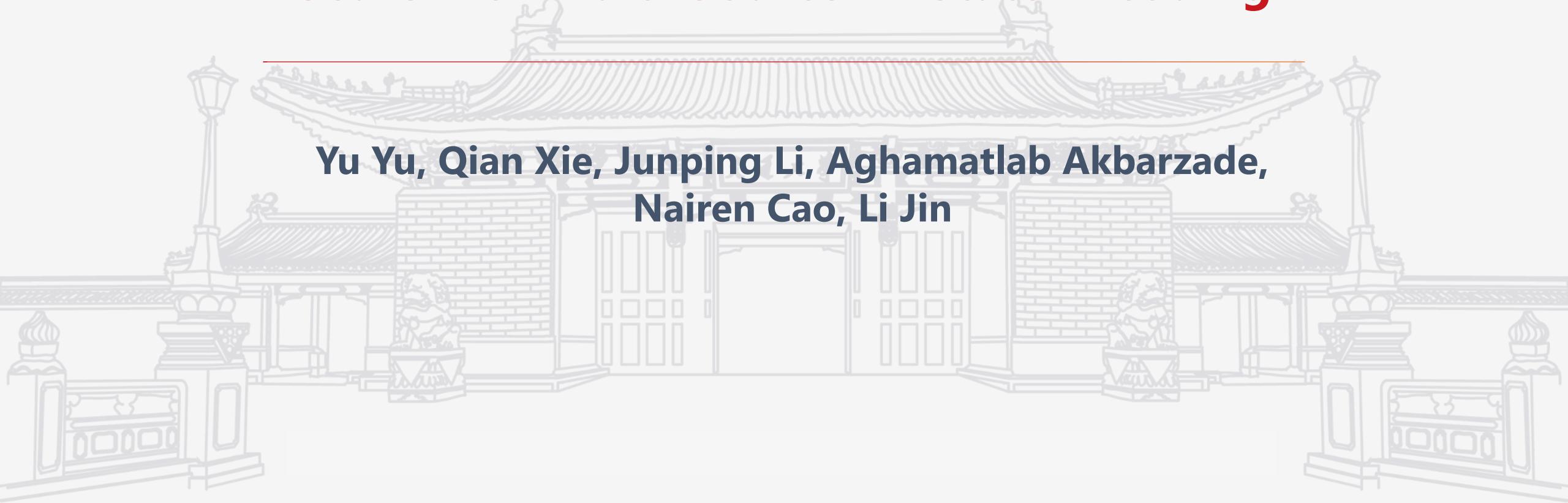


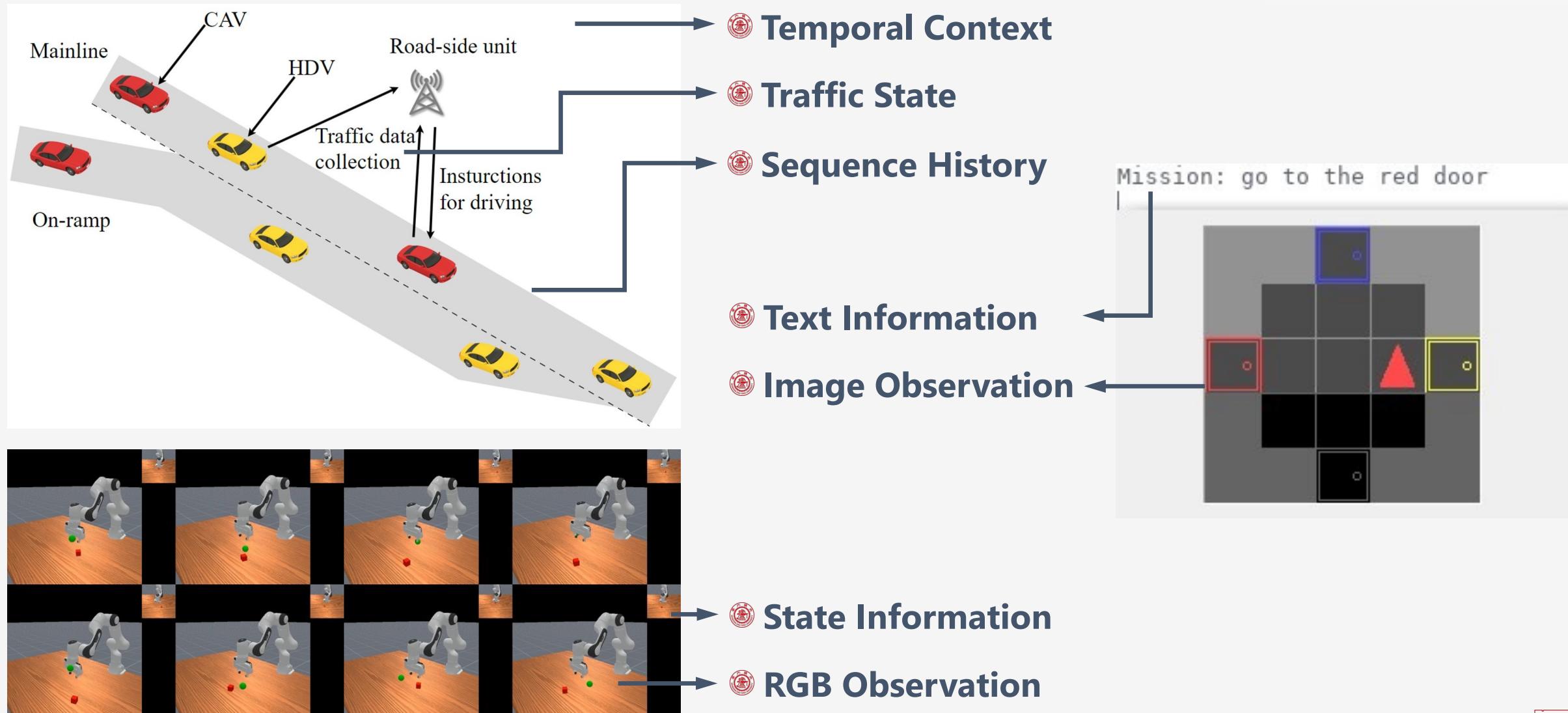
---

# **LLM-Driven Composite Neural Architecture Search for