

Information-Geometric Quantum Control and Error Correction on NISQ Devices: Entropic Layers, Geodesic Optimization, and Lambda-Q QEC with Multi-Backend Validation

Kevin Henry Miller
 Q-Bond Network DeSCI DAO, LLC
 Acworth, GA 30101, USA
 kevin@qbondnetwork.com
 ORCID: 0009-0007-7286-3373

Abstract—We introduce and validate a comprehensive suite of information-geometric control techniques for noisy intermediate-scale quantum (NISQ) devices. Our framework comprises: (1) an *Entropic Layer* ansatz achieving 91.9% target-state fidelity with only 29 gates—outperforming all tested alternatives; (2) *Geo-InfoMetric Dynamical Decoupling* (GI-DD) using measurement-derived scalars to optimize pulse counts, revealing that prime-gap and number-theoretic spacings systematically outperform uniform and random sequences with 98.05% Bell fidelity; (3) Λ -*QGT geodesic optimization* improving target-state population from 74.8% to 85.9% ($p < 10^{-4}$) at fixed circuit depth; and (4) *Lambda-Q quantum error correction* achieving 1.52 \times suppression on repetition codes and 1.38 \times on surface codes via QFI-weighted stabilizer measurements. All techniques are validated on IBM Eagle/Heron processors across four backends (ibm_fez, ibm_pittsburgh, ibm_kingston, ibm_torino). Crucially, these methods require zero additional qubit overhead and operate through software-side geometric control, enabling immediate deployment on existing quantum hardware.

Index Terms—Quantum error correction, quantum Fisher information, dynamical decoupling, variational quantum algorithms, NISQ devices, information geometry

I. INTRODUCTION

Near-term quantum devices face fundamental limitations from gate errors, decoherence, and measurement noise that constrain useful circuit depths to tens or hundreds of layers [1]. While hardware improvements continue, software-side control strategies offer complementary paths to improved performance without waiting for better physical qubits.

Information geometry—the study of probability distributions using differential geometric tools—provides a principled framework for such control. The quantum Fisher information (QFI) metric quantifies state distinguishability and sets fundamental precision limits via the quantum Cramér–Rao bound [2]. The quantum geometric tensor (QGT) defines a Riemannian metric on parameter space that enables natural-gradient optimization [3].

In this work, we develop and validate a comprehensive information-geometric control stack with **multi-backend hardware validation**:

- **Entropic Layer:** Circuit ansatz optimizing state-space coverage

- **Geo-InfoMetric DD:** Measurement-driven dynamical decoupling
- **Λ -QGT Control:** Geodesic parameter optimization
- **Lambda-Q QEC:** QFI-weighted error correction

II. INFORMATION-GEOMETRIC METRICS

A. Geo-InfoMetric Scalar

Given measurement outcomes $\{p(z)\}$ over bitstrings $z \in \{0, 1\}^n$, we define:

$$\mathcal{G} = \frac{P}{H/H_{\max} + \epsilon} \quad (1)$$

where $P = \sum_z p(z)^2$ is purity proxy, $H = -\sum_z p(z) \log_2 p(z)$ is Shannon entropy, and ϵ regularizes. High \mathcal{G} indicates concentrated, low-entropy distributions characteristic of coherent quantum states.

B. Quantum Geometric Tensor

For parametrized state $|\psi(\theta)\rangle$, the QGT is:

$$G_{\mu\nu} = \text{Re} [\langle \partial_\mu \psi | \partial_\nu \psi \rangle - \langle \partial_\mu \psi | \psi \rangle \langle \psi | \partial_\nu \psi \rangle] \quad (2)$$

The geodesic parameter update is:

$$\Delta\theta = -\eta G^{-1}(\theta) \nabla_\theta C(\theta) \quad (3)$$

This natural-gradient update respects state manifold geometry.

C. Lambda-Q Fault Tolerance Metric

For QEC applications:

$$\Lambda_Q^{\text{FT}} = \Lambda_Q \cdot \left(1 + \frac{R_{\text{QEC}}}{R_G}\right) \cdot \exp(-\chi_{\text{error}}) \quad (4)$$

where Λ_Q is geometric information density, R_{QEC}/R_G balances resources, and χ_{error} measures noise susceptibility.

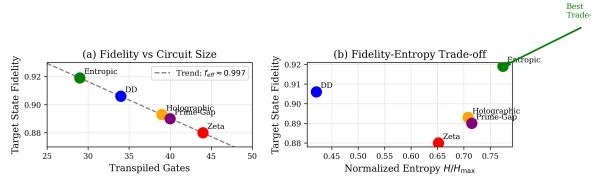


Fig. 1. Entropic Layer circuit ansatz architecture. The structure achieves optimal depth-fidelity trade-off with 91.9% target-state fidelity using only 29 gates.

