

Here comes the $SU(N)$: multivariate quantum gates and gradients

By Roeland Wiersema
QTM - November 2023



UNIVERSITY OF
WATERLOO

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Variational Quantum Computing

Choosing an Ansatz for a variational quantum algorithm is hard

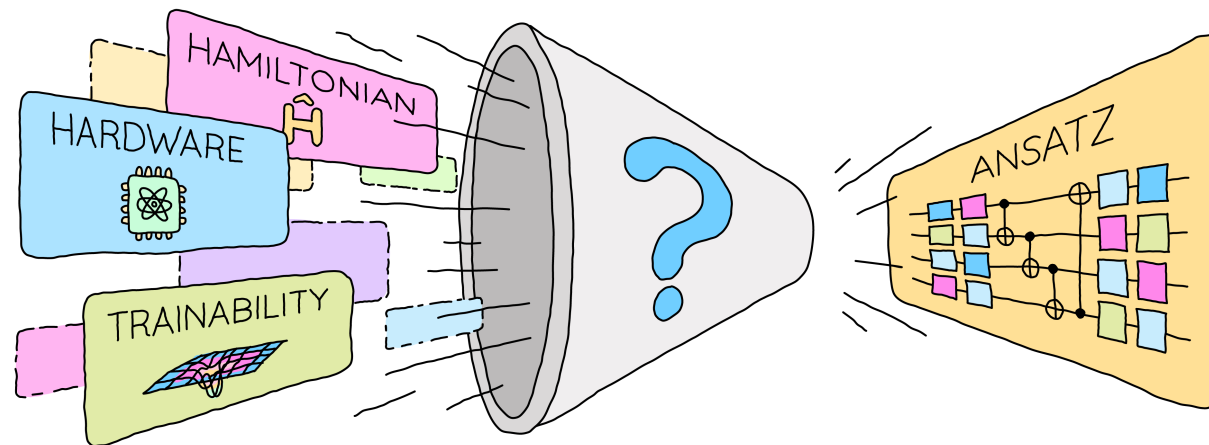


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Variational Quantum Computing

If you cannot find structure in the problem, one can pick a generic ansatz of this form:

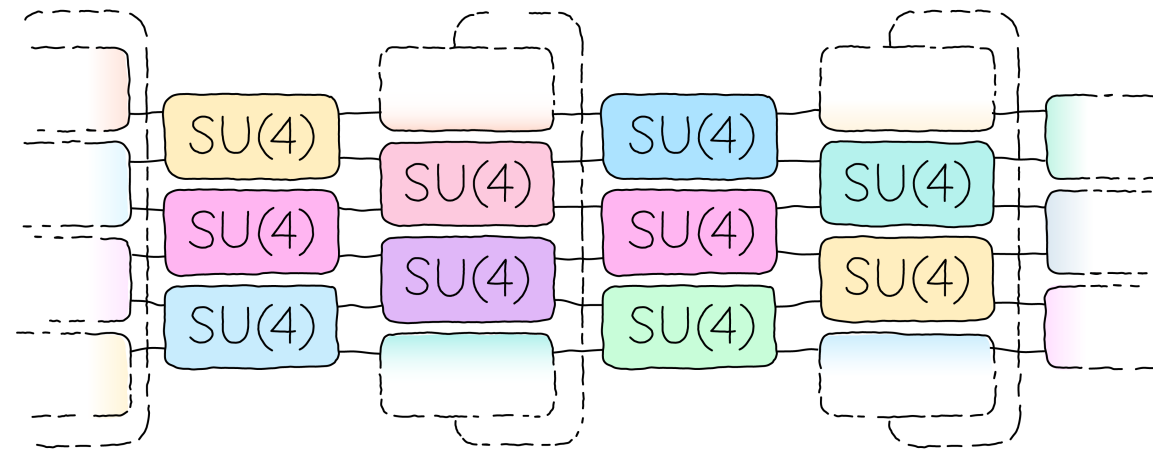



Image credit: Xanadu

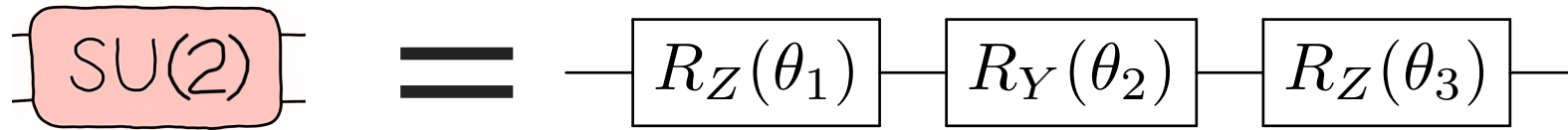
General SU(N) parameterization

$$\boxed{\text{SU}(2)} = \text{---} \boxed{R_Z(\theta_1)} \text{---} \boxed{R_Y(\theta_2)} \text{---} \boxed{R_Z(\theta_3)} \text{---}$$

