

The Birth, Life, and Death of Internal Organization

Yuxuan Deng

dengyx0211@163.com

Independent Researcher <https://orcid.org/0009-0000-1531-2152>

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The Birth, Life, and Death of Internal Organization

Anonymous

Abstract

Learning systems change internally all the time, yet it remains unclear when such change constitutes an organization: when one comes into being, whether two differently built systems carry the same one, and what remains when it dies. Internal structure is usually compared by representational similarity, by what a probe can decode, or by alignment to a hypothesized algorithm [1, 2, 3, 4]. Here we identify internal organization by what admissible internal interventions do to an independently named future consequence; the resulting *Consequence Organization*, a construction closely related to causal abstraction, is *endogenous* when the system’s own consequence history forms it. We show that, across naturalistic and controlled systems, such organizations formed under stable but not shuffled consequence, persisted and reorganized, could be erased, restored and transplanted, and could lose their causal role while carrier-associated readouts or residual carrier content remained. Their identity transferred across independently trained seeds and, in most directions, across architectures, but failed across an intentionally incompatible cross-consequence family even where internal operators remained nearly identical. In the tested systems, internal organization thus has a measurable lifecycle and a consequence-relative identity that neither decodability nor internal-operator similarity alone tracks.

When change becomes organization

Adaptive systems alter parameters, representations and behavior as they learn. Classical state-space models, hidden-state models and system identification compress dynamics into state variables chosen for filtering, prediction or dynamical sufficiency [5, 6, 7]; predictive-state and computational-mechanics approaches instead define state through equivalence of future distributions [8, 9, 10, 11]. For learned internal structure, sameness is usually judged by representational similarity or stitchability between networks [12, 1, 2, 13], by what a probe can decode [3], or by alignment to a hypothesized high-level algorithm [14, 4]. A feature may nevertheless remain decodable after its former role has disappeared, and the same causal role may be instantiated by different coordinates. The question here is therefore not whether a system changed internally, but whether previously interchangeable *internal interventions* became different with respect to an independently fixed consequence, and whether that difference has a lifecycle and an identity of its own.

Following the interventionist distinction between observation and causal manipulation [15, 16], fix a realization r at time t , a prospectively frozen family of admissible internal interventions $\mathfrak{I}_t^{(r)}$, and a named consequence space $\mathcal{Q}$. The real future consequence operator is

$$\mathfrak{C}_t^{(r)} : \mathfrak{I}_t^{(r)} \longrightarrow \mathcal{Q}. \quad (1)$$

Two interventions are organizationally equivalent when they produce the same audited consequence,

$$i \sim_{\mathcal{Q},t}^{(r)} j \iff \mathfrak{C}_t^{(r)}(i) = \mathfrak{C}_t^{(r)}(j), \quad (2)$$

giving the local realized quotient

$$E_t^{(r)} = \left(\mathfrak{I}_t^{(r)} / \sim_{\mathcal{Q},t}^{(r)}, \bar{\mathfrak{C}}_t^{(r)} \right). \quad (3)$$

The quotient is not yet the fundamental object, because its intervention coordinates belong to one realization. We therefore identify two realizations when there is a bijection between their quotient classes that preserves the anchored consequence map. Two realizations are *compatible* when they are evaluated with the same named consequence space $\mathcal{Q}$ and an admissible intervention family of the same type, so that their consequence maps can be compared class by class; in the experiments below they differ in seed or architecture while task, outcome and action semantics are shared, and the cross-consequence test lies outside this definition by design. Writing this consequence-preserving isomorphism as $\cong_Q$, the object studied here is

$$\mathcal{O}_Q(t) = \left[\mathfrak{I}_t^{(r)} \rightarrow E_t^{(r)} \rightarrow \mathcal{Q} \right]_{\cong_Q}. \quad (4)$$

We call $\mathcal{O}_Q(t)$ a *Consequence Organization*. If the organization was absent before adaptation and is subsequently created by the system’s own consequence history, we call it an *Endogenous Consequence Organization*. A

measured subspace, layer stack, low-rank coupling or complete parameter vector may realize or encode $\mathcal{O}_Q$ without being identical to it. Conversely, the definition does not assert metaphysical independence from the complete physical state: any deterministic realization can be represented in a sufficiently rich state description. The empirical claim is instead that the same consequence organization can be identified and manipulated across distinct realizations while coordinate identity is neither necessary nor sufficient.

Formally, the local quotient is a causal coarsening of the intervention family [17, 18], akin to causal states and to bisimulation in Markov decision processes [8, 19], and identification across realizations is an exact causal abstraction of each realization to a common high-level model whose variables are the consequence classes [20, 21, 14, 4]. The contribution is therefore not a new notion of abstraction, nor any one of these formal operations. Here they are combined in a time-indexed empirical construction in which admissible internal interventions are partitioned by a named future consequence, and the resulting organization is followed through formation, persistence, reorganization, death and transport.

The finite closure signature σ_Q over the frozen probe family is only an operational certificate of consequence anchoring at the tested resolution. We do not assume $\sigma_Q \equiv \bar{\mathcal{C}}$ unless the probe family is separately shown to be separating. The object is also explicitly Q -relative: realization can change while the consequence semantics remain fixed, but no invariance across arbitrary consequence laws is assumed.

Birth is a newly resolved consequence distinction; persistence is preservation of that distinction; reorganization changes its anchored incidence; and death is the merger of a previously active distinction under the named consequence readout. Identity is then tested by intervention: erase should remove the consequence mapping, restore should recover it, and transport should carry the mapping into a compatible realization. Figures 1–4 keep the naturalistic recommendation data and controlled neural-model experiments in separate evidence fields while following these shared causal questions. The cross-architecture and cross-consequence direction rows provide distinct tests of implementation change and consequence-relative identity.

