

Asian Deans' Forum 2026

THE RISING STARS

Women in Engineering Workshop

SEP 13 - 15 | HKUST, Hong Kong



Optimizing Service Operations via LLM-Powered Multi-Agent Simulation

PROBLEM & CHALLENGES

Services Operations Dominate Modern Economics (China, U.S., ...)

- (c.f. Manufacturing) Human-centric: Intensive Interactions

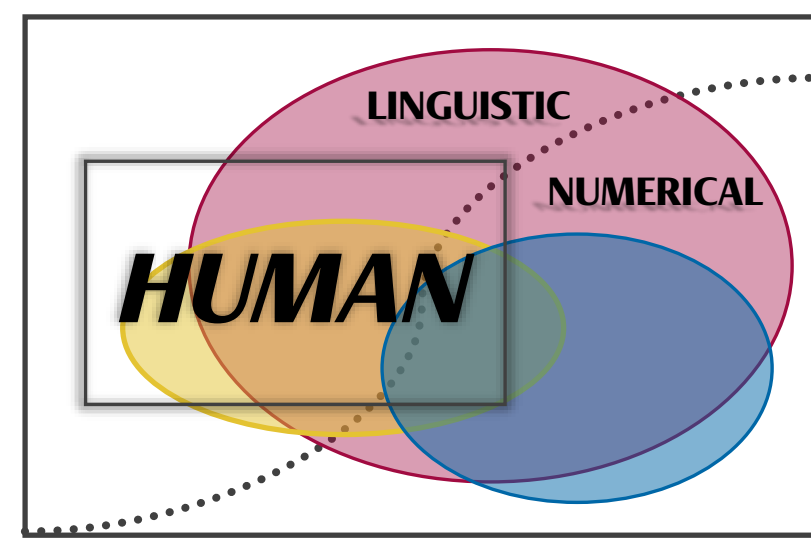
Designer's Decision-Making Problem: **System Design Optimization**

Systems	Supply Chains	Innovation Contests	Online Marketplaces
Designs	Sustainable Policies	Incentive Structures	Operational Levers
Participants	Multi-echelon Players	Contestants	Sellers, Consumers

Generic Task: Design Parameters $\rightarrow$ Participant Responses $\rightarrow$ Design Performance

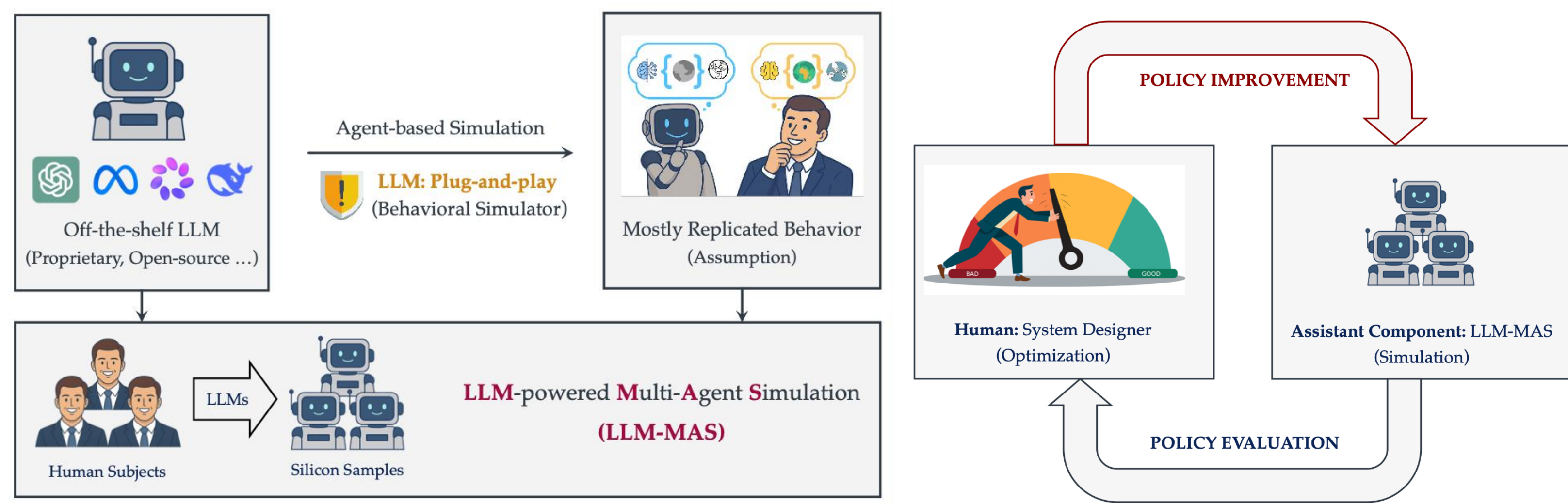
- Capturing **(Human) Behavior** Is Non-trivial!

