

Sensorimotor Games: Dynamical Models and Experiments

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Model: Learning is an optimization process

- ▶ Reward: $R : \Theta \rightarrow \mathbb{R}$
- ▶ Variable: $\theta \in \Theta$

Optimality: $R(\theta^*) \geq R(\theta), \theta \in \Theta_{x^*}$ (locally)

$$\max_{\theta \in \Theta} R(\theta)$$

Single objective

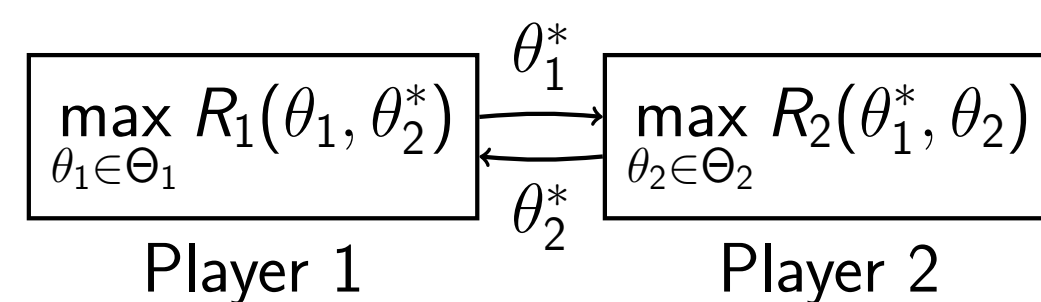
Game (two players):

- ▶ Game: $\mathcal{G} = (R_1, R_2)$.
- ▶ Reward: $R_i : \Theta_1 \times \Theta_2 \rightarrow \mathbb{R}, i \in [1, 2]$
- ▶ Variables: $(\theta_1, \theta_2) \in \Theta_1 \times \Theta_2$.

Optimality:

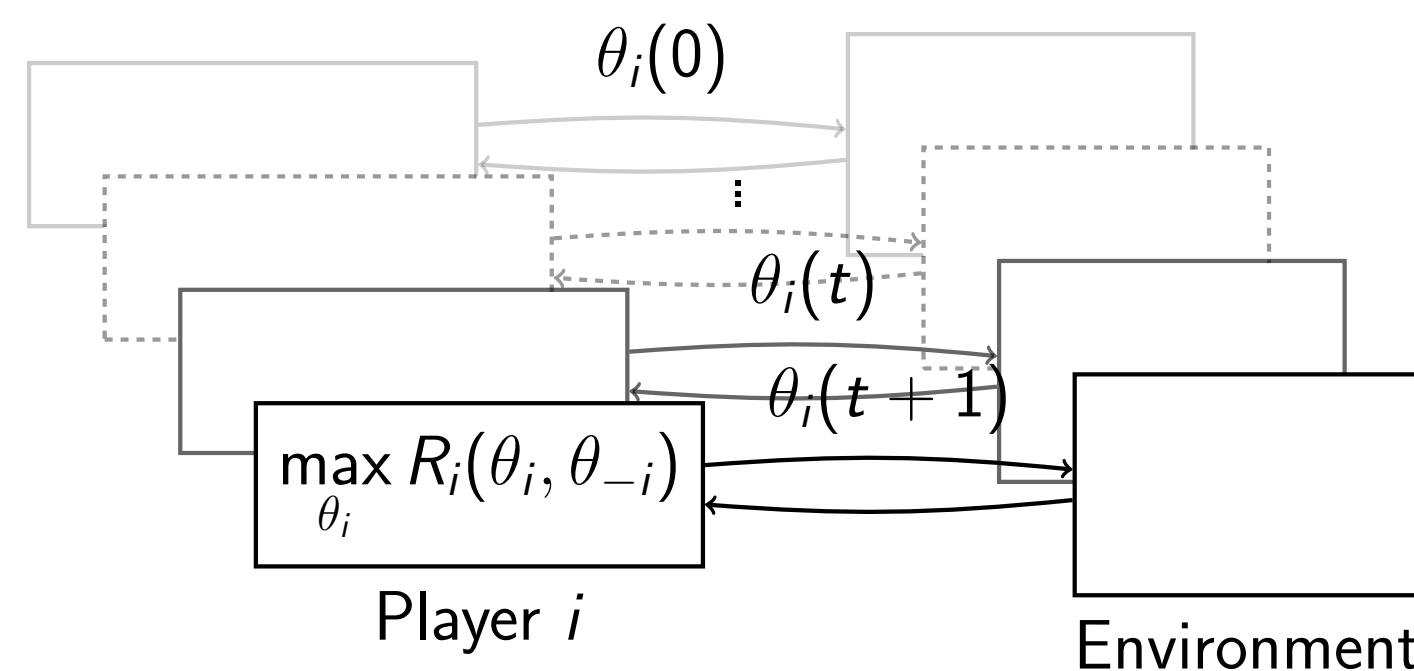
$$R_1(\theta_1^*, \theta_2^*) \geq R_1(\theta_1, \theta_2^*), \theta_1 \in \Theta_1,$$