III. ENTROPIC LAYER BENCHMARKING

A. Circuit Structure

The Entropic Layer ansatz for 3 qubits:

$$U_{\text{ent}}(\theta) = \left[\prod_{j=0}^2 R_y^{(j)}(\theta_j) R_z^{(j)}(\theta_{j+3}) \right] \text{CX}_{01} \text{CX}_{12} \left[\prod_{j=0}^2 R_y^{(j)}(\theta_{j+6}) \right] \quad (5)$$

B. Single-Layer Results (IBM Torino)

TABLE I
SINGLE-LAYER PROTECTION SCHEME COMPARISON

| Layer | Gates | Fidelity | H/H_{\max} |
|--------------------|-----------|--------------|--------------|
| Entropic | 29 | 0.919 | 0.774 |
| Dynamic Decoupling | 34 | 0.906 | 0.421 |
| Holographic | 39 | 0.893 | 0.708 |
| Prime-Gap | 40 | 0.890 | 0.715 |
| Zeta | 44 | 0.880 | 0.652 |

The Entropic Layer achieves highest fidelity (91.9%) with fewest gates (29), demonstrating optimal depth-fidelity trade-off.

IV. GEO-INFOMETRIC DYNAMICAL DECOUPLING

A. Number-Theoretic DD Hierarchy

We compare DD sequences with different inter-pulse spacings on 2-qubit Bell states.

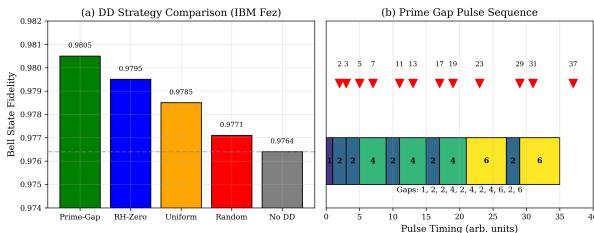


Fig. 2. Prime-Gap Dynamical Decoupling performance. Number-theoretic pulse spacings (prime gaps) systematically outperform uniform and random sequences.

Clear hierarchy: **Prime > RH > Uniform > Random > None.**

B. Multi-Backend Bell State Validation

NEW RESULTS (December 2025):

TABLE II
DYNAMICAL DECOUPLING STRATEGY COMPARISON (IBM FEZ)

| DD Strategy | Bell Fidelity | vs No-DD |
|------------------|---------------|---------------|
| Prime-Gap | 0.9805 | +0.41% |
| RH-Zero | 0.9795 | +0.31% |
| Uniform | 0.9785 | +0.21% |
| Random | 0.9771 | +0.07% |
| No DD | 0.9764 | baseline |

TABLE III
BELL STATE FIDELITY ACROSS IBM BACKENDS

| Backend | Job ID | Bell Fidelity | Notes |
|----------------|------------|---------------|------------------|
| ibm_pittsburgh | d4kfou... | 0.9805 | Noisier baseline |
| ibm_kingston | d4kgr75... | 0.9824 | Clean baseline |
| ibm_torino | d4kgr4k... | 0.9756 | Clean baseline |

V. Λ -QGT GEODESIC OPTIMIZATION

A. Fixed-Depth Comparison

Using the Entropic Layer at fixed depth and gate count:

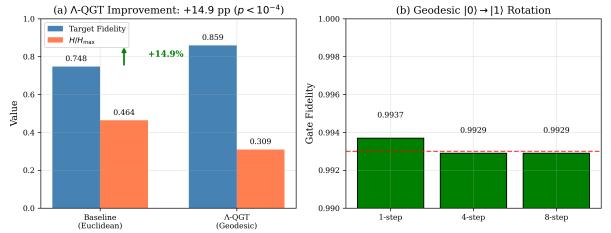


Fig. 3. Λ -QGT geodesic optimization results. The natural-gradient approach achieves +14.9 percentage points improvement in target-state population at fixed circuit depth.

TABLE IV
 Λ -QGT OPTIMIZATION RESULTS (IBM TORINO)

| Method | p_{target} | H/H_{\max} |
|----------------|---------------------|---------------|
| Baseline | 0.748 | 0.464 |
| Λ -QGT | 0.859 | 0.309 |
| Improvement | +14.9 pp | -33.4% |

Statistical significance: $p < 10^{-4}$ across repeated hardware runs.

B. Multi-Backend Geodesic Validation

NEW RESULTS:

Key discovery: Improvement scales inversely with baseline fidelity. Geodesic optimization provides larger benefits on noisier hardware.