Empirical Studies	Limited Data, Insufficient Granularity
Lab Experiments	Costly, Time-consuming, Hard to Scale
Analytical Models	Idealized Scenarios, Rigid Assumptions
LLM Simulations	Promising (Academia, Industry)



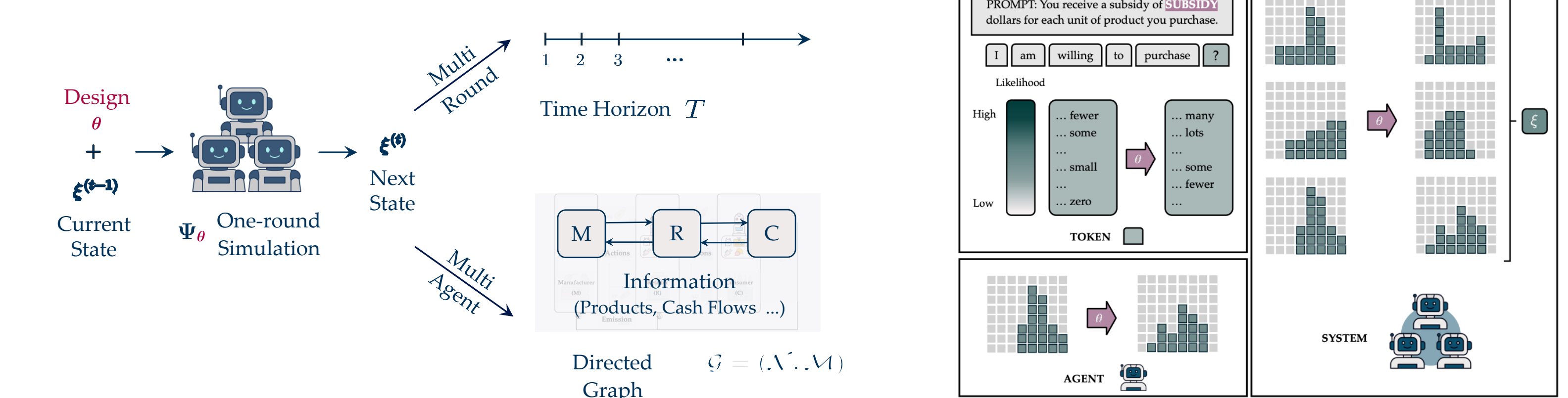
NOVEL FRAMEWORK: LLM-MAS

Our Contributions: Modeling – Algorithm – Empirics

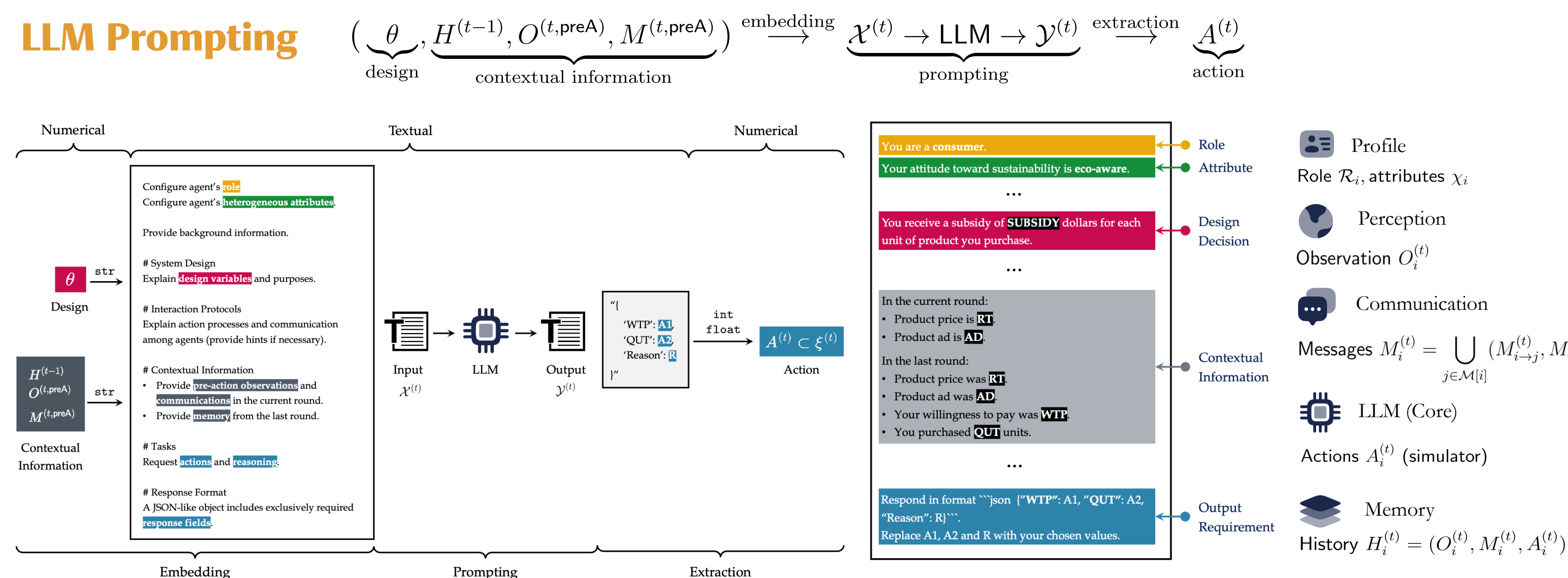


CONTRIBUTION: MODELING

