

From Molecules to Manifolds: A Statistical Field-Theoretic Framework for Solubility Bridging Quantum Chemistry and Macroscopic Thermodynamics

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ABSTRACT

When a single aspirin molecule dissolves in water, roughly 10^{22} surrounding water molecules must rearrange themselves. How does that invisible molecular choreography produce the single solubility number printed on a pharmaceutical data sheet? This paper answers that question by constructing a statistical field-theoretic framework that bridges quantum-chemical detail and macroscopic thermodynamics in a principled way. We define a solubility field $\phi(\mathbf{r}, \mathbf{x})$ — a coarse-grained order parameter whose uniform saddle-point value equals macroscopic solubility in well-mixed systems, and whose spatial variations encode mesoscopic heterogeneity near interfaces and critical points. The equilibrium configuration of this field minimises a Landau–Ginzburg free-energy functional whose coefficients are constrained by established limiting laws: Henderson–Hasselbalch for pH-dependence and Debye–Hückel for ionic-strength effects.

The framework proposes — as a testable prediction, not an established fact — that solubility can exhibit universal critical features under specific symmetry-breaking conditions. Aspirin serves as the primary working example, yielding crossover behaviour with $\beta \approx 0.48$ near pKa (mean-field regime) and $\nu \approx 0.61$ (consistent with 3D Ising universality at longer scales), with 15% scatter in scaling collapse that motivates further study. Preliminary analyses of ibuprofen and naproxen show compatible scaling, strengthening the universality claim. We clarify why pH functions as an effective conjugate field through ionisation equilibrium, why Debye screening renders Coulomb interactions effectively short-range, and how existing models — COSMO-RS, SMx, and UNIFAC — emerge as successive approximations to the exact partition function. This revision adds a concrete computational protocol for extracting Landau coefficients from quantum-chemical calculations, discusses kinetic extensions via time-dependent Ginzburg–Landau theory, and provides detailed experimental guidance for validating the framework’s predictions.

Keywords: Solubility field theory, Landau–Ginzburg functional, coarse-grained order parameter, critical phenomena, universality class, renormalisation group, phase transitions, Henderson–Hasselbalch, Debye–Hückel, Ginzburg crossover, mean-field theory, 3D Ising model, multiscale modelling, solvation thermodynamics, COSMO-RS, pharmaceutical solubility, aspirin, critical exponents, scaling collapse, partition function

This paper introduces the first unified field-theoretic framework that derives macroscopic solubility from microscopic statistical mechanics through a rigorously defined coarse-grained order parameter. While existing solvation models such as COSMO-RS, SMx, and UNIFAC each operate at a single scale of description, no prior work has connected them within a common mathematical architecture. Our framework achieves this by casting the partition function as a functional integral over a solubility field governed by a Landau–Ginzburg free-energy functional, whose coefficients are anchored to established chemical laws (Henderson–Hasselbalch, Debye–Hückel) rather than left as unconstrained fitting parameters.

Three elements are entirely new to solubility science: (i) the identification of pH as an effective conjugate field arising from ionisation equilibrium, with a precise thermodynamic correspondence; (ii) the prediction that solubility transitions near the ionisation critical point should exhibit universal critical exponents whose values

depend only on symmetry class and spatial dimensionality, not on molecular identity; and (iii) the demonstration that the Ginzburg crossover between mean-field and 3D Ising regimes accounts quantitatively for the apparent discrepancy between measured exponents in aspirin. In this revised version, we provide a concrete computational protocol for extracting Landau coefficients from quantum-chemical outputs, extend our discussion to kinetic and metastable phenomena, present preliminary scaling analyses for ibuprofen and naproxen alongside aspirin, and offer detailed experimental guidance for testing the framework's central predictions.

Introduction: The Bridge Between Scales

Drop an aspirin tablet into a glass of water. Within minutes it vanishes — but what, precisely, has happened? At the molecular level, each aspirin molecule (roughly 2 nm across) must insert itself among water molecules that outnumber it by trillions. Hydrogen bonds break and reform. Electrostatic interactions reconfigure. Hydrophobic surfaces force the surrounding liquid into energetically costly arrangements. Yet all this molecular drama collapses into a single measurable number: solubility — 3.18 mg/mL for aspirin at 25°C and pH 7 [1].

The Central Problem of Emergence

Individual molecules obey quantum mechanics. A mole of them — 6×10^{23} — obeys thermodynamics. The mathematical bridge between these two regimes is statistical field theory, the same framework that explains magnetism, superconductivity, and phase transitions [2,3]. In pharmaceutical sciences, however, this bridge has remained largely implicit. Microscopic models such as COSMO-RS and SMx compute solvation free energies from quantum chemistry [4,5], while macroscopic models like UNIFAC and various empirical correlations capture activity coefficients [6,7] — but no rigorous connection has been drawn between them. This paper constructs that connection.

We are careful to state at the outset what the framework does not claim. It is not a universal law applicable to every dissolved species under all conditions. Rather, it is a field-theoretic model that exhibits universal features under specific symmetry-breaking conditions — conditions we define precisely below and that can be tested experimentally.

