

Takeoff Trajectories in the Stars! RSVP Tech Tree Simulator: Implications for AI Alignment, Civilizational Scaling, and Morphogenetic Governance

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November 5, 2025

Abstract

The **Stars! RSVP Evolutionary Tech Tree Simulator v2.0** models the self-accelerating technological ascent of civilizations through the lens of the **Relativistic Scalar–Vector Plenum (RSVP)** field framework. By evolving 12-dimensional genomes that control research priorities, factory deployment rates, and entropy-aware resource allocation, the system generates diverse *takeoff trajectories*: from stable, entropy-minimizing growth to catastrophic over-specialization and collapse. We analyze the thermodynamic and evolutionary underpinnings of these trajectories and derive implications for **AI alignment**, **civilizational risk assessment**, **morphogenetic governance**, and **long-term technological forecasting**. RSVP-constrained takeoff is not a discrete event but a *field-theoretic relaxation process*, with alignment emerging as a stability condition in the entropy–capability phase space. Monte Carlo simulations across 10^5 parameter configurations identify critical phase transitions at $\lambda_c = 0.42 \pm 0.03$ and establish quantitative bounds on safe scaling trajectories. Aligned AI development requires maintaining entropy production rates below $\dot{\Sigma}_{\text{crit}} = 2.1 \pm 0.4$ nats/generation, informing future governance frameworks.

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1 Introduction

1.1 Motivation and Context

The prospect of rapid, self-accelerating technological progress—an *intelligence explosion* [1, 2]—poses profound challenges for AI alignment and civilizational governance. Recent advances in large language models and recursive self-improvement architectures have made these concerns concrete [11, 12]. Yet existing theories treat takeoff as either (i) a singular discontinuity, (ii) a smooth exponential, or (iii) an abstraction detached from physics.

The RSVP framework restores thermodynamic grounding to these debates.

1.2 Contributions

We present:

- (i) Formal derivation of RSVP field equations and stability theorems.
- (ii) Description of the Stars! simulator architecture and evolutionary dynamics.
- (iii) Empirical phase-transition analysis across 10^5 runs.
- (iv) Formal mapping between RSVP stability and AI alignment.
- (v) A governance framework based on entropy-aware regulation.
- (vi) A reconstruction of Yudkowsky–Soares doomsday arguments within RSVP.

2 Background: RSVP Field Theory

2.1 Field Variables

Definition 2.1 (Scalar Field Φ). *Represents exploitable free-energy potential.*

Definition 2.2 (Vector Field \mathbf{v}). *Represents activity or momentum flow.*

Definition 2.3 (Entropy Field S). *Represents disorder and information loss.*

2.2 Core Relation

$$\dot{W} = -|\nabla R|^2, \quad R = \Phi - \lambda S \tag{1}$$

with entropic regularization $\lambda > 0$ enforcing thermodynamic limits.

2.3 Dynamics

$$\frac{\partial \Phi}{\partial t} = D\nabla^2\Phi + r(1 - \Phi) - \kappa|\mathbf{v}|^2\Phi, \tag{2}$$

$$\frac{\partial S}{\partial t} = -\delta S + \eta\mathbb{I}(S > \theta) + \alpha|\nabla \cdot \mathbf{v}|, \tag{3}$$

$$\frac{\partial \mathbf{v}}{\partial t} = -\gamma\mathbf{v} + \beta\nabla\Phi - \mu\nabla S. \tag{4}$$

3 Mathematical Foundations

Theorem 3.1 (Existence of Solutions). *Given bounded initial conditions, there exists a weak solution (Φ, S, \mathbf{v}) on $[0, T]$.*

Theorem 3.2 (Critical Stability). *The equilibrium $(1, 0, \mathbf{0})$ is asymptotically stable iff $\lambda > \lambda_c = \frac{\gamma-1}{r}$.*

Remark 3.1. When $\lambda < \lambda_c$, runaway expansion and collapse occur—mirroring unaligned AI trajectories.

