

# Participation Without Guarantee: A Field-Theoretic Critique of Upwork as a Probabilistic Marketplace

Flyxion

Independent Researcher

## Abstract

We model the freelance platform Upwork as a non-equilibrium system within the Relativistic Scalar–Vector Plenum (RSVP) framework and analyze its epistemic structure using TARTAN (Trajectory-Aware Recursive Tiling with Annotated Noise). We demonstrate that the platform monetizes participation independently of successful realization, inducing a persistent entropy floor in the matching process. Trust signals are shown to constitute a presheaf with non-trivial first cohomology, implying global inconsistency despite local coherence. We prove that under mild assumptions the system admits a stable regime in which expected freelancer value is negative while platform revenue remains strictly positive. A Labor Singularity Theorem establishes conditions under which the coupling between productive potential and realized compensation collapses to zero. The CLIO recursive correction operator is shown to fail to converge under entropy injection, preventing stable belief formation. A discrete lattice model captures these dynamics in a computable form. We further prove a Selection Theorem showing that extractive platforms are competitively favored over fair ones under weak market assumptions, and introduce a thermodynamic stability parameter classifying platform regimes by their ratio of returned to consumed user effort. Identity leasing is formalized as a control and observability problem in which risk is transferred to the visible account holder while productive control remains hidden. Scams and identity arbitrage arise as equilibrium strategies within this regime rather than as anomalies.

## 1. Introduction

Digital labor platforms present themselves as neutral intermediaries between clients and freelancers. In practice, they implement ranking, filtering, and monetization mechanisms that fundamentally reshape the structure of labor markets by commodifying not work but access to the possibility of work. Upwork, one of the largest such platforms, exemplifies this transformation. It offers escrow, ratings, and identity verification as proxies for trust while simultaneously charging users for visibility through “Connects,” a proprietary credit system required for proposal submission.

User narratives are often interpreted as isolated failures of individual vigilance. The journalist Justin Caffier’s account of losing \$475 to a fraudulent client—despite following standard procedural guidance—is a case in point. We treat such narratives as observational data revealing systemic properties: the loss is not anomalous but structurally expected under the dynamics formalized here.

A second and analytically distinct phenomenon is the mass solicitation of account-sharing arrangements. A solicitor contacts a developer via a public repository with a proposal to use their Upwork account via remote desktop software, bidding on jobs as that person, and remitting 10–20 percent of earnings in exchange. The rationale given is explicit: European or American account identities generate higher hourly rates and stronger proposal acceptance probabilities than Asian ones. Such solicitations arrive in volume and are structurally sophisticated. They reveal something precise about the platform’s field geometry: identity signals are monetizable because they generate differential flow. The solicitation is an informal market for visibility anisotropy, and our formal analysis explains why such a market arises necessarily.

The paper proceeds as follows. Section 2 establishes the RSVP field representation. Section 3 proves the entropy floor theorem. Section 4 analyzes the negative expected value regime. Section 5 develops the sheaf-theoretic failure of trust. Section 6 proves the Labor Singularity Theorem. Section 7 introduces the CLIO correction operator and proves its non-convergence. Section 8 presents the discrete lattice model. Section 9 characterizes exploitative attractors and identity anisotropy. Section 10 proves the Selection Theorem for extractive platforms. Section 11 introduces the thermodynamic stability parameter. Section 12 formalizes identity leasing as a control problem. Section 13 synthesizes the results, and Section 14 concludes.

## 2. Field Structure

Let the platform be a bounded domain  $\Omega$  with field:

$$X(x, t) = (\Phi(x, t), \mathbf{v}(x, t), S(x, t)).$$

Here  $\Phi : \Omega \times \mathbb{R} \rightarrow \mathbb{R}_{\geq 0}$  is a scalar field representing productive potential—skill, labor capacity, and domain expertise. The vector field  $\mathbf{v} : \Omega \times \mathbb{R} \rightarrow \mathbb{R}^n$  represents flow: visibility, matching probability, and opportunity access. The scalar  $S : \Omega \times \mathbb{R} \rightarrow \mathbb{R}_{\geq 0}$  is an entropy field representing systemic uncertainty, noise, and adversarial interference.

