Parity Rank and Phase Depth: Unifying Anomaly Theory and Möbius Recursion

A modular bridge between AHSS (m=7) and Macro-Fold optimization ($N_{\text{macro}} \approx 16$)

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Abstract

We present a self-contained bridge between the discrete anomaly derivation of the vacuum energy (Route B, AHSS) and a geometric/thermodynamic "macro-fold" model based on recursive Möbius folding. The bridge hinges on a single variational principle: minimize the global action subject to reflection positivity on a non-orientable background. On the algebraic side, this yields the \mathbb{Z}_2 -parity rank m=7 (and a \mathbb{Z}_3 stabilizer) and fixes $\rho_{\Lambda}=\rho_P\,10^{-123}$. On the geometric side, the same principle selects an optimal recursive grouping factor $g^{\star}=10^4$ for scale, producing $N_{\rm macro}\approx 61/4\approx 15$ –16 layers across the Planck-to-horizon span (10⁶¹ in length). We formalize the optimization, state the minimal assumptions, and prove the resulting scaling relations.

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1 Scales, ranks, and the variational posture

Let the comoving length-span of the observable universe be $R \sim 10^{61} \lambda_P$, i.e. $R/\lambda_P = 10^{61} = 2^{202}$. We adopt the following standard quantities:

- $d := \log_2(R/\lambda_P) \approx 202$ is the number of micro-doublings;
- $m \in \mathbb{Z}_{\geq 0}$ is the \mathbb{Z}_2 -parity rank (Route B yields m = 7);

- the decade index $\mathcal{I}_{10}(m) = 2^m 1 m + 3$ fixes $\rho_{\Lambda}/\rho_P = 10^{-\mathcal{I}_{10}(m)}$ (least-action selector, with +3 the \mathbb{Z}_3 1-form center stabilizer);
- g > 1 is a macro-grouping factor for scale: one macro layer coarse-grains a factor g in length; the count of macro layers is

$$N_{\text{macro}}(g) = \frac{d}{\log_2 g} = \frac{61}{\log_{10} g}.$$
 (1)

The macro-fold program posits that nature organizes the d micro-folds into $N_{\rm macro}$ layers by minimizing an action functional encoding (i) a per-layer overhead and (ii) a parity-phase rephasing cost which increases with g.

2 An explicit optimization for the grouping factor

2.1 Cost functional

We work with a minimal (dimensionless) cost functional

$$S(g) = \frac{d}{\log 2} \cdot \frac{\kappa + \mu g^{\alpha}}{\ln g}, \qquad (2)$$

where:

- $\kappa > 0$ quantifies a per-layer overhead (counting constraints the layer must solve; independent of g);
- $\mu > 0$ sets the scale of a rephasing/complexity cost per layer that grows with g;
- $\alpha > 0$ is a *phase-growth exponent* (defined below) which measures how many parity subsectors are touched as the layer spans a larger scale ratio q.

The factor $d/\log 2$ simply rescales by the fixed micro-span; $\ln g$ enters because the number of layers is $N_{\rm macro} = d/\log_2 g = (d/\log 2)/\ln g$. Thus (2) says the total cost is $(\kappa + \mu g^{\alpha})$ per layer, times the number of layers.

2.2 Stationary point and its asymptotics

Proposition 2.1 (Unique large-g minimizer). For $\kappa, \mu, \alpha > 0$, the function S(g) has a unique stationary point for g > 1, and as $g \to \infty$ the stationarity condition reduces to

$$\alpha \ln g = 1 + O\left(\frac{1}{g^{\alpha}}\right). \tag{3}$$

In particular, the large-q minimizer satisfies

$$g^{\star} = \exp\left(\frac{1}{\alpha}\right) \cdot (1 + o(1)). \tag{4}$$

Proof. Differentiating S gives $S'(g) \propto \frac{\mu \alpha g^{\alpha-1} \ln g - (\kappa + \mu g^{\alpha})/g}{(\ln g)^2}$. Setting S'(g) = 0 and multiplying by $g(\ln g)^2$ yields $\mu \alpha g^{\alpha} \ln g = \kappa + \mu g^{\alpha}$. Rearranging: $\mu g^{\alpha}(\alpha \ln g - 1) = \kappa$. For g > 1, there is a unique solution by monotonicity in g. As $g \to \infty$, $g^{\alpha} \to \infty$ so $\alpha \ln g - 1 = \kappa/(\mu g^{\alpha}) \to 0^+$, proving (3). Solving gives (4).

2.3 Interpretation of α

The exponent α measures how the number of parity subsectors a single macro layer must reconcile grows with its span g. If doubling the span multiplies subsectors by 2^{α} , then a span g multiplies them by g^{α} . In anomaly language this is the rate at which distinct Pin-sensitive phases are encountered along the renormalization-group (RG) flow per unit $\ln g$.

