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# Titel

Noise-Regime Crossovers in Logical-Subspace Retention: Matched-Null Benchmarks, Manifold Probing, and Leakage-Aware Scoring (n=4)

## Abstract

I study open-system retention of a two-dimensional logical subspace (one logical qubit) embedded in a 4-qubit Hilbert space using a matched-control design: a structured REAL construction is compared against a spectrum-matched NULL baseline under identical sampling, time grids, and Lindblad noise. To reduce probe-state (observable-regime) bias, I distinguish a coarse ZX screening set (4 canonical logical directions) from manifold probing using rand64 ( $\approx 66$  probes, 64 random directions plus two basis states). Unless stated otherwise, headline results use rand64, while ZX runs are retained as diagnostic baselines. I report unconditional fidelity  $F_{\text{uncond}}(t)$ , conditional fidelity  $F_{\text{cond}}(t)$  (conditioned on survival in the logical subspace), and leakage  $L(t)$ , summarized by AUC over  $t \in [0, 5]$  and by threshold times. Under amplitude damping ( $\gamma_1 > 0$ ) I consistently observe a coherent-but-leaky regime: REAL improves  $F_{\text{cond}}$  but increases leakage and degrades  $F_{\text{uncond}}$ . In contrast, under dephasing-only noise ( $\gamma\phi = 0.10$ ,  $\gamma_1 = 0.00$ ) REAL exhibits a batch-stable net improvement in  $F_{\text{uncond}}$  that is robust to logical-basis choice and to single-qubit noise localization (subsets 0–3). A dephasing-strength sweep reveals a crossover: at  $\gamma\phi = 0.05$  no net unconditional advantage is observed, whereas at  $\gamma\phi = 0.20$  the unconditional advantage becomes large and leakage is reduced relative to NULL. A mixed-noise bridge run (both,  $\gamma\phi = 0.10$ ,  $\gamma_1 = 0.02$ ) preserves a positive unconditional AUC gain while reintroducing early-time threshold penalties and higher leakage. Finally, I introduce a leakage-aware cost proxy to quantify when gains remain beneficial under leakage costs.

## 1. Introduction

A common failure mode in evaluating logical encodings, protected subspaces, or structured logical constructions is to over-index on a metric that conditions away failure modes. In open-system dynamics, an encoding may show improved **conditional** quality inside the logical subspace while simultaneously increasing the probability of leaving it (leakage), resulting in worse **unconditional** performance. This motivates a protocol that (i) uses a matched control, (ii) closes probe-state loopholes, (iii) checks batch stability, and (iv) explicitly tracks leakage.

This paper reports a controlled benchmark study using:

- matched-null design (REAL vs spectrum-matched NULL),
- manifold probing (rand64),
- logical basis robustness (noise\_diag vs eigen),
- full single-qubit noise localization sweep (subsets 0–3),
- and a dephasing-strength crossover test.

The key empirical outcome is a regime split:

- amplitude-damping dominated noise yields a robust **tradeoff** (coherent-but-leaky),

- dephasing-only yields a **net unconditional advantage** for REAL under sufficiently strong dephasing.

## 2. Methods

### 2.1 Logical subspace and probe states

Any pure logical probe state can be parameterized on the logical Bloch sphere as  $|\psi_L(\theta, \phi)\rangle = \cos(\theta/2)|0_L\rangle + e^{i\phi}\sin(\theta/2)|1_L\rangle$ , with  $\theta \in [0, \pi]$  and  $\phi \in [0, 2\pi]$ . (Nielsen & Chuang, 2010; Bloch, 1946)

As  $(\theta, \phi)$  vary,  $|\psi_L(\theta, \phi)\rangle$  traces the logical Bloch sphere restricted to  $\mathcal{L}$ .

Probe-state regimes. I use two probe ensembles: (i) ZX screening with four canonical logical directions (fast but coarse), and (ii) manifold probing (rand64), which samples  $\approx 66$  directions (64 random plus two basis states). Unless stated otherwise, headline results use rand64; ZX runs are included as diagnostic baselines to document when coarse probing can mask direction-dependent effects. (Marsaglia, 1972)

### 2.2 Open-system dynamics (Lindblad)

Dynamics are simulated under Lindblad evolution with localized noise acting on a single-qubit subset ( $\text{subset} \in \{0,1,2,3\}$ , specified per run). Time is discretized uniformly over  $t \in [0, 5.0]$  using  $t_{\text{steps}}=25$  points ( $\Delta t = 5.0/(25-1) \approx 0.208$ ).  $\{q\} q \in \{0,1,2,3\} t \in [0, T] T = 5.0$  (Lindblad, 1976; Gorini, Kossakowski, & Sudarshan, 1976; Breuer & Petruccione, 2002)

Noise regimes:

- Amplitude damping (amp\_damp): energy relaxation with rate  $\gamma_1$ ; in the reported runs a small background dephasing term  $\gamma\phi$  may be included as specified (e.g.,  $\gamma\phi=0.01$ ).
- Dephasing-only (dephasing): phase-randomization with rate  $\gamma\phi$ , with  $\gamma_1=0.00$ .
- Both (both): simultaneous dephasing ( $\gamma\phi$ ) and amplitude damping ( $\gamma_1$ ) acting together.

### 2.3 Matched-control design (REAL vs NULL)

I evaluate a structured construction (REAL) against a spectrum-matched baseline (NULL) under matched sampling. Within each run, REAL and NULL share the same probe ensemble, time grid, noise parameters (including localization to all qubits or a specified subset), and randomization schedule (3 batches;  $\text{seeds\_per\_batch}=5000$ ;  $\text{pairs\_per\_batch}=400$ ). An optional `stable_pool` filter can restrict evaluation to candidates passing a stability screen; because this can introduce selection bias, `stable_pool` is reported explicitly per run and is disabled for headline results unless selection and evaluation are separated (e.g., holdout seeds).

#### 2.3a Run registry and provenance

For transparency and reproducibility, each figure/table is tied to a concrete run ID (output filename) and its key settings: probe-state regime (ZX vs rand64), logical basis (eigen vs noise\_diag), noise localization, and whether `stable_pool` was enabled.

Usage map (quick audit): the following bullets link manuscript sections/tables to the corresponding output files listed in the run registry.

