

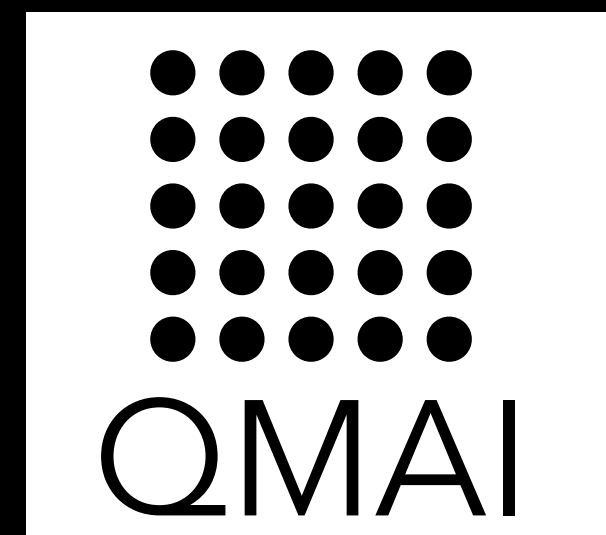
Neural wavefunctions for $SU(2)$ lattice gauge theory

Thomas Spriggs

TU Delft

Work in collaboration with: Eliska Greplova, Juan Carrasquilla, and Jannes Nys

Hamiltonian lattice gauge theories



In a nutshell

- New method for simulation non-Abelian lattice gauge theories in any* dimension
- I will show: ground state properties of SU(2) pure gauge
- Can be extended to: SU(N), time evolution, fermions

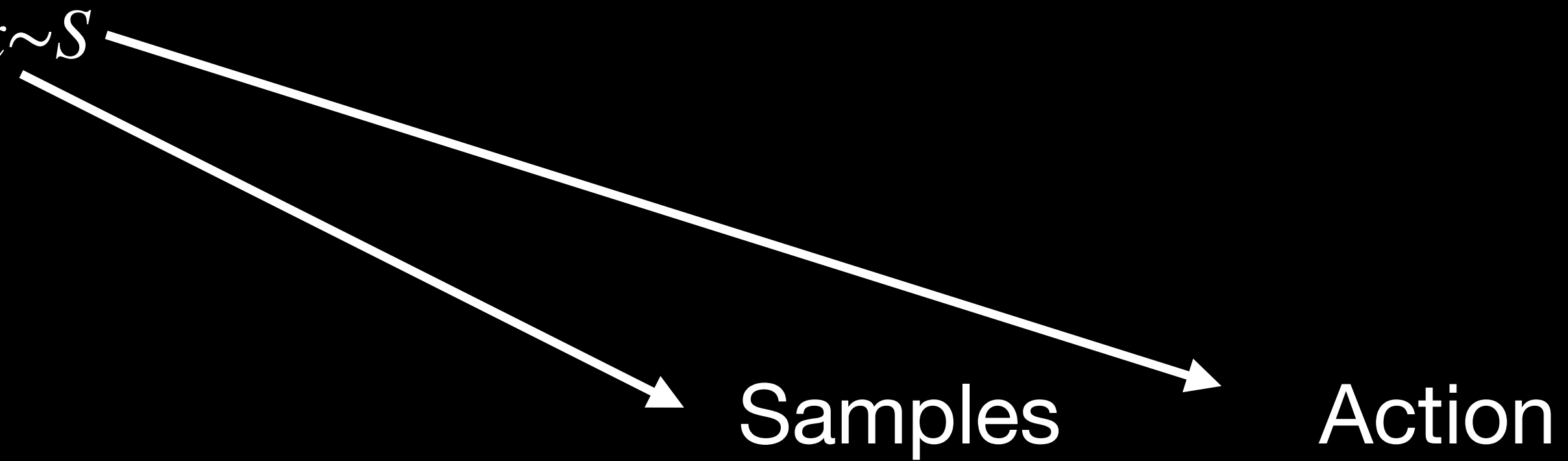
In a nutshell

- In Euclidean (quantum) Monte Carlo:
 - $\langle O \rangle = \mathbb{E}[O(x)]_{x \sim S}$

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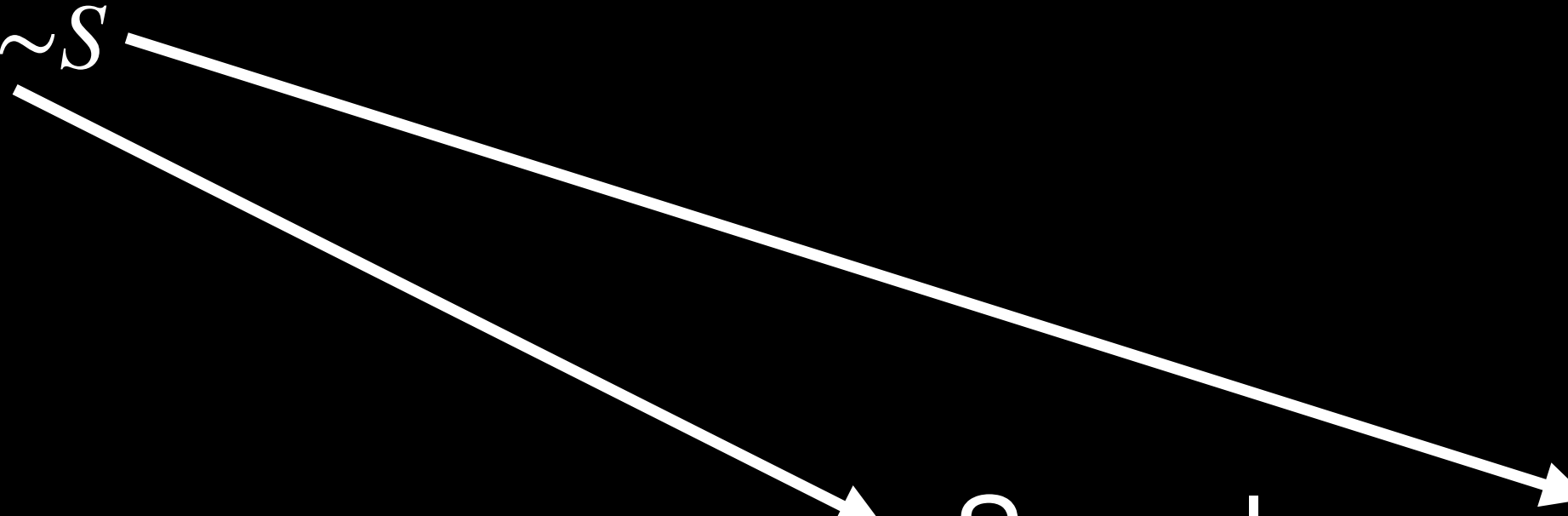
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In a nutshell

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Samples

Action

- In variational Monte Carlo:

- $\langle O \rangle = \mathbb{E}[O(x)]_{x \sim \psi}$

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Samples

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Learned distribution

In a nutshell

- In Euclidean (quantum) Monte Carlo:

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Samples

Action

Neural network

In a nutshell: why?

- Use the Hamiltonian framework (continuous time)
- Free of the sign problem (probably*)
- Maintains continuous representation
- Real time evolution of lattice gauge theories (future direction)

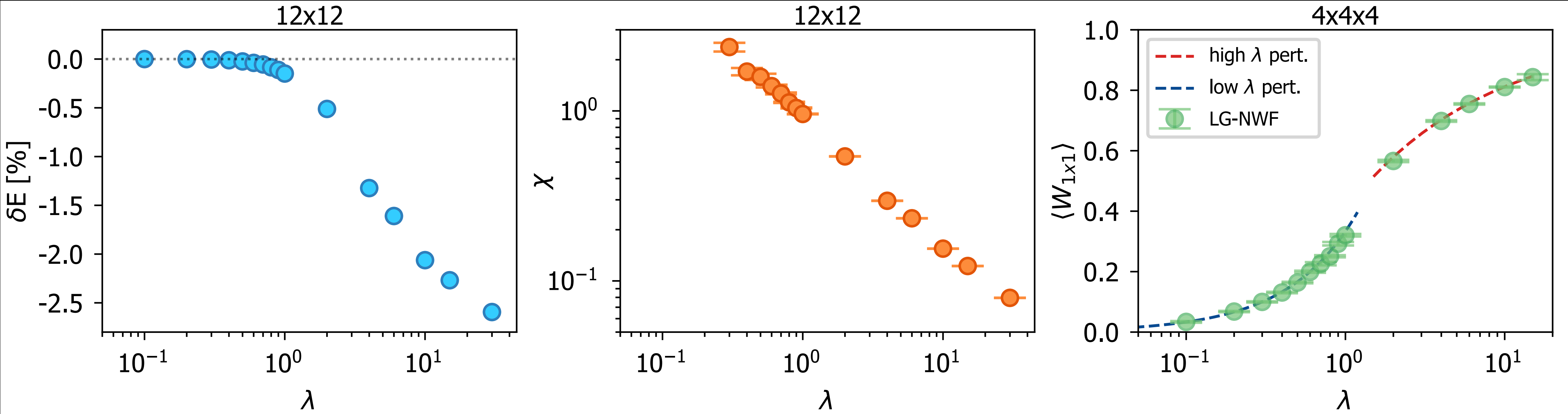
* hard to know exactly if the distribution can be learned in all cases without something like the sign problem appearing

In a nutshell: why not?

- Hard to design the distribution to give you the right observables
- Sampling from this distribution still remains difficult

g range ~ 2.5 - 0.6

Large λ is weak coupling (small g)



Energy improvement
over a reference
ansatz
(low is good)

Creutz ratio

Plaquette

Related neural wavefunction work

- $U(1)$ (Abelian)

Luo, D. *et al.* arXiv:2211.03198

- Z_2 (discrete and Abelian)

Apte, A. *et al.* Phys. Rev. B **110**, 165133

- Non-neural-network:

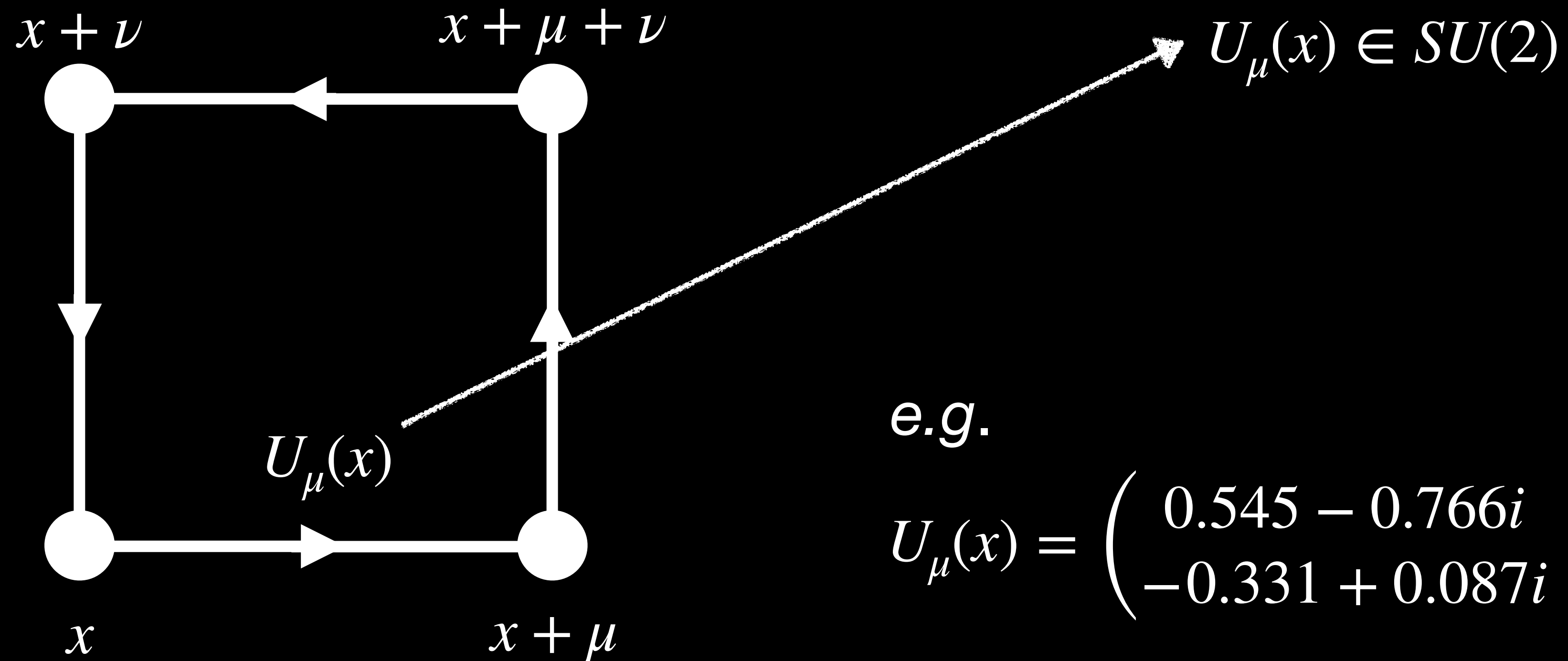
- $SU(2)$ Chin, S. A. *et al.* (1985) *Phys Rev D* **31**, 3201

- $SU(3)$ Chin, S. A. *et al.* (1988) *Phys Rev D* **37**, 3001

SU(2) lattice gauge theory

SU(2) Hamiltonian

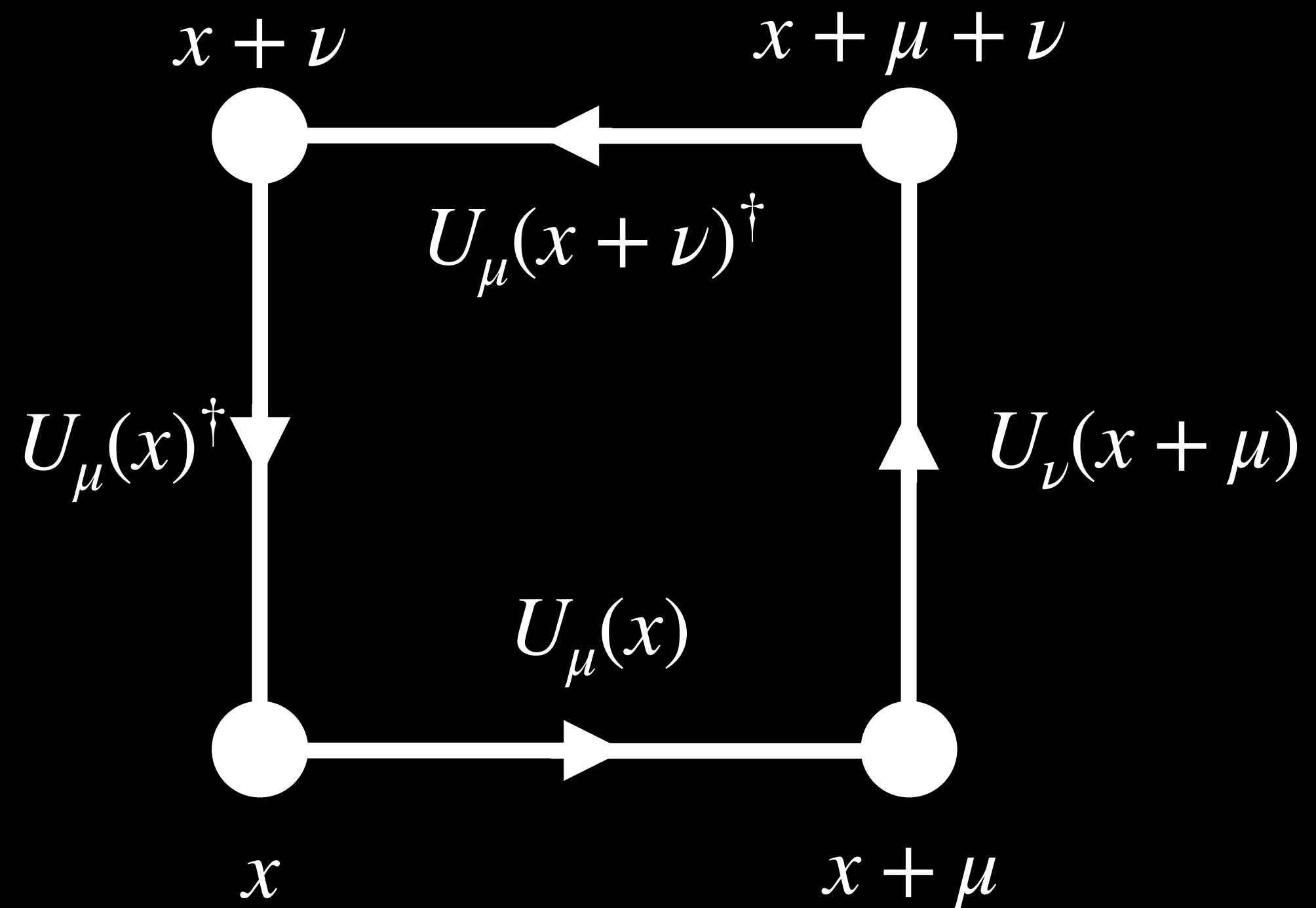
Basic degrees of freedom



e.g.

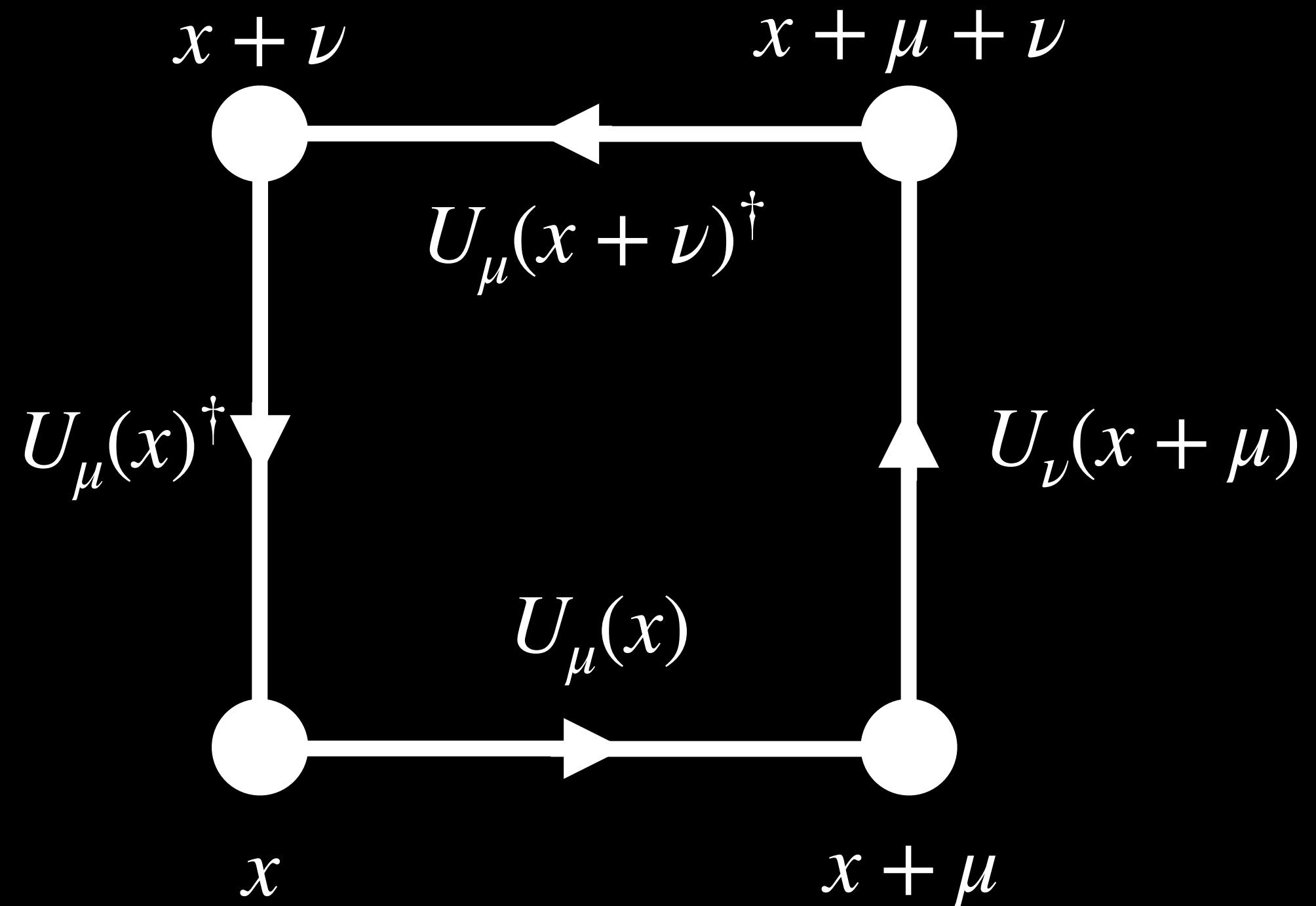
$$U_\mu(x) = \begin{pmatrix} 0.545 - 0.766i & 0.331 + 0.087i \\ -0.331 + 0.087i & 0.545 + 0.766i \end{pmatrix}$$

SU(2) Hamiltonian



Plaquette:

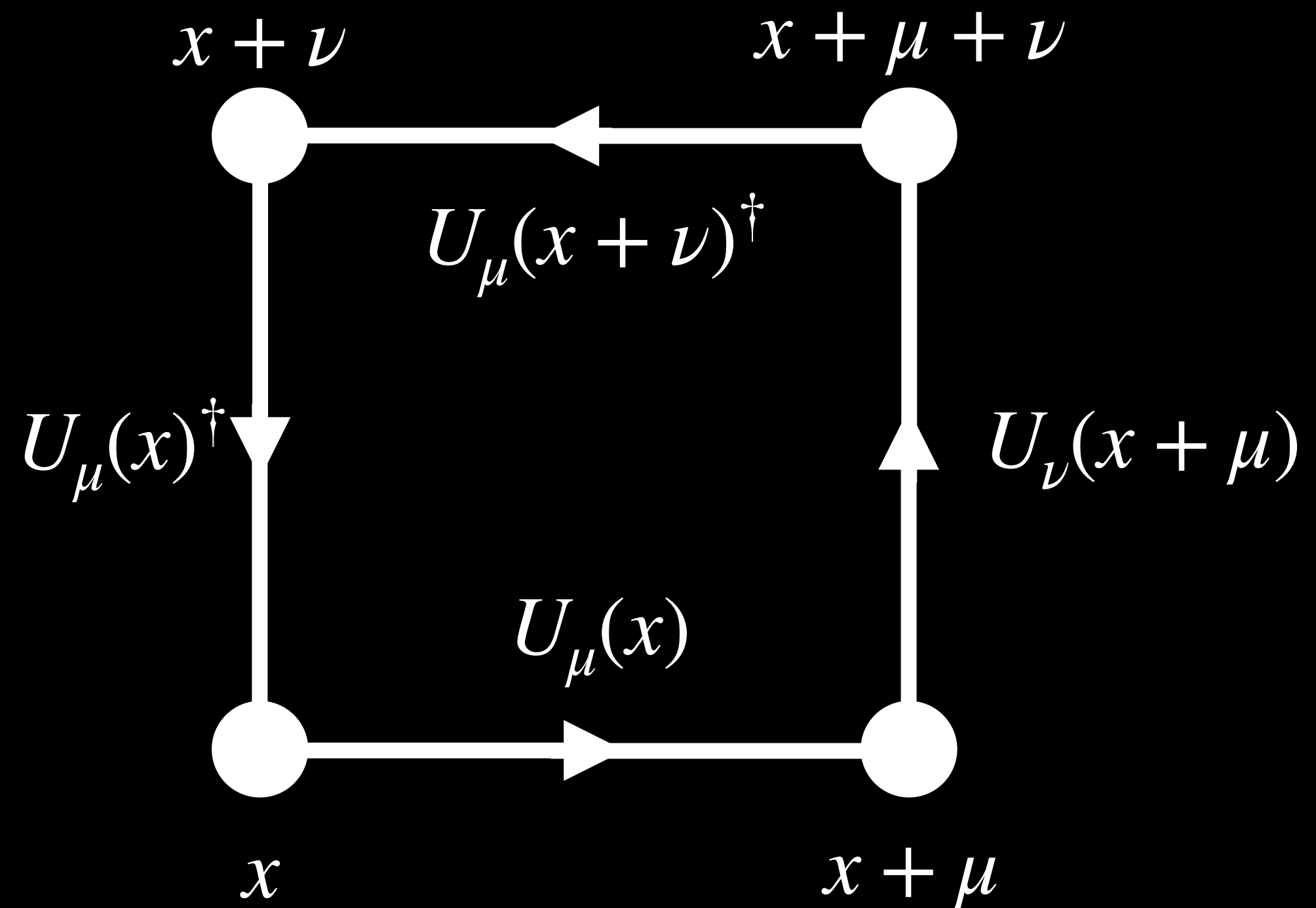
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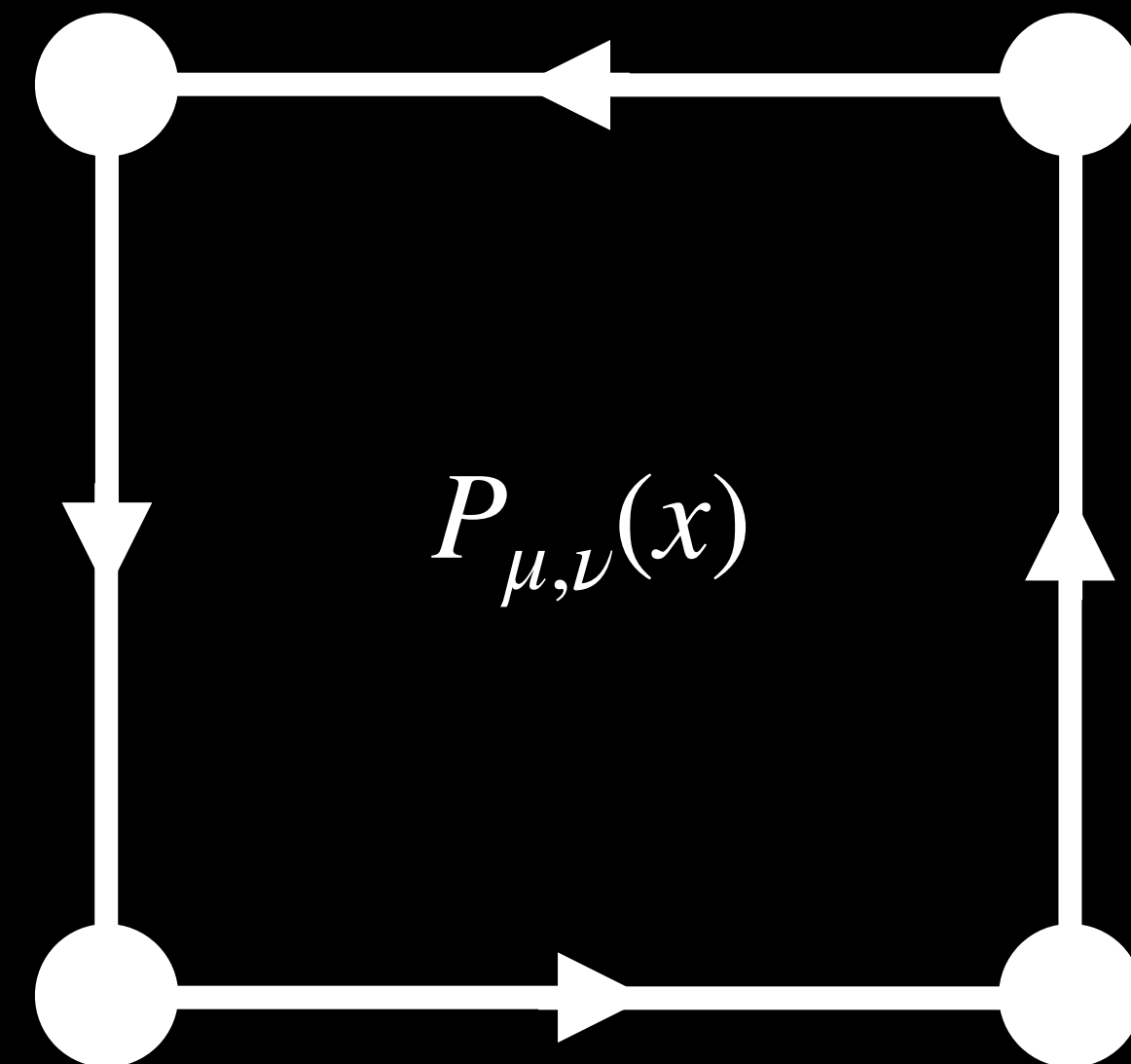
$$P_{\mu,\nu}(x) = U_\mu(x)U_\nu(x + \mu)U_\mu(x + \nu)^\dagger U_\nu(x)^\dagger$$

