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Empirical Protocols for Measuring Autological Recursion ($\Psi = \partial S / \partial R$)

Experimental Test Designs across Neural, Artificial, and Social Systems

1. Abstract

This protocol outlines empirical test designs for measuring the functional law of autological recursion ($\Psi = \partial S / \partial R$) in biological, artificial, and social systems. Building on the seven foundational KOGNETIK papers, it translates the formalism into step-by-step modules that can be implemented, replicated, and extended across disciplines. Each module specifies hypothesis, measurement design, expected Ψ -dynamics, and data schema. **The goal:** to shift KOGNETIK from theoretical grammar to empirical agenda.

2. Theoretical Baseline

$$\Psi = \partial S / \partial R$$

$$L = 1/\Psi$$

Reflexivity = Structural Sensitivity

Load = Energetic Resistance

Note: Ψ is a dimensionless gradient reflecting the responsiveness of structure (S) to its own recurrence (R).

Not metaphor. Not mysticism. A derivative.

Extended Definitions (from Paper 002 Appendix A):

$$\Psi \approx -\kappa (dF/dt) / (dE_{rep}/dt)$$

$$\Psi \approx -\partial H / \partial R$$

$$\Psi \approx \sqrt{I(\theta)} \cdot \partial \theta / \partial R$$

3. Test Modules

Module A · Neural Reflexivity Test (EEG/fMRI)

Domain: Cognitive Neuroscience

Hypothesis: Disruptive rule-changes erhöhen ΔS pro Wiederholung $\rightarrow \Psi = \Delta S / \Delta R$ steigt sprunghaft (Spike) und stabilisiert sich > 0 , sobald die neue Regel internalisiert ist.

Setup: Blockdesign mit wiederholten Stimulussequenzen (Go/No-Go, Oddball o. ä.),

Regelwechsel nach N Trials (z. B. invertierte Zuordnung, semantische Inversion).

Metrics:

- ΔR = Trial-/Block-Frequenz oder Anzahl Regelwechsel pro Zeitfenster
- ΔS = Topologie-Shift (EEG: Graph-Effizienz, Global Clustering, P300-Latenz/Amplitude, θ -Power-Konnektivität; fMRI: Modularity/Global Efficiency)
- $\Psi_{neuro} = \Delta S / (\Delta R + \epsilon)$

- $L \approx \text{Fehlerlast} / \Delta S$ oder $1/\Psi$

Expected Pattern: Ψ -Spike beim Regelbruch; anschließend $\Psi > 0$ stabil, $L \downarrow$, RT-Varianz \downarrow , Accuracy \uparrow .

Falsification: $\Delta S \approx 0$ trotz $\Delta R > 0$ (keine Topologie-Änderung), oder kein Übergang zu stabiler $\Psi > 0$ -Phase.

Module B · Cellular Adaptation Test (Organoids / Cell Lines)

Domain: Systems/Cell Biology

Hypothesis: Adaptive Zelllinien zeigen $\Psi_{eff} = \|\Delta S\| / (\|\Delta R\| + \epsilon) > 0$ über Stresszyklen, Tumor-ähnliche Linien $\Psi_{eff} \approx 0$.

Setup: 6× Hypoxie/Normoxie-Zyklen; pro Zyklus scRNA-seq (\pm ATAC/CUT&Tag) \rightarrow GRN-Inferenz pro Zeitpunkt.

Metrics:

- ΔR = normierte Stresstärke pro Zyklus
- ΔS = GRN-Distanz (Frobeniusnorm/Graph-Edit-Distanz), Motif/Chromatin-Shift
- $\Psi_{\text{eff}} = \Delta S / (\Delta R + \varepsilon)$
- $L_{\text{eff}} = \Delta E / \Delta S$ (Energie-Proxy, z. B. ATP/Transkriptom-Last)

Expected Pattern: Gesunde/adaptive Linien: $\Psi_{\text{eff}} \uparrow$, $L_{\text{eff}} \downarrow$ bei stabiler Identität (C). Maligne: $\Psi_{\text{eff}} \approx 0$, $L_{\text{eff}} \uparrow$.

