	Jiachen Hu		
No.5 Yiheyuan Road, Haidian District, Beijing, P.R.China 100871		Email: NickH@pku.edu.cn Tel: (+86) 15911068937 Website: https://nickhclos.github.io/	
EDUCATION	School of Computer Science, Peking Ph.D candidate. Advisor: Liwei Wang	g University 2020.9 -	
	Turing Class, School of EECS, Pekin Bachelor of Science, Summa Cum Lau GPA: 3.815/4	ng University 2016.9 - 2020.6 de (Honorable BS)	
RESEARCH INTERESTS	• Understanding practical deep RL f based RL from a theoretical perspe- decision making, the theoretical pro-	ramework: the pros and cons of Transform- ective, the possibility of foundation model of operty of sim-to-real transfer.	
	• Machine learning assisted mathem (e.g., MCTS) in extremal combinat	atical discovery: search-based ML methods orics.	
	• Theoretical analysis of RL in quantum computers: the benefits of quantum accessible environments of RL, the fundamental separation of classical environments and quantum environments in terms of sample complexity of exploration.		
	• The application of RL in economic methods with RL in automated ma oretical algorithms in sequential eco ogenous noises).	• The application of RL in economical models: designing empirical data-driven methods with RL in automated market makers, designing provably efficient the- oretical algorithms in sequential economical models (e.g., in the presence of ex- ogenous noises).	
	• Provable sample efficient RL: online exploration in MDP and structured POMDP, offline RL		
	• Classical online learning: multi-arm	ed bandits, linear bandits	
SELECTED PROJECTS	Distributed Acceleration of Bandits Advisor: Liwei Wang	2019	
	This project focuses on understanding theoretical acceleration of reinforcement learning in a collaborated distributed system. We hope to figure out two crucial quantity: the optimal speedup in terms of sample complexity, and the minimum communication over- head in order to achieve this optimal speedup. Starting from multi-armed bandits and linear bandits, we showed that the optimal linear speedup is possible. We constructed several sophisticated communication protocols to reduce the communication overhead, and proved that the communication overhead is optimal (excluding logarithmic terms) in multi-armed bandits setting.		
	The paper was accepted in ICLR 2020, wh the tradeoff between communication over RL. It was followed by many theoretical	he paper was accepted in ICLR 2020, which opened up a new direction of investigating e tradeoff between communication overhead and sample complexity in bandits and L. It was followed by many theoretical works since then.	
	Understanding the Theoretical Bend Advisor: Chi Jin, Lihong Li, Liwei Wang	efits of Sim-to-Real Transfer 2021-2022	
	We investigated the theoretical benefits of ogy to reduce the high cost of obtaining r	of sim-to-real transfer, a promising methodol- real-world training data. As the first attempt	

ogy to reduce the high cost of obtaining real-world training data. As the first attempt to analyze the theoretical benefits, we formalize the simulator as a set of models with tunable parameters. We study the sim-to-real gap of two famous algorithms: domain randomization and robust adversarial training. Our results showed that these two algorithms are highly efficient when the simulator set has some structures and the policy is history-dependent.

Two papers are published in ICLR 2022/2023, the former of which is accepted as a spotlight. This project provides insight of the crucial component in sim-to-real transfer.

The limitation of Transformers on HMM-like models Advisor: Chi Jin

2023 - 2024

Understanding the efficiency of Transformers in Hidden Markov Models (HMM) and its variants is crucial for developing robust Transformer-based RL agents, given that HMM serves as one of the fundamental sequential models. This study investigates the predictive capabilities of Transformers and Recurrent Neural Networks (RNN) in HMM-like models, focusing on both next observation and hidden state predictions. We designed HMM-like instances with varying levels of complexity and evaluated the performance of standard RNNs and Transformers. While RNNs consistently demonstrated strong performance across all tasks, shallow Transformers exhibited slightly larger prediction errors in simpler and medium complexity HMM-like models, and struggled to adapt to more challenging scenarios. Complementing our empirical findings, we constructed a theoretical framework for shallow Transformers tailored to simple and medium HMM-like models, demonstrating its efficacy in fitting these models accurately.

PREPRINTS (* denotes equal contributions)

On Limitation of Transformer for Learning HMMs

- Jiachen Hu, Qinghua Liu, Chi Jin
- Arxiv preprint

PUBLICATIONS (* denotes equal contributions)

Provably Efficient Exploration in Quantum Reinforcement Learning with Logarithmic Worst-Case Regret

- Han Zhong*, Jiachen Hu*, Yecheng Xue, Tongyang Li, Liwei Wang
- ICML 2024

Quantum Non-Identical Mean Estimation: Efficient Algorithms and Fundamental Limits

- Jiachen Hu, Tongyang Li, Xinzhao Wang, Yecheng Xue, Chenyi Zhang, Han Zhong (alphabetical order)
- TQC 2024

ZeroSwap: Data-driven Optimal Market Making in DeFi

- Viraj Nadkarni, Jiachen Hu, Ranvir Rana, Chi Jin, Sanjeev Kulkarni, Pramod Viswanath
- FC 2024

Provable Sim-to-real Transfer in Continuous Domain with Partial Observations

- Jiachen Hu*, Han Zhong*, Chi Jin, Liwei Wang
- ICLR 2023

Understanding Domain Randomization for Sim-to-real Transfer

- Xiaoyu Chen*, Jiachen Hu*, Chi Jin, Lihong Li, Liwei Wang
- ICLR 2022 (spotlight, 6% 176/3328)

Near-Optimal Reward-Free Exploration for Linear Mixture MDPs with Plug-in Solver

- Xiaoyu Chen, Jiachen Hu, Lin F. Yang, Liwei Wang
- ICLR 2022 (spotlight, 6% 176/3328)

	 Near-optimal Representation Learning for Linear 2 Jiachen Hu*, Xiaoyu Chen*, Chi Jin, Lihong Li, Liv ICML 2021 	Bandits and Linear RL vei Wang	
	Efficient Reinforcement Learning in Factored MDPs with Application to Constrained RL		
	• Xiaoyu Chen, Jiachen Hu, Lihong Li, Liwei Wang		
	• ICLR 2021		
	Distributed Bandit Learning: Near-Optimal Regret with Efficient Commu- nication		
	• Yuanhao Wang*, Jiachen Hu*, Xiaoyu Chen, Liwei Wang		
	• ICLR 2020		
INTERNSHIPS	Summer intern at UIUC Host: Nan Jiang	2019.7 - 2019.8	
	Visiting scholar at Princeton University Host: Chi Jin	2023.6 - 2023.11	
SERVICES	Conference reviewer: ICLR (2021,2022,2023), ICML (2022), NeurIPS (2021,2023)		
	Journal reviewer: AURO		
HONORS	Excellent Academic Research Reward (2022)		
	Peking University President Scholarship (2020-2021)		
	Turing Class Summit Scholarship (2019)		
	Jane Street Electronic Trading Challenge 2018 in Tsinghua University, 3rd Place		
	Wu Si Scholarship (2018)		
	Merit Student of Peking University (top 5%, 2017)		
	Robin Li Scholarship (top 1%, 2017)		
	Third-class annually New Students Scholarship (2016)		
	National Olympic of informatics (NOI) 2015, Sliver Medal		