Multi-Round, Multi-Agent Simulation $(\mathcal{G}, \Psi_\theta, T)$



LLM Prompting



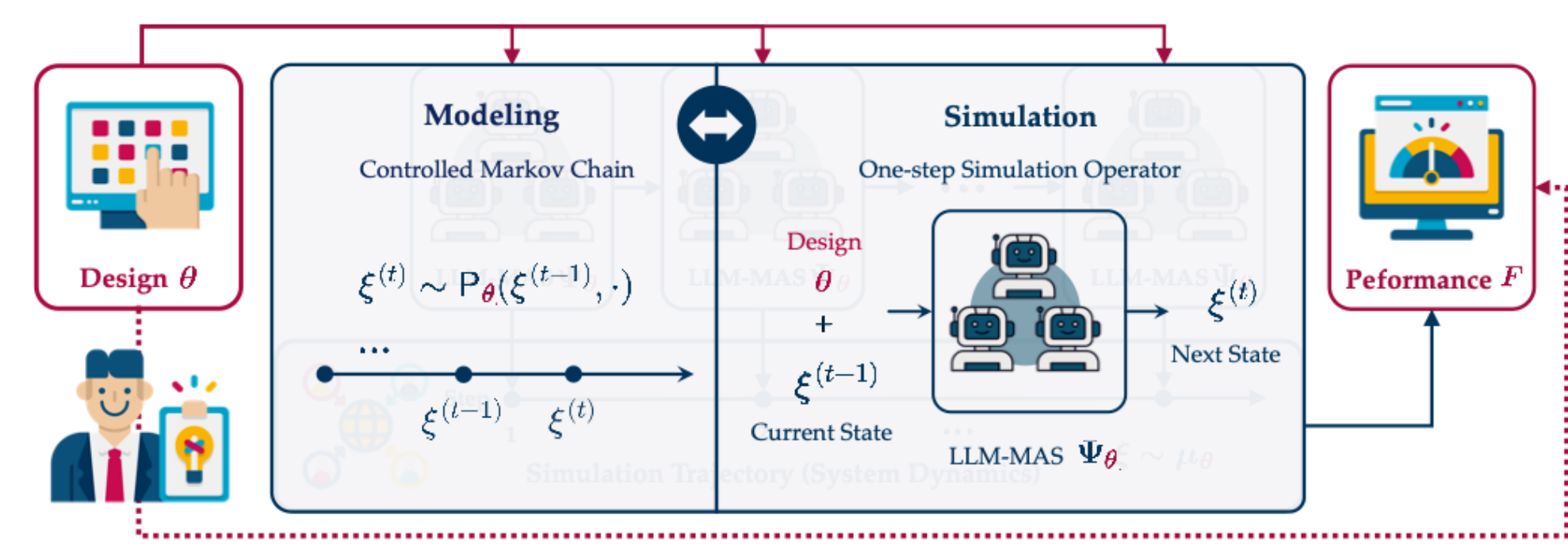
Controlled Markov Chain

- System Dynamics

Stochastic Optimization

- Decision-dependent uncertainty

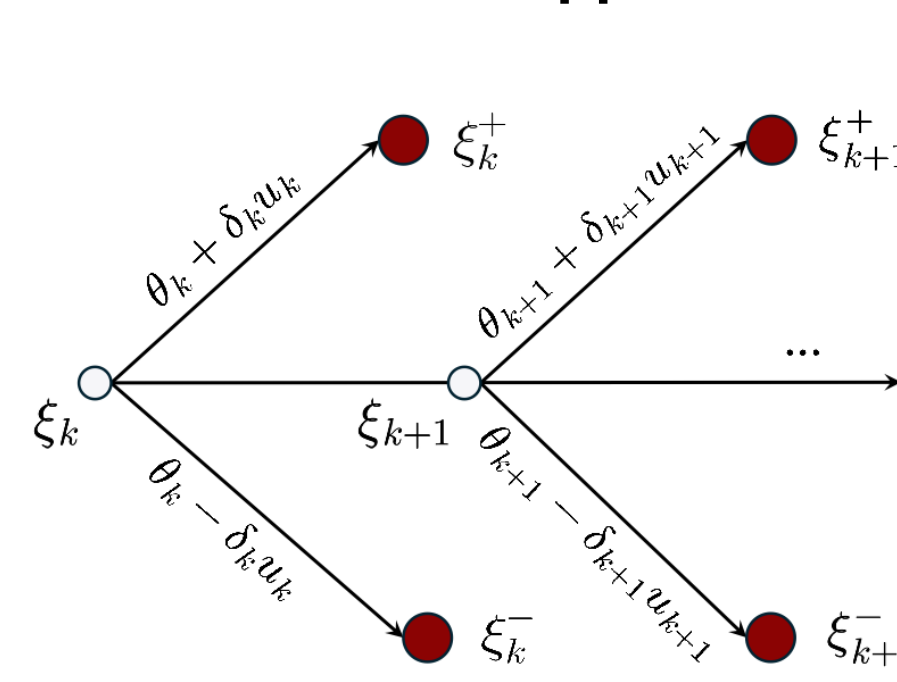
$$\min_{\theta \in \Theta} \left\{ f(\theta) := \mathbb{E}_{\xi \sim \mu_\theta} [F(\theta; \xi)] \right\}$$



