

# Variational Monte Carlo with Proximal Policy Optimization

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  - Requirement
  - Neural-network quantum state (NQS)
  - Optimization: Natural gradient descent
    - PPO algorithm
    - Stochastic reconfiguration (TDVP)

## Requirement

pyTorch >= 1.8.0 + cu111

## Neural-network quantum state (NQS)

algos.core

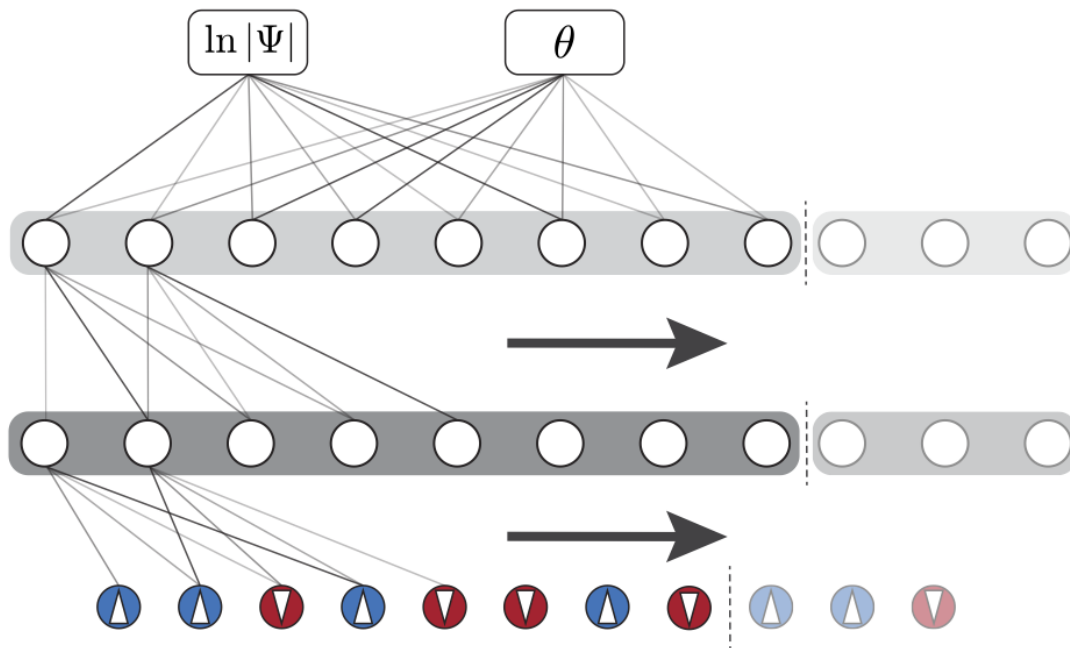
Two types of convolution neural network are implemented as NQS.