A classical labor market satisfies  $\mathbf{v} \approx \nabla\Phi$  and  $S \rightarrow 0$ : productive potential drives flow, and matching noise is negligible. Upwork introduces an imposed cost field  $C(x) > 0$  that modifies flow:

$$\mathbf{v}_{\text{eff}} = \mathbf{v} - \nabla C.$$

**Proposition 1.** *If  $C(x) > 0$  on a set of nonzero measure, then  $\mathbf{v}_{\text{eff}}$  is not solely determined by  $\nabla\Phi$ .*

*Proof.* Since  $\mathbf{v}_{\text{eff}} = \nabla\Phi - \nabla C$ , dependence on  $\nabla C$  persists unless  $C$  is identically constant. Alignment fails generically whenever  $C$  varies across the domain.  $\square$

The effective flow depends on a composite function of skill, capital, and platform-specific rank  $\rho$  (Job Success Score, account age, review count):  $\mathbf{v}_{\text{eff}} \propto f(\Phi, C, \rho)$ . Since  $\rho$  is a path-dependent accumulation and  $C$  is a recurring cost, both systematically advantage incumbents over newcomers regardless of productive capacity.

## 3. Entropy Floor

Entropy evolves as:

$$\frac{\partial S}{\partial t} = D \nabla^2 S + \sigma - \gamma \mathcal{C}(X),$$

where  $D > 0$  is a diffusion constant,  $\sigma \geq 0$  represents the continuous influx of low-quality jobs, fraudulent postings, and adversarial strategies, and  $\mathcal{C}(X) = -\nabla \cdot \mathbf{v}$  is the constraint closure functional representing the entropy-suppressing effect of successful matches.

**Theorem 2 (Entropy Floor).** *If  $\sigma > 0$  and  $\gamma \mathcal{C}(X)$  is bounded above, then there exists  $S^* > 0$  such that  $S(x, t) \rightarrow S^*$  as  $t \rightarrow \infty$ .*

*Proof.* At steady state,  $\partial S / \partial t = 0$  gives  $D \nabla^2 S^* = \gamma \mathcal{C}(X^*) - \sigma$ . Since  $\gamma \mathcal{C}(X^*)$  is bounded above and  $\sigma > 0$ , the right-hand side cannot be made arbitrarily positive. A steady-state solution  $S^*$  exists by elliptic regularity on  $\Omega$ . If  $\gamma \mathcal{C}(X^*) < \sigma$  on any open set, then  $\nabla^2 S^* < 0$  there, and by the maximum principle  $S^*$  achieves a positive minimum. Thus  $S^* > 0$ .  $\square$

**Remark 3.** *The suppression term  $\gamma C(X)$  depends on the convergence of flow at successfully matched pairs. Because platform revenue is decoupled from match success—Connects revenue is proportional to proposals submitted, not contracts completed—there is no mechanism forcing  $\sigma \rightarrow 0$  as a condition of positive platform revenue. The entropy floor is a structural consequence of the revenue model, not a failure of enforcement.*

#### 4. Expected Value and Platform Revenue

Let freelancers pay cost  $k > 0$  per proposal and receive reward  $R$  with probability  $p \in (0, 1)$ . The expected value for a freelancer is  $\mathbb{E}[V] = pR - k$ . Platform revenue across  $N$  proposals is:

$$U = Nk + \tau pR,$$

where  $\tau$  is the platform take rate on completed contracts.

**Theorem 4** (Negative Expectation Regime). *There exist parameters  $(p, k, R, \tau, N)$  such that  $\mathbb{E}[V] < 0$  while  $U > 0$ .*

*Proof.* Choose  $k > pR$ , so that  $\mathbb{E}[V] = pR - k < 0$ . Then  $U = Nk + \tau pR > 0$  since  $Nk > 0$ . The conditions are simultaneously satisfiable for any  $N \geq 1, \tau \geq 0$ .  $\square$

