



Quantum Computation for the Understanding of Mass

Simulating Quantum Field Theories

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Contents

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

- 1 Introduction
 - Mass generation
- 2 Quantum computing formulation of the NJL model
 - Fermion-qubit mapping
 - Space parametrization
 - State preparation
- 3 Algorithmic solution
 - Ground state energy



Introduction

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

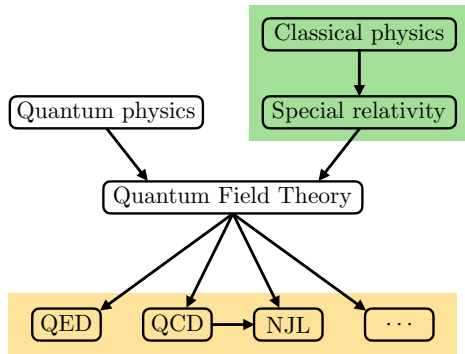
State preparation

Algorithmic solution

Ground state energy



- Quantum Chromodynamics (QCD) is the theory of the strong nuclear force, and it holds many mysteries such as **mass generation**.
- QCD is currently studied using brute-force numerics on the world's largest supercomputers, nonetheless many of its aspects cannot be reproduced by classical means.
- The **NJL model** is an effective field theory regarded as a low-energy approximation to QCD. It retains certain key features like the so called Goldstone modes, and **dynamical chiral symmetry breaking**; and can also be solved nonperturbatively for verification.



Mass generation

Introduction

Mass generation

Quantum computing formulation of the NJL model

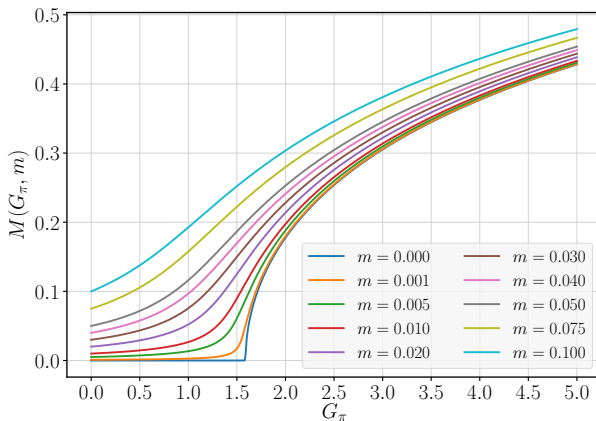
Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



N_{Dirac}	N_{color}	N_{flavor}	Λ_{IR}	Λ_{UV}
$1 + 1 \rightarrow 2$	1	1	0.240 GeV	0.645 GeV

Lattice formulation I

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

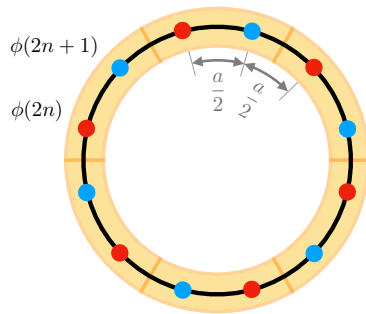
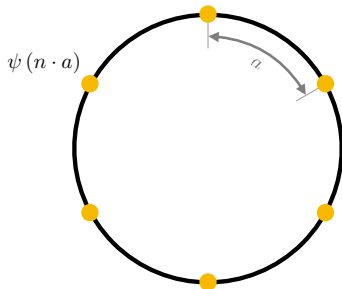
State preparation

Algorithmic solution

Ground state energy



First of all, to discretize our system, we will implement **staggered fermion lattices**; which use two computational lattice sites for each theoretical value of ψ . The resulting newly defined operators obey the **canonical commutation relations for fermions**.



Lattice formulation II

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



All components of the Hamiltonian are written in terms of **Dirac bilinears**:

$$H_N = H_N^{(M)} + H_N^{(K)} + H_N^{(I)}$$

$$H_N^{(M)} = m \sum_{n=0}^{2N-1} (-1)^n \phi^\dagger(n) \phi(n)$$

$$H_N^{(K)} = \frac{i}{a} \sum_{n=0}^{2N-1} \left[\phi^\dagger(n) \phi(n+1) - \phi^\dagger(n+1) \phi(n) \right]$$

$$H_N^{(I)} = -\frac{1}{2a} G_\pi \sum_{n=0}^{N-1} \left[\phi^\dagger(2n) \phi(2n) - \phi^\dagger(2n+1) \phi(2n+1) \right]^2$$

Fermion-qubit mapping

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



Generally speaking, quantum computers cannot measure any given operator directly. Therefore, in order to simulate any Hamiltonian in a quantum processor, one needs to efficiently map its component operators onto ones suitable for evaluation in such machines (e.g. **Pauli operators** and the identity).