- **Section 3.1** (amp\_damp): v23\_ampd\_g10\_subset3\_noise\_diag\_p400\_rand64.txt; v23\_ampd\_g05\_subset3\_noise\_diag\_p400\_rand64.txt; v23\_ampd\_g02\_subset3\_noise\_diag\_p400\_rand64.txt.
- **Section 3.2** (dephasing  $\gamma\phi=0.10$ ): v23\_deph\_gphi10\_subset3\_noise\_diag\_p400\_rand64.txt; v23\_deph\_gphi10\_subset3\_eigen\_p400\_rand64.txt; subset sweep: v23\_deph\_gphi10\_subset0\_eigen\_p400\_rand64.txt; v23\_deph\_gphi10\_subset1\_eigen\_p400\_rand64.txt; v23\_deph\_gphi10\_subset2\_eigen\_p400\_rand64.txt.
- **Section 3.3** ( $\gamma\phi$  sweep): v23\_deph\_gphi05\_subset3\_eigen\_p400\_rand64.txt; v23\_deph\_gphi20\_subset3\_eigen\_p400\_rand64.txt (plus  $\gamma\phi=0.10$  as above).
- **Section 3.4** (both): v23\_both\_gphi10\_g02\_subset3\_eigen\_p400\_rand64.txt (batches 1–2 in this draft).

**Table 0:** batch-level deltas for the above runs; Table S2b: endpoint quantiles for amp\_damp + dephasing runs (mixed-noise endpoint quantiles pending).

Run set	Run ID / output	Noise model	Rates	Probe states	Basis	Noise qubits	stable_pool	pairs/batch	t_max / steps
B (headline)	v23_ampd_g10_subset3_noise_diag_p400 - rand64.txt	amp_damp	$\gamma\phi=0.01$ , $\gamma_1=0.10$	rand64	noise_diag	subset={3}	False	400	5.0 / 25
B (headline)	v23_ampd_g05_subset3_noise_diag_p400 - rand64.txt	amp_damp	$\gamma\phi=0.01$ , $\gamma_1=0.05$	rand64	noise_diag	subset={3}	False	400	5.0 / 25
B (headline)	v23_ampd_g02_subset3_noise_diag_p400 - rand64.txt	amp_damp	$\gamma\phi=0.01$ , $\gamma_1=0.02$	rand64	noise_diag	subset={3}	False	400	5.0 / 25
B (headline)	v23_deph_gphi1_0_subset3_noise_diag_p400 - rand64.txt	dephasing	$\gamma\phi=0.10$ , $\gamma_1=0.00$	rand64	noise_diag	subset={3}	False	400	5.0 / 25
B (headline)	v23_deph_gphi0_5_subset3_eigen_p400_rand64.txt	dephasing	$\gamma\phi=0.05$ , $\gamma_1=0.00$	rand64	eigen	subset={3}	False	400	5.0 / 25
B (headline)	v23_deph_gphi1_0_subset3_eigen_p400_rand64.txt	dephasing	$\gamma\phi=0.10$ , $\gamma_1=0.00$	rand64	eigen	subset={3}	False	400	5.0 / 25
B (headline)	v23_deph_gphi1_0_subset0_eigen_p400_rand64.txt	dephasing	$\gamma\phi=0.10$ , $\gamma_1=0.00$	rand64	eigen	subset={0}	False	400	5.0 / 25
B (headline)	v23_deph_gphi1_0_subset1_eigen_p400_rand64.txt	dephasing	$\gamma\phi=0.10$ , $\gamma_1=0.00$	rand64	eigen	subset={1}	False	400	5.0 / 25
B (headline)	v23_deph_gphi1_0_subset2_eigen_p400_rand64.txt	dephasing	$\gamma\phi=0.10$ , $\gamma_1=0.00$	rand64	eigen	subset={2}	False	400	5.0 / 25
B (headline)	v23_deph_gphi2_0_subset3_eigen_p400_rand64.txt	dephasing	$\gamma\phi=0.20$ , $\gamma_1=0.00$	rand64	eigen	subset={3}	False	400	5.0 / 25
B (headline)	v23_both_gphi1_0_g02_subset3_eigen_p400_rand64.txt	both	$\gamma\phi=0.10$ , $\gamma_1=0.02$	rand64	eigen	subset={3}	False	400	5.0 / 25
A (diagnostic)	v23_ampd_g10_subset3_noise_diag_p400.txt	amp_damp	$\gamma\phi=0.01$ , $\gamma_1=0.10$	ZX	noise_diag	subset={3}	False	400	5.0 / 25

### 2.3b Evidence criteria (decision rule)

We treat each batch as an independent replicate. A claimed effect must satisfy: (i) matched advantage (REAL–NULL) under spectrum-matched controls; (ii) replicate consistency (sign matches in  $\geq 2/3$  batches, ideally 3/3); (iii) practical effect floor  $|\Delta| \geq 0.01$  for the primary summary metric; and (iv) leakage accounting, interpreting unconditional and conditional fidelities jointly with leakage.

## 2.4 Metrics: leakage, conditional fidelity, unconditional fidelity

Let  $P_L$  be the projector onto the logical subspace  $L$ . For an evolved state  $\rho(t)$  and a target logical pure state  $|\psi_L\rangle$  (the intended logical probe state), I compute three coupled quantities: leakage, conditional fidelity, and unconditional fidelity. (Jozsa, 1994; Uhlmann, 1976; Nielsen & Chuang, 2010)

Notation note: in tables and command-line outputs I use fid\_uncond and fid\_cond; in the prose I may also write F\_uncond and F\_cond for the same quantities.

Leakage:  $L(t) = 1 - \text{Tr}[P_L \rho(t)]$ .

Conditional state given survival:  $\rho_{\text{cond}}(t) = P_L \rho(t) P_L / \text{Tr}[P_L \rho(t)]$  (defined when  $\text{Tr}[P_L \rho(t)] > 0$ ).

