

Applications of Topological Data Analysis to Condensed Matter and High Energy Physics

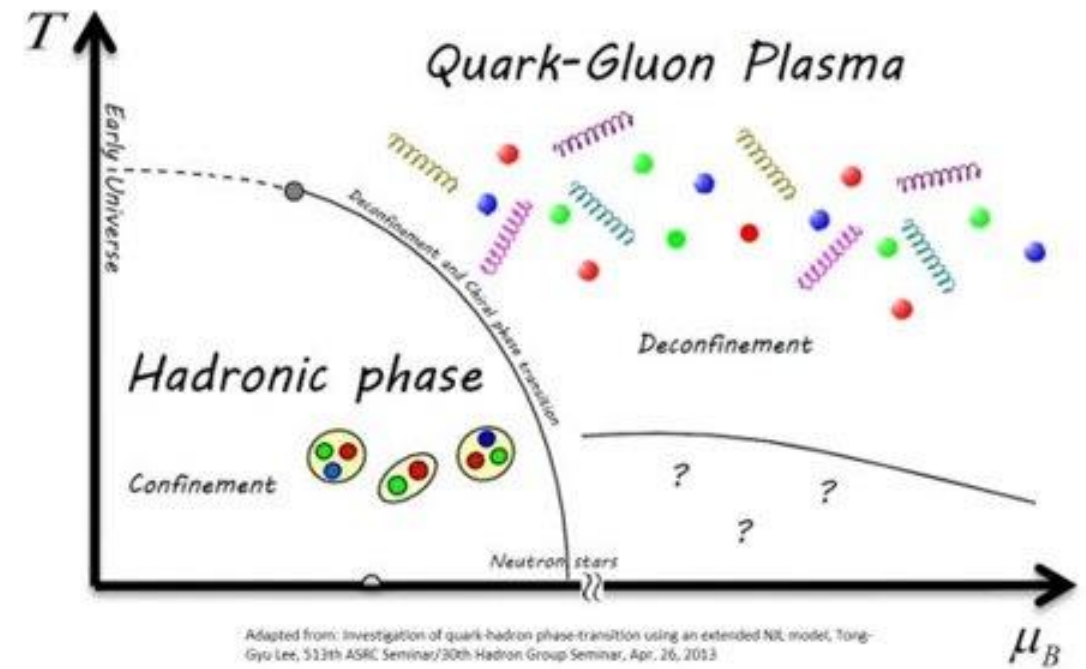
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Motivation

- Quantum Chromodynamics (QCD)
- Mechanism for confinement and the deconfinement phase transition
- Center vortices
- Want new data analysis tools to study these in sampled configurations



Outline

- Topological data analysis and persistent homology
- Vortices in the XY model and the BKT transition
- Quantitative analysis via persistent homology

- Center vortices in SU(2) lattice gauge theory
- Detecting center vortices with persistent homology

- The road to QCD and other future work

Topological data analysis

- Tagline: applying tools from topology to learn the "shape of data"
- Topology is about connectivity (but TDA tells us about geometry too)
- Applications across the sciences
- Several different tools, but principally persistent homology



Persistent Homology

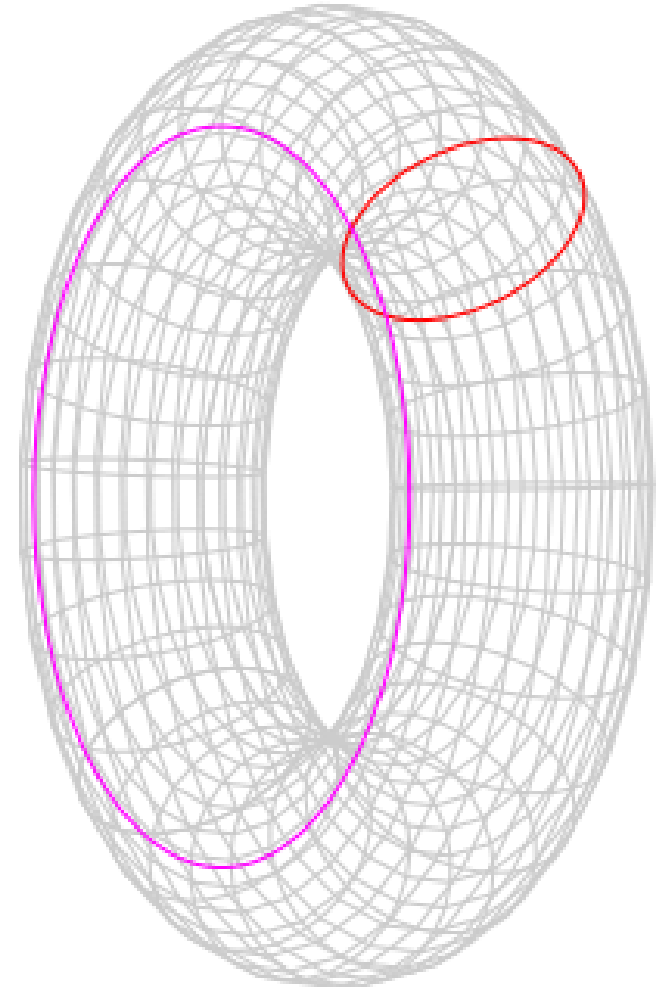
- Homology

- Vector space $H_k(X)$ with basis in 1-to-1 correspondence with k -dimensional "holes" in X

- E.g. 2-torus has

$$\dim(H_0(X)) = 1, \quad \dim(H_1(X)) = 2, \quad \dim(H_2(X)) = 1$$

- Easy to compute (given a triangulation)
- Functorial: $X \rightarrow Y$ induces $H_k(X) \rightarrow H_k(Y)$

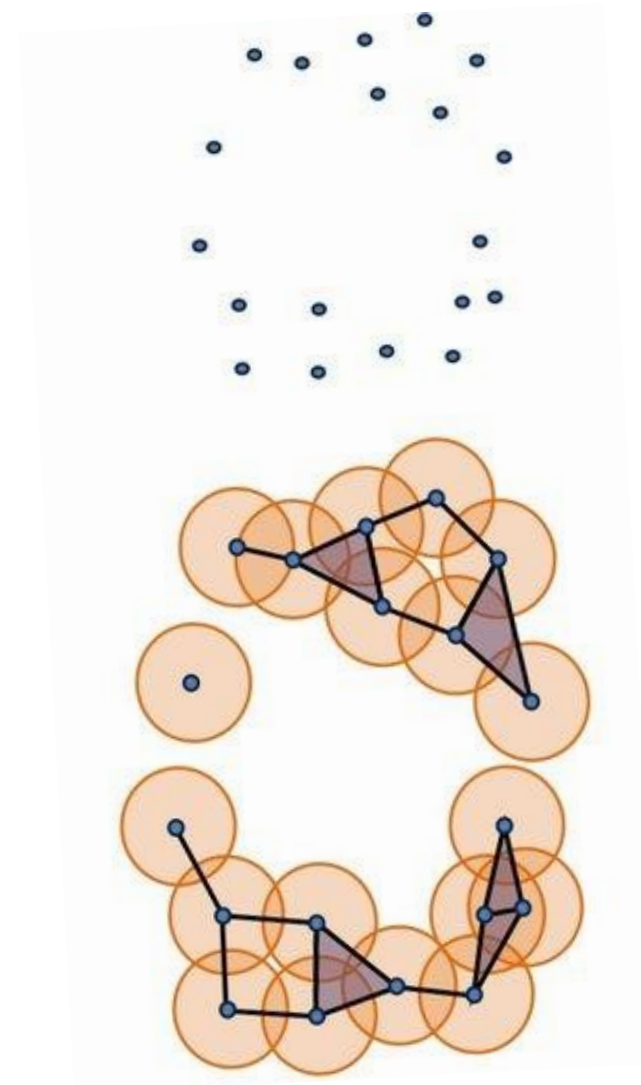


Persistent Homology

- How do we apply this to data?
 - E.g. Vietoris-Rips complex $VR_s(X)$
- What scale do we work on?
- Why not multiple at once:

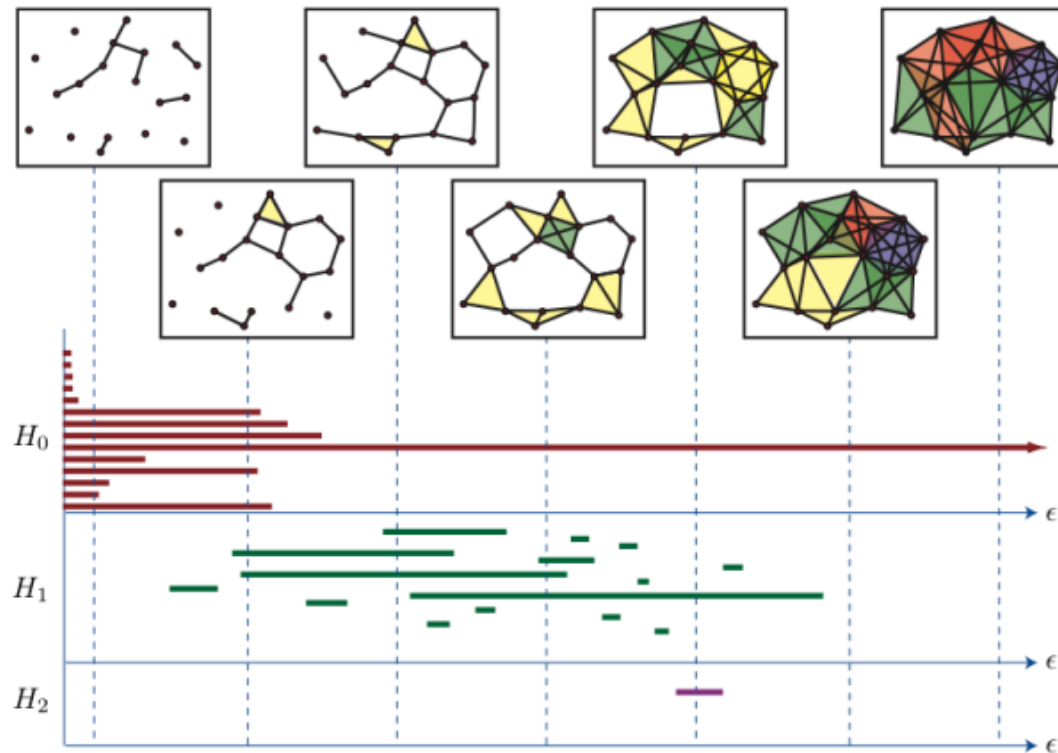
$$VR_s(X) \hookrightarrow VR_t(X) \text{ for } s \leq t$$

$$\text{induces } H_k(VR_s(X)) \rightarrow H_k(VR_t(X))$$



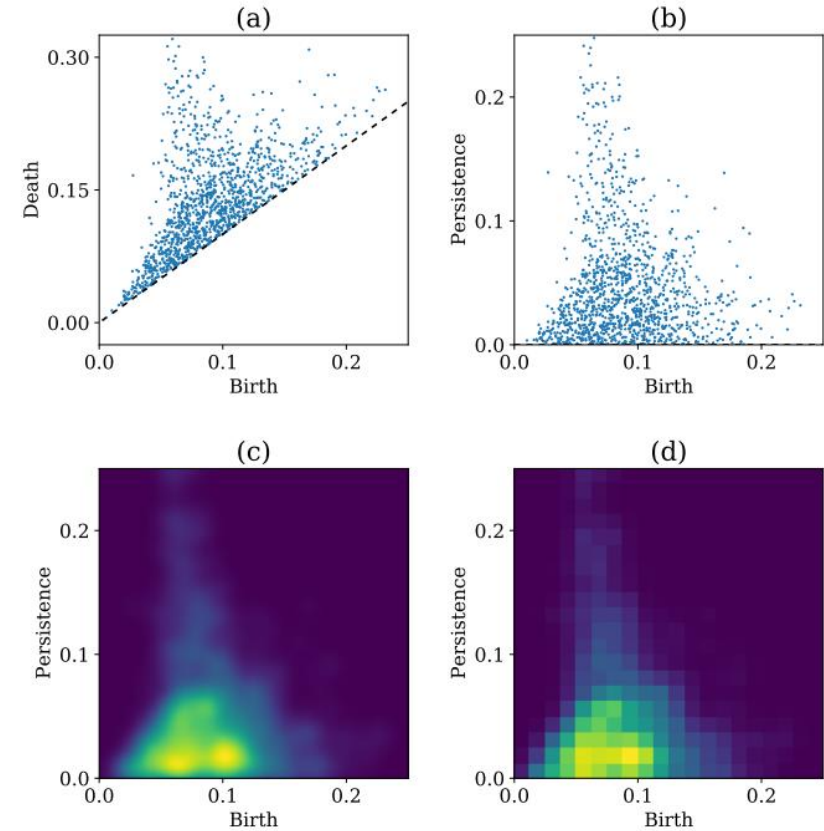
Persistent Homology

$$H_k(\text{VR}_{s_0}(X)) \rightarrow H_k(\text{VR}_{s_1}(X)) \rightarrow \dots \rightarrow H_k(\text{VR}_{s_n}(X))$$



Persistent Homology

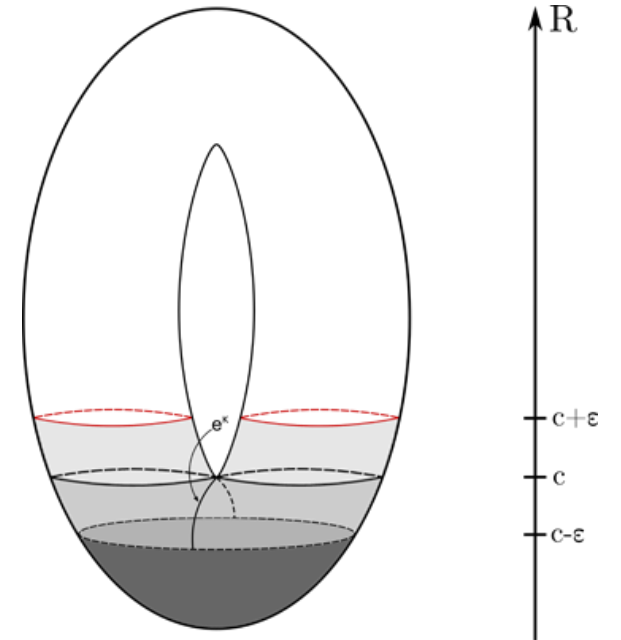
- Note: Vietoris-Rips wasn't necessary, only that we have a nested sequence of topological spaces (a filtration)
- Can also represent the output with a persistence diagram
- Vectorisation
 - E.g. persistence images



Persistent Homology for Phase Transitions

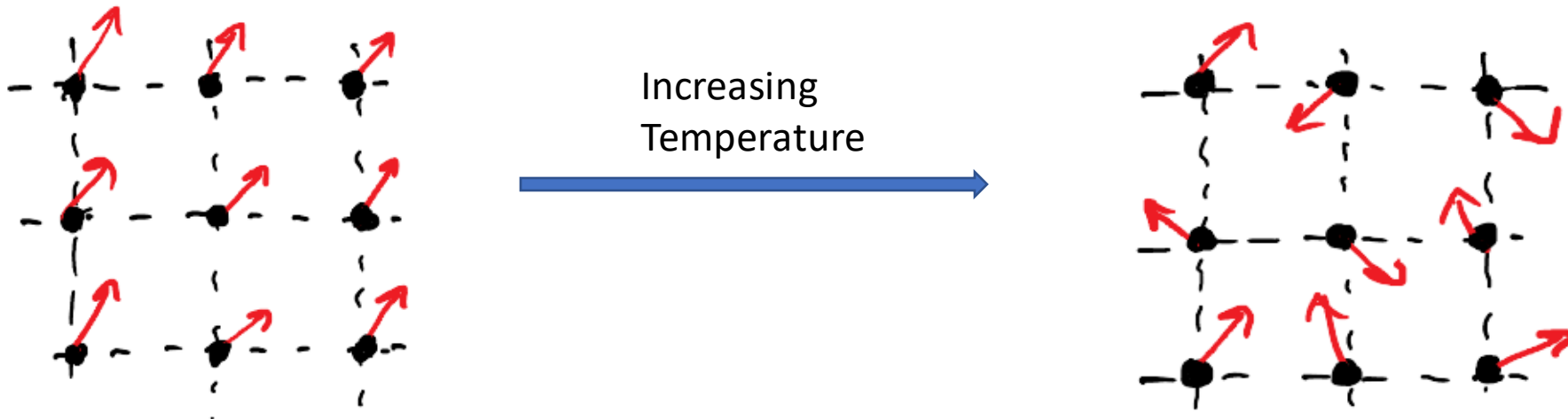
Two paradigms:

- *Persistent homology of configuration space*
 - Topology hypothesis
 - Very high dimension problem
- *Persistent homology as an observable*
 - Topological structure of individual configurations
 - Investigate contribution of such structure



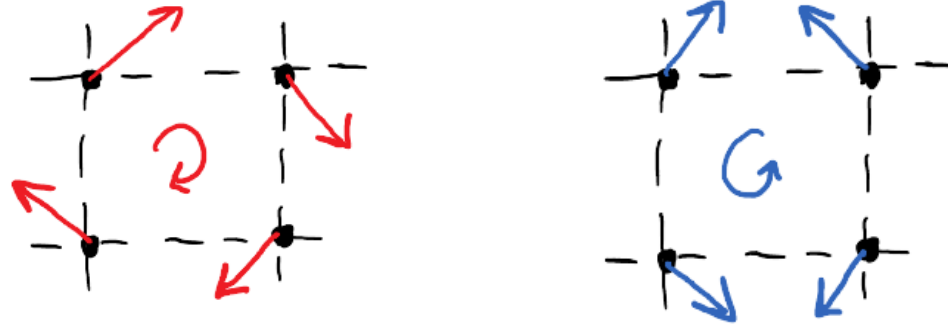
XY Model

- 2D lattice spin model with Hamiltonian $H(\theta) = -J \sum_{\langle ij \rangle} \cos(\theta_i - \theta_j)$
- A configuration is sampled with probability $\propto \exp(-H / T)$
- Transition from quasi-long range ordered phase to disordered phase



XY Model

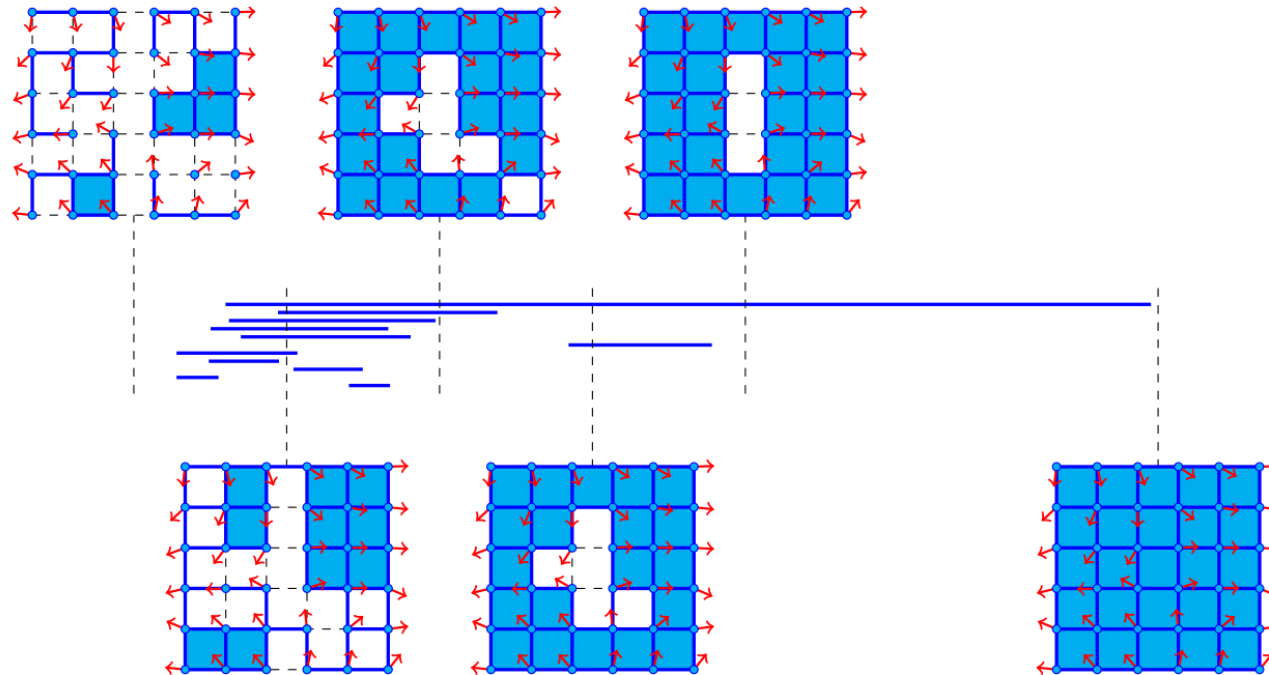
- Phase transition at $T = 0.893$ driven by topological defects: vortices



- Low temperatures: bound vortex-antivortex pairs
- High temperatures: sea of free vortices

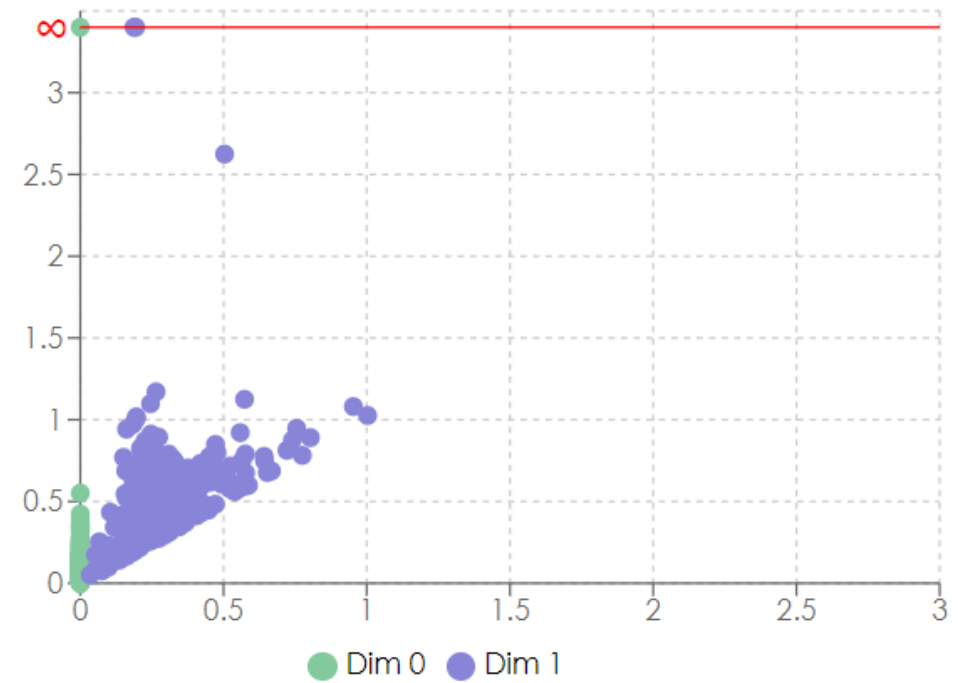
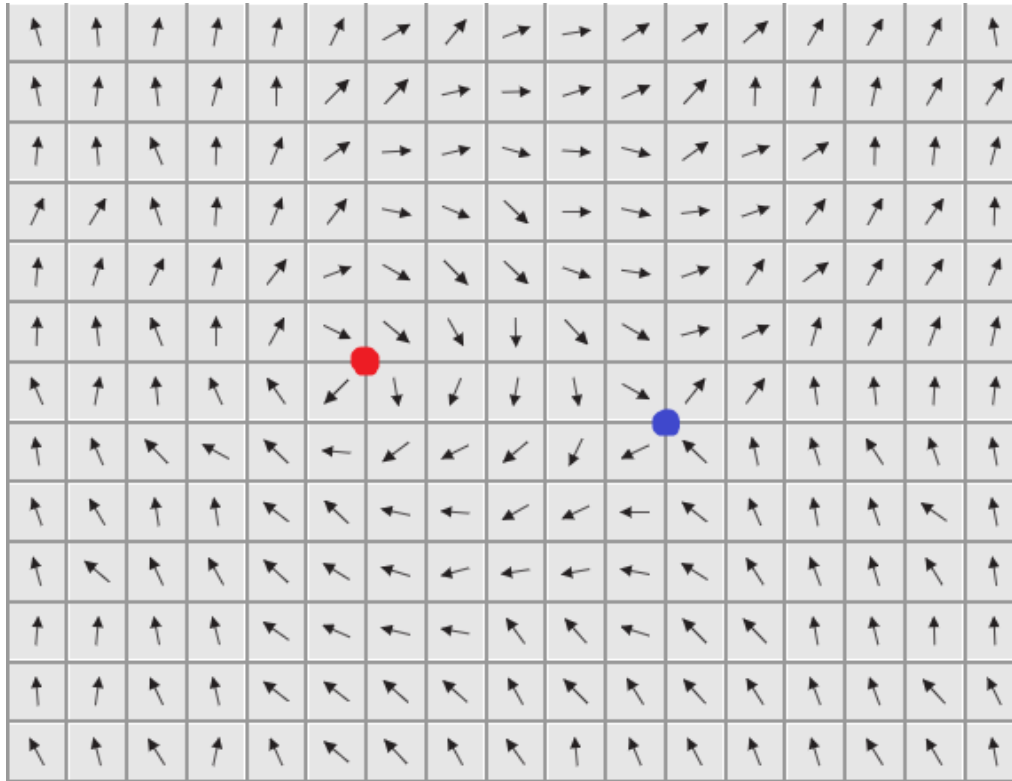
Persistent Homology of XY Model