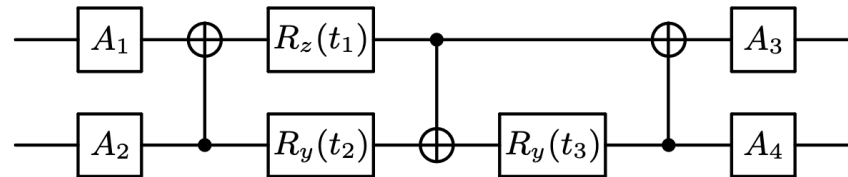

$$R_z(\theta) = \exp\{iZ\theta\}$$

Quantum gate

General SU(N) parameterization



Theorem 5. *Every two-qubit quantum gate in $U(4)$ can be realized, up to a global phase, by a circuit consisting of 15 elementary one-qubit gates and 3 CNOT gates.*



$$R_z(\theta) = \exp\{iZ\theta\}$$

Quantum gate

Are there better parameterizations out there?

General SU(N) parameterization

- Parameterize an element of the Lie algebra $\mathfrak{su}(N)$

$$A(\boldsymbol{\theta}) = \sum_m^{4^N-1} \theta_m G_m$$
$$U(\boldsymbol{\theta}) = e^{A(\boldsymbol{\theta})}$$

- With $\theta_m \in \mathbb{R}$ and G_m a Pauli string multiplied by the imaginary unit i .

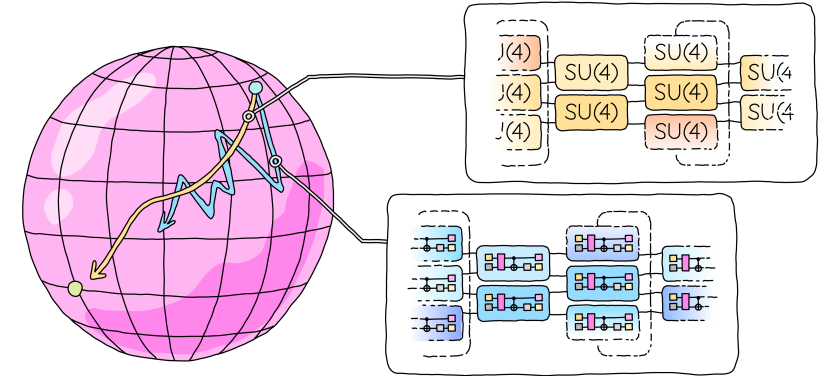
Example 1

$$R_x(\theta) = e^{i\theta X}$$

Example 2

$$A(\theta_1, \theta_2) = e^{i(\theta_1 X + \theta_2 Y)}$$

Gradients of SU(N) gates



$$\min_{\theta} C(\theta) = \min_{\theta} \text{Tr}\{U(\theta)\rho U^{\dagger}(\theta)H\}$$

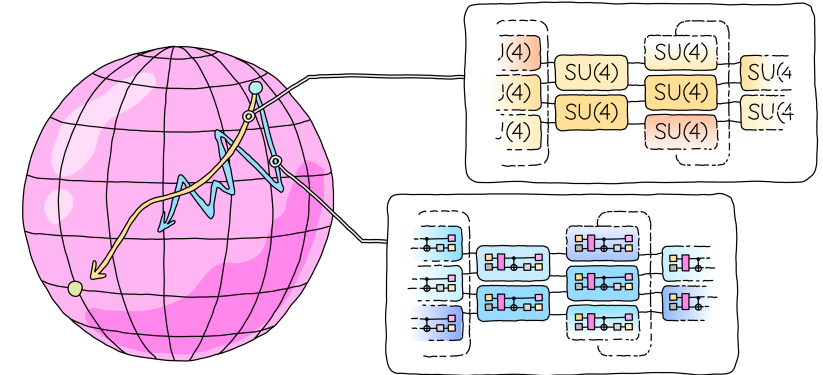
VQE Optimization

Schuld, M., Bergholm, V., Gogolin, C., Izaac, J., & Killoran, N. (2019). *Physical Review A*, 99(3), 032331.
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Izmaylov, A. F., Lang, R. A., & Yen, T. C. (2021). *Physical Review A*, 104(6), 062443.