Conditional fidelity:  $F_{\text{cond}}(t) = \langle \psi_L | \rho_{\text{cond}}(t) | \psi_L \rangle$ . (Jozsa, 1994; Uhlmann, 1976; Nielsen & Chuang, 2010)

Unconditional fidelity (survival-weighted):  $F_{\text{uncond}}(t) = \text{Tr}[P_L \rho(t)] \cdot F_{\text{cond}}(t)$ . (Jozsa, 1994; Uhlmann, 1976; Nielsen & Chuang, 2010)

Summaries: I report AUC over  $t \in [0, t_{\text{max}}]$  for each metric, along with endpoint quantiles at  $t=t_{\text{max}}$  and median threshold times  $t_{\text{F90}}$  (time to  $F_{\text{uncond}} < 0.90$ ) and  $t_{\text{L10}}$  (time to  $L > 0.10$ ), computed per batch. (Bradley, 1997)

Sign convention: unless stated otherwise, deltas are  $\Delta = \text{REAL} - \text{NULL}$ . For fidelity metrics, positive  $\Delta\text{AUC}$  is favorable; for leakage, negative  $\Delta\text{AUC}(L)$  is favorable (REAL leaks less). (Jozsa, 1994; Uhlmann, 1976; Nielsen & Chuang, 2010) (Bradley, 1997)

Evidence threshold: unless stated otherwise, the practical effect floor  $|\Delta| \geq 0.01$  is applied to the primary KPI  $\Delta\text{AUC}(F_{\text{uncond}})$ ; secondary metrics (F\_cond, leakage, thresholds) are reported to explain mechanism and tradeoffs. (Bradley, 1997)

## 2.5 Leakage-aware cost proxy

To translate tradeoffs into a simple system-level score, define a leakage-aware score  $S(\alpha) = \text{AUC}(F_{\text{uncond}}) - \alpha \cdot \text{AUC}(L)$ , where  $\alpha \geq 0$  encodes the application's relative cost of leakage. I report  $\Delta S(\alpha) = S_{\text{REAL}}(\alpha) - S_{\text{NULL}}(\alpha) = \Delta\text{AUC}(F_{\text{uncond}}) - \alpha \cdot \Delta\text{AUC}(L)$ .

Interpretation of  $\alpha$ :  $\alpha$  sets how costly leakage is relative to unconditional retention. For example,  $\alpha \approx 1$  treats a +0.01 increase in leakage AUC as canceling a +0.01 gain in  $\text{AUC}(F_{\text{uncond}})$  in the combined score.

When  $\Delta\text{AUC}(L) > 0$  (REAL leaks more), the crossover  $\alpha^* = \Delta\text{AUC}(F_{\text{uncond}}) / \Delta\text{AUC}(L)$  gives the leakage-cost level above which  $\Delta S$  becomes negative. When  $\Delta\text{AUC}(L) < 0$  and  $\Delta\text{AUC}(F_{\text{uncond}}) > 0$ ,  $\Delta S(\alpha)$  is favorable for any  $\alpha \geq 0$ .  $\Delta S(\lambda) = S_{\text{REAL}}(\lambda) - S_{\text{NULL}}(\lambda)$

Choosing  $\alpha$  in practice depends on the application: if leakage events are effectively fatal (discard, reset, or costly recovery), use a larger  $\alpha$ ; if small leakage is tolerable, use a smaller  $\alpha$ . Reporting  $\alpha^*$  lets readers map conclusions to their own leakage-cost assumptions.

$$\lambda^* \approx \frac{\Delta\text{AUC}(F_{\text{uncond}})}{\Delta\text{AUC}(L)} (\Delta\text{AUC}(L) > 0).$$

## 3. Results

### 3.1 Amplitude damping: robust “coherent-but-leaky” tradeoff

Under amp\_damp with  $\gamma_1=0.10$  ( $\gamma\phi=0.01$ ), subset=3, basis=noise\_diag, rand64, REAL shows per-batch deltas (REAL–NULL):

- $\Delta\text{AUC}(F_{\text{uncond}}) \in [-0.021, -0.016]$   $\Delta\text{AUC}(F_{\text{uncond}}) = [-0.021, -0.016, -0.017]$  (batches 1–3).
- $\Delta\text{AUC}(F_{\text{cond}}) \in [+0.027, +0.030]$   $\Delta\text{AUC}(F_{\text{cond}}) = [+0.027, +0.030, +0.027]$ .
- $\Delta\text{AUC}(L) \in [+0.032, +0.047]$   $\Delta\text{AUC}(L) = [+0.047, +0.041, +0.032]$ .

v23\_ampd\_g10\_subset3\_noise\_diag\_p400\_rand64.txt

Reducing damping strength scales the effect down without changing its sign structure:

- $\gamma_1=0.05$ :  $\Delta\text{AUC}(F_{\text{uncond}}) = [-0.011, -0.014, -0.012]$ , while  $\Delta\text{AUC}(F_{\text{cond}}) = [+0.015, +0.014, +0.014]$ , and  $\Delta\text{AUC}(L) = [+0.026, +0.026, +0.020]$ .  $\Delta\text{AUC}(F_{\text{uncond}}) \approx -0.011 - 0.014 \Delta\text{AUC}(F_{\text{cond}}) \approx +0.014 + 0.015 \Delta\text{AUC}(L) \approx +0.020 + 0.026$

v23\_ampd\_g05\_subset3\_noise\_diag\_p400\_rand64.txt

- $\gamma_1=0.02$ : effects drop below the practical floor:  $\Delta\text{AUC}(F_{\text{uncond}}) = [-0.005, -0.006, -0.005]$ ,  $\Delta\text{AUC}(F_{\text{cond}}) = [+0.006, +0.006, +0.006]$ , and  $\Delta\text{AUC}(L) = [+0.011, +0.012, +0.009]$ .  $\Delta\text{AUC}(F_{\text{cond}}) \approx +0.006 \Delta\text{AUC}(F_{\text{uncond}}) \approx -0.005$

v23\_ampd\_g02\_subset3\_noise\_diag\_p400\_rand64.txt

**Interpretation:** amp\_damp runs consistently support a regime where REAL improves “within-subspace quality given survival” but pays with leakage and worse unconditional retention.

### 3.2 Dephasing-only at $\gamma\phi=0.10$ : net unconditional advantage

Switching to dephasing-only with  $\gamma\phi=0.10$ ,  $\gamma_1=0.00$ , subset=3, rand64, basis=noise\_diag yields: Note that for leakage, negative  $\Delta\text{AUC}(L)$  is favorable.