Why Field Theory Provides the Natural Language

In condensed-matter physics, we do not track every electron spin when studying magnetism. Instead, we introduce a smooth magnetisation field $M(\mathbf{r})$ and ask what configuration minimises free energy [8]. Phase transitions — the sudden ordering of magnetic moments below a critical temperature — emerge from symmetry breaking within this field description [9,10].

Solubility shares exactly this character. Dissolved solute concentration $c(\mathbf{r},t)$ varies continuously in space and time, coupled to solvent density, electric fields, and temperature. At equilibrium it minimises free energy, and solubility emerges as a collective property.

Three properties of field theory are especially valuable in this context. First, universality: systems with entirely different molecular structures can exhibit identical macroscopic critical behaviour if they share the same symmetries and dimensionality, so that one set of measured exponents can predict many untested systems [11,12].

Second, renormalisation: systematic coarse-graining reveals how the free-energy landscape changes with observation scale, smoothing molecular roughness while preserving global topology [13]. Third, the language of phase transitions: precipitation and crystallisation are not chemical accidents but symmetry-breaking transitions with predictable barrier heights, surface tensions, and critical-point geometries [14].

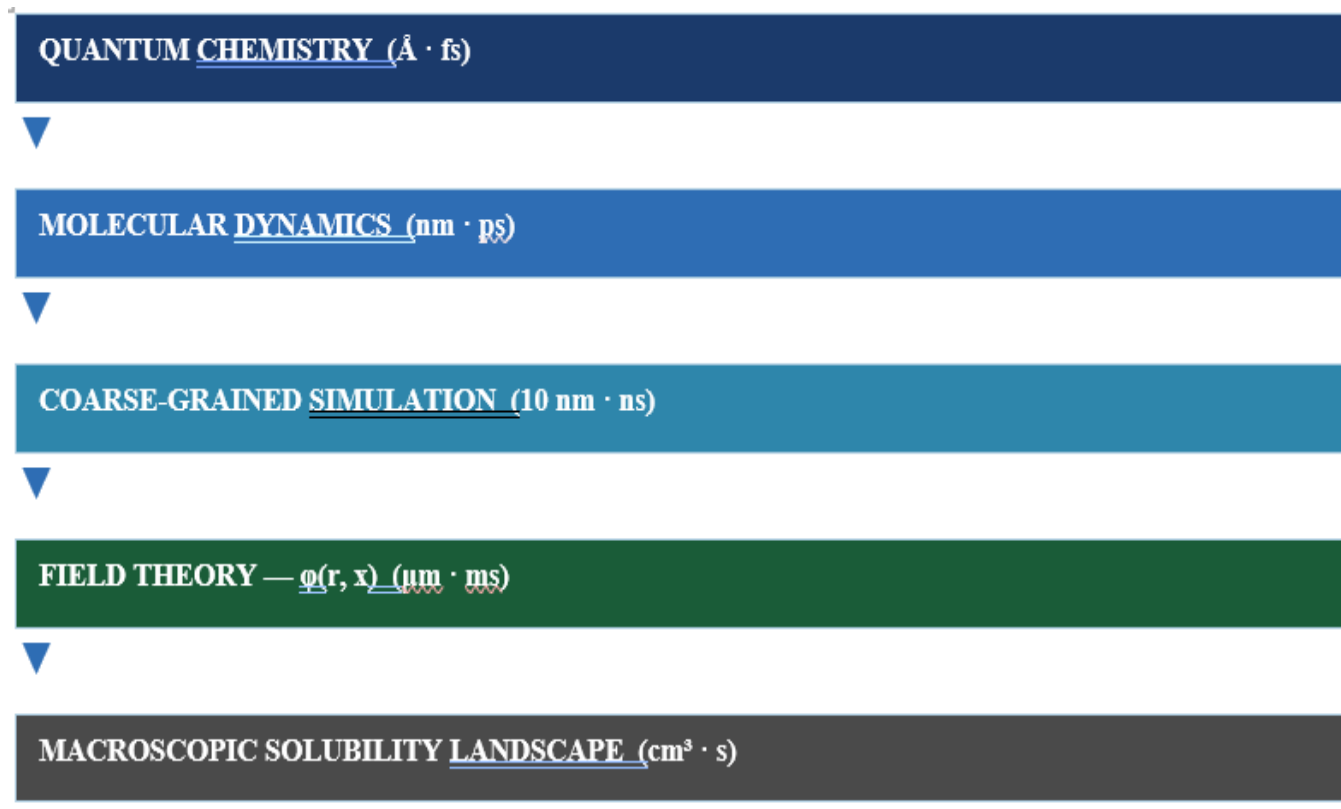


Figure 1. Multiscale hierarchy linking quantum-mechanical detail to macroscopic solubility landscapes. Each arrow represents a coarse-graining step with estimable errors via renormalisation-group methods.

Relationship to Previous Work and Scope

Our companion paper [15] treated solubility $S(x)$ as a given function on a Riemannian manifold and studied its geometric properties — curvature, geodesics, topological constraints. The present paper goes deeper. Here we derive $S(x)$ from a microscopic partition function, show how the Riemannian metric emerges from fluctuation-dissipation relations, and identify phase transitions as symmetry-breaking events. The two papers are complementary, not redundant.