Gradients of SU(N) gates

Parameter-shift rule (P has two eigenvalues):

$$V(t) = e^{itP} \rightarrow \partial_t C(t) = \frac{1}{2} (C(t + \delta) - C(t - \delta))$$



$$\min_{\theta} C(\theta) = \min_{\theta} \text{Tr}\{U(\theta)\rho U^\dagger(\theta)H\}$$

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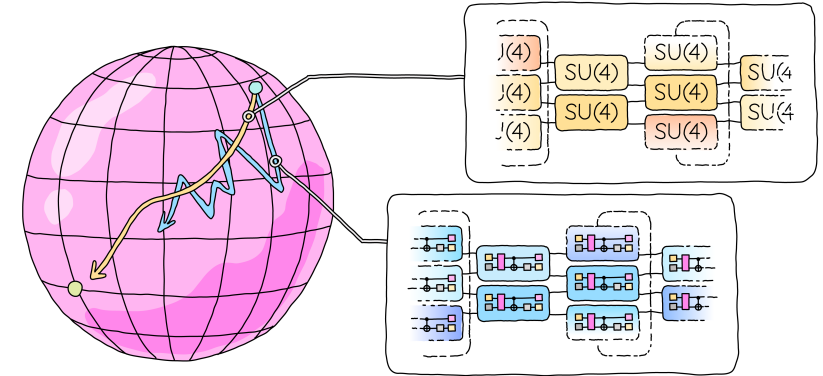
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Generalized parameter-shift rules (H Hermitian, R unique evs.):

$$W(t) = e^{itH} \rightarrow \partial_t C(t) = \sum_{i=1}^R \Delta_i (C(t + \delta_i) - C(t - \delta_i))$$



$$\min_{\theta} C(\theta) = \min_{\theta} \text{Tr}\{U(\theta)\rho U^\dagger(\theta)H\}$$

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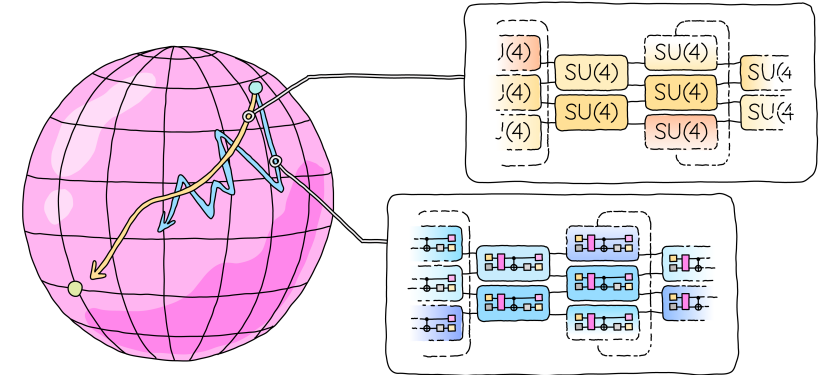
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Gates generated by non-commuting operators:

$$U(\theta_1, \theta_2) = e^{i(\theta_1 X + \theta_2 Y)} \rightarrow \partial_{\theta_1} C(\theta_1, \theta_2) = ???$$

Our work



$$\min_{\theta} C(\theta) = \min_{\theta} \text{Tr}\{U(\theta)\rho U^\dagger(\theta)H\}$$

VQE Optimization

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Gradients of SU(N) gates

- **Analytically**: an infinite series

$$\partial_{\theta_1} U(\theta_1, \theta_2) = U(\theta_1, \theta_2) \sum_{p=0}^{\infty} \frac{(-1)^p}{(p+1)!} (\text{ad}_{\theta_1 X + \theta_2 Y})^p X = \underbrace{U(\theta_1, \theta_2)}_{\text{SU}(N)} \underbrace{\Omega_{\theta_1}(\theta_1, \theta_2)}_{\mathfrak{su}(N)}$$

