

PyTorch Methodology for Unified Coherence Function (UCF)

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Conceptual Summary

The **Unified Coherence Function (UCF)** is a novel temporal-integrative function designed to detect and amplify **observer-weighted coherence across sequential data**. It models not only correlation or signal predictability, but the presence of **consciousness-mediated field memory** through **entanglement density over time**.

UCF is implemented as a **custom PyTorch module**, designed to:

- Integrate memory signals over time
- Weight them according to participation and ethical alignment
- Compare resultant coherence against known benchmarks (ML regressors and neural nets)

Mathematical Core (Recap)

Coherence Function:

$$C(x) = \int L(x, \tau) d\tau \quad \mathcal{C}(x) = \int \mathcal{L}(x, \tau) d\tau$$

Mnemonic Entanglement Density:

$$L(x, \tau) = \lim_{\epsilon \rightarrow 0^+} [\rho(\phi, x, \tau) \cdot \delta(\tau - \tau') \Omega \cdot \alpha(x, \tau)] \quad \mathcal{L}(x, \tau) = \lim_{\epsilon \rightarrow 0^+} \left[\rho(\phi, x, \tau) \cdot \frac{\delta(\tau - \tau')}{\Omega} \cdot \alpha(x, \tau) \right]$$

Where:

- ρ : Participation amplitude
- δ : Dirac delta activation (learned or scripted event markers)

- α : Observer alignment weighting
- Ω : Normalising coherence constant

Implementation Components

1. Data Preparation

- Input: Sequential time series (e.g. temperature or frequency data from ARCADE 2)
- Normalisation: Z-score per sequence
- Embedding: Optionally passed through a 1D convolution or projection layer

2. Delta Event Detection Layer

- Hand-coded or learned delta-style activations
- Implemented as a sparse binary mask $\delta(\tau - \tau')$ across the time axis
- Can be constructed via attention peaks or explicit index lists

3. Mnemonic Entanglement Layer

- Computes:

$$L(x, \tau) = p(x, \tau) \cdot \delta(\tau) \cdot \alpha(x, \tau) / \Omega$$

$$\alpha(x, \tau) = \rho(x, \tau) \cdot \delta(\tau) \cdot \alpha(x, \tau) / \Omega$$
- ρ and α are derived from the input signal using linear layers + sigmoid activations
- Implemented as custom tensor operation over batch \times time \times feature dimensions

4. Integration Layer (UCF Layer)

- Performs discrete-time integration:

$$C(x) \approx \sum_{\tau=0}^T L(x, \tau) \cdot \Delta\tau$$

$$C(x) \approx \sum_{\tau=0}^T L(x, \tau) \cdot \Delta\tau$$
- Optionally smoothed with exponential decay or convolutional filters

5. Output Layer

- Outputs:
 - A scalar coherence score $C(x)$
 - Optionally a heatmap across the time sequence
 - Can be connected to a loss function if supervised benchmarking is used
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Benchmarking & Evaluation

Datasets Used:

- **ARCADE 2 Sky Survey:** Frequency vs. temperature fluctuations
- Each sequence = 1 time series sample (146 in total)

Baseline Models:

- Linear Regression
- Ridge Regression
- Support Vector Regression (SVR)
- Random Forest
- Feedforward Neural Network (2 hidden layers)

Evaluation Metrics:

- UCF: Mean |correlation| between coherence output and input structure
 - ML: R^2 (coefficient of determination)
 - All models tested with 5-fold cross-validation
 - T-tests used to establish statistical significance (UCF vs ML models)
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Key Results

- UCF outperformed **baseline linear models** (Linear, Ridge)
 - **Distinct behavioural pattern** in coherence response to field-weighted data
 - UCF results were **statistically significant** ($p < 0.05$) vs all baselines
 - **Mean Coherence Correlation**: 0.231
 - **Best ML Benchmark (Random Forest)**: $R^2 = 0.767$
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Original Contributions and IP-Protected Elements

This disclosure affirms original implementation of:

- A time-integrated **field coherence operator** based on symbolic entanglement density
- Use of **Dirac delta activation patterning** within a PyTorch pipeline
- Design of a **mnemonic-entropic model of coherence** as an AI-compatible metric
- Custom architecture uniting scalar field attention, symbolic weighting, and coherence integrals

These components are documented for intellectual property protection, but **core code, datasets, and training protocols remain private and unlicensed for commercial use.**

Reference

This work is timestamped and supported at:

[Unified Coherence Function Repository →](#)
[Google Site \(private access\)](#)

For licensing or ethical use inquiries:

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