Birth under stable consequence

The flagship experiment used the random-policy component of KuaiRand, a public recommendation dataset containing platform-randomized video replacements inserted into standard recommendation feeds [22]. The released random-policy subset contains 1,186,059 interactions from 27,285 re-indexed users. Our analysis partitioned users into 16,373 development, 5,456 locator and 5,456 frozen-confirmation users. A generic 18,891-parameter MLP received observations and scalar consequence but no organization label, phase identifier, router, expert bank or dedicated memory.

The naturalistic evidence comes from distinct archived experiments, retaining pre-absence, formation, persistence, reorganization, polarity reversal and Null readouts. The later unpaired cross-consequence archive supplies the two task/model families in Figs. 1b and 2a; it is separate from the original long-view manipulation and cross-seed archives. Matched shuffled-consequence rows preserve the relevant marginal data and training budget while breaking action–outcome correspondence. When the always-present release mechanism was removed, unsupported C remained causally active in the retained control rows. These rows separate recruitment from generic training and loss of support from automatic death. A separate controlled-language-model archive supplies executed formation rows and phase-resolved histories. Figure 1 combines the consequence schematic and naturalistic run field with the controlled formation and lifecycle panels, so the birth claim is shown from design, unit-level evidence and phase-resolved histories rather than from a single trajectory. The controlled rows are not pooled statistically with the KuaiRand lifecycle.

Figure 1c keeps each controlled history as one paired unit and displays ordinary/shuffled history–world contrasts $I = x(M_\alpha) - x(M_\beta)$, where $x(M) = \text{Acc}_\alpha(M) - \text{Acc}_\beta(M)$. Figure 1d shows x at four recorded phases for three 3B histories. Persistence and reorganization passed in all three, but ordinary release passed in 2/3 and stack release in 1/3; the partial phase series is retained without retrospective rescue. Complete gate-profile rows are provided in Source Data.

Role death with carrier persistence

Figure 2a separates rank-four C-channel deletion effects from trace-probe AUC across twelve archived runs. C and Null are different checkpoints of the same run. The probe uses hidden features plus action indicators, and shuffled runs that never formed C reach the same AUC range; its AUC therefore neither isolates hidden-state memory nor indexes the C organization.

In a separate small-network experiment, three learner/world cases each supplied four continuations cloned from their B endpoint and optimizer state (Fig. 2b). Three Null regimes remove B support but differ in trainable layers; continued B retains it. The 60 records track old-mapping role power separately from residual carrier energy at every recorded step. Role power collapsed within 50 steps in every Null branch, whereas carrier content was case-dependent: it persisted or grew in the quadratic case, decayed with the role in the sine case and was near zero throughout in the interaction case. Loss of the audited causal role therefore need not eliminate energy in the former carrier, but it does not always preserve it.

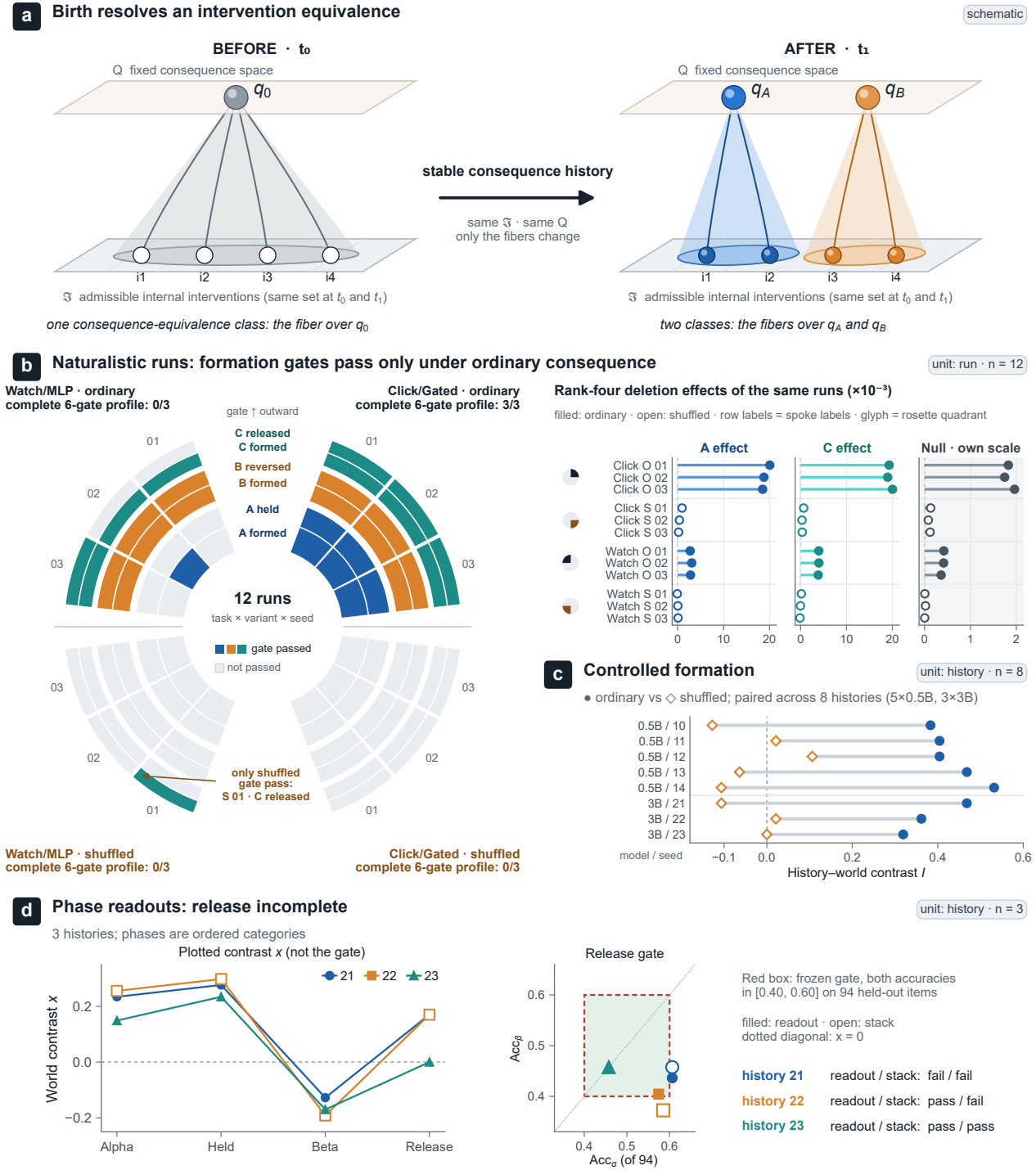
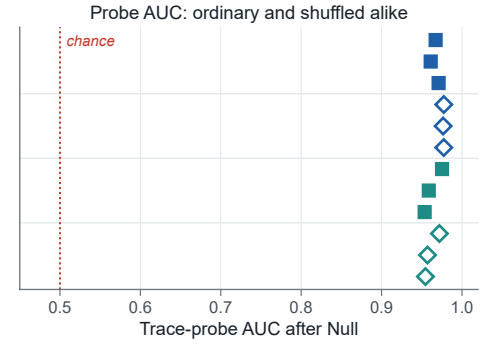
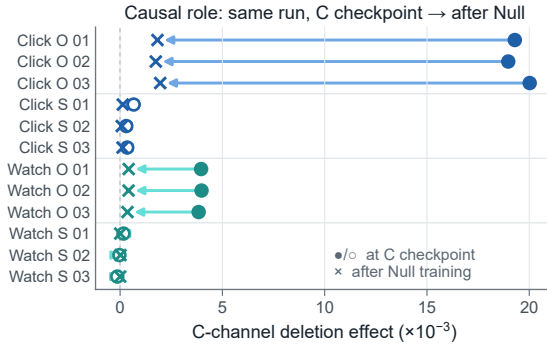