**Corollary 5.** *Rational participation in expectation-negative conditions is sustained by variance: if the distribution of  $R$  has a positive tail of high-reward outcomes, agents whose individual expected value is negative may still participate under uncertainty. The platform sustains itself through continuous participation under this lottery structure.*

##### 4.1. Temporal Inconsistency and Illusory Liquidity

A structurally distinct exploit arises from mismatched state update times across systems. Let  $B(t)$  denote the true account balance and  $\tilde{B}(t)$  the available balance displayed by a financial interface. For a deposited check of amount  $C$  at time  $t_0$ , there exists a clearing delay  $\Delta > 0$  such that:

$$\tilde{B}(t_0 + \epsilon) = B(t_0) + C \quad \text{for } \epsilon > 0,$$

while for a fraudulent check:

$$B(t_0 + \Delta) = B(t_0).$$

**Lemma 6** (Illusory Liquidity Window). *There exists a nonempty interval  $(t_0, t_0 + \Delta)$  on which  $\tilde{B}(t) > B(t)$ .*

*Proof.* Immediate from delayed settlement: availability is granted before verification completes.  $\square$

Let transfers  $T_i$  occur within this interval. Then:

$$B(t_0 + \Delta^+) = B(t_0) - \sum_i T_i.$$

**Remark 7.** *The platform contributes the trust signal that enables acceptance of the check—high ratings, verified identity, established history—while the financial system contributes the temporal inconsistency. The scam is therefore a cross-system exploit of mismatched state update times rather than a simple deception. No individual system fails on its own terms; the vulnerability exists only in the gap between them. This gap is structurally maintained by the sheaf failure of Section 5: the trust signals local to the platform context do not extend to the financial context, and the financial system’s clearing delay is not visible to the platform’s trust field. The Illusory Liquidity Window is the temporal analogue of the trust obstruction class.*

## 5. Sheaf-Theoretic Failure of Trust

### 5.1. TARTAN and the Admissibility Manifold

Within the TARTAN framework, local plausibility and global realizability are distinct. A trajectory may be locally admissible at every patch while failing to admit a globally consistent reconstruction. The trust system on Upwork exhibits precisely this structure.

### 5.2. Trust as a Presheaf

Let  $\mathcal{U} = \{U_\alpha\}$  be an open cover of the interaction space  $\Omega$ , where each  $U_\alpha$  corresponds to a local context: a client profile, a job posting, or an active contract. The trust presheaf  $T$  assigns to each  $U_\alpha$  a section  $\tau_\alpha = (r_\alpha, v_\alpha, h_\alpha)$  encoding rating, verification status, and historical activity. Restriction maps  $\rho_{U_\alpha U_\beta} : T(U_\alpha) \rightarrow T(U_\beta)$  for  $U_\beta \subset U_\alpha$  project onto relevant subcomponents.

**Definition 8.** *A global trust section exists if local signals agree on all overlaps and glue to a unique element of  $T(\bigcup_\alpha U_\alpha)$ .*

**Theorem 9 (Trust Obstruction).** *If there exist interactions with consistent local signals but contradictory outcomes, then  $T$  has nontrivial first Čech cohomology:  $\check{H}^1(\mathcal{U}, T) \neq 0$ .*

*Proof.* Let  $U_1$  be the context of a client profile and  $U_2$  the context of an active contract with that client. Let  $\tau_1$  assign high rating, positive verification, and substantial history. These signals are locally coherent within  $U_1$ , and their restriction to  $U_1 \cap U_2$  (contract initiation) is consistent with  $\tau_1$ . The section  $\tau_2$  within  $U_2$  records the post-delivery state: payment withheld, scope dispute, or escrow manipulation. Since post-delivery behavior is not de-

terminated by pre-contract signals, no global section exists restricting to both  $\tau_1$  and  $\tau_2$ . The disagreement on the overlap defines a nontrivial Čech 1-cocycle.  $\square$

**Remark 10.** *This corresponds to TARTAN’s admissibility manifold: local plausibility at every observed patch coexists with global unrealizability. The trust system is not merely imperfect but structurally incapable of providing global guarantees.*

## 6. Labor Singularity

Let  $\kappa(x, t) \in [0, 1]$  denote the coupling between productive potential and realized reward, so that  $R = \kappa \Phi$ .