SU(2) Hamiltonian

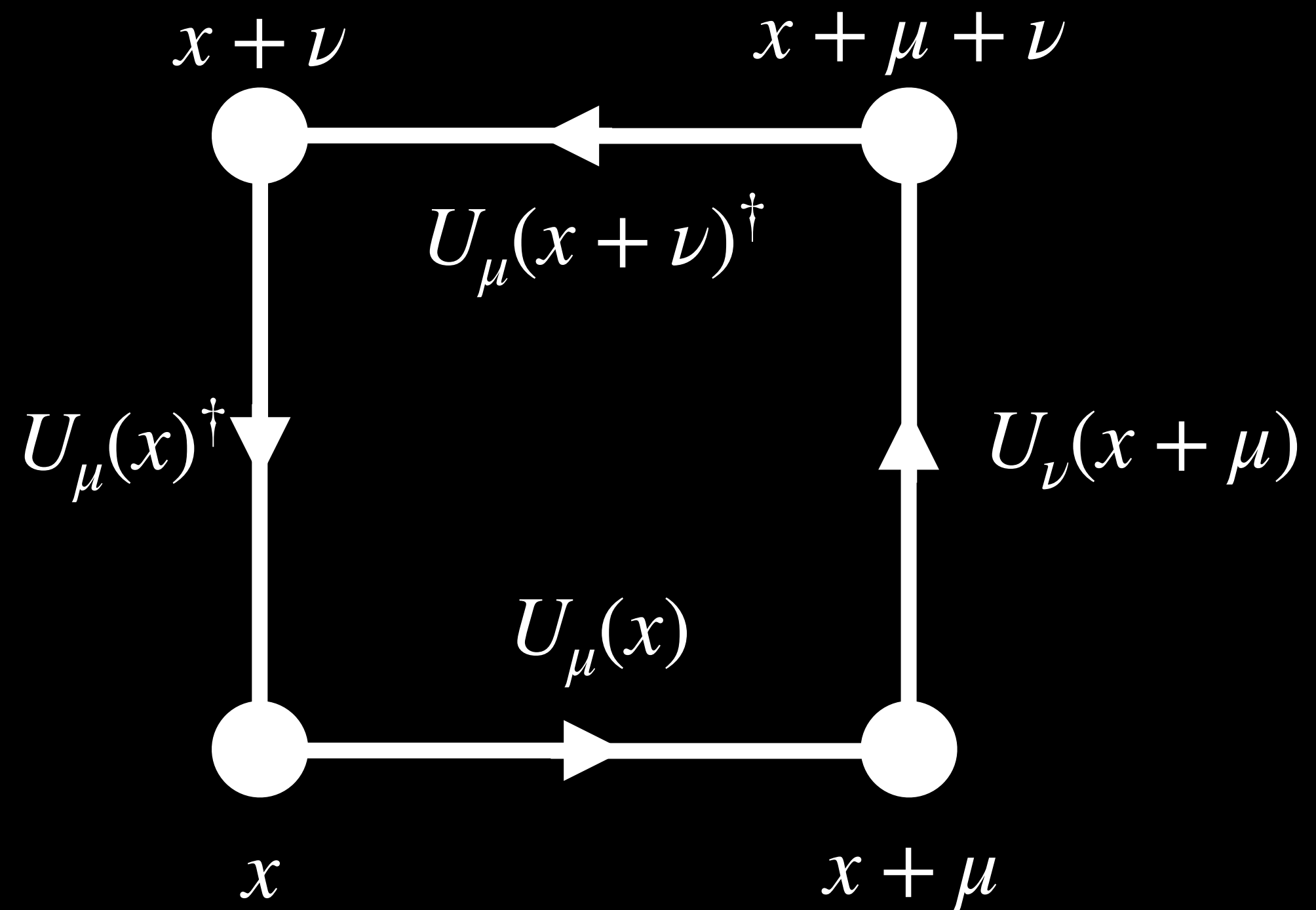


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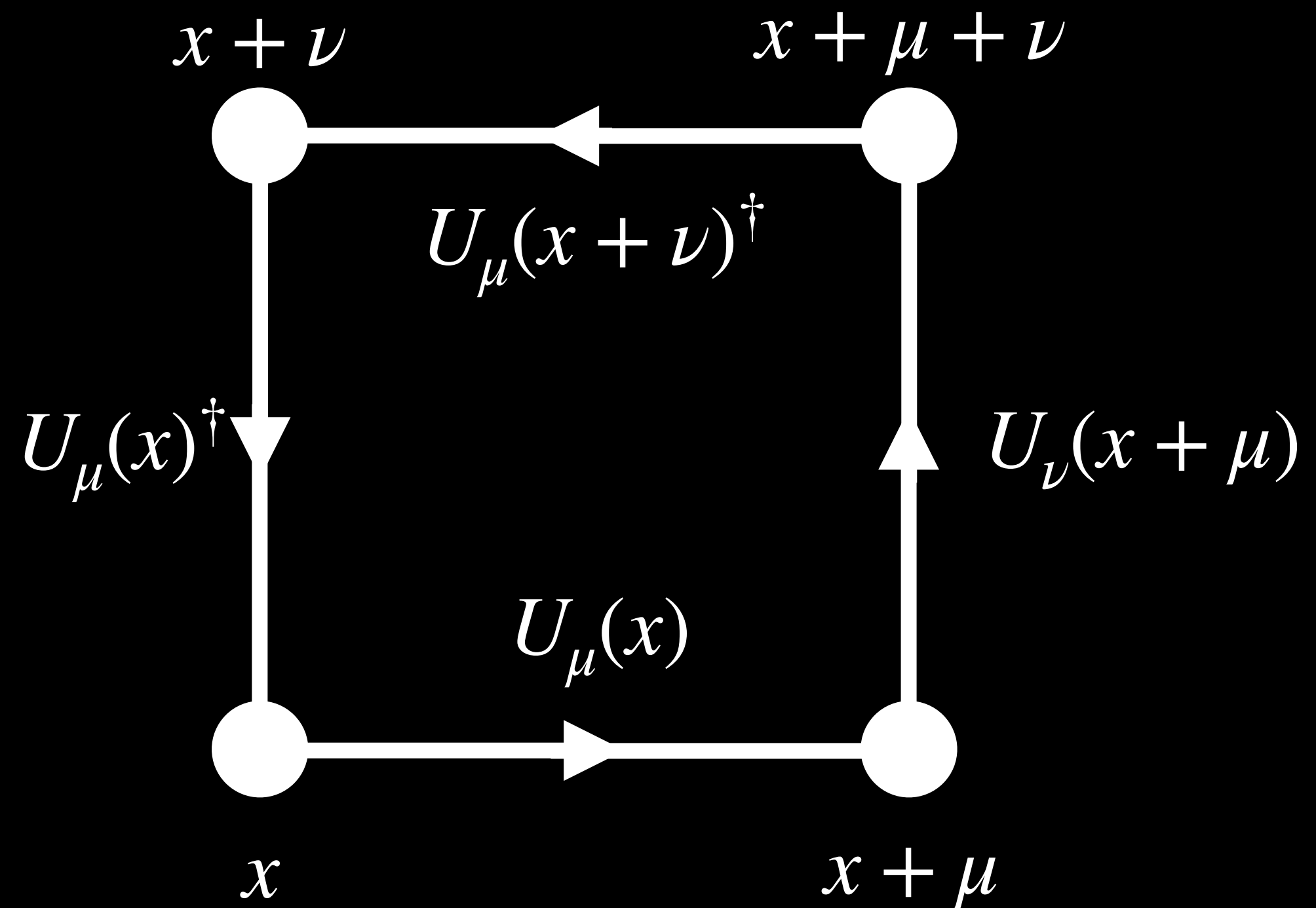


SU(2) Hamiltonian



Derivative:

SU(2) Hamiltonian

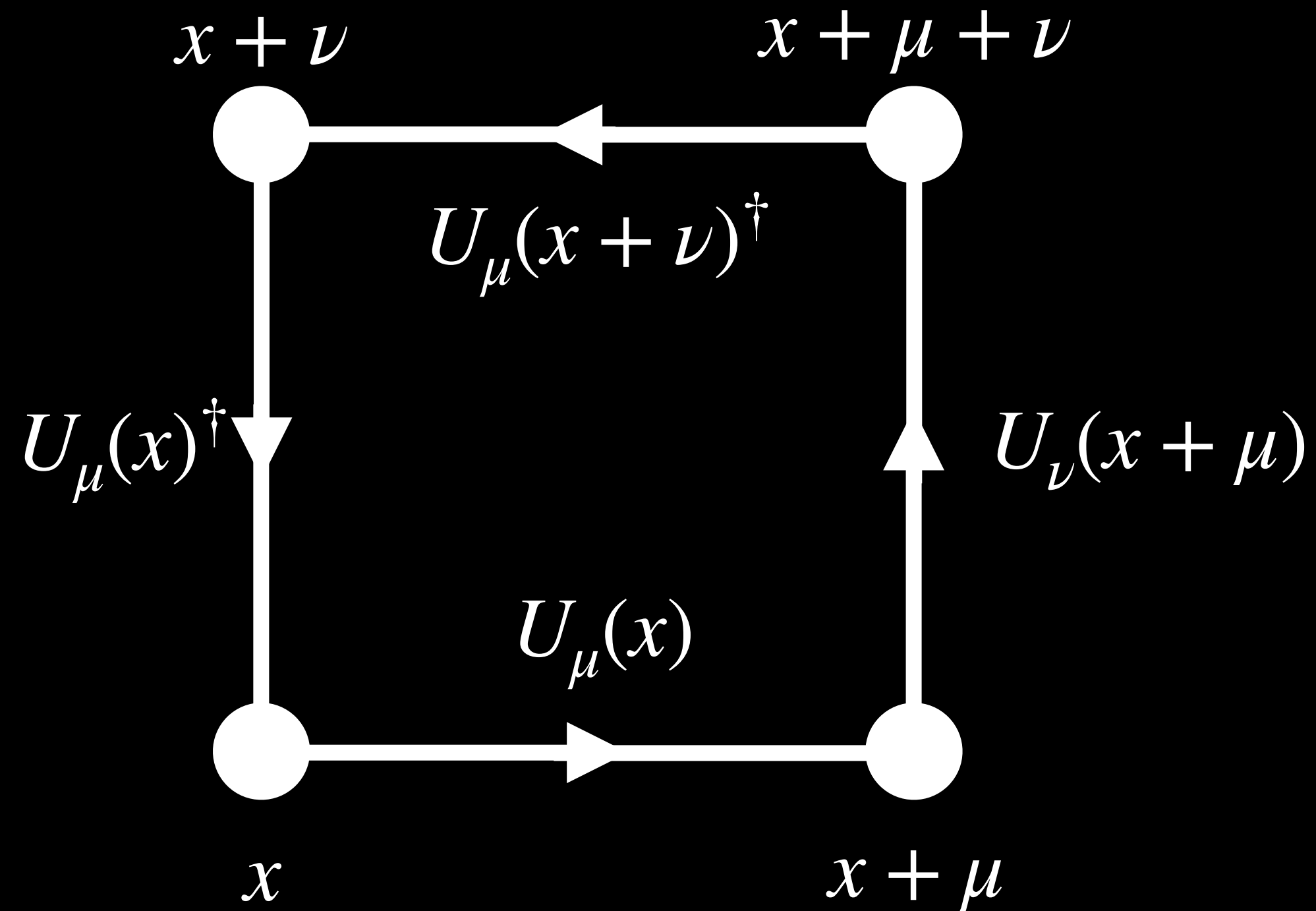


Derivative:

$$U_\mu(x) = \exp(-i\sigma_a A_\mu^a(x))$$

$$\nabla \sim \frac{\partial}{\partial A_\mu}$$

SU(2) Hamiltonian



Hamiltonian:

$$\mathcal{H} = -\frac{1}{2} \sum_l \nabla_l^2 + \lambda \sum_p \left(1 - \frac{1}{2} \text{Tr} (P_p) \right)$$

kinetic + λ potential

electric + λ magnetic

Measure ground state properties of the SU(2) LGT

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Variational Monte Carlo

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Some definitions: **variational** wavefunction

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 - “Finding ground state” == finding **a function that gives** four* numbers

$$|\psi_\theta\rangle = \sum_i f_\theta(\phi_i) |\phi_i\rangle$$

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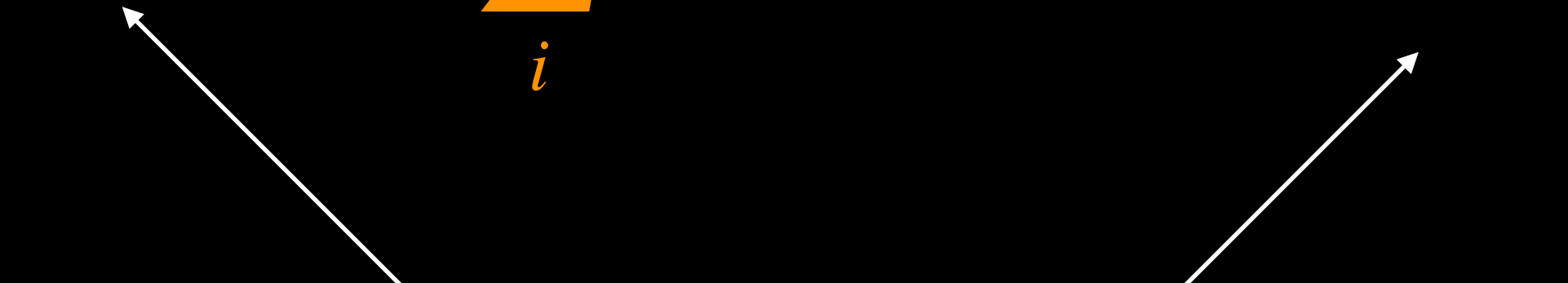
$$\langle\phi_0|\psi_\theta\rangle = \sum_i \langle\phi_0|f_\theta(\phi_i)|\phi_i\rangle = f_\theta(\phi_0)$$

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“Configuration in, amplitude out”

