

## Introduction

⇒ Min-max optimization is a problem of interest in several communities including Optimization, Game Theory and Machine Learning.

$$(\theta^*, \phi^*) = \arg \min_{\theta \in \mathbb{R}^n} \arg \max_{\phi \in \mathbb{R}^m} r(\theta, \phi)$$

⇒ Unfortunately for non-convex non-concave  $r$  little is known, not even when the problem has meaningful solutions.

⇒ Min-max optimization has attracted renewed interest thanks to the application of training GANs.

## Bilinear Games vs GANs

⇒ Bilinear games

$$\arg \min_{\theta \in \Delta_n} \arg \max_{\phi \in \Delta_m} \theta^T U \phi$$

→ Convex Concave

→ Players control probabilities directly

⇒ GANs

$$\arg \min_{\theta \in \mathbb{R}^n} \arg \max_{\phi \in \mathbb{R}^m} r(\theta, \phi)$$

→ Non-Convex Non-Concave

→ Players control weights of Generators and Discriminators that compete with each other.

*How can we bridge the gap in theoretical understanding between bilinear games and GANs?*

## Hidden Bilinear Games

$$r(\theta, \phi) = \mathbf{F}(\theta)^T \mathbf{U} \mathbf{G}(\phi)$$

⇒ Each player has a smooth function not necessarily concave or convex  $F$  and  $G$

⇒ They get rewarded based on a hidden bilinear game whose Nash equilibria map to equilibria of this game

**But:** Can players use gradient descent to find them?

## Gradient-Descent-Ascent

$$\underbrace{\begin{cases} \dot{\theta} = -\nabla_{\theta} r(\theta, \phi) \\ \dot{\phi} = \nabla_{\phi} r(\theta, \phi) \end{cases}}_{\text{Continuous time: CGDA}} \text{ VS } \underbrace{\begin{cases} \theta_{k+1} = \theta_k - \alpha \nabla_{\theta} r(\theta_k, \phi_k) \\ \phi_{k+1} = \phi_k + \alpha \nabla_{\phi} r(\theta_k, \phi_k) \end{cases}}_{\text{Discrete time: DGDA}}$$

## Hidden Bilinear Games with two strategies

⇒ Hidden Bilinear Game setup

$$\mathbf{F}(\theta) = \begin{pmatrix} f(\theta) \\ 1 - f(\theta) \end{pmatrix} \quad \mathbf{U} = \begin{pmatrix} u_{0,0} & u_{0,1} \\ u_{1,0} & u_{1,1} \end{pmatrix} \quad \mathbf{G}(\phi) = \begin{pmatrix} g(\phi) \\ 1 - g(\phi) \end{pmatrix}$$

⇒ Nash equilibrium is  $p, q$  and there is a scalar  $v$

$$\dot{\theta} = -v \nabla f(\theta)(g(\phi) - q) \quad \& \quad \dot{\phi} = v \nabla g(\phi)(f(\theta) - p)$$

## Safety of an initialization

⇒ Observe that  $\theta(t)$  cannot traverse stationary points of  $f$ .

⇒ This traps  $f(\theta(t))$  in an interval  $f_{\theta(0)}$ . Same thing applies to  $g$ .

⇒ We call  $(\theta(0), \phi(0))$  safe if  $\nabla f(\theta(0)) \neq \mathbf{0}$  and  $\nabla g(\phi(0)) \neq \mathbf{0}$  and  $p \in f_{\theta(0)}$  and  $q \in g_{\phi(0)}$

## Possible behaviours of $f(\theta(t))$ and $g(\phi(t))$

⇒ Periodic orbits around the  $(p, q)$  equilibrium

### Theorem (Periodical Orbits)

Let  $\theta(0)$  and  $\phi(0)$  be safe initial conditions. Then the orbit  $(\theta(t), \phi(t))$  is periodic. Additionally

$$\lim_{T \rightarrow \infty} \frac{\int_0^T f(\theta(t)) dt}{T} = p, \quad \lim_{T \rightarrow \infty} \frac{\int_0^T g(\phi(t)) dt}{T} = q$$

⇒ Or even convergence to stationary points of  $f$  and  $g$

### Theorem (Spurious equilibria)

One can construct functions  $f$  and  $g$  so that for a positive measure set of initial conditions the trajectories converge to fixed points that do not correspond to equilibria of the hidden game.

Full version at arxiv:TBA

## Main ideas behind the proof

⇒ Even if  $f, g$  are non-invertible, there exist smooth functions such that  $\theta(t) = X_{\theta(0)}(f(t))$ ,  $\phi(t) = X_{\phi(0)}(g(t))$ .

⇒ The hidden game of  $f, g$  mimics properties and behaviors of conservative systems. The following quantity is time invariant:

$$H(f, g) = \int_p^f \frac{z - p}{\|\nabla f(X_{\theta(0)}(z))\|^2} dz + \int_q^g \frac{z - q}{\|\nabla g(X_{\phi(0)}(z))\|^2} dz$$

⇒ Apply the Poincaré-Bendixon theorem on the planar dynamical system of  $(f, g)$  to get periodicity.