$$R_2(\theta_1^*, \theta_2^*) \geq R_2(\theta_1^*, \theta_2), \theta_2 \in \Theta_2 \text{ (locally).}$$



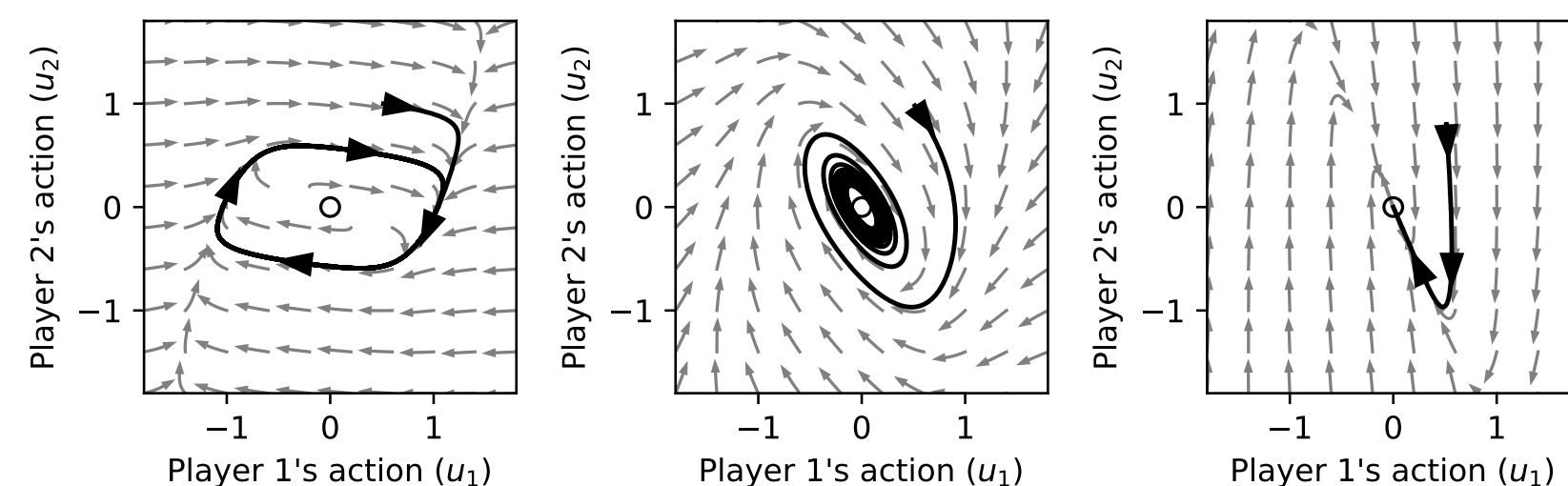
Dynamics of learning process

Observe learning $\theta_i(0), \dots, \theta_i(t)$ over time...



