

# Learning Memory Mechanisms for Decision Making through Demonstrations

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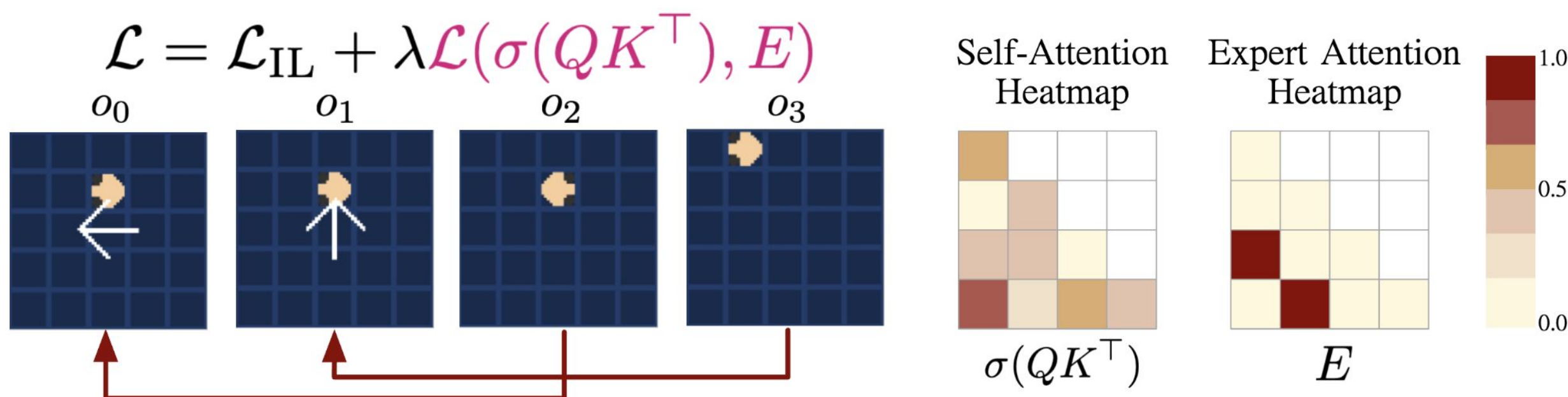


