# Machine Learning in Lattice Gauge Theory

Andreas Athenodorou, Davide Vadacchino and Maria Paola Lombardo











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## **Topics to be covered**

- Basics on Machine Learning
- Phase Transition in Lattice Gauge Theories
- Renormalisation Group and Machine Learning
- Machine Learning for Quantum Field Theories with sign problem
- Machine Learning and Spectral Functions in Lattice QCD
- Flow Based Sampling
- Machine Learning and Effective Field Theories
- Estimation of Lattice QCD observables using Machine Learning

# **Proposed Syllabus – The basics**

Basic Structure of Artificial Neural Networks Training a Network Supervised Learning Convolutional Neural Networks Unsupervised learning Recurrent Networks Transformers Boltzmann Machines Support Vector Machines ✤Reinforcement learning

# **Proposed Syllabus – Phase transitions**

Supervised and Unsupervised phase transition recognition in Spin Models
 Examples of Applications in LGT

 Phase transition recognition in SU(2) (Sebastian Wetzel)
 Phase transition recognition in SU(3) (Denis Boyda)
 Phase transition recognition in Lattice QCD (Maria Paola Lombardo)

## **Proposed Syllabus – Renormalization Group**

Machine Learning and Renormalization Group (Youtube lectures)
 Deep Learning and Renormalisation Group (Youtube lectures)
 Inverse Renormalisation Group in Lattice Field Theories (Dimitrios Bachtis)

# **Proposed Syllabus – Machine Learning for Quantum Field Theories with sign problem**

Contour Deformations (Thomas Luu, Michael Wagman)
 Complex Normalizing Flow (Yukari Yamauchi)
 Applications

# Proposed Syllabus – Machine Learning for Spectral Reconstruction

- Machine Learning and the Inverse Problem
- Machine Learning Hadron Spectral Functions In Lattice QCD (Heng-Tong Ding)
- Applications

#### **Proposed Syllabus – Flow Based Sampling**

Normalizing Flows (Elia Cellini)
Stochastic Normalizing Flows (Elia Cellini)
Fourier Flows (Lingxiao Wang)
Equivariant Flows (Matteo Favoni, Daniel Schuh, Phiala Shanahan)
Applications

## **Proposed Syllabus – Effective Field Theories**

- Symbolic Computation using Machine Learning (Abdulhakim Alnuqayqdan)
- Bayesian estimates for high orders in perturbation theory (Aleksas Mazeliauskas)
- Generative models in Effective Field Theories (Marina Marinkovic)
- Differentiable programming in Effective Field Theories (Fernando Romero-Lopez, Phiala Shanahan)

# **Proposed Syllabus – Estimation of Lattice QCD observables using Machine Learning**

Boosted decision tree regression ML algorithm (Boram Yoon)
 Single Point Neural Networks & Global Neural Networks(Giovanni Pederiva)

#### **Pre-existing Material**

The basics by Florian Marquardt: <a href="http://machine-learning-for-physicists.org/">http://machine-learning-for-physicists.org/</a>
 Introduction to Machine Learning Approaches for Simulating Lattice Field Theories, by Lena Funcke – Lattice Practises 2021 - <a href="https://indico.cyi.ac.cy/event/1/timetable/#20211012">https://indico.cyi.ac.cy/event/1/timetable/#20211012</a>

Machine Learning for Lattice QCD, by Phiala Shanahan – INT Summer School on Problem Solving in Lattice QCD

- Many general presentations on Machine Learning for Lattice QCD can be found online - however not lectures
  - Phiala Shanahan YouTube
  - Kurtej Kanwar YouTube

The basics exist in several lectures

Need to create lectures from scratch - ask volunteers

# **Pre-requisites**

Basic Data Science skills
 Computational Skills – Programming languages,

# **People involved so far in Machine Learning**

- Andreas Athenodorou
- Davide Vadacchino
- Maria Paola Lombardo
- More to follow after discussions in

Machine Learning approaches in Lattice QCD - An interdisciplinary exchange

- Week 27 February 2023 to 3 March 2023
- Covers a broad range of applications