In one spatial dimension, spin- $\frac{1}{2}$ particles (i.e. qubits) behave much like fermions. The **Jordan-Wigner transform** associates spin *down/up* with *occupied/unoccupied* fermion states:

$$|\uparrow\rangle \cong |0\rangle, \quad |\downarrow\rangle \cong |1\rangle$$

$$|\downarrow\rangle \cong \phi^\dagger |0\rangle, \quad |\uparrow\rangle \cong \phi |1\rangle$$

$$S(n)\phi(n) \rightarrow \sigma^+(n), \quad \phi^\dagger(n)S^\dagger(n) \rightarrow \sigma^-(n)$$

Particularly, choosing a gauge which makes the **string operator** $S(n)$ hermitian $\forall n$:

$$\phi(n) \rightarrow \left[\prod_{l < n} \sigma^3(l) \right] \sigma^+(n), \quad \phi^\dagger(n) \rightarrow \left[\prod_{l < n} \sigma^3(l) \right] \sigma^-(n)$$

Refactoring the NJL Hamiltonian

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

$$H_N^{(M)} \rightarrow \frac{m}{2} \sum_{n=0}^{2N-1} (-1)^{n+1} \sigma^3(n)$$

$$H_N^{(K)} \rightarrow \frac{i}{a} \sum_{n=0}^{2N-1} [\sigma^-(n) \sigma^+(n+1) - \sigma^-(n+1) \sigma^+(n)]$$

$$H_N^{(I)} \rightarrow \frac{G_\pi}{4a} \sum_{n=0}^{N-1} [\sigma^3(2n) \sigma^3(2n+1) - N]$$

With periodic boundary conditions $\sigma^p(N) = \sigma^p(0)$, and dropping the adiabatic modification term $\frac{G_\pi N}{4a}$, this Hamiltonian will adopt the following form in the **Chiral limit** (i.e. $m = 0$):

$$P_N \triangleq 2aH_N = \sum_{n=0}^{2N-1} [X_{n+1} Y_n - Y_{n+1} X_n] + \frac{G_\pi}{2} \sum_{n=0}^{N-1} Z_{2n+1} Z_{2n}$$

The number of terms in this operator **grows polynomially** with the size of the system N .



Space parametrization

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

Once we have ways of measuring our Hamiltonian, we need to be able to explore different quantum states. This can be achieved by parametrizing the Hilbert/Fock space of states representing the system. To do this efficiently, we will analyze the two distinct parts in our Hamiltonian independently; since these will dominate in two **different regimes**:

INFINITELY STRONG INTERACTIONS

Interaction term dominates (i.e. $G_\pi \rightarrow \infty$)

$$G_N \triangleq \sum_{n=0}^{N-1} Z_{2n+1} Z_{2n}$$

INFINITELY WEAK INTERACTIONS

Kinetic term dominates (i.e. $G_\pi \rightarrow 0$)

$$K_N \triangleq \sum_{n=0}^{2N-1} [X_{n+1} Y_n - Y_{n+1} X_n]$$

From the symmetries of these two terms for the case $N = 2$, we can extract the following **symmetry-based parametrization ansatz (SBP)**:

$$|\text{SBP}_2(\theta, \eta)\rangle \triangleq \sin(\theta) \sin(\eta) \left| \gamma_{\max}^2 \right\rangle - \sin(\theta) \cos(\eta) \left| \gamma_{\min,1}^2 \right\rangle + i \cos(\theta) \left| \gamma_{\min,2}^2 \right\rangle$$