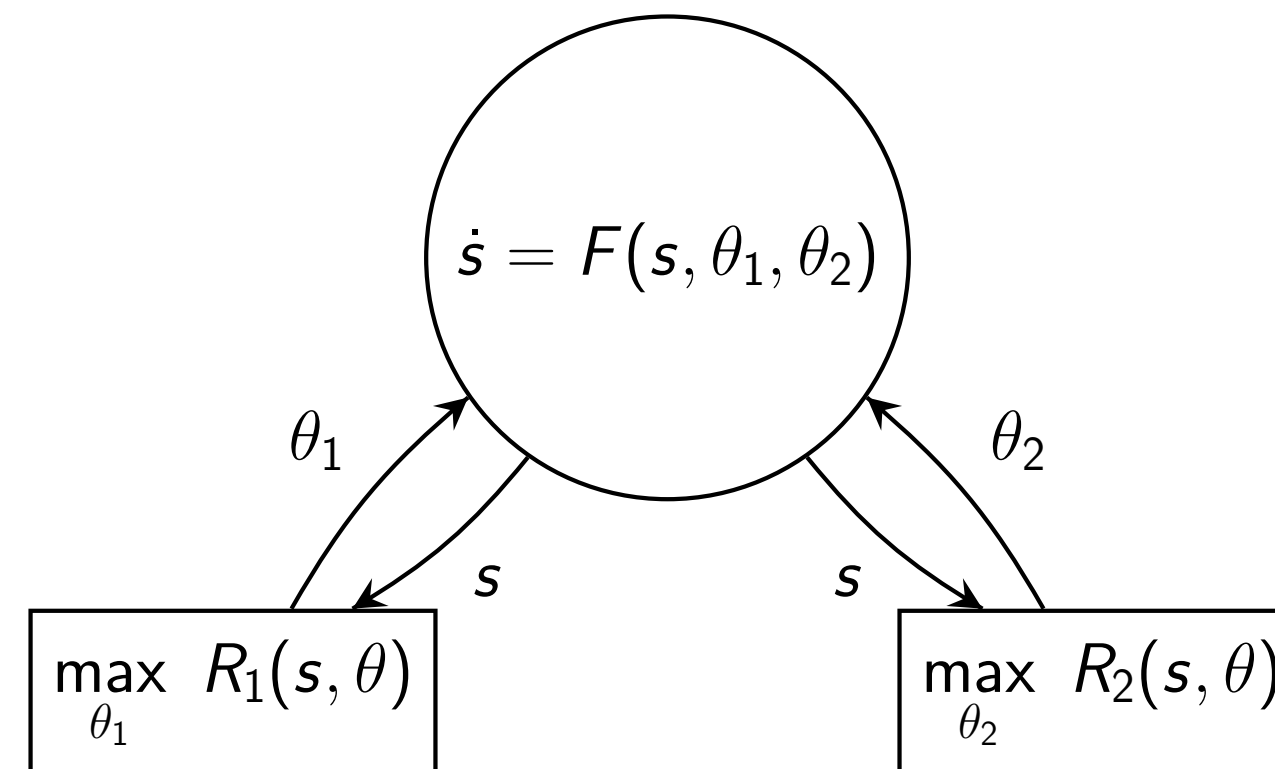
Prediction: periodic orbits and spurious attractors

- ▶ Vector field of $\omega \equiv (D_1 R_1, D_2 R_2)$ with rewards
- ▶ $R_1(\theta_1, \theta_2) = \frac{1}{4}\theta_1^4 - \frac{1}{2}\theta_1^2 - \theta_1\theta_2$
- ▶ $R_2(\theta_1, \theta_2) = \frac{1}{2}\theta_2^2 + 2\theta_1\theta_2$



A Dynamical Systems Perspective

A coupled set of dynamics describes (multi-agent) learning in environments.



- ▶ parameters $\theta = (\theta_1, \theta_2)$
- ▶ shared state s evolves via dynamics F .

Theory: Stability and Convergence Guarantees

- ▶ Asymptotic convergence: as $t \rightarrow \infty$, does $u(t) \rightarrow u^*$?
- ▶ Finite-time bounds: for $t \geq T$, what is $\|u(t) - u^*\|$?
- ▶ Regret/no-regret learning: "best action *in hindsight*."