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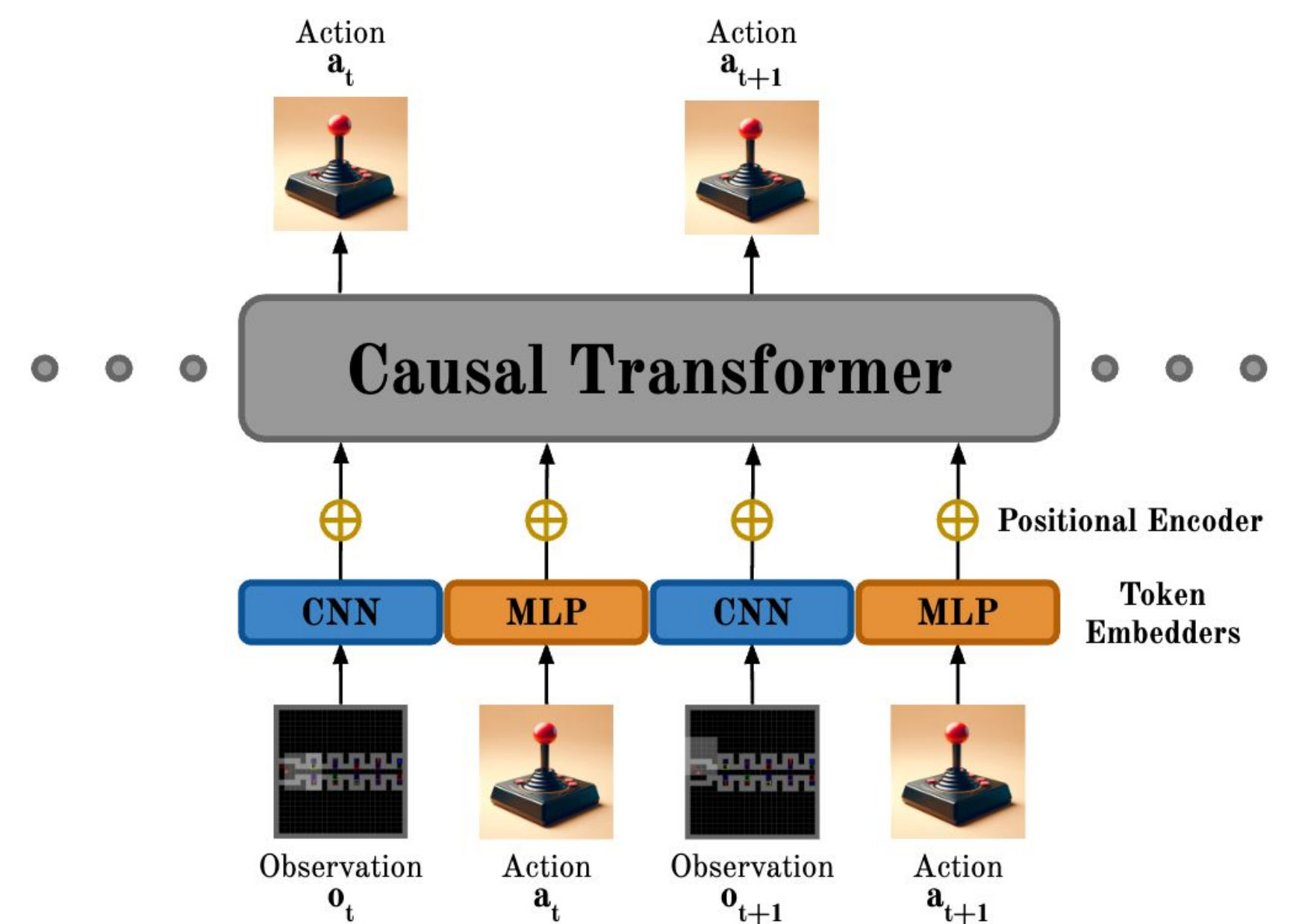


Teaching a Transformer how to use memory leads to much faster learning than letting it learn on its own.