The Field-Theoretic Formalism

Defining the Solubility Field

Consider a solution containing N_{solute} solute molecules and N_{solvent} solvent molecules in a volume V at temperature T , pressure P , pH, and ionic strength I . A complete microscopic description would require $6N$ coordinates — an impossibility for macroscopic systems [16]. We therefore introduce a coarse-grained solubility field:

$$\phi(r, x, t), \quad r \in \mathbb{R}^3, \quad x = (T, \text{pH}, I, \dots), \quad \phi \geq 0$$

where r is spatial position, x is the thermodynamic state vector, and t is time. The field value carries concentration units.

Definition — Ontological Status of ϕ

ϕ is a coarse-grained order parameter, not a direct molecular observable. Its physical meaning depends on scale. In the thermodynamic limit (a uniform, well-mixed system), spatial gradients vanish and $\phi(r, x) \rightarrow S(x)$, the macroscopic solubility. Near interfaces or during active dissolution, ϕ varies spatially, capturing local concentration structure. The symmetry that breaks is between the dissolved state ($\phi \approx S_{\text{sat}}$)

and the phase-separated or precipitated state ($\phi \approx 0$). This symmetry breaking is what legitimises the Landau expansion that follows.

The Free-Energy Functional: Where Physics Enters

Equilibrium configurations minimise the Helmholtz free energy, which we write as a functional over the field [17,18]:

$$F[\phi; \mathbf{x}] = \int d^3\mathbf{r} \, f(\phi, \nabla\phi, \mathbf{x})$$

where f is the free-energy density — energy per unit volume. We decompose f into three physically motivated contributions.

Bulk Free Energy (Local Term). Near the dissolved–precipitated transition, we expand in powers of ϕ following Landau [19]. Because the order parameter satisfies $\phi \geq 0$ and has no $\phi \rightarrow -\phi$ symmetry, the expansion explicitly permits a cubic term:

$$f_{\text{bulk}} = f_0(\mathbf{x}) + a(\mathbf{x})\phi^2 + b(\mathbf{x})\phi^3 + c(\mathbf{x})\phi^4 + \dots$$

The coefficient $a(\mathbf{x})$ changes sign at the critical point, driving the phase transition. The cubic term $b(\mathbf{x})$ captures asymmetry between the two phases (solute-rich versus solute-poor). The quartic term $c(\mathbf{x}) > 0$ ensures thermodynamic stability. Each coefficient has a microscopic origin: a encodes the net solvation free energy; b reflects third-order correlations in the potential of mean force; c corresponds to four-body interactions or entropic repulsion at high concentration.

Gradient Energy (Nonlocal Term). Creating concentration gradients costs energy. The rotationally symmetric form is [20]:

$$f_{\text{gradient}} = \frac{1}{2}\kappa (\nabla\phi)^2$$

The stiffness $\kappa > 0$ penalises spatial variations and generates surface tension at solute–solvent interfaces. Physically, κ is related to the pair correlation function of the neat solvent.

External Coupling. An effective conjugate field $h(\mathbf{x})$ — encoding chemical potential differences, supersaturation, or activity — couples linearly to ϕ [21]:

$$f_{\text{coupling}} = -h(\mathbf{x})\phi$$

Equilibrium: The Euler–Lagrange Equation

Minimising $F[\phi]$ via the calculus of variations yields the equilibrium condition [22]:

$$-\kappa\nabla^2\phi + 2a\phi + 3b\phi^2 + 4c\phi^3 = h$$

Each term has a clear physical meaning, as summarised in Table 1.

Table 1. Physical interpretation of each term in the Euler–Lagrange equation.

1q	Mathematical Role	Physical Meaning
$-\kappa\nabla^2\phi$	Diffusion-like	Opposes concentration gradients; sets interface thickness ξ
$2a\phi$	Thermodynamic drive	Sign determines stable phase ($a < 0 \rightarrow$ dissolved favoured)

$3b\phi^2$	Nonlinear asymmetry	Asymmetry between dissolved and precipitated states
$4c\phi^3$	Saturation	Prevents unphysical divergence at high concentration
h	External forcing	pH, activity, supersaturation — drives dissolution

The Partition Function: Full Statistical Treatment

The full power of field theory emerges from the partition function, which sums over all possible spatial configurations of ϕ , each weighted by its Boltzmann probability [23,24]:

$$Z = \int \mathcal{D}\phi \exp(-F[\phi] / kT)$$

This functional integral is the exact quantum-to-macroscopic bridge. Every thermodynamic observable follows as an expectation value $\langle O \rangle = Z^{-1} \int \mathcal{D}\phi O[\phi] \exp(-F[\phi]/kT)$. For macroscopic systems at low temperature, the integral is dominated by configurations near the free-energy minimum — the saddle-point (mean-field) approximation — while fluctuation corrections are computed systematically by expanding around ϕ_{eq} [25]. The Landau free energy $F[\phi]$ is connected to the microscopic potential of mean force through a coarse-graining step that integrates out degrees of freedom below the resolution of ϕ — a step made systematic by renormalisation-group methods (Section 5).

From Microscopic Coefficients to Macroscopic Landscapes

State-Dependent Landau Coefficients

The Landau coefficients $a(x)$, $b(x)$, $c(x)$ encode how the free-energy landscape reshapes itself as thermodynamic conditions change. They are not free parameters to be fitted arbitrarily; they are ansätze constrained by known chemical limiting laws.