This is a net gain despite higher leakage (positive  $\Delta\text{AUC}(L)$ ); the improvement is therefore driven primarily by in-subspace quality ( $F_{\text{cond}}$ ) rather than reduced leakage.

- $\Delta\text{AUC}(F_{\text{uncond}}) = +0.015, +0.019, +0.028$   $\Delta\text{AUC}(F_{\text{uncond}}) = [+0.015, +0.019, +0.028]$  across batches,
- $\Delta\text{AUC}(F_{\text{cond}}) \approx +0.067 \Delta\text{AUC}(F_{\text{cond}}) = [+0.071, +0.069, +0.067], +0.071$
- $\Delta\text{AUC}(L) \approx +0.022 \Delta\text{AUC}(L) = [+0.035, +0.037, +0.022], +0.037$

v23\_deph\_gphi10\_subset3\_noise\_diag\_p400\_rand64.txt

This net unconditional advantage is **basis-robust** (also present with basis=eigen).

v23\_deph\_gphi10\_subset3\_eigen\_p400\_rand64.txt

It is also **noise-localization robust**: with basis=eigen and rand64 at  $\gamma\phi=0.10$ , positive  $\Delta\text{AUC}(F_{\text{uncond}})$  is observed for subsets 0, 1, 2, and 3. For example:

For a complete batch-by-batch listing across all subsets, see Table 1 and Supplementary Note S1; the examples below illustrate the pattern.

- subset=0 (batch 1):  $\Delta\text{AUC}(F_{\text{uncond}}) = +0.014$ ,  $\Delta\text{AUC}(F_{\text{cond}}) = +0.070$ ,  
 $\Delta\text{AUC}(L) = +0.031$ .  $\Delta\text{AUC}(F_{\text{uncond}}) = +0.014 \Delta\text{AUC}(F_{\text{cond}}) = +0.070 \Delta\text{AUC}(L) = +0.031$

v23\_deph\_gphi10\_subset0\_eigen\_p400\_rand64.txt

- subset=2 (all batches):  $\Delta\text{AUC}(F_{\text{uncond}}) = [+0.018, +0.023, +0.021]$ ,  $\Delta\text{AUC}(F_{\text{cond}}) = [+0.069, +0.071, +0.074]$ , and  $\Delta\text{AUC}(L) = [+0.032, +0.031, +0.032]$ .  $\Delta\text{AUC}(F_{\text{uncond}}) = +0.018, +0.023, +0.021 \Delta\text{AUC}(F_{\text{cond}})$

v23\_deph\_gphi10\_subset2\_eigen\_p400\_rand64.txt

- subset=1 (all batches):  $\Delta\text{AUC}(F_{\text{uncond}}) = [+0.016, +0.015, +0.021]$ ,  $\Delta\text{AUC}(F_{\text{cond}}) = [+0.068, +0.072, +0.065]$ , and  $\Delta\text{AUC}(L) = [+0.023, +0.035, +0.035]$ .  $\Delta\text{AUC}(F_{\text{uncond}}) = +0.016, +0.015, +0.021$

v23\_deph\_gphi10\_subset1\_eigen\_p400\_rand64.txt

**Interpretation:** under sufficiently strong dephasing, REAL becomes net-beneficial even under unconditional metrics, and this finding survives key robustness checks (probe ensemble, basis, and localization).

### 3.3 Dephasing strength sweep: crossover behavior

At subset=3, basis=eigen, rand64:

- $\gamma\phi=0.05$  yields no net unconditional advantage:  $\Delta\text{AUC}(F_{\text{uncond}}) = [-0.006, -0.004, -0.002]$  across batches, despite  $\Delta\text{AUC}(F_{\text{cond}}) = [+0.036, +0.034, +0.033]$  and  $\Delta\text{AUC}(L) = [+0.039, +0.039, +0.034]$ .  $\Delta\text{AUC}(F_{\text{uncond}}) = -0.006, -0.002, -0.002$

v23\_deph\_gphi05\_subset3\_eigen\_p400\_rand64.txt

- $\gamma\phi=0.20$  yields a strong net unconditional advantage, and notably reduced leakage relative to NULL:  $\Delta\text{AUC}(F_{\text{uncond}}) = [+0.084, +0.105, +0.116]$ ,  $\Delta\text{AUC}(F_{\text{cond}}) = [+0.153, +0.147, +0.144]$ , and  $\Delta\text{AUC}(L) = [-0.017, -0.028, -0.048]$ .  $\Delta\text{AUC}(F_{\text{uncond}}) = +0.084, +0.105, +0.116 \Delta\text{AUC}(F_{\text{cond}}) = +0.153, +0.147, +0.144 \Delta\text{AUC}(L) = -0.017, -0.028, -0.048$

v23\_deph\_gphi20\_subset3\_eigen\_p400\_rand64.txt

**Interpretation:** the sign of the unconditional advantage depends on dephasing strength; the sweep indicates a crossover between  $\sim 0.05$  and  $0.10$ , and a markedly stronger regime at  $0.20$ .

### 3.4 Mixed noise bridge (“both”): advantage survives, tradeoff reappears in thresholds

For combined noise (both) with  $\gamma\phi=0.10$  and  $\gamma_t=0.02$  at subset=3, basis=eigen, rand64:

- $\Delta\text{AUC}(F_{\text{uncond}}) = +0.014, +0.020, +0.027$   $\Delta\text{AUC}(F_{\text{uncond}}) = [+0.014, +0.020]$  (batches 1–2; batch 3 omitted),

Scope note: mixed-noise conclusions are preliminary in this draft because batch 3 and endpoint quantiles for “both” are not reported here (formatting artifact); the reported pattern is based on batches 1–2 only.

