Flow equations on quantum circuits

Unified variational methods for quantum many-body systems

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Mathematical Sciences

An optimal control approach to the **Plow equations on quantum circuits** *Classical simulation of* Unified variational methods for quantum many-body systems

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Overview

- 1. Introduction the quantum many-body problem (sort of)
- 2. Approximations and variational methods
 - Variational ansatzes as quantum circuits
 - Examples
- 3. Flow equations
 - History
 - Applications to quantum-many body problems
- 4. Flow equations on quantum circuits
 - Universal variational method
 - Optimal gate generators
 - Example: application to 1D Heisenberg model
- 5. Conclusions and TODO

The quantum many-body problem

- Describe the ground and low-energy eigenstates of a system of *N* interacting particles
- Generic model: particles arranged on a lattice with nearestneighbour interactions.



$$H = \sum_{\langle j,k \rangle} K_{jk} + \sum_{j} H_{j}$$

"two-local" Hamiltonian

"Describing" quantum systems

- *What* do we mean by a description of these low-energy states?
 - Quantitative: energy spectrum and gap, local observables, correlation functions.
 - Insight: emergence of unusual excitations or quasiparticles and their statistical properties.
- *How* do we provide these descriptions?
 - Hardly anything is exactly soluble.
 - The ground state of a system of 30 spin-½ particles (qubits) expanded in the usual basis requires 16 GB worth of coefficients.

Approximations and variational methods

- "Hilbert space is a big place" for a locally interacting system with lots of symmetry only a small part should be relevant.
- In a variational method we posit a relevant subclass of states, and then try to optimize approximations within that subclass.
- e.g.
 - Mean Field Theory $|\psi\rangle = \bigotimes_{k=1}^{N} |\psi_k\rangle$
 - Density Matrix Renormalization

$$|\psi\rangle = \sum_{k_1,\dots,k_N} \operatorname{Tr}\left(A_1^{k_1}\cdots A_N^{k_N}\right)|k_1,\dots,k_N\rangle$$

– General Tensor Networks

Variational classes as quantum circuits

- A *quantum circuit class* specifies the location of gates, any ancillary systems, and refinement parameters.
- All existing ansatz states have equivalent descriptions as quantum circuits.
 - e.g. "Staircase" circuit, corresponding to matrix product states



[Fannes, Nachtergaele, Werner 1989; Osterlund and Roemer, 1992]

Variational classes as quantum circuits

e.g. "MERA" circuit [Vidal 2006]



More variational classes

e.g. 1D Quantum cellular automata model (fixed depth)



Desiderata: Efficient local expectations



- 1. Cancel $U_4^{\dagger}U_4 = I$
- 2. Evaluate $U_3^{\dagger}KU_3$ then set $A_2 = (I \otimes \langle 0 |) U_3^{\dagger}KU_3 (I \otimes |0 \rangle)$
- 3. Iterate $A_{k-1} = (I \otimes \langle 0 |) U_k^{\dagger} A_k U_k (I \otimes |0\rangle)$ until we obtain A_0 acting only on the auxiliary system.
- 4. Read out the (1,1) element.

Flow Equations

• Analytic techniques for transforming a Hamiltonian via a continuously parameterized unitary transformation

$$U(t) = \mathcal{T} \exp\left(-i \int_0^t G(s) ds\right)$$
$$U(0)$$

• For
$$H(t) = U(t)^{\dagger} H U(t)$$
 we find

$$\frac{dH}{dt} = -i [G(t), H(t)]$$

• With appropriate choice of generator G(t) we can obtain useful limiting forms $\lim_{t\to\infty} H(t)$

History of Flow Equations

- Introduced independently by lots of people, in particular Brockett, Glazek and Wilson, Wegner.
- e.g. with G = [H(t), N] we have *double-bracket* flow

$$\frac{dH}{dt} = -i\left[\left[H, N\right], H\right],$$

and if *N* is a diagonal matrix with increasing entries, then

$$\lim_{t\to\infty} H(t) = \operatorname{diag}\left(E_0, E_1, \dots, E_n\right).$$

