Imitative Follower Deception in Stackelberg Games

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Background: Stackelberg Games & Learning

- A leader (*L*) vs. a follower (*F*)
- Stackelberg equilibrium $\langle x^*, y^* \rangle$ --- the optimal leader commitment:
- $\langle \mathbf{x}^*, \mathbf{y}^* \rangle = \operatorname{argmax}_{\mathbf{x}, \mathbf{y} \in \underline{BestResp}(\mathbf{x})} U_L(\mathbf{x}, \mathbf{y})$
- $BestResp(\mathbf{x}) \coloneqq \operatorname{argmax} U_F(\mathbf{x}, \mathbf{y})$
- Efficient computation of optimal leader commitment
- Applications: security, exam design, contract design, mechanism design

When Follower Type (Payoffs) is Uncertain...

— Learn the optimal commitment by observing follower best responses [Letchford et al., 2009; Blum et al., 2014; Haghtalab et al., 2016; Roth et al., 2016; Peng et al., 2019]



Imitative Follower Deception: an Example



A defender (the leader, row player) wants to defend two areas 1 and 2, which a poacher (the follower, column player) wants to attack. The poacher may be of Types A or B as his payoffs depend on animal prices on the black market, which fluctuate and are held private by the poacher.

Imitate When the follower is truthful Туре В Туре А Туре В $\left(\frac{1}{2}-\epsilon,\ \frac{1}{2}+\epsilon\right)$ $\left(\frac{3}{4}-\epsilon, \frac{1}{4}+\epsilon\right)$ Optimal commitment: Follower response: Leader utility: 1/20 Follower utility: 0 0



Our Model: Play Against Follower Deception

• A naïve playbook when deception is ignored



Leader Policy: a Better Playbook

- A policy-based framework
- Stage 1: Leader commits to a policy that specifies the strategy he will play for each reported (learned) follower type.



- But when the follower is **untruthful**...
- A *Type-A* follower has an incentive to imitate *Type B*, which makes the leader play $(1/2 - \epsilon, 1/2 + \epsilon)!$
- A Type-A follower gets ≈ 1 , but the leader only gets ≈ 0 N

Computing Optimal Policy: Algorithmic Results

• A complete view of the complexity: **OptPly** is hard to approximate, and hard still under **incentive compatibility (OptPly-IC)**

Theorem. For any $\epsilon > 0$, no poly-time $\frac{1}{(|\Theta|-1)^{1-\epsilon}}$ -approximation for **OptPly** unless P=NP, even when the number of follower actions is fixed to 3.

Theorem. For any $\epsilon > 0$, no poly-time $\frac{1}{|\Theta|^{1-\epsilon}}$ -approximation for **OptPly-IC** unless P=NP, even when the number of follower actions is fixed to 3.

Theorem. There exists a poly-time $\frac{1}{|\Theta|}$ -approximation algorithm for both w/o IC.

Theorem. Both **OptPly** and **OptPly-IC** are tractable for a fixed $|\Theta|$.

Generalization to Mixed Policies

• A higher level of randomization, able to improve leader utility further



Feel free to choose how to behave.

• **Stage 2**: **Follower** optimally reports (imitates) a type *T*, so that the strategy the leader will play according to her policy maximizes the follower's utility in Stage 3.

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• **Stage 3**: Leader plays a strategy **x** as prescribed by her policy and **Follower** best responds to **x** as if he is of type **T**.

A *Type-A* follower now has **no** incentive to misreport *Type B* !!

Theorem. Mixed policies with support size msuffice for achieving the optimality.

Theorem. With mixed policy, OptPoly remains hard to approximate, but **OptPoly-IC** becomes tractable.

Experiments

• Comparison of leader utility obtained with different approaches

