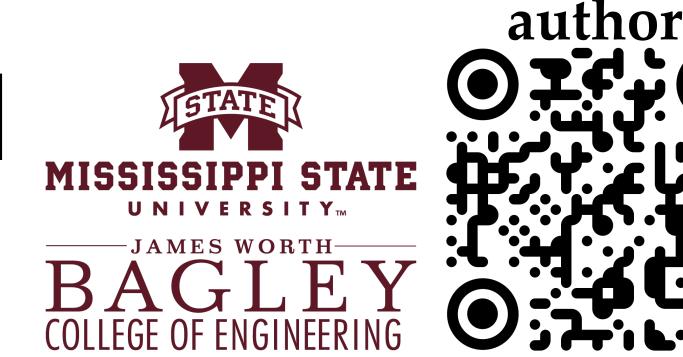
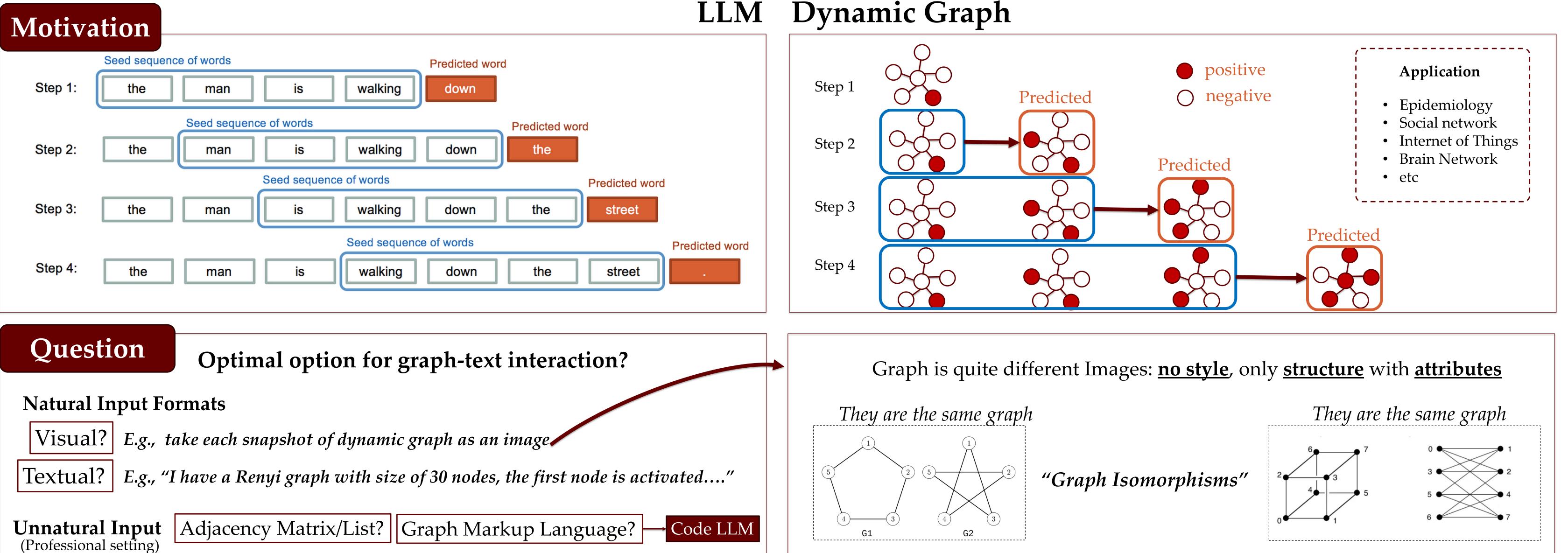
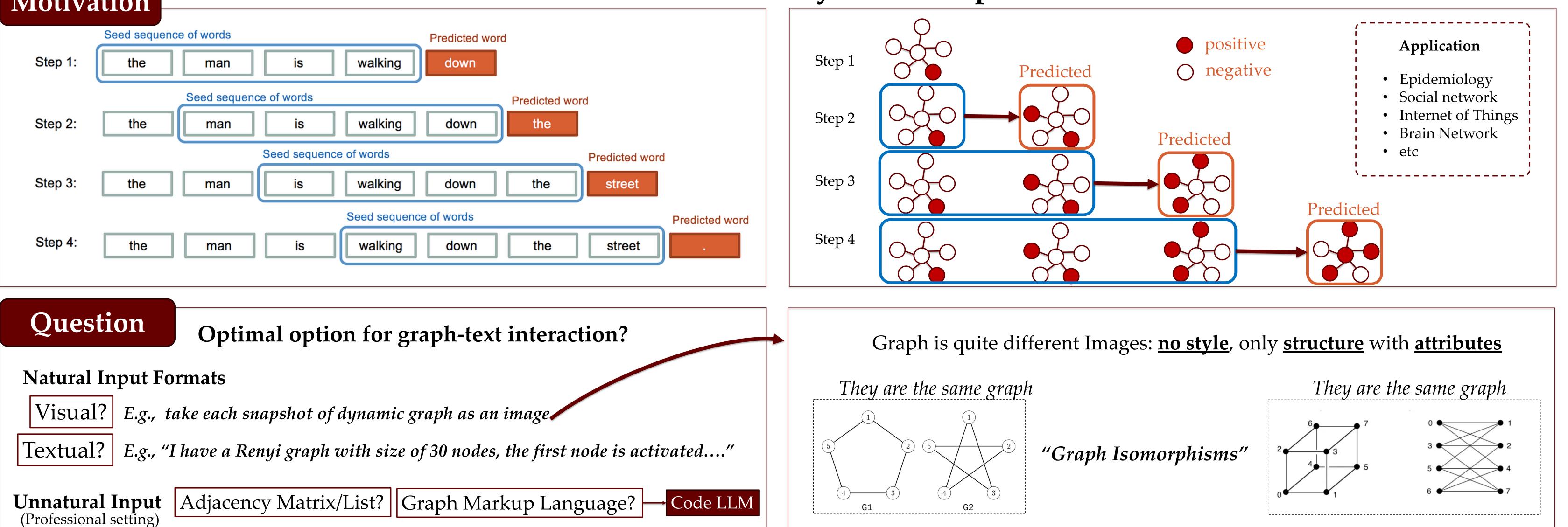
# FlowGPT: How Long Can LLMs Trace Back and Predict the Trends of Graph Dynamics?