- Idea: filter the tiling of the 2-torus corresponding to the lattice according to difference in neighbouring spins



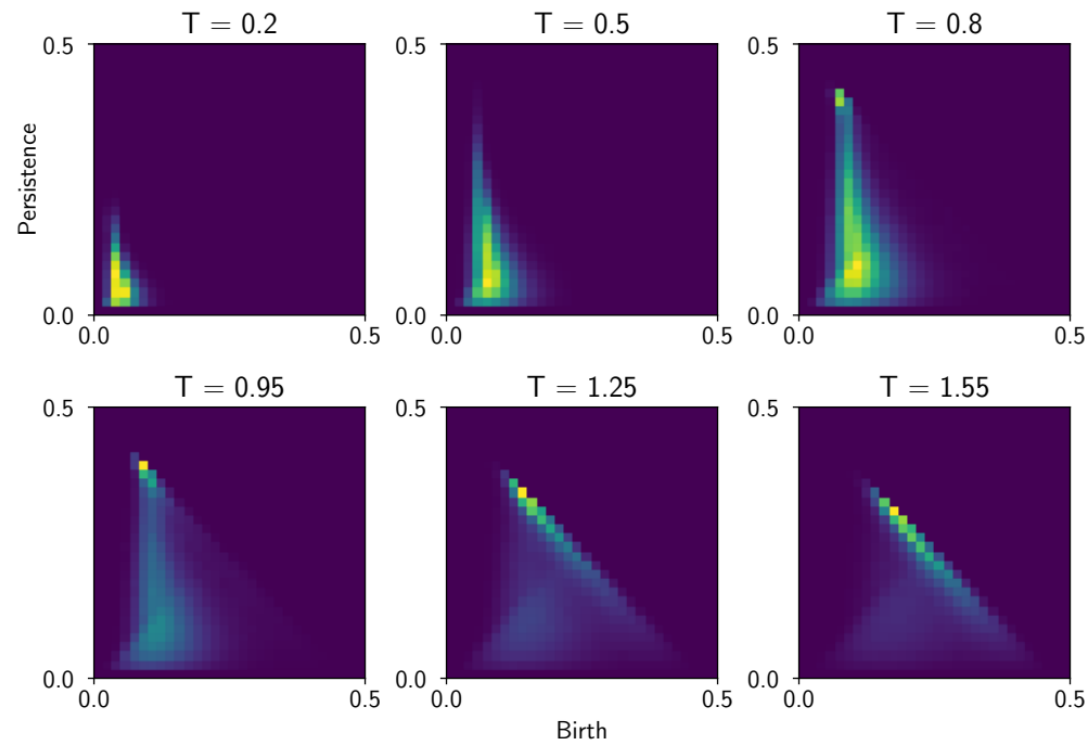
Persistent Homology of XY Model

- Example



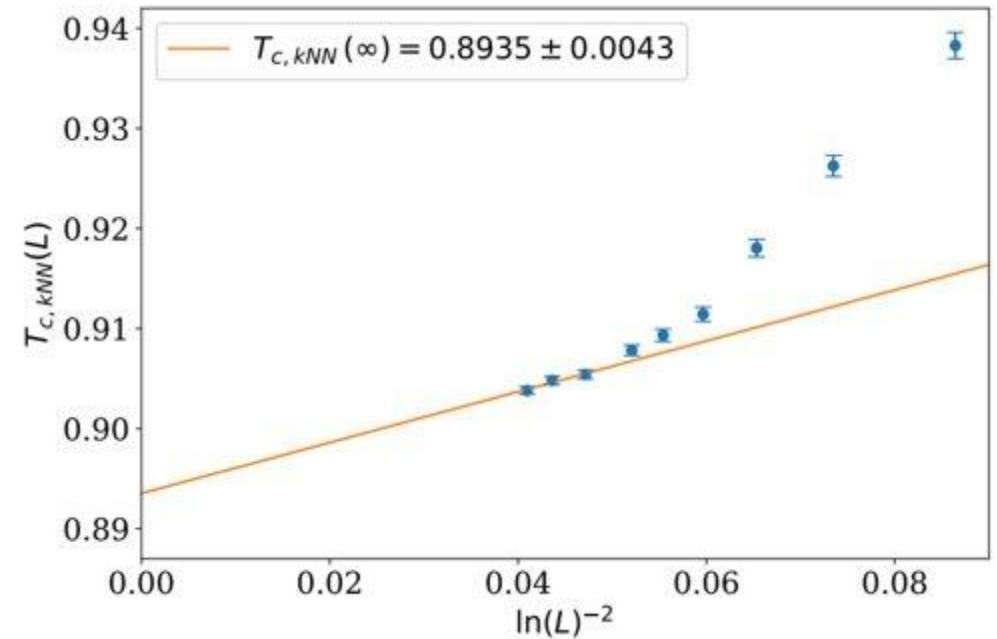
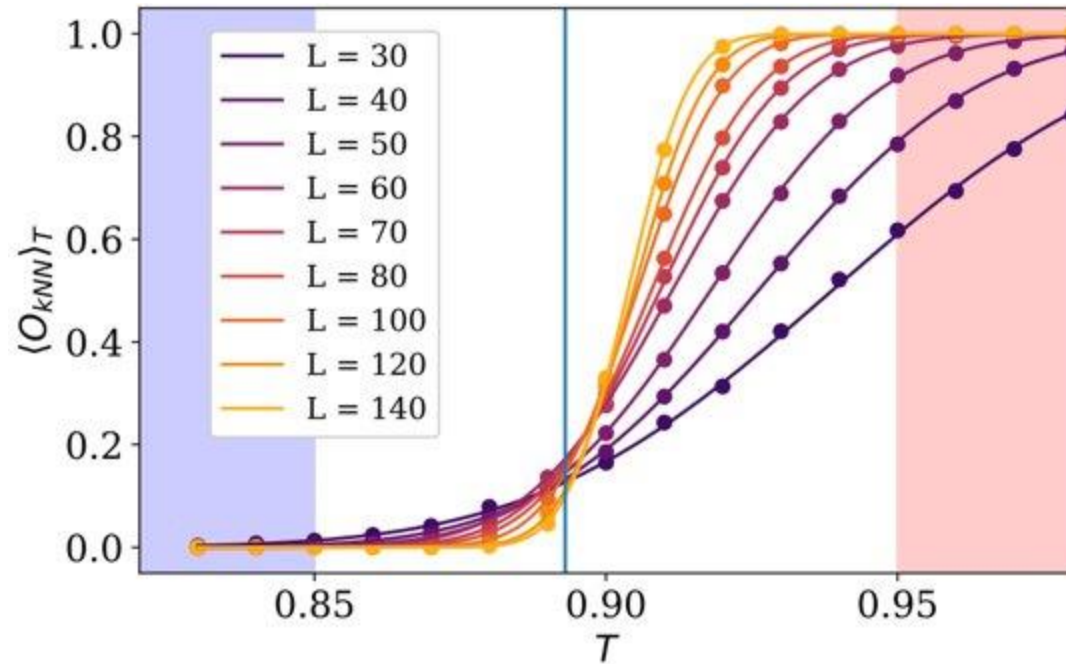
Persistent Homology of XY Model