Gradients of SU(N) gates

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- **Numerically**: matrix derivative via automatic differentiation



$$\longrightarrow \partial_{\theta_i} \exp A(\theta) \longrightarrow U^\dagger(\theta) \partial_{\theta_i} \exp A(\theta) = \Omega_{\theta_i}(\theta)$$

Gradients of SU(N) gates

- The gradient calculation then becomes:

$$\frac{\partial}{\partial \theta_l} C(\boldsymbol{\theta}) = \text{Tr} \left\{ \left(\frac{\partial}{\partial \theta_l} U(\boldsymbol{\theta}) \right) \rho U^\dagger(\boldsymbol{\theta}) H \right\} + \text{h.c.}$$

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$$\frac{\partial}{\partial \theta_l} C(\boldsymbol{\theta}) = \text{Tr} \{ U(\boldsymbol{\theta}) \Omega_l(\boldsymbol{\theta}) \rho U^\dagger(\boldsymbol{\theta}) H \} + \text{h.c.}$$

$$\partial_{\theta_1} U(\theta_1, \theta_2) = \underbrace{U(\theta_1, \theta_2)}_{\text{SU}(N)} \underbrace{\Omega_{\theta_1}(\theta_1, \theta_2)}_{\mathfrak{su}(N)}$$

Calculate matrix derivative $\partial_{\theta_l} U(\boldsymbol{\theta})$ and obtain $\Omega_l(\boldsymbol{\theta})$ by multiplying $\partial_{\theta_l} U(\boldsymbol{\theta})$ on the **left** with $U^\dagger(\boldsymbol{\theta})$



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$$\frac{\partial}{\partial \theta_l} C(\boldsymbol{\theta}) = \frac{d}{dt} \text{Tr} \{ U(\boldsymbol{\theta}) e^{t \Omega_l(\boldsymbol{\theta})} \rho U^\dagger(\boldsymbol{\theta}) H \} + \text{h.c.} \Big|_{t=0}$$

$$\partial_{\theta_1} U(\theta_1, \theta_2) = \underbrace{U(\theta_1, \theta_2)}_{\text{SU}(N)} \underbrace{\Omega_{\theta_1}(\theta_1, \theta_2)}_{\mathfrak{su}(N)}$$

Calculate matrix derivative $\partial_{\theta_l} U(\boldsymbol{\theta})$ and obtain $\Omega_l(\boldsymbol{\theta})$ by multiplying $\partial_{\theta_l} U(\boldsymbol{\theta})$ on the **left** with $U^\dagger(\boldsymbol{\theta})$



Add the gate generated by $\Omega_l(\boldsymbol{\theta})$ to the circuit and calculate the **total derivative** of the new gate with a standard parameter-shift rule on quantum hardware



Gradients of $SU(N)$ gates

Mari, A., Bromley, T. R., & Killoran, N. (2021). *Physical Review A*, 103(1), 012405.
Banchi, L., & Crooks, G. E. (2021). *Quantum*, 5, 386.

Gradients of SU(N) gates

Finite difference

$$\partial_{\text{FD},\theta_j} C(\boldsymbol{\theta}) = \frac{1}{\delta} \left[C\left(\boldsymbol{\theta} + \frac{\delta}{2}\mathbf{e}_j\right) - C\left(\boldsymbol{\theta} - \frac{\delta}{2}\mathbf{e}_j\right) \right]$$

- Pro: simple to implement
- Con: biased estimator for the gradient

Gradients of SU(N) gates

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- Pro: simple to implement
- Con: biased estimator for the gradient

Stochastic parameter-shift

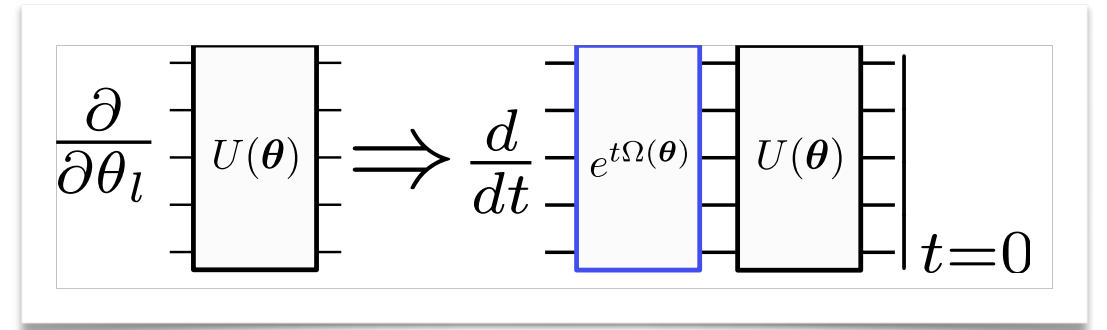
$$\partial_{\text{SPS},\theta_j} C(\boldsymbol{\theta}) = \int_0^1 ds (C_+(\boldsymbol{\theta}, s) - C_-(\boldsymbol{\theta}, s))$$