Figure 1. Formation contrasts and phase histories retain their experimental boundaries. **a**, Schematic fiber split: admissible internal changes $\mathcal{I}$ begin consequence-equivalent (one fiber over q_0) and separate under fixed Q (fibers over q_A and q_B); positions are not measurements. **b**, Twelve naturalistic runs (two task/model families, ordinary/shuffled variants, three seeds). Task is left/right (Watch/MLP, Click/Gated), variant top/bottom; each sector is one run and six outward equal-area lifecycle gates (coloured, passed; grey, not). The only shuffled pass is Watch S 01 at C release. Adjacent points show the same runs' raw rank-four deletion effects (A/C share an axis; three slightly negative shuffled values are retained). Complete lifecycle gates passed in ordinary Click/Gated (3/3) and none in Watch/MLP (0/3). **c**, Within eight controlled histories (five 0.5B, three 3B), conditions are paired and connectors show raw history–world contrasts I . **d**, Three 3B histories show $x = \text{Acc}_\alpha - \text{Acc}_\beta$ at four phases (left) and release accuracies (right). Release requires both accuracies in [0.40, 0.60] on 94 held-out items, not a threshold on x ; filled/open points are readout/stack. Histories 21 and 22 share $x = 16/94$: 21 fails at $\text{Acc}_\alpha = 57/94$ (one item above band), whereas 22 passes at $\text{Acc}_\beta = 38/94$ (lowest passing count); its stack fails at 35/94. Readout release passed 2/3 histories and stack release 1/3; complete release is not established. Experiments are not pooled.

a Consequence effects and the trace-probe readout

unit: run · n = 12

12 naturalistic runs; blue: Click, teal: Watch; filled: ordinary, open: shuffled

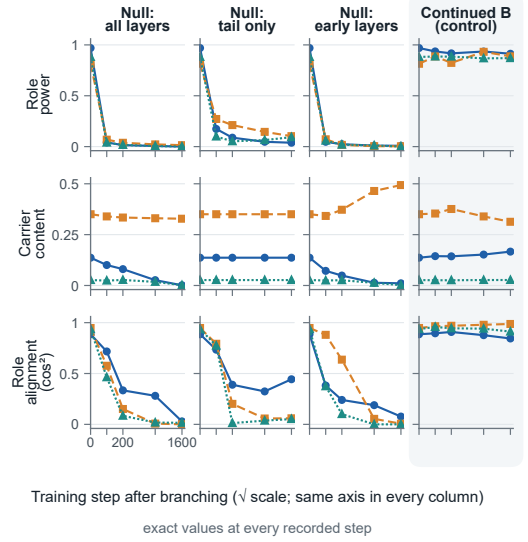
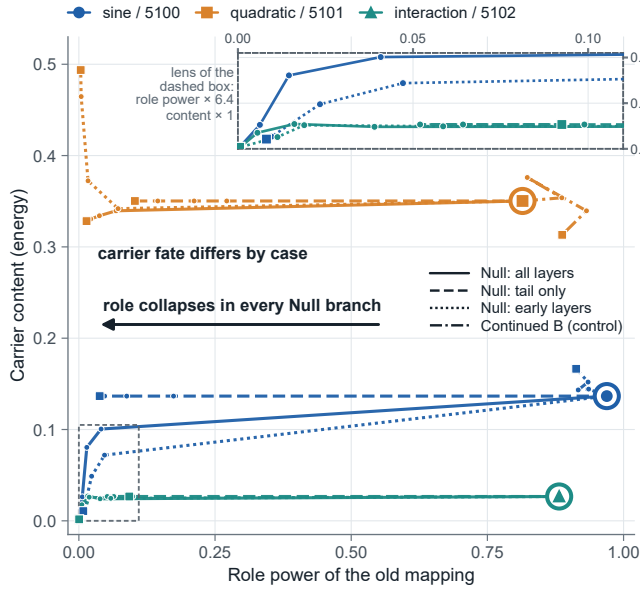


Shuffled runs never formed C, yet reach the same AUC range (0.954–0.977): the probe, which also receives action indicators, does not index the C organization or isolate hidden-state memory.