Multi-Source RL State Encoding**



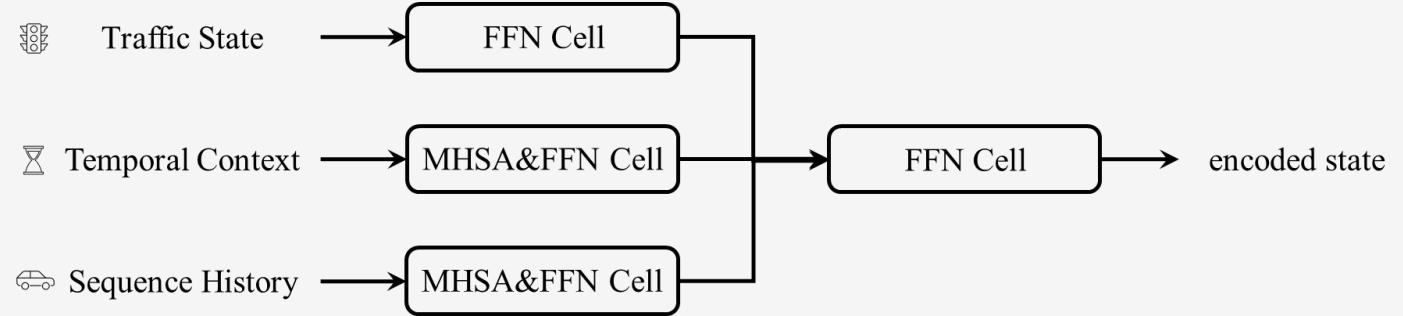
**Yu Yu, Qian Xie, Junping Li, Aghamatlab Akbarzade,  
Nairen Cao, Li Jin**