- Average H_1 persistence images show vortex formation and qualitative change in behavior at different temperatures



Quantitative Analysis

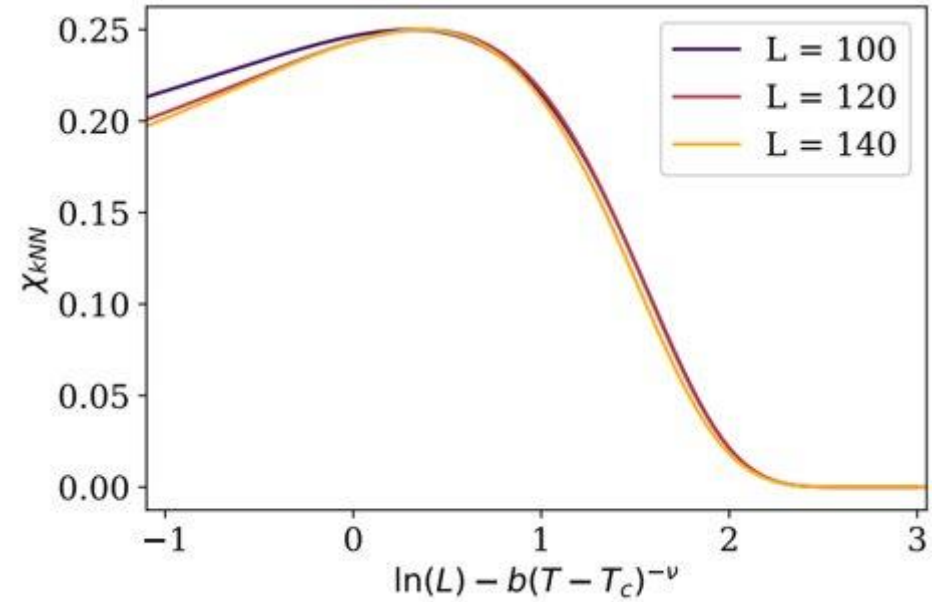
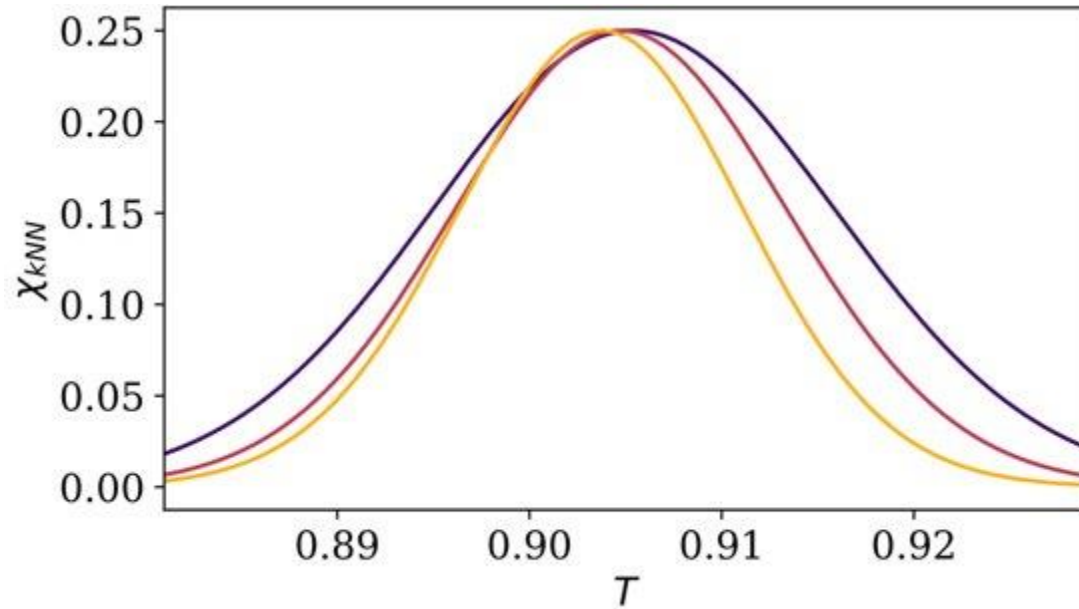
- Train a k-nearest neighbour classifier on persistence images
- Use finite-size scaling to extract critical temperature



Quantitative Analysis

- Curve collapse to get critical exponent

$$T_c = 0.8918 \pm 0.0033$$
$$\nu = 0.4972 \pm 0.0264$$

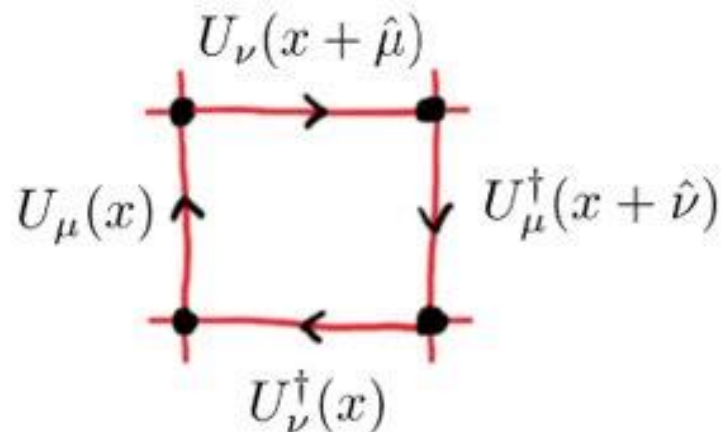


XY Summary

- Persistent homology easily allowed us to detect vortices
- It captured degrees of freedom relevant to the phase transition
- Quantitative analysis of the phase transition was possible with a simple machine learning approach

SU(2) Lattice Gauge Theory

- Simplified version of QCD (gluons only)
- Yang-Mills \rightarrow Path Integral \rightarrow Wick Rotation \rightarrow Lattice Discretisation
- SU(2) matrix on each lattice link
- Only traces of products along closed paths are gauge invariant

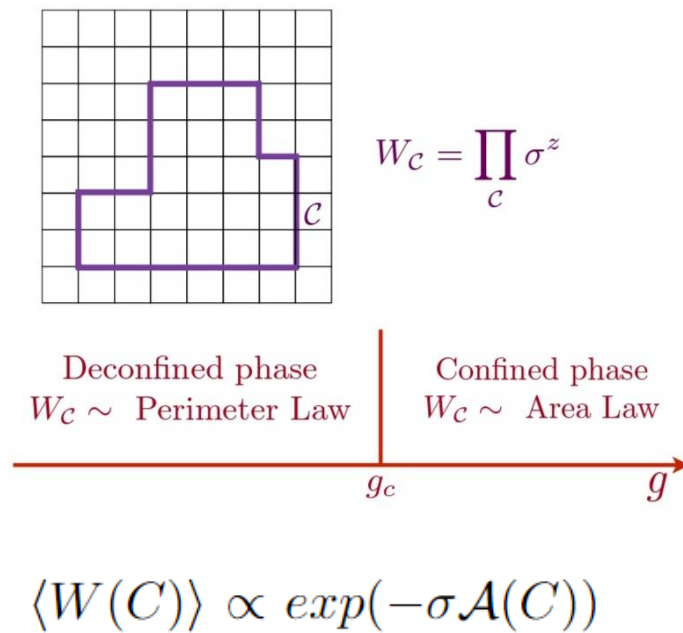


$$W_{x,\mu,\nu} = \text{tr} \left[U_{\mu}(x) U_{\nu}(x + \hat{\mu}) U_{\mu}^{\dagger}(x + \hat{\nu}) U_{\nu}^{\dagger}(x) \right]$$