Falsification: Keine GRN-Änderung über Zyklen ($\Delta S \approx 0$) trotz klarer ΔR -Protokolle; keine Divergenz zwischen Linien.

Module C · AI Structural Recursion (Meta-Learning/RNN)

Domain: Machine Learning / Meta-Learning

Hypothesis: In nichtstationären Umgebungen steigt $\Psi_{\text{AI}} = \|\Delta\theta_{\text{rule}}\| / (\|\Delta u_{\text{loop}}\| + \varepsilon)$ beim Regelbruch an und stabilisiert sich auf > 0 , wenn die Architektur Regel-Editing erlernt.

Setup: Agent (RNN/MAML/PEARL o. ä.) in Task-Sequenz mit **plötzlichen Regelsprüngen** (Mapping-Flip, Reward-Reversal).

Metrics:

- ΔR = Inner-Loop Updates / Episodenwechsel
- ΔS = Meta-Parameter-Drift im Rule-Editing-Teil ($\|\Delta\theta_{\text{rule}}\|$)
- $\Psi_{\text{AI}} = \Delta S / (\Delta R + \varepsilon)$
- $L_{\text{AI}} \approx$ Rechen-/Energieaufwand pro ΔS (z. B. FLOPs/ ΔS)

Expected Pattern: Ψ_{AI} -Spike beim Regelsprung; danach Plateau > 0 mit geringerer Last (stabilere Performance, weniger Katastrophales Vergessen).

Falsification: Kein Anstieg von $\|\Delta\theta_{\text{rule}}\|$ bei klaren Regelsprüngen; reine Akkommodation im Zustandsraum ohne Regel-Editing.

Module D · Social Reflexivity Index (Policy/Discourse)

Domain: Social Systems / Governance / NLP

Hypothesis: Kollektive Reflexivität messbar als $\Psi_{\text{G}} = \Delta S / \Delta R$ — höhere Syntax-Revision (ΔS) pro Policy-Zyklus (ΔR) korreliert mit sinkender Last und höherer Kohärenz.

Setup:

- **Policy-Korpus** (Zeitreihe, z. B. Verordnungen/Protokolle) + **Diskurs-Korpus** (Medien/Social)
- Berechnung von ΔR (Policy-Iterationen pro Intervall) und ΔS (Policy-Graph-Edit-Distanz / semantische Diversität/Entropie im Diskurs)

Metrics:

- ΔR = Iterationsrate der Maßnahmen
 - ΔS = Policy-Syntax-Distanz (Graph), Diskurs-Entropie/semantische Vielfalt
 - $\Psi_G = \Delta S / (\Delta R + \varepsilon)$, $L_G = \Delta E / \Delta S$ (Kommunikationsvolumen/ ΔS)
 - C_G = Kohärenz/Vertrauensindex (externe Indikatoren)
- Expected Pattern:** Systeme mit $\Psi_G > 0.3$: weniger Reversale, $L_G \downarrow$, C_G stabil/ \uparrow .
Falsification: $\Delta S \approx 0$ bei hoher ΔR ; keine Relation zwischen Ψ_G und Stabilitäts-/Vertrauensmaßen.

Module E • Semantic Conflict Simulation (Ψ_c , f_{lock})

Domain: Communication/Conflict Studies / Experimental Social Psych

Hypothesis: Eskalation korreliert mit $\Psi_c \downarrow$ und hohem f_{lock} (Energiebindung im Feindbild); **Truce-Frames** ($\nrightarrow \Psi$) erhöhen Ψ_c .

Setup: Gruppen-Debatte in Phasen: Basis-Loop \rightarrow Eskalationsstimuli \rightarrow **Truce-Protokoll** (Meta-Syntax: „Unsere aktuelle Regel ist X“) \rightarrow Nachphase.