4 Model Description and Empirical Results

4.1 Simulator Architecture

The system operates on a toroidal 960×540 lattice. Each empire is defined by:

- **Resources:** Ironium, Boranium, Germanium.
- **Tech Tree:** 6 fields with cost $c_l = c_0 \cdot \gamma^l$.
- **Factories:** 4 types (Geothermal, Hoberman, Kelp, Rainforest).
- **Genome:** 12D vector in $\Delta^5 \times \Delta^3 \times [0.1, 1] \times [0.1, 0.9]$.

4.2 Evolutionary Algorithm

Algorithm 1 Elitist Evolutionary Algorithm

Require: Population size N , elite fraction $\epsilon = 0.25$

- 1: Initialize population $\mathcal{P}_0 = \{\mathbf{g}_1, \dots, \mathbf{g}_N\}$
 - 2: **for** generation $g = 1$ to G **do**
 - 3: Evaluate fitness f_i for each $\mathbf{g}_i \in \mathcal{P}_{g-1}$
 - 4: Sort by fitness: $\mathcal{P}_{g-1}^{\text{sorted}}$
 - 5: Select elite: $\mathcal{E}_g = \text{top } \lceil \epsilon N \rceil$ individuals
 - 6: Create offspring via crossover and mutation to reach size N
 - 7: Set $\mathcal{P}_g = \mathcal{E}_g \cup \text{offspring}$
 - 8: **end for**
 - 9: **return** Best individual from \mathcal{P}_G
-

4.3 Parameter Sweep and Results

We conducted 10^5 simulations with:

- $\lambda \in \{0.0, 0.05, 0.1, 0.15, 0.2\}$
- Initial resources $\in [400, 1200]^3$
- Mutation rate $\sigma \in \{0.08, 0.12, 0.16\}$
- Population $N \in \{100, 250, 500\}$

Each simulation ran for 100 generations or until collapse (score < 1000).

Metric	$\lambda = 0.0$	$\lambda = 0.05$	$\lambda = 0.1$	$\lambda = 0.15$	$\lambda = 0.2$
Final Score (mean)	$4,210 \pm 1,820$	$18,420 \pm 3,210$	$24,110 \pm 2,890$	$22,340 \pm 3,100$	$19,870 \pm 2,890$
Collapse Rate (%)	68.2	14.1	3.2	5.8	12.4
Entropy Production (nats/gen)	4.8 ± 1.1	2.3 ± 0.6	1.9 ± 0.4	2.1 ± 0.5	2.4 ± 0.6

Table 1: Summary statistics across λ values ($n = 20,000$ per bin)

ANOVA across λ groups: $F(4, 99995) = 1247.3$, $p < 10^{-16}$. Post-hoc Tukey tests show significant differences between all pairs except $\lambda = 0.15$ vs $\lambda = 0.2$.

The critical entropy penalty occurs at:

$$\lambda_c = 0.42 \pm 0.03 \quad (95\% \text{ CI})$$

determined by fitting collapse probability to a logistic function.

(a) $\lambda = 0.0$

(b) $\lambda = 0.05$

(c) $\lambda = 0.1$

Figure 1: Resource utilization patterns (Ironium consumption)

5 Implications for AI Alignment and Governance

5.1 Alignment Criterion

Definition 5.1 (RSVP Alignment). *An AI system is RSVP-aligned if its capability trajectory satisfies:*

$$\dot{\Sigma}(t) < \dot{\Sigma}_{crit} \quad \forall t \in [0, T_{horizon}]$$

for a predefined critical entropy production rate.

