

The Geometry of Adversarial Learning

Mirror Flows, Poincaré Recurrence, and No-Regret Learning in Continuous Games

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Motivation: Competition Beyond Min-Max?

Potential and **min-max** games anchor two ends of a behavioral spectrum:

- **Cooperation** among players and **convergent** learning dynamics;
- **Competition** among players and **cycling** learning dynamics.

We argue: *min-max is not the right model for adversarial learning/competition.*

- Min-max \cap Potential $\neq \emptyset$
- What about > 2 agents?

What is a structurally sound notion of multi-agent competition, and how should adaptive agents learn in such environments?

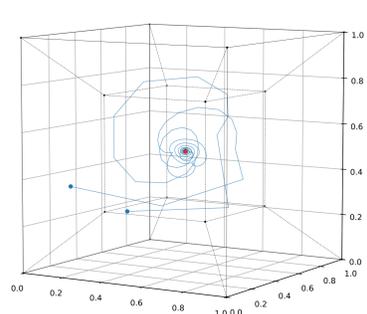
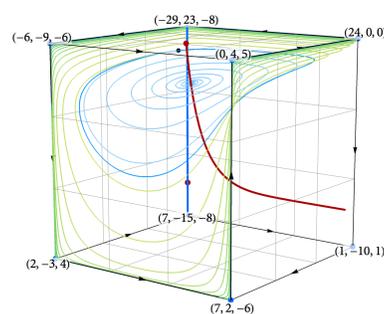
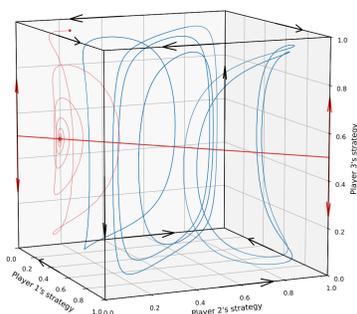
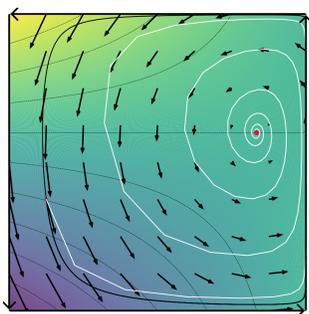


Figure: Long-run behavior of entropic mirror flow (**recurrent**) and FTRL+ dynamics (**convergent**) in 2-, 3-, and 4-player **incompressible** games. **Red**: Nash equilibria.

Hessian Flows and Poincaré Recurrence

« In a **Poincaré-recurrent** dynamical system trajectories return **arbitrarily close to their starting points** infinitely often. »

- **Poincaré's + Liouville's theorems:** $\begin{cases} v^\sharp \text{ is divergence-free} \\ \text{ri } \mathcal{X} \text{ has finite volume} \end{cases} \implies \dot{x} = v^\sharp(x) \text{ is recurrent}$

FTRL+: Last-Iterate Convergence and Optimal Regret

$$\begin{aligned} y_{i,n+1/2} &= y_{i,n} + \hat{v}_{i,n} & y_{i,n+1} &= y_{i,n} + v_i(x_{n+1/2}) \\ x_{i,n+1/2} &= Q_i(\eta_{i,n+1/2} y_{i,n+1/2}) & x_{i,n+1} &= Q_i(\eta_{i,n+1} y_{i,n+1}) \end{aligned}$$

- **Extrapolation step:** Encompasses several known algorithms (DEX, Optimistic DA, ...)
- **Adaptive learn-rate:** Environment-agnostic

Continuous Games

- **Players:** $i = 1, \dots, N$
- **Strategies:** compact convex $\mathcal{X}_i \subset \mathbb{R}^{n_i}$
- **Payoffs:** $u_i: \mathcal{X} = \prod_{i \in \mathcal{N}} \mathcal{X}_i \rightarrow \mathbb{R}$
- **Gradient field:**
 $v_i(x) = \nabla_i u_i(x) \in \mathbb{R}^{n_i}$

Take-Away: Characterize and Stabilize Multi-Agent Cycles

We propose **incompressible games** to model continuous **multi-agent competition**.

- 1 Continuous games admit a **unique decomposition** into potential and incompressible components \implies Incompressible \cap Potential $= \emptyset$.
- 2 In incompressible games, mirror flows are **Poincaré recurrent**.
- 3 Optimistic dynamics recover **last-iterate convergence** and $\mathcal{O}(1)$ **regret**.

Hodge Decomposition

- On a **compact** manifold, $v = d\phi + v_{\text{inc}}$ with $\delta v_{\text{inc}} = 0$
 - **Obstacle:** $\text{ri } \mathcal{X}$ is **not compact!**
- We propose:**
Novel compactification procedure via **isometric smoothing map**.