## Extensions

**More than two strategies:**

⇒ Let vectors  $F(\theta), G(\phi)$  have size  $(N, M)$  whose each coordinate is controlled by disjoint subsets of the vectors  $\theta, \phi$ , i.e.

$$\mathbf{F}(\theta) = [f_1(\theta_1) \cdots f_N(\theta_N)] \quad \mathbf{G}(\phi) = [g_1(\phi_1) \cdots g_M(\phi_M)]$$

⇒ CGDA is bounded and diffeomorphic to a volume-preserving system.

⇒ **Poincaré recurrence** theorem states that volume-preserving bounded systems will, after a sufficiently long but finite time, recurturn to a state very close to their initial state.

**Question: Discretization is critical? Answer: No**

### Theorem (Spurious equilibria)

One can choose a learning rate  $\alpha$  and functions  $f, g$  (i.e. sigmoids) for DGDA so that its energy  $H$  is always non-decreasing and for a positive measure set of initial conditions the trajectories converge to fixed points that do not correspond to equilibria of the hidden game.

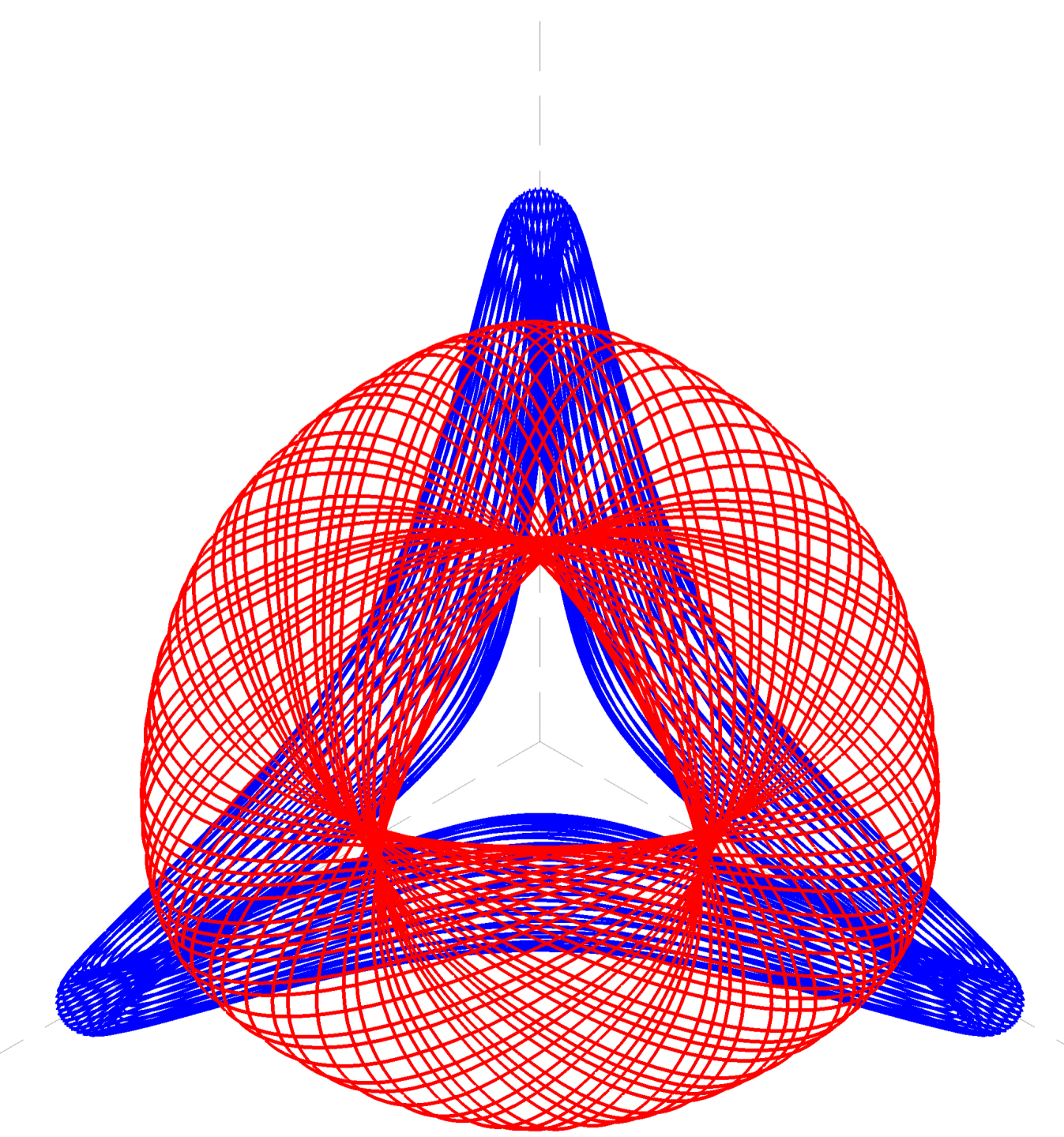


Figure: Rock-Paper-Scissors dynamics with sigmoid functions.