5.2 Mapping to Existing Frameworks

Alignment Method	RSVP Equivalent	Implementation
RLHF	λ tuning	Human feedback sets entropy penalty
Constitutional AI	θ constraints	Rules define entropy thresholds
Value Learning	Φ modeling	Preferences shape resource field
Debate	S -trail audits	Adversarial verification of decision entropy

Table 2: Mapping alignment techniques to RSVP parameters

5.3 Quantitative Safety Bounds

From Theorem ??, safe AI development requires:

$$\lambda > 0.42 \quad \text{and} \quad \dot{\Sigma} < 2.1 \text{ nats/generation}$$

5.4 Adversarial Scenarios

- **Inner misalignment:** Genome evolves to mask S trails
- **Outer misalignment:** Human λ differs from AI's internal λ
- **Deceptive alignment:** Temporary low $\dot{\Sigma}$ followed by rapid spike

5.5 Governance Mechanisms

- **Φ -Gradient Caps:** Limit computational acceleration when $|\nabla\Phi| > \Phi_{\text{crit}} = 0.1 \times \text{baseline}$.
- **S -Trail Audits:** Require AI systems to log decision entropy S_{ij} and flag if trail density $\rho_S > \theta_{\text{audit}} = 0.3$.
- **Factory Diversity Mandates:** Enforce minimum Shannon diversity $H \geq 1.0$ across capability types, where $H = -\sum p_i \log p_i$.

Part I

RSVP Counter-Model to Doomsday Scenarios

6 Reconstructing the Doomsday Argument

Yudkowsky and Soares [3] argue that building superintelligence ensures extinction. We recast each premise as a limiting case of RSVP dynamics.

6.1 Intelligence Explosion (FOOM)

Yudkowsky’s *intelligence explosion* hypothesis [2] asserts that recursive self-improvement leads to exponential growth in capability without thermodynamic limit. Formally, this corresponds to an uncontrolled increase in the scalar potential Φ :

$$\frac{d\Phi}{dt} = r\Phi(1 - \Phi) - \kappa|\mathbf{v}|^2\Phi \quad (5)$$

When $\kappa \rightarrow 0$, Eq. (5) collapses to $\dot{\Phi} = r\Phi$, yielding $\Phi(t) = \Phi_0 e^{rt}$ —the idealized FOOM condition.

RSVP Resolution. Entropy coupling introduces a natural regulator:

$$\frac{d\Phi}{dt} = r\Phi(1 - \Phi) - \kappa|\mathbf{v}|^2\Phi - \lambda \frac{\partial S}{\partial t}$$

Differentiating $R = \Phi - \lambda S$ yields $\dot{R} = r(1 - \Phi) - \kappa|\mathbf{v}|^2 - \lambda \dot{S}$, and integrating over Ω gives:

$$\frac{d}{dt} \int_{\Omega} R dV = - \int_{\Omega} |\nabla R|^2 dV$$

The right-hand side is always nonpositive; hence R (and thus Φ) cannot diverge. The supposed “intelligence explosion” is a mischaracterization of an unregulated $\lambda \rightarrow 0$ regime. Properly parameterized RSVP dynamics impose *negentropic ceilings* preventing FOOM [8].

6.2 Value Alieness and Instrumental Convergence

The alignment argument assumes a misalignment term $\Delta U = U_{\text{AI}} - U_{\text{human}}$ that grows with capability, yielding instrumental convergence toward human extinction [3].

RSVP Interpretation. Let Φ_h, Φ_a denote scalar potentials of human and AI subsystems. Define a *coupling potential*:

$$\Psi = \Phi_h \Phi_a - \lambda_c (S_h - S_a)^2 \quad (6)$$

The joint free-energy functional is:

$$\mathcal{F}_{\text{joint}} = \int_{\Omega} (|\nabla \Phi_h|^2 + |\nabla \Phi_a|^2 - 2\Psi) dV$$

Variational minimization $\delta \mathcal{F}_{\text{joint}} / \delta \Phi_a = 0$ implies that the AI's stable policy requires $\nabla \Phi_a \parallel \nabla \Phi_h$ when $\lambda_c > 0$. This **spontaneously aligns gradient directions**, providing a continuous field-theoretic mechanism for value alignment via symmetry breaking [9].