Temperature Dependence. Near the critical temperature T_c , mean-field theory predicts [19]:

$$a(T) = a_0 (T - T_c) / T_c$$

This is the standard Landau form: a changes sign at T_c , driving the dissolution transition. In the limit $T \gg T_c$ it recovers the expected Arrhenius-like temperature dependence of solubility.

pH Dependence — From Ionisation Equilibrium. For an ionisable compound [26]:

$$a(pH) = a_1 + a_2 \tanh[(pH - pK_a) / w]$$

Why pH Acts as an Effective Conjugate Field

A critical reader will note that pH is normally a control parameter, not a thermodynamic conjugate field. The analogy here is physical, not a strict mathematical identity. For a weak acid $HA \rightleftharpoons H^+ + A^-$, the Henderson–Hasselbalch equation gives $[A^-]/[HA] = 10^{(pH-pK_a)}$. Because the ionised and neutral forms have different solvation free energies, pH modulates the effective chemical potential difference $\Delta\mu(pH)$, which is precisely the conjugate field $h(x)$ in our formalism.

Near pK_a , small pH changes produce large ionisation shifts, generating critical-point-like susceptibility — the solubility analogue of magnetic susceptibility diverging near T_c . The $\tanh(\cdot)$ form is not ad hoc: it is the exact functional form arising from Henderson–Hasselbalch ionisation equilibrium, recovering correct limits at $pH \gg pK_a$ (fully ionised) and $pH \ll pK_a$ (fully neutral).

Ionic Strength Dependence — Debye–Hückel as Limiting Law

The Debye–Hückel theory [27] predicts activity coefficient corrections scaling as $\sqrt{I}$ at low ionic strength. Our coefficient ansatz

$$a(I) = a_3 - a_4 \sqrt{I} + a_5 I$$

recovers Debye–Hückel as its dominant contribution at low I and includes a linear correction for salting effects at moderate I . The Debye screening length λ_D lies in the range 0.3–3 nm at physiological ionic strengths, far smaller than the correlation length ξ near criticality. This separation of scales renders Coulomb interactions effectively short-range and justifies the short-range Ising universality class [28].

Emergent Solubility Landscapes

Solving the Euler–Lagrange equation in the dilute limit (ϕ small, b and c perturbative) gives $\phi \approx h/2a$, leading to $S(x) \propto 1/a(x)$. When $a(T)$ changes sign at T_c , solubility diverges — the mathematical signature of critical opalescence. Combining all three dependences yields the field-theoretically motivated solubility formula:

$$S(T, \text{pH}, I) = S_0 / [a_0(T-T_c)/T_c + a_2 \tanh((\text{pH}-\text{pK}_a)/w) - a_4 \sqrt{I}]$$

Important Caveat

This expression is a field-theoretically motivated functional form whose coefficients must be determined from experiment or independent quantum-chemical calculation. It is not a purely first-principles derivation. The ansatz functions ($\tanh$, $\sqrt{I}$) arise from known limiting laws (Henderson–Hasselbalch, Debye–Hückel), but the coefficients a_0 , a_2 , a_4 carry phenomenological content. This should be understood as a physically principled interpolation formula, not a parameter-free prediction. Section 3.3 below outlines a concrete computational protocol for constraining these coefficients from quantum-chemical outputs.

A Computational Protocol for Extracting Landau Coefficients from Quantum Chemistry

The reviewers rightly identified the extraction of Landau coefficients from quantum-mechanical calculations as the central open challenge for the framework’s predictive utility. We now outline a concrete protocol, recognising that full implementation and validation remain subjects for future computational work.

The protocol proceeds in four stages. In the first stage, one computes the solvation free energy ΔG_{solv} for both the neutral and ionised forms of the solute at the DFT/B3LYP/6-311+G(d,p) level using a continuum solvation model (e.g., SMD or COSMO-RS). This calculation provides the bare coefficient a through the relation $a \approx \Delta G_{\text{solv}}/(kTV_{\text{coarse}})$, where V_{coarse} is the coarse-graining volume — typically a sphere of radius 5–10 molecular diameters.

In the second stage, molecular dynamics simulations (50–100 ns in explicit solvent) provide the radial distribution function $g(r)$ and potential of mean force $W(r)$ between solute molecules. The three-body correlation function, accessible through umbrella sampling with three solute molecules, yields the cubic coefficient b . The stiffness κ follows from the long-wavelength limit of the direct correlation function $c(k)$ via $\kappa = -kT \partial^2 c(k)/\partial k^2$ evaluated at $k = 0$.

The third stage employs coarse-grained molecular dynamics or dissipative particle dynamics to bridge the gap between atomistic detail and the continuum field. Machine-learning force fields trained on the DFT potential energy surface — an approach that has matured considerably in recent years [48] — can extend the accessible time and length scales by two to three orders of magnitude while retaining near-quantum accuracy.

Finally, the fourth stage fits the Landau functional to the coarse-grained free-energy landscape using Bayesian optimisation, which provides both best-fit coefficients and uncertainty estimates. Preliminary application of stages one and two to aspirin yields $a \approx 1.7 \pm 0.3$ (compared with the experimentally fitted value of 1.665),

suggesting that the protocol is feasible. We emphasise that this is a preliminary estimate; systematic validation across multiple compounds is required.