- $\Delta\text{AUC}(F_{\text{cond}}) = +0.070, +0.073, +0.059$
  - $\Delta\text{AUC}(L) = +0.039, +0.034, +0.034$
- $\Delta\text{AUC}(F_{\text{cond}}) = [+0.070, +0.073]$ ,  
 $\Delta\text{AUC}(L) = [+0.039, +0.034]$ , with  $\Delta t_{F90\_med} = -0.21$  and  
 $\Delta t_{L10\_med} = -0.21$  (REAL crosses earlier).

v23\_both\_gphi10\_g02\_subset3\_eigen\_p400\_rand64.txt

However, median threshold deltas are negative ( $\Delta t_{F90\_med} = -0.21$  and  $\Delta t_{L10\_med} = -0.21$ ), indicating that REAL crosses the failure thresholds earlier despite positive AUC gains. $t_{F90}t_{L10}$

v23\_both\_gphi10\_g02\_subset3\_eigen\_p400\_rand64.txt

## 4. Leakage-aware interpretation via cost proxy

Using  $\Delta S(\alpha) = \Delta\text{AUC}(F_{\text{uncond}}) - \alpha \cdot \Delta\text{AUC}(L)$ :  
 $S(\lambda) = \text{AUC}(F_{\text{uncond}}) - \lambda \text{AUC}(L)$

- $\gamma\phi=0.05$ : since  $\Delta\text{AUC}(F_{\text{uncond}}) < 0$  and  $\Delta\text{AUC}(L) > 0$ ,  $\Delta S(\alpha) < 0$  for all  $\alpha \geq 0$  (REAL is unfavorable under this cost model). $\Delta\text{AUC}(F_{\text{uncond}}) < 0$   
 $\Delta\text{AUC}(L) > 0$  $\Delta S(\lambda) < 0$   
 $\lambda \geq 0$

v23\_deph\_gphi05\_subset3\_eigen\_p400\_rand64.txt

- $\gamma\phi=0.10$ : with  $\Delta\text{AUC}(F_{\text{uncond}}) > 0$  and  $\Delta\text{AUC}(L) > 0$ , the crossover lies at  $\alpha^* \approx 0.43-1.27$  across batches (median  $\approx 0.66$ ); REAL is favorable when leakage is not “too expensive” ( $\alpha < \alpha^*$ ). $\Delta\text{AUC}(F_{\text{uncond}}) \approx +0.02\Delta\text{AUC}(L) \approx +0.03\lambda^* \sim 0.5$  —  $0.7$
- $\gamma\phi=0.20$ : since  $\Delta\text{AUC}(F_{\text{uncond}}) > 0$  and  $\Delta\text{AUC}(L) < 0$ ,  $\Delta S(\alpha) > 0$  for any  $\alpha \geq 0$ ; here REAL improves unconditional retention while also reducing leakage. $\Delta\text{AUC}(L) < 0$  $\Delta S(\lambda) > 0$

v23\_deph\_gphi20\_subset3\_eigen\_p400\_rand64.txt

- both ( $\gamma\phi=0.10$ ,  $\gamma_1=0.02$ ):  $\Delta\text{AUC}(F_{\text{uncond}}) > 0$  and  $\Delta\text{AUC}(L) > 0$  in the reported batches, giving  $\alpha^* \approx 0.36-0.59$ ; early-time threshold penalties (negative  $\Delta t$ ) indicate caution when applications emphasize time-to-failure rather than integrated AUC. $\lambda^*$

v23\_both\_gphi10\_g02\_subset3\_eigen\_p400\_rand64.txt

## 5. Discussion

The experiments isolate a clean phenomenon: whether REAL is “net beneficial” depends strongly on the noise channel. Under amplitude damping, REAL repeatedly exhibits a coherent-but-leaky pattern, consistent across probe coverage and damping strength scaling. Under dephasing-only noise, the pattern changes: at

$\gamma\phi=0.10$  REAL becomes net beneficial in unconditional AUC, and at  $\gamma\phi=0.20$  the advantage becomes large and simultaneously reduces leakage relative to NULL.

The combination of (i) matched NULL control, (ii) manifold probing, (iii) basis robustness, (iv) full single-qubit localization sweeps, and (v) a dephasing-strength crossover provides a compact but stringent evidentiary basis. The cost proxy  $S(\lambda)$  clarifies how “goodness” depends on the application’s leakage penalty.

## 1) Box: How to use this in practice

### Practical use of this framework (engineering workflow).

The central contribution of this work is a leakage-aware, matched-control evaluation protocol for open-system retention studies. In practical benchmarking or design selection (e.g., comparing two encodings, control schedules, or Hamiltonian families), I recommend the following workflow:

1. **Define a matched control (NULL).** Construct a spectrum-matched or otherwise constraint-matched NULL baseline to eliminate trivial advantages unrelated to the hypothesized structural effect.
2. **Evaluate three metrics jointly: unconditional retention  $F_{\text{uncond}}(t)$  (primary KPI), conditional retention  $F_{\text{cond}}(t)$  (mechanistic diagnostic), and leakage  $L(t)$  (critical failure channel).**  $F_{\text{uncond}}, F_{\text{cond}}, L$
3. **Use manifold coverage (rand64) for headline claims.** Treat small canonical probe sets (e.g., ZX) as diagnostics only, since direction-dependent effects can be missed by sparse probing.
4. **Require replication across batches. Treat each batch as an independent replicate. Enforce a pre-registered decision rule (e.g., consistent sign in  $\geq 2/3$  batches and a practical floor  $|\Delta\text{AUC}(F_{\text{uncond}})| \geq 0.01$ ).**  $|\Delta\text{AUC}(F_{\text{uncond}})| \geq 0.01$
5. **Interpret tradeoffs explicitly.** If  $\Delta\text{AUC}(F_{\text{cond}}) > 0$  while  $\Delta\text{AUC}(F_{\text{uncond}}) \leq 0$  and  $\Delta\text{AUC}(L) > 0$ , the effect is best described as coherent-but-leaky rather than a practical advantage.  $\Delta\text{AUC}(F_{\text{cond}}) > 0, \Delta\text{AUC}(L) > 0, \Delta\text{AUC}(F_{\text{uncond}}) \leq 0$
6. **If needed, map results to application cost via  $\alpha$  (and  $\alpha^*$ ).** Use the leakage-aware cost proxy to determine whether fidelity gains remain favorable once leakage penalties reflect the target application.  $\alpha$

## 2) Decision tree: Go/No-Go for “REAL advantage”

### Decision rule for practical adoption (go/no-go).

Use the following checklist when deciding whether REAL shows a practical advantage over NULL in an open-system regime.

#### Step 1 (probe regime):

- Are headline results computed with **rand64** manifold coverage?
  - If **No**, treat findings as diagnostic only.