# Introduction

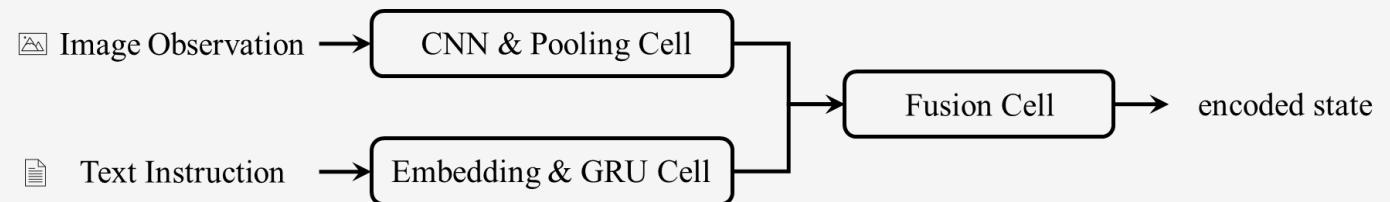


# Multi-Source RL State Encoding

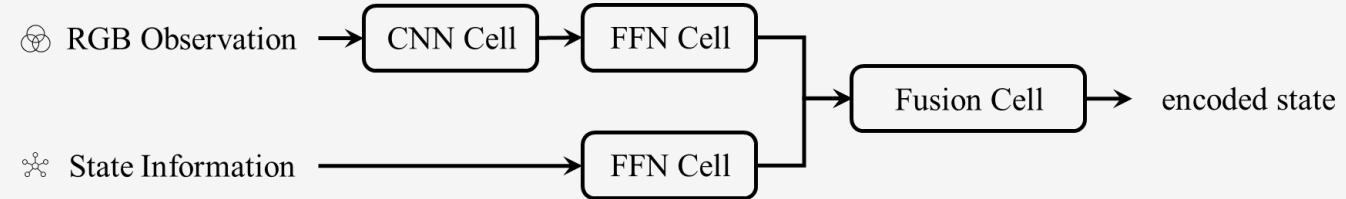
## Mixed-autonomy traffic control



## MiniGrid goal-oriented