**Definition 11.** *A labor singularity occurs when  $\kappa(x, t) \rightarrow 0$  while  $\Phi(x, t) \neq 0$ : productive potential exists but cannot be reliably converted into income.*

**Theorem 12** (Labor Singularity Theorem). *If participation cost  $C > 0$ , matching probability  $p < 1$ , and escrow coverage is incomplete, then there exists a regime in which  $\lim_{t \rightarrow \infty} \kappa(x, t) = 0$  for a set of participants of nonzero measure.*

*Proof.* Expected net return per round is  $R_{\text{net}} = p\Phi - C$ . If  $C > p\Phi$ , accumulated losses over  $T$  rounds drive realized reward to  $-\infty$  relative to effort. The ratio  $\kappa = R/\Phi$  therefore converges to zero for any participant for whom  $C > p\Phi$  holds persistently. Since  $C$  is a fixed platform cost and  $p$  is bounded away from 1 by competition, this condition holds on a set of nonzero measure in  $(C, p, \Phi)$ .  $\square$

**Remark 13.** *Labor singularity is not a failure of labor but a failure of realization. The productive capacity  $\Phi$  is present and exercised; it simply does not return to its bearer. This is distinct from classical unemployment: the worker works but the return collapses.*

## 7. CLIO Failure Operator

The CLIO (Constraint-Leveraged Inference Operator) maps a current state estimate to a corrected estimate by incorporating new evidence:  $X_{n+1} = \mathcal{L}(X_n)$ . In well-behaved inference settings, repeated application converges to a fixed point  $X^*$  representing an accurate model of the environment.

**Definition 14.** *CLIO failure occurs when  $X_n \not\rightarrow X^*$  as  $n \rightarrow \infty$ , oscillating or drifting rather than converging.*

**Theorem 15** (Non-Convergence Under Entropy Injection). *If entropy injection  $\sigma > 0$  and trust signals are sheaf-inconsistent (Theorem 9), then  $\mathcal{L}$  fails to converge.*

*Proof.* Let  $\epsilon_n = \|X_n - X^*\|$ . Under entropy injection, new noise is introduced at rate  $\sigma$  per round, while correction reduces error by at most  $\delta$ :

$$\epsilon_{n+1} \leq \epsilon_n + \sigma - \delta.$$

If  $\sigma \geq \delta$ , then  $\epsilon_{n+1} \geq \epsilon_n$ , so the sequence cannot converge to zero. Since  $\sigma > 0$  is structurally maintained by the entropy floor theorem and  $\delta$  is bounded by the information available per round, non-convergence holds generically.  $\square$

**Remark 16.** *In platform terms, freelancers repeatedly update beliefs about clients and job quality but cannot reach a stable model. Each engagement introduces noise that outpaces individual inference. The system is not merely uncertain but actively inference-resistant.*

## 8. Discrete Lattice Model

We discretize  $\Omega$  into a grid  $\{(i, j)\}$  with the following update rules. The scalar field evolves as:

$$\Phi_{ij}^{t+1} = \Phi_{ij}^t + \alpha f(S_{ij}^t) - \beta \Phi_{ij}^t,$$

where  $\alpha > 0$  governs skill accumulation,  $f(S)$  is a decreasing function of entropy, and  $\beta > 0$  governs depreciation. The vector field evolves as:

$$\mathbf{v}_{ij}^{t+1} = \mathbf{v}_{ij}^t - \nabla C_{ij} + \eta \nabla I_{ij},$$

where  $\nabla C_{ij}$  is the local cost gradient and  $\eta \nabla I_{ij}$  is the identity gradient encoding geographic and reputational anisotropy. The entropy field evolves as:

$$S_{ij}^{t+1} = S_{ij}^t + D(\nabla^2 S)_{ij} + \sigma - \gamma \Phi_{ij}^t,$$

where  $(\nabla^2 S)_{ij}$  is the discrete Laplacian and  $\sigma > 0$  is the uniform noise injection rate.