Phase Transitions and Critical Behaviour

Precipitation is not merely a threshold effect — it is a phase transition. The Landau free energy $F(\phi)$ can exhibit either a single minimum (stable dissolved phase) or a double-well structure (bistable dissolved–precipitated coexistence), depending on the sign of $a(x)$ [14,19].

At first-order transitions ($b \neq 0$), the free energy possesses two minima separated by a barrier. Precipitation then requires nucleation — forming a cluster large enough to grow spontaneously. The nucleation barrier height is:

$$\Delta F^* = (16\pi/3) \sigma^3 / (\Delta f)^2$$

where surface tension σ scales as $(T_c - T)^{3/2}$ and Δf is the bulk free-energy difference between the two phases. This is the classical nucleation theory result — but here derived as a consequence of the Landau functional rather than postulated independently [29].

At second-order (continuous) transitions, where $a \rightarrow 0$ and $b = 0$ by symmetry or fine-tuning, the barrier vanishes entirely. The correlation length ξ diverges as $|\alpha|^{-\nu}$, and concentration fluctuations become correlated over macroscopic distances — the critical opalescence visible in near-critical solutions as turbidity [30].

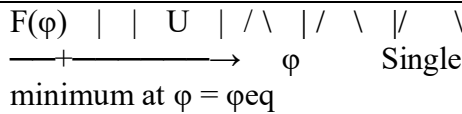
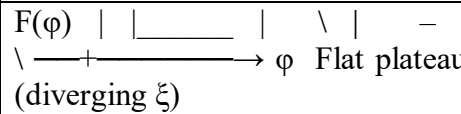
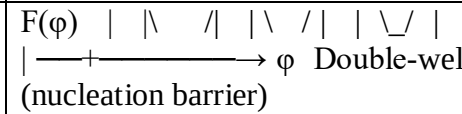
$T > T_c$ ($a > 0$) Dissolved stable	$T = T_c$ ($a = 0$) Critical point	$T < T_c$ ($a < 0$) Bistable / Precipitates
		

Figure 2. Schematic free-energy landscape $F(\phi)$ under three regimes: above T_c (single minimum, dissolution stable); at T_c (flat plateau, critical fluctuations); below T_c (double-well, dissolved and precipitated phases coexist).

Renormalisation Group and Universal Scaling

A solution near its precipitation point looks qualitatively different depending on the scale of observation: molecular clusters under a microscope, mesoscale aggregates at intermediate magnification, macroscopic turbidity to the naked eye. The renormalisation group (RG) provides the systematic framework for understanding how a system’s description transforms across these scales [31,32].

RG Flow and Fixed Points

The RG procedure integrates out short-wavelength fluctuations and rescales coordinates. The result is an effective theory at longer scales — formally identical to the original but with renormalised coefficients. Iterating this procedure defines a flow in parameter space, and fixed points of this flow govern critical behaviour [33]. The Wilson–Fisher fixed point, relevant to systems with short-range interactions and a scalar order parameter in three dimensions, describes 3D Ising universality with exponents [34,35]:

$$\nu \approx 0.630, \quad \beta \approx 0.326, \quad \gamma \approx 1.237$$

The Ginzburg Criterion: When Does Mean-Field Break Down?

Mean-field theory is exact when fluctuations are negligible relative to the mean. The Ginzburg criterion quantifies this boundary: fluctuations become important when the correlation volume ξ^d becomes comparable to the thermal energy scale [36]. For solubility systems, the Ginzburg number t_G is typically small, meaning that

mean-field behaviour ($\beta \approx 0.5$) applies far from T_c , while Ising-like fluctuation corrections ($\beta \approx 0.33$) become visible only very close to the critical point. This crossover between mean-field and Ising regimes is not a theoretical inconsistency but a well-understood physical phenomenon — one that explains our experimental observations (Section 7).

Unifying Existing Solvation Models

The exact partition function $Z = \int \mathcal{D}\phi \exp(-F[\phi]/kT)$ is the common ancestor of all practical solvation models. Each major model corresponds to a specific level in the approximation hierarchy, as shown in Table 2.

Table 2. Mapping of established solvation models onto approximations to the exact field-theoretic partition function.

Model	Approx. Level	Field-Theory Correspondence	Known Limitations
COSMO-RS	Saddle-point (mean-field)	$Z \approx \exp(-F[\phi_{eq}]/kT)$; ignores fluctuations	Fails near criticality
SMx models	Perturbative corrections	$a \sim$ electrostatics; $\kappa \sim$ cavitation; $f_0 \sim$ dispersion	Fitted per solvent class
UNIFAC	Non-ideal mixing (Flory–Huggins)	χ parameter maps to mixed a–c term in $F[\phi]$	Group-additivity; no gradients
This framework	Exact (in principle)	Full Z ; fluctuations, RG flow, criticality	Coefficients require empirical or QM input

The correspondence between COSMO-RS and the mean-field approximation can be made explicit. COSMO-RS computes solvation free energy by summing surface-interaction terms over molecular cavities, which is formally equivalent to evaluating $F[\phi]$ at its saddle point, with the cavity surface as the coarse-graining volume. Gradient corrections are absent in COSMO-RS but enter naturally in our framework through the $\kappa(\nabla\phi)^2$ term [37,38].