State preparation I

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



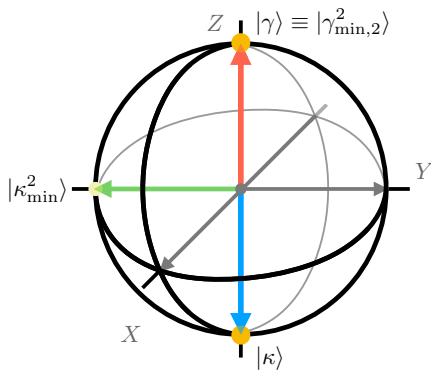
In order to implement this parametrization on any of the IBM-Q quantum computers, we need to be able to write it down as a quantum circuit in Qiskit:

$$|\text{SBP}_2(\theta, \eta)\rangle = U(\theta, \eta) |\text{SR}\rangle$$

For positive values of the coupling constant (i.e. we will only need the minimum eigenstates) we can simplify even further the parametrization:

$$|\gamma\rangle \equiv |\gamma_{\min,2}^2\rangle \triangleq \frac{|5\rangle - |10\rangle}{\sqrt{2}} \equiv |\text{SBP}_2(0, \frac{\pi}{4})\rangle$$

$$|\kappa\rangle \triangleq \frac{|3\rangle - |6\rangle - |9\rangle + |12\rangle}{2} \equiv |\text{SBP}_2(\frac{\pi}{2}, \frac{\pi}{4})\rangle$$



State preparation II

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

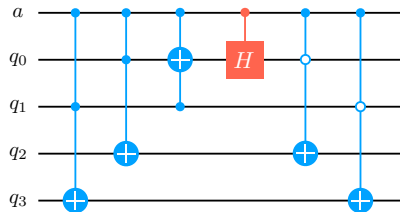
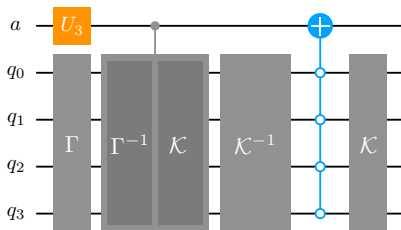


Figure: (Left) Quantum circuit to map the ancilla qubit onto the target qubit Hilbert space in our system. (Right) Simplified controlled $\mathcal{K}\Gamma^{-1}$ gate.



Variational quantum eigensolver algorithm

Introduction

Mass generation

Quantum computing formulation of the NJL model

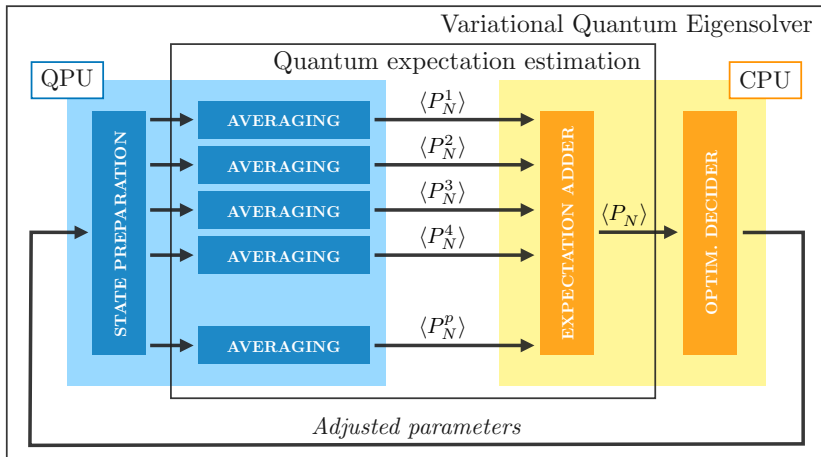
Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



Optimal sampling regression algorithm

Introduction

Mass generation

Quantum computing formulation of the NJL model

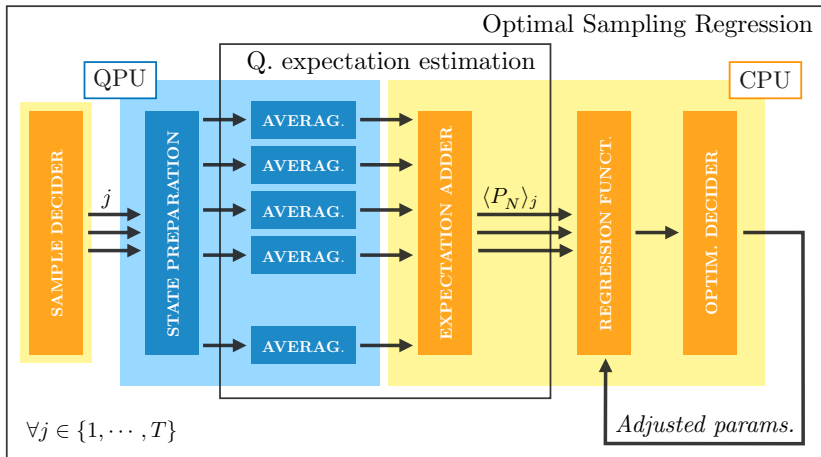
Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