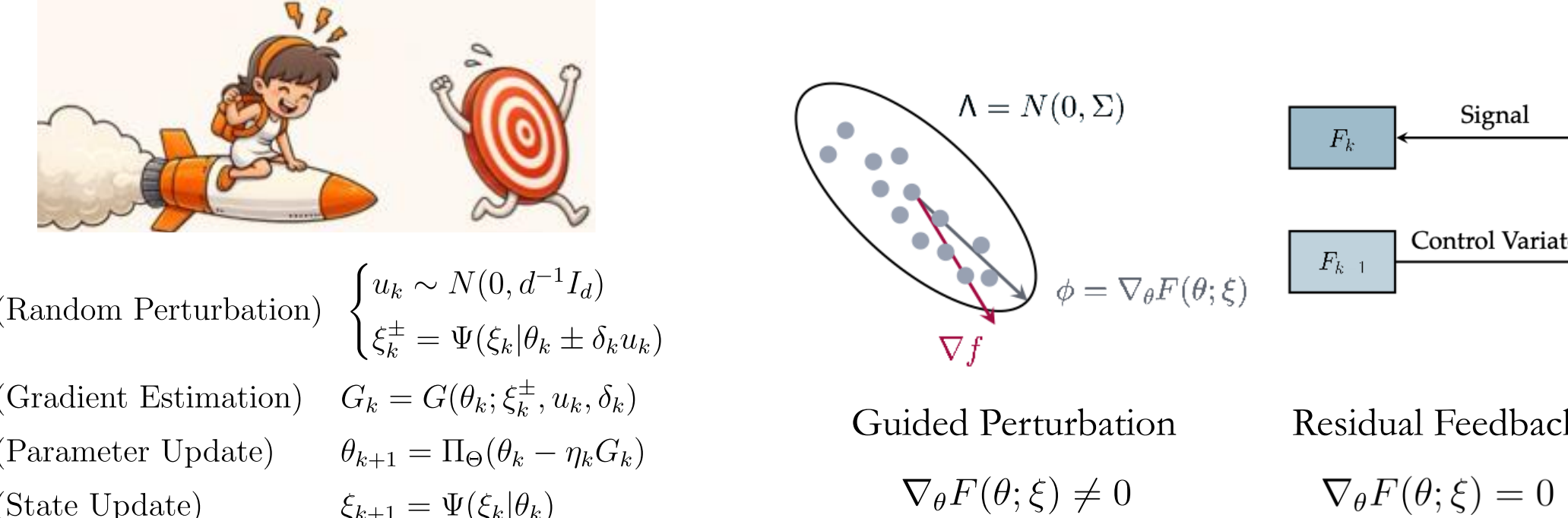
CONTRIBUTION: ALGORITHM

On-Trajectory Learning (OTL) Tackles LLM Deep Coupling, High Inference Latency