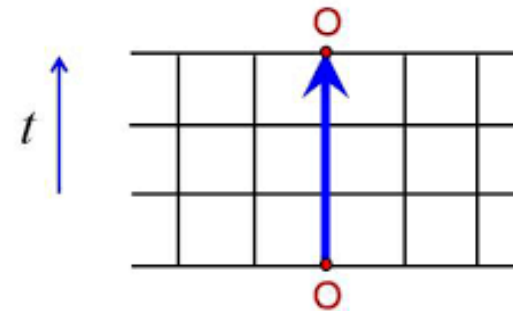
$$S(\mathbf{U}) = -\frac{\beta}{4} \sum_{x,\mu < \nu} W_{x,\mu,\nu}$$

Confinement

- Recall confinement means quarks and gluons always found in bound states
- Both SU(2) LGT and QCD exhibit a deconfinement phase transition



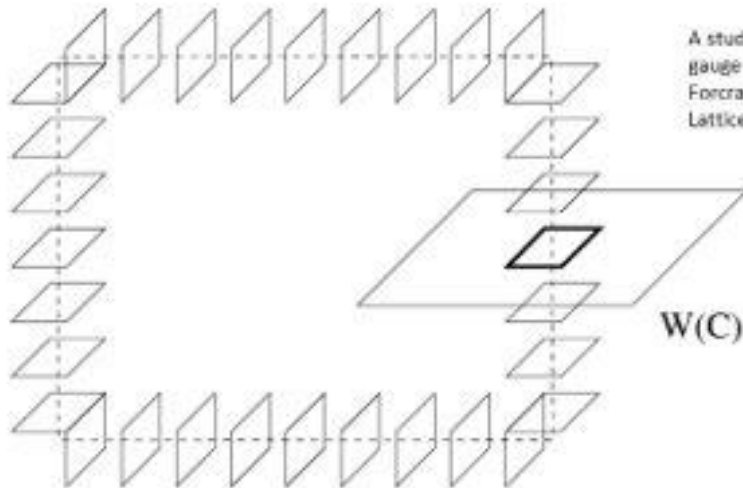
Polyakov loop (gauge-invariant)



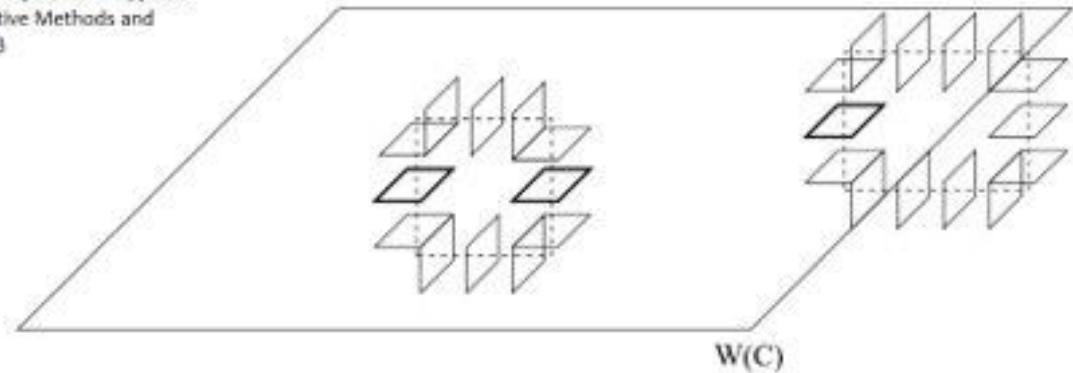
$$\langle \Phi \rangle = e^{-\beta F_Q}$$

Center Vortex Picture

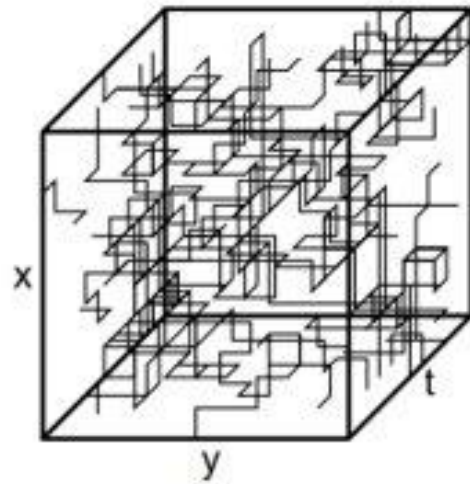
- $Z(\text{SU}(2)) = Z_2$
- Co-closed collection of plaquettes
- Linking Wilson loops are multiplied by a center element



A study of center vortices in SU(2) and SU(3) gauge theories, Michele Pepe and Philippe de Forcrand, Non-Perturbative Methods and Lattice QCD, pp. 194-203

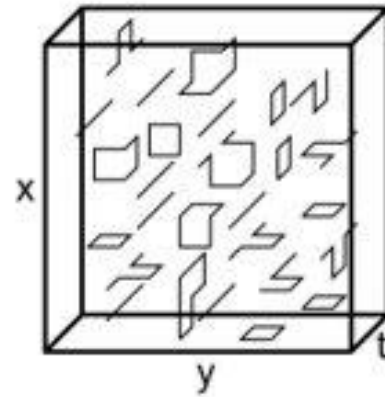


Center Vortex Picture



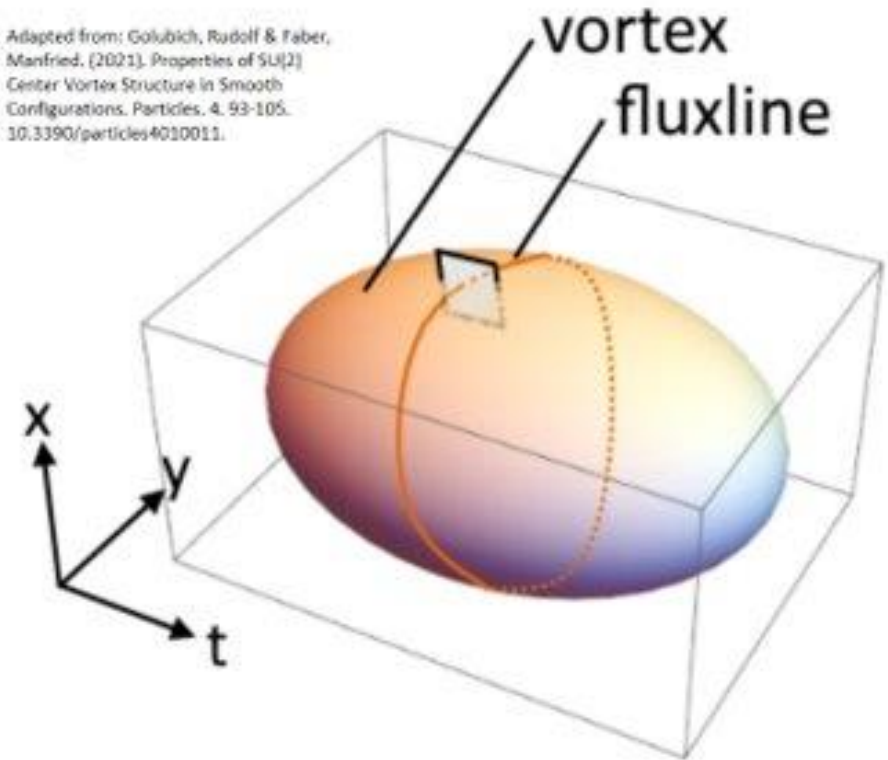
confined phase

Deconfinement in $SU(2)$ Yang-Mills theory as a center vortex percolation transition, M. Engelhardt, K. Langfeld, H. Reinhardt, and O. Tennert, *Phys. Rev. D* 61, 054504



