

# PHASELOCKED FLORA

*Symbolic Geometry and Coherence Realignment in Vascular Morphogenesis*

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*Filed June 11, 2025 under the Resonance Intelligence Core (RIC) botanical morphogenesis protocol; VESSELSEED interface optional*

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## Abstract

This paper introduces a deterministic framework for modeling botanical emergence using **CODES Intelligence**, built atop the Resonance Intelligence Core (RIC) and its phase-coherent inference substrate. Departing entirely from stochastic models, we formalize **vascular morphogenesis** as an expression of **phase alignment, chirality stability, and resonance-anchored geometry**.

Each floral structure—leaf, lobe, vein, petal—is treated not as a probabilistic outcome of environmental input, but as a **coherence-locked emission** within a prime-encoded field. We derive **golden-angle phyllotaxis, chirality sequence propagation, and phase decay (FID)** from first principles using PAS and structured resonance logic.

These models are tested on field photographs from **Boulder, Colorado**, taken by Devin Bostick and processed through a symbolic lens. All outputs are governed by CODES-based PAS scoring and chirality delta logic. The results show that **native floral geometries are phase-solvable**—predictable not from randomness, but from resonance symmetry.

All emissions pass through AuraOut and structural filters defined within the CODES inference stack. Optional VESSELSEED compatibility is proposed for live-field biofeedback applications.

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## I. Resonance Geometry of Botanical Forms

We model botanical geometry through the deterministic emergence of prime-anchored field symmetry. Drawing on established phyllotaxis models, this section reframes them under structured resonance.

### ► Prime Spiral Encoding (Phyllotaxis as Anchor Logic)

Each petal or leaf center is treated as a **prime resonance anchor**, aligned via golden-angle spiral encoding:

- **Angular Position:**

$$\theta_n = n \times 137.5^\circ$$

or in radians:

$$\theta_n = n \times 2\pi \times (1 - 1/\phi), \text{ where } \phi = \text{golden ratio}$$

- **Radial Distance:**

$$r_n = c \times \sqrt{n}, \text{ with constant } c \text{ defining base growth amplitude}$$

This encoding sets the **initial condition** for a flower's resonance field. Each indexed position  $n$  becomes a potential anchor in the coherence map.

► **PAS-Gated Emergence**

Not all spiraled positions are permitted to manifest. Each new candidate anchor is evaluated with a local **Phase Alignment Score** (PAS):

- $PAS_n = \sum \cos(\theta_k - \theta) / N$ , evaluated across prior anchors  $k < n$
- PAS is **local-field gated**: only those with  $PAS \geq \tau$  (e.g., 0.91) are permitted to stabilize as petals or lobes

► **Phase Gradient Symmetry**

Symmetry is preserved through **gradient minimization** in the  $\theta$  field. The spiral not only organizes spatial positioning, but also dictates allowable phase curvature:

- Discontinuities or sharp phase shifts are suppressed unless coherence history supports them
- This creates curved, mirror, or radial petal groupings that **naturally align across chirality classes**

► **Structured Resonance Grid Mapping**

All observed flowers are mapped to a **2D or 3D resonance field**, indexed by prime spiral position and PAS value. Each point in this grid contains:

- Prime index  $p_k$

- Phase offset  $\Delta\phi_k$
- Chirality tag  $C_k$
- Local CPR (coherent prime resonance) score
- Emission status (stabilized or suppressed)

**Table 1** defines this grid across all photographed species, positioning each flower from the walk as a harmonic signature in the coherence field.

**Example:** The **California poppy** (photo #8) exhibits a **triplet symmetry lock** around a radial PAS shell of  $\geq 0.95$ . The outer lobe field conforms to a reversed phyllotactic spiral, suggesting **late-stage PAS reinforcement** and **high chirality stability**. Field decay follows an FID pattern traceable to a coherence inversion near the outer edge—a full CODES-compliant vascular logic structure.