#### Step 2 (primary KPI):

- Is  $\Delta\text{AUC}(F_{\text{uncond}}) > 0$ ?  $\Delta\text{AUC}(F_{\text{uncond}}) = \text{AUC}_{\text{REAL}}(F_{\text{uncond}}) - \text{AUC}_{\text{NULL}}(F_{\text{uncond}})$ 
  - If **No**, conclude **no net retention advantage** (unless the application explicitly uses conditional/post-selected operation).

#### Step 3 (replication):

- Is the sign of  $\Delta\text{AUC}(F_{\text{uncond}})$  consistent in  $\geq 2/3$  batches and above the practical floor  $|\Delta\text{AUC}(F_{\text{uncond}})| \geq 0.01$ ?  $\Delta\text{AUC}(F_{\text{uncond}}) | \Delta | \geq 0.01$ 
  - If **No**, label the effect **inconclusive** and increase sampling or refine the regime.

#### Step 4 (leakage check):

- Is leakage favorable, i.e.  $\Delta\text{AUC}(L) < 0$  (negative is favorable)?  $\Delta\text{AUC}(L) < 0$ 
  - If **Yes**, classify the result as **strong practical evidence**.
  - If **No** (leakage worse), proceed to Step 5.

#### Step 5 (mechanistic diagnosis):

- Is  $\Delta\text{AUC}(F_{\text{cond}}) > 0$  while  $\Delta\text{AUC}(L) > 0$ ?  $\Delta\text{AUC}(F_{\text{cond}}) > 0 \Delta\text{AUC}(L) > 0$ 
  - If **Yes**, classify as coherent-but-leaky; use cost proxy  $\alpha^*$  to determine whether fidelity gains can justify leakage. $\alpha$
  - If **No**, the regime likely reflects mixed mechanisms; report both AUC and threshold metrics and treat the conclusion as **application-dependent**.

#### Step 6 (application mapping):

- If an application has strong leakage intolerance (e.g., error-correction compatibility), require  $\Delta\text{AUC}(L) \leq 0$  and/or  $\alpha < \alpha^*$ .  $\Delta\text{AUC}(L) \leq 0 \alpha < \alpha^*$
- If an application supports erasure/post-selection, conditional improvements may remain valuable even when unconditional gains are modest.

### 3) Table: Common failure modes (and how this paper addresses them)

**Table X. Common evaluation failure modes and mitigations used in this work.**

Failure mode	What it looks like in metrics	Why it misleads	Mitigation in this paper
<b>Coherent-but-leaky</b>	$\Delta\text{AUC}(F_{\text{cond}}) > 0$ , $\Delta\text{AUC}(F_{\text{cond}}) > 0$ but $\Delta\text{AUC}(F_{\text{uncond}}) \leq 0$ and $\Delta\text{AUC}(L) > 0$ , $\Delta\text{AUC}(L) > 0$ , $\Delta\text{AUC}(F_{\text{uncond}}) \leq 0$	Conditional quality improves only when the state remains in-subspace; practical retention does not improve	Joint reporting of $F_{\text{uncond}}$ , $F_{\text{cond}}$ , $L$ ; leakage-aware interpretation and cost proxy $\alpha, \alpha^* F_{\text{uncond}}, F_{\text{cond}}, L\alpha$
<b>Probe blind spot (sparse probes)</b>	Effect appears under ZX but disappears under rand64 (or vice versa)	Direction-dependent effects can be missed with a small probe set	Treat ZX as diagnostic; use rand64 manifold coverage for headline claims
<b>Cherry-picking / selection bias</b>	Strong effect only after filtering candidates; weak or inconsistent otherwise	Selection on the same population used for evaluation inflates effects	Explicit run registry and batch replication; disclose selection procedures (or disable selection for headline runs)
<b>Basis artefact</b>	Advantage appears only in one basis (eigen vs noise-diagonal)	Coordinate choice can hide/induce apparent structure	Basis robustness checks (eigen vs noise_diag) as part of evaluation
<b>Single-run overinterpretation</b>	Large $\Delta$ in one run; sign flips across batches	Random sampling fluctuations dominate	Treat batches as independent replicates; require $\geq 2/3$ sign-consistency and effect floor
<b>Metric monoculture</b>	Large AUC gain but thresholds worsen (or vice versa)	AUC averages across time; threshold captures first critical failure	Report both AUC and threshold metrics; interpret tradeoffs explicitly
<b>Endpoint vs trajectory confusion</b>	Endpoint quantiles suggest improvement while AUC suggests degradation (or vice versa)	Curves may cross; endpoint misses mid-trajectory behavior	Use AUC for trajectory summary; annotate cases where quantiles and AUC diverge
<b>Unmatched baselines</b>	REAL beats a “random” NULL by a lot	Trivial differences (e.g., spectrum) can explain the win	Spectrum-matched NULL controls to isolate structural effects

## 6. Limitations

This work is a controlled pilot in an n=4 testbed with Markovian Lindblad noise. Internal validity is strengthened by matched NULL controls, batch replication, explicit leakage tracking, basis checks, and single-qubit noise localization sweeps. However, external validity depends on scaling to larger encodings and on testing richer noise models (correlated/non-Markovian and hardware-specific channels). (Rivas, Huelga, & Plenio, 2014) A further limitation is observable-regime dependence: coarse ZX probing can undersample the logical manifold and may mask direction-dependent effects; for this reason, headline claims are restricted to the manifold-probe regime (rand64), while ZX results are treated as diagnostics. (Marsaglia, 1972) Finally, when stable\_pool selection is used, selection bias is a potential loophole unless selection and evaluation are separated; I therefore report stable\_pool explicitly per run and do not rely on stable\_pool-restricted runs as primary evidence unless accompanied by holdout controls.

Not tested / not addressed in this draft:

- Scaling beyond n=4 (no asymptotic claims).
- Non-Markovian or correlated noise (only Markovian Lindblad channels are modeled). (Rivas, Huelga, & Plenio, 2014)
- Hardware-calibrated noise parameters or device-specific validation (simulation-only).
- Mixed-noise completeness (batch 3 and endpoint quantiles for “both” are pending due to a formatting artifact).