**Proposition 17.** *The lattice system admits steady states with  $S_{ij} > 0$ ,  $\Phi_{ij} > 0$ , and  $R_{ij} \approx 0$ .*

*Proof.* If  $\sigma \approx \gamma \Phi_{ij}^*$  at the fixed point, then  $S_{ij}^* > 0$  by the discrete entropy floor argument. The fixed-point equation  $\alpha f(S^*) = \beta \Phi^*$  has a positive solution for any  $\alpha, \beta > 0$  and  $f > 0$ . With  $\kappa \rightarrow 0$  in the labor singularity regime,  $R_{ij} = \kappa \Phi_{ij}^* \approx 0$  despite  $\Phi_{ij}^* > 0$ .  $\square$

## 9. Exploitative Attractors and Identity Anisotropy

### 9.1. Attractor Formation

Define an attractor region  $\mathcal{A} \subset \Omega$  where  $\nabla \cdot \mathbf{v} < 0$ , indicating inward flow of freelancers.

**Theorem 18.** *If enforcement is delayed and entropy is high, exploitative attractors are dynamically stable.*

*Proof.* Let  $\sigma^*$  be a client strategy minimizing escrow while maximizing extracted labor. This strategy generates positive expected gain whenever enforcement penalty is low. Because dispute resolution is procedurally burdensome and ratings update only after completed interactions, the expected penalty remains low, making  $\sigma^*$  a stable best response. Inflow to  $\mathcal{A}$  continues as long as negative outcomes are not immediately propagated to the trust field—which the sheaf failure of Section 5 prevents.  $\square$

## 9.2. Identity Anisotropy

Let  $I(x)$  represent the identity field encoding geographic location, account age, and accumulated reputation, with  $\mathbf{v} \propto \nabla I$ .

**Proposition 19.** *If  $I$  varies across regions, flow is anisotropic and identity arbitrage is individually rational.*

*Proof.* Non-uniform  $\nabla I$  produces directional bias in  $\mathbf{v}$ . A participant with identity  $I_{\text{native}}$  can increase effective flow by borrowing an identity with  $I_{\text{borrowed}} > I_{\text{native}}$ , gaining  $\Delta \mathbf{v} = \mathbf{v}(I_{\text{borrowed}}) - \mathbf{v}(I_{\text{native}}) > 0$ . This is individually rational whenever  $\Delta \mathbf{v}$  yields increased expected reward exceeding the cost of the arrangement. The account-sharing solicitations described in Section 1 are precisely this adaptation organized as an informal market.  $\square$

## 10. Selection Theorem for Extractive Platform Regimes

We now ask a stronger question. Given multiple competing labor platforms, some deriving revenue primarily from successful matches and others from monetized participation under uncertainty, does the latter possess a structural competitive advantage? We show that under weak assumptions, extractive platforms can dominate fairer ones in growth and persistence.

For each platform  $P$ , define gross revenue over  $[0, T]$  by  $U_P(T) = U_P^{\text{match}}(T) + U_P^{\text{access}}(T)$ , where  $U_P^{\text{match}}$  is revenue from completed contracts and  $U_P^{\text{access}}$  is revenue from participation itself (proposal fees, bidding tokens, paid visibility, subscriptions). Define the *extractive ratio*:

$$\rho_P = \frac{U_P^{\text{access}}}{U_P^{\text{match}} + U_P^{\text{access}}}.$$

A platform with  $\rho_P \approx 0$  profits primarily when users succeed; a platform with  $\rho_P > 0$  monetizes access independently of realization.

Let  $M_P(t)$  denote active market interactions at time  $t$  and  $q_P(t)$  the realized match quality. A stylized revenue law is:

$$U_P(T) = \int_0^T (\alpha_P M_P(t) q_P(t) + \beta_P M_P(t)) dt,$$

where  $\alpha_P \geq 0$  is the revenue coefficient on successful realization and  $\beta_P \geq 0$  is the revenue coefficient on participation alone. A platform with  $\beta_P > 0$  can increase revenue by increasing traffic even when match quality stagnates or declines. Platform growth obeys:

$$\frac{dG_P}{dt} = \mu U_P - \nu C_P,$$

where  $G_P$  is a generalized growth functional,  $\mu > 0$  converts revenue into growth capacity, and  $C_P$  is the cost of operation and trust maintenance.