• Can also be used to numerically sort lists, solve linear programming problems [*Brockett 1988*]

Flow Eqns and quantum many-body systems

- Numerical flow techniques are obviously not directly practical for $2^N \times 2^N$ Hamiltonians.
- Used as an approximate analytic method by differentiating in parameters
 - Make a clever choice for the generator G(t)
 - Truncate resulting system of DEs
 - Solve by whatever means necessary.
- Used to effectively diagonalize Hamiltonians, calculate correlation functions, and to take controlled expansions in strong-coupling models. [*Kehrein & Mielke, Wegner*]

Flow Eqns as universal variational method

- <u>Our approach</u>: write variational classes for quantum manybody problems as quantum circuits, and use flow equations as a general purpose optimization method.
- Quantum circuit class specified by M gates $U_j(t)$ so the overall unitary is $U(t) = \prod_{j=1}^{M} U_j(t)$. Aim to minimize the expectation.

$$E(t) = \langle 0 | U(t)^{\dagger} H U(t) | 0 \rangle$$

• <u>Method</u>: Calculate infinitesimal generators $G_j(t)$ individually to minimize the derivative dE/dt.

Finding optimal generators

• This amounts to minimizing the real part of the quantity:

$$-i\sum_{j=1}^{M} \langle 0|U(t)^{\dagger} H\Big(\prod_{k=j+1}^{M} U_k(t)\Big) G_j(t)\Big(\prod_{k=1}^{j} U_k(t)\Big)|0\rangle$$

• After various rearrangements, it can be shown that the optimal generator is given by $G_j = -2(F_j + F_j^{\dagger})$ with

$$F_{j} = \operatorname{Tr}_{R_{j}}\left[\left(\prod_{k=1}^{j} U_{j}\right)|\mathbf{0}\rangle\langle\mathbf{0}|U^{\dagger}H\left(\prod_{k=j+1}^{M} U_{k}\right)\right]$$

(i.e. a partial trace over the particles not acted on by U_j)

• Given $H = \sum_{\langle l,m \rangle} H_{lm}$ the circuit class must admit an efficient method of evaluating these operators.

Example (i) Contracting staircase circuits

• Contribution to the optimal generator of $U_3(t)$ due to H_{23}



• Can be evaluated by sequentially tracing out qubits from matrices of dimension at most $4D \times 4D$.

Example (ii) Contracting MERA circuits



• Slightly more complicated, but same basic idea allows us to evaluate the partial trace via a sequence of smaller partial traces on (at most) 16×16 matrices.

The Flow Algorithm

• Starting from some initial configuration of the circuit $\{U_j(0)\}\$ we can implement a downhill algorithm as follows

```
Set n = 0

Initialize each U_k(0)

Set E(0) = \langle 0 | U^{\dagger}(0) H U(0) | 0 \rangle

do:

for each k = 1, \dots, M:

Calculate optimal generator G_k(n)

Calculate optimal flow strength t

Set U_k(n+1) = \exp(-iG_k(n)t) U_k(n)

end

n = n + 1

Set E(n) = \langle 0 | U^{\dagger}(n) H U(n) | 0 \rangle

while E(n) < E(n-1)
```

Performance with "staircase" circuits (i)

Flow convergence for the 30-qubit Heisenberg chain



Performance with staircase circuits (ii)

Flow convergence for the 28-qubit Heisenberg ring



QAP Maria Laach 2007

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Conclusions and TODO

- 1. By re-expressing ansatz states for quantum many-body systems as quantum circuits, we can use the method of flow equations as a general purpose optimization method.
- 2. Appeal:
 - Flexibility
 - QI has a lot to say about quantum circuits and gates
 - "The wisdom of optimal control"
- 3. Future work:
 - MERA version in "development"
 - 1D and 2D cellular automata circuits.