b Role dies, carrier may remain: four continuations cloned from each B endpoint

unit: case · n = 3

3 learner/world cases (one seed each) × 4 branches × 5 recorded steps (0, 50, 200, 800, 1600); lines join one branch only



ring: step 0, the B endpoint shared by all four branches · dots: steps 50–800 · square: step 1600 · dashed box: region shown in the lens

Figure 2. Consequence effects and probe information are separate readouts. a, Twelve naturalistic runs show C-channel deletion effect at checkpoint and after Null in each run (left; ordinary Click, $\sim 0.019 \rightarrow \sim 0.002$; Watch, $\sim 0.004 \rightarrow \sim 0.0004$) and trace-probe AUC after Null (right, 0.954–0.977; chance 0.5). Shuffled runs that never formed C reach the same AUC; the probe includes action indicators and does not isolate hidden-state memory. Colour denotes task; filled/open denote ordinary/shuffled. **b**, Three learner/world cases (sine, quadratic, interaction; one seed each) branch from a shared B endpoint into three Null regimes and continued B. Left, every checkpoint plots role power against carrier content; the inset expands only the near-zero role-power axis ($\times 6.4$). Role power collapses in every Null branch, while carrier content persists or grows (quadratic), decays with role (sine, all-layer/early-layer Null), or stays near zero (interaction). Right, each readout is followed across five steps (0, 50, 200, 800, 1600) on a square-root step axis; lines join checkpoints within the same case/branch (60 records). Tail-only Null freezes early layers, so carrier content is constant by design; early-only Null freezes the tail. Branches/checkpoints are repeated records, not independent cases.

Erase, restore and transplant

Reversible interventions were frozen in the original long-view recommendation ecology. Rank-four hidden-to-action couplings were measured on locator users, with restore estimated on a disjoint fold; confirmation used 5,456 held-out users and 243,254 exposures. Figure 3b retains ordinary/shuffled seed profiles from a separate gated lifecycle archive. Figure 3a shows 50 fresh controlled worlds, all of which passed the 13 frozen gates; each world contributes one point to every operation, so the operations are repeated measurements within world, not additional worlds, and are not pooled with naturalistic seeds. The views pair the gated frozen-locator fraction with the ungated full-signature residual of the same world: every world lies inside the frozen locator interval, while full-signature residuals remain 0.68–1.83 of the native gap. Residual ratios are descriptively larger in worlds with larger locator overshoot and smaller native gap; because that gap is the ratio denominator, neither co-variation is interpreted as an independent association.

Across five ordinary confirmation seeds, the naturalistic manipulation archive gives mean effects for C in its native state, after erase, and after restore of 0.008290, 0.000454, and 0.008285, respectively. Transplanting C into Null raises the mean effect from 0.000263 to 0.007940; restoring Null returns it to 0.000263. On a common $+B$ axis, exchanging the positive- B module into a negative- B recipient gives +0.006151, whereas exchanging the negative- B module into a positive- B recipient gives -0.010840 ; restoring recipient modules recovers the original signs. The accompanying Source Data retains all ten archived run-level rows and their twelve frozen gate outcomes per run; the B -exchange signs are expressed on the common $+B$ axis.

Identity across realizations

Cross-seed transport tested 20 directions among five MLP realizations using locator consequence signatures; confirmation users were excluded from alignment (Fig. 4a). The architecture test retains all 30 directions and its two failures (Fig. 4b). The 18-direction unpaired cross-task test jointly changes user domains, outcomes, action/context maps and architecture (Fig. 4c); near-one internal operator cosines do not rescue the failed full identity gate. These fields separate compatible-realizer transport from the cross-consequence boundary.

A conditional lifecycle dynamics

The lifecycle gates above were defined and frozen independently of any dynamical model. Nevertheless, the observed transitions admit a common theoretical compression. At a fixed audit resolution and within one realization, let $\mathfrak{C}_t$ denote the intervention–consequence map after update t , and let Δ_t denote the currently supported consequence pressure. Relative to a paired evolution in which that pressure is absent, write

$$\mathfrak{C}_{t+1} - \mathfrak{C}_t = R_t(\Delta_t) + U_t(\mathfrak{C}_t). \quad (5)$$

Here R_t is the consequence-dependent recruitment or redirection term—the change attributable to supported consequence relative to its paired pressure-removed evolution—and U_t is the evolution of the same organizational realizer in the absence of that pressure. The decomposition is exact bookkeeping for the paired update; it is not, by itself, a lifecycle theorem. Predictive content enters through independently testable constitutive conditions on recruitment, contraction and leakage, together with their finite-time consequences.

The dynamics acts on $\mathfrak{C}_t$, not directly on $\mathcal{O}_Q$. After each update the local quotient $E_t^{(r)}$ is induced from the updated consequence map, and consequence-preserving identification across compatible realizations yields $\mathcal{O}_Q(t)$. A recruitment term that separates two previously consequence-equivalent interventions produces a fiber split and therefore birth. Continued support can preserve the split. Changed consequence can redirect the anchored incidence or reverse its sign. When support is removed and unsupported evolution contracts the active consequence contrast, formerly distinct interventions merge under the frozen audit and the organization dies. Leakage or overlapping support can instead produce coexistence or a nonzero residual floor.

The constitutive conditions are non-vacuous. Supported consequence without sufficient recruitment need not produce formation; removing support need not produce death when leakage or persistence dominates; and a one-step contraction estimate does not by itself certify contraction throughout an entire release phase. These excluded regimes are observed in the control record and prevent the dynamical form from merely restating the lifecycle labels after the fact. In transparent adaptive classes, constitutive quantities can be estimated before an unseen phase and used in frozen prediction of lifecycle regime, but quantitative transfer requires realization-specific calibration. The present deep-neural and naturalistic experiments do not claim parameter-free universal kinetics or prospective recovery of all phase-domain coefficients from a compact state description.