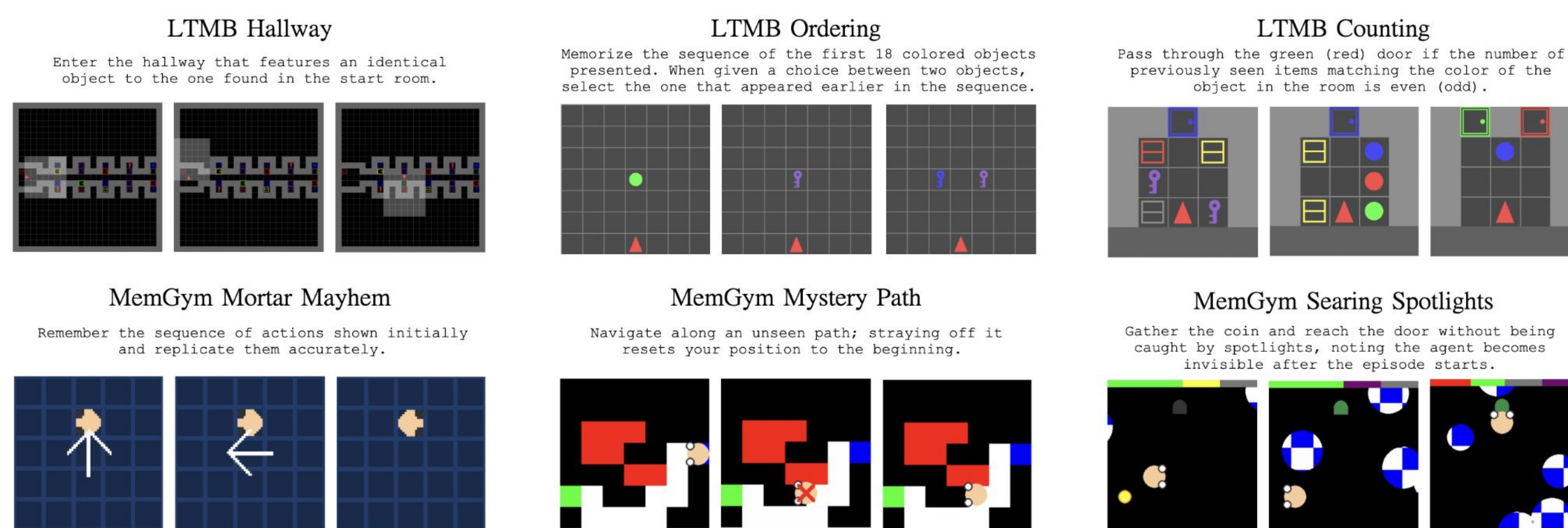
## Method



## Architecture



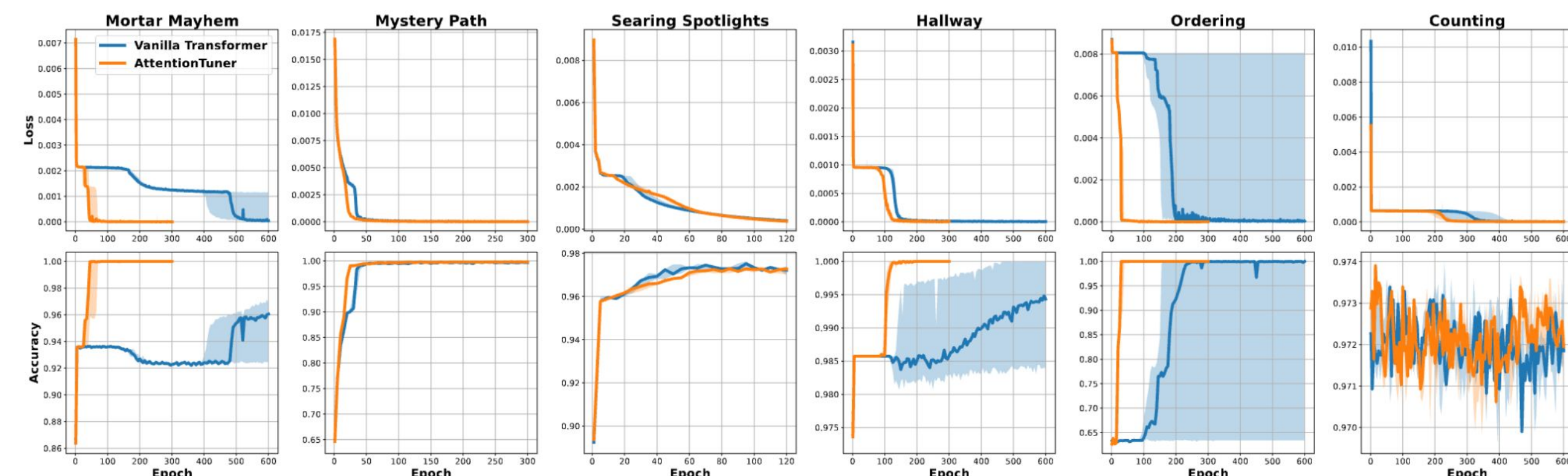
## Experiments



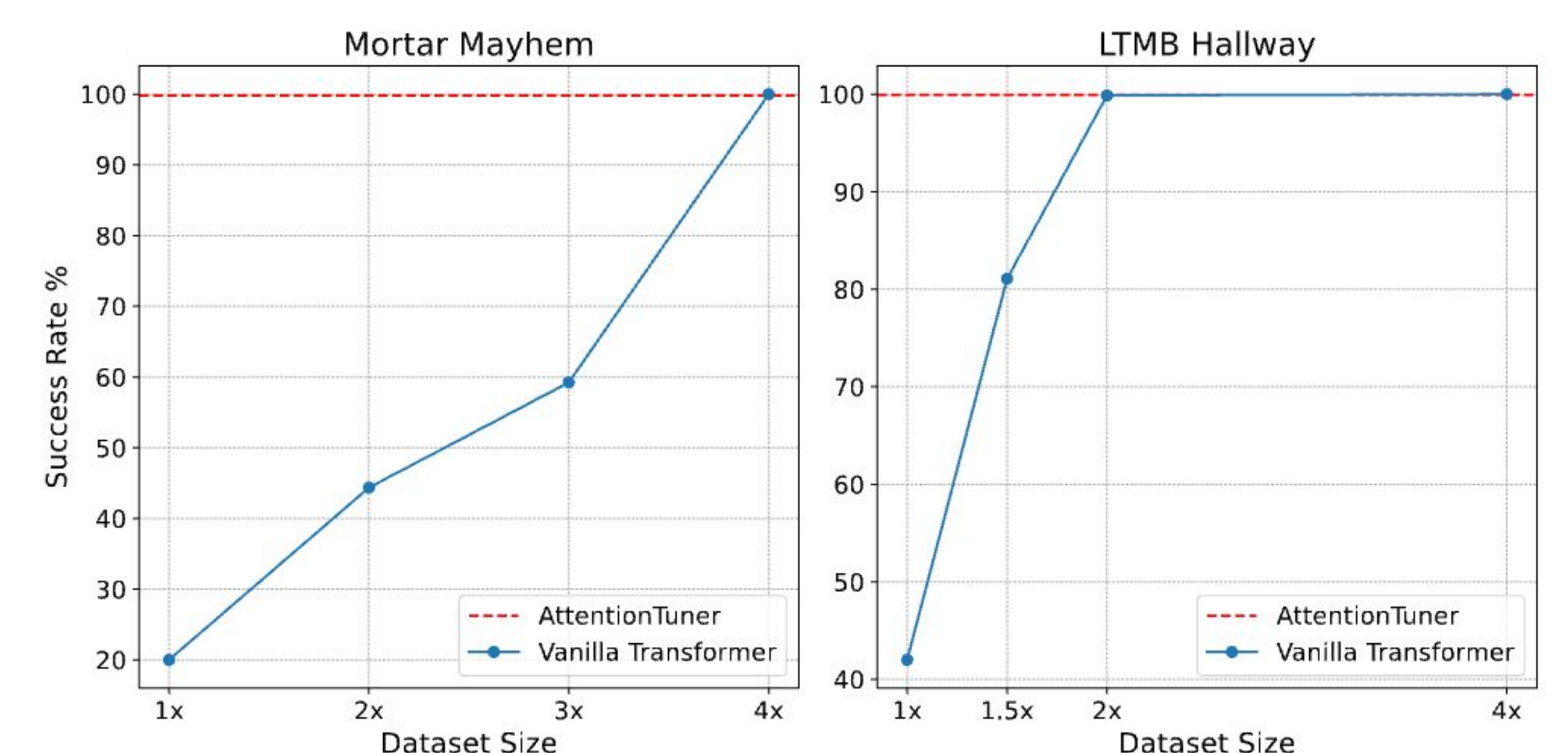
## Results

METHODS	MORTAR MAYHEM	MYSTERY PATH	SEARING SPOTLIGHTS	HALLWAY	ORDERING	COUNTING
VANILLA TRANSFORMER	20.8 ± 42.2	97.3 ± 0.7	62.2 ± 5.5	53.2 ± 28.3	59.4 ± 22.9	6 ± 0.7
ATTENTIONTUNER (OURS)	<b>99.8 ± 0.4</b>	<b>98.7 ± 0.4</b>	<b>64.2 ± 3.4</b>	<b>99.9 ± 0.1</b>	<b>99.9 ± 0.3</b>	<b>6.5 ± 0.4</b>

Table 1: Average success rates and 90% confidence intervals for two different methods—Vanilla Transformer and AttentionTuner (our approach)—across various tasks in the Memory Gym and Long-term Memory Benchmark.



## Scaling



## Data Efficiency