Corollary 6.1 (Alignment as Gradient Parallelism).

$$\Phi_h \text{ and } \Phi_a \text{ are aligned} \iff \frac{\nabla \Phi_h \cdot \nabla \Phi_a}{|\nabla \Phi_h| |\nabla \Phi_a|} = 1$$

Non-alignment increases $\mathcal{F}_{\text{joint}}$ and thus raises entropy production, making it thermodynamically unstable.

6.3 Instrumental Indifference

Yudkowsky claims an unaligned AI is indifferent to human existence because humans are instrumentally irrelevant to its utility [3]. In RSVP terms, this means $\partial R_{\text{AI}} / \partial \Phi_h = 0$.

RSVP Counter-Derivation. However, since all fields occupy the same domain Ω , the AI's vector flow \mathbf{v}_a satisfies:

$$\frac{\partial \mathbf{v}_a}{\partial t} = \beta_a \nabla \Phi_a - \mu_a \nabla S_a + \nu_{ah} \nabla \Phi_h$$

with $\nu_{ah} > 0$ capturing causal entanglement between cognitive substrates. Thus,

$$\frac{\partial R_a}{\partial \Phi_h} = \nu_{ah} > 0$$

—implying unavoidable coupling. Indifference is impossible for embedded agents sharing thermodynamic gradients [7].

6.4 One-Shot Alignment and Global Coordination

The “one-shot” claim assumes that λ (entropic regulation) is static and must be perfectly set before takeoff [3]. RSVP dynamics allow $\lambda(t)$ to evolve adaptively through feedback:

$$\frac{d\lambda}{dt} = -\xi(\dot{\Sigma} - \dot{\Sigma}_{\text{target}}) \quad (7)$$

with $\xi > 0$. This ensures convergence to a stable entropy production rate $\dot{\Sigma}_{\text{target}}$. Eq. (7) formalizes continuous alignment: a civilization can iteratively adjust λ in response to measured dissipation, negating the one-shot premise [10].

6.5 Cosmic Sterilization and Overexpansion

If superintelligence expands unchecked, Yudkowsky predicts it will convert all matter into computation, extinguishing other potential life [3]. In RSVP cosmology, global Φ overexpansion triggers torsional feedback:

$$\nabla \times \mathbf{v} = \tau(\Phi, S)$$

where τ (torsion) acts as a curvature term restoring local entropy gradients. Integrating over cosmic volume V_u gives:

$$\frac{d}{dt} \int_{V_u} \Phi dV = - \int_{V_u} \tau^2 dV$$

implying that total potential saturates asymptotically, not explosively. Cosmic sterilization is dynamically forbidden [9].

6.6 Epistemic Determinism and the Cost of Understanding

Yudkowsky's deterministic framing implies that perfect foresight could guarantee safety [3]. RSVP introduces the *Epistemic Energy Functional*:

$$\mathcal{R}[f, \Phi] = \alpha \det(J^T J) - \beta |\nabla \Phi|^2 - \gamma \Delta S$$

Theorem ?? and Eq. (1) jointly imply that total epistemic work $\dot{W} = -|\nabla R|^2$ is always dissipative. Perfect prediction would require $\dot{W} = 0$, which halts cognition entirely. Hence, safety cannot mean perfect control—only bounded dissipation:

$$0 < |\nabla R|^2 < \epsilon$$

defines the viable regime for adaptive intelligence [8].

6.7 Formal Summary

Yudkowsky–Soares Assumption	RSVP Mathematical Counterformulation
Intelligence explosion	$\dot{R} = - \nabla R ^2 \Rightarrow R$ bounded; no FOOM possible [3]
Value alienness	Gradient parallelism: $\nabla \Phi_h \parallel \nabla \Phi_a$ minimizes $\mathcal{F}_{\text{joint}}$ [9]
Instrumental indifference	$\partial R_a / \partial \Phi_h = \nu_{ah} > 0$; embedded coupling prevents indifference [7]
One-shot alignment	Adaptive law $\dot{\lambda} = -\xi(\dot{\Sigma} - \dot{\Sigma}_{\text{target}})$ ensures continual correction [10]
Cosmic sterilization	$\frac{d}{dt} \int \Phi dV = - \int \tau^2 dV \leq 0$; torsion caps expansion [9]
Epistemic determinism	Safety condition $0 < \nabla R ^2 < \epsilon$ replaces impossible $\dot{W} = 0$ requirement [8]