The Mirror Descent Family

« Replace Euclidean gradient steps with **mirror / Bregman ones** »

- K -strongly convex $h: \mathcal{X} \rightarrow \mathbb{R}$
- Prox-mapping
 $P_x(v) = \arg \min_{q \in \mathcal{X}} \{ \langle v, x - q \rangle + D(q, x) \}$
- **Mirror Descent**
 $x_{i,n+1} = P_{x_{i,n}}(\gamma_{i,n} v_{i,n})$

- Choice map (soft-max)
 $Q(y) = \arg \max_{x \in \mathcal{X}} \{ \langle y, x \rangle - h(x) \}$

- **Follow-The-Regularized-Leader**
 $y_{i,n+1} = y_{i,n} + \gamma_{i,n} v_i(x_n)$
 $x_{i,n+1} = Q_i(y_{i,n+1})$

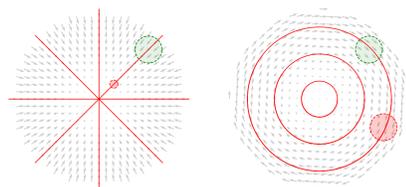
Mirror Flows

- Mirror flow
 $\dot{y}_i(t) = v_i(x(t))$
 $\dot{x}_i(t) = Q_i(y_i(t))$
- Induced **Hessian flow**
 $\dot{x}_i = v_i^\sharp(x) := \Pi_{x_i} \text{Hess } h_i^{-1} v_i(x)$
- Riemannian metric $g_i := \text{Hess } h_i$

Incompressible Games

« **Hessian-Riemannian divergence** = 0 »

$$0 \equiv \text{div } v^\sharp(x) = \delta v(x) = \sum_{i=1}^N \frac{1}{\sqrt{\det g_i(x_i)}} \sum_{\alpha_i=1}^{n_i} \frac{\partial}{\partial x_{i\alpha_i}} \left(\sqrt{\det g_i(x_i)} v_{i\alpha_i}^\sharp \right)$$



Competition Beyond Min-Max

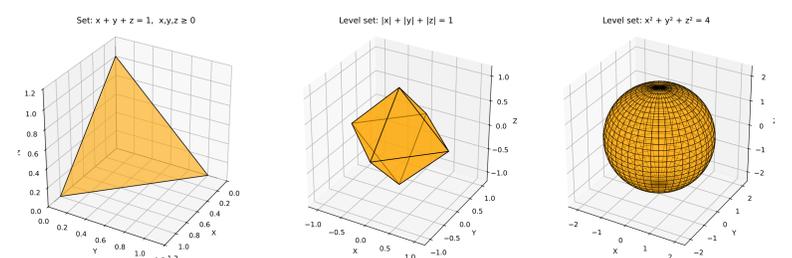
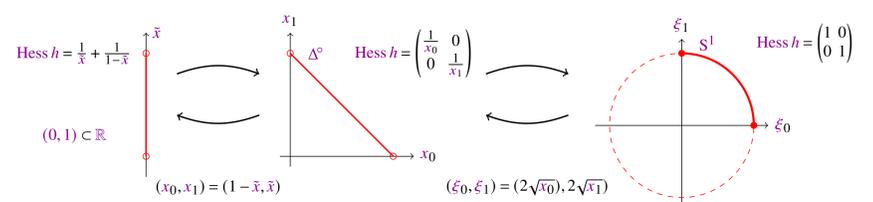
« **Local deviations by one player induce transverse responses by the others** »

$$\sum_{i \in \mathcal{N}} m_i \langle v_i(x), q_i - x_i \rangle = 0 \quad \# \forall x; m_i > 0; q \in \text{ri } \mathcal{X}$$

- Encompasses min-max games
- **3-player example**

« Two players **compete** for users engagement; a third player **moderates**. »

$$\begin{aligned} u_1(x) &= (x_3 - 1/2) x_1 \\ u_2(x) &= (1/2 - x_3) x_2 \\ u_3(x) &= (x_2 - x_1) x_3 \end{aligned}$$



Our Contributions

Thm. — Potential-Incompressible Hodge Decomposition

The gradient field v of any continuous game can be **uniquely decomposed** as $v = d\phi + v_{\text{inc}}$ where $\phi: \text{ri } \mathcal{X} \rightarrow \mathbb{R}$ is a smooth function and $\delta v_{\text{inc}} = 0$.

Thm. — Mirror Flows are Poincaré Recurrent

In incompressible games, if $\text{vol } \text{ri } \mathcal{X} < \infty$, then almost every **mirror flow** orbit $x(t)$ returns arbitrarily close to its starting point $x(0)$ infinitely many times.

Thm. — Last-Iterate Convergence and Optimal Regret

In finite incompressible games, **FTRL+ converges** to the set of Nash equilibria, guaranteeing **constant** $\text{Reg}_i(T) = \mathcal{O}((\max h_i - \min h_i) + 1/K_i)$.