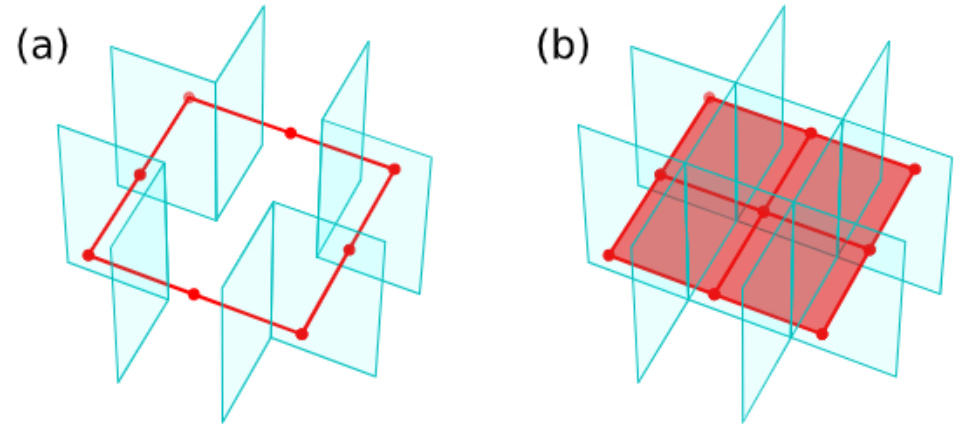
deconfined phase

Adapted from: Golubich, Rudolf & Faber, Manfred. (2021). Properties of $SU(2)$ Center Vortex Structure in Smooth Configurations. *Particles*. 4, 93-105. [10.3390/particles4010011](https://doi.org/10.3390/particles4010011).



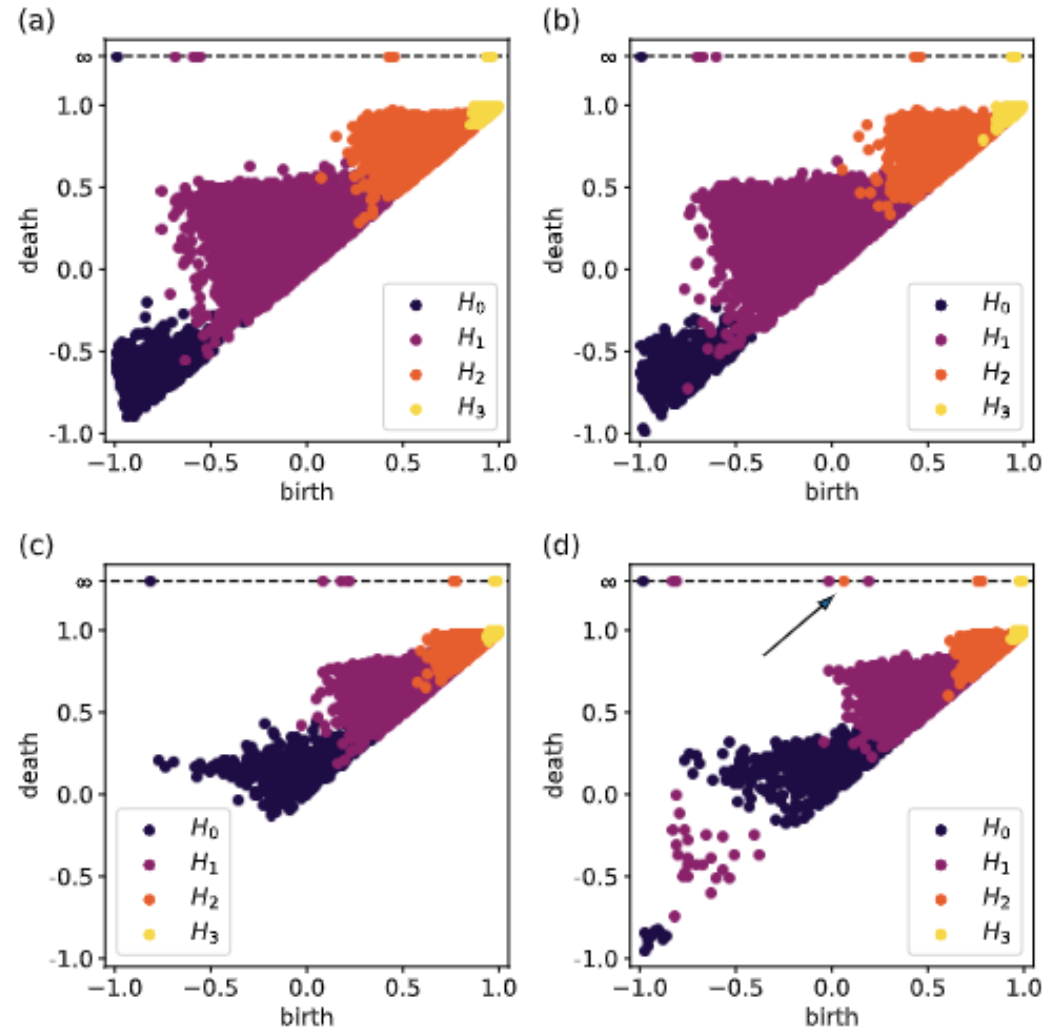
Persistent Homology of $SU(2)$ LGT

- Idea: filter the cubical tiling of the 4-torus corresponding to the dual lattice according to Wilson loop around plaquettes
- Introduce each plaquette (2-cell) at time equal to the WL of the plaquette it links with
- Introduce 1-cells and 0-cells as needed
- Introduce 3-cells and 4-cells introduced according to a "clique" rule



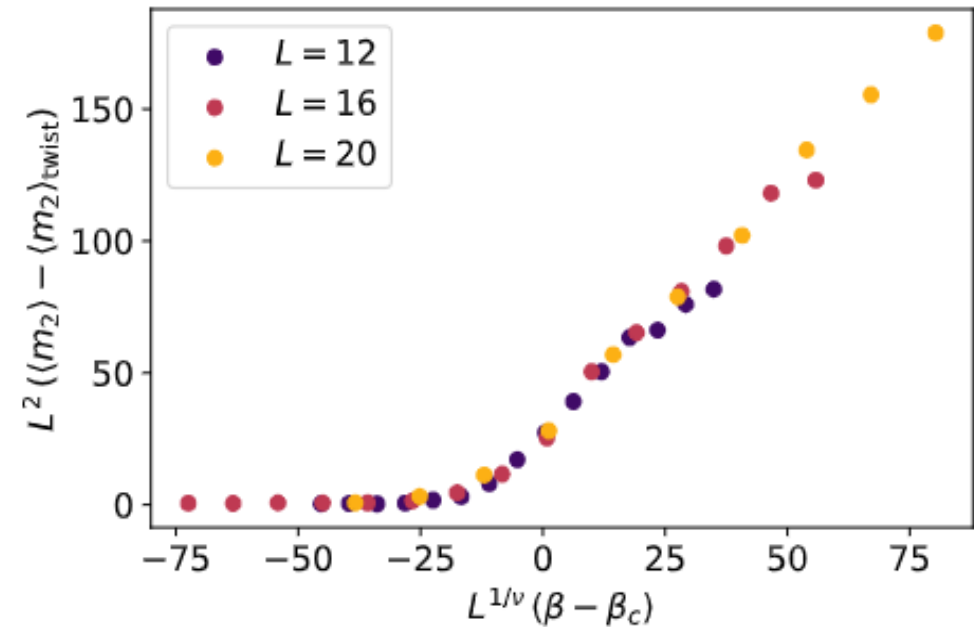
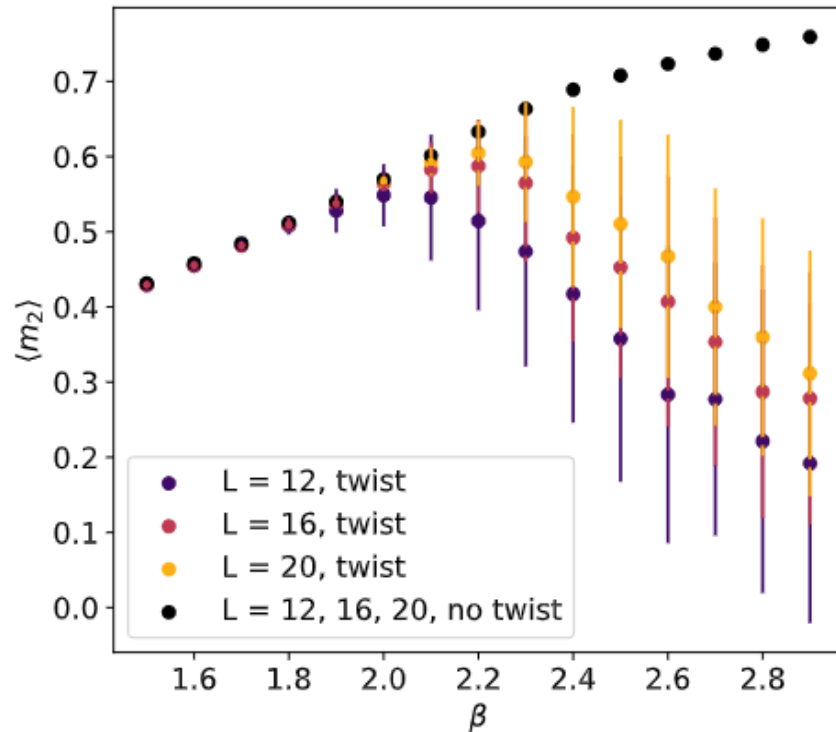
Persistent Homology of SU(2) LGT