Algorithm comparison

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

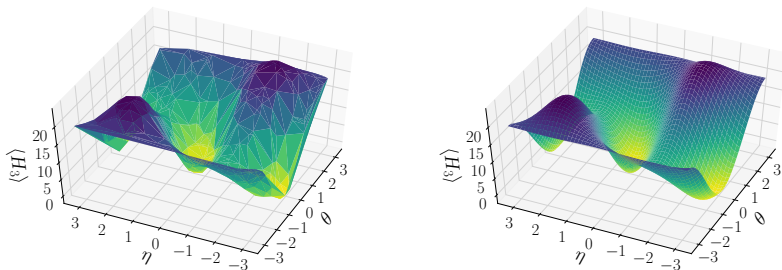


Figure: Comparison between the VQE and OSR algorithms, when reproducing an external model with two parameters. (Left) Triangulation of the expectation value function from raw samples. (Right) Approximate function obtained through the Optimal Sampling Regression method with $S_q = S_{\max} = 2 \forall q$.

N_{params}	VQE samples	OSR samples	VQE error	OSR error
1	24	3	3.5%	1.0%
2	153	25	0.3%	0.2%



Ground state energy I

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

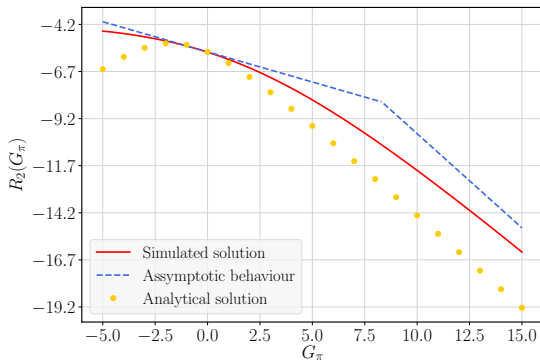
State preparation

Algorithmic solution

Ground state energy



At last, we have everything that we need to solve for the ground state energy of our system using a quantum computer. For simplicity, we will do so first through a **quantum simulator**.



Ground state energy II

Introduction

Mass generation

Quantum computing formulation of the NJL model

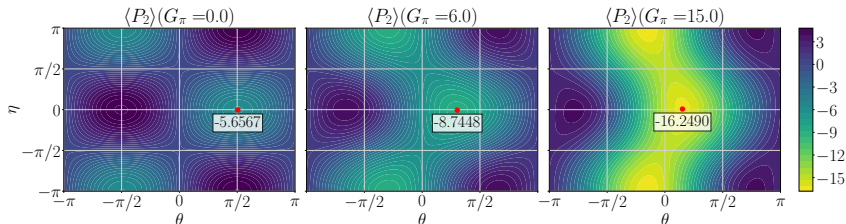
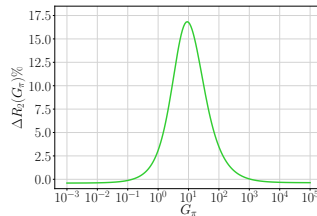
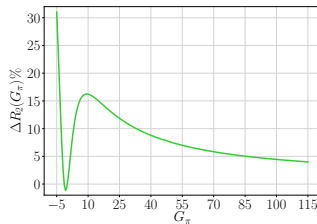
Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy





NJL model and the gap equation

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

NJL LAGRANGIAN DENSITY

Simplest version that reproduces a condensate

$$\mathcal{L}(x) \triangleq \bar{\psi}(x)(i\not{\partial} - m)\psi(x) + \mathcal{L}_I(x)$$

$$\mathcal{L}_I(x) = \frac{1}{2}G_\pi [\bar{\psi}(x)\psi(x)]^2$$

NJL HAMILTONIAN DENSITY

Obtained through the Lengendre transform

$$\mathcal{H}(x) \triangleq \bar{\psi}(x)\left(m - i\gamma^1\partial_1\right)\psi(x) + \mathcal{H}_I(x)$$

$$\mathcal{H}_I(x) = -\frac{1}{2}G_\pi [\bar{\psi}(x)\psi(x)]^2$$

The **bare and dressed masses** appear on the bare quark propagator S_0 , and the NJL dressed quark propagator S respectively. We can find a relationship between these two by solving the **gap equation**:

$$S^{-1} = S_0^{-1} - 2iG_\pi \int \frac{d^2p}{(2\pi)^2} N_{\text{color}} N_{\text{flavor}} \text{Tr}_D[S]$$

$$M \simeq m + 4iG_\pi N_{\text{color}} N_{\text{flavor}} \int \frac{d^2p}{(2\pi)^2} \frac{M}{p^2 - M^2}$$



NJL Hamiltonian in 1 + 1 dimensions

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



We can define the Hamiltonian of the system as the integral over space of the Hamiltonian density:

$$H = \int \mathcal{H}(x) dx = \int \left\{ \bar{\psi}(x) (m - i\gamma^1 \partial_1) \psi(x) - \frac{1}{2} G_\pi [\bar{\psi}(x) \psi(x)]^2 \right\} dx$$

For a basis where:

$$\psi = \begin{bmatrix} \psi_+ \\ \psi_- \end{bmatrix}, \quad \bar{\psi} \triangleq \psi^\dagger \gamma^0, \quad \gamma^0 = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \quad \gamma^1 = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

We can write the kinetic term as:

$$\bar{\psi} (-i\gamma^1 \partial_1) \psi = \frac{i}{2} \left\{ \left[\psi_+^\dagger (\partial_1 \psi_-) - (\partial_1 \psi_+^\dagger) \psi_- \right] + \left[\psi_-^\dagger (\partial_1 \psi_+) - (\partial_1 \psi_-^\dagger) \psi_+ \right] \right\}$$

Dirac equation from staggered fermion lattice

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



From the kinetic term of the discretized Hamiltonian, we can now recover the **massless Dirac equation** in the continuum limit; which serves as proof of correctness:

$$\dot{\phi}(n) = i \left[H_N^{(K)}, \phi(n) \right]_- = \frac{\phi(n+1) - \phi(n-1)}{a}$$

In terms of the original fields, this is:

$$\dot{\psi}_+ = \frac{\Delta\psi_-}{\Delta x}, \quad \dot{\psi}_- = \frac{\Delta\psi_+}{\Delta x}$$

Lastly, taking the limit when $a \rightarrow 0$:

$$\frac{\partial}{\partial t} \psi = \hat{\alpha}_1 \frac{\partial}{\partial x} \psi$$
$$\hat{\alpha}_1 \triangleq \gamma_0 \gamma_1 = \gamma^0 \gamma_1 = -\gamma^0 \gamma^1 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

Space parametrization II

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



From the symmetries of these two terms for the case $N = 2$, we can extract the following **symmetry-based parametrization ansatz (SBP)**:

$$|\text{SBP}_2(\theta, \eta)\rangle \triangleq \sin(\theta) \sin(\eta) |\gamma_{\max}^2\rangle - \sin(\theta) \cos(\eta) |\gamma_{\min,1}^2\rangle + i \cos(\theta) |\gamma_{\min,2}^2\rangle$$

$$|\gamma_{\max}^2\rangle \triangleq \frac{|3\rangle + |12\rangle}{\sqrt{2}}, \quad |\gamma_{\min,1}^2\rangle \triangleq \frac{|6\rangle + |9\rangle}{\sqrt{2}}, \quad |\gamma_{\min,2}^2\rangle \triangleq \frac{|5\rangle - |10\rangle}{\sqrt{2}}$$

As a matter of fact, this state can indeed evaluate to the minimum and maximum eigenstates of the operator:

$$\begin{aligned} |\kappa_{\max}^2\rangle &\equiv \left| \text{SBP}_2\left(\frac{3\pi}{4}, \frac{\pi}{4}\right) \right\rangle, & |\kappa_{\min}^2\rangle &\equiv \left| \text{SBP}_2\left(\frac{\pi}{4}, \frac{\pi}{4}\right) \right\rangle \\ |\gamma_{\max}^2\rangle &\equiv \left| \text{SBP}_2\left(\frac{\pi}{2}, \frac{\pi}{2}\right) \right\rangle, & |\gamma_{\min,1}^2\rangle &\equiv \left| \text{SBP}_2\left(\frac{\pi}{2}, 0\right) \right\rangle, & |\gamma_{\min,2}^2\rangle &\equiv \left| \text{SBP}_2(0, 0) \right\rangle \end{aligned}$$