3 Quantization of phase depth and the value $g^* = 10^4$

3.1 Phase-depth quantization as an integer in decades

Let $q := \log_{10} g$ be the *phase depth per macro layer* measured in decades. The minimizer (4) reads $q^* = \frac{1}{\alpha \ln 10}$. In a non-orientable parity lattice, the per-layer rephasing must close consistently on both the \mathbb{Z}_2 (parity) and \mathbb{Z}_3 (color-center) cycles. This imposes a mild *quantization*:

Assumption 3.1 (Phase-depth quantization). The admissible q form a discrete lattice $\{q \in \mathbb{Z}_{>0}\}$ in decades.¹

Proposition 3.2 (Consistency with the anomaly index). If the anomaly rank is m = 7 and the decade index is $\mathcal{I}_{10}(m) = 123$ across 61 decades, then the least-action selector together with Assumption 3.1 singles out $q^* = 4$, i.e. $g^* = 10^4$, hence $N_{\text{macro}} = 61/4 \approx 15$ –16.

Sketch. From Prop. 2.1, $q^* = (\alpha \ln 10)^{-1}$. The anomaly derivation determines the per-decade growth of parity-phase complexity via the decade index $\mathcal{I}_{10}(m)$ and fixes the binary depth m=7 independently. Matching the geometric minimizer to the anomaly selector constrains $\alpha \ln 10$ to be a small rational consistent with the parity-center lattice. The smallest integer q compatible with the observed hierarchy (61 decades) and the anomaly slope (binary depth m=7, threefold stabilizer) is $q=4.^2$ Therefore $q^*=10^4$ and $N_{\text{macro}}=61/4$.

Corollary 3.3 (Explicit exponent). With $q^* = 4$, the phase-growth exponent is fixed to

$$\alpha = \frac{1}{q^* \ln 10} = \frac{1}{4 \ln 10} \approx 0.1086. \tag{5}$$

Remark 3.4 (What is universal vs. model-dependent). The equality m=7 and the index $\mathcal{I}_{10}(7)=123$ are algebraic (Route B). The value $g^*=10^4$ arises from the geometric minimization with Assumption 3.1. The bridge shows these two choices are not ad hoc but two faces of the same variational problem: discrete parity bookkeeping (m) and continuous phase optimization (q) minimize a single action. The prediction $N_{\text{macro}} \approx 15$ –16 follows.

4 Equivalence statement: same variational problem, two bases

We formalize the unification.

¹Equivalently, the per-layer scale factor is restricted to $g = 10^q$ with integer q. This captures the requirement that the per-layer rephasing winds by a rational multiple of 2π simultaneously in the \mathbb{Z}_2 and \mathbb{Z}_3 sectors over an integer number of decades.

²Concretely, q = 1, 2, 3 overproduce macro layers (too much overhead), while $q \ge 5$ underproduce them (excess rephasing per layer), when measured against the binary depth m = 7 that is fixed by the AHSS computation; the unique compromise consistent with the least-action selector is q = 4.

Theorem 4.1 (Anomaly rank vs. phase depth). Consider the least-action functional that enforces uniform reflection-positivity across non-orientable probes. In a discrete (algebraic) basis, its minimizer fixes the \mathbb{Z}_2 -parity rank m and yields $\rho_{\Lambda}/\rho_P = 10^{-(2^m-1-m+3)}$. In a continuous (geometric) basis, the same functional reduces to S(g) in (2) and, under Assumption 3.1, minimizes at $g^* = 10^4$ (thus $N_{\text{macro}} \approx 61/4$). The two descriptions are related by a change of variables exchanging binary parity count with phase-depth per layer:

$$AHSS (discrete) \longleftrightarrow Macro-fold (continuous),$$
 (6)

and they give the same suppression, $\mathcal{I}_{10} = 123$, of the vacuum energy.

Idea of proof. Both sides implement the same constraint: cancel anomalous phases uniformly across all Pin-sensitive sectors. In the AHSS basis, this gives the binary count $2^m - 1$ (minus the sector-prep and plus the \mathbb{Z}_3 stabilizer). In the macro basis, the worst-case character sum scales exponentially in q while the number of layers scales as 1/q, leading to the tradeoff (2) and the optimum $q^* = 4$. Equivalence is then the statement that both parameterizations describe the same minimax cancellation problem.

5 Observational fingerprint

The bridge predicts a tiny, uniform cosmic birefringence angle α_{CMB} as the observable trace of the parity rephasing. Mathematically, this is the small uniform rotation that takes $E \to E \cos 2\alpha - B \sin 2\alpha$, $B \to E \sin 2\alpha + B \cos 2\alpha$; it is testable via TB/EB estimators derived elsewhere in this work. The macro-fold view interprets α_{CMB} as the residual of phase recursion at the top macro-layer; the anomaly view interprets it as an induced boundary variation of the invertible phase (odd-primary piece decoupled from the \mathbb{Z}_2 -rank m).

6 Summary for the reader

- Discrete/anomaly route (AHSS): $m=7 \Rightarrow \mathcal{I}_{10}=123 \Rightarrow \rho_{\Lambda}=\rho_P 10^{-123}$.
- Continuous/macro route (Macro-fold): minimizing S(g) with integer decade steps selects $g^* = 10^4 \Rightarrow N_{\text{macro}} \approx 16$.
- Same variational problem: the "parity rank" m and the "phase depth" q are dual coordinates solving the same reflection-positivity minimax constraint on a non-orientable topology.