- Test with twisted boundary conditions
- Thin vs thick vortices



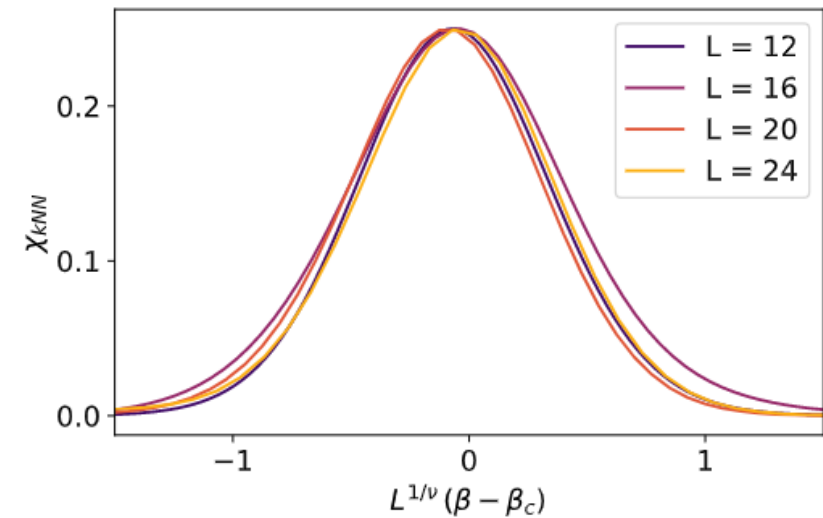
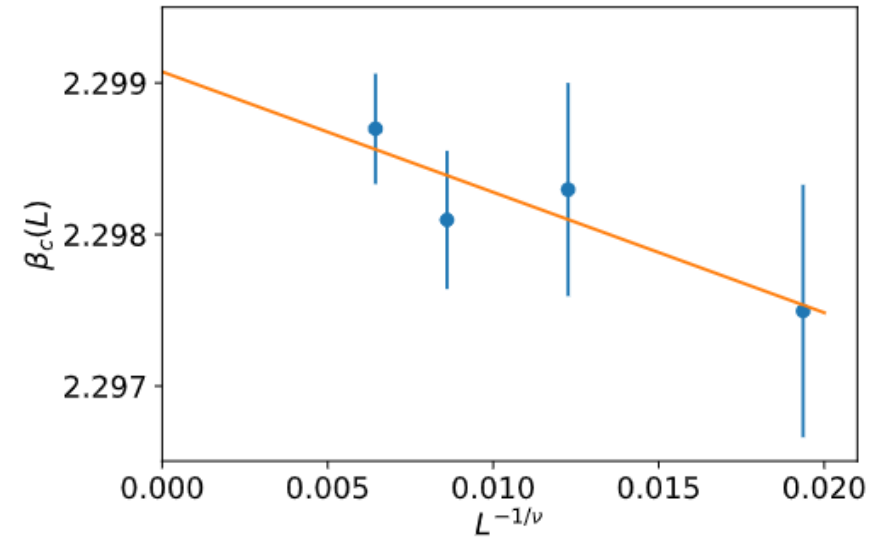
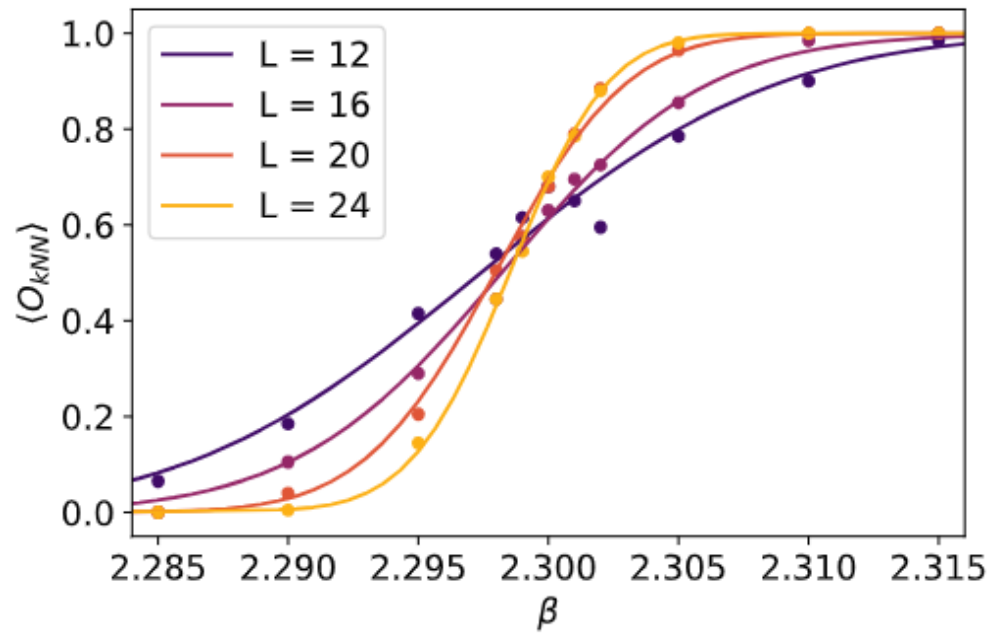
Quantitative Analysis

- Twisted boundary conditions allow us to define an order parameter



Quantitative Analysis

- Also possible without TBC



SU(2) LGT Summary

- Persistent homology let us detect thin vortices
- We hypothesise that it's sensitive to thick vortices too
- It captured degrees of freedom relevant to the phase transition
- Quantitative analysis of the phase transition was possible with a simple machine learning approach

- Evidence on the role of vortices in deconfinement?

Future Work

- $SU(2) \rightarrow SU(3) \rightarrow \text{QCD}$
- $Z(SU(3)) = Z_3$ (2 types of vortex)
- Other pictures of confinement
- Computational innovations allowing persistent homology of configuration space approach?

Thank You!

- "Quantitative analysis of phase transitions in two-dimensional XY models using persistent homology" Nicholas Sale, Jeffrey Giansiracusa, Biagio Lucini
- Paper on SU(2) LGT work coming very soon
- <https://nicksale.github.io/>
- nicholas.j.sale@gmail.com
- Questions?