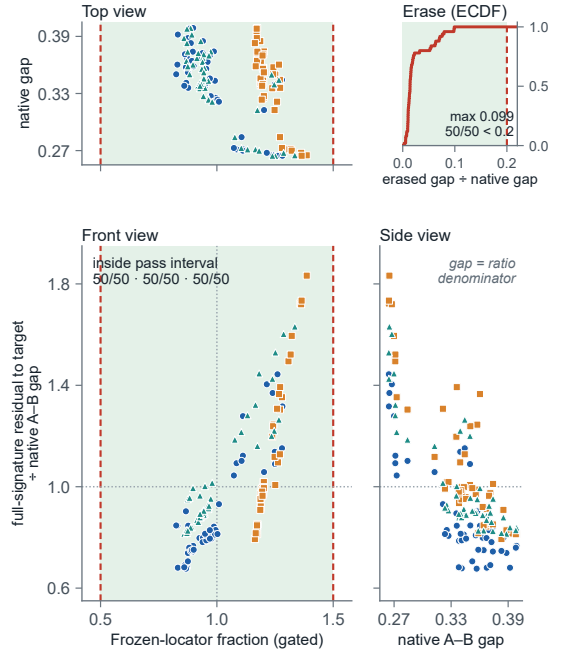
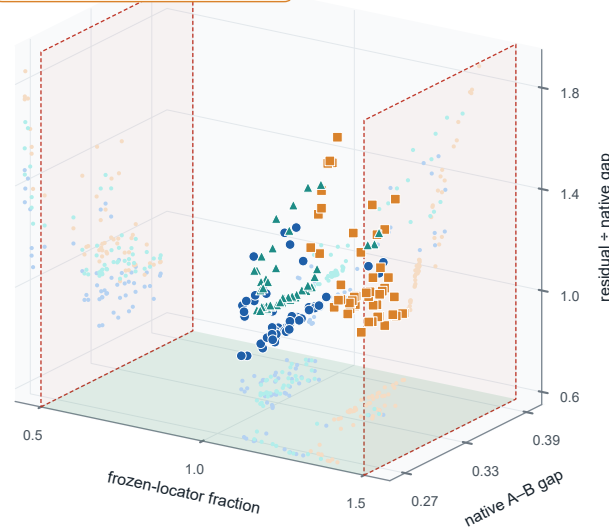
a Frozen-locator gate and full-signature residual in 50 controlled worlds: a three-variable view

unit: world · n = 50

Three measured axes per world and operation; the same 50 worlds recur in every operation. All 50 passed all 13 gates.

● Transplant → A ■ Swap → B ▲ Restore → A ▨ frozen pass interval 0.5–1.5

native gap also normalizes the residual

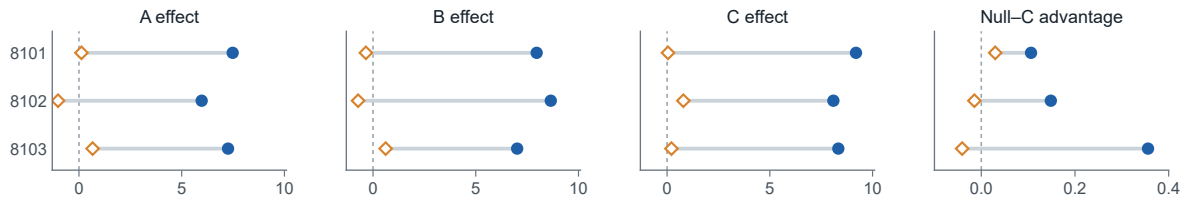


Linked views share axes with their neighbours (top/front: locator fraction; front/side: residual). The three operations of each world share its native gap (same height in top and side views). Residuals use all twelve $\log(1+\text{gradient-energy})$ readouts; dotted lines: target ($x = 1$) and native gap ($y = 1$). A locator fraction near 1 does not imply full-signature identity; the native gap is the ratio denominator. Its inverse relation with the residual (side view) is descriptive, not an independent association.

b Naturalistic paired contrasts: ordinary versus shuffled

unit: seed · n = 3 (paired)

3 paired gated-model seeds; effects $\times 10^{-3}$. First three axes share a scale; Null-C uses a separate scale.



● ordinary ◇ shuffled (paired within seed) Naturalistic seeds are separate from the 50 controlled worlds and are not pooled with them.

Figure 3. Intervention readouts retain their world and seed units. **a**, Each point is one of 50 controlled worlds under one operation; the same worlds recur in every operation, so the panel holds 50 units, not 150 or 200. Three measured variables are shown for transplant, swap and restore: the gated frozen-locator fraction (green: frozen pass interval 0.5–1.5; 1 = target reached along the locator), the ungated full-signature residual distance to the A, B and A target, respectively, divided by the native A–B gap, and that native gap. Left, an orthographic three-dimensional view draws the frozen interval as two criterion planes and every observation also as its projection on floor and walls; right, linked top, front and side views share their axes and give exact readings; the three operations of a world share its native gap. The native gap is the ratio denominator; its inverse relation with the residual is therefore descriptive and is not interpreted as an independent association. Signatures contain $\log(1+\text{gradient energy})$ readouts for six frozen future tasks. The erase inset is the right-continuous empirical distribution of the erased-state gap divided by the native gap (frozen criterion < 0.2). All 50 worlds passed all 13 frozen gates; a locator fraction near 1 is not full-signature identity. **b**, A, B and C deletion effects and Null–C advantage relative to uniform policy are paired across ordinary/shuffled training within three naturalistic gated-model seeds. The naturalistic experiment is separate from panel a.