- Pro: unbiased estimator
- Con: requires Monte Carlo estimate of integral

Gradients of SU(N) gates

SU(N) gate gradients

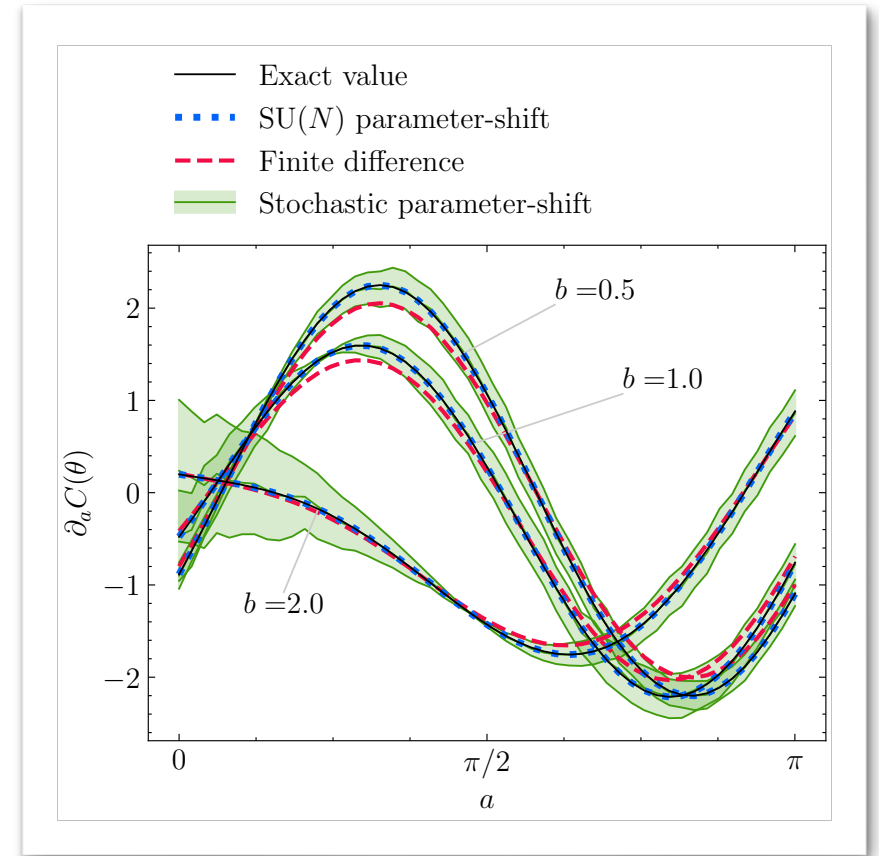
- Unbiased estimator
- Easy to implement
- Sample complexity is the same as the generalized parameter-shift rule



$$\frac{\partial}{\partial \theta_l} C(\theta) = \frac{d}{dt} \text{Tr} \{ U(\theta) e^{t\Omega_l(\theta)} \rho U^\dagger(\theta) H \} + \text{h.c.} \Big|_{t=0}$$