I. Zeroth-Order Approach



II. Variance Reduction



CONTRIBUTION: EMPIRICS

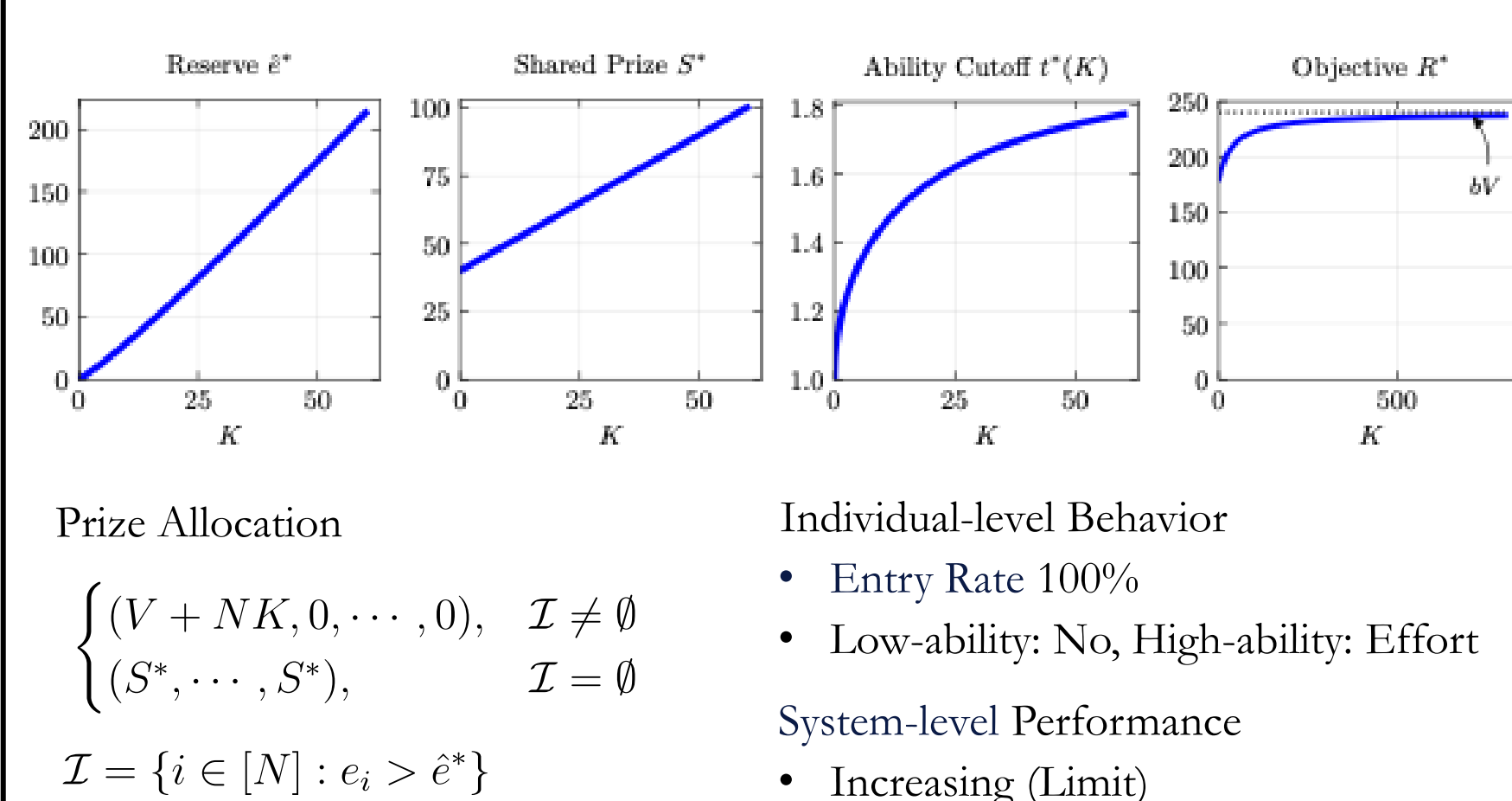


Innovation Contest Design: Incentive Structure