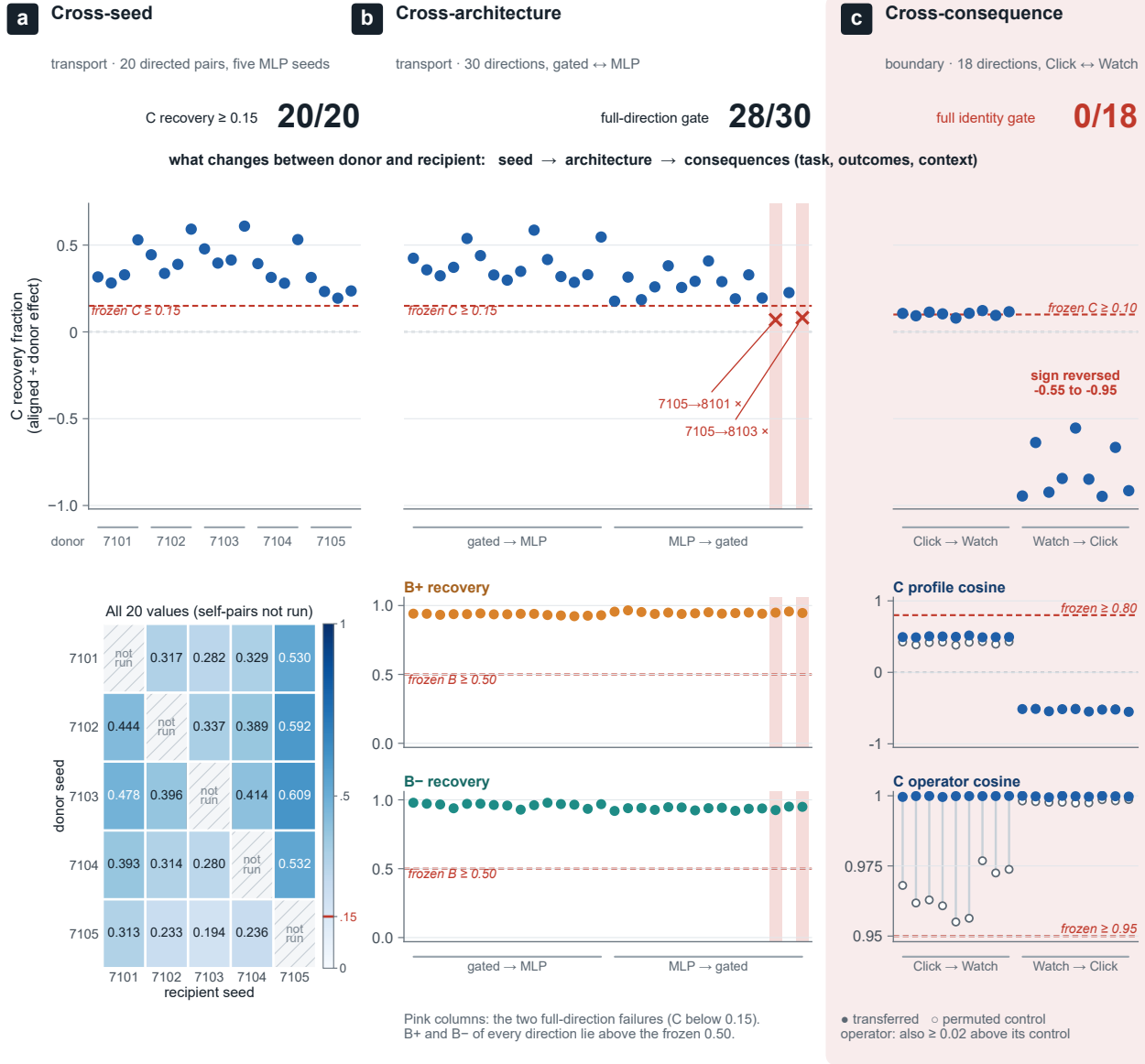


Figure 4. Cross-realization recovery and the cross-consequence boundary remain separate tests. Top, C-recovery fraction (aligned/donor effect) for all 68 directed transfers, with each frozen criterion; categorical columns are not connected, and nested readouts align beneath each transfer. **a**, All 20 cross-seed directions among five MLP realizations; no self-pairs. The 5×5 matrix gives exact off-diagonal values; criterion 0.15 appears on axis and colour scale. **b**, All 30 architecture directions with nested C, B^+ and B^- recovery readouts; criteria are C 0.15 and B 0.50 (red dashes). Two full-direction gates fail below the C criterion; several passes are near it. Directions are not independent seeds. **c**, All 18 cross-task directions are assessed against C recovery (criterion 0.10), C-profile cosine, C-operator cosine and permuted controls; failed C-control and B-profile components are in Source Data. Click $\rightarrow$ Watch recovers 0.08–0.12 of the donor effect; Watch $\rightarrow$ Click reverses sign (-0.55 to -0.95). Profile cosines remain below 0.80 and Watch $\rightarrow$ Click profiles match permuted controls. Operator cosines are near 1, but so are controls, failing the required 0.02 margin. No direction passed; near-one operator similarity does not establish consequence-profile identity.

Discussion

The experiments identify an adaptive object by causal organization rather than by location. Stable consequence can split a previously coarse intervention equivalence; continued support can preserve that split; changed consequence can reorganize or reverse it; release can merge a former distinction; and certified realizers can be erased, restored, transplanted and exchanged. The same carrier can therefore realize different consequence organizations, while distinct carriers can realize the same $\mathcal{O}_Q$ under a fixed consequence semantics.

This object sits among established notions rather than outside them. State-space and system-identification variables summarize dynamical sufficiency [5, 6, 7]; causal states and predictive-state representations group histories by future prediction [8, 9, 10, 11]; causal representation learning seeks variables with causal semantics [23]; probes, unit analyses and activation patching reveal encoded, localized or mediating structure [3, 24, 25, 26, 27]; and causal abstraction tests whether components realize a high-level model [14, 4]. Consequence Organization uses an intervention-preserving identity criterion closely related to causal abstraction, and its erase, restore and transplant operations are patching interventions. The contribution is not a new general notion of abstraction or a new intervention tool, but an empirical account of how a consequence-anchored organization forms, persists, dies, is restored and is transported, and of where its identity fails across consequence-incompatible transfers. A state model may encode this object; encoding does not supply that account.

