

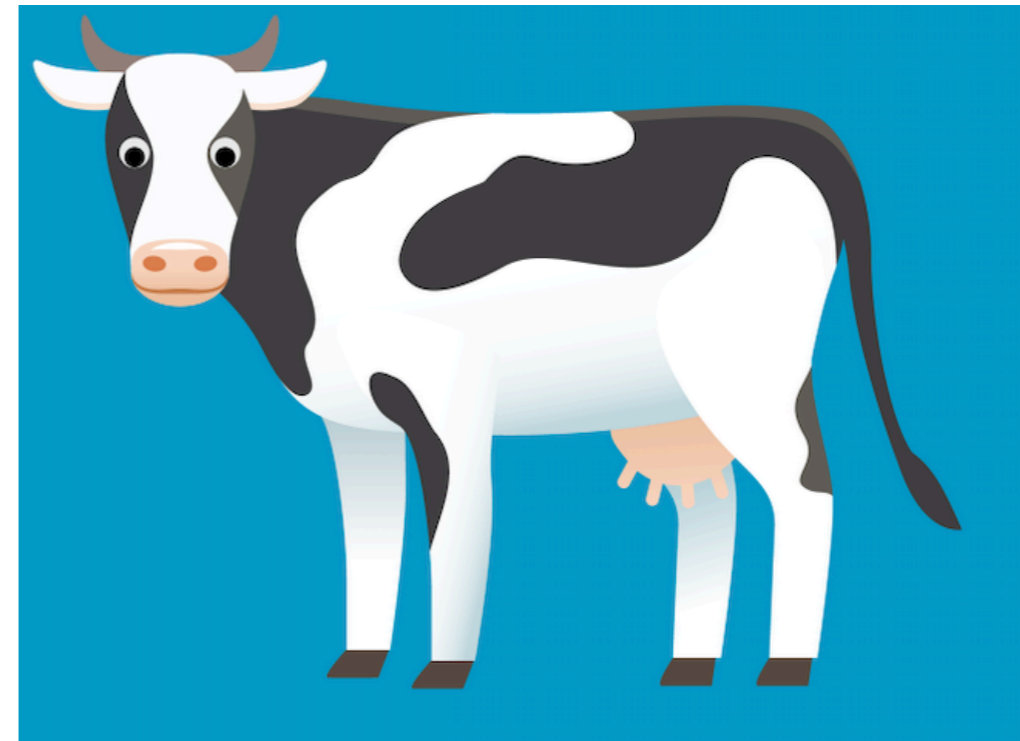
Linear Regression Games:
Convergence Guarantees to Approximate
Out-of-Distribution Solutions

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Failures due to **spurious correlations**



Usual cow: green background



Unusual cow: blue background

Deep neural networks exploit background color!



Motivation

- **Invariant risk minimization** (IRM) recently proposed to address out-of-distribution generalization
- Can IRM or subsequent methods **provably outperform** ERM even when data is generated from linear models?
 - No! due to **non-convex loss**
- How to build methods that provably outperform ERM?

Linear Regression Game

- Two environments, linear predictors, and square loss
- Environment 1 selects $\theta_1 \in \Theta$ to minimize its risk

$$\min_{\theta_1 \in \Theta} R_1(\theta_1, \theta_2) = \mathbb{E}_{\mathbb{P}_1}[(Y_1 - \theta_1^\top X_1 - \theta_2^\top X_1)^2]$$

- Environment 2 selects $\theta_2 \in \Theta$ to minimize its risk

$$\min_{\theta_2 \in \Theta} R_2(\theta_1, \theta_2) = \mathbb{E}_{\mathbb{P}_2}[(Y_2 - \theta_1^\top X_2 - \theta_2^\top X_2)^2]$$

- Environment 1 and 2 playing a **linear regression game (LRG)**
- Simultaneous minimization leads to a **Nash equilibrium (NE)** $(\theta_1^\dagger, \theta_2^\dagger)$
- **NE-based ensemble predictor** $\bar{\theta}^\dagger = \theta_1^\dagger + \theta_2^\dagger$

Linear Regression Game: Key Results

- Θ is ℓ_∞ ball: NE of LRG closer to ideal OOD solution
- Θ is ℓ_∞ ball: best response dynamics learn the NE

Thank you!