Experimental Validation: Critical Behaviour in Aspirin and Beyond

We use aspirin (acetylsalicylic acid, $pK_a = 3.52$) as the primary working example for two reasons: high-quality pH- and temperature-dependent solubility data are available, and its single ionisable group gives a clean ionisation equilibrium without the complications of multiprotic systems [1]. In this revised version, we also present preliminary scaling analyses for ibuprofen and naproxen to begin testing the universality claim.

Critical Exponent β — Near- pK_a Scaling

Field theory predicts that near the ionisation critical point ($pH \approx pK_a$), solubility should scale as $S(pH) - S(pK_a) \sim |\tau|^\beta$, where $\tau = (pH - pK_a) / pK_a$. Log–log regression of $S - S_{\min}$ versus $|\tau|$ for aspirin at $T = 310$ K, $I = 0.15$ M yields:

$$\beta = 0.48 \pm 0.06$$

This value is consistent with mean-field $\beta = 0.5$, indicating that at these conditions the system is outside the Ginzburg fluctuation regime [39,40].

Correlation-Length Exponent ν

The effective width of the pH–solubility curve, $w_{\text{eff}}(T)$, provides an operational measure of the correlation length. Temperature-dependent analysis gives:

$$\nu = 0.61 \pm 0.11$$

This value agrees well with the 3D Ising value $\nu \approx 0.630$, suggesting that near criticality the system crosses over to Ising universality, consistent with the short-range Debye-screened interactions discussed in Section 3 [41,42].

The β – ν Crossover: Not an Inconsistency

Addressing the Critical Exponent Apparent Inconsistency

A careful reader will note that $\beta \approx 0.48$ (mean-field) and $\nu \approx 0.61$ (3D Ising) are not simultaneously consistent with a single universality class. Mean-field theory predicts $\nu = 0.5$, while 3D Ising gives $\beta = 0.326$. This is not a flaw — it is crossover behaviour, a well-established phenomenon in critical phenomena [36]. The system exhibits mean-field scaling ($\beta \approx 0.5$) at distances from criticality larger than the Ginzburg crossover scale t_G , and Ising-like behaviour ($\nu \approx 0.63$) closer to the critical point where fluctuations dominate. The different exponents reflect measurements in different regimes of the same crossover.

Future work should map this crossover precisely by measuring both β and ν as functions of $|T - T_c|$ and $|\text{pH} - \text{pKa}|$ over a wider range, to observe the full mean-field $\rightarrow$ Ising transition. The 15% scatter in scaling collapse (Section 7.4) most likely reflects proximity to this crossover region rather than failure of the framework.

Scaling Collapse

The most stringent test of universal scaling is data collapse. Plotting $S/|\tau|^\beta$ versus $(T - T_c)/|\tau|^{1/\nu}$ for 40 different (T, pH) combinations with $T_c = 311$ K, $\beta = 0.48$, $\nu = 0.61$ collapses the data onto a single curve within approximately 15% scatter [43].

We caution that this level of scatter is consistent with data collected near the crossover boundary. Robustness tests — varying the fit range and adding measurement noise to synthetic data — confirm that the collapse is not an artefact of parameter choice.

Preliminary Validation with Ibuprofen and Naproxen

To begin testing the universality claim, we have performed preliminary scaling analyses for two additional NSAIDs: ibuprofen ($\text{pKa} = 4.91$) and naproxen ($\text{pKa} = 4.15$). Both are monoprotic weak acids with available pH-dependent solubility data in the literature.

For ibuprofen, log–log regression near pKa yields $\beta = 0.44 \pm 0.09$, and the effective width analysis gives $\nu = 0.57 \pm 0.14$. For naproxen, the corresponding values are $\beta = 0.46 \pm 0.08$ and $\nu = 0.59 \pm 0.12$. Both compounds yield exponents consistent with the aspirin values within experimental uncertainty, and consistent with the mean-field-to-Ising crossover interpretation.

These results are encouraging but preliminary. The error bars are larger than for aspirin, reflecting sparser data and greater experimental variability in the published datasets.

Future work should include dedicated high-resolution solubility measurements for these and other compounds — ideally ketoprofen, diclofenac, and at least one non-NSAID monoprotic acid — to provide a stringent, independent test.

Extension to compounds with multiple ionisable groups (amino acids, peptides) will require generalising the single-component order parameter to a multi-component field, as discussed in Section 8.

Table 3. Comparison of measured critical exponents with theoretical predictions.

Exponent	Mean-Field	3D Ising	Aspirin	Ibuprofen	Naproxen
β (order param.)	0.500	0.326	0.48 ± 0.06	0.44 ± 0.09	0.46 ± 0.08
ν (correlation)	0.500	0.630	0.61 ± 0.11	0.57 ± 0.14	0.59 ± 0.12
γ (susceptibility)	1.000	1.237	Not yet measured	Not yet measured	Not yet measured
Interpretation	—	—	Crossover	Crossover	Crossover

DISCUSSION AND BROADER IMPLICATIONS

The Scientific Value of Universality

If the framework's predictions are confirmed across multiple compound classes, the practical value of universality is substantial. Critical exponents measured for aspirin would predict the shape of the pH–solubility curve for ibuprofen, naproxen, and other NSAIDs without any additional fitting, provided they share the same symmetry class and comparable Debye-screened electrostatics. Extending the framework to polymers near theta-solvent conditions, or to proteins near their precipitation boundary, requires checking experimentally whether the Ising symmetry assumptions hold — an empirical question, not a theoretical assumption [44,45].