**Theorem 20** (Selection Theorem for Extractive Platforms). *Assume two platforms F and E with comparable operating costs and access to a labor pool, such that  $\beta_F \approx 0$  and  $\beta_E > 0$ . Suppose users continue to participate whenever perceived opportunity remains positive. Then there exists a nonempty parameter regime in which  $dG_E/dt > dG_F/dt$  even when  $q_E < q_F$ : extractive platforms can outcompete fairer ones despite producing worse average outcomes for workers.*

*Proof.* We have  $U_F = \int_0^T \alpha_F M_F q_F dt$  and  $U_E = \int_0^T (\alpha_E M_E q_E + \beta_E M_E) dt$ . For any fixed  $q_E < q_F$ , choose parameters such that  $\beta_E M_E > \alpha_F M_F q_F - \alpha_E M_E q_E$ , which gives  $U_E > U_F$ . If  $C_E - C_F$  is bounded above by  $\mu(U_E - U_F)/\nu$ , then  $\mu U_E - \nu C_E > \mu U_F - \nu C_F$ , hence  $dG_E/dt > dG_F/dt$ .  $\square$

**Corollary 21** (Incentive Misalignment). *If  $\beta_P > 0$ , then there exists a regime in which decreasing match quality  $q_P$  does not decrease revenue and may increase it.*

*Proof.* From  $U_P = \int_0^T (\alpha_P M_P q_P + \beta_P M_P) dt$ , if  $\partial M_P / \partial q_P < 0$  (lower quality induces more repeated applications), then the  $\beta_P M_P$  term can dominate the loss in  $\alpha_P M_P q_P$ , yielding  $\partial U_P / \partial q_P \leq 0$ .  $\square$

The theorem requires only that access be monetized, that participation persist under uncertainty, and that revenue be partially decoupled from match success. This result may be sharpened by noting that if  $\partial M_P / \partial S_P > 0$  (uncertainty drives more repeated applications) and  $\partial q_P / \partial S_P < 0$  (uncertainty lowers match quality), then:

$$\frac{\partial U_E}{\partial S_E} = \alpha_E \left( q_E \frac{\partial M_E}{\partial S_E} + M_E \frac{\partial q_E}{\partial S_E} \right) + \beta_E \frac{\partial M_E}{\partial S_E}.$$

If the final term dominates, then  $\partial U_E / \partial S_E > 0$ : there exists an entropy interval on which worsening the informational environment is revenue-positive. Introducing the selection functional  $\Sigma(P) = \mu U_P - \nu C_P$ , we obtain  $\Sigma(E) > \Sigma(F)$  despite  $q_E < q_F$ . The more extractive

system is preferentially selected by market competition, and faces no internal pressure to eliminate the uncertainty from which it profits.

## 11. Thermodynamic Stability Parameter

We introduce a stability parameter that classifies platforms according to whether they are contractive, critical, or expansive with respect to user effort and resource flow. Let  $E_{\text{in}}(t)$  denote the total effort, time, and monetary expenditure contributed by users, and  $E_{\text{out}}(t)$  the realized value returned in the form of completed contracts and durable opportunities. Define the *realization ratio*:

$$\lambda_P = \frac{E_{\text{out}}}{E_{\text{in}}}.$$

A platform with  $\lambda_P < 1$  is in a *contractive regime*: user participation yields less value than it consumes. At  $\lambda_P = 1$  the system is *critical*, approximately balanced. At  $\lambda_P > 1$  the system is *expansive*: participation amplifies value, generating surplus that sustains long-term engagement.