via **Sslido** 

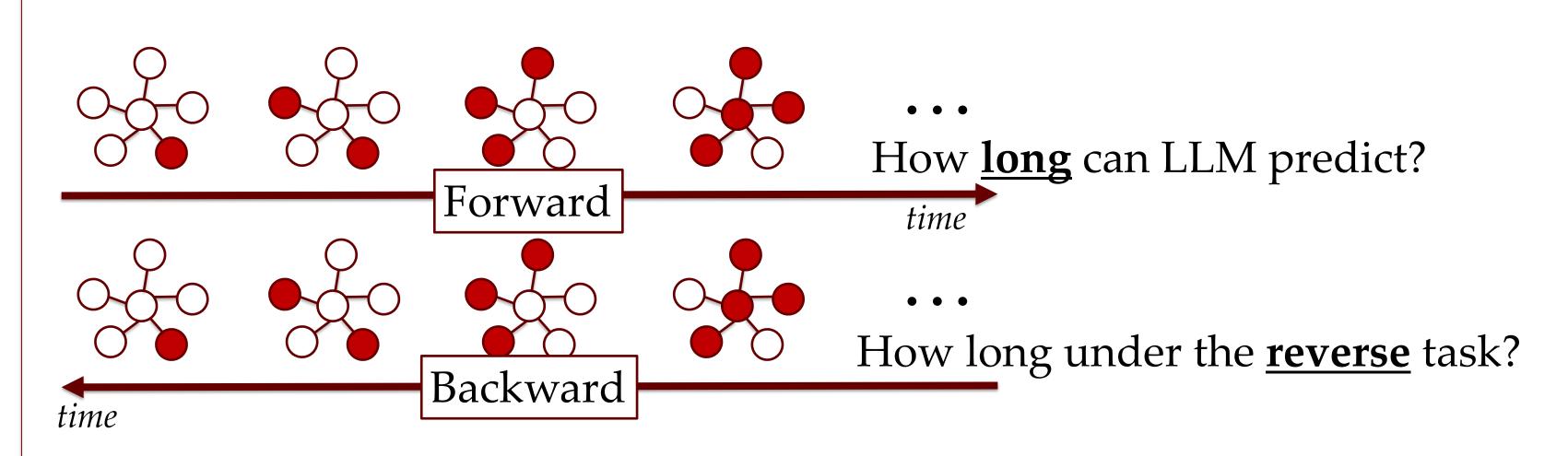


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## Task: Dynamic Graph



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Can ask anonymously

### **Direction and step size combination**

**FW1**: Step differences is 1 and forward prediction. including step  $1 \rightarrow 2, 2 \rightarrow 3$  ... **BW2**: Step differences is 2 and backward prediction. including step  $3 \rightarrow 1, 5 \rightarrow 3$  ... FW1\_2: mixed with FW1 and FW2