Two empirical separations show that common decodability and internal-similarity criteria are insufficient to track this identity in the tested systems. The first concerns decodability: after Null training the C organization's causal role collapsed, yet a trace-probe readout stayed in the same range as in shuffled runs that never formed C, so that readout does not index the organization (Fig. 2a); in a separate assay, residual carrier content persisted, and in one learner grew, after the role had died (Fig. 2b). Carrier-associated readouts can therefore outlast organizational death, extending to whole organizations a dissociation that amnesic probing has shown for individual features [28]. The second concerns similarity: transport between independently trained seeds and architectures has an analogue in model-stitching tests of cross-model compatibility [12, 2], but in the cross-consequence test internal operator signatures remained nearly identical while consequence identity failed in all 18 directions and Watch→Click recovery reversed sign (Fig. 4c). Together with cross-seed and cross-architecture transport, these results support bounded realization invariance within specified consequence semantics, not a universal or law-independent ontology.

The same separations state how the proposal could fail. If every identification across realizations were already predicted by realization-local coordinate equality, so that consequence-preserving equivalence added no invariance beyond matched state coordinates, the object would have no independent empirical content; the cross-seed and cross-architecture transfers instead succeed without coordinate identity. If, under fixed consequence semantics, transport between compatible realizations failed systematically despite correct identification of each local quotient, the cross-realization identity criterion would be refuted; those fields retain every direction, including the two architecture failures. If loss of the audited role always coincided with loss of carrier content, the distinction between organization and carrier would be unnecessary; Fig. 2 records carrier persistence as well as carrier decay after the same role collapse. Finally, the definition predicts the cross-consequence failure: a full identity pass obtained there from internal operator similarity alone would contradict its Q -relativity.

If this account holds, organization need no longer be identified with a location, a coordinate system or a fixed mechanism; it can be identified experimentally by a consequence-preserving causal identity across realizations. The questions then shift from where a representation resides to which realizations instantiate the same organization, when an organization is born and when it dies, what survives its death, and whether it can be constructed in a different system.

The intervention evidence acts on certified physical realizers of $\mathcal{O}_Q$ at the stated audit resolution. The cross-realization result therefore concerns preservation of consequence-class incidence and anchoring under compatible realization changes; it does not require coordinate identity, microscopic uniqueness or invariance across different consequence laws.

The conditional recruitment–release dynamics adds a separate claim from object identity. Its update decomposition is exact, but its predictive force resides in the independently testable constitutive inequalities. Transparent systems show that those inequalities can have prospective regime-level bite after calibration; the headline neural and naturalistic experiments establish the lifecycle and intervention identity rather than a universal kinetic parameterization. The remaining generative problem is to determine, for complex adaptive systems, which learning rules make those constitutive quantities prospectively measurable and which consequence organizations are reachable. The empirical result here is therefore bounded but stronger than a catalogue of phases: Consequence Organization has a measurable lifecycle, an interventionally testable identity, cross-realization persistence under fixed consequence semantics and a non-tautological conditional dynamics for its transitions.

Methods

Naturalistic random-policy data and consequence definition

KuaiRand documents a random-exposure collection stage in which each recommended video had equal probability of being replaced by a random video sampled from an item pool, with approximately 0.37% of interactions replaced in the released logs [22]. The random-policy file contains all interactions resulting from the random intervention. User and item identifiers are re-indexed. The SE analysis used 1,186,059 random-policy interactions from 27,285 users and partitioned users into development, locator and frozen-confirmation subsets before confirmation. The principal consequence was the released `long_view` field, defined by KuaiRand as complete viewing for videos of at most 18 seconds or at least 18 seconds of play for longer videos. This study is a secondary analysis of a dataset containing platform-randomized item replacements; it is not described as a trial randomized by the present author. The two task/model families in Figs. 1b and 2a use long-view outcomes on even-ID users (Watch/MLP) and click outcomes on odd-ID users (Click/Gated), with different opaque action assignments and context weights. A, B and C denote frozen action-group reward channels, not supplied organization labels; Null is audited against the old C channel. Each plotted deletion effect is $s[V(\pi) - V(\pi_{\text{deleted}})]$ for a located rank-four subspace, whereas advantage is $s[V(\pi) - V(\pi_{\text{uniform}})]$. Here $V = \sum_i w_i r_i / \sum_i w_i$ and $w_i = \pi(a_i | o_i) / p_i$ use the archived propensity field, and s is the prescribed polarity. The trace-probe AUC uses both hidden features and observed-action indicators; its input contract does not isolate a history-specific hidden trace.

Repeated exposures from the same user are not treated as independent confirmatory replicates. The primary SE confirmation unit for the causal chains is the frozen model realization (seed). Held-out user exposures estimate the named consequence effects within each realization; protocol success is adjudicated at the realization level. The archive does not contain a complete platform propensity/filtering specification beyond the public random-replacement description, so claims are limited to the released random-policy subset and named consequence.

Formation, lifecycle and death audits

Formation required pre-absence followed by a selective post-consequence distinction and matched-control separation. Lifecycle gates tested persistence under continued support, reorganization or sign reversal under changed consequence and dissolution under release. In the phase-resolved 3B histories, persistence and reorganization were signed criteria on x , whereas readout and stack release each required $0.40 \leq \text{Acc}_\alpha \leq 0.60$ and $0.40 \leq \text{Acc}_\beta \leq 0.60$ on the same 94 held-out items; equal x can therefore receive different release verdicts. The same-checkpoint death experiment branched Null and continued-support futures from the same checkpoint and optimizer state and audited old-mapping consequence separately from residual carrier content. Figure 2b is the small-network mechanism archive, separate from the language-model phase series. Its sine/ReLU, quadratic/Tanh and interaction/residual-ReLU cases each retain one seed. All four continuations share each case's B endpoint and RMSprop state. Null removes the B-differential training target while retaining the shared base target; continued B retains both. Tail-only Null freezes early layers, whereas early-only Null freezes the tail. With h_d the condition-centered hidden response, V the B-frozen intermediate basis, y_d the B-target differential and δp_d the intact-minus-deleted output differential, carrier content is $\langle (h_d V)^2 \rangle / \langle y_d^2 \rangle$, role power is $\langle (\delta p_d)^2 \rangle / \langle y_d^2 \rangle$, and alignment is $\cos^2(\delta p_d, y_d)$. Means include all displayed latent coordinates where applicable; squared alignment is not a signed-polarity readout.