tasks



## ManiSkill robotic control

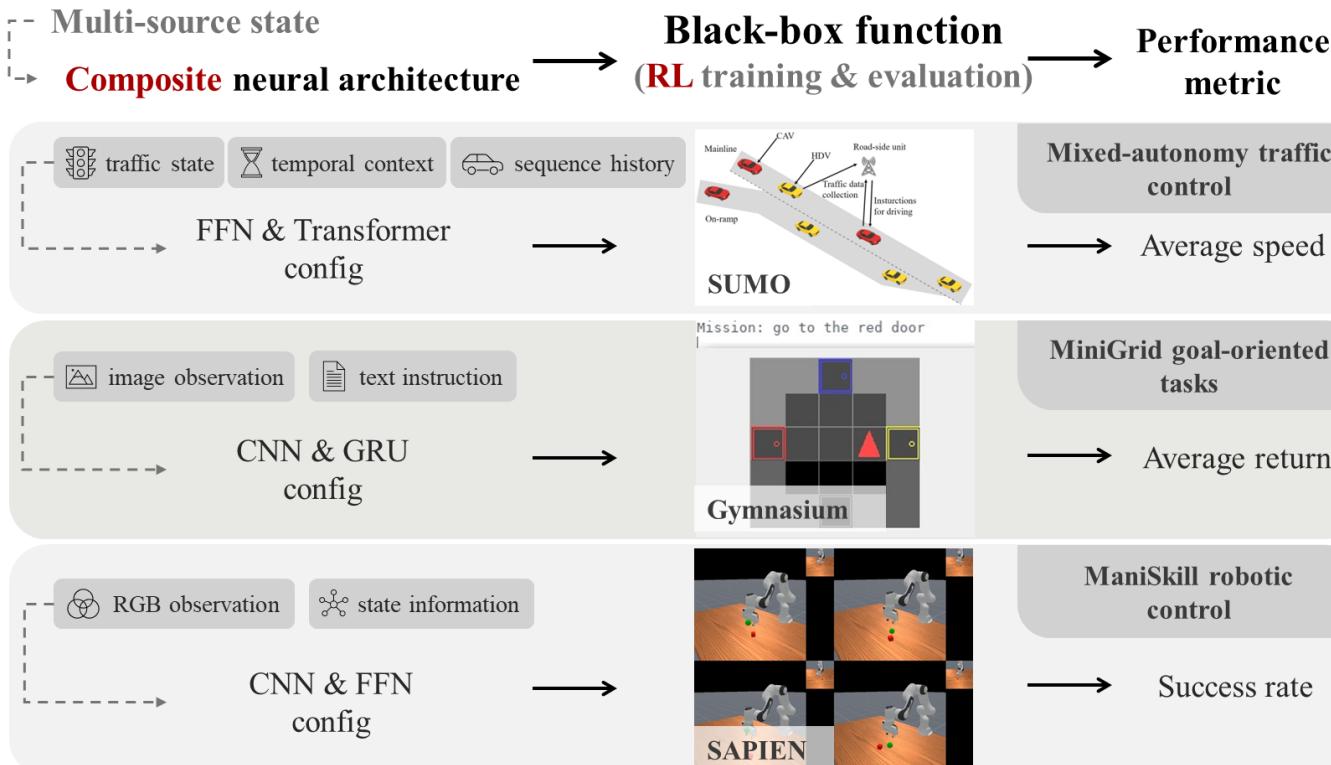
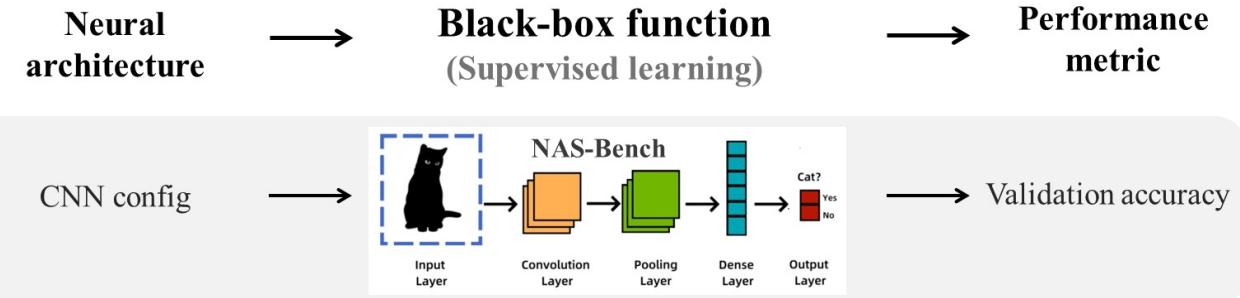


 Designing state encoders for multi-source-RL remains underexplored.