#### **Data and Metrics**

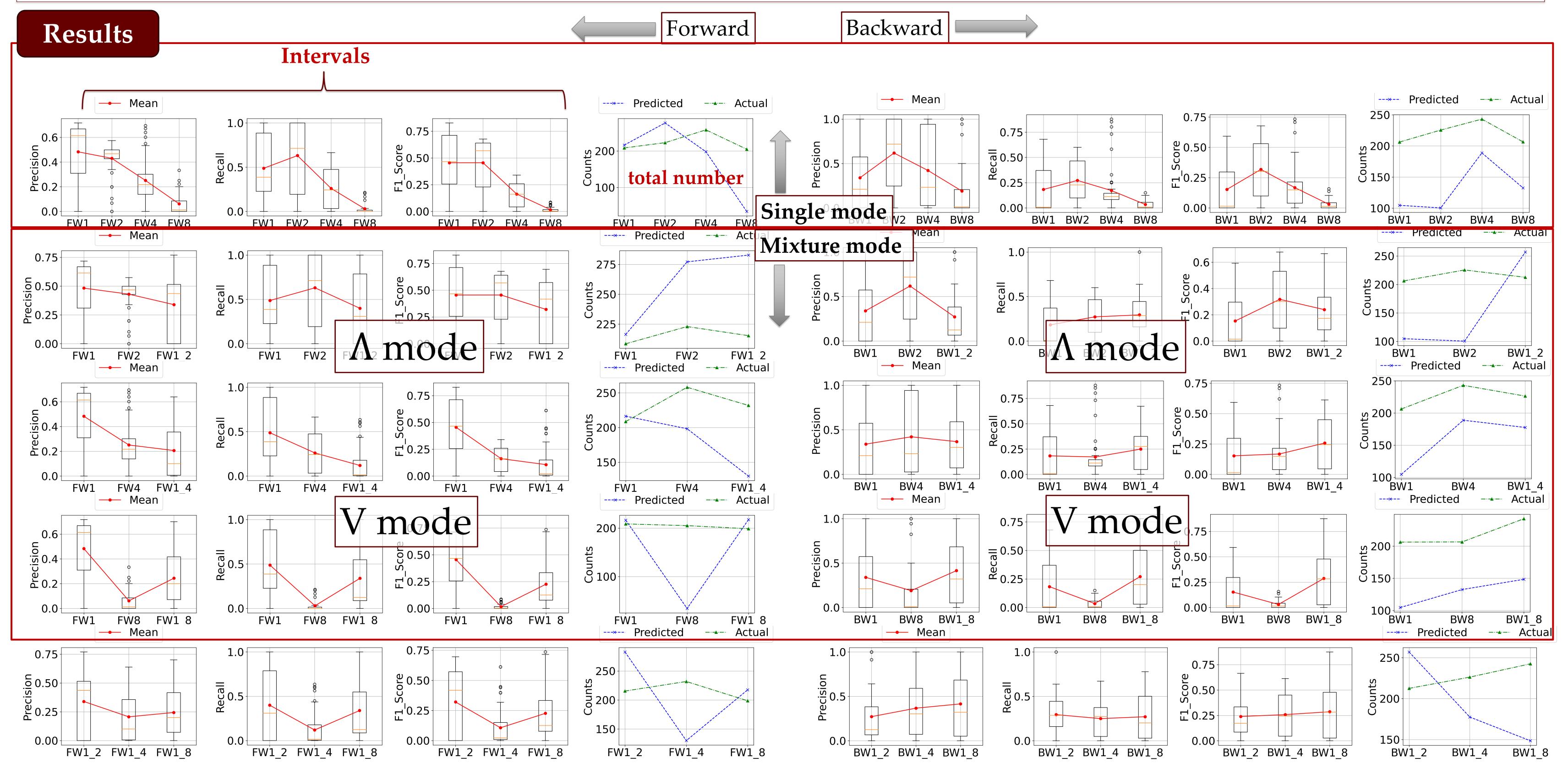
Flow Tasks	Shared Input	Variable Input	Output
Forward	graph $G$ , diffusion model $D$ , observation intervals $I$	source nodes $\Omega$	target nodes $\omega$
Backward	graph $G$ , diffusion model $D$ , observation intervals $I$	target nodes $\omega$	source nodes $\Omega$

Graph: Watts-Strogatz small-world graphs Diffusion: SIR

**Implementation:** https://github.com/XGraph-Team/XFlow

LLM: ChatGPT-4, GUI version, 10 results for each test

**Metrics:** count the distance to real results; count the total number



**Future Plan** 

- Long dependency is still limited with text input, even with short intervals.
- Single-mode performance drops down significantly as the interval increases.
- Mixture-model: from mixing interval 1 with 2/4/8: from V to flat and then to  $\Lambda$  ${\color{black}\bullet}$

#### Text mode: GPT API; Other LLMs;

Professional mode: graph format & Customized Transformer