**Inputs:** spin configurations.

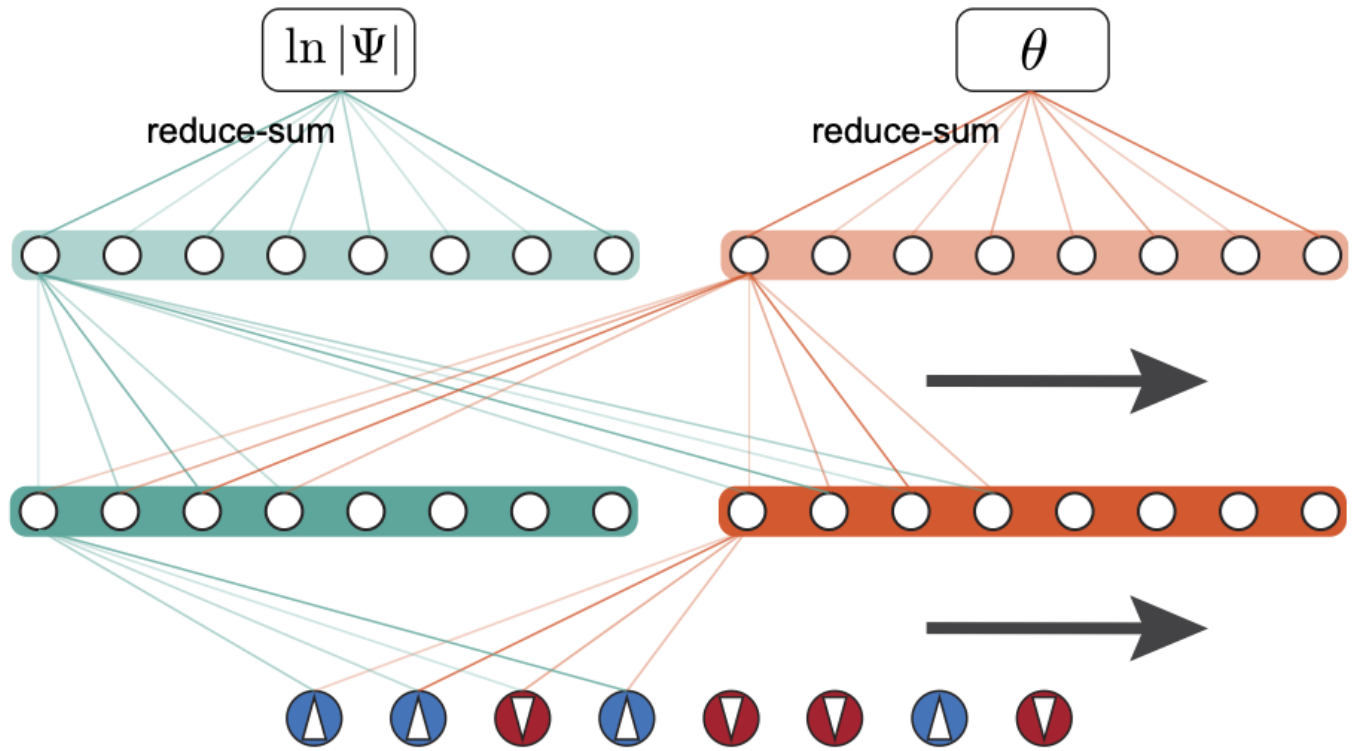
**Outputs:**

$\log |\Psi|$  and  $\theta$

- pseudocomplex CNN with only real parameters.



- complex CNN contains two sublayers in each layer, one stands for real part and the other denotes the imag part of the complex layer.



## Optimization: Natural gradient descent

Optimization object,

$$\text{minimize}_w E_w(\Psi(s; w)) \quad \text{s.t. } D_{FS}^2(\Psi_{\text{old}}(s; w_{\text{old}}), \Psi(s; w)) \leq \delta.$$

where

$E_w$

is the energy estimated via importance sampling and

$D_{FS}$

is the Fubini-Study distance.

## PPO algorithm

`algos.pseudocomplex_ppo` or `algos.complex_ppo`

NGD can be approximately solved by PPO-clip and PPO-clip updates wavefunctions via,

$$w_{k+1} = \text{argmin}_w \mathbb{E}[L(w_k, w)]$$

with the loss function,

$$L(w_k, w) = \max \left( \frac{|\Psi_w|^2}{|\Psi_{w_k}|^2} E_{w \text{ or } w_k}, \text{clip} \left( \frac{|\Psi_w|^2}{|\Psi_{w_k}|^2}, 1 - \epsilon, 1 + \epsilon \right) E_{w \text{ or } w_k} \right),$$

where

$E_w$

is estimated by the current wavefunction

$\Psi_w$

and

$E_{w_k}$

is estimated by the old wavefunction

$\Psi_{w_k}$

.

## Stochastic reconfiguration (TDVP)

`algos.pseudocomplex_sr` or `algos.complex_sr`

The optimization object of NGD is also equivalent to,

$$\text{minimize}_{\Delta w} \{E_w + \nabla_w E_w \Delta w\} \quad \text{s.t.} \quad \frac{1}{2} \Delta w^\dagger \mathbf{S} \Delta w < \delta,$$

due to,

$$D_{FS}^2 \approx \sum_{ij} dw_i^* dw_j [\langle \mathcal{O}_i^* \mathcal{O}_j \rangle - \langle \mathcal{O}_i^* \rangle \langle \mathcal{O}_j \rangle],$$

where

$$S_{ij} = [\langle \mathcal{O}_i^* \mathcal{O}_j \rangle - \langle \mathcal{O}_i^* \rangle \langle \mathcal{O}_j \rangle]$$

is the stochastic reconfiguration matrix and

$$\mathcal{O}_i = \partial_w \log \Psi$$

.

Such a conditional minimal problem is hence equivalent to:

$$\text{minimize}_{\Delta w} \{E_w + \nabla_w E_w + \lambda (\frac{1}{2} \Delta w^\dagger \mathbf{S} \Delta w - \epsilon)\},$$

with a Lagrange multiplier

$\lambda$

. Its minimal satisfies,

$$\mathbf{S} \Delta w = -\alpha \nabla_w E_w,$$

which is the exact TDVP equation.