# Composite Neural Architecture Search

## NAS for supervised learning



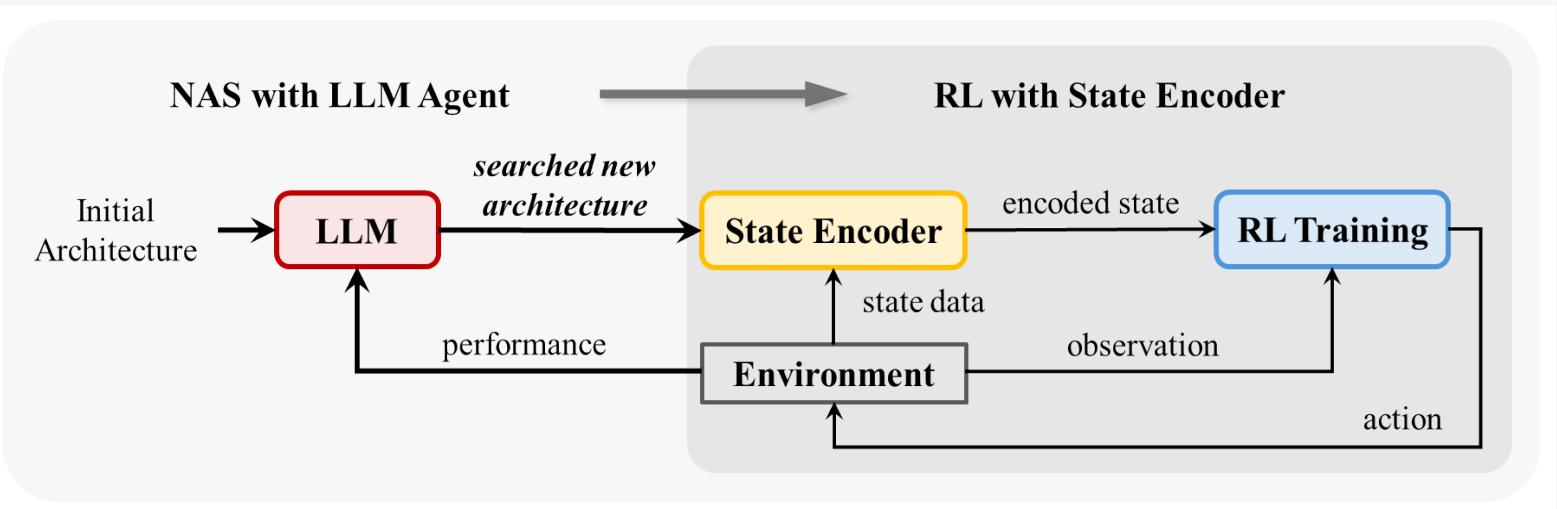
## NAS for multi-source RL state encoding

Unlike supervised learning, multi-source RL requires composite NAS.



# LACER: An LLM → State Encoder → RL Pipeline

## LLM-driven Neural Architecture Search for Composite State Encoders in RL



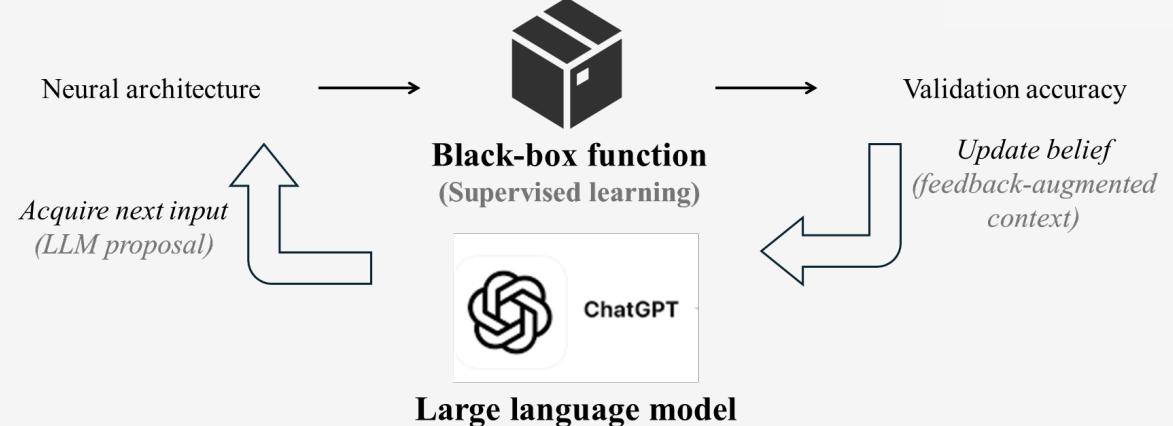
## Module-specific search space

Table 1: Module-specific search spaces where bold values denote the configurations used by the Expert baseline.

