## Advances in Lattice QCD

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5th Symposium of the Division for Physics of Fundamental Interactions of the Polish Physical Society

### Lattice regularization

Discretize your theory with a minimal lattice spacing a (UV) and a finite volume (usually torus) (IR). Evaluate numerically the finite path integral to estimate quantum expectation values.

### Recovery of relevant continuum (Euclidean) physics

- statistical limit
- infinite volume limit
- physical quark masses limit
- continuum (renormalization) limit
- other required limits (large hadron's momentum)
- ...

All this takes a lot of human and computer time!

⇒ but allows to keep all systematic uncertainties under full control!

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# Lattice Quantum Field Theory: where are we now?



New strategies in computing and simulation, e.g., machine learning and quantum computing

Taken from Snowmass 2022 report: https://arxiv.org/pdf/2209.10758.pdf

Advances in Lattice QCD

# Improving algorithms and computational strategies



Advances in Lattice QCD

# Improving algorithms and computational strategies

### Coordinated Lattice Simulations effort

#### Low lying baryon spectrum

 $\odot$  Octet baryon spectrum from BChPT, finite *a* and *FV* fits (determine  $\sqrt{8t_{0.ab}}$ ) Agreement with QCD "expt." masses within 1% overall uncertainty.

△ Octet and decuplet masses from SSE BChPT, finite a and FV fits.



Unstable decuplet baryons: grey bands indicate the expt. Breit-Wigner width. Proper treatment via the Lüscher formalism required.

#### S. Collins (U. Regensburg) Lattice2022 presentation

# Phenomenology

FLAG (Flavor Averaging Group) reports and decay constants

 $f_+(0) = \langle K | Q^{ar{u}s} | \pi 
angle, \qquad f_\pi = \langle \pi | Q | 0 
angle, \qquad f_K = \langle K | Q | 0 
angle$ 

#### Progress



 $f_+(0) = 0.9706(27)$ 

 $f_+(0) = 0.9698(17)$ 

### QED on the lattice: theoretically challenging!

- Gauss law does not allow charged states on a periodic torus
- massless photon  $\Rightarrow$  long range interactions

### $C^*$ boundary conditions

QED with C-parity boundary conditions (QED<sub>C</sub>): the matter fields are multiplied by the charge conjugation at the edge of the lattice in the spatial directions. This provides a consistent QFT. Charged interpolating operators can be constructed in a gauge-invariant way  $\Rightarrow$  correlation functions of charged operators can be estimated.

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# $QCD+QED_C$

### $C^*$ boundary conditions



Figure 7: Baryon effective masses and effective  $\phi$  observables for the ensemble A360a50b324+RW2, with the selected plateaux and the fits to a constant. Values in MeV are obtained by using the reference value  $(8t_0)^{1/2} = 0.415$  fm.

Taken from RC\* collaboration, https://arxiv.org/pdf/2209.13183.pdf

# *x*-dependence of structure functions

#### Twist-2 PDF



Taken from K.Cichy, https://arxiv.org/pdf/2111.04552.pdf

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### Further exploratory studies

- flavor-singlet quark PDFs
- gluon PDFs
- twist-3 quark PDFs
- generalized PDFs (GPD)
- transverse momentum dependent PDFs (TMD)

# Quantum computing

### Working principle

#### How to simulate these field theories?

Example: hybrid quantum-classical algorithms

#### Key concept

Classical computer: main computation Quantum computer: classically hard/intractable part Advantages: even for small quantum hardware!



1 Peruzzo et al. (2014)

Variational Quantum Eigensolver (VQE)

#### Goal

Find ground state and excited states of Hamiltonian  $\mathcal{H}$ Variational approach Minimize  $E(\vec{a}) = (\psi(\vec{a})|\mathcal{H}|\psi(\vec{a}))$  w.r.t. parameters  $\vec{a}$ Classical computer Given  $E(\vec{a})$ , find optimized parameters  $\vec{a}_{i+1}$ Quantum device Given  $\vec{a}_i$ , prepare  $|\psi(\vec{a})\rangle$  and measure  $E(\vec{a}_i)$ 