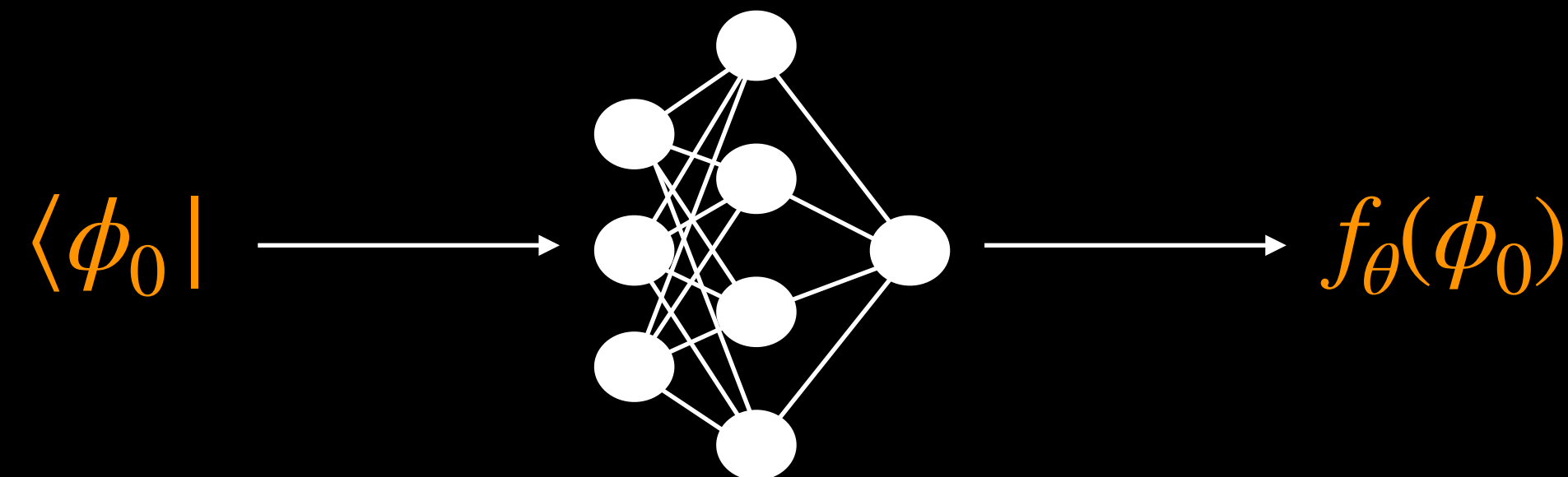


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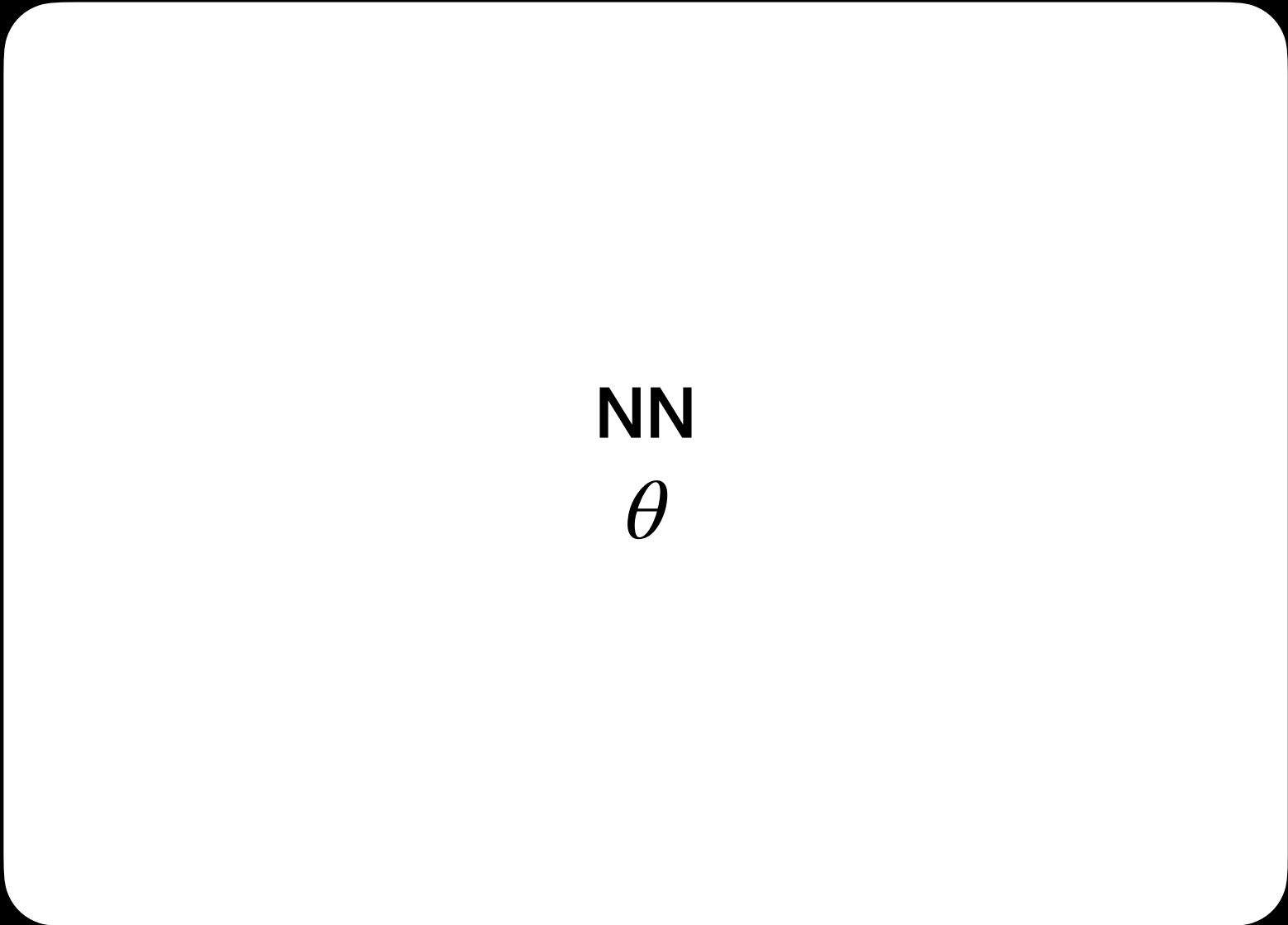
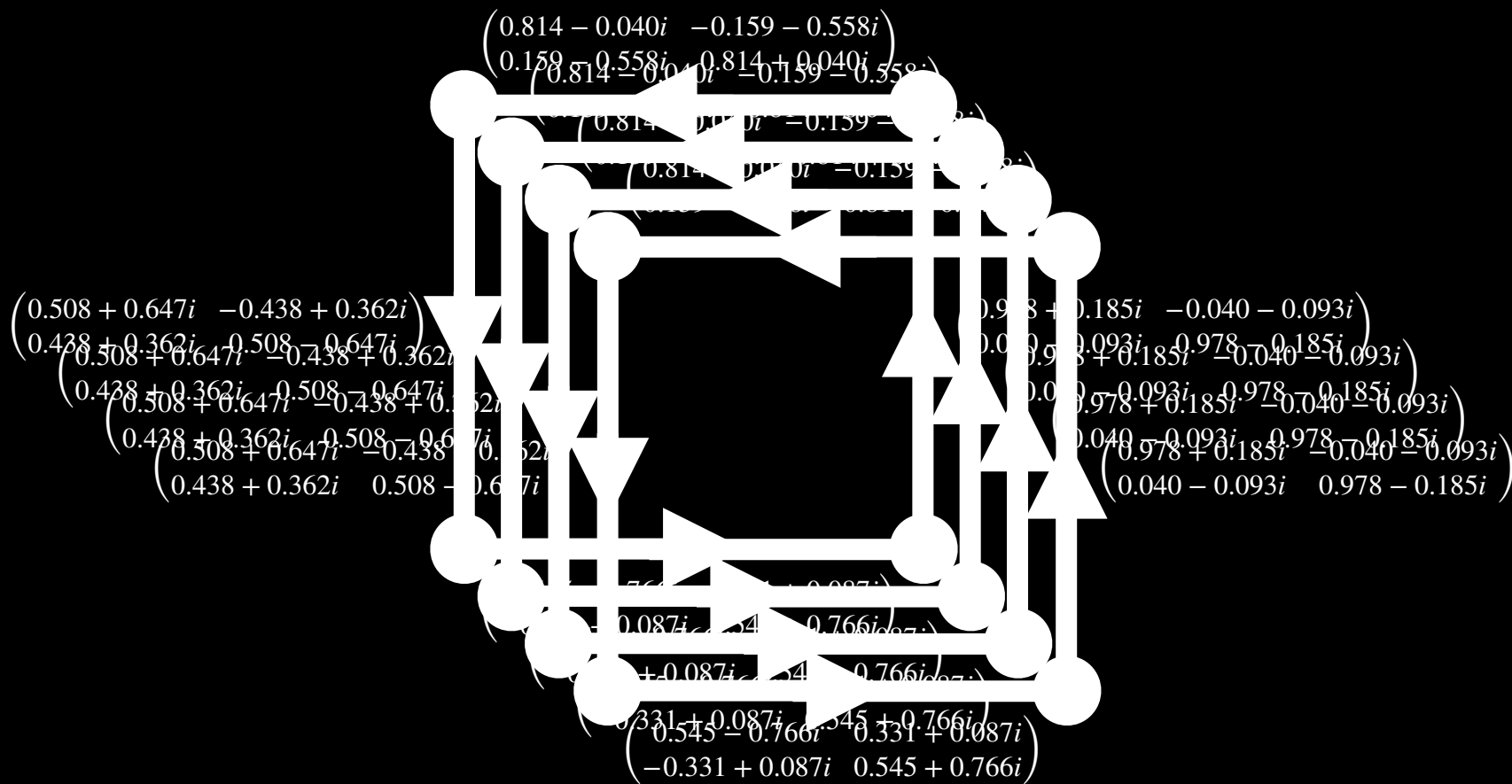
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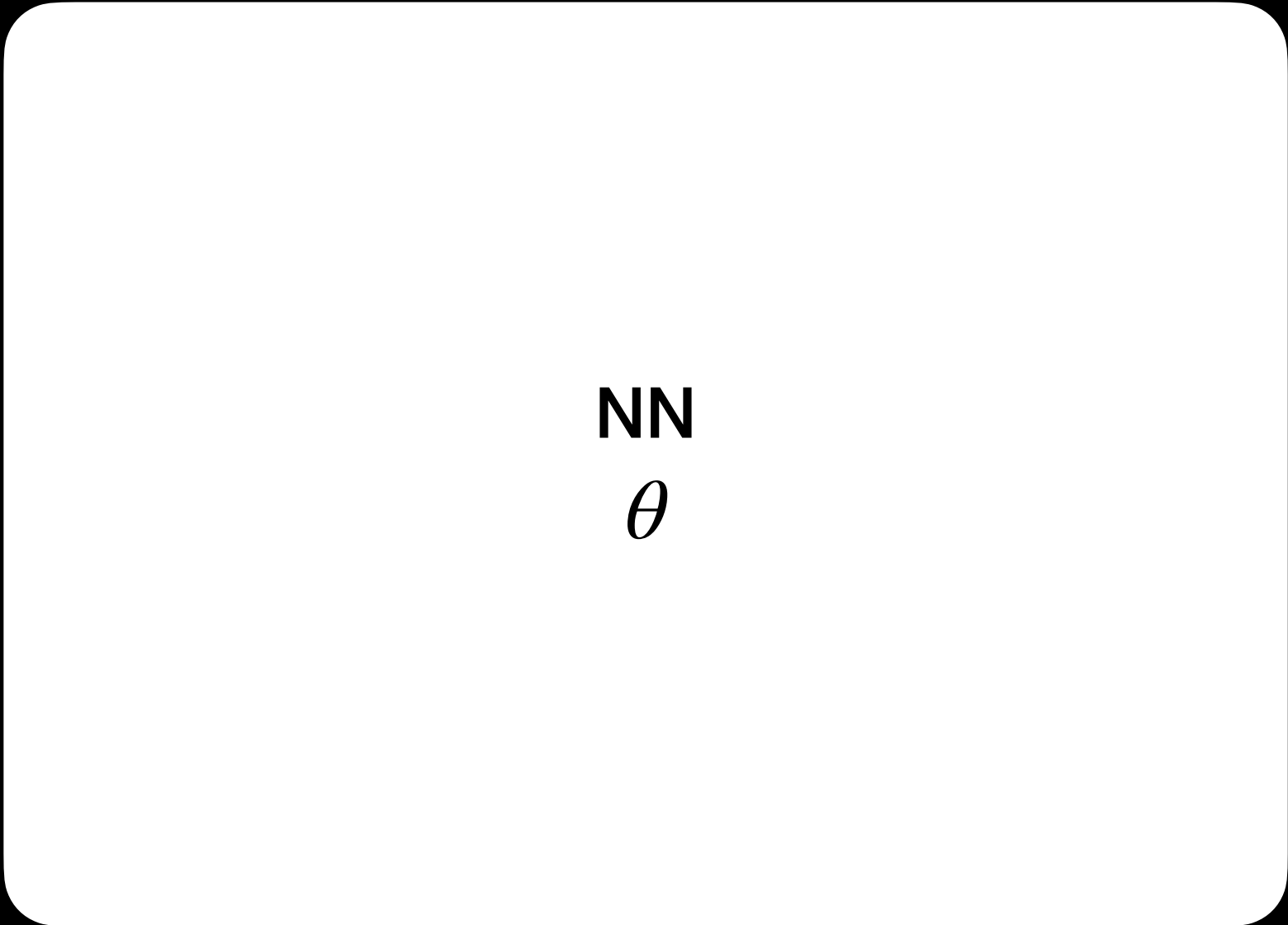
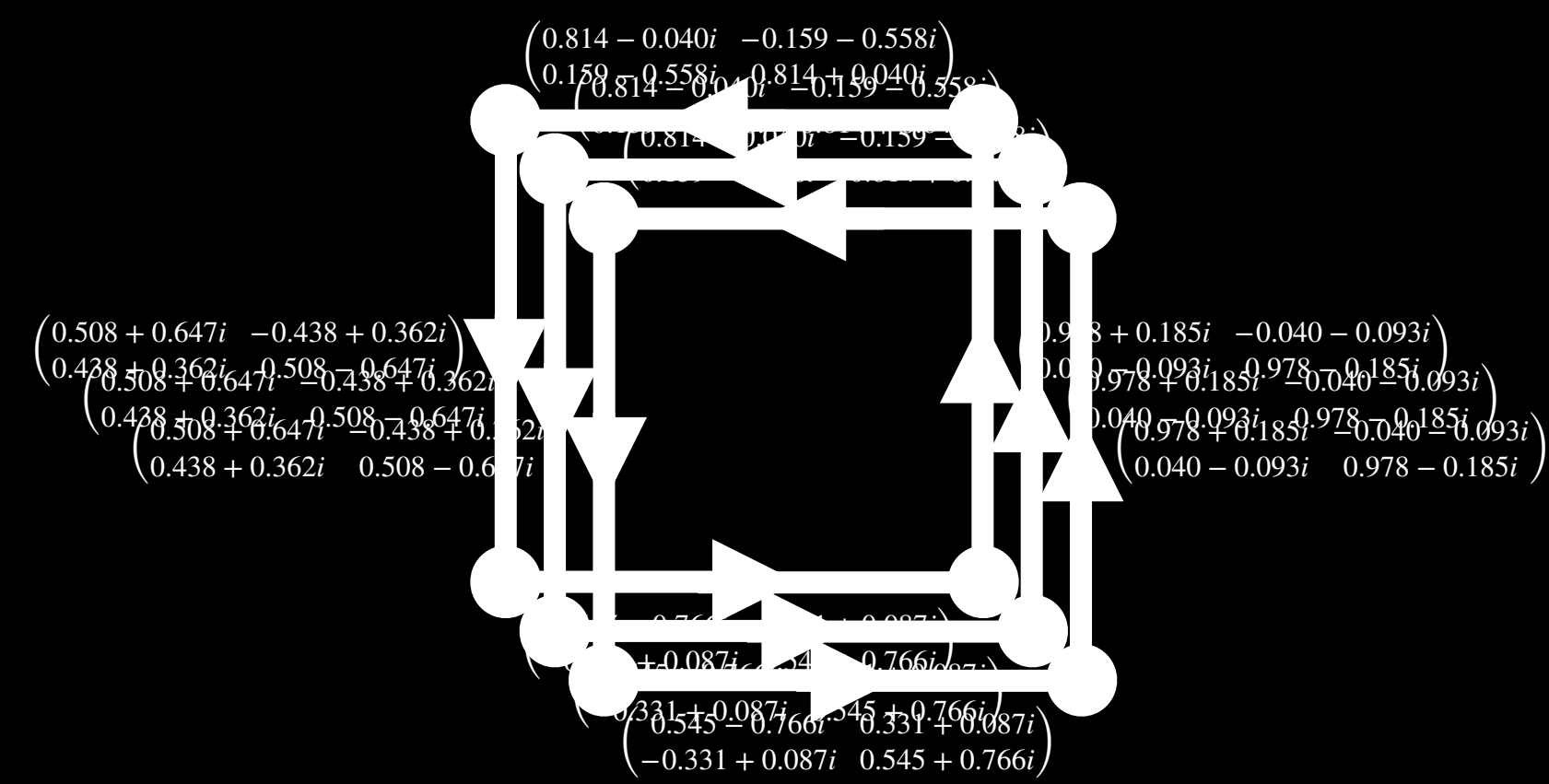
NQS workflow for LGTs

$$\{U_i\} \sim |\psi_\theta\rangle$$



NQS workflow for LGTs

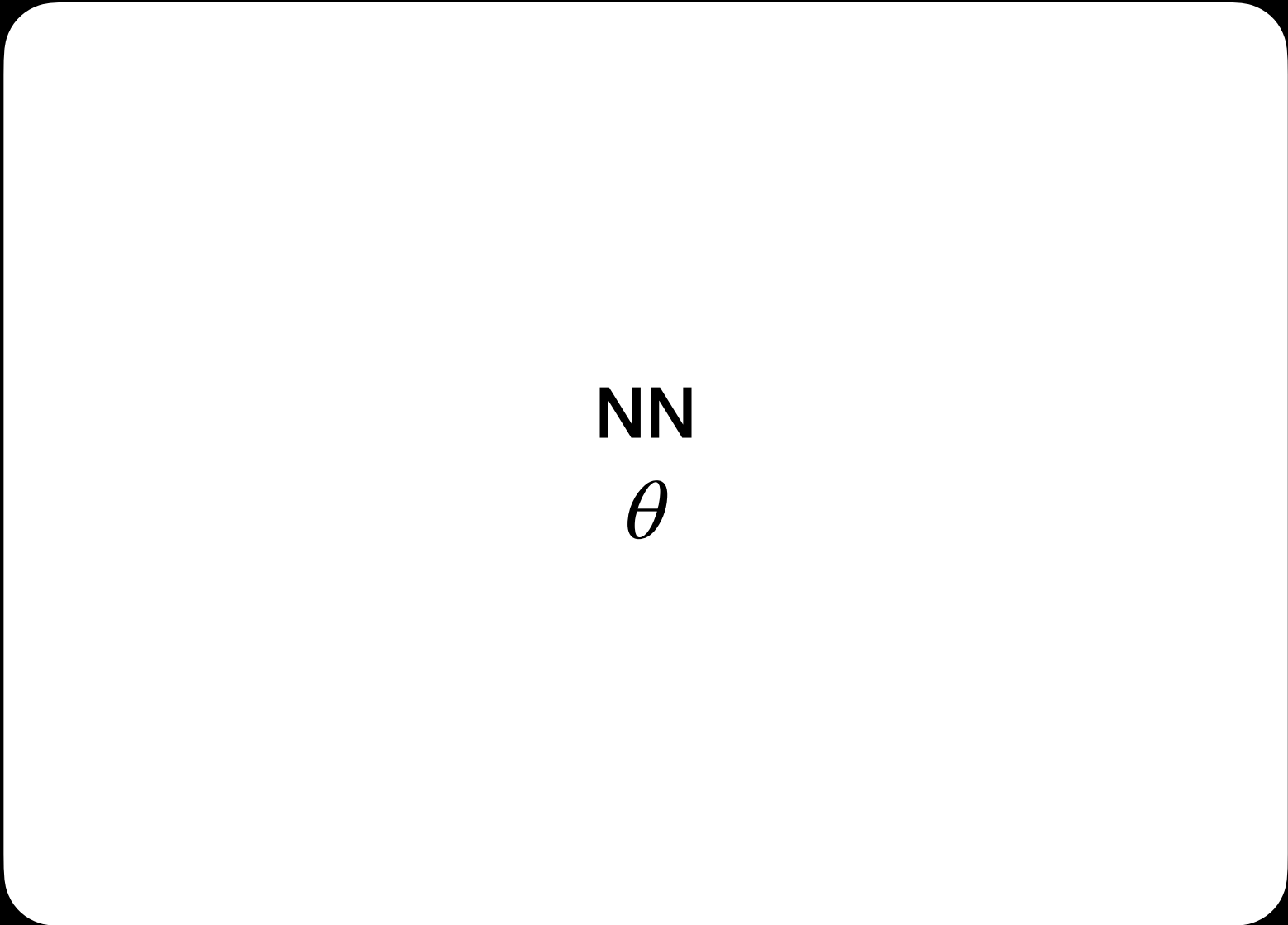
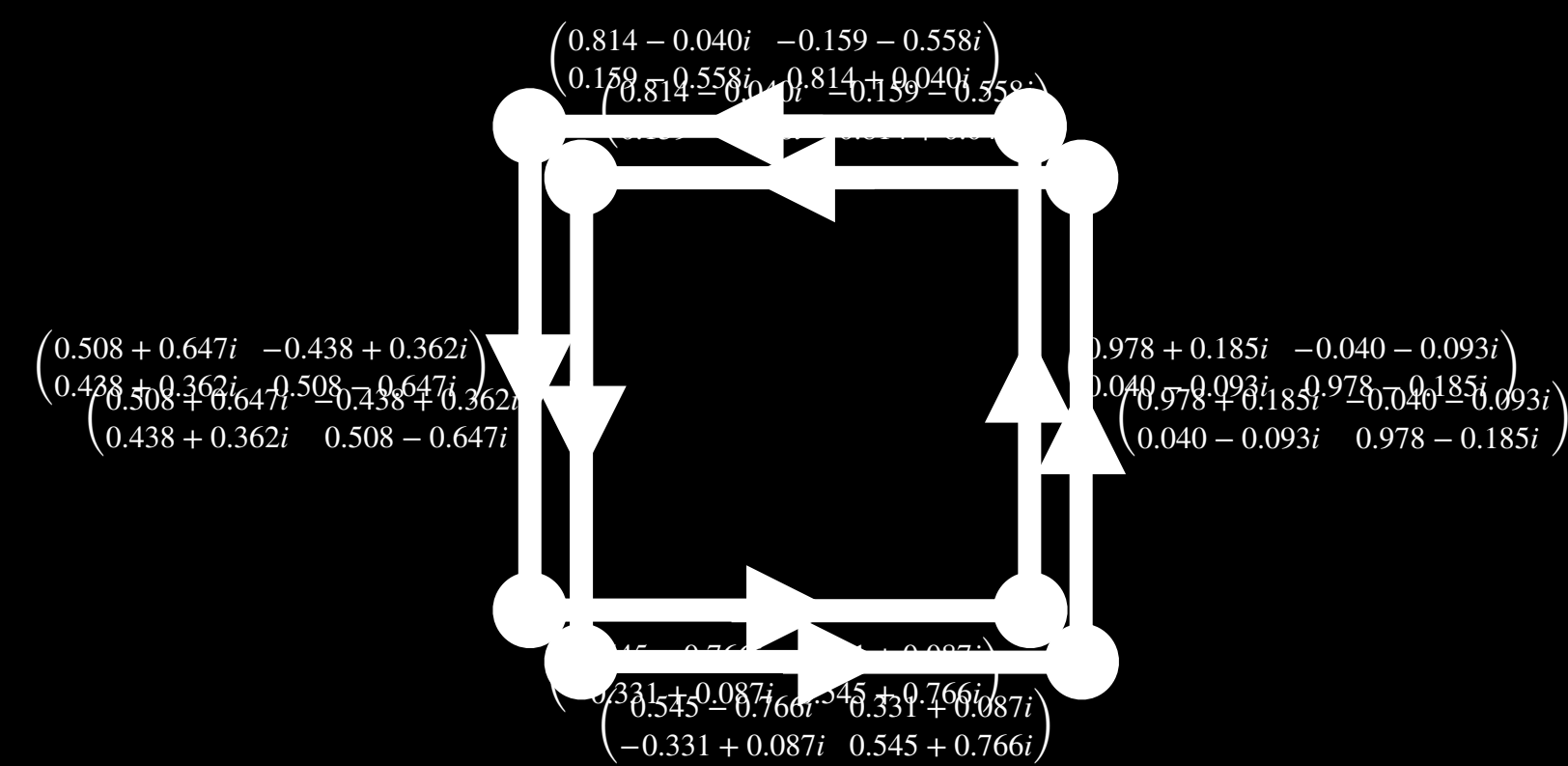
$$\{\mathbf{U}_i\} \sim |\psi_\theta\rangle$$



$$f_\theta(\mathbf{U}_i)$$

NQS workflow for LGTs

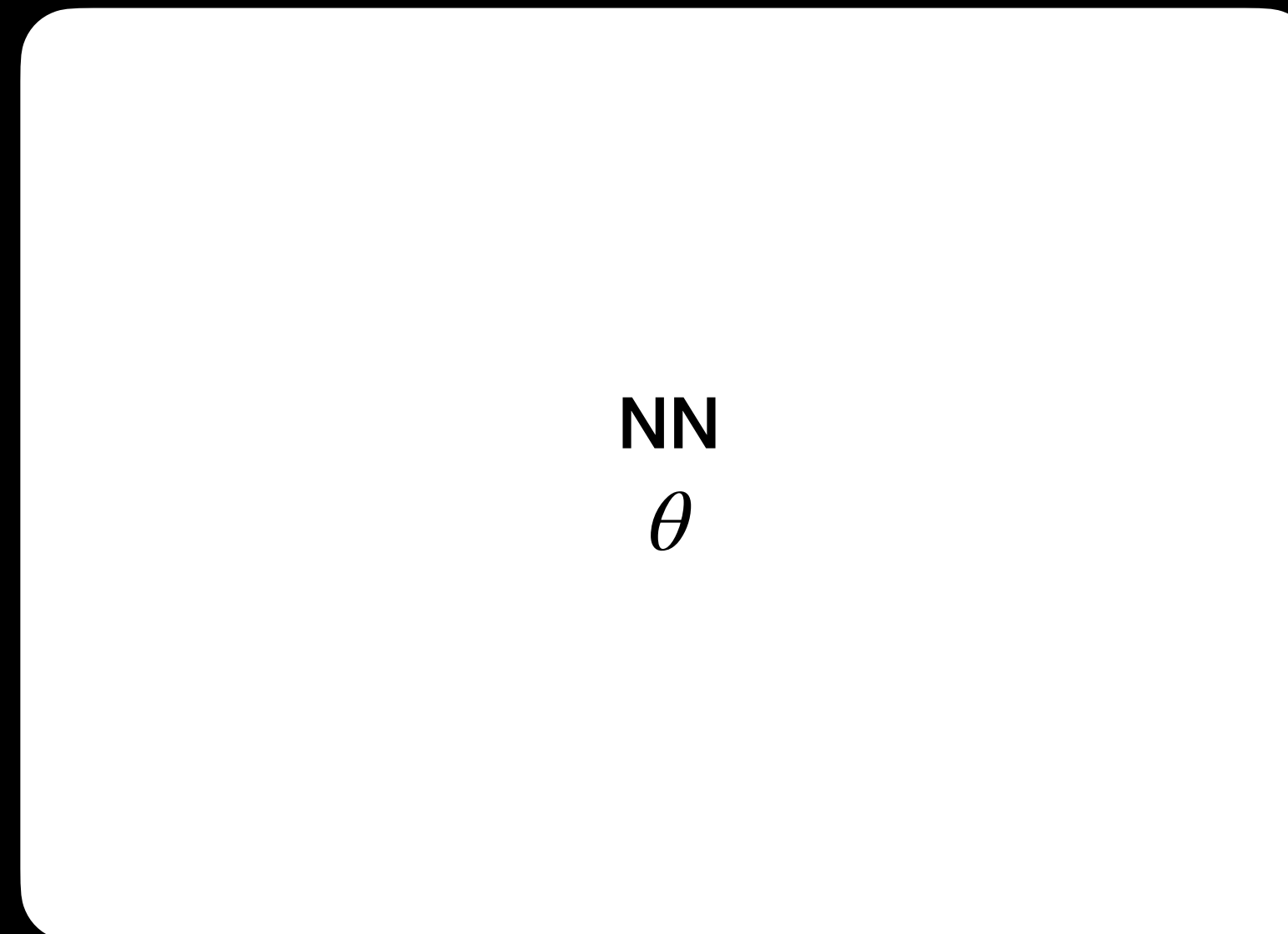
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$$f_\theta(\mathbf{U}_i)$$

