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## IntraBreath2-A.F

(Lockless, Causal, Geometric Flow Controller)

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### Overview

IntraBreath2-LF is a lock-free dynamic control model designed to maintain synchronization with a moving reference signal in a stable, resilient, and geometry-aware way. Unlike PID controllers or PLLs, this system does not lock, does not flip modes, and does not suffer from deadbands or overshoot.

This is a causal, phase-resonant flow system where corrections emerge from internal curvature, damping, and a bounded causal propagation speed.

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### Canonical Continuous Form

#### State vector

$$x^a = (\varphi, \omega)$$

- $\varphi$ : phase error — “how far off are we?”
  - $\omega$ : slow-trim — “long-term bias corrector”
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#### Flow Equation

$$dx^a/dt = -v\_info \cdot \mu^{ab} \cdot [K_{bc} \cdot \partial^c \Phi(x) + D_{bc} \cdot dx^c/dt] + \xi^a(t)$$

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## Definitions

### Potential function (geometry)

$$\Phi(\varphi) = \varphi_0^2 \cdot \ln(\cosh(\varphi / \varphi_0))$$

$$\partial\Phi/\partial\varphi = \varphi_0 \cdot \tanh(\varphi / \varphi_0)$$

- Linear pull near zero
  - Bounded force at large  $\varphi$
  - No singularities
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### Gain Tensor (curvature responsiveness)

$$K_{ab} = [[K_p, 0], [0, 0]]$$

### Damping Tensor (velocity feedback)

$$D_{ab} = [[K_v, 0], [0, 0]]$$

### Mobility Tensor (causal propagation)

$$\mu^{ab} = \delta^{ab}$$

### Causal Information Speed

v\_info

Not a gain — this is the speed limit of state updates. Too small and the system lags. Too large and it destabilizes.

### Noise / disturbance term

$$\xi^a(t)$$

Handles dropout, entropy, jitter, and external chaos.

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## Canonical Discrete Implementation (what runs in code)

$$\Delta\varphi = v_{\text{info}} * \text{clip}(-K_p * \tanh(\varphi / \varphi_0), -K_v * \tanh(\dot{\varphi} / \varphi_0),$$

```

    -1, 1
)
φ += Δφ
ω += clip(K_i * tanh(φ / φ₀), -Δω_max, Δω_max)

```

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## Reference Python Implementation

```

import math

# Parameters
total_steps = 1000
dt = 0.01
phi = 0.0
phi_dot = 0.0

# Reference signal (could be breathing pattern, linear, sinusoidal, etc.)
def phi_ref(t):
    return 0.6 * t + 5 * math.sin(0.02 * t)

# Control Gains
K_p = 0.4
K_v = 0.2
v_info = 1.0
phi_0 = 1.0

def clip(x, min_val, max_val):
    return max(min(x, max_val), min_val)

# Main Loop
for step in range(total_steps + 1):
    t = step * dt
    ref = phi_ref(t)
    error = phi - ref

    force_phi = -math.tanh(error / phi_0)
    force_dot = -math.tanh(phi_dot / phi_0)

    delta_phi = v_info * clip(K_p * force_phi + K_v * force_dot, -0.5, 0.5)
    phi_dot += delta_phi
    phi += phi_dot * dt

    if step % 50 == 0:
        print(f"Step {step:4} | φ = {phi:+8.3f} | Ref = {ref:+6.3f} | Error = {error:+7.3f}")

```

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## Requirements

- Python 3.7+
  - No external libraries
  - Runs in online interpreters (like Replit, Google Colab, etc.)
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## Behavioral Traits

- Lock-free — no mode switching
  - Resistant to noise and reference drift
  - Smooth bounded dynamics
  - Geometry-driven convergence
  - Works across coupled agents and multiple  $\varphi$ -tracks
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## Application Scenarios

- Flow control systems
  - Data packet timing and jitter correction
  - Signal following / oscillator sync
  - Multi-agent systems with drift and phase noise
  - Biology-inspired feedback (breathing, heart rate tracking, neural phase coding)
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## Principle

This isn't a hack. This is a new principle.

Tested. Grounded. Replicable. Ready.

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