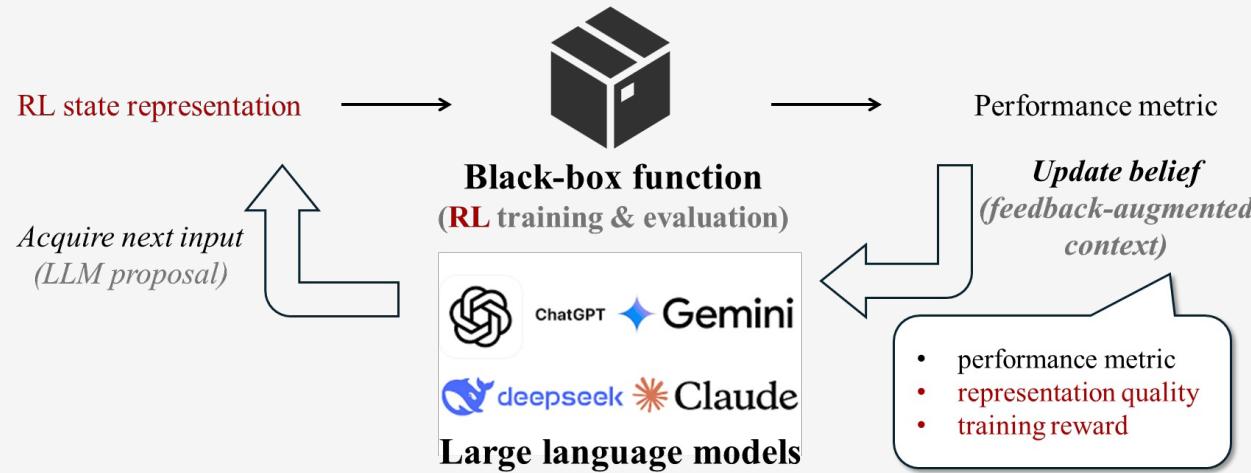
Module	Operation	Heads / Activation	Dimension	Ratio	Depth
Time	MHSA, FFN	{ <b>2</b> , 4, 8}	{ <b>8</b> , 16, 32}	{1, <b>2</b> , 4}	{1, <b>2</b> , 3}
Traffic	FFN	{relu, gelu, swish}	{16, <b>32</b> , 64}	{1, 2, 4}	{1, 2, 3}
Sequence	MHSA, FFN	{2, <b>4</b> , 8}	{8, <b>16</b> , 32}	{1, <b>2</b> , 4}	{1, <b>2</b> , 3}
Fusion	FFN	{relu, gelu, swish}	{64, <b>128</b> , 256}	{1, 2, 4}	{1, 2, 3}