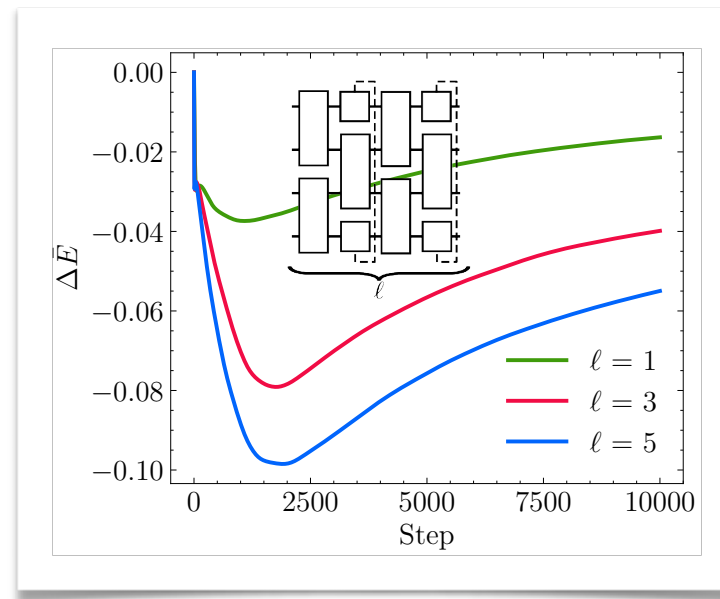
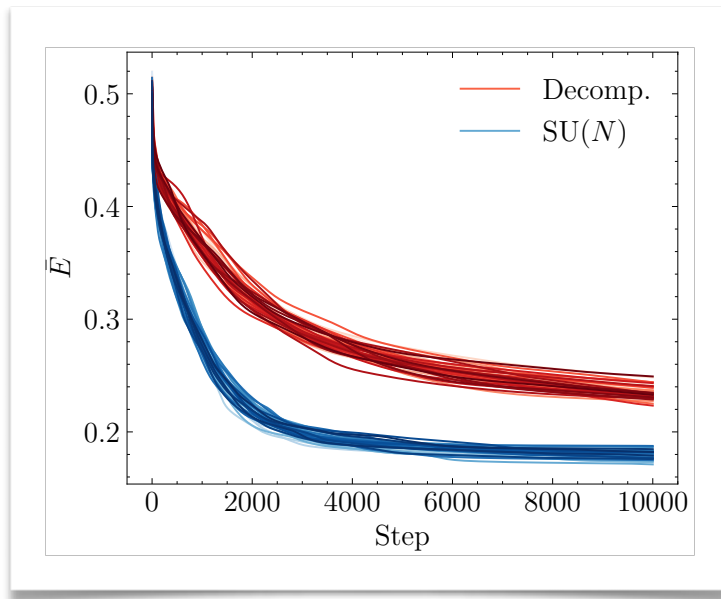
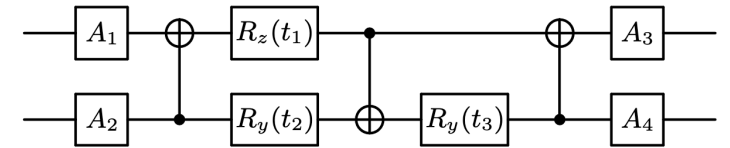
Toy example

- We consider a random single qubit Hamiltonian H
- The generator of our gate is given by $A(\theta) = iaX + ibY$
- Then, we calculate the partial derivative with respect to a

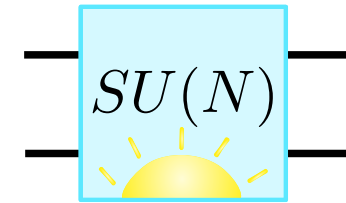


Numerical study

- We consider 100 random 10-qubit Hamiltonians with a bricklayer ansatz



V.S.



