

Beyond Task-Oriented and Chitchat Dialogues : Proactive and Transition-Aware Conversational Agents

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