Solution Concepts: Differential Notions

Twice-continuously differentiable objective $c \in \mathcal{C}^2$.

Local optimality of $u^* = (\theta_1^*, \theta_2^*)$ [1]:

First player's condition

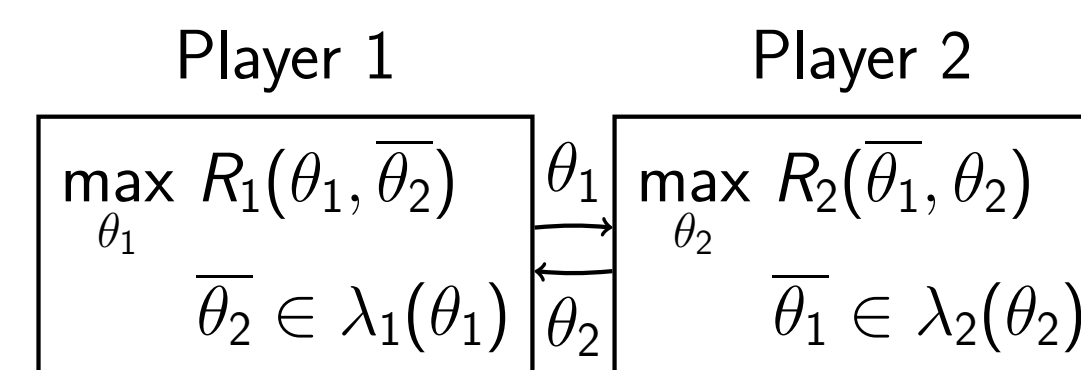
$$D_1 R_1(\theta_1^*, \theta_2^*) = 0, D_1^2 R_1(\theta_1^*, \theta_2^*) < 0$$

Second player's condition

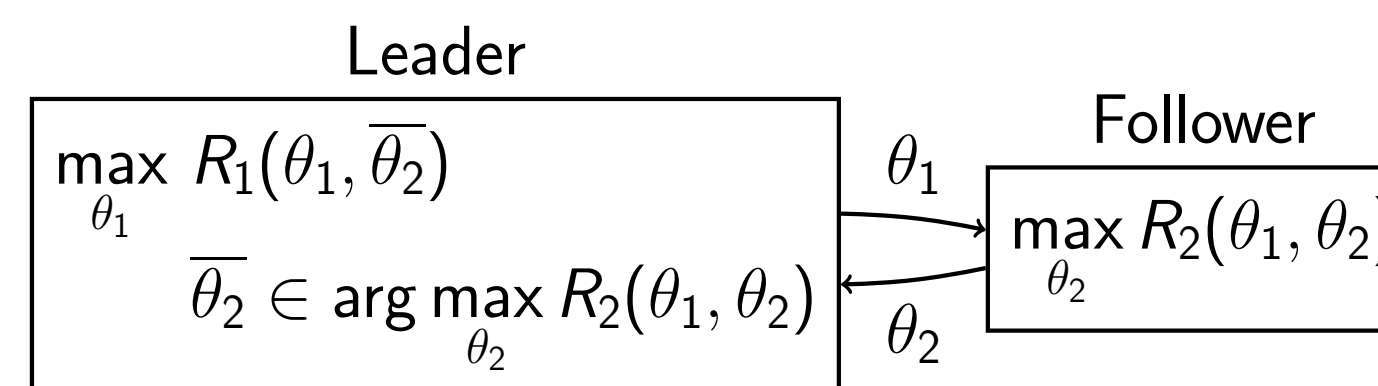
$$D_2 R_2(\theta_1^*, \theta_2^*) = 0, D_2^2 R_2(\theta_1^*, \theta_2^*) < 0$$

Novel Framework: Bounded Rationality

Players form *internal models* about others [2]:



For example, a Stackelberg game can be represented as

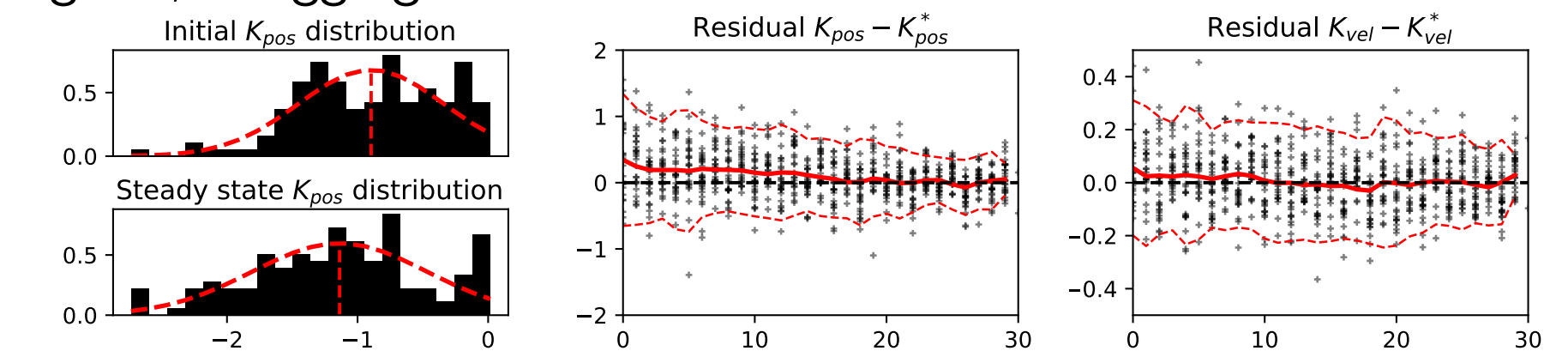


Experiments: Learning to Integrate

Operating a double integrator model $\dot{s} = (A + B\theta_t)s$. How do subjects learn θ_t over time? [2]

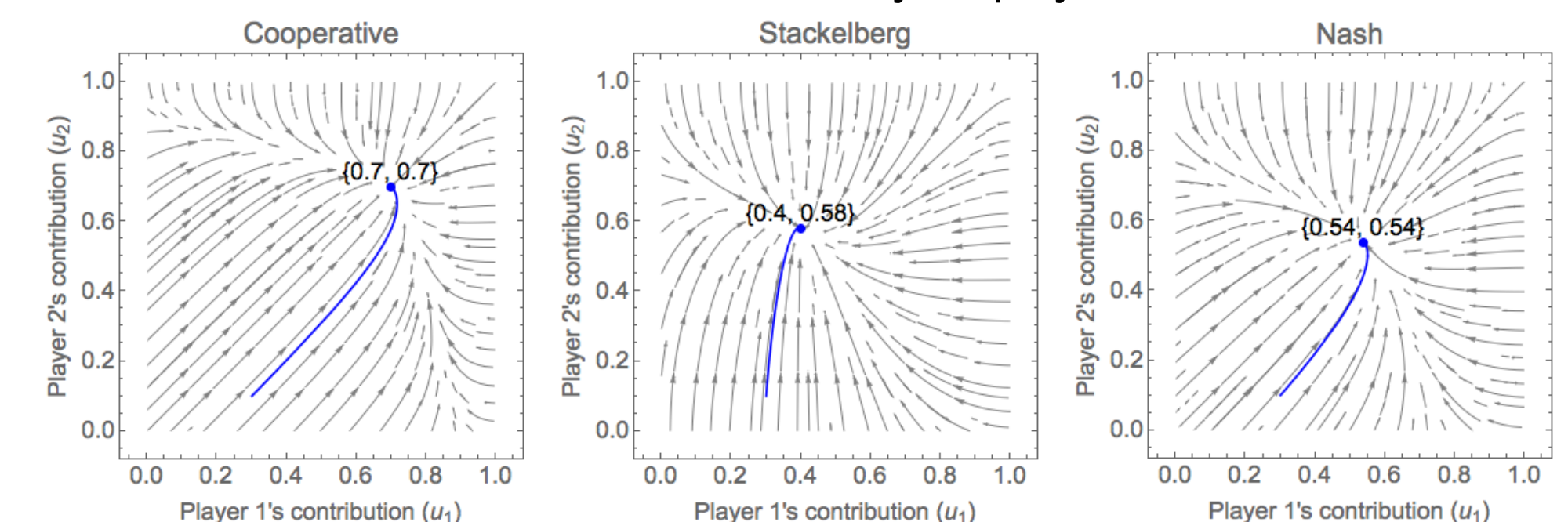


We observe subjects' learning curve for feedback and feedforward gains, in aggregate.