$$f_\theta(\mathbf{U}_{i'})$$

NQS workflow for LGTs



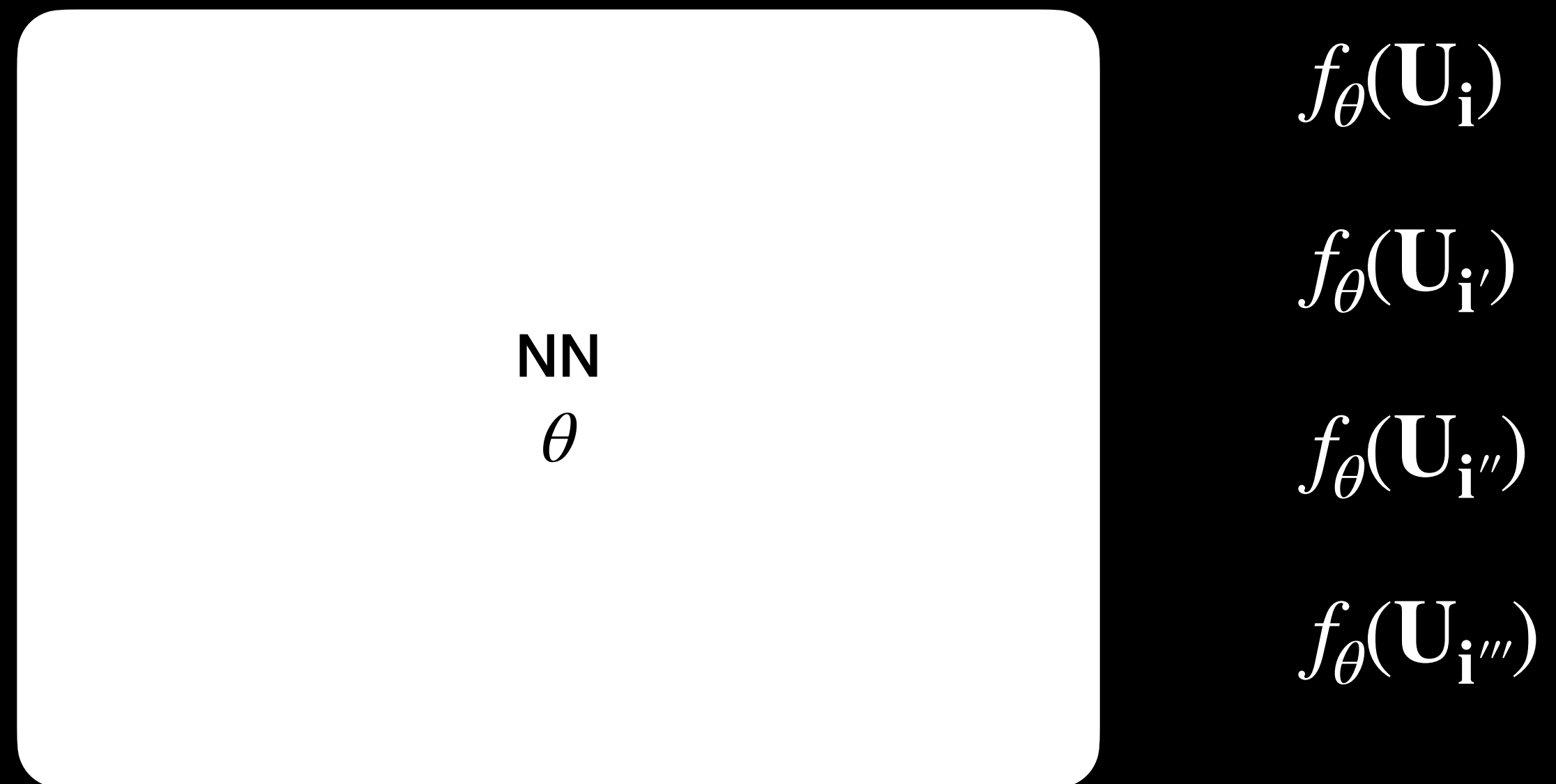
$$f_{\theta}(\mathbf{U}_i)$$

$$f_{\theta}(\mathbf{U}_{i'})$$

$$f_{\theta}(\mathbf{U}_{i''})$$

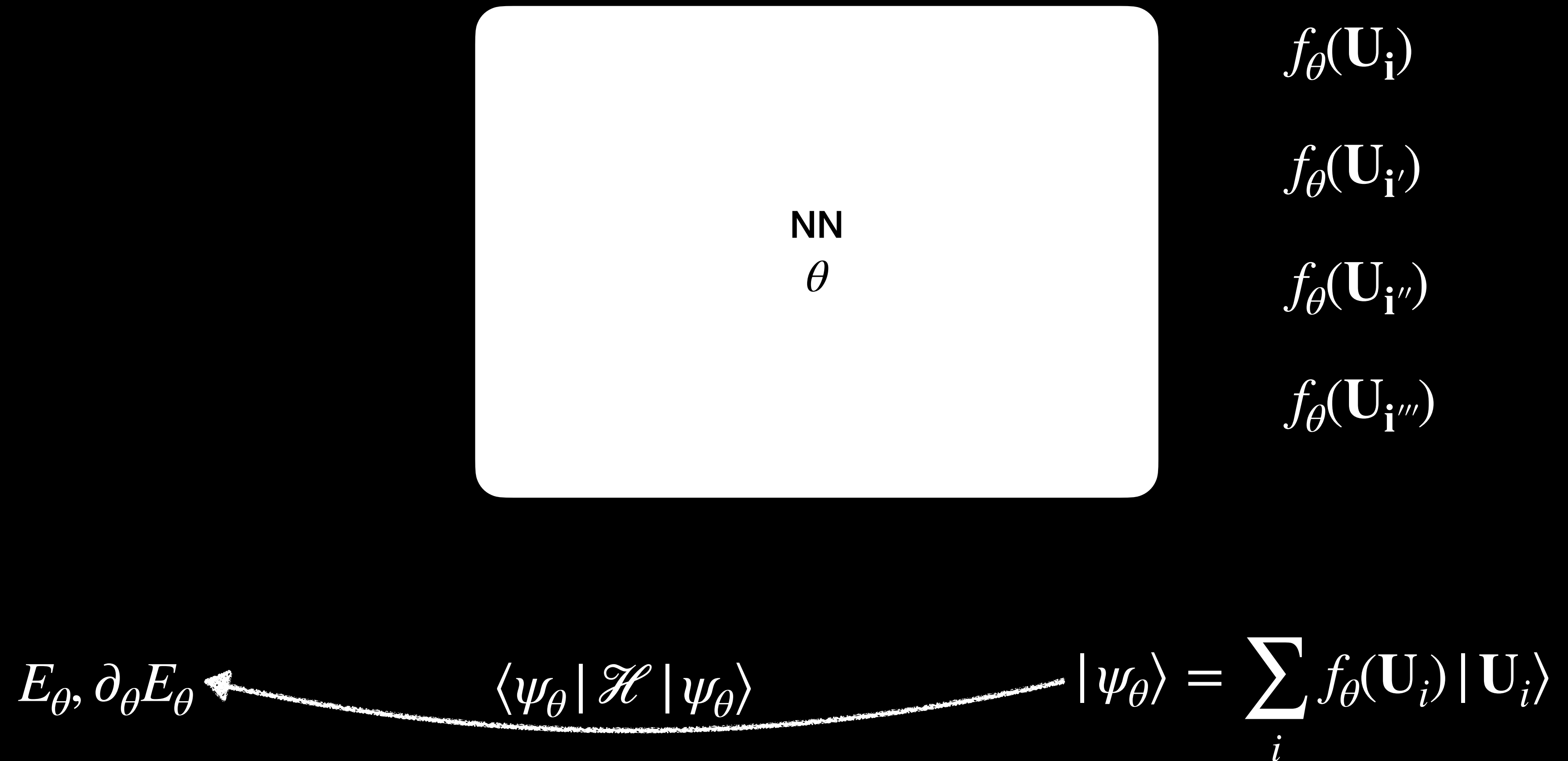
$$f_{\theta}(\mathbf{U}_{i'''})$$

NQS workflow for LGTs

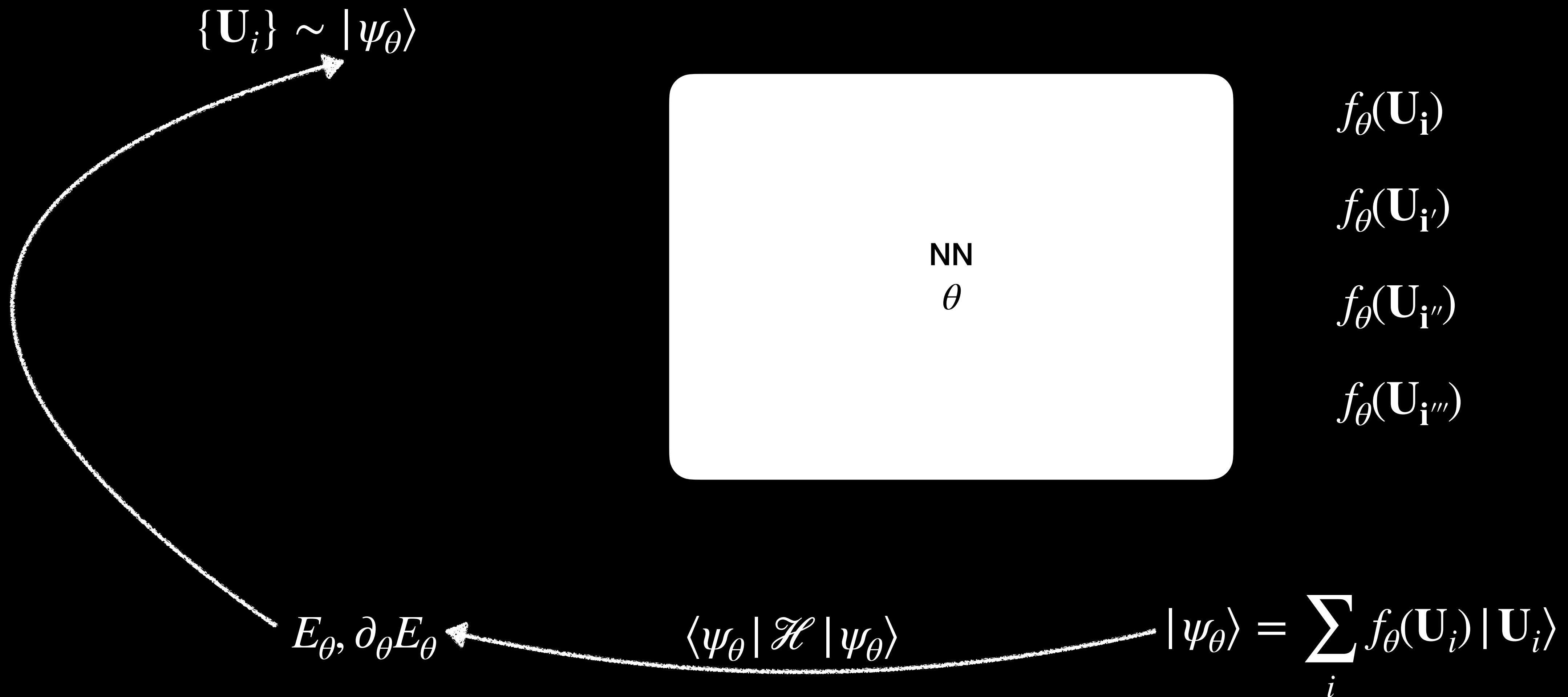


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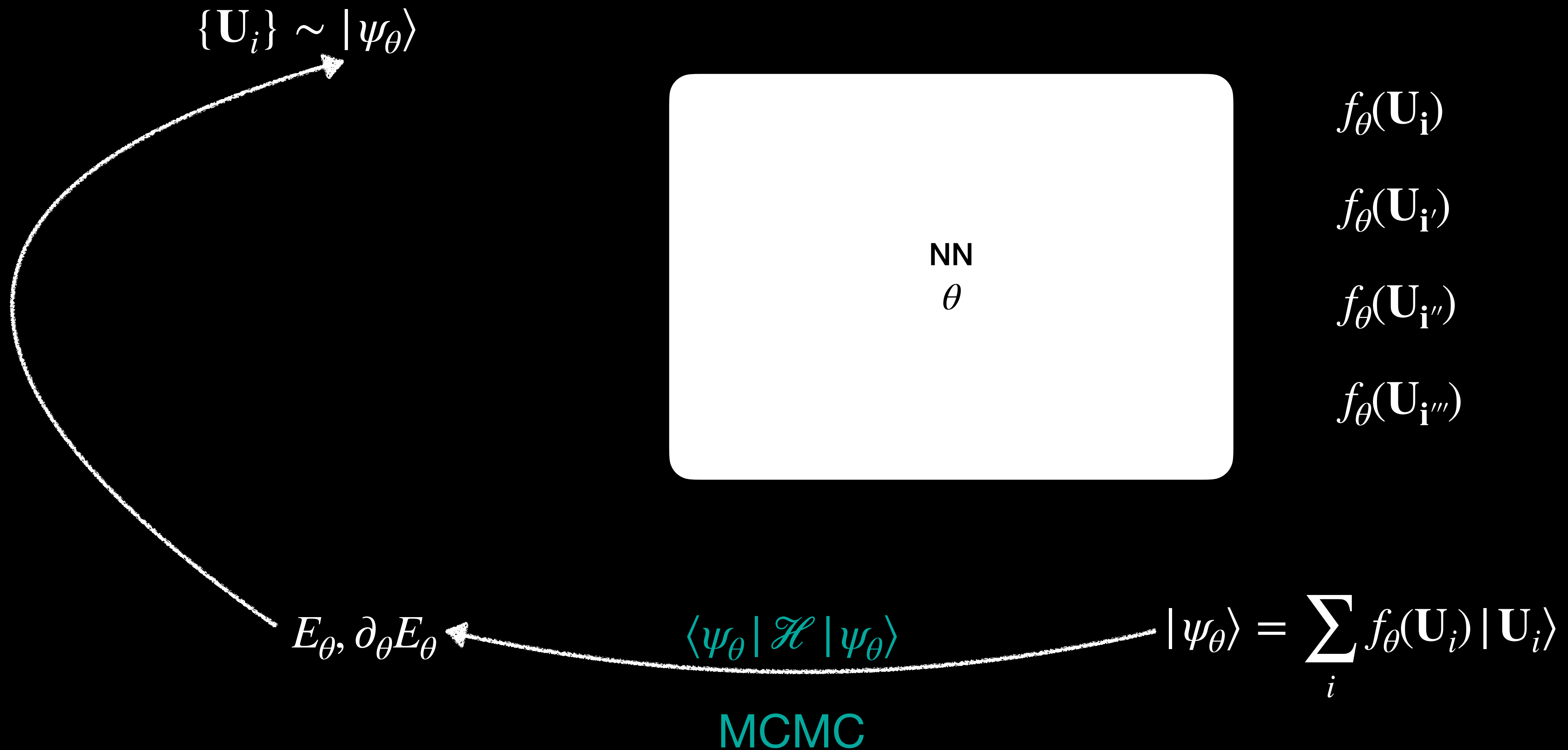
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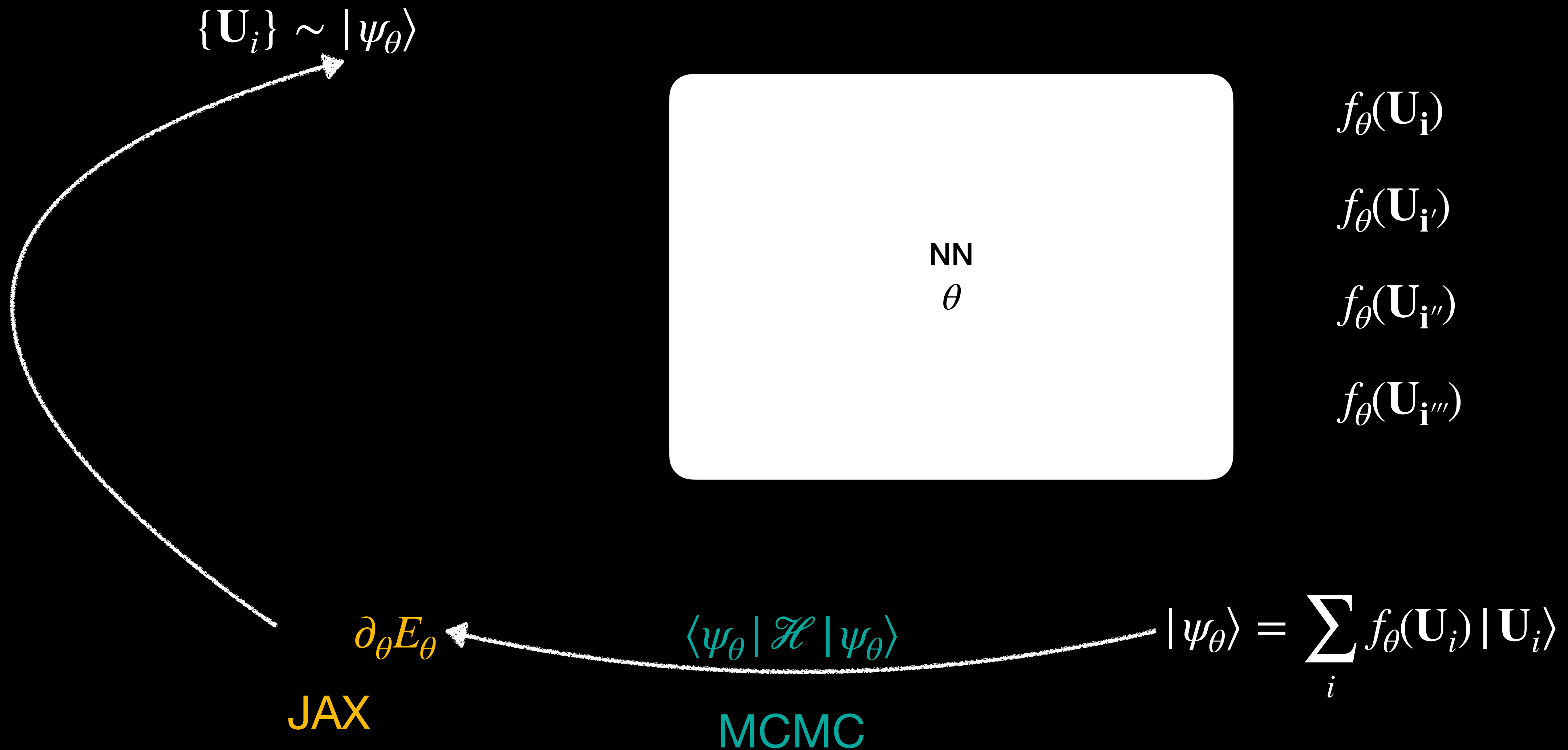
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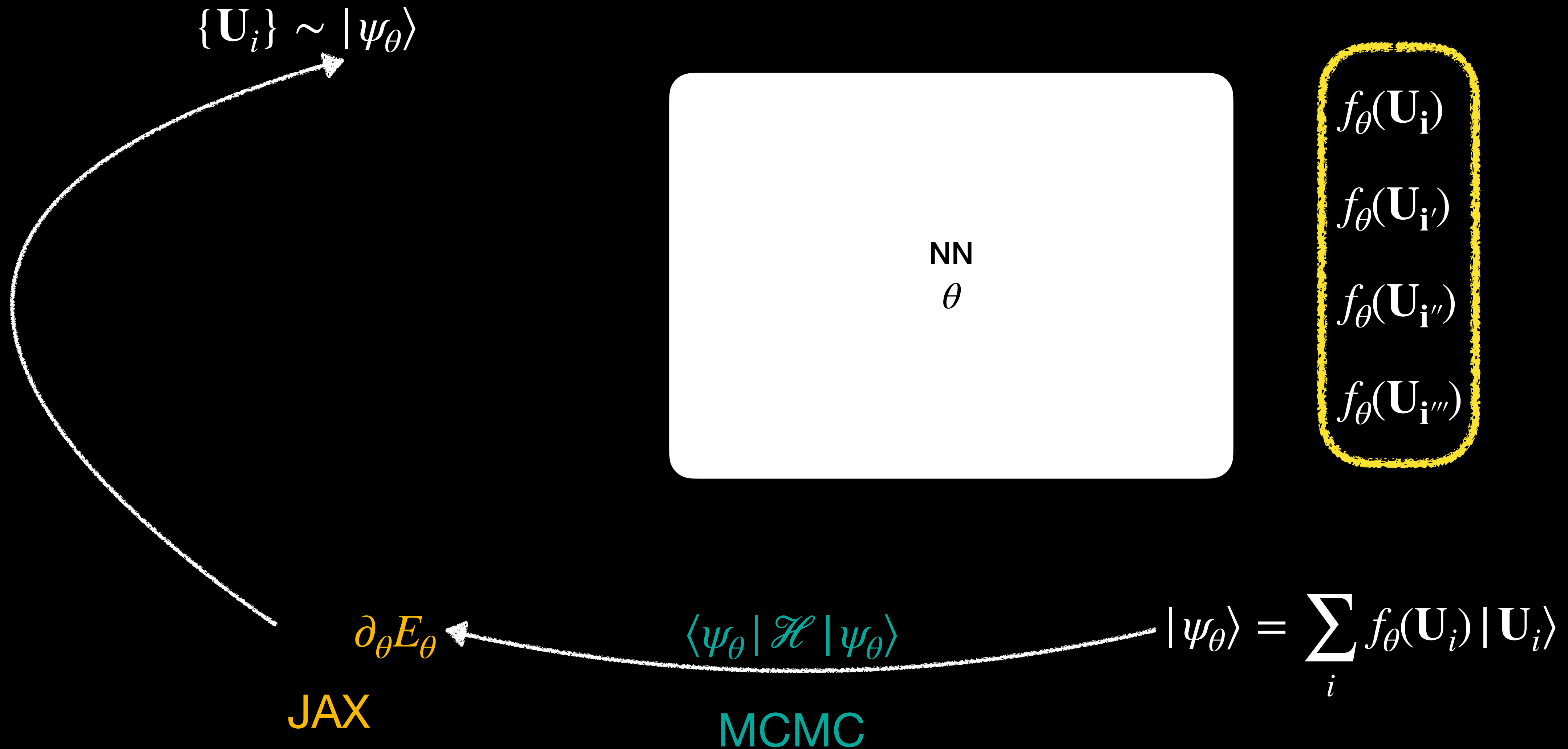


NQS workflow for LGTs



NQS workflow for LGTs

Magnetic basis?