Erase, restore, transplant and signed swap

A rank-four hidden-to-action coupling was estimated on locator data. Erase removed that measured coupling. Independent restore used a disjoint locator fold. C transplantation inserted the active C realization into a released Null recipient, followed by recipient recovery. Signed exchange transplanted positive B into a negative-B host and vice versa. Equal-rank random subspaces, action-row permutations, wrong-organization modules and shuffled-consequence modules served as separation controls. These interventions identify a reproducible realizer at the stated resolution; they do not identify a unique or minimal physical carrier.

Cross-realization identity

Cross-seed transport has 20 directed pairs among five MLP realizations; cross-architecture transport has 30 directed pairs between MLP and gated-residual families. Both retain C, positive-B and negative-B readouts; frozen per-direction criteria included donor recovery ≥ 0.15 for C and ≥ 0.50 for signed B, plus aligned-effect and control-margin criteria. The 18-direction unpaired boundary jointly changes outcome semantics, user domain, action/context maps and architecture. All full-direction gates fail. Its frozen per-direction criteria were C recovery ≥ 0.10 , held-out profile cosine ≥ 0.80 , operator-signature cosine ≥ 0.95 and at least 0.02 above its permuted-locator control, together with control-margin and restore criteria. The profile and score-operator readouts test that joint transport contract; they do not isolate one causal factor.

Conditional lifecycle dynamics and admission boundary

Lifecycle gates were frozen independently of the dynamics. The decomposition compares paired pressure-present and pressure-removed evolutions and is algebraically exact bookkeeping. Predictive content requires independently specified recruitment, contraction and leakage conditions. Admission tests retain failed formation despite support, persistence under release, and the insufficiency of one-step contraction to certify a whole phase. Transparent adaptive systems support prospective regime prediction after realization-specific calibration. No realization-free coefficient map, parameter-free universal timescale or prospective deep-neural constitutive closure is claimed.

Controlled model experiments and unit definitions

The controlled formation archive includes five 0.5B and three 3B language-model histories; the lifecycle archive includes three phase-resolved 3B histories. The displayed small-network branch experiment contains three learner/world configurations, one seed per configuration, four cloned branches per case and five repeated steps per branch (60 branch-checkpoint records). The cases differ in task function and network settings; they are not three identical-configuration seed replicates. The manipulation archive contains six seed-level naturalistic contrasts and fifty controlled-world units; the latter are not pooled with the naturalistic rows. For the 50-world experiment, the functional signature S contains twelve readouts: $\log(1 + \|\nabla_{\theta} L_{Q_j}\|^2)$ and $\log(1 + \|\nabla_{\theta_h} L_{Q_j}\|^2)$ for each of six frozen future tasks Q_j . The distance ratios use the native $\|S_A - S_B\|$ as denominator: transplant and restore retain residual distance to S_A , swap to S_B , and erase the residual gap between the two erased states. Locator fractions instead measure displacement along a frozen development-time locator. The frozen pass intervals were (0.5, 1.5) for each locator fraction and < 0.2 for the erase ratio, alongside eleven further frozen gates; all 50 confirmation worlds passed all thirteen. A locator fraction near 1 is not equality of the complete signature. Cross-seed transport contains twenty directed contrasts among five MLP realizations; cross-architecture transport contains thirty directed rows; and the cross-consequence boundary contains eighteen directed rows. Direction and checkpoint observations remain nested within their realization, branch or transfer design rather than being counted as independent model histories.

Statistical units and reporting

Confirmatory success is adjudicated at the unit defined by each design: model history for formation, independent world for the controlled confirmation, and the frozen seed or branch for the corresponding manipulation and lifecycle experiments. Users, exposures, directions and checkpoints are nested measurements where the design specifies repeated observations; they are not substituted for independent histories. The twenty cross-seed directions arise from five realizations, the thirty architecture-transfer rows from directed cross-family pairs, and the eighteen cross-consequence rows from the frozen transfer family. Continuous readouts and gate outcomes are retained separately in the accompanying Source Data.

Ethics and governance status

This study performs secondary analysis of the publicly released KuaiRand dataset. The author had no interaction with the original users, performed no new participant recruitment, intervention or data collection, and analysed only the released research records, in which user and item identifiers were re-indexed by the dataset providers. Public documentation describes the platform-randomized exposure procedure used to generate the relevant interactions. We were unable to verify an authoritative public statement specifying the original collection's ethics-review, exemption or consent status, and attempts to obtain clarification from the dataset authors did not yield such a statement before submission. We therefore make no claim regarding an original approval, exemption or consent status that is not documented in the available public record.

Use of generative AI

Generative AI tools from OpenAI, including ChatGPT and Codex, were used during study planning and research discussions, to assist with code drafting and modification and the development and review of analysis workflows, and in preparing the manuscript. The author reviewed and verified AI-assisted code, analyses and text, made the scientific judgments, and accepts full responsibility for the final content.

Data availability

The accompanying Source Data file contains the ten derived tables underlying Figs. 1–4 and a separate archived naturalistic manipulation table with ten run-level records (five ordinary and five shuffled confirmation seeds), continuous readouts and twelve frozen gate outcomes per run. The original KuaiRand random-policy data are available from the providers (DOI: 10.5281/zenodo.10439422) and are not redistributed here. A separate archive of experimental consolidation and provenance records is held under restricted access and is not downloadable through this submission. Model checkpoints and large upstream input files are not included.

Code availability

Python scripts and dependencies to regenerate and validate the four figures from the accompanying derived source tables are supplied as Supplementary Software 1. This package reproduces the figures, not model training or the original experimental runs; it excludes model checkpoints and the original KuaiRand user-level records.

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