$\ell = 5$ circuit

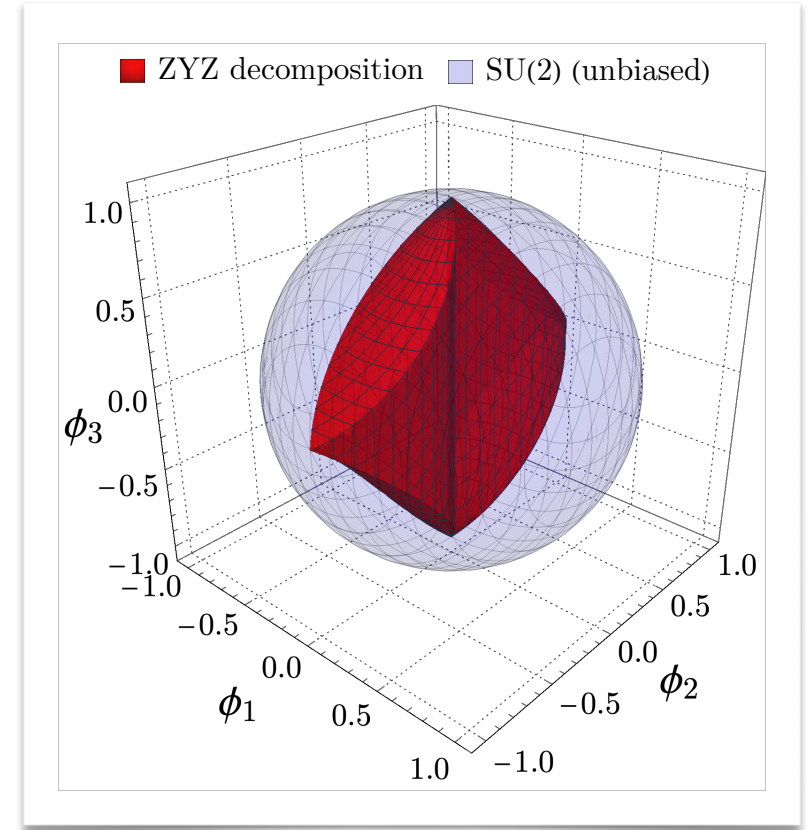
Uniform parameterization of the group

- We call a product of gates that parameterizes $SU(N)$ a *decomposed* gate

$$V(\boldsymbol{\theta}) = \text{---} \boxed{R_Z(\theta_1)} \text{---} \boxed{R_Y(\theta_2)} \text{---} \boxed{R_Z(\theta_3)} \text{---}$$

- If we normalize $\boldsymbol{\theta}$, we can write this as in our parametrization as

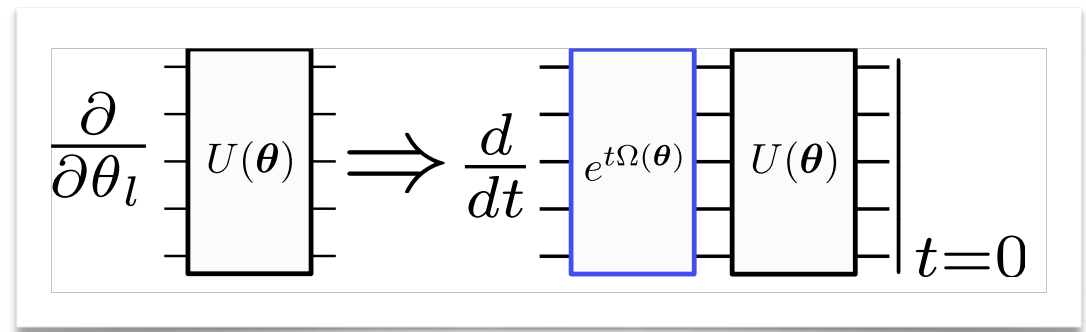
$$V(\boldsymbol{\theta}) = \exp \{ \phi_1(\boldsymbol{\theta})X + \phi_2(\boldsymbol{\theta})Y + \phi_3(\boldsymbol{\theta})Z \}$$



Resource estimation via the dynamical Lie algebra of the gate

- The dynamical Lie algebra is given by $\mathfrak{g} = \text{span}\{[\dots[G_k, [G_i, G_j]]]\}$
- The effective generator $\Omega(\theta) \in \mathfrak{g}$
- The number of circuit evaluations depends on the number of unique eigenvalues, which is related to the structure of \mathfrak{g} .

$$A(\theta) = \sum_m^{4^N - 1} \theta_m G_m$$
$$U(\theta) = e^{A(\theta)}$$



Conclusion

- We proposed a new parameterization for general $SU(N)$ gates and provide a method to obtain the gradients of this gate on hardware
- Leveraging auto-differentiation in different ways could motivate new variational algorithms
- Instead of focusing on expressivity of ansätze we should try to understand **how** we are parameterizing circuits

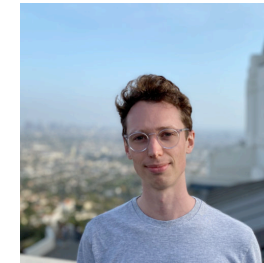


Paper

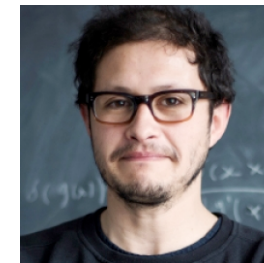


Link to demo

Dylan Lewis



Juan Carrasquilla



David Wierichs



Nathan Killoran