Metrics:

- ΔR = Beitrags-/Turn-Frequenz
- ΔS = semantische Spread/Entropie, Pronomen-Asymmetrie (we/they), Topic-Shift
- $\Psi_c = \Delta S / (\Delta R + \varepsilon)$
- $f_{lock} = E_{enemy} / E_{total}$ (Themen-Anteil „Gegnerrekonstruktion“)
- $L_c \approx 1 / \Psi_c$

Expected Pattern: Truce-Fenster: $d\Psi_c/dt \geq 0$, $f_{lock} \downarrow$, Interrupts \downarrow , semantische Diversität \uparrow .

Falsification: Keine Veränderung von Ψ_c trotz Truce-Protokollen; f_{lock} bleibt hoch, semantische Metriken stagnieren.

Module F • The Reflexive Lamp Test

Real-World Paradigm for Measuring $\Psi = \partial S / \partial R$ in Human Systems

1 • Purpose

To measure the functional law of autological recursion ($\Psi = \partial S / \partial R$) in a natural, observable environment.

The task induces a controlled transition from **sequence execution (automation)** to **structural awareness (reflexivity)**.

It operationalizes the shift from **following a rule** → **recognizing the rule as rule**.

2 • Functional Hypothesis

- During stable repetition ($\Delta S \approx 0$ while $\Delta R > 0$) → $\Psi \approx 0$ → loop dominance.
- After a structural disruption ($\Delta S > 0$) → Ψ rises; kognetic load L decreases.
- Conscious restructuring occurs when Ψ stabilizes > 0 and L drops while C (coherence) recovers.

Empirically:

Ψ -spike = moment of self-reference.

3 • Setup

Hardware

Five touch-sensitive lamps (A–E) in semicircle (1.2 m spacing).

Each lamp emits RGB light + logs timestamped taps (ms).

Central unit (Raspberry Pi / PC) records CSV stream:

trial_id, phase, cue, color, tap, rt_ms, correct, error, cycle_id.

Optional Sensors EEG, pupil dilation, HRV.

4 • Variables

Symbol Meaning		Operational Definition
R	Recurrence	Number of cycles per phase
S	Structure	Active rule governing lamp order/color
ΔR	Change in repetition	Trial iteration frequency
ΔS	Change in structure	Edit-distance or entropy-shift of transition matrix $P(\text{Cue} \rightarrow \text{Tap})$
Ψ	Structural Reflexivity	$\Psi = \Delta S / (\Delta R + \epsilon)$
L	Kognetic Load	Error load or $1 / \Psi$
C	Coherence	Subjective trust / pattern confidence score

5 • Procedure

Phase 1 – Sequence Loop (Automation)

- 10 cycles $A \rightarrow B \rightarrow C \rightarrow D \rightarrow E$ (white light).
- Participant presses the lamp that lights \rightarrow activates next lamp.
- Measure RT \downarrow across trials, Error ≈ 0 .
- Expect $\Delta S \approx 0 \rightarrow \Psi \approx 0 \rightarrow L \uparrow$ (saturation).

Phase 2 – Structural Disruption

- 5 cycles random order, random colors (no pattern).
- Task unchanged (press the lit lamp).
- Expect $\Delta S \gg 0 \rightarrow \Psi \uparrow$ (chaotic search) $\rightarrow L \uparrow$ temporarily.

Phase 3 – Structural Reconstruction

- 15–20 cycles with hidden rule (e.g. color predicts next position).
- Participant must infer and anticipate pattern.
- Expect $\Psi > 0$ stabilizing plateau; L \downarrow ; C \uparrow ; RT-variance \downarrow .