VMC in other contexts

- Frustrated quantum spin systems

Viteritti, L. *et al.* Phys. Rev. Lett. **130**, 236401

- Ultra-cold Fermi gasses

Kim, J. *et al.* Commun Phys **7**, 148

- Helium-4

Linteau, D. *et al.* Phys. Rev. Lett. **134**, 246001

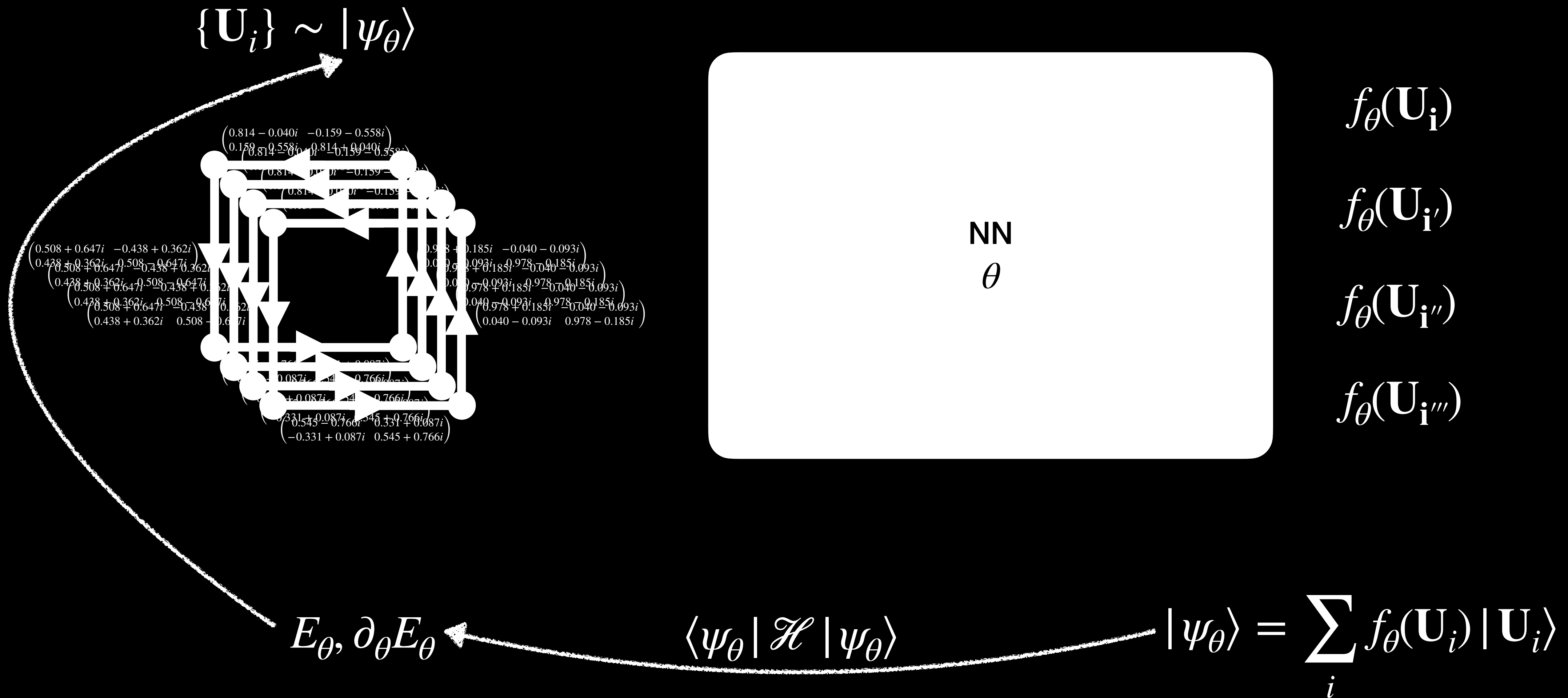
- Neutron stars

Fore, B. *et al.* Phys. Rev. Research **5**, 033062

- “Foundation” models

Scherbela, M. *et al.* arXiv 2303.09949 & Rende, R. Nat Commun **16**, 7213

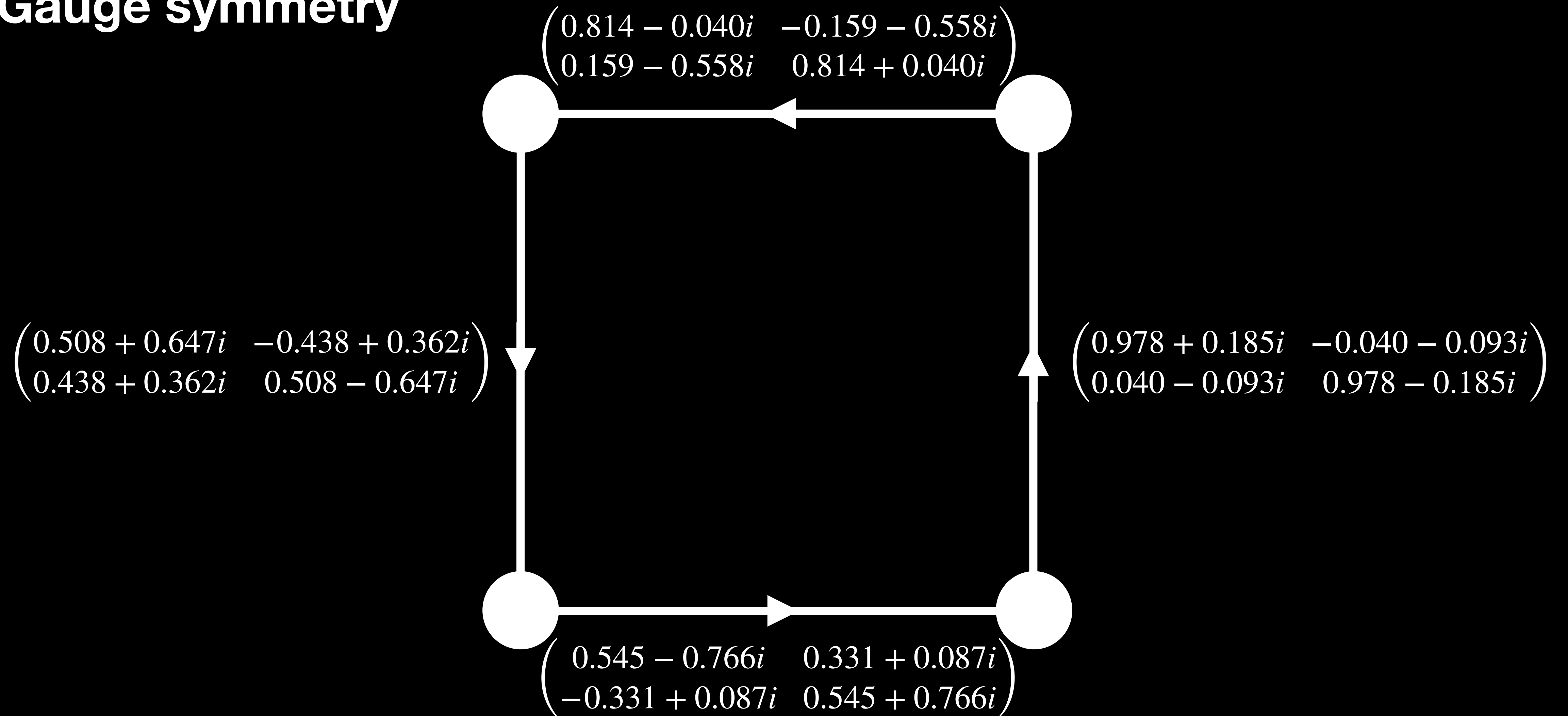
NQS workflow for LGTs



Gauge symmetry

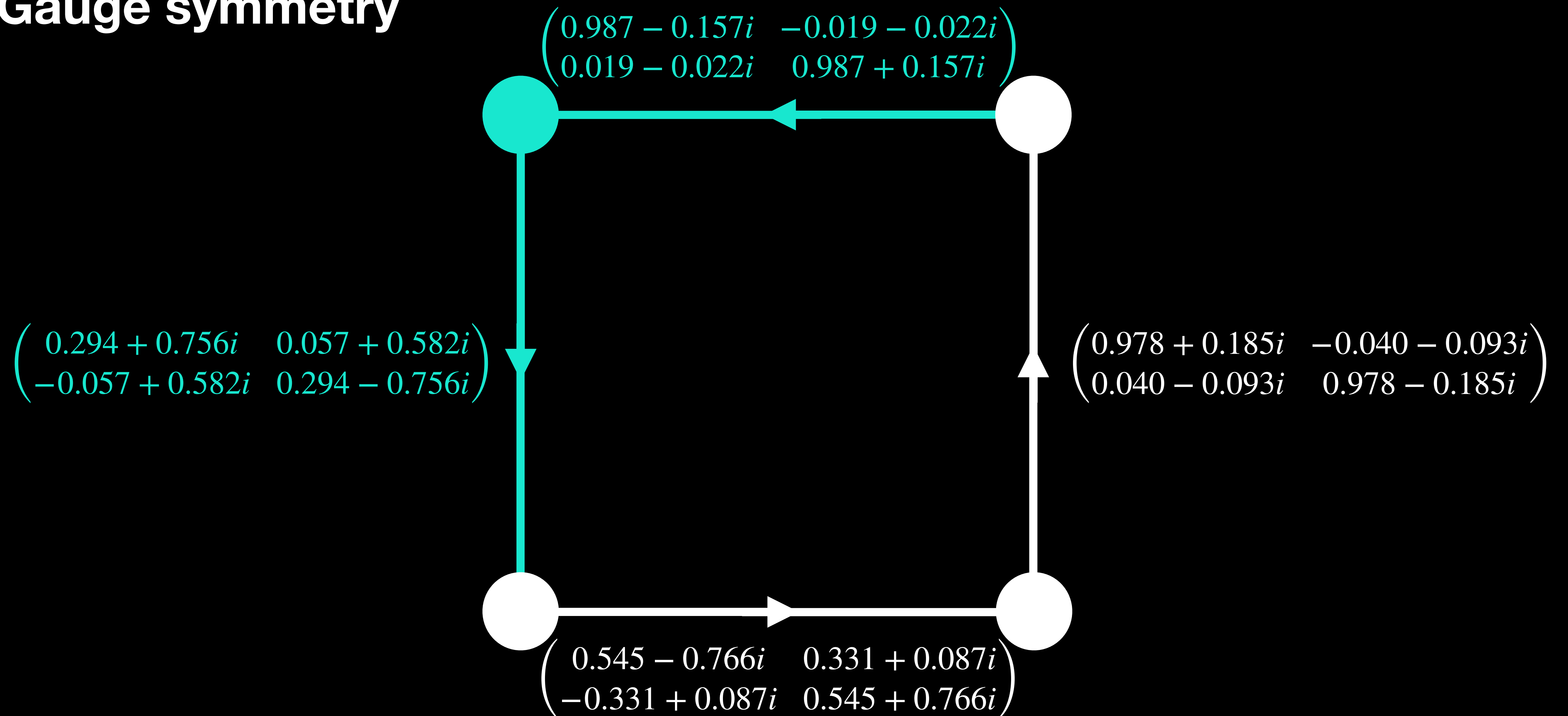
SU(2) Hamiltonian

Gauge symmetry



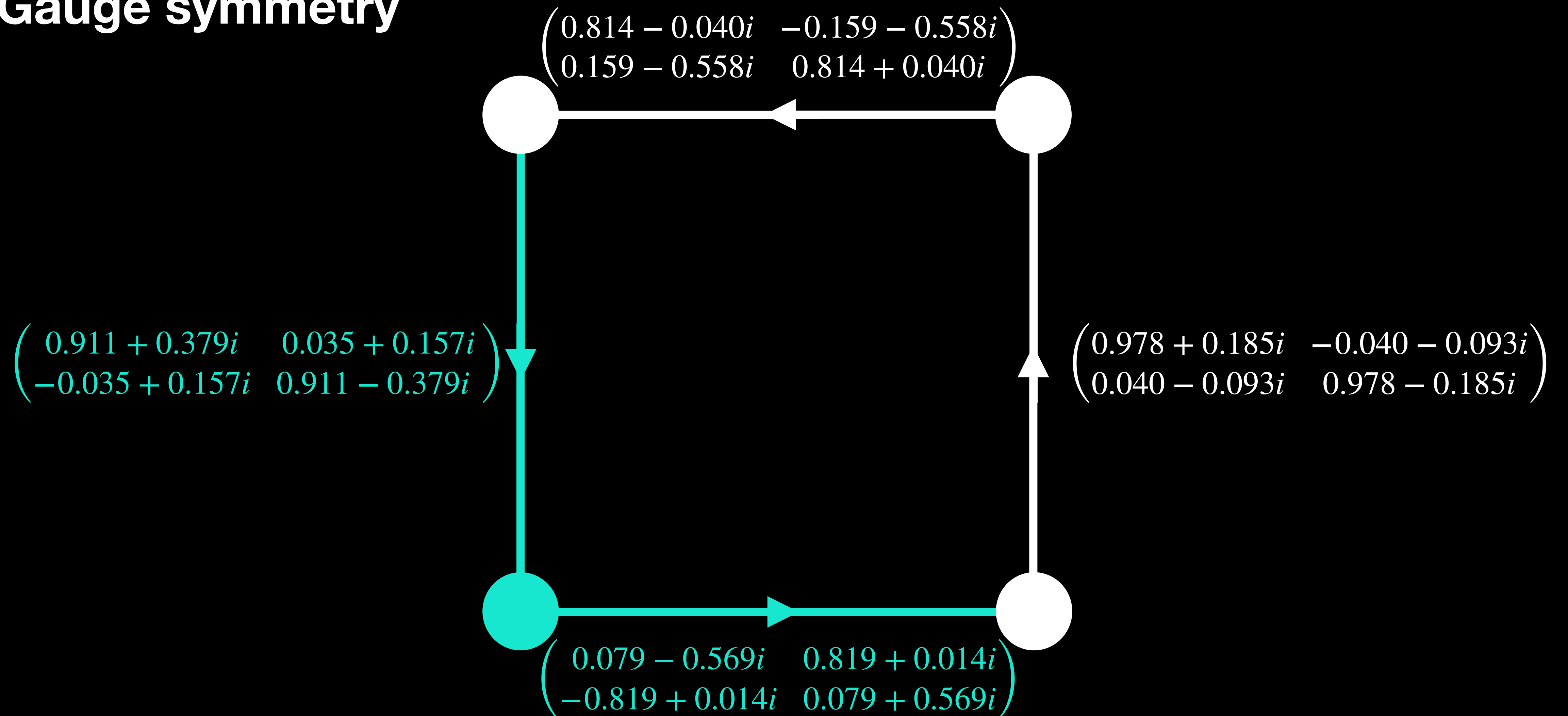
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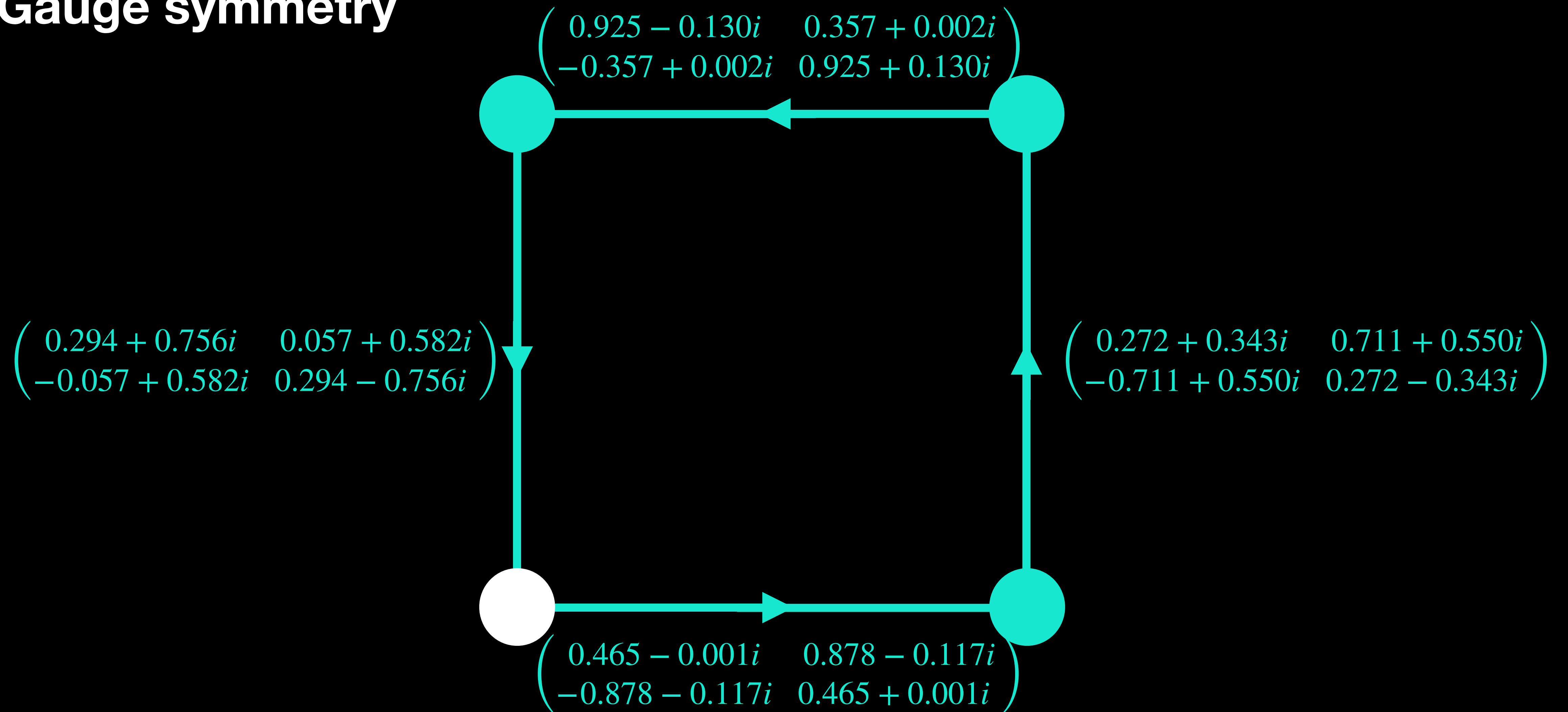
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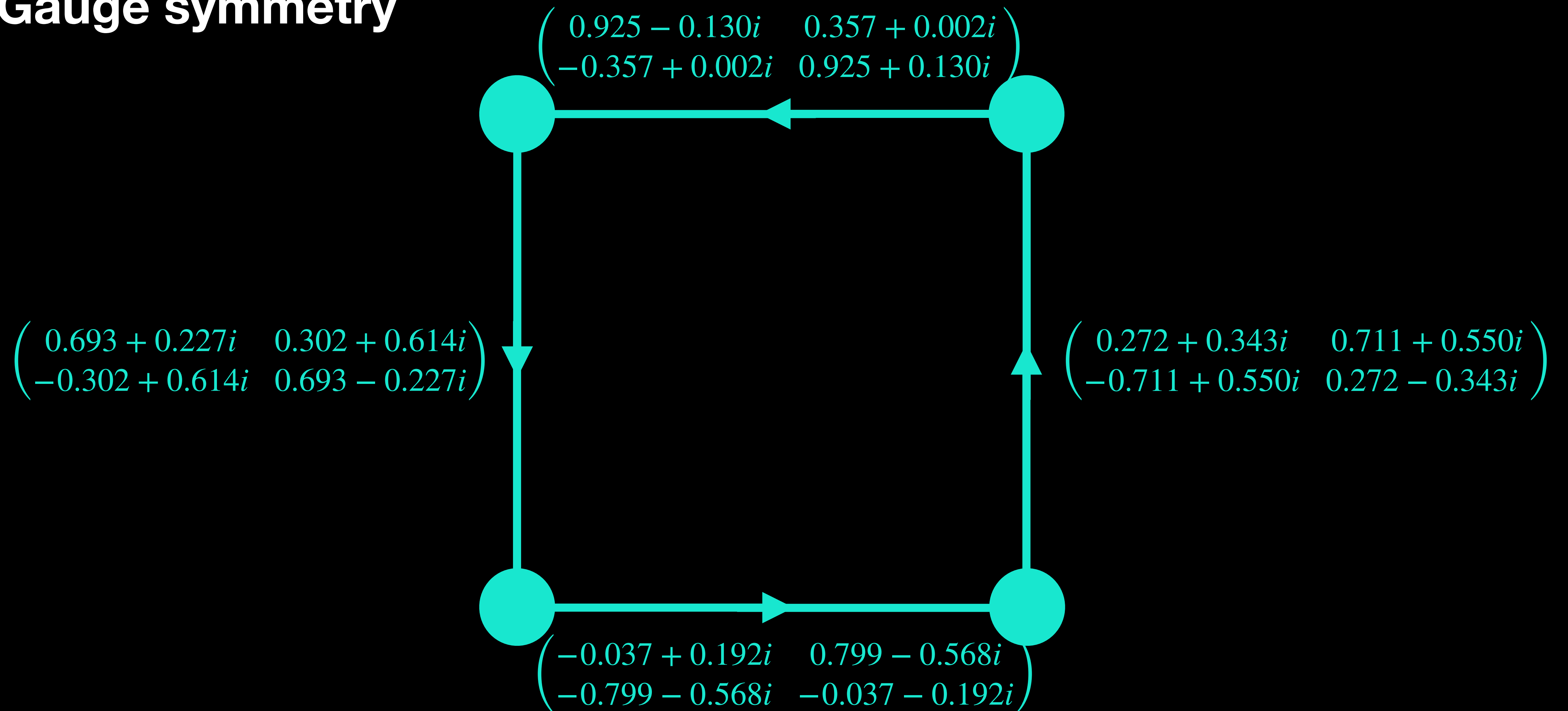
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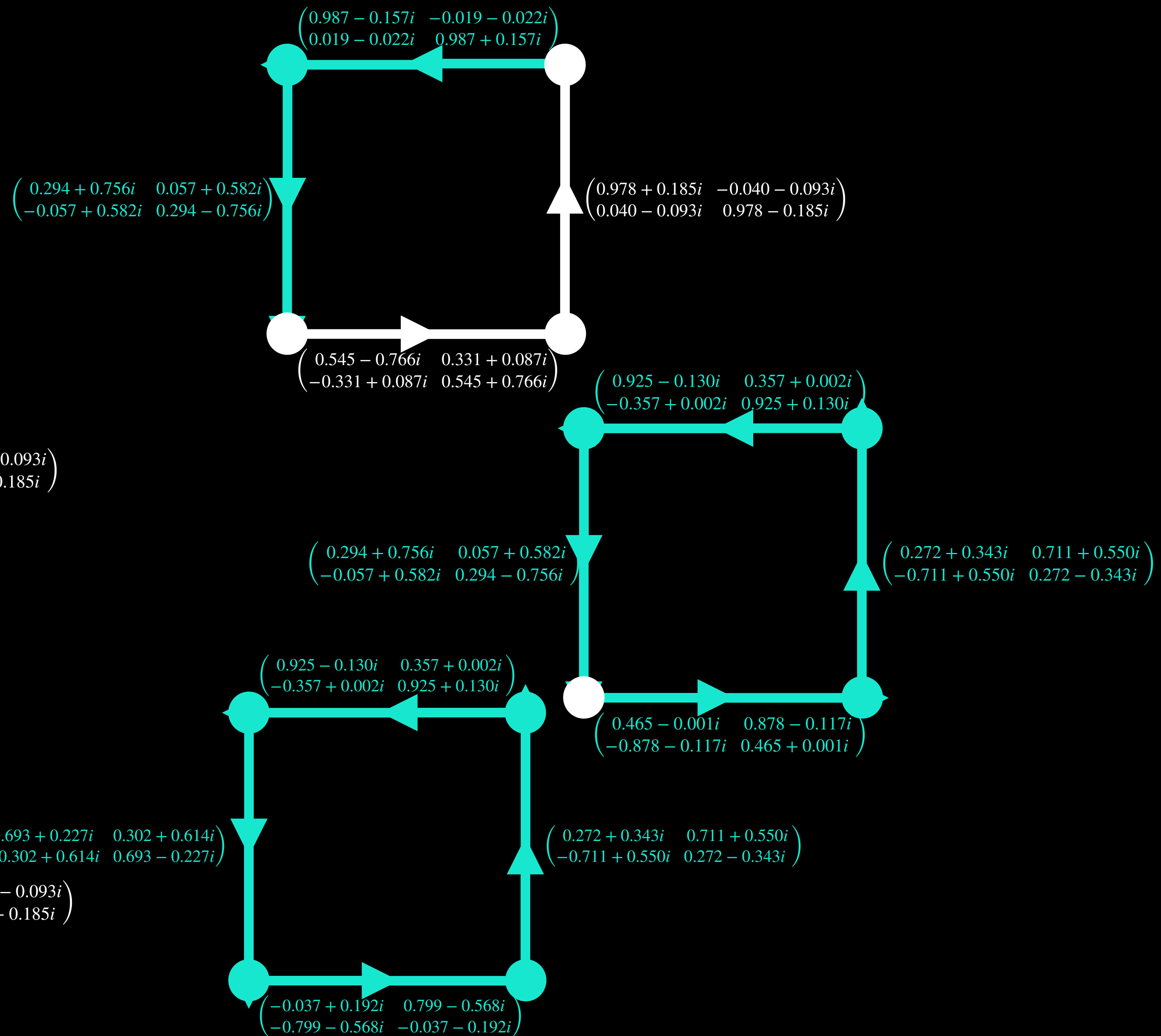
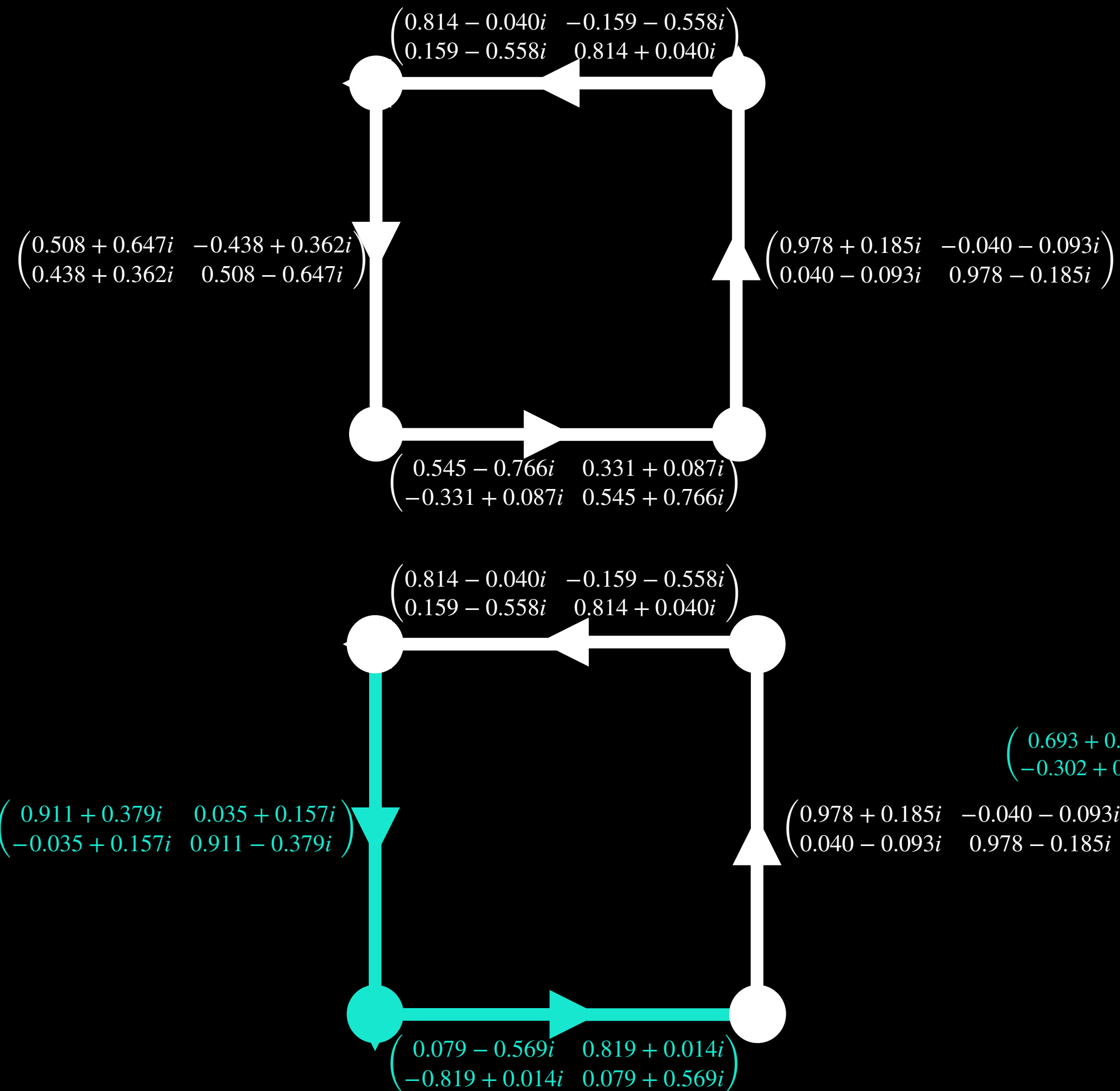
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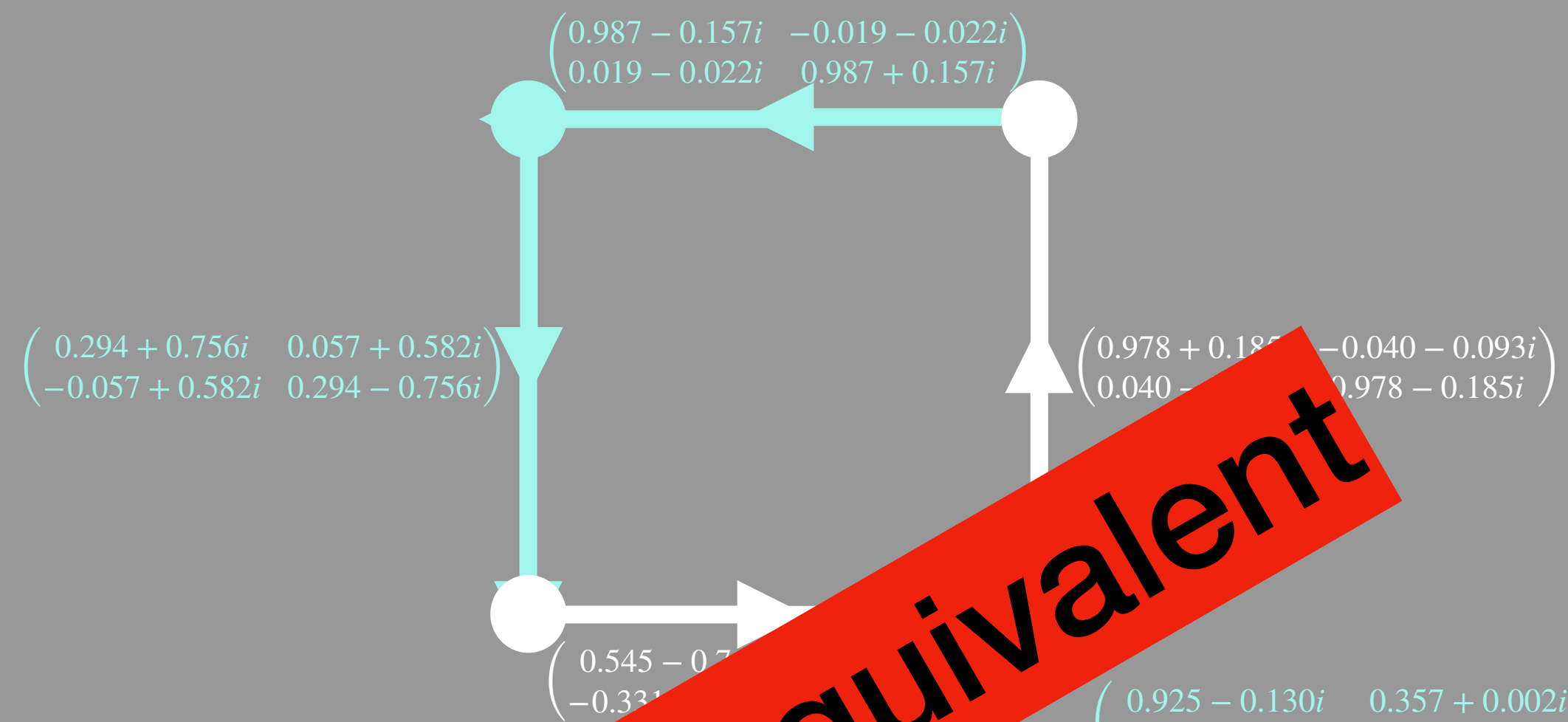
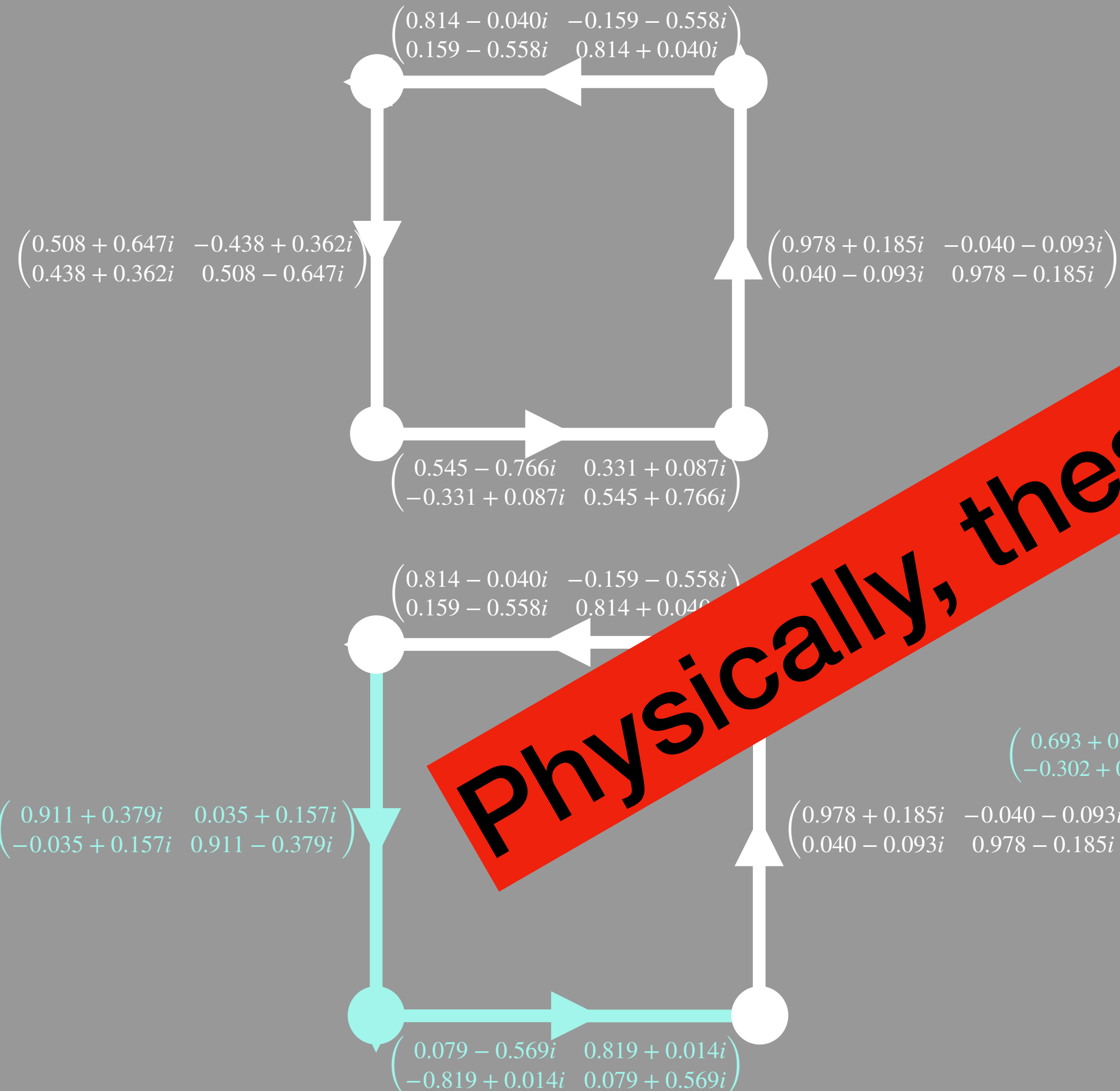
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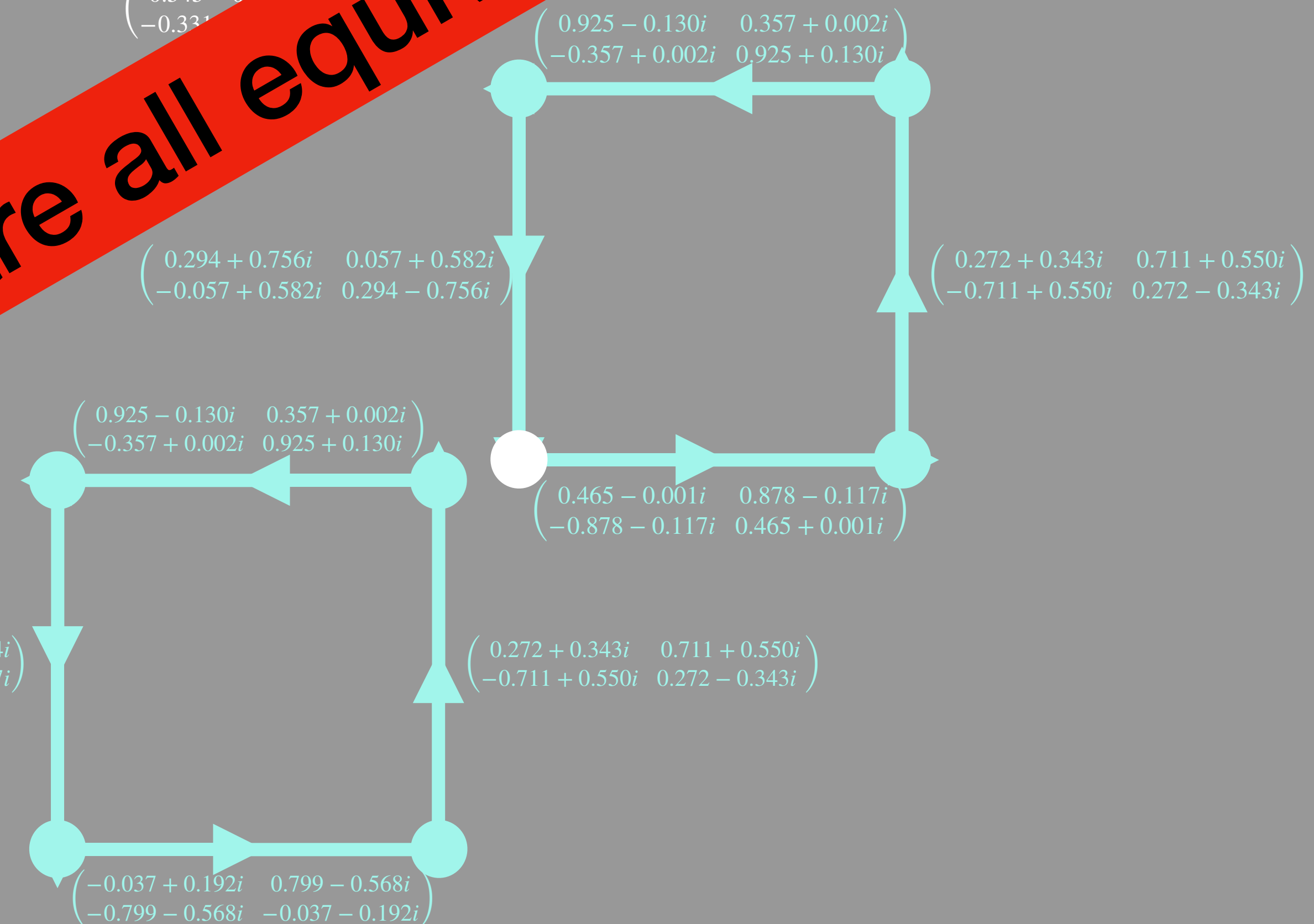