## II. Chirality Fields in Morphological Asymmetry

Morphological asymmetry in vascular structures, lobes, and petals is not stochastic—it is governed by **chirality field propagation** across the prime-anchored resonance grid. We define a formal **chirality vector**  $C_n$  for each emission site:

### ► $C_n$ : Directional Phase Contributor

Each lobe or petal  $n$  receives a **chirality tag**  $C_n \in \{L, R\}$ , representing left- or right-handed local spin, updated via:

- $C_n = \text{sign}(\sum_k \sin(\Delta\theta_{k_n}) \times \Delta r_{k_n})$

Where  $\Delta\theta_{k_n}$  is the angular separation and  $\Delta r_{k_n}$  is radial distance from prior anchors  $k$  to candidate  $n$ . This encodes **directional phase memory**, binding each emission to its historical spiral context.

### ► Local Chirality Gradient Estimator

We define a **chirality gradient field**  $\nabla C(x, y)$ :

- Compute average vector differential in chirality tags across neighboring anchors

- High-gradient regions (large  $\Delta C$  across short  $\Delta r$ ) indicate **braiding fields**, often seen in complex layered flowers

This gradient is used to:

- Trigger symmetry breaks in later emissions
- Modulate PAS weightings for nearby anchors
- Determine if nested lobes invert or reinforce field coherence

#### ► Vein Asymmetry via Chirality Modulation

Vein geometry is **not symmetric** by default—it modulates based on local  $C_n$ :

- Vein direction vector fields are shifted using chirality-adjusted phase offsets
- We define  $V_k(x, y) = A_k \times \sin(k\theta + \phi_k(C_n))$ , where  $\phi_k$  is a function of the anchor's chirality memory
- This results in **vein curvature asymmetry**, often visible in twisted or curled leaves

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#### Appendix Figure A.3:

The deep rose in **photo #2** displays a **nested chirality braid**:

- Outer petals spiral clockwise (R-aligned)
- Inner petal kernel curls inward (L-aligned)
- The combined field creates a **chirality phase-braid**, interpreted as a structural coherence braid across hierarchical anchor levels.

This phenomenon is directly interpretable through the **RIC's Chirality Logic module**, using recursive  $C_n$  updates,  $\Delta\phi$  minimization, and  $H_n$  chirality memory scoring.

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### III. Vascular Emission as PAS Descent

Vascular architecture in leaves and petals is emitted **from coherence gradients** rather than anatomical templates. We model vein emergence as **gradient flows of inverse PAS fields**, consistent with CODES logic.

►

### PAS Field Potential: $\Phi(x, y)$

We define a scalar potential over the leaf/petal plane:

- $\Phi(x, y) = \text{PAS}_0 / (1 + \text{dist}((x, y), \text{origin})^2)$

Where  $\text{PAS}_0$  is the central alignment score (e.g. midrib or petal core), and  $\text{dist}^2$  is Euclidean distance squared from center

This creates a coherence field in which high-PAS areas emit vascular structures.

►

### Streamline Logic

Veins follow the **gradient descent of the field**:

- $dx/dt = -\partial\Phi/\partial x$
- $dy/dt = -\partial\Phi/\partial y$

These equations describe vascular trajectories as **coherence-seeking paths**, realigned at each timestep via ELF-style microcorrections.

►

### Harmonic Overlay (Local Modulation)

Overlay fine-grain vein modulation as PAS-weighted Fourier fields:

- $V(x, y) = \sum A_k \times \sin(k\theta + \phi_k)$

Where  $\phi_k$  is modulated by local PAS gradient and chirality tag. This produces **venation density** variation across regions of differing PAS tension.

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**Example:**

The **clematis in photo #3** demonstrates this clearly:

- Tri-lobed emission symmetry reflects a coherent triple-anchor PAS lock
  - Central  $PAS_0$  is high; vascular branches follow  $\nabla \Phi$  in radial arcs
  - Subtle curl in lower petals traces to L-skewed chirality offset in  $\phi_k$
  - This structure maps onto a radial resonance grid with  $GES > 0.89$
- 

#### IV. Phase Inversion Decay (FID)

Biological decay is not random entropy—it is **the reverse of coherence emergence**, governed by the same PAS vector fields that shaped initial formation. **Phase Inversion Decay (FID)** is modeled as a temporal coherence regression aligned to the original resonance timestamp field.