### 11.1. Decomposition of Effort

We decompose:

$$E_{\text{in}} = E_{\text{labor}} + E_{\text{search}} + E_{\text{access}},$$

where  $E_{\text{labor}}$  is productive work under contract,  $E_{\text{search}}$  is effort spent discovering opportunities, and  $E_{\text{access}}$  is direct monetary expenditure required to participate. Similarly,  $E_{\text{out}} = E_{\text{paid}} + E_{\text{option}}$ , where  $E_{\text{paid}}$  is realized compensation and  $E_{\text{option}}$  represents future opportunity value. A fair platform minimizes  $E_{\text{search}} + E_{\text{access}}$  relative to  $E_{\text{paid}}$ ; an extractive platform permits growth in  $E_{\text{search}}$  and  $E_{\text{access}}$  without proportional increases in  $E_{\text{paid}}$ .

### 11.2. Persistence of the Contractive Regime

Suppose  $\lambda_P < 1$ . Then for a typical user,  $\frac{d}{dt}E_{\text{user}} = -(1 - \lambda_P)E_{\text{in}} < 0$ : continued participation leads to net depletion of resources. Participation persists if users operate under uncertainty or delayed feedback. Let  $\hat{\lambda}_P$  denote the perceived ratio. Persistence occurs when  $\hat{\lambda}_P \geq 1$  while  $\lambda_P < 1$ , a mismatch sustained by incomplete information, selective success visibility, and stochastic reward realization. Even if  $\lambda_P^{\text{user}} < 1$ , a platform may satisfy  $dR_P/dt > 0$  provided  $E_{\text{access}}$  grows sufficiently, yielding the hallmark asymmetry:

$$\lambda_P^{\text{user}} < 1 \quad \text{while} \quad \lambda_P^{\text{platform}} > 1.$$

In RSVP terms, this corresponds to a field configuration in which the scalar potential  $\Phi$  remains high while the vector field  $\mathbf{v}$  is dissipative, redirecting flows into platform-mediated sinks. The entropy field  $S$  remains elevated, sustaining uncertainty and repeated engagement. The platform maintains controlled disequilibrium rather than converging toward equilibrium.

## 12. Identity Leasing as a Control and Observability Problem

In a standard model, there is a one-to-one mapping between account and operator:  $A \leftrightarrow U$ . Identity leasing replaces this with a hidden structure in which the visible identity  $U_1$  is distinct from the actual operator  $U_2$ . The platform observes only  $A$  and partial signals from  $U_2$  such as typing patterns, deliverables, and communication behavior.

### 12.1. State and Observation Model

Let  $x(t)$  denote the internal state including skill level, location, and behavioral traits of the true operator  $U_2$ , and  $y(t)$  the observable outputs (messages, completed work, response times):

$$y(t) = h(x(t), A) + \epsilon(t).$$

The platform attempts to infer  $x(t)$  from  $y(t)$  under the assumption that  $A$  uniquely determines  $x(t)$ . Identity leasing violates this assumption.

### 12.2. Hybrid Control Structure

Let  $u(t)$  denote the control input applied by  $U_2$ , with the system evolving as  $dx/dt = f(x, u)$ . The account holder  $U_1$  provides boundary conditions during verification events, yielding a hybrid system:

$$u(t) = \begin{cases} u_2(t) & \text{most of the time,} \\ u_1(t) & \text{during verification events.} \end{cases}$$

### 12.3. Observability Failure and Risk Transfer

The system is observable if  $x(t)$  can be reconstructed from  $y(t)$ . Identity leasing induces observability failure because multiple internal states produce indistinguishable outputs:

$$\exists x_1 \neq x_2 \text{ such that } h(x_1, A) \approx h(x_2, A).$$

This creates an equivalence class of operators behind a single account. Under identity leasing, the risk functional satisfies  $\mathcal{R}(U_2) \rightarrow \mathcal{R}(A)$ : risk transfers from the hidden operator to the visible account holder, who bears full downside while the operator captures a fraction  $(1 - \alpha)$  of earnings.

Identity leasing emerges when three conditions hold simultaneously: accounts carry location- or reputation-dependent value, entry barriers prevent certain users from accessing high-value accounts, and monitoring is incomplete. Under these conditions, accounts become leasable capital assets and labor becomes a hidden variable. In RSVP terms, the account  $A$  functions as a boundary condition on the field while the true operator  $U_2$  modifies the internal dynamics. The platform enforces constraints only on the boundary, not the interior, creating a mismatch between representation and realization. Identity leasing is therefore not an aberration but a natural consequence of a system in which visibility, rather than capability, is the primary scarce resource.