State preparation low level circuits I

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

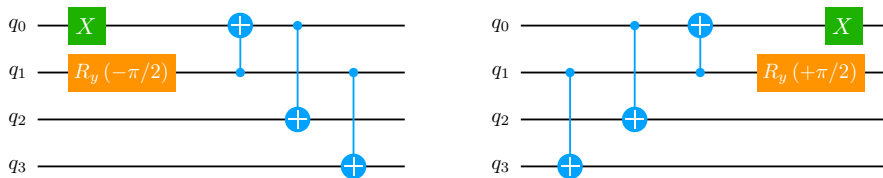


Figure: (Left) Preparation Γ of state $|\gamma\rangle$. (Right) Quantum gate Γ^{-1} for reversing state $|\gamma\rangle$.



State preparation low level circuits II

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

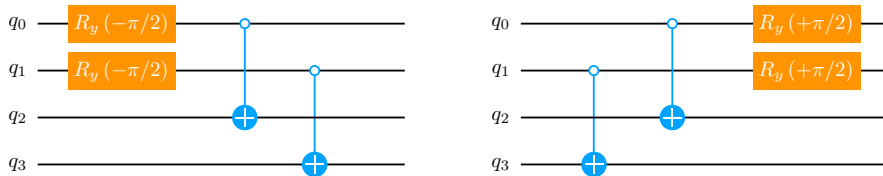


Figure: (Left) Preparation \mathcal{K} of state $|\kappa\rangle$. (Right) Quantum gate \mathcal{K}^{-1} for reversing state $|\kappa\rangle$.



Optimal sampling regression algorithm I

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy

The method that we have used to parametrize space will naturally return cycles in the the states that we are parametrizing. Such **periodic nature** will transfer to the expectation value function, which in turn allows us to consistently apply Fourier analysis to fully describe it:

$$f(\theta) \equiv a_0 + \sum_{s=1}^S [a_s \cos(s\theta) + b_s \sin(s\theta)]$$

$$\begin{bmatrix} 1 & \cos(\theta_1) & \sin(\theta_1) & \cos(2\theta_1) & \cdots & \sin(S\theta_1) \\ 1 & \cos(\theta_2) & \sin(\theta_2) & \cos(2\theta_2) & \cdots & \sin(S\theta_2) \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \cos(\theta_{2S+1}) & \sin(\theta_{2S+1}) & \cos(2\theta_{2S+1}) & \cdots & \sin(S\theta_{2S+1}) \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ b_1 \\ a_2 \\ \vdots \\ b_S \end{bmatrix} = \begin{bmatrix} f(\theta_1) \\ f(\theta_2) \\ \vdots \\ f(\theta_{2S+1}) \end{bmatrix}$$

$$Fc = f \rightarrow F^\dagger Fc = F^\dagger f$$



Optimal sampling regression algorithm II

Introduction

Mass generation

Quantum computing formulation of the NJL model

Fermion-qubit mapping

Space parametrization

State preparation

Algorithmic solution

Ground state energy



Generally $S \rightarrow \infty$, however, if the bandwidth is bounded, S will be finite and it will be possible to evaluate this expression exactly. Theoretically, the power of this method is demonstrated through the **Nyquist-Shannon sampling theorem**; which states that if a function $f(\theta)$ contains no angular frequencies higher than ω_S , it is completely determined by giving its ordinates at a series of points $1/2\omega_S$ apart:

$$\omega_{\text{sampling}} > 2\omega_S$$

Extending these results to **higher dimensions** is straight forward considering multidimensional Fourier series. In this case, we may have a different bandwidth S_q for each parameter. Calling the total number of parameters Q , and the maximum bandwidth S_{\max} , the total number of samples T required by this method is:

$$T = \prod_{q=1}^Q (2S_q + 1) = \mathcal{O}(S_{\max}^Q)$$