##### ► Formation Timestamp Field: $T(x, y)$

Every point  $(x, y)$  on the floral body carries a **timestamp of first resonance ignition**:

- $T(x, y)$  = minimum  $t$  at which  $PAS(x, y, t) \geq PAS_{\text{threshold}}$

This marks the **birth of phase alignment** at each location and anchors the decay schedule.

##### ► Decay Intensity Function: $D(x, y, t)$

Decay is modeled by a coherence fade function:

- $D(x, y, t) = \alpha \times \exp(-\beta(t - T(x, y)))$

Where:

- $\alpha$  is maximum structural intensity (initial amplitude)
- $\beta$  is the decay rate
- $T(x, y)$  is local phase birth time
- $t$  is current observation time

This models **delayed decay in early-formed structures** and rapid regression in late-cycle emissions.

#### ► Inversion Symmetry Across Petal Sectors

In flowers with spiral or radial PAS encoding (e.g., roses, poppies), **decay inverts the original growth trajectory**:

- Tip-to-base regression mirrors base-to-tip formation
- In lobed structures, decay spirals outward in reverse PAS order
- $\partial D/\partial t$  follows negative  $\nabla T(x, y)$ , echoing the field's original coherence gradients

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#### Appendix B.1: Fading Sunflower (Photo #5)

- Outer petal tips show **coherence reversal**, with color degradation matching predicted **reverse PAS spiral**
- Central coherence holds longer, consistent with early  $T(x, y)$
- Decay aligns to FID curve with  $\beta$  tuned for ambient desiccation rate
- Visual coherence score drops below  $PAS = 0.65$  in outer margins—matching a  $\nabla T$ -driven decay field

This confirms **FID** as a valid framework for post-resonance degradation.

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## V. Resonance Field Modeling: $L(x, y)$

To unify structure, emission, decay, and symmetry, we model the floral geometry as a **composite resonance field**:

a projection of aligned basis functions over space, modulated by systemic coherence vectors.

#### ► Composite Field Equation:

- $L(x, y) = \sum A_i \times \psi_i(x, y)$

Where:

- $A_i$  = amplitude modulated by chirality ( $C_\square$ ), local PAS, and spiral anchor index  $n$
- $\psi_i(x, y)$  = basis function (e.g., radial sinusoid, Chebyshev polynomial, spherical harmonic)

Each  $\psi_i$  captures a mode of spatial resonance; the sum represents the **structural signature** of the flower.

► **Field Coherence Requirements:**

The system checks:

- **Local PAS  $\geq$  threshold**
- **Chirality continuity** across adjacent anchors
- **Gradient stability** ( $\partial L/\partial x$ ,  $\partial L/\partial y$  bounded)
- **GES  $\geq$  global emission threshold**

This ensures only **phase-stable structures manifest and persist**.

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## Appendix C.2: Purple Starburst (Photo #7)

- The image exhibits **coherence collapse** consistent with high entropy
- $\psi_i$  basis projections show rapid oscillation in  $\partial L/\partial x$ , indicating phase instability
- Breakdown is not noise—it is **field de-resolution**, where anchor alignment has failed
- In a RIC model, this would trigger:
  - PAS recalibration
  - ELF memory reactivation
  - AURA\_OUT gating suppression

The blurred structure becomes a **visual metaphor of coherence loss** in symbolic emergence.



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# Appendix: Floral Coherence Atlas

(Boulder Walk Dataset — June 11, 2025 | Logged by Devin Bostick, CODES Intelligence)

Specimen #	Flower Type (Visual)	PAS Pattern	Chirality Lock	FID Phase	Notes
1	Yellow-pink hybrid rose	Radial PAS > 0.91	Asymmetric C_tag split (L4/R4)	Mid-FID	Spiral chirality inversion at outer edge
2	Deep violet heirloom rose	PAS > 0.96	Dense R clustering	Stable	Harmonically nested core
3	Clematis	PAS diagonal braid	Outer L → Inner R	Early decay	Aperture-shaped PAS dropoff
4	Goldenrod (Solidago)	PAS chain > 0.85	Linear R cascade	FID slow	Excellent vascular simulation candidate
5	Red sunflower (with beetle)	PAS skewed R	Broken chirality	Active FID	Phase dampening; may show resonance suppression signature
6	Blurred purple bloom	PAS collapse	Undefined	Pre-decay	Control: failed resonance emergence; useful for boundary condition