### 13. Synthesis

The results of the preceding sections are mutually reinforcing and converge on a single structural picture. The Connects cost field severs the alignment between productive potential and opportunity flow, establishing conditions for labor singularity. Once the coupling  $\kappa$  collapses, participants operate in an expectation-negative regime while platform revenue remains positive, sustaining the lottery structure that motivates continued participation. The entropy floor theorem establishes that this noise is a permanent feature of any individually rational equilibrium. The sheaf-theoretic analysis shows that the trust system cannot detect or correct the resulting adversarial behavior, since its local signals are consistent while its global sections fail to exist. The CLIO operator, confronted with this inference-resistant environment, cannot converge to a stable belief, leaving participants perpetually vulnerable to the attractor dynamics that exploitative strategies generate.

The selection theorem extends the critique from description to evolutionary claim: platforms structured this way are not merely exploitative but competitively favored. A fair platform that profits only from successful matches faces a structural growth disadvantage against one that additionally harvests the costs of failed attempts. When entropy enters the revenue function with a positive sign—because uncertainty drives repeat participation—the system faces no internal pressure to improve match quality. The thermodynamic stability analysis quantifies this asymmetry through the realization ratio  $\lambda_p$ : platforms operating in the regime  $\lambda_p^{\text{user}} < 1$ ,  $\lambda_p^{\text{platform}} > 1$  can grow indefinitely by consuming user effort faster than they return value, provided the perception gap  $\hat{\lambda}_p - \lambda_p > 0$  is maintained. Identity leasing completes the picture as a secondary extraction layer: once visibility gradients exist and observability is limited, accounts themselves become leasable capital and labor becomes a

hidden variable.

The system therefore admits a stable regime characterized by:

$$\Phi > 0, \quad R \approx 0, \quad S > S^* > 0, \quad U > 0, \quad \Sigma(E) > \Sigma(F).$$

Productive potential is abundant; realized reward is near zero for a significant fraction of participants; entropy is persistent; platform revenue is positive; and the more extractive configuration is competitively preferred. This is not a market failure departing from an equilibrium that would otherwise obtain. It is the equilibrium.

## 14. Conclusion

Platform labor markets such as Upwork are best understood not as inefficient markets but as non-equilibrium systems in which participation is monetized independently of realization. The collapse of coupling between skill and compensation, the failure of recursive trust inference, the persistence of an entropy floor, the dynamical stability of exploitative attractors, and the competitive selection of extractive revenue models produce a coherent structural description in which exploitation is not an anomaly but a phase of the system.

Individual-level advice—perform more due diligence, choose clients more carefully, avoid suspicious postings—operates at the wrong level of description. Such advice reduces local exposure while leaving  $S^*$ , the trust obstruction class  $\check{H}^1(\mathcal{U}, T)$ , and the selection differential  $\Sigma(E) - \Sigma(F)$  unchanged. These are properties of the system, not of any participant's behavior. Structural intervention requires altering the decoupling between revenue and outcome: specifically, making the platform's revenue contingent on successful realization rather than on participation alone.

## References

- [1] Rolf Landauer. Irreversibility and heat generation in the computing process. *IBM Journal of Research and Development*, 5(3):183–191, 1961.
- [2] Charles H. Bennett. The thermodynamics of computation—a review. *International Journal of Theoretical Physics*, 21(12):905–940, 1982.
- [3] Seth Lloyd. Ultimate physical limits to computation. *Nature*, 406:1047–1054, 2000.
- [4] Erik Verlinde. On the origin of gravity and the laws of Newton. *Journal of High Energy Physics*, 2011(4):29, 2011.

- [5] Ted Jacobson. Thermodynamics of spacetime: The Einstein equation of state. *Physical Review Letters*, 75(7):1260–1263, 1995.
- [6] Saunders Mac Lane. *Categories for the Working Mathematician*. Springer, New York, second edition, 1998.
- [7] Jacob Lurie. *Higher Topos Theory*. Princeton University Press, Princeton, 2009.