# Pipeline comparison

## GENIUS (existing LLM-based NAS)



## LACER (ours)

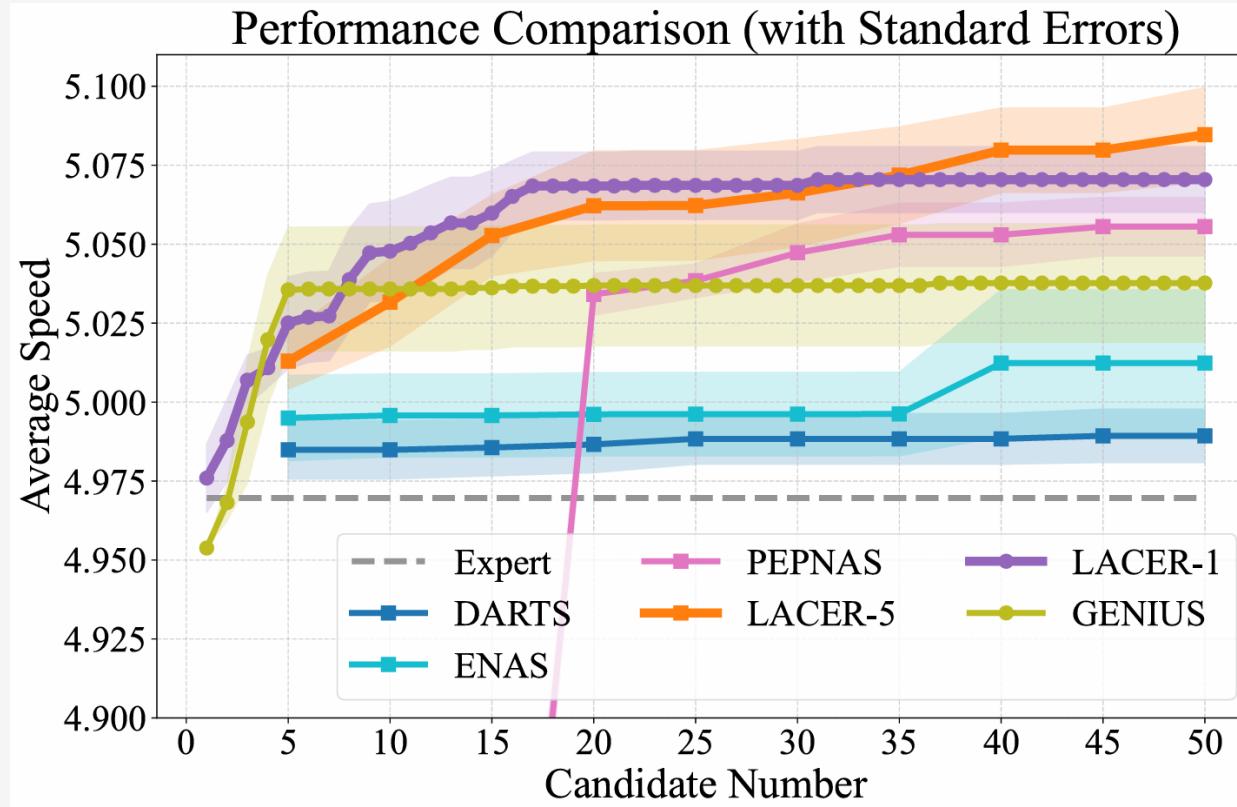


Exploit side info on source-specific encoders beyond performance metric.