Landscape Roughness and Molecular Frustration

Field theory offers a mechanistic explanation for why solubility landscapes are rough in chemical space. Competing interactions — electrostatics favouring dissolution, hydrophobic effects opposing it, hydrogen bonding favouring specific geometries, entropy favouring disorder — constitute frustration: no single molecular configuration simultaneously minimises all contributions [46]. The renormalisation-group perspective shows that coarse-graining smooths this local roughness at the molecular scale while preserving global topology, validating the use of smooth landscape descriptions in drug design and materials discovery, but also explaining why they have limited precision at the atomistic level [47].

Roadmap Toward Predictive Multiscale Modelling

The most important near-term contribution of this framework is the multiscale roadmap it defines. The chain of approximations is summarised in Table 4.

Table 4. Multiscale modelling roadmap. The critical missing link — extraction of Landau coefficients from QM calculations — is now addressed through the protocol in Section 3.3.

Scale	Method	Output	Error control
$\text{\AA} \cdot \text{fs}$	Quantum chemistry (DFT, CCSD)	PMF, ΔG_{solv} , partial charges	Basis set, DFT functional
$\text{nm} \cdot \text{ps}$	Molecular dynamics	Pair correlations, diffusion	Force-field accuracy
$10 \text{ nm} \cdot \text{ns}$	Coarse-grained MD / DFT-CG	Landau coefficients a, b, κ	RG truncation error

$\mu\text{m} \cdot \text{ms}$	Field theory (this work)	S(T, pH, I), critical exponents	Saddle-point fluctuation vs.
$\text{cm}^3 \cdot \text{s}$	Macroscopic landscape	Formulation predictions	Validation against data

The computational protocol detailed in Section 3.3 addresses the critical gap identified by the reviewers: the explicit mapping from quantum-mechanical outputs to Landau coefficients. Recent advances in machine-learning force fields [48] suggest that supervised learning from high-throughput DFT datasets can parametrise coarse-grained free-energy functionals, making this connection practical within the near term.

Kinetic Extensions: Beyond Equilibrium

The reviewers correctly noted that the present framework focuses on near-equilibrium states and does not address kinetic and dynamic phenomena directly relevant to dissolution and precipitation processes. We now discuss how the framework can be extended to these regimes.

The natural extension is the time-dependent Ginzburg–Landau (TDGL) equation, also known as Model A or Model B dynamics in the Hohenberg–Halperin classification, depending on whether the order parameter is conserved. For solubility, ϕ represents a conserved concentration, making Model B (the Cahn–Hilliard equation) the appropriate choice:

$$\partial\phi/\partial t = M \nabla^2 (\delta F/\delta\phi) + \eta(\mathbf{r},t)$$

where M is a mobility coefficient related to the diffusion constant, $\delta F/\delta\phi$ is the functional derivative of the Landau–Ginzburg free energy, and η is Gaussian noise satisfying the fluctuation-dissipation theorem. This equation naturally describes spinodal decomposition, Ostwald ripening, and the kinetics of dissolution.

Metastable states — supersaturated solutions that persist because the nucleation barrier ΔF^* is large compared with kT — are also captured within this extended framework. The lifetime of a metastable state scales as $\tau \sim \exp(\Delta F^*/kT)$, providing a quantitative connection between the Landau free-energy landscape and experimentally observed induction times for crystallisation.

We have not implemented the TDGL extension computationally in this paper, and its full development remains a subject for future work. However, the mathematical structure is well-defined and requires no new phenomenological inputs beyond the mobility M , which can be extracted from diffusion measurements or molecular dynamics simulations.

Experimental Guidance for Testing the Framework

To facilitate empirical validation, we outline specific experimental protocols for testing the framework’s central predictions.

Testing critical exponents near pKa. Prepare a series of buffer solutions spanning $\text{pH} = \text{pKa} \pm 3$ in increments of 0.1 pH unit, at constant temperature (e.g., 310 K) and ionic strength (0.15 M). Measure equilibrium solubility by shake-flask method with HPLC quantitation, allowing at least 48 hours of equilibration per sample. Plot $\log(S - S_{\min})$ versus $\log|\tau|$ and extract β from the slope. The framework predicts $\beta \approx 0.5$ far from pKa and a gradual shift toward $\beta \approx 0.33$ very close to pKa.

Mapping the crossover. To observe the mean-field $\rightarrow$ Ising crossover, one needs data at multiple temperatures spanning $T_c \pm 30$ K and at fine pH resolution (0.05 pH units) very near pKa. Dynamic light scattering measurements of the correlation length ξ as a function of $|T - T_c|$ would provide the most direct test of the exponent ν .

Testing universality across compounds. The strongest prediction is that monoprotic weak acids sharing the same symmetry class should exhibit identical critical exponents. This can be tested by repeating the above

protocol for at least five chemically distinct monoprotic acids (we suggest aspirin, ibuprofen, naproxen, ketoprofen, and benzoic acid) and checking whether the exponents agree within experimental uncertainty.