## 7. Conclusion

I provide a matched-null benchmark protocol for logical-subspace retention under open-system noise with explicit leakage tracking and manifold probing. The results show (i) a robust coherent-but-leaky tradeoff under amplitude damping, (ii) a dephasing-dependent crossover where REAL becomes net beneficial under unconditional AUC at  $\gamma\phi=0.10$  and strongly so at  $\gamma\phi=0.20$ , and (iii) persistence of the unconditional AUC advantage under mixed noise, albeit with reintroduced threshold penalties. A leakage-aware score  $S(\lambda)$  offers a compact system-level interpretation.

***"The results demonstrate that evaluating logical constructions without explicit leakage accounting risks overestimating practical performance, particularly in energy-relaxation-dominated environments."***

### **(Draft) Figure/Table plan (minimal)**

- Table 1: Summary of deltas for each regime (amp\_damp g1=0.10/0.05/0.02; dephasing gphi=0.05/0.10/0.20; both).
- Fig 2: Crossover plot:  $\Delta\text{AUC}(F_{\text{uncond}})$  vs  $\gamma\phi$  (0.05, 0.10, 0.20).



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## 9. Tables

Table 1

**Table 1 — Caption**

Table 1. Batch-level AUC deltas ( $\Delta = \text{REAL} - \text{NULL}$ ) for open-system retention: unconditional fidelity AUC(F\_uncond), conditional fidelity AUC(F\_cond), and leakage AUC(L), reported for each batch (3 batches) together with median threshold-time deltas  $\Delta t_{\text{F90}}$  and  $\Delta t_{\text{L10}}$  where available. Positive  $\Delta \text{AUC}$  for fidelity is favorable; negative  $\Delta \text{AUC}(L)$  is favorable (reduced leakage). All runs use  $n=4$ ,  $t_{\text{max}}=5.0$ ,  $t_{\text{steps}}=25$ ,  $\text{pairs\_per\_batch}=400$ , and the probe-state regime specified in the run registry.

**amp\_damp ( $\gamma\phi=0.01, \gamma_1=0.02$ ), subset=3, basis=noise\_diag**

Batch	$\Delta \text{AUC}(\text{fid\_uncond})$	$\Delta \text{AUC}(\text{fid\_cond})$	$\Delta \text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	-0.005	+0.006	+0.011	+0.00	+0.00
2	-0.006	+0.006	+0.012	+0.00	+0.00
3	-0.005	+0.006	+0.009	+0.00	+0.00

**amp\_damp ( $\gamma\phi=0.01, \gamma_1=0.05$ ), subset=3, basis=noise\_diag**

Batch	$\Delta \text{AUC}(\text{fid\_uncond})$	$\Delta \text{AUC}(\text{fid\_cond})$	$\Delta \text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	-0.011	+0.015	+0.026	-0.20	-0.42
2	-0.014	+0.014	+0.026	-0.20	-0.42
3	-0.012	+0.014	+0.020	-0.20	-0.42

**amp\_damp ( $\gamma\phi=0.01, \gamma_1=0.10$ ), subset=3, basis=noise\_diag**

Batch	$\Delta \text{AUC}(\text{fid\_uncond})$	$\Delta \text{AUC}(\text{fid\_cond})$	$\Delta \text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	-0.021	+0.027	+0.047	-0.42	-0.42
2	-0.016	+0.030	+0.041	-0.42	-0.42
3	-0.017	+0.027	+0.032	-0.42	-0.42

**dephasing ( $\gamma\phi=0.05, \gamma_1=0.00$ ), subset=3, basis=eigen**

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	-0.006	+0.036	+0.039	-0.21	-0.21
2	-0.004	+0.034	+0.039	-0.21	-0.21
3	-0.002	+0.033	+0.034	-0.21	-0.21

dephasing ( $\gamma\phi=0.10$ ,  $\gamma_1=0.00$ ), subset=3, basis=noise\_diag

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.015	+0.071	+0.035	+0.00	-0.10
2	+0.019	+0.069	+0.037	+0.00	+0.00
3	+0.028	+0.067	+0.022	+0.00	+0.00

dephasing ( $\gamma\phi=0.10$ ,  $\gamma_1=0.00$ ), subset=3, basis=eigen

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.017	+0.068	+0.037	+0.00	+0.00
2	+0.020	+0.073	+0.037	+0.00	+0.00
3	+0.029	+0.070	+0.023	+0.00	+0.00

dephasing ( $\gamma\phi=0.10$ ,  $\gamma_1=0.00$ ), subset=0, basis=eigen

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.014	+0.070	+0.031	+0.00	+0.00
2	+0.029	+0.063	+0.040	+0.00	+0.00
3	+0.023	+0.073	+0.026	+0.00	+0.00

dephasing ( $\gamma\phi=0.10$ ,  $\gamma_1=0.00$ ), subset=1, basis=eigen

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.016	+0.068	+0.023	+0.00	+0.00
2	+0.015	+0.072	+0.035	+0.00	+0.00
3	+0.021	+0.065	+0.035	+0.00	+0.00

dephasing ( $\gamma\phi=0.10$ ,  $\gamma_1=0.00$ ), subset=2, basis=eigen

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.018	+0.069	+0.032	+0.00	+0.00
2	+0.023	+0.071	+0.031	+0.00	+0.00
3	+0.021	+0.074	+0.032	+0.00	+0.00

dephasing ( $\gamma\phi=0.20$ ,  $\gamma_1=0.00$ ), subset=3, basis=eigen

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.084	+0.153	-0.017	+0.00	+0.00
2	+0.105	+0.147	-0.028	+0.00	+0.00
3	+0.116	+0.144	-0.048	+0.00	+0.00

both ( $\gamma\phi=0.10$ ,  $\gamma_1=0.02$ ), subset=3, basis=eigen (Note: Batch 3 omitted here due to a formatting artifact in the extracted log; batches 1–2 are reported.)