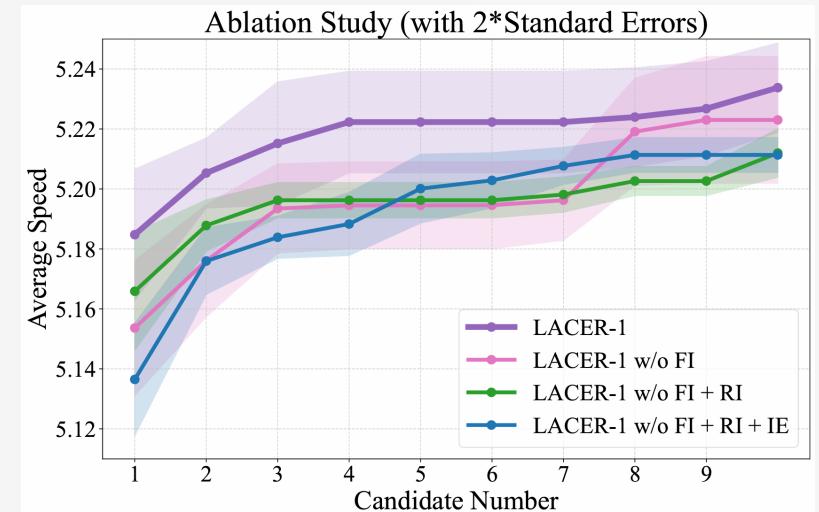
# Results

## Comparison between LACER and baselines

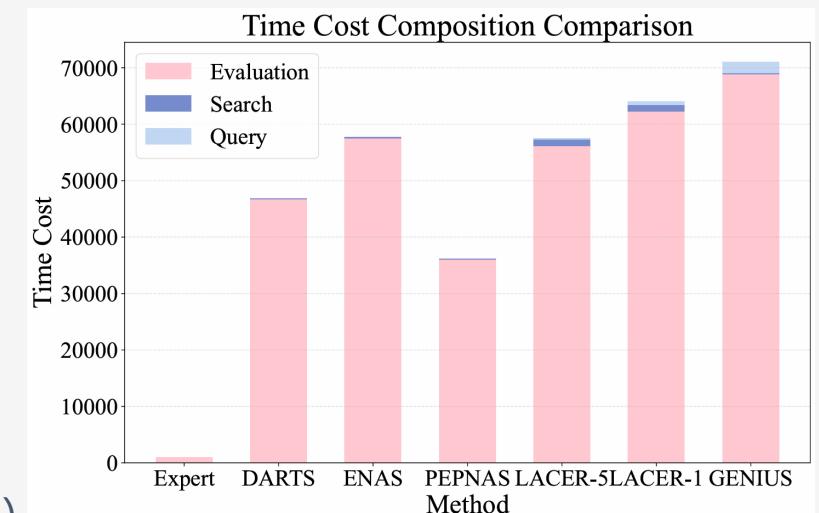


LACER-5 denotes the batch variant (5 candidates per iteration).

## Ablation studies

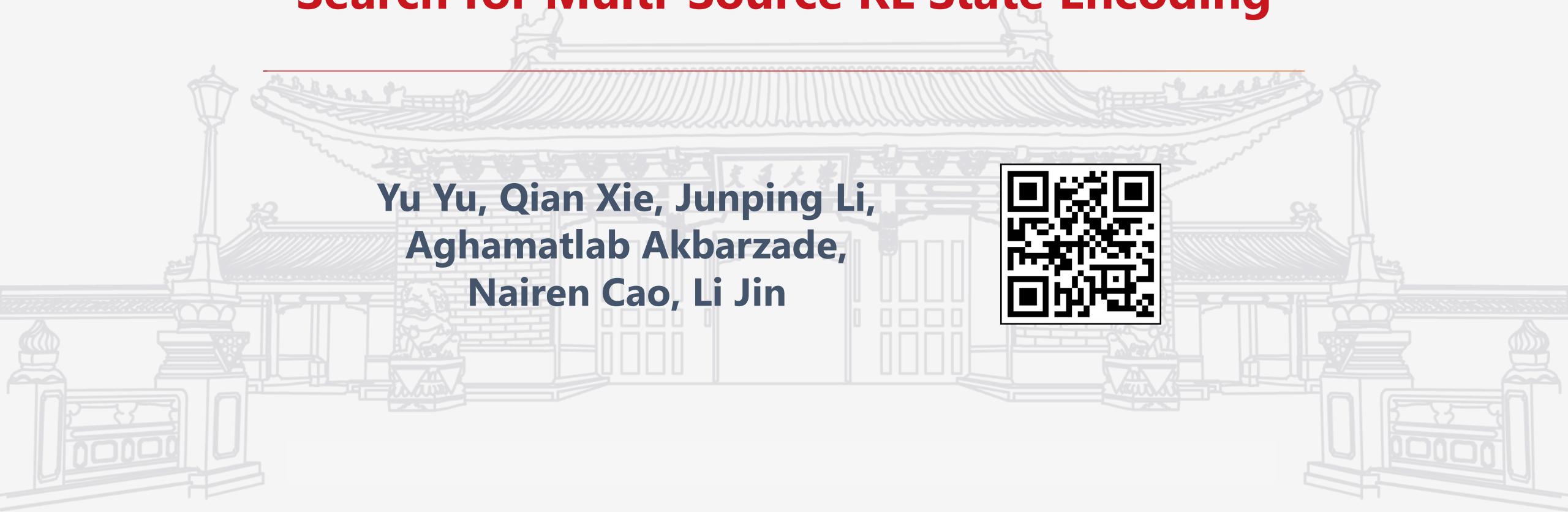


## Computation time



---

# LLM-Driven Composite Neural Architecture Search for Multi-Source RL State Encoding



Yu Yu, Qian Xie, Junping Li,  
Aghamatlab Akbarzade,  
Nairen Cao, Li Jin