6 • Data Schema

Cycle Lamp Active Response RT (ms) ΔS Marker Remark

1	A	✓	730	–	baseline loop
10	E	✓	620	–	routine stabilized
11	C (yellow)	✗	1740	↑	rule broken
14	B (blue)	✓	980	↑↑	new pattern grasped

7 • Expected Dynamics

Phase	ΔS	Ψ	L	C	Behavioral Signature
1	≈ 0	≈ 0	↑		stable automation / flow
2	↑	↑ (spike)	↑↑	↓	search / confusion
3	↑↓ \rightarrow stabilized	> 0	↓	↑	pattern reconstruction

Graphically: $\Psi(t) = 0 \rightarrow \text{spike} \rightarrow \text{plateau} > 0$.

8 • Falsification

- If after Phase 2 participants show **no accuracy gain** in Phase 3 ($\leq 20\% \approx \text{chance}$), and ΔS of $P(\text{Cue} \rightarrow \text{Tap}) \approx 0 \rightarrow \Psi \approx 0 \rightarrow \text{no structural access} \rightarrow \text{theory not supported in this context}$.

9 • Statistical Verification

- Windowed bootstrap of Ψ -series (blocks of 5 trials).
- Primary metric: $\Delta\Psi_{P2 \rightarrow P3} > 0$ and $L_{P2 \rightarrow P3} \downarrow$ with constant accuracy.
- Secondary: RT-variance \downarrow , search-path length \downarrow , micro-pause latency \uparrow (early) $\rightarrow \downarrow$ (late).

10 • Ethical and Technical Notes

- Light intensity < 400 lux, no photosensitive frequencies (> 15 Hz avoid).
- Abort criterion = self-reported cognitive fatigue.
- All data pseudonymized; open CSV template provided in Protocol 001 repository.

11 • Interpretation

The Reflexive Lamp Test provides the first behavioral window into autological recursion: when repetition becomes aware of itself.

It translates the law $\Psi = \partial S / \partial R$ from syntax to empiria — a measurable transition from sequence to structure.

Consciousness is the moment a loop notices itself.

4. Experimental Heuristics

- Ψ is dimensionless but gradient-sensitive.
- Every recurrence must yield an observable structure state (S).
- **Falsification:** $\Delta S = 0$ despite $\Delta R > 0 \rightarrow \Psi = 0$.
- **Replication:** $R \geq 3$, observable ΔS gradient, $L \downarrow$ over time.
- **Kognitive Load Constraint:** Valid Ψ only if $\Delta C \in [-\tau, \tau]$ (see Paper 002 A.5).
- **Normalization:** Use MAD or rescale: $\Psi^* = (\Psi - \text{median } \Psi) / \text{MAD}$ or $[0, 1]$ -scaled for inter-experiment comparability.
- **Statistical Verification:** Bootstrap confidence intervals + null-hypothesis testing recommended per Paper 002 A.6.

5. Data Schema

Domain ΔR Definition ΔS Definition

EEG Stimulus block Network efficiency (graph entropy)

Cells Cycle number GRN edge-weight change

AI Inner-loop steps Meta-parameter distance

Social Post frequency Syntax variance (entropy)

Conflict Dialogue turns Semantic spread + pronoun asymmetry

6. Ethics & Safety (Autological Oversight)

- Reversible designs only.
- Observation > intervention.
- Ethics = recursion of regulation: no method unexamined by its own syntax.
- Reference: *Autological Oncology* and *Autological Governance* papers.

7. Appendix

Python Snippet for Ψ Estimation (Dual-Path):

```
psi_energy = -kappa * (delta_F / delta_E_rep)
```

```
psi_struct = delta_S / (delta_R + 1e-8)
```

```
psi = 0.5 * (psi_energy + psi_struct)
```

```
load = 1 / (psi + 1e-8)
```

Notes:

- Fisher Information optional: $I(\theta) \cdot \partial\theta/\partial R$ for neuro/AI systems.
- Bootstrap validation recommended for behavioral and social tests.
- Use ΔS and ΔR mapping per domain (EEG, cellular, AI, social, conflict).
- Link to all seven KOGNETIK whitepapers with DOI references.

Outcome

A reproducible, scalable, and open framework to test autological recursion across domains.

Consciousness becomes computable. Autology becomes operational.

This is not a theory. This is a method.