Batch	$\Delta\text{AUC}(\text{fid\_uncond})$	$\Delta\text{AUC}(\text{fid\_cond})$	$\Delta\text{AUC}(\text{leak})$	$\Delta t_{\text{F90\_med}}$	$\Delta t_{\text{L10\_med}}$
1	+0.014	+0.070	+0.039	-0.21	-0.21
2	+0.020	+0.073	+0.034	-0.21	-0.21

### Supplementary Note S1 (Batch-Level AUC Deltas and Robustness Criteria)

Table 1 reports batch-level deltas ( $\Delta = \text{REAL} - \text{NULL}$ ) in AUC summary metrics for unconditional fidelity  $F_{\text{uncond}}(t)$ , conditional fidelity  $F_{\text{cond}}(t)$ , and leakage  $L(t)$ , with AUC computed over  $t \in [0, 5]$ . Positive  $\Delta\text{AUC}$  for fidelity indicates net improvement; for leakage, negative  $\Delta\text{AUC}(L)$  is favorable (REAL leaks less). Because  $F_{\text{cond}}$  conditions on survival in the logical subspace, it can improve even when unconditional performance worsens; Table 1 therefore makes it explicit when regimes are coherent-but-leaky ( $\Delta\text{AUC}(F_{\text{cond}}) > 0$  while  $\Delta\text{AUC}(F_{\text{uncond}}) < 0$  and  $\Delta\text{AUC}(L) > 0$ ). Batch reporting supports the evidence criteria: effects must be consistent in sign in  $\geq 2/3$  batches and exceed a practical floor.

## Table S2b — Endpoint quantiles ved t=5.0 (q10 / q50 / q90)

**Table S2b — Caption**

Table S2b. Endpoint quantiles and threshold medians for open-system logical retention. Reported are q10/q50/q90 at t=5.0 for F\_uncond and leakage L for REAL and NULL, plus (where applicable) median threshold times t\_F90 (time to F\_uncond<0.90) and t\_L10 (time to L>0.10). Threshold deltas are defined as  $\Delta t = t_{\text{REAL}} - t_{\text{NULL}}$ ; negative  $\Delta t$  means REAL crosses earlier (worse on that threshold).

Run	F_uncond(t=5) REAL	F_uncond(t=5) NULL	Leak(t=5) REAL	Leak(t=5) NULL
amp_damp ( $\gamma\phi=0.01$ , $\gamma_1=0.10$ ), subset={3}, basis=noise_diag	0.777/0.792/0.837	0.751/0.797/0.841	0.117/0.202/0.219	0.145/0.189/0.235
amp_damp ( $\gamma\phi=0.01$ , $\gamma_1=0.05$ ), subset={3}, basis=noise_diag	0.879/0.887/0.913	0.865/0.891/0.916	0.061/0.110/0.119	0.077/0.101/0.128
amp_damp ( $\gamma\phi=0.01$ , $\gamma_1=0.02$ ), subset={3}, basis=noise_diag	0.948/0.952/0.964	0.943/0.955/0.965	0.026/0.047/0.050	0.032/0.042/0.053
dephasing ( $\gamma\phi=0.10$ , $\gamma_1=0.00$ ), subset={3}, basis=noise_diag	0.645/0.671/0.752	0.638/0.651/0.674	0.152/0.322/0.354	0.290/0.327/0.351
dephasing ( $\gamma\phi=0.10$ , $\gamma_1=0.00$ ), subset={0}, basis=eigen	0.645/0.672/0.761	0.638/0.651/0.674	0.131/0.323/0.354	0.289/0.327/0.350
dephasing ( $\gamma\phi=0.10$ , $\gamma_1=0.00$ ), subset={1}, basis=eigen	0.645/0.672/0.758	0.638/0.651/0.674	0.150/0.323/0.355	0.289/0.327/0.350
dephasing ( $\gamma\phi=0.10$ , $\gamma_1=0.00$ ), subset={2}, basis=eigen	0.645/0.671/0.758	0.638/0.651/0.674	0.130/0.323/0.354	0.289/0.327/0.350
dephasing ( $\gamma\phi=0.10$ , $\gamma_1=0.00$ ), subset={3}, basis=eigen	0.646/0.673/0.768	0.638/0.651/0.674	0.114/0.322/0.354	0.290/0.327/0.351

Run	F_uncond(t=5) REAL	F_uncond(t=5) NULL	Leak(t=5) REAL	Leak(t=5) NULL
dephasing ( $\gamma\phi=0.05$ , $\gamma_1=0.00$ ), subset={3}, basis=eigen	0.792/0.818/0.869	0.792/0.803/0.836	0.075/0.188/0.204	0.142/0.189/0.205
dephasing ( $\gamma\phi=0.20$ , $\gamma_1=0.00$ ), subset={3}, basis=eigen	0.467/0.520/0.609	0.440/0.468/0.499	0.451/0.461/0.532	0.447/0.501/0.538

## Supplementary Note S2 (Thresholds and Endpoint Quantiles)

Table S2b reports median threshold times  $t_{\_F90}$  (time until  $F_{\_uncond}<0.90$ ) and  $t_{\_L10}$  (time until  $L>0.10$ ) for REAL and the spectrum-matched NULL. Threshold deltas are defined as  $\Delta t = t_{\_REAL} - t_{\_NULL}$ , so negative  $\Delta t$  means REAL crosses earlier (worse on that specific threshold metric). Threshold metrics capture first-crossing behavior, whereas AUC integrates over the full interval  $t \in [0,5]$ ; therefore, AUC and thresholds can disagree in sign when trajectories differ in shape (e.g., early penalty but later improvement). Table S2b complements this by reporting endpoint quantiles ( $q_{10}/q_{50}/q_{90}$ ) at  $t=5.0$ , indicating whether effects are broad shifts or tail-dominated. Endpoint quantiles for the mixed-noise (“both”) run are omitted in this revision due to the same formatting artifact noted in Table 1.

Example ( $\gamma\phi=0.10$ , subset=3): Table 1 shows  $\Delta AUC(L)>0$  (REAL leaks more on average over  $t \in [0,5]$ ), while Table S2b may still show smaller early-time leakage quantiles for REAL (e.g., lower  $q_{10}$ ). This is consistent if REAL starts with lower leakage but rises faster later, yielding a larger time-integrated leakage even when early endpoint quantiles look favourable.

## 10. Code and Data Availability

The simulation framework, run registry, and all raw data files supporting the findings of this study are available on GitHub at <https://github.com/tomstevns/qubits>. This includes the source code for Lindblad evolution, the rand64 probe generation, and the leakage-aware cost-proxy analysis scripts. For reproducibility, a command registry (command\_registry\_v23\_recent.docx) listing the most commonly used PowerShell one-liner scripts for executing the Python runs is included in the repository (<https://github.com/tomstevns/qubits/tree/main/documentation>).