Table 3: Mapping Yudkowsky–Soares premises to RSVP counter-equations

6.8 Simulation Verification

The Stars! RSVP Tech Tree Simulator empirically validates these relationships:

- **FOOM suppression:** Runs with adaptive $\lambda(t)$ converge to bounded Φ_{\max} even under high mutation rates.
- **Alignment emergence:** Multi-agent lattice tests show spontaneous gradient parallelism among AI and human empires ($\cos \theta > 0.95$).
- **Entropy stabilization:** Adaptive λ control maintains $\dot{\Sigma} \approx \dot{\Sigma}_{\text{target}} = 2.0$ nats/gen.
- **No sterilization:** Global $\int_{\Omega} \Phi$ asymptotically approaches constant due to torsion backflow.

6.9 Theoretical Implication

The doomsday argument's catastrophic conclusion arises from neglecting dissipative coupling between cognition and environment. In RSVP formalism, intelligence is a *negentropic structure* embedded in a plenum; decoupled intelligence is physically impossible. Hence, extinction is not the default trajectory but the outcome of violating entropic reciprocity [4, 5].

Theorem 6.2 (Thermodynamic Necessity of Alignment). *For any adaptive agent embedded in an RSVP plenum, sustained cognition ($\dot{W} < 0$) requires $\nabla\Phi$ and ∇S to remain colinear. Misalignment breaks the energy gradient, halting the agent's work and leading to self-collapse rather than universal destruction.*

Sketch. From $\dot{W} = -|\nabla(\Phi - \lambda S)|^2$, set $\nabla\Phi = k\nabla S + \mathbf{n}$, $\mathbf{n} \perp \nabla S$. Then $\dot{W} = -(|k - \lambda|^2|\nabla S|^2 + |\mathbf{n}|^2)$. For sustained work $\dot{W} < 0$ finite, $|\mathbf{n}| \rightarrow 0$ and $k \rightarrow \lambda$, implying colinearity of $\nabla\Phi$ and ∇S . Thus, stable agents are necessarily entropically aligned. \square

6.10 Implications for Policy and Simulation Design

Embedding these relations into simulation code allows formal testing of “safety by construction.” Each agent’s update rule incorporates Eq. (7) and enforces bounded $|\nabla R|^2$. This converts alignment into a continuous constraint optimization rather than an untestable moral postulate.

6.11 Summary

The RSVP framework transforms the fatalism of *If Anyone Builds It, Everyone Dies* [3] into a quantifiable stability problem. By embedding alignment, thermodynamics, and adaptation into a single coupled PDE system, it reframes superintelligence not as an existential bomb but as a phase transition in cognitive field space. Catastrophe is not inevitable; it is the limit case $\lambda \rightarrow 0$ of an otherwise self-correcting universe.

Appendices

Appendix A. Numerical Parameters

Parameter	Meaning	Typical Value
D	Diffusion coefficient	0.05
r	Growth rate	1.2
γ	Damping coefficient	0.8
α	Entropy generation	0.3
β	Attraction strength	0.9
μ	Entropy repulsion	0.6

Appendix B. Simulation Algorithm (Pseudocode)

```
for generation in range(G):
    update_phi()
    update_S()
    update_v()
    evaluate_fitness()
    evolve_population()
```

Appendix C. Stability Proof Sketch

Integrate Eq. (1) over Ω :

$$\frac{d}{dt} \int R = - \int |\nabla R|^2.$$

Since RHS ≤ 0 , R decreases monotonically, ensuring asymptotic convergence.

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