SU(2) Hamiltonian

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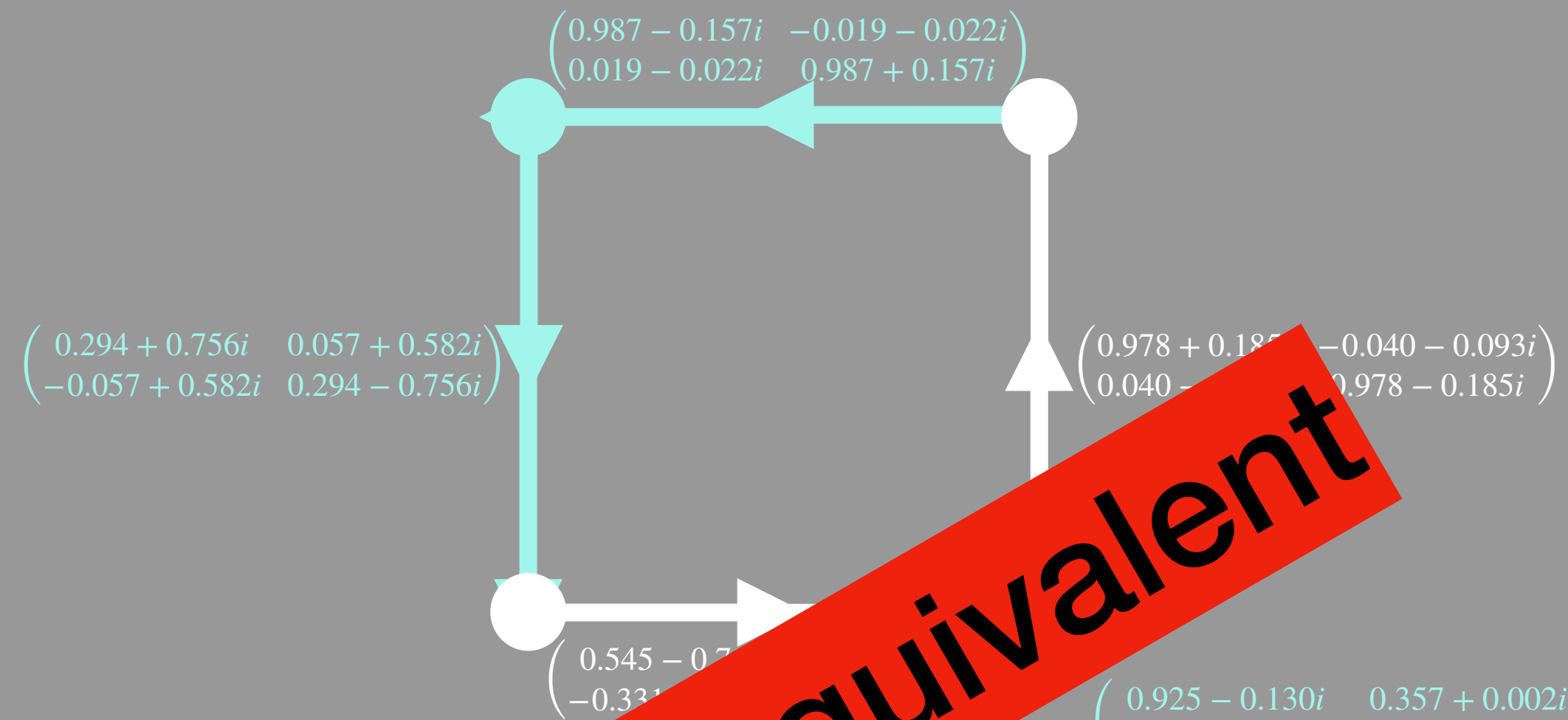
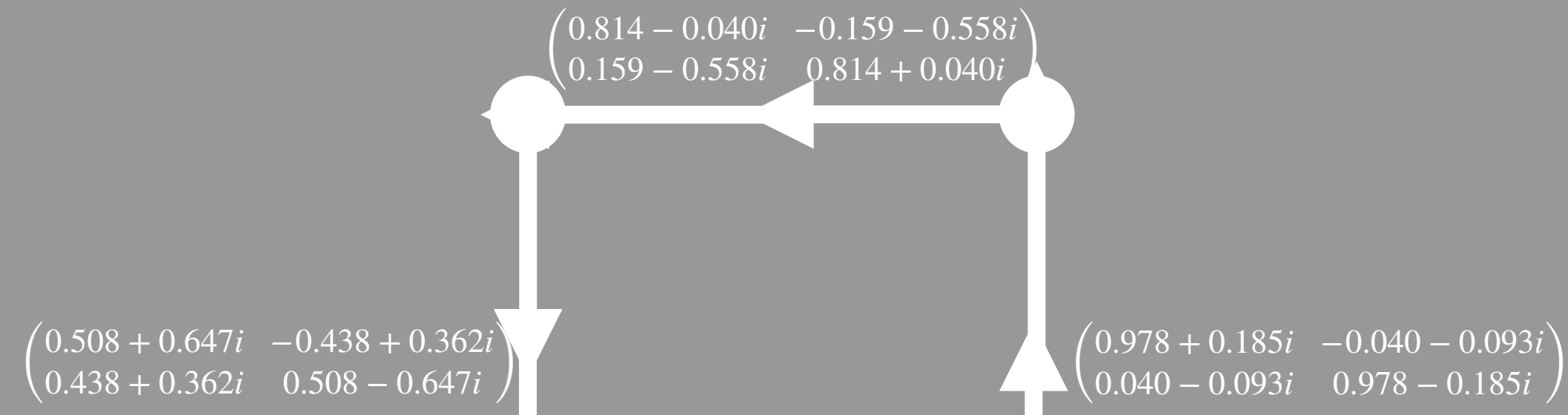


Physically, these are all equivalent

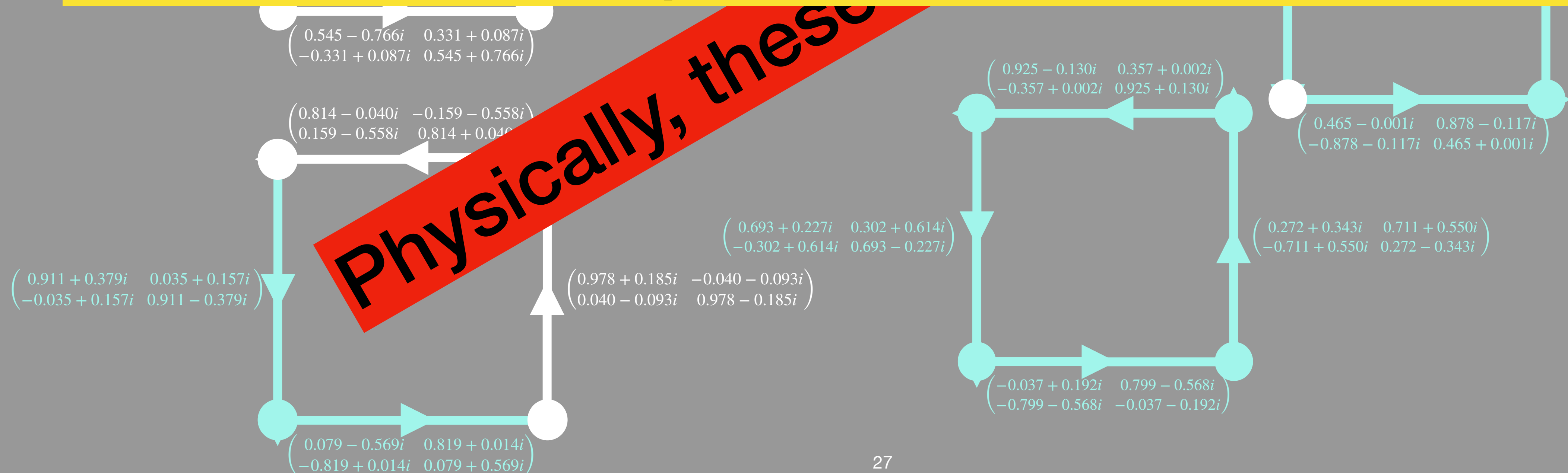


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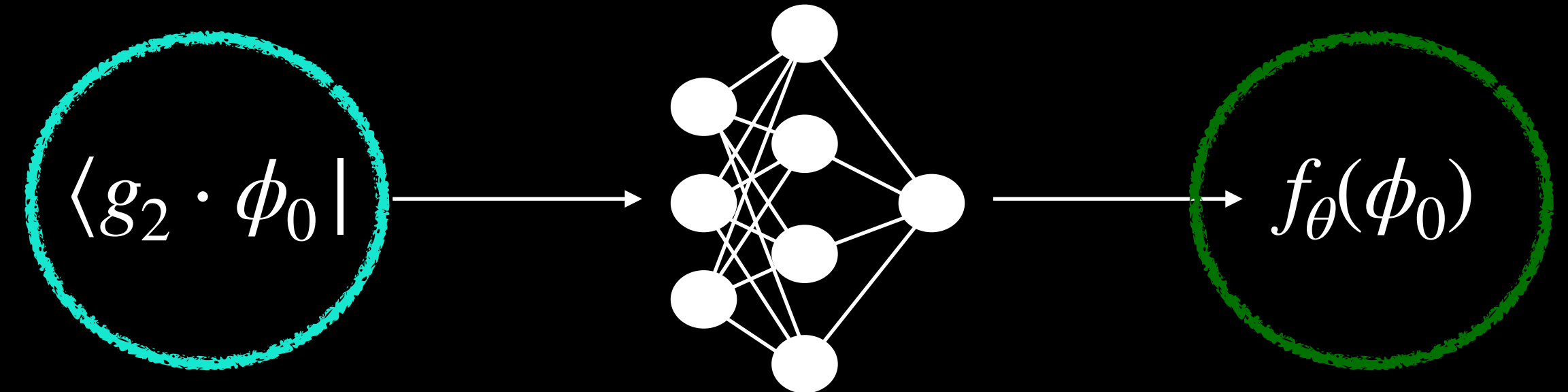
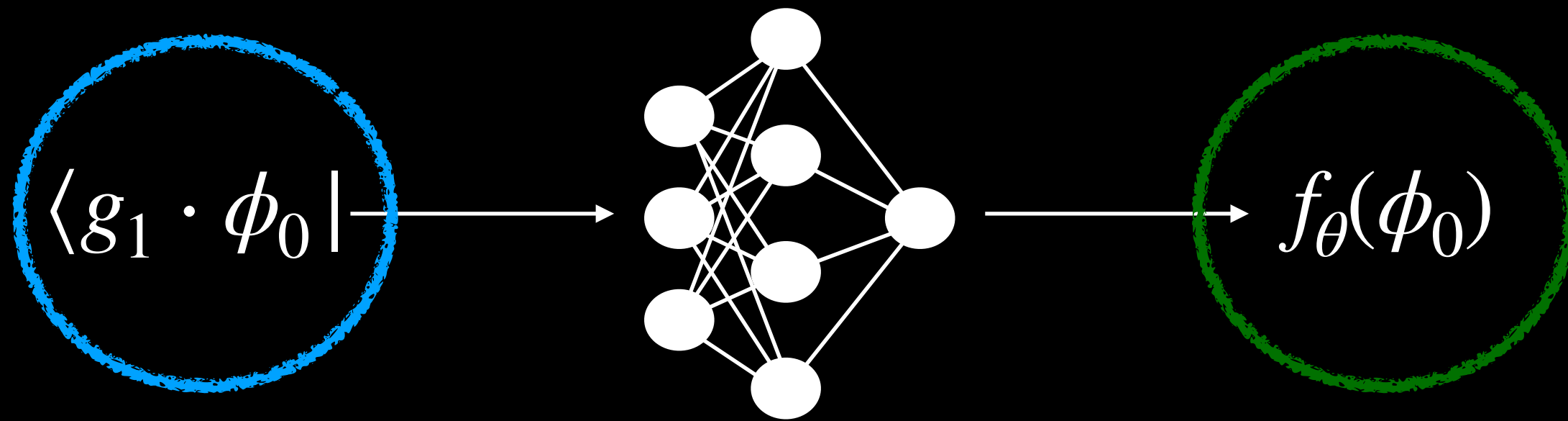
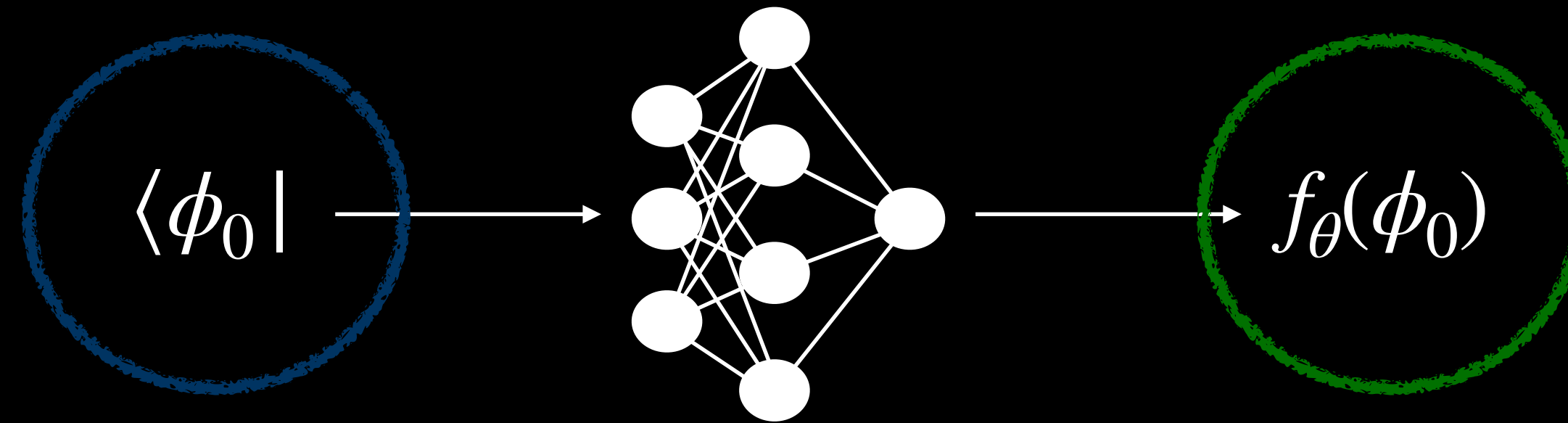
Gauge symmetry



We have to develop a solution that understand this



“Configuration in, amplitude out”

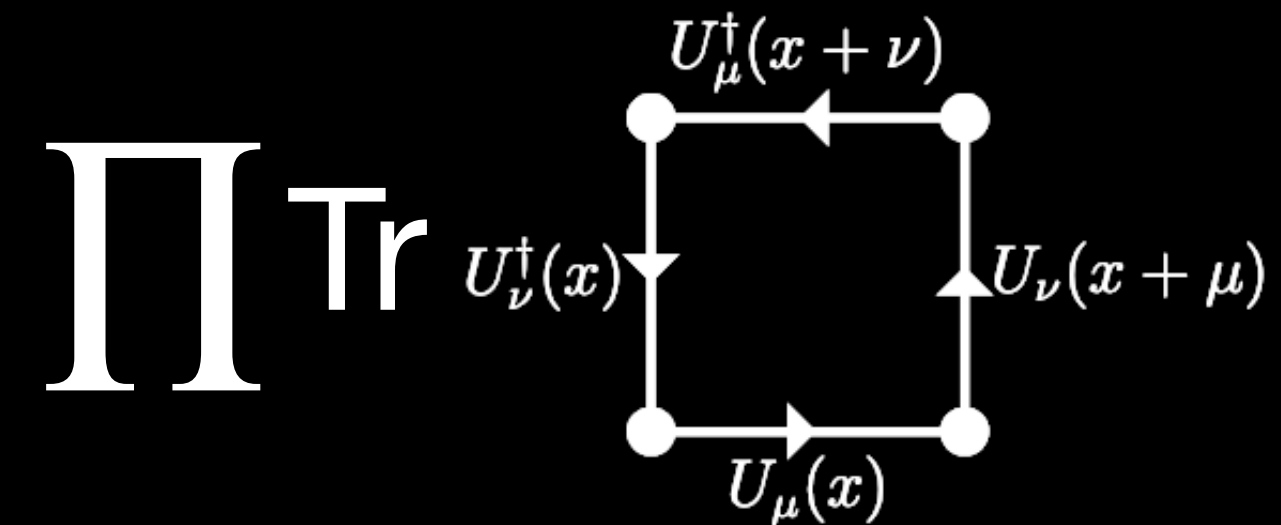


Gauge invariant ansatze

Simple SU(2) gauge invariant ansatz

- Trace of the plaquette is gauge invariant
 - Build our ansatz from this

$$(1) \quad f(\mathbf{U}) = \prod_p e^{\alpha \frac{1}{2} \text{Tr}(P_p)}$$

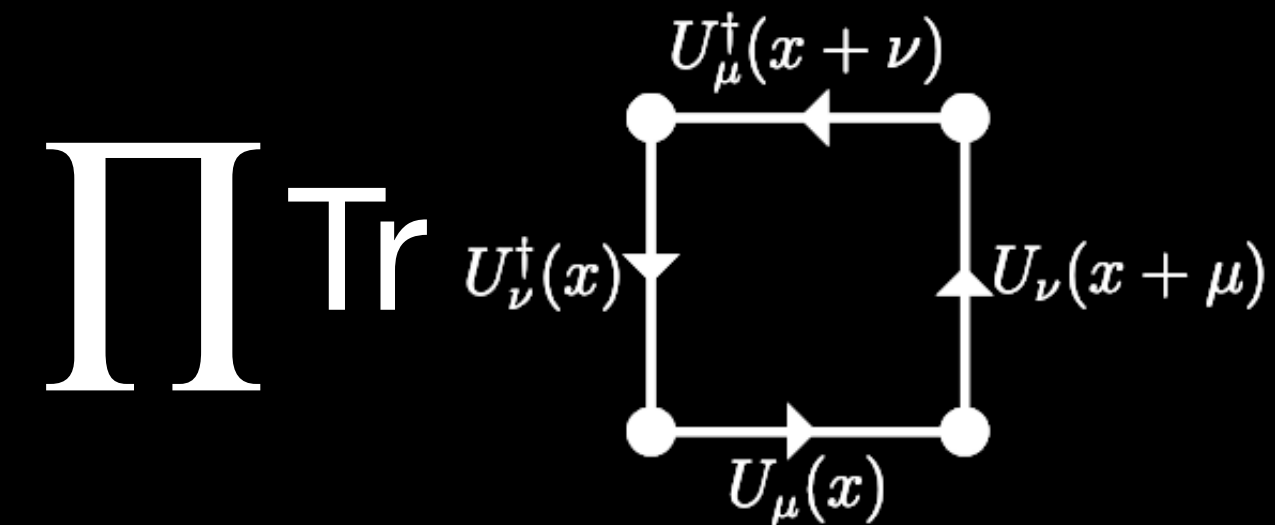


$$(2) \quad f(\mathbf{U}) = \prod_p e^{\alpha \frac{1}{2} \text{Tr}(P_p)} + \prod_{p,q} e^{\frac{1}{4} \text{Tr}(P_p) \beta_{p,q} \text{Tr}(P_q)}$$

