

## 1. ALGORITHM COMPARISON: TACKLING THE BURGERS' EQUATION

We dive into a head-to-head comparison of three cutting-edge computational approaches to solve the 1D viscous Burgers' equation, a cornerstone of fluid dynamics defined by  $\partial_t u + u\partial_x u = \nu\partial_x^2 u$ , with viscosity  $\nu = 0.1$ , domain  $x \in [0, 1]$ , boundary conditions  $u(0, t) = 1$ ,  $u(1, t) = 0$ , and initial condition  $u(x, 0) = \sin(\pi x)$ . The spatial domain is discretized into  $N$  grid points ( $dx = 1/(N - 1)$ ). The contenders are:

- **QSDE-Trotter:** Our quantum-powered method, transforming the Burgers' equation into a linear diffusion equation, implemented on a 4-qubit circuit ( $N = 16$ ) using Qiskit's AerSimulator with `ibm_torino` noise model and Zero Noise Extrapolation (ZNE).
- **HSE:** The Hydrodynamic Schrödinger Equation approach, leveraging a Madelung transform to cast the equation into a quantum-like framework, solved via a prediction-correction algorithm (Meng and Yang, 2023).
- **QTN:** The Quantum Tensor Network approach, harnessing Matrix Product States (MPS) to compress solutions, with a DMRG-type solver and open boundary conditions (Nature Communications Physics, 2024).

We also evaluate a hybrid **QTN-HSE** approach, merging HSE's physics with QTN's compression. Each comparison explores Core Approach, Computational Cost, Scalability Power, NISQ Readiness, and Winning Trade-offs, delivering a clear verdict for computational fluid dynamics (CFD) applications.

**1.1. Quantum Simplicity vs. Turbulent Ambition: QSDE-Trotter vs. HSE.** We pit our QSDE-Trotter against HSE, comparing their prowess in solving the Burgers' equation with efficiency and precision.

- **Core Approach:**
  - *QSDE-Trotter:* Uses the Cole-Hopf transform to convert the nonlinear Burgers' equation into a linear diffusion equation ( $\partial_t \psi = \nu\partial_x^2 \psi$ ,  $u = -2\nu\frac{\partial_x \psi}{\psi}$ ). A quantum circuit evolves the initial state  $\psi_0$  under a pseudo-Hamiltonian  $H = -i\nu L$  (with tridiagonal Laplacian  $L$ ) using Trotterized  $R_{xx}$  gates, transpiled to `cz`, `rz`, and `sx`.
  - *HSE:* Employs a generalized Madelung transform to map the equation to a hydrodynamic Schrödinger equation, evolving a two-component wave function under a nonlinear Hamiltonian with a prediction-correction algorithm, capturing vorticity and dissipation.
- **Computational Cost:**
  - *QSDE-Trotter:* Needs  $n = \lceil \log_2 N \rceil$  qubits (4 for  $N = 16$ ), with circuit depth scaling as  $O(N_{\text{steps}} \cdot n)$ , where  $N_{\text{steps}} = 20$  for  $t = 0.01$ . Observed depths range from 109 to 866, with `cz` gates (35–149) and single-qubit gates (120–753). Runtime is 48.29 seconds with 8192 shots. ZNE triples `cz` gates, boosting depth by 1.5-fold.
  - *HSE:* Requires more gates due to nonlinear Hamiltonian Trotterization, with no specific counts reported. The prediction-correction algorithm adds overhead, likely surpassing QSDE-Trotter's cost for  $N = 16$ .
- **Scalability Power:**
  - *QSDE-Trotter:* Boasts logarithmic qubit scaling ( $n = \lceil \log_2 N \rceil$ ), but linear depth and NISQ noise (error rate 1.498 for  $N = 16$ ) cap scalability at  $N \approx 2^{10}$ . Fault-tolerant hardware is needed for larger grids.
  - *HSE:* Faces scalability hurdles on NISQ devices due to complex gate sequences, but its unitary evolution may shine in fault-tolerant settings for 3D turbulent flows.
- **NISQ Readiness:**
  - *QSDE-Trotter:* Excels in NISQ environments with standard Qiskit gates and ZNE, reducing errors by 14.8–46.7

- *HSE*: Struggles with NISQ due to complex gate decompositions for nonlinear terms, increasing error rates ( $\sim 0.5\%$  for *cz*) and overhead from prediction-correction steps.

- **Winning Trade-offs:**

- *QSDE-Trotter* delivers simplicity and NISQ compatibility but misses vorticity modeling, limiting its scope for complex flows.
- *HSE* captures turbulent features (e.g., five-thirds energy spectrum) but demands complex circuits, making it less practical for NISQ.
- For small grids ( $N \leq 128$ ), *QSDE-Trotter* wins for efficiency; *HSE* holds promise for fault-tolerant, 3D applications.

