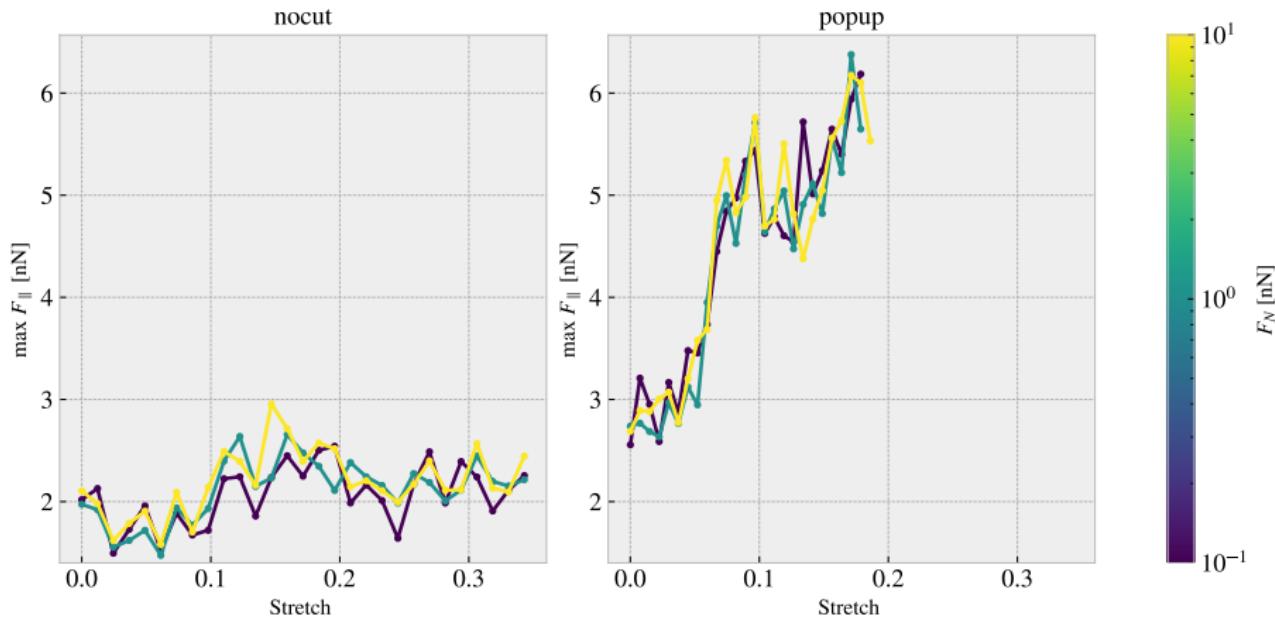
## Friction stretch dependency



**Figure:** Mean friction force  $F_{\parallel}$  parallel to drag direction as a function of stretch of the sheet with and without cuts.

# Master's thesis

## Friction stretch dependency

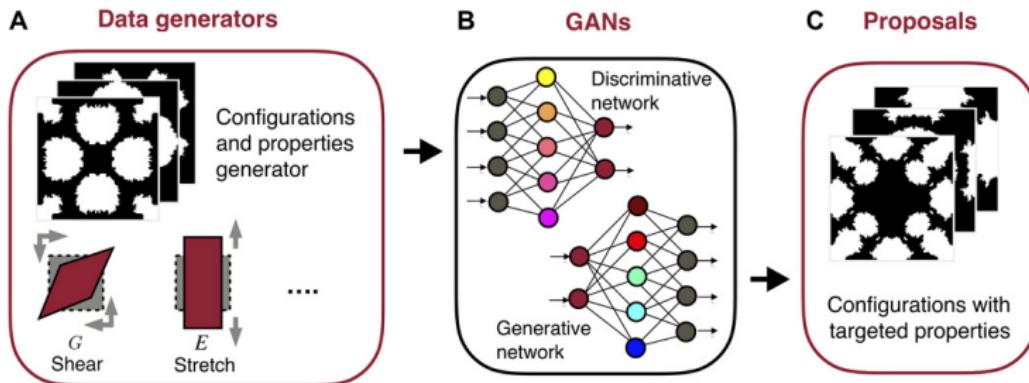


**Figure:** Max friction force  $F_{\parallel}$  parallel to drag direction as a function of stretch of the sheet with and without cuts.

# Master's thesis

## Inverse design

*Designing complex architectured materials with generative adversarial networks, YUNWEI MAO, 2020.*



**Figure:** (A) Data generators to generate datasets of configurations and properties of architectured materials. (B) GANs trained by the datasets. (C) New designs of architectured materials with the targeted properties proposed by the GANs.

# Master's thesis

Possible application: Nanomachine for negative friction coefficient

$$\left. \begin{array}{l} \text{Normal force : } F_f = k \cdot F_N \\ \text{Stretch : } F_f \sim s \cdot \text{stretch} \\ \text{Nanomachine : stretch} = \pm R \cdot F_n \end{array} \right\} \Rightarrow F_f \propto \underbrace{(k \pm sR)}_{\mu} \cdot F_n$$

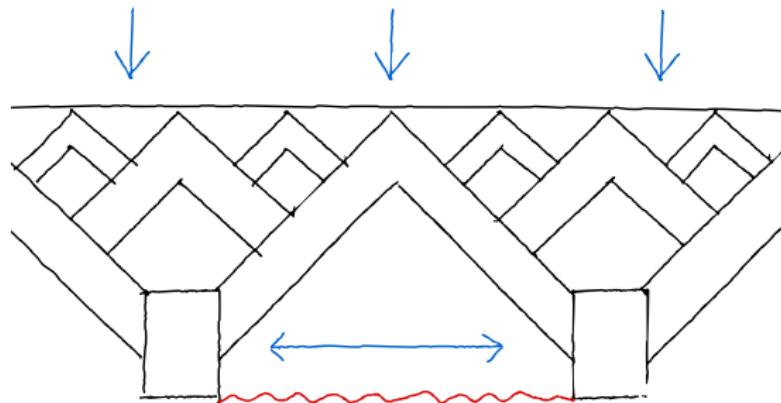


Figure: Sketch for nanomachine coupling normal force and stretch. Black represents nanomachine components and red the sheet.