

# Auto-Intent: Automated Intent Discovery and Self-Exploration for Large Language Model Web Agents



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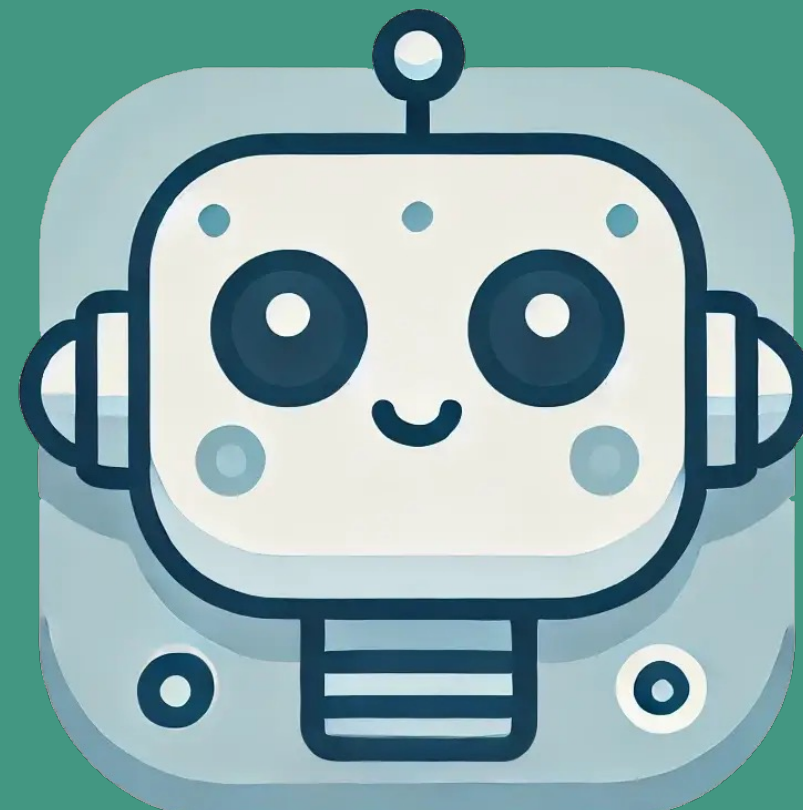
## Auto-Intent: Data-driven, efficient approach to turning pre-trained LLMs into target domain decision-making agents

### Pre-trained LLM Agent



- Insufficient domain knowledge
- Few-shot examples give limited info

### Auto-Intent



Pre-trained LLM  

Small, domain-fine-tuned intent predictor



Top-k predicted "intents" as hint

1. Specifying job type

2. Specifying salary range

3. Opening salary filter

⋮

### Small, domain-fine-tuned agent



- Weak reasoning
- Weak generalization capabilities



Decision-making tasks in target domains

### Intents $z_t$

Very concise natural language phrases (2-3 words)

- Natural language → well generalizable
- Concise → better intent space exploration  
→ easier learning with small LMs

### Intent Discovery $z_t = \mathcal{M}_{\text{extract}}(o_t, a_t, z_{1:t-1})$

Location Boston, NY	Date Thu, Mar 16	Time Now
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#### Observation $O_t$ :

Task: Check pickup restaurant available in Boston, NY on March 18, 5pm with one guest

Web page: `<html> ... <div> <input id=3 text date thu, mar 16 /> <button id=4 button date, selected value is thu,> <svg id=5 /> </button> </div> ... </html>`

Previous intents  $z_{1:t-1}$ : (1) selecting service type, (2) selecting location, ..., (4) finalizing location

#### Action $a_t$ :

CLICK  
`<svg id=5 />`

Intent Extractor (LLM)  
 $\mathcal{M}_{\text{extract}}$

Intent  $z_t$ :  
opening date picker

### Intent Predictor $\mathcal{M}_{\text{intent}}$

is trained to predict intents discovered for demo data.  
For inference, it predicts top-k intents via beam search:

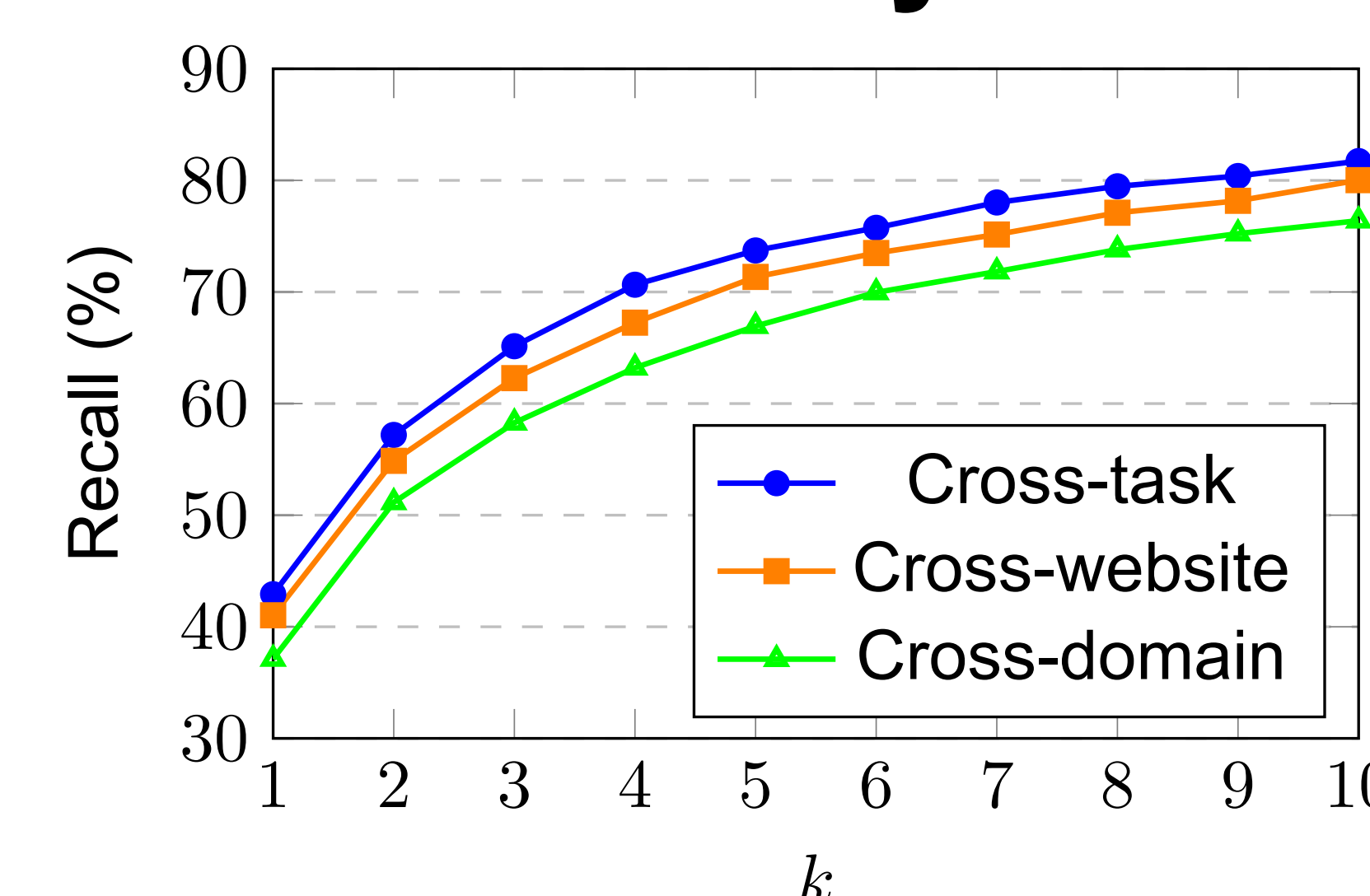
$$\hat{z}_t^1, \dots, \hat{z}_t^k \sim \mathcal{M}_{\text{intent}}(o_t, a_{1:t-1}, z_{1:t-1})$$

### Mind2Web Experiments

Method	Cross-task		Cross-website		Cross-domain	
	Elem. acc	Step SR	Elem. acc	Step SR	Elem. acc	Step SR
SeeAct (GPT-4V)	46.4	40.2	38.0	32.4	42.4	36.8
MindAct (Mistral-7B)	53.7	50.1	41.7	38.1	43.5	40.3
ICL (GPT-4)	46.9	41.7	45.0	40.0	45.3	41.3
+ Ours (Mistral-7B)	53.3	47.3	49.3	42.0	48.8	44.1
ICL (Llama-405B)	50.4	43.6	46.8	39.9	47.1	41.6
+ Ours (Mistral-7B)	56.3	50.4	51.1	43.6	49.5	44.6

### Generalization Results and Analysis

Method	Task SR
ICL (GPT-4)	19.0%
+ Ours (Mistral-7B)	23.8%
ICL (Llama-405B)	14.3%
+ Ours (Mistral-7B)	19.0%



Cross-benchmark  
generalization from  
Mind2Web to WebArena

Intent recall increases  
as k increases

Please check out our paper for details & more results!