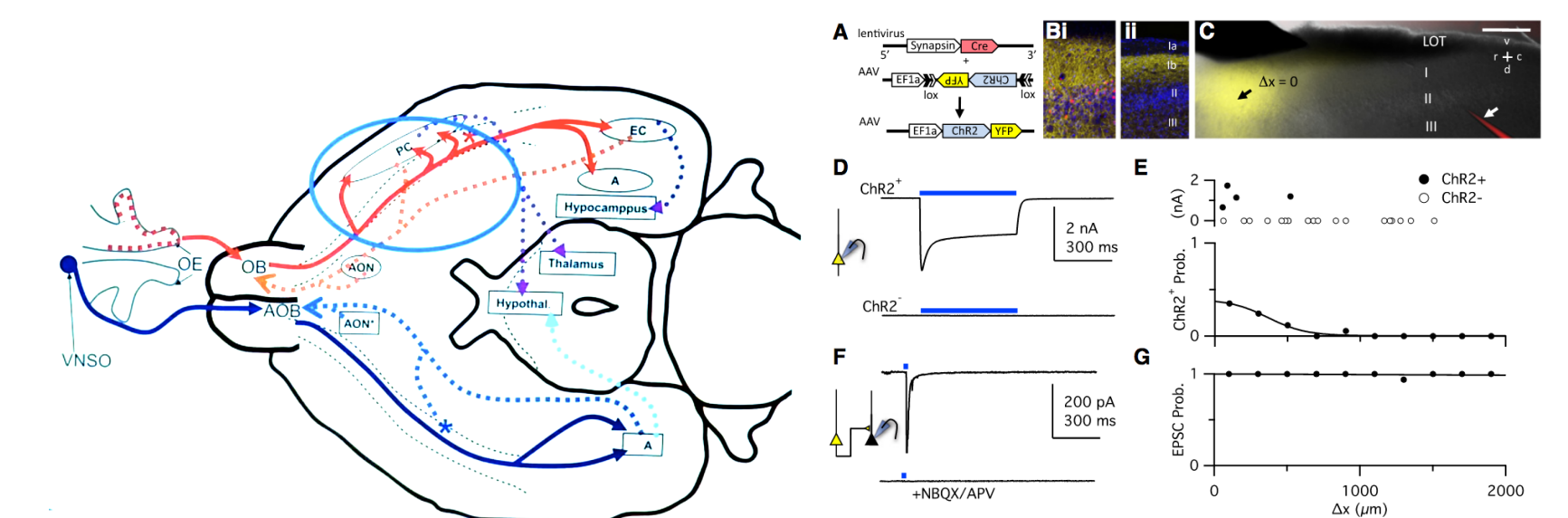
Experiments: Computational Predictions

- ▶ Individual contribution to public goods.
- ▶ Utilities: $-R_i(\theta_i, \theta_{-i}) = (I_i - \theta_i)^{\alpha_i}(\theta_i + \theta_{-i})^{1-\alpha_i}$
- ▶ Internal models dictate the hierarchy of play.



Experiments: Recurrent Circuitry in Piriform Cortex

Future work: modeling neurons in the piriform cortex "voting" for consensus on odors. [3]



References

- [1] Ratliff, Burden, Sastry. *Characterization and computation of local nash equilibria in continuous games*. 2013.
- [2] Chasnov, Yamagami, Parsa, Ratliff, Burden. *Experiments with sensorimotor games in dynamic human/machine interaction*. 2019.
- [3] Franks, Russo, Sosulski, Mulligan, Siegelbaum, Axel. *Recurrent Circuitry Dynamically Shapes the Activation of Piriform Cortex*. 2011.