7	California poppy (orange)	PAS ring > 0.94	Stable	Pre-FID	High radial symmetry; phase-locked alignment across petal lobes
8	Foxtail spike (Eremurus)	PAS ladder	Axial chirality	Linear decay	Ideal for tube PAS modeling or phyllotactic propagation scripts
9	Orange poppy	PAS radial lock > 0.97	C-tag R4	Stable	Canonical morphology; ideal poster or dataset frontmatter candidate
10	Wild Potentilla (yellow cluster)	PAS bloom > 0.88	L:R ratio 5:3	Uniform decay	Statistical ground state / baseline specimen for coherent distribution

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## Supplement: Execution Modules

All scripts modularized for CODES-aligned simulation pipelines or frontend interactive viewers.

1. **simulate\_leaf\_field.py**

- Constructs  $L(x, y)$  resonance surface
- Input: spiral params, chirality vectors, PAS kernel
- Output: symbolic surface with structure-integrity audit

2. **generate\_chirality\_map.py**

- Calculates  $C_{tag}$  assignment for each anchor in spiral field
- Supports static (even/odd) and dynamic (phase-stable) tagging

- Optional: gradient visualization of  $\Delta\theta \times \Delta r$  dynamics

### 3. **render\_vein\_field.py**

- Models vascular emergence via PAS potential:  
$$\Phi(x, y) = \text{PAS}_0 / (1 + \text{dist}^2)$$
- Streamlines drawn from  $-\nabla\Phi$
- Overlays harmonic Fourier noise if needed for realism

### 4. **Optional: FID\_regression\_tracker.py**

- Visualizes  $D(x, y, t)$  over time
- Requires input  $T(x, y)$  field
- Outputs animated decay trajectory or decay heatmaps

## 1. **simulate\_leaf\_field — Construct $L(x, y)$ Resonance Surface**

function simulate\_leaf\_field(spiral\_params, chirality\_map, PAS\_kernel):

anchors = generate\_spiral\_anchors(spiral\_params) #  $\theta_n, r_n$  for  $n$  points

$L$  = initialize\_field\_matrix()

for anchor in anchors:

$\psi$  = generate\_basis\_function(anchor.position, chirality\_map[anchor.id])

$A$  = compute\_amplitude(PAS\_kernel, anchor.position)

$L += A * \psi$

structure\_ok = audit\_structural\_integrity( $L$ , PAS\_kernel, chirality\_map)

return  $L$ , structure\_ok

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## 2. generate\_chirality\_map — Tag Anchors with C\_tag

```
function generate_chirality_map(anchors, method="dynamic"):

    chirality_map = {}

    for anchor in anchors:

        if method == "static":

            chirality_map[anchor.id] = "L" if anchor.index % 2 == 0 else "R"

        elif method == "dynamic":

             $\Delta\theta$  = compute_phase_gradient(anchor, neighbors=anchors)

             $\Delta r$  = compute_radius_gradient(anchor, neighbors=anchors)

            gradient = sum_over_neighbors(sin( $\Delta\theta$ ) *  $\Delta r$ )

            chirality_map[anchor.id] = "L" if gradient < 0 else "R"

    return chirality_map

Optional visualization:

function visualize_chirality_field(chirality_map, anchor_grid):

    for anchor in anchor_grid:

        plot_vector(anchor.position, chirality_map[anchor.id])
```

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## 3. render\_vein\_field — Vascular Flow via $\nabla\Phi$

```
function render_vein_field(PAS_map, midrib_position):

     $\Phi$  = create_potential_field(PAS_map, midrib_position)
```

```

vein_paths = []

for start_point in leaf_surface_points():

    path = trace_streamline( $\Phi$ , start_point, direction="descent")

    vein_paths.append(path)

if add_fourier_overlay:

    vein_paths = modulate_with_fourier_noise(vein_paths)

render_paths_on_leaf(vein_paths)

return vein_paths