**1.2. Quantum Speed vs. Classical Compression: QSDE-Trotter vs. QTN.** We compare *QSDE-Trotter* with *QTN*, evaluating quantum speedup against classical compression for CFD.

- **Core Approach:**

- *QSDE-Trotter*: Solves the diffusion equation on a quantum circuit, leveraging low entanglement for short times ( $t \leq 0.01$ ) and simple boundary conditions.
- *QTN*: Compresses velocity fields as MPS with bond dimension  $\chi$ , using finite differences, DMRG for the Poisson equation, and MPS masks for no-slip boundary conditions, adept at handling complex geometries classically.

- **Computational Cost:**

- *QSDE-Trotter*: Uses 4 qubits, depth 109–866, and runtime 48.29 s for  $N = 16$ .
- *QTN*: Scales as  $O(N\chi^6)$  per time step, reducible to  $O(N\chi^4)$  with variational methods. For  $N = 19$ ,  $\chi = 45$ , it uses 18,800 parameters (27.9X compression), with runtime seconds per iteration, competitive but slower for small  $N$ .

- **Scalability Power:**

- *QSDE-Trotter*: Limited by NISQ noise and linear depth, effective up to  $N \approx 2^{10}$ .
- *QTN*: Tackles large grids ( $N = 2^{19}$ ) with moderate  $\chi$ , but  $\chi$  growth with flow complexity increases costs.

- **NISQ Readiness:**

- *QSDE-Trotter*: Built for NISQ with standard gates and ZNE, ideal for quantum hardware.
- *QTN*: Classical, sidestepping quantum noise but missing quantum speedup opportunities.

- **Winning Trade-offs:**

- *QSDE-Trotter* offers quantum speedup potential but lacks flexibility for complex geometries.
- *QTN* shines in compression and boundary condition handling, perfect for high-dimensional flows (e.g., 2D/3D).
- *QSDE-Trotter* is simpler for 1D problems; *QTN* dominates for complex, large-scale simulations.

**1.3. Simplicity vs. Hybrid Power: QSDE-Trotter vs. QTN-HSE.** We evaluate *QSDE-Trotter* against the hybrid *QTN-HSE*, blending *HSE*'s physics with *QTN*'s compression.

- **Core Approach:**

- *QSDE-Trotter*: Relies on a linear diffusion equation for straightforward circuit design, ideal for 1D problems with minimal vorticity.
- *QTN-HSE*: Maps the Burgers' equation to a Schrödinger-like form, compressing the two-component wave function as an MPS, evolved via hybrid quantum-classical algorithms (e.g., variational or DMRG).

- **Computational Cost:**

- *QSDE-Trotter*: As above, with runtime 48.29 s for  $N = 16$ .

- *QTN-HSE*: Combines HSE’s high gate counts with QTN’s  $O(N\chi^6)$  or  $O(N\chi^4)$  complexity. For  $N = 16$ ,  $\chi = 30$ , runtime exceeds QSDE-Trotter due to hybrid iterations.
- **Scalability Power:**
  - *QSDE-Trotter*: Efficient for small  $N$ , but NISQ noise limits larger grids.
  - *QTN-HSE*: Handles large  $N$  (e.g.,  $2^{19}$ ) via MPS compression, but hybrid complexity and  $\chi$  growth reduce scalability compared to QTN alone.
- **NISQ Readiness:**
  - *QSDE-Trotter*: Highly NISQ-compatible with simple gates and ZNE.
  - *QTN-HSE*: Less suitable due to complex quantum circuits for nonlinear terms and classical MPS updates, increasing error risks.
- **Winning Trade-offs:**
  - QSDE-Trotter is efficient and NISQ-friendly for 1D problems.
  - QTN-HSE captures complex flow physics with compression but is weighed down by hybrid complexity.
  - QTN-HSE may lead in fault-tolerant settings for 3D turbulent flows, while QSDE-Trotter excels in small-scale NISQ simulations.

**1.4. Conclusion.** Our QSDE-Trotter approach delivers a lean, NISQ-ready solution for the 1D Burgers’ equation, with logarithmic qubit scaling and straightforward circuits. HSE brings ambitious turbulent modeling but stumbles in NISQ environments due to complex gates. QTN offers unmatched compression for high-dimensional flows, shining in classical settings with complex geometries. The hybrid QTN-HSE merges advanced physics with compression but struggles with hybrid complexity. For small grids ( $N \leq 128$ ), QSDE-Trotter and QTN take the lead, while HSE and QTN-HSE await fault-tolerant quantum hardware to unleash their full potential in large-scale, turbulent CFD challenges.