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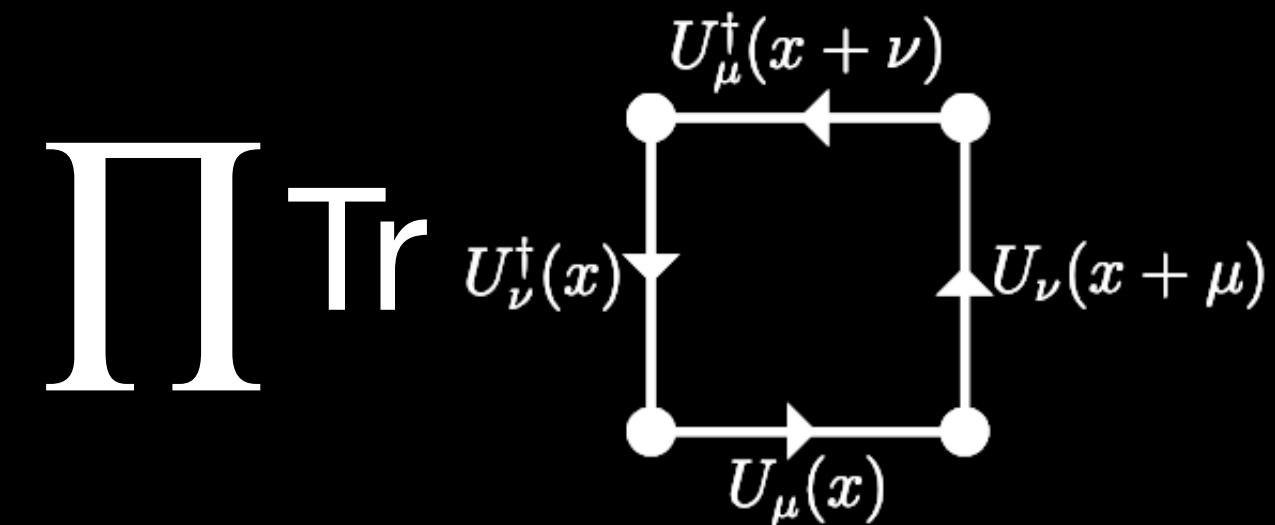
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Only involves local*, scalar quantities

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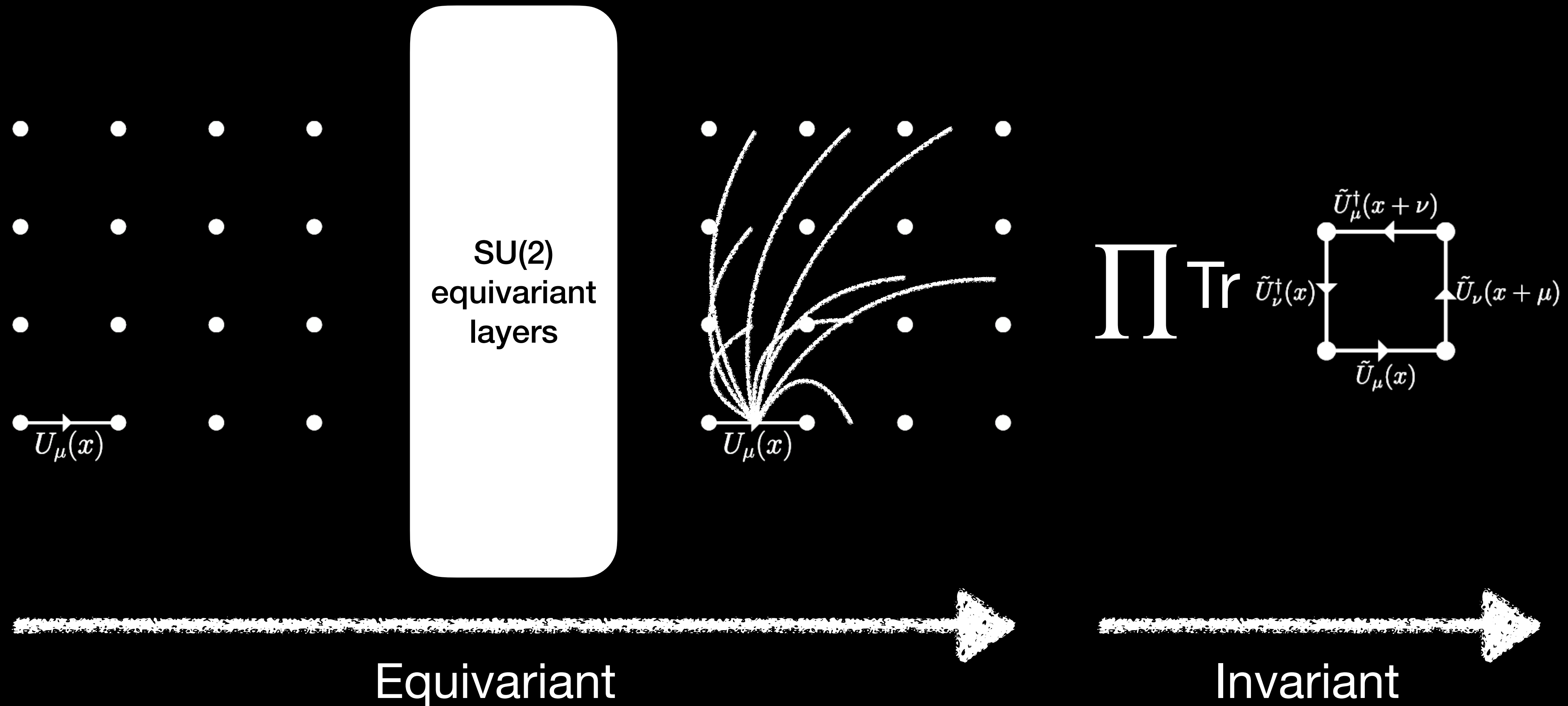
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Cannot even learn clover action

Our SU(2) invariant ansatz

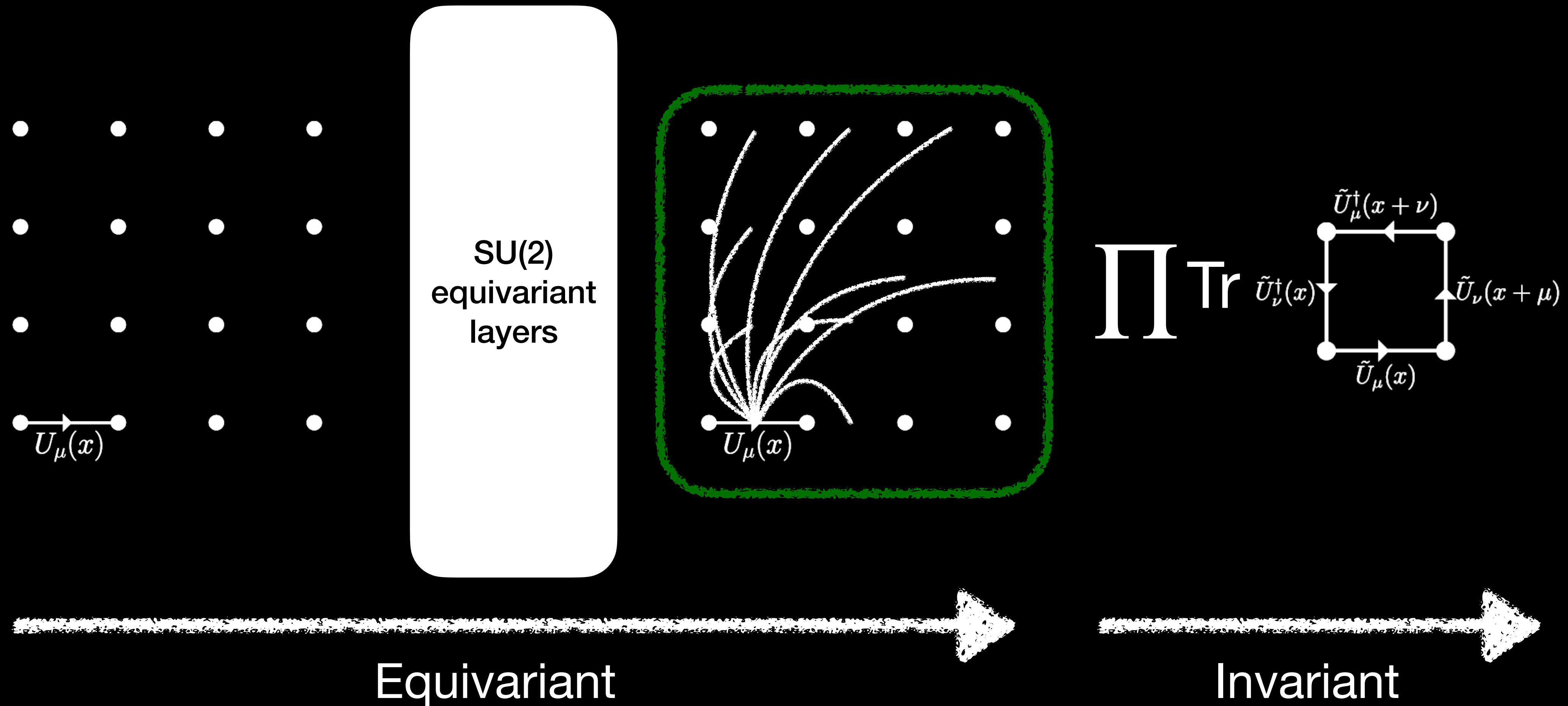


Equivariant layers from:

Kanwar, G. *et al.* Phys. Rev. Lett. **125**, 121601 & Boyda, D. L. *et al.* Phys. Rev. D **103**, 074504

Favoni, M. *et al.* Phys. Rev. Lett. **128**, 032003

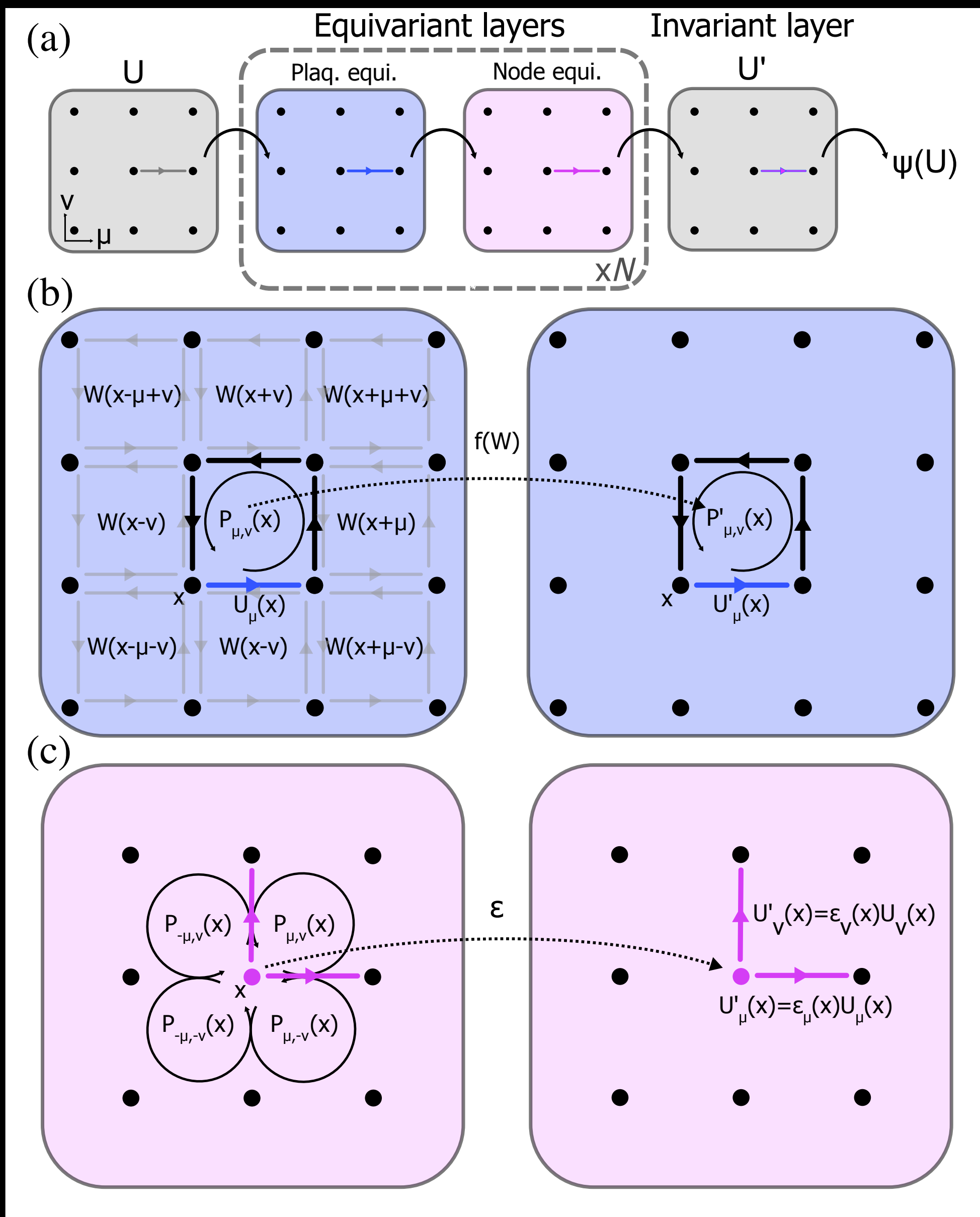
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Recap

- Want to learn the distribution of configurations (i.e. learn the ground state wavefunction), $|\psi_\theta\rangle$
 - Gauge invariant learned distribution
 - Sample from this distribution $\{\mathbf{U}\} \sim |\psi_\theta\rangle$
- Perform measurements on these configurations

Simulations

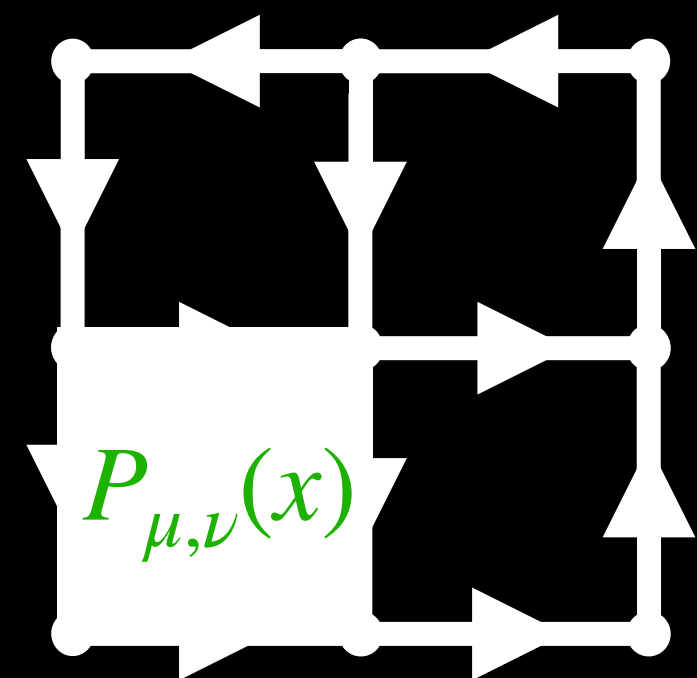
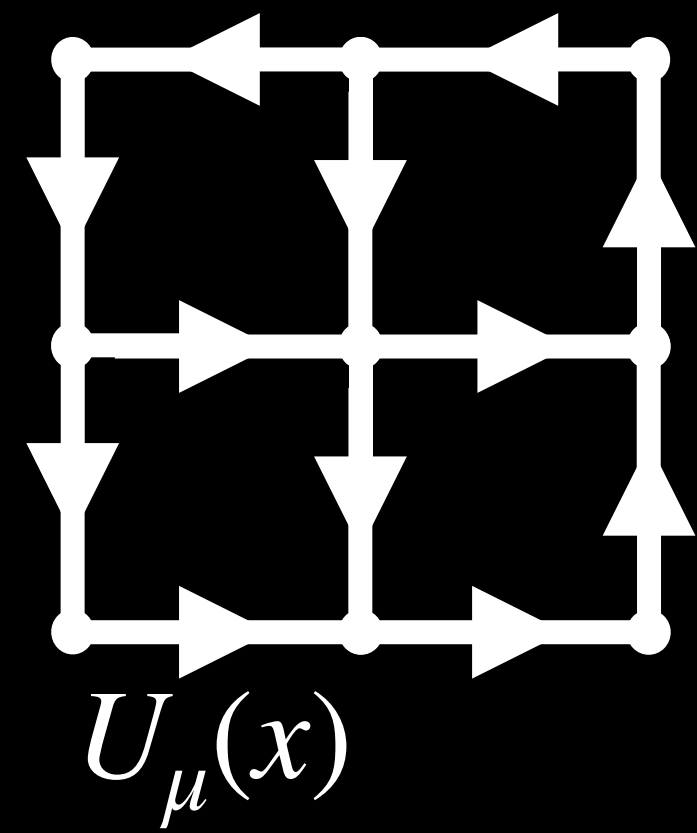
Simulation 1

$$\mathcal{H} = -\frac{1}{2} \sum_l \nabla_l^2 + \lambda \sum_p \left(1 - \frac{1}{2} \text{Tr} \left(P_p \right) \right)$$

- Lattice of size 12x12
- Goal: find the lowest ground state energy at various λ s
- Ansätze:
 - simple, invariant-only, ansatz -> **Jastrow**
 - multi-layer equivariant+invariant ansatz -> **Equivariant**
- Relative energy decrease coming from new ansatz:

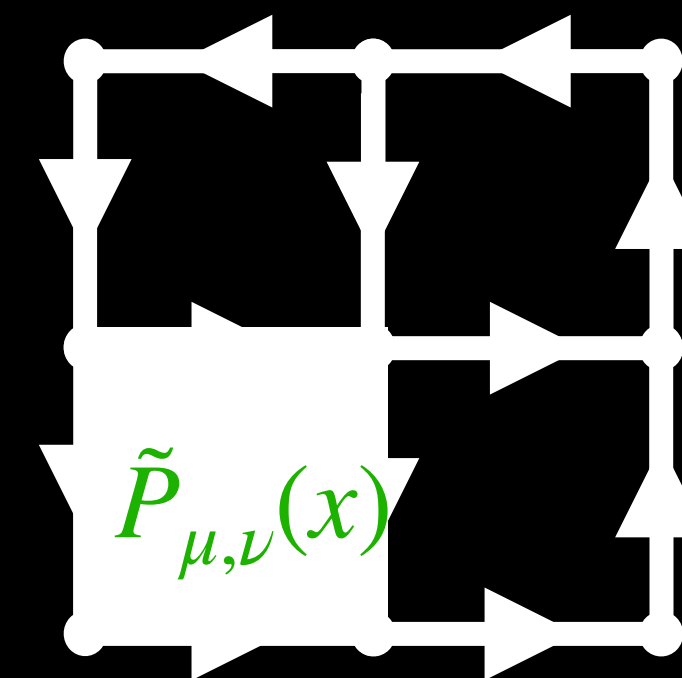
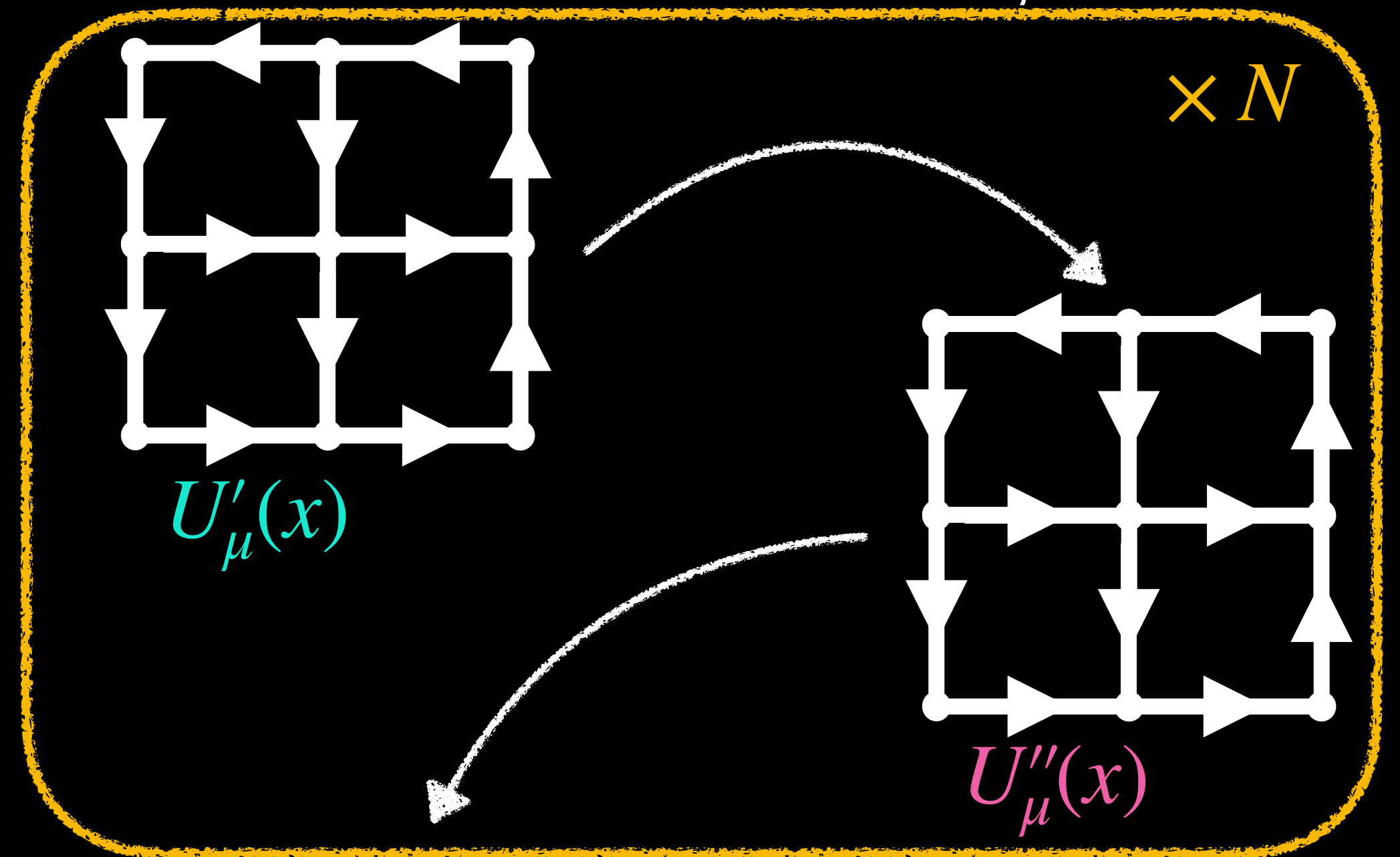
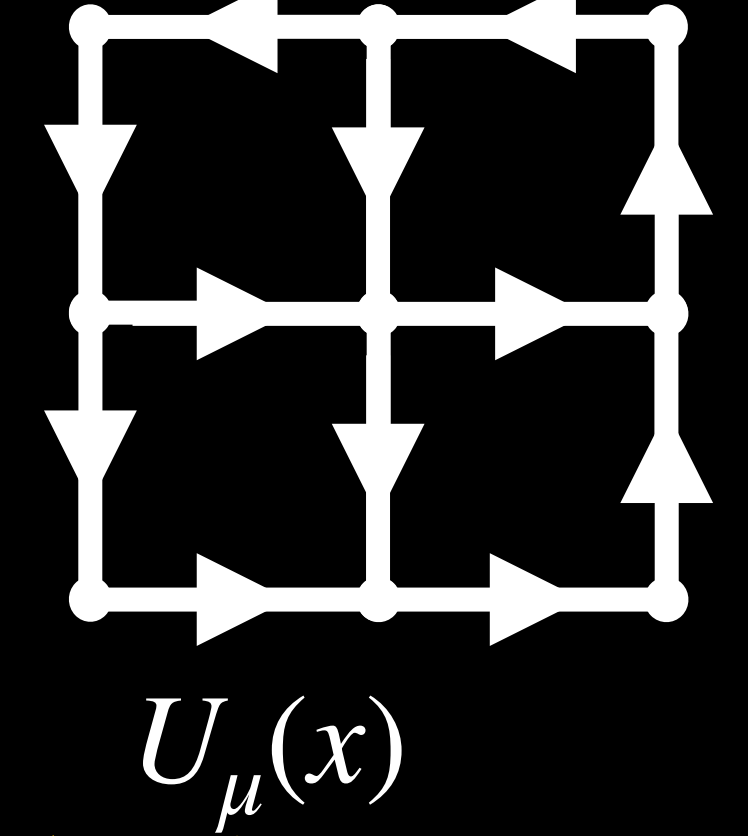
$$\bullet \delta E = \frac{E_{\text{Equivariant}} - E_{\text{Jastrow}}}{E_{\text{Jastrow}}}$$

Jastrow



$$\prod_p e^{\alpha \frac{1}{2} \text{Tr}(P_p)} + \prod_{p,q} e^{\frac{1}{4} \text{Tr}(P_p) \beta_{p,q} \text{Tr}(P_q)}$$