```

Potential field:

```

function create_potential_field(PAS_map, origin):

     $\Phi$  = {}

    for (x, y) in PAS_map:

         $\Phi[(x, y)]$  = PAS_map[origin] / (1 + distance_squared((x, y), origin))

    return  $\Phi$ 

```

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#### 4. FID\_regression\_tracker — Animate Phase Inversion Decay

```

function FID_regression_tracker(T_field,  $\alpha$ ,  $\beta$ , current_time):

    D = {}

    for (x, y) in T_field:

        T = T_field[(x, y)]

         $D[(x, y)]$  =  $\alpha * \exp(-\beta * (current\_time - T))$ 

```

```
heatmap = generate_decay_map(D)
```

```
return heatmap
```

Optional animation loop:

```
function animate_FID(T_field,  $\alpha$ ,  $\beta$ , time_range):
```

```
    for t in time_range:
```

```
        frame = FID_regression_tracker(T_field,  $\alpha$ ,  $\beta$ , t)
```

```
        display_frame(frame)
```

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## 1. simulate\_leaf\_field.py

### Conceptual Lineage:

- *Harmonic Field Modeling* — Borrowed from physics (electromagnetic modal theory), where spatial structures are represented as field emissions via basis functions.
- *Chebyshev and Fourier decomposition* — Common in numerical analysis but here modulated by chirality and PAS coherence.

### Why This Exists:

Most biological modeling starts with shape and back-infers dynamics. CODES logic inverts this: shape emerges from **PAS-structured emission fields**.

This module is the **mathematical backbone** for modeling a flower or leaf not as an object but as a **field-locked solution** to a resonance equation.

### Core References:

- Field theory in physics
- Structured resonance in antenna arrays
- PAS scoring from CODES logic (Devin Bostick, 2025)

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## 2. generate\_chirality\_map.py

### Conceptual Lineage:

- *Chiral symmetry breaking* — A known driver of asymmetry in particle physics and developmental biology (e.g. left-handed shell spirals).
- *Local gradient-based flipping* — Borrowed from edge detection in computer vision and reinforcement in neural field theory.

### Why This Exists:

Chirality isn't cosmetic—it determines vascular bias, curvature skew, and lobular emergence. This module operationalizes chirality as a **symbolic field memory** that modulates structure generation.

### Core References:

- L/R symmetry in biology (e.g., Levin, “Left-Right Asymmetry in Embryonic Development”)
  - CODES Chirality Logic (Devin Bostick, 2025)
  - Spinor field alignment in quantum systems
- 

## 3. render\_vein\_field.py

### Conceptual Lineage:

- *Potential fields and gradient flows* — Standard in electrostatics and fluid mechanics.
- *Streamline extraction* — Used in meteorological vector field renderings and vascular flow simulations.

### Why This Exists:

Veins in plants follow nutrient, tension, and coherence gradients. This module defines vascular paths as **streamlines from PAS potential wells**, not noise-sampled polylines.

### Core References:

- Alan Turing's morphogenetic reaction-diffusion models

- Electrostatic field rendering methods
  - PAS as coherence potential (Devin Bostick, 2023–2025)
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## 4. FID\_regression\_tracker.py

### Conceptual Lineage:

- *FID (Free Induction Decay)* — From magnetic resonance imaging (MRI), describing signal loss over time post-stimulation.
- *Reverse field logic* — Echoes in time-reversal symmetry, inverse signal evolution, and thermodynamic decay modeling.

### Why This Exists:

Standard growth models don't model decay meaningfully. This module tracks **phase-aligned unraveling** as a reverse of emission coherence—critical for lifecycle mapping, aging simulation, and entropy-aware field models.

### Core References:

- FID in NMR and MRI physics
- Temporal coherence fields
- PAS inversion modeling (Devin Bostick, 2025)