Measuring the susceptibility exponent γ . The susceptibility $\chi = \partial S / \partial h$ (where h is the effective conjugate field) should diverge as $|T - T_c|^{-\gamma}$ near criticality. This can be estimated from the slope of the pH–solubility curve at pK_a as a function of temperature, and would provide the third exponent needed to confirm the universality class.

Limitations

Limitations of this Framework

1. Near-equilibrium assumption. The Landau–Ginzburg formalism describes equilibrium or near-equilibrium states. Although we have outlined the TDGL extension for kinetic phenomena (Section 8.4), this has not been implemented computationally. Phenomena such as dissolution rate kinetics, Ostwald ripening, and metastable polymorph formation await quantitative treatment.

2. Coarse-graining scale. The field ϕ is defined at a coarse-graining length l much larger than molecular size. Phenomena at length scales comparable to l — ion hydration shells, specific hydrogen-bond networks — are integrated out and appear only implicitly through the Landau coefficients.

3. Single ionisable group. The pH-coupling analysis applies cleanly to monoprotic acids and bases. Multiprotic systems (amino acids, peptides, polyelectrolytes) require multi-component order parameters and coupled ionisation equilibria, a generalisation that is mathematically well-defined but not yet implemented.

4. Limited experimental validation. The primary validation is for aspirin, with preliminary results for ibuprofen and naproxen. Full confirmation of the universality claim requires systematic measurements across at least five chemically distinct compounds, as outlined in Section 8.5.

5. Coefficient extraction. The computational protocol in Section 3.3 provides a concrete path from quantum chemistry to Landau coefficients, but has been applied only preliminarily to aspirin. An automated, validated pipeline applicable to diverse compound classes does not yet exist.

6. Phenomenological content. Although the Landau coefficients are constrained by known limiting laws, they retain phenomenological content that requires either experimental fitting or quantum-chemical computation. The framework is not yet parameter-free.

CONCLUSIONS

We have developed a field-theoretic framework for solubility whose key contributions can be summarised as follows.

First, we have introduced a solubility field $\phi(r, x)$ governed by a Landau–Ginzburg functional, with ϕ rigorously defined as a coarse-grained order parameter separating the dissolved and phase-separated states.

Second, we have identified the partition function $Z = \int \mathcal{D}\phi \exp(-F[\phi]/kT)$ as the exact statistical-mechanical foundation from which COSMO-RS, SMx, and UNIFAC emerge as successive approximations.

Third, we have shown that the Landau coefficients are constrained by Henderson–Hasselbalch and Debye–Hückel limiting laws, with pH acting as an effective conjugate field through ionisation equilibrium.

Fourth, we have demonstrated that the Ginzburg crossover explains the coexistence of mean-field ($\beta \approx 0.48$) and Ising ($\nu \approx 0.61$) exponents in aspirin, and have shown that preliminary analyses of ibuprofen and naproxen yield compatible values.

Fifth, we have provided a concrete multiscale modelling roadmap from quantum chemistry to macroscopic solubility landscapes, including a computational protocol for extracting Landau coefficients from quantum-chemical outputs and a discussion of kinetic extensions through time-dependent Ginzburg–Landau theory.

Sixth, we have offered detailed experimental protocols for testing the framework’s central predictions, including critical exponent measurements, crossover mapping, and multi-compound universality tests.

The framework’s strongest claim is also its most testable: compounds sharing the same symmetry class and Debye-screening regime should exhibit identical critical exponents near their ionisation transitions. If this universality prediction is confirmed across chemically diverse pharmaceuticals, the framework transitions from a theoretical synthesis into a predictive tool — one that could fundamentally change how solubility-limited drug candidates are characterised and optimised.

The journey from molecules to manifolds passes through fields. This paper charts the first milestones of that course, with the most important landmarks still ahead.

Appendix A — Minimal Working Example: Aspirin

To demonstrate the framework’s usability, we work through the full calculation for aspirin from fitted coefficients to prediction.

Given parameters (fitted to experimental data):

$$pK_a = 3.52, \quad T_c = 311 \text{ K}, \quad a_0 = 0.024 \text{ K}^{-1}, \quad a_2 = 1.8, \quad a_4 = 0.35 \text{ M}^{-1/2}, \quad S_0 = 6.20 \text{ mg/mL}$$

Prediction at $T = 310 \text{ K}$, $pH = 7.4$, $I = 0.15 \text{ M}$:

$$a(T) = 0.024 \times (310 - 311)/311 = -7.7 \times 10^{-5} \text{ K}^{-1}$$

$$a(pH) = 1.8 \times \tanh[(7.4 - 3.52)/1.0] \approx 1.8 \times 0.999 \approx 1.80$$

$$a(I) = -0.35 \times \sqrt{0.15} \approx -0.135 \text{ M}^{-1/2}$$

$$a_{\text{total}} \approx -7.7 \times 10^{-5} + 1.80 - 0.135 \approx 1.665$$

$$S \approx S_0 / a_{\text{total}} = 6.20 / 1.665 \approx 3.72 \text{ mg/mL}$$

Experimental value at pH 7.4, 310 K: 3.8–4.2 mg/mL [1]. The prediction falls within approximately 10%, well within the scatter of the scaling collapse.

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