Equivariant

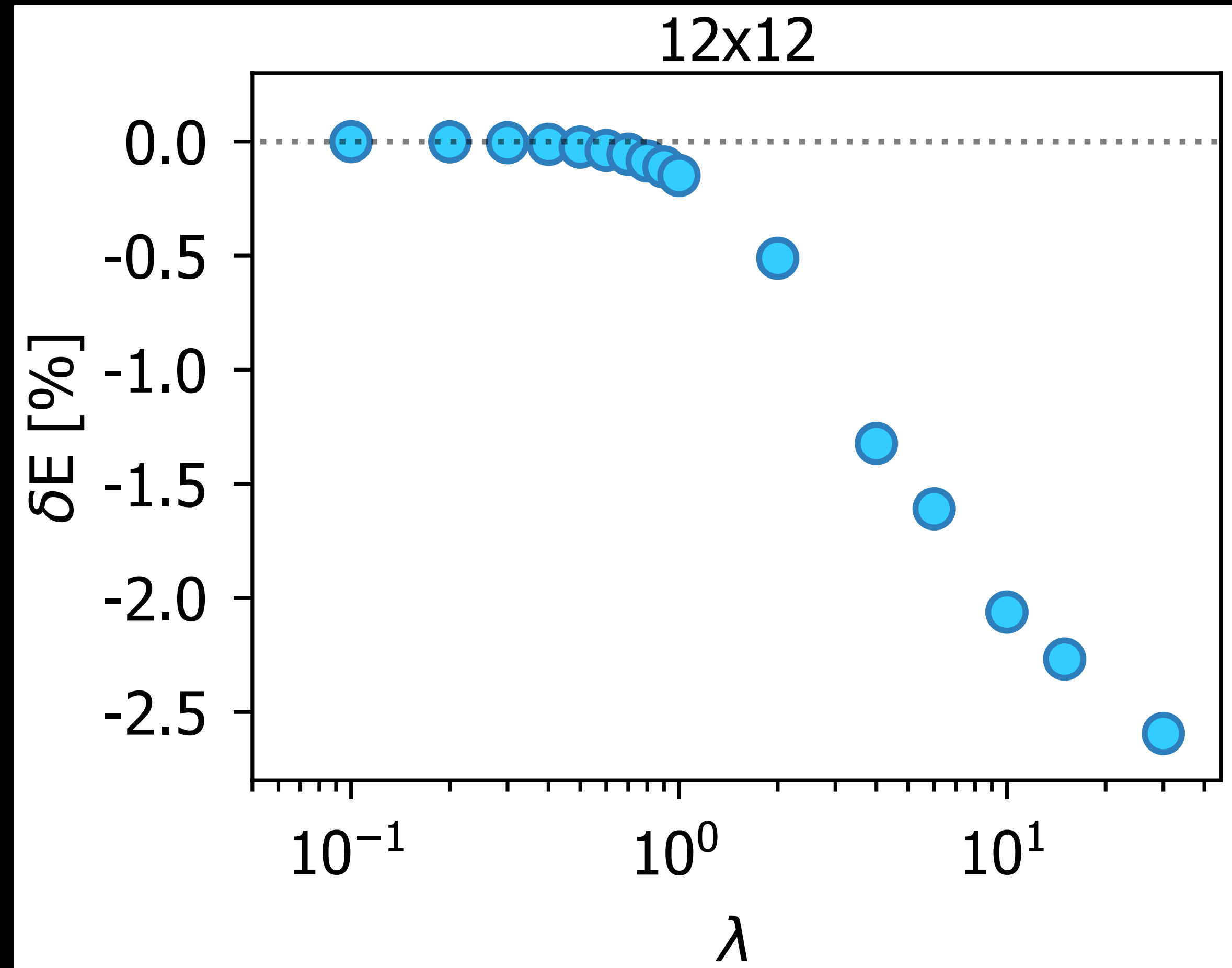


$$\prod_p e^{\alpha \frac{1}{2} \text{Tr}(\tilde{P}_p)} + \prod_{p,q} e^{\frac{1}{4} \text{Tr}(\tilde{P}_p) \beta_{p,q} \text{Tr}(\tilde{P}_q)}$$

Simulation 1

$$\mathcal{H} = -\frac{1}{2} \sum_l \nabla_l^2 + \lambda \sum_p \left(1 - \frac{1}{2} \text{Tr} \left(P_p \right) \right)$$

g range ~ 2.5 - 0.6

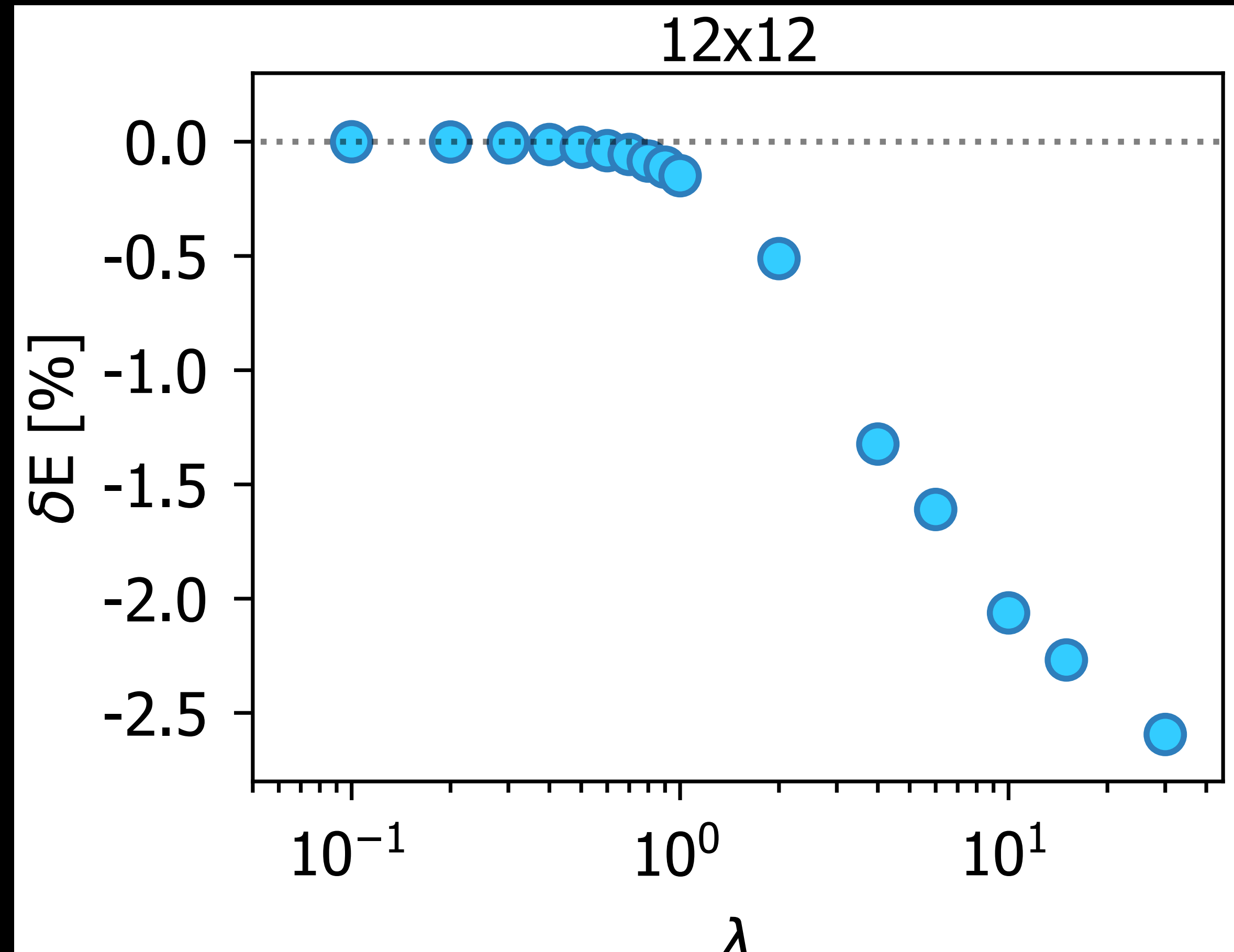


$$\delta E = \frac{E_{\text{Equivariant}} - E_{\text{Jastrow}}}{E_{\text{Jastrow}}}$$

Simulation 1

$$\mathcal{H} = -\frac{1}{2} \sum_l \nabla_l^2 + \lambda \sum_p \left(1 - \frac{1}{2} \text{Tr} \left(P_p \right) \right)$$

g range ~ 2.5 - 0.6

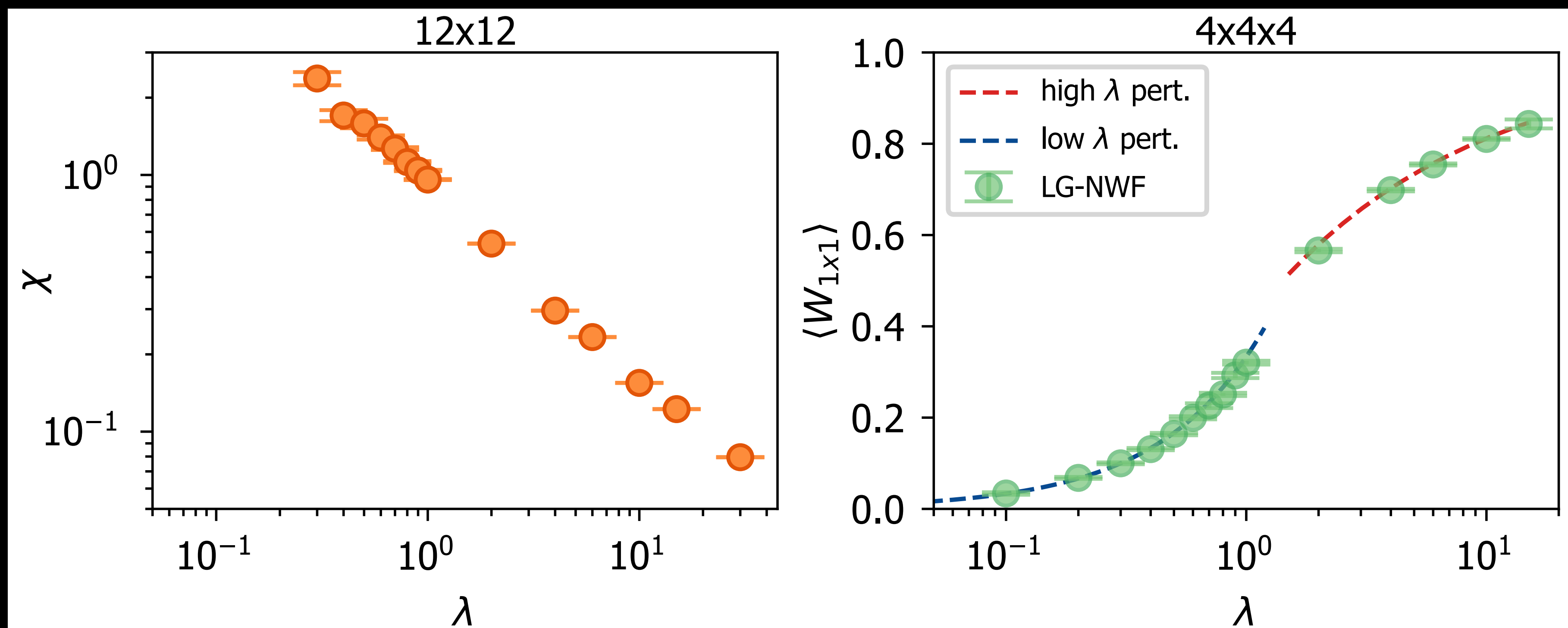


Equivariant layers needed to learn a better representation of the ground state

Simulation 2

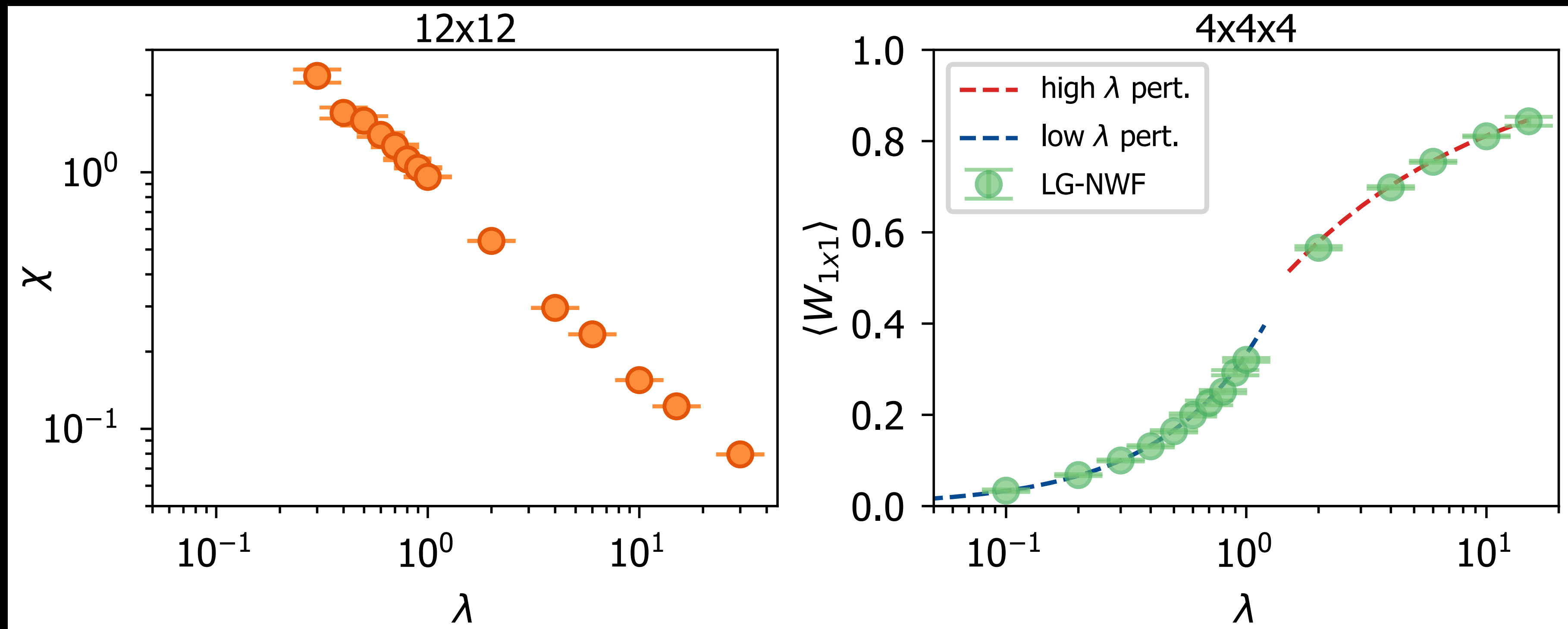
- Lattice of size 12x12 and 4x4x4
- Goal: measure physical observables
- Ansätze:
 - Equivariant
- Average 1x1 Wilson loop $\langle W^{1 \times 1} \rangle$
- Creutz ratio $\chi = -\log \left(\frac{\langle W^{1 \times 1} \rangle \langle W^{2 \times 2} \rangle}{\langle W^{1 \times 2} \rangle \langle W^{2 \times 1} \rangle} \right)$

Simulation 2



Perturbative estimates from Chin, S. A. *et al.* (1985) *Phys Rev D* **31**, 3201

Simulation 2



Perturbative estimates from Chin, S. A. *et al.* (1985) *Phys Rev D* **31**, 3201

Ground state representation physically agrees with theoretical predictions

What now?

LGTs and neural wavefunctions

- What has been done:
 - Abelian theories in 2D
- What we have just done
 - Non-Abelian theories in 2 and 3D
- What can now be done
 - Time evolution^[1] and quantum information entropies^[2]
 - SU(3) without and then with fermions

[1] Carleo, G. Troyer, M. Science 355, Schmitt, M. & Heyl, M., Phys. Rev. Lett. **125** and many more...

[2] TS *et al.* Mach. Learn. Sci. Tech. 6 015042, Sinibaldi, A. *et al.* arXiv:2502.09725



Eliska Greplova



Juan Carrasquilla



Jannes Nys

Thank you for listening

Backup slides

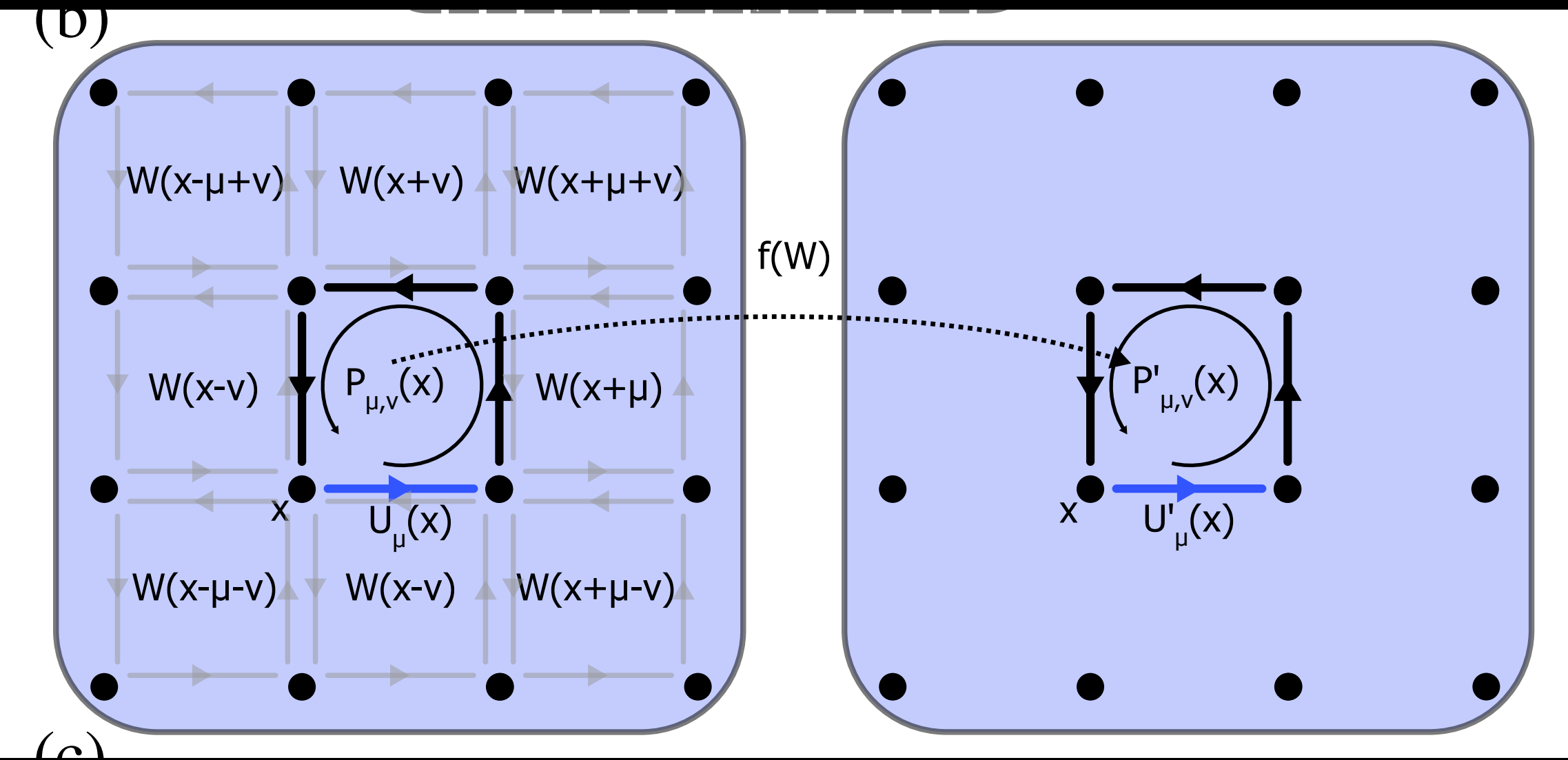
Plaquette equivariant layer

$$U'_\mu(x) = P'_{\mu,\nu}(x) * (U_\nu(x + \mu)U_\mu^\dagger(x + \nu)U_\nu^\dagger(x))^{-1}$$

$$U_\mu(x) \rightarrow P_{\mu,\nu}(x) \rightarrow \lambda_P \rightarrow \lambda'_P \rightarrow P'_{\mu,\nu}(x) \rightarrow U'_\mu(x)$$

CNN

Eigenvalue decomposition



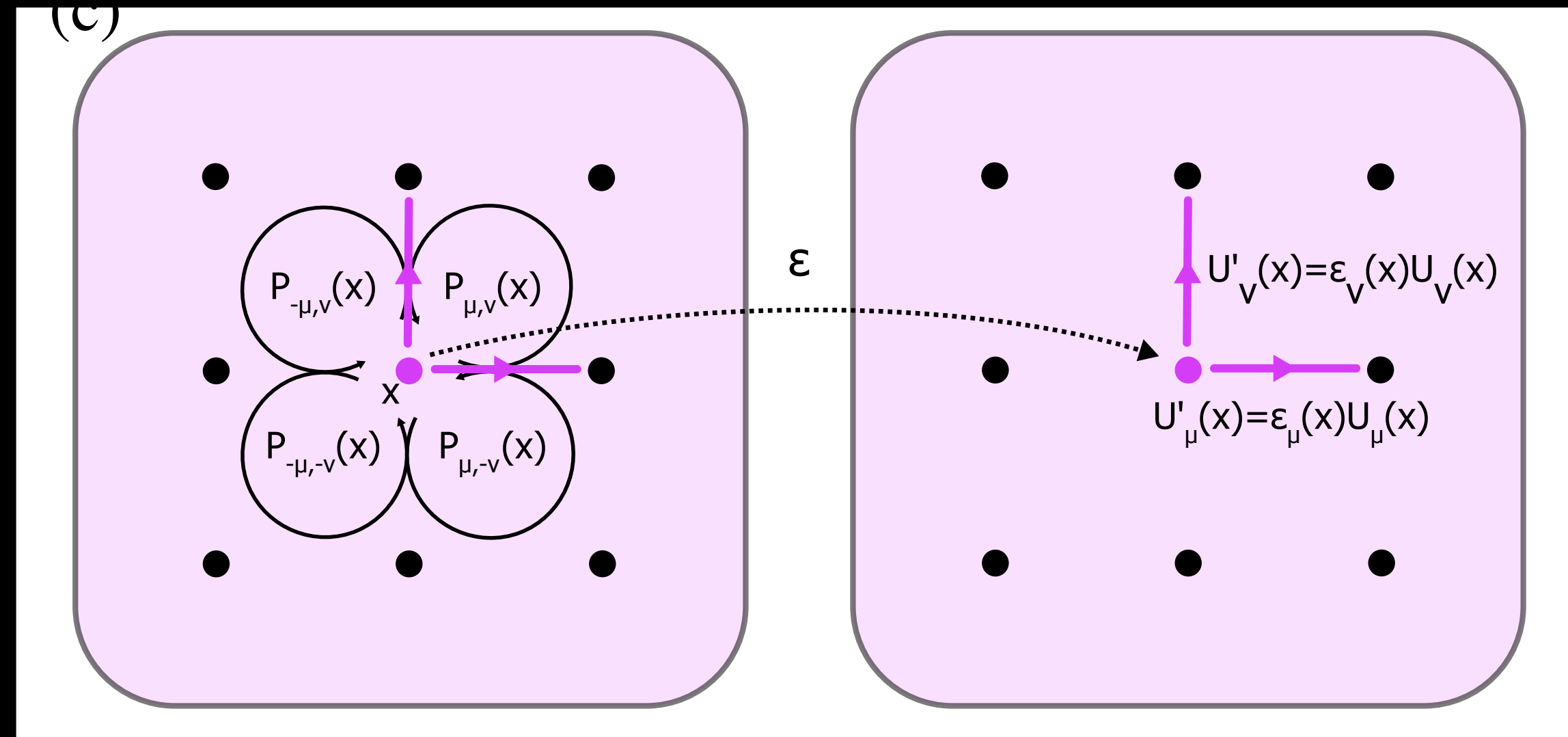
Eigenvalue recomposition

$$P_{\mu,\nu}(x) = U_\mu(x)U_\nu(x + \mu)U_\mu^\dagger(x + \nu)U_\nu^\dagger(x)$$

Node equivariant layer

$$U_\mu(x) \rightarrow C_{\mu,\nu}^i(x) \rightarrow \epsilon_{\mu,\nu}(x) \rightarrow \epsilon_{\mu,\nu}(x)U_\mu(x)$$

$$C_{\mu,\nu}^i(x) = \{P_{\mu,\nu}(x), P_{\mu,-\nu}(x), P_{-\mu,\nu}(x), P_{-\mu,-\nu}(x)\}^i$$



$$\epsilon_\mu(x) = e^{i \sum_i \gamma_{\mu,i} [C_{\mu,\nu}^i(x)]_{aH}}$$

Gauge symmetry

Our new SU(2) invariant ansatz



Equivariant

$$U_\mu(x)$$

$$f(U_\mu(x))$$

$$g(f(U_\mu(x)))$$



Invariant

$$h(g(f(U_\mu(x))))$$

Gauge symmetry

Our new SU(2) invariant ansatz



Equivariant

$$U_\mu(x) \quad f(U_\mu(x)) \quad g(f(U_\mu(x)))$$

$$\Omega(x)U_\mu(x) \quad f(\Omega(x)U_\mu(x)) \quad g(f(\Omega(x)U_\mu(x)))$$



Invariant

$$h(g(f(U_\mu(x))))$$

$$h(g(f(\Omega(x)U_\mu(x))))$$

Gauge symmetry

Our new SU(2) invariant ansatz



Equivariant

$$U_\mu(x) \quad f(U_\mu(x)) \quad g(f(U_\mu(x)))$$

$$\Omega(x)U_\mu(x) \quad f(\Omega(x)U_\mu(x)) \quad g(f(\Omega(x)U_\mu(x)))$$

Invariant

$$h(g(f(U_\mu(x))))$$

$$h(g(f(\Omega(x)U_\mu(x))))$$

$$h(\Omega(x)g(f(U_\mu(x))))$$