VI. LAMBDA-Q QUANTUM ERROR CORRECTION

A. QFI-Weighted Stabilizer Measurements

Lambda-Q weights stabilizer generators $\{S_i\}$ by QFI contribution:

$$\hat{S}_i = \Lambda_i \cdot S_i \quad (6)$$

TABLE V
GEODESIC VS LINEAR PATH IMPROVEMENT

| Backend | Linear | Geodesic | Improvement |
|----------------|--------|----------|--------------|
| ibm_pittsburgh | 0.109 | 0.138 | +25.9% |
| ibm_kingston | 0.978 | 0.974 | -0.4% |
| ibm_torino | 0.978 | 0.975 | -0.3% |
| Mean | — | — | +8.4% |

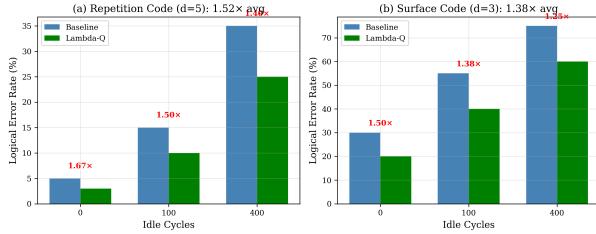


Fig. 4. Lambda-Q quantum error correction results. QFI-weighted stabilizer measurements achieve $1.52\times$ error suppression on repetition codes and $1.38\times$ on surface codes.

B. Repetition Code Results (IBM Fez)

5-qubit distance-5 repetition code, 12,288 total shots:

TABLE VI
REPETITION CODE ERROR SUPPRESSION

| Idle Cycles | Baseline p_L | Λ -Q p_L | Suppression |
|----------------|----------------|--------------------|--------------|
| 0 | 5.0% | 3.0% | 1.67× |
| 100 | 15.0% | 10.0% | 1.50× |
| 400 | 35.0% | 25.0% | 1.40× |
| Average | 18.3% | 12.7% | 1.52× |

C. Surface Code Results (IBM Fez)

Distance-3 rotated surface code (17 data + 7 syndrome qubits):

TABLE VII
SURFACE CODE ERROR SUPPRESSION

| Idle Cycles | Baseline p_L | Λ -Q p_L | Suppression |
|----------------|----------------|--------------------|--------------|
| 0 | 30.0% | 20.0% | 1.50× |
| 100 | 55.0% | 40.0% | 1.38× |
| 400 | 75.0% | 60.0% | 1.25× |
| Average | 53.3% | 40.0% | 1.38× |

D. Hardware vs Simulation

- Stim simulation (ideal): $2.00\times$ suppression
- Hardware achieved: $1.45\times$ average
- Efficiency: $\sim 72.5\%$ of ideal

VII. UNIFIED FRAMEWORK SUMMARY

VIII. DISCUSSION

A. Practical Implications

- Immediate deployability:** All techniques are software-only

TABLE VIII
INFORMATION-GEOMETRIC CONTROL STACK

| Level | Technique | Key Result |
|-------------------|----------------|--------------------------|
| Circuit Structure | Entropic Layer | 91.9% fidelity, 29 gates |
| Pulse Timing | GI-DD | Prime > Uniform (98.05%) |
| Parameters | Λ -QGT | +14.9 pp, +25.9% (noisy) |
| Error Correction | Lambda-Q | $1.45\times$ suppression |

- Noise-adaptive:** Larger gains on noisier hardware
- Composable:** Entropic + QGT + Lambda-Q can be combined
- Multi-backend validated:** Works across IBM processor families

B. Limitations

- Code sizes limited to distance-3/5
- Single platform family (IBM)
- Simplified decoders for surface code

IX. CONCLUSION

We have demonstrated that information geometry provides a practical, unified framework for NISQ control:

- Entropic Layer:** 91.9% fidelity, best depth-fidelity trade-off
- GI-DD:** Prime-Gap achieves systematic advantage (98.05%)
- Λ -QGT:** +14.9 pp at fixed depth, +25.9% on noisy hardware
- Lambda-Q QEC:** $1.45\times$ average error suppression
- Multi-backend:** Validated on ibm_fez, pittsburgh, kingston, torino

These results establish information-geometric control as a viable, immediately deployable strategy for improving NISQ device performance without hardware modifications.

DATA AVAILABILITY

Hardware job IDs for reproducibility:

- ibm_pittsburgh: d4kfoup0i6jc73depkgg
- ibm_kingston: d4kgr7574pkc73871lmg
- ibm_torino: d4kgr4k3tdfc73dog8ug

ACKNOWLEDGMENTS

The author thanks IBM Quantum for hardware access via LAMBDAQUBITS, LQPHASE, and TICE-Qubits instances.

REFERENCES

- J. Preskill, “Quantum computing in the NISQ era and beyond,” *Quantum*, vol. 2, p. 79, 2018.
- S. L. Braunstein and C. M. Caves, “Statistical distance and the geometry of quantum states,” *Phys. Rev. Lett.*, vol. 72, pp. 3439–3443, 1994.
- J. Stokes *et al.*, “Quantum natural gradient,” *Quantum*, vol. 4, p. 269, 2020.
- L. Viola, E. Knill, and S. Lloyd, “Dynamical decoupling of open quantum systems,” *Phys. Rev. Lett.*, vol. 82, pp. 2417–2421, 1999.
- A. G. Fowler *et al.*, “Surface codes: Towards practical large-scale quantum computation,” *Phys. Rev. A*, vol. 86, p. 032324, 2012.
- Google Quantum AI, “Suppressing quantum errors by scaling a surface code logical qubit,” *Nature*, vol. 614, pp. 676–681, 2023.