Taken from L.Flucke, Lattice2022 plenary talk

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### Quantum computing

### Revolution is happening now!

#### More details: parallel talks @ this conference

Quantum Electrodynamics in 1+1D	Discrete and higher-dimensional gauge theories
1+1D Schwinger model with chemical potential	1+1D $D_n$ gauge theory
Studying multi-flavor chemical potential with VQE	Preparing ground state & evolution on quantum annealer
→ talk by Stefan Kühn tomorrow at 2:20pm	$\rightarrow$ talk by Michael Fromm on Wednesday at 2pm
1+1D Schwinger model with θ-term	1+1D Z <sub>2</sub> gauge theory
Comparing quantum algorithms for state preparation	Developing quantum algorithms for thermal states
→ talk by Alexei Bazavov tomorrow at 3:20pm	→ talk by Connor Powers tomorrow at 3:20pm
1+1D Schwinger model with Wilson fermions	2+1D U(1) gauge theory
Comparing Wilson and staggered fermions	Finding a resource-efficient implementation
→ talk by Takis Angelides tomorrow at 5:30pm	→ talk by Christopher Kane tomorrow at 2pm
1+1D Schwinger model at finite temperature	
Mapping out $T - \mu$ phase diagram with VQE	
ightarrow talk by Akio Tomiya on Thursday at 12:30pm	

16

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#### Taken from L.Flucke, Lattice2022 plenary talk

### Quantum computing

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SU(N) lattice gauge theories	Lattice field theories
1+1D SU(2) gauge theory	1+1D Ising model
Implementing real-time evolution on IBM-Q hardware	Studying NISQ algorithms for open quantum systems
→ talk by Emanuele Mendicelli tomorrow at 4:30pm	$\rightarrow$ talk by Bharath Sambasivam tomorrow at 6:10pm
Comparing Schwinger-boson & loop-string hadron form.	1+1D extended O(2) model
$\rightarrow$ talk by Jesse Stryker on Thursday at 11:50am	Studying discrete approximations of continuous groups
SU(2) gauge theories	$\rightarrow$ talk by Leon Hostetler on Wednesday at 3:40pm
Developing efficient digitization via finite subgroups	1+1D Wess-Zumino model
→ talk by Timo Jakobs tomorrow at 2:40pm	Studying dynamical breaking of supersymmetry
Defining canonical momenta for discretized gauge fields	$\rightarrow$ talk by Christopher Culver tomorrow at 2:20pm
→ talk by Carsten Urbach tomorrow at 3pm	
SU(3) gauge theories	
Developing efficient digitization & improved Hamiltonians	
→ talk by Henry Lamm tomorrow at 5:50pm	
	17

#### Taken from L.Flucke, Lattice2022 plenary talk

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# Machine learning



# Fields via flow models

Train the model:

Example application: Embarrassingly parallel direct sampling

Flow model as an approximate trivialising map

 $\begin{array}{c} z & f(z) \\ f(z) & \varphi \\ f(z) & e^{i\varphi} \end{array}$ "Base" distribution: "Model" distribution Efficient to sample  $q(\varphi) \simeq L_{\varphi} - S(\varphi)$ 

Efficient to sample  $q(\phi) \approx rac{1}{Z} e^{-S(\phi)}$  e.g., Haar-uniform

 Independent samples of the base distribution map to independent samples of the model distribution Gradient descent to minimise "loss function" with minimum at  $q(\phi) = \frac{1}{Z}e^{-S(\phi)}$  $L(q) = \int \underline{d\phi} q(\phi) [\log q(\phi) + S(\phi)]$ 

> Estimate stochastically by sampling from the model, i.e., "self training"

• Guarantee exactness: Reweight or form a Markov chain with Metropolis-Hastings accept/reject step



Phiala Shanahan, MIT

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Taken from P.Shanahan, Lattice2022 parallel talk

Advances in Lattice QCD

# Machine learning

### Another revolution is happening now!

# Flow models for QCD in 4D

Initial QCD demonstration [this talk +upcoming manuscripts on scaling and 4D]

- Direct combination of published results on gauge-equivariant flows and pseudofermions (Boyda et al., 2008.05456, Abbott et al., 2207 08945)
- Illustration at straightforward parameters V=4<sup>4</sup>, N<sub>f</sub>=2, β=1, κ=0.1
- Observables from flow ensemble in precise agreement with HMC at high statistics (65k samples)



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### A **VERY** subjective subset:

- improving algorithms and computational strategies ⇒ increasing precision: permille precision on stable light hadron masses
- such precision requires control of corrections which so far were neglected  $\Rightarrow$  QCD+QED
- $\bullet\,$  improved precision proliferates on other quantities  $\Rightarrow\,$  structure functions
- ever increasing computational cost ⇒ will machine learning or quantum computers help?