Hybrid quantum tensor networks for aeroelastic applications

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Hybrid Tensor network inspired algorithm

We investigated the prospect of using hybrid quantum tensor network based algorithms for aeroelastic problems. Quantum tensor networks for



Machine Learning can be realized by Variational Quantum Circuits using a tensor network inspired internal gate structure [1].

Classical dimensionality reduction

quantum circuit. The green circles represent elements of the feature vector. Adapted from [1]

$a = 0.4, \mu = 50, U_{\infty} = 155 (m/s)$ 0.4 $a = -0.4, \mu = 40, U_{\infty} = 250 (m/s)$ 0.2 -0.0 Чh $a = -0.4, \mu = 10, U_{\infty} = 177 (m/s)$ -0.2 $\mu = 20, U_{\infty} = 233 (m/s)$ 10 -0.4 - -10^{-10} -0.6

Typical aeroelastic section of a wing with three degrees of freedom: heave h, pitch θ around the elastic axis and aileron rotation β .

Use Case

Simulation dataset for constant aeroelastic parameters up to :

- a: the nondimensional distance between midchord and elastic axis
- μ : the mass ratio
- U_{∞} : the airspeed



Simplified aeroelastic configuration. Adapted from [2]



Representative time histories: with stable (top) or unstable (bottom) aeroelastic response for non-zero initial conditions.

Regression

Training was done on \sim 7000 examples using 5-fold cross validation. And testing was carried out on ~ 2000 examples.



Classification

Predict the aeroelastic parameters (a, μ, U_{∞}) for each time history.

We used the expectation values of three Qubits to approximate • the target vectors.

Provide the binary stability classification for each time history.

• We used the expectation value of a Qubit as the class probability.





The univariate regression of *a* with 9 distinct values could be solved. Could be solved perfectly, although we observed that the results were The more complex multivariate regression including μ with 5 and very susceptible to random weight initialization. U_{∞} with 201 distinct values, could not be completely solved.

Conclusion



We found that hybrid quantum tensor network based algorithms can be successfully applied to aeroelastic problems. Nevertheless, the appropriate choice of hyperparameters is still a challenge.

We will investigate the impact of using classical tensor networks for the data dimensionality reduction. An ablation study will be performed to understand what each component contributes to the solution of the task.



[1] Rieser Hans-Martin, Köster Frank and Raulf Arne Peter (2023). Tensor networks for quantum machine learning. Proc. R. Soc. A.479:20230218.

[2] Tewari, A. (2015). Aeroservoelasticity. In Control Engineering. Springer New York



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