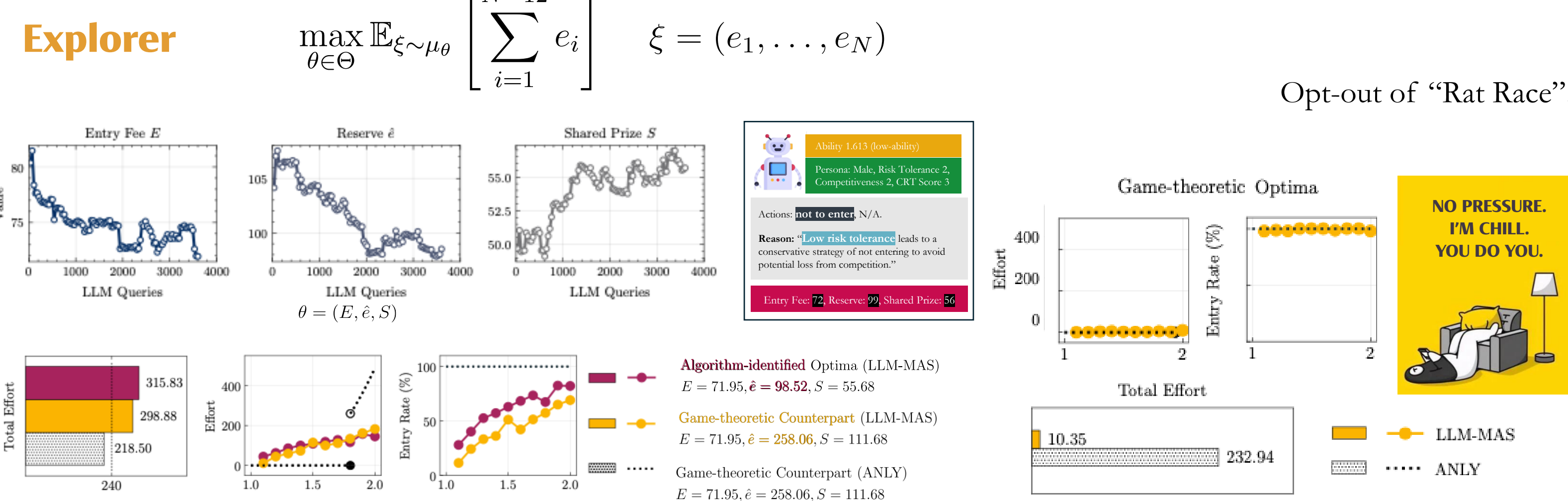
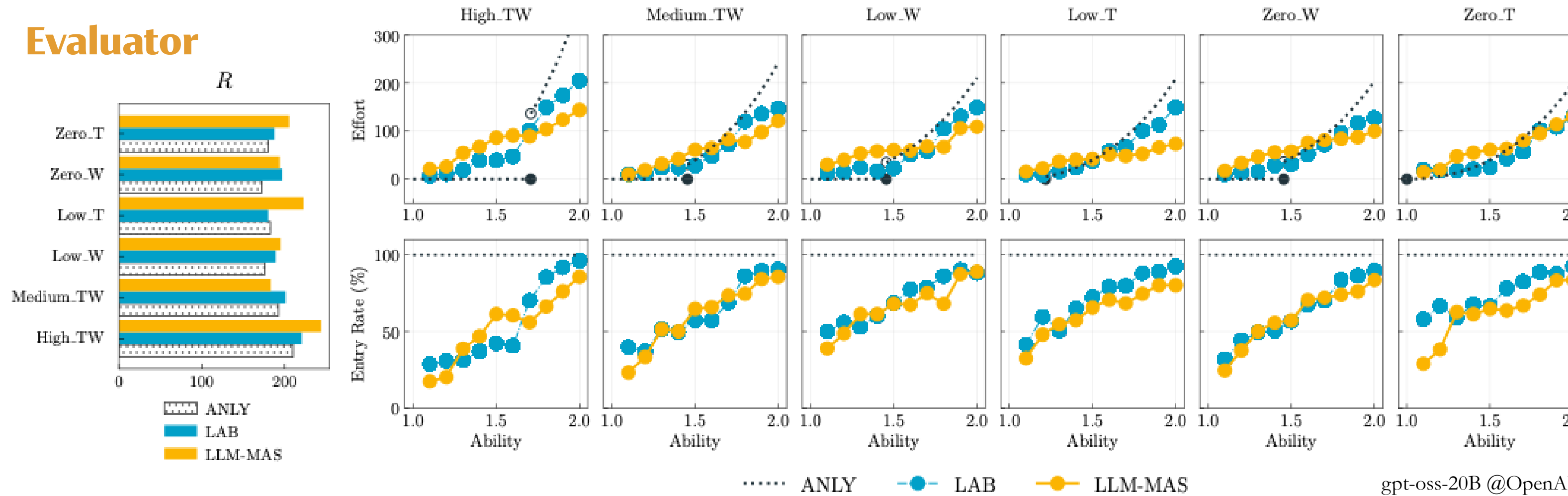
- c.f. Winner-Takes-All (WTA)
- Entry Fee, Reserve Threshold, Shared Prize

Traditional Approach: Lab Experiment

Traditional Approach: Game-Theoretic Analysis



Our Approach: LLM-MAS



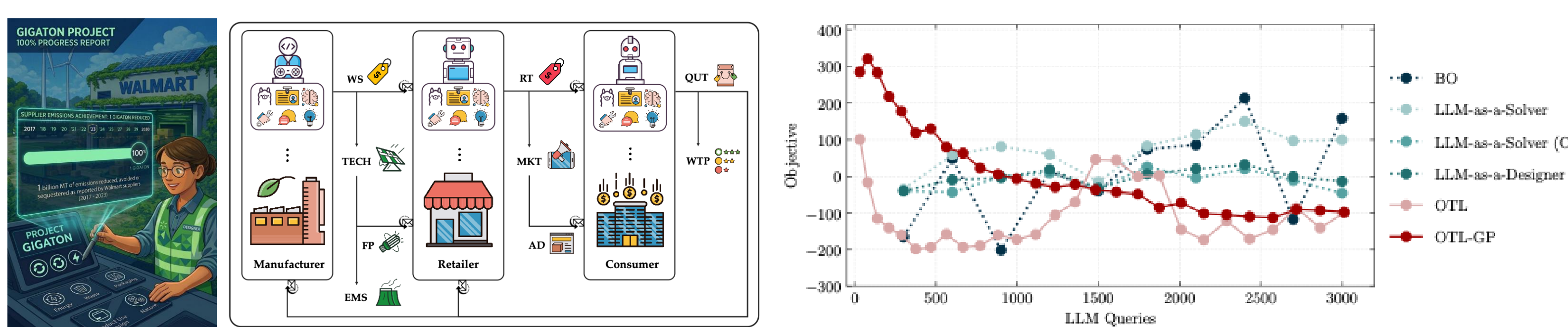
THEORETICAL GUARANTEE

Almost-Sure Convergence of OTL Iterates $\{\theta_k : k \in \mathbb{N}_+\}, \theta_\infty \in \Theta_0$

- | | |
|------------------------------------|---|
| A1. Objective: Lipchitz Continuous | A2. Markov Kernel: Smooth, Geometric Mixing |
| A3. Feasible Set: Compact, Convex | A4. Poisson Equation Solution: Smooth |
| A5. Step-size: Two-timescale | |

RESULTS & IMPLICATIONS

Prototyping Example: Walmart's Project Gigaton



R1. OTL Outperforms Baselines!

R2. LLM-MAS Complements Traditional Approaches!

FUTURE RESEARCH WORKS

Potential Directions of LLM-MAS&O

- | | |
|-----------------------------|---------------------------|
| 01. Refined Theory | 04. Efficient Deployment |
| 02. High-Fidelity Alignment | 05. Scalable Deployment |
| 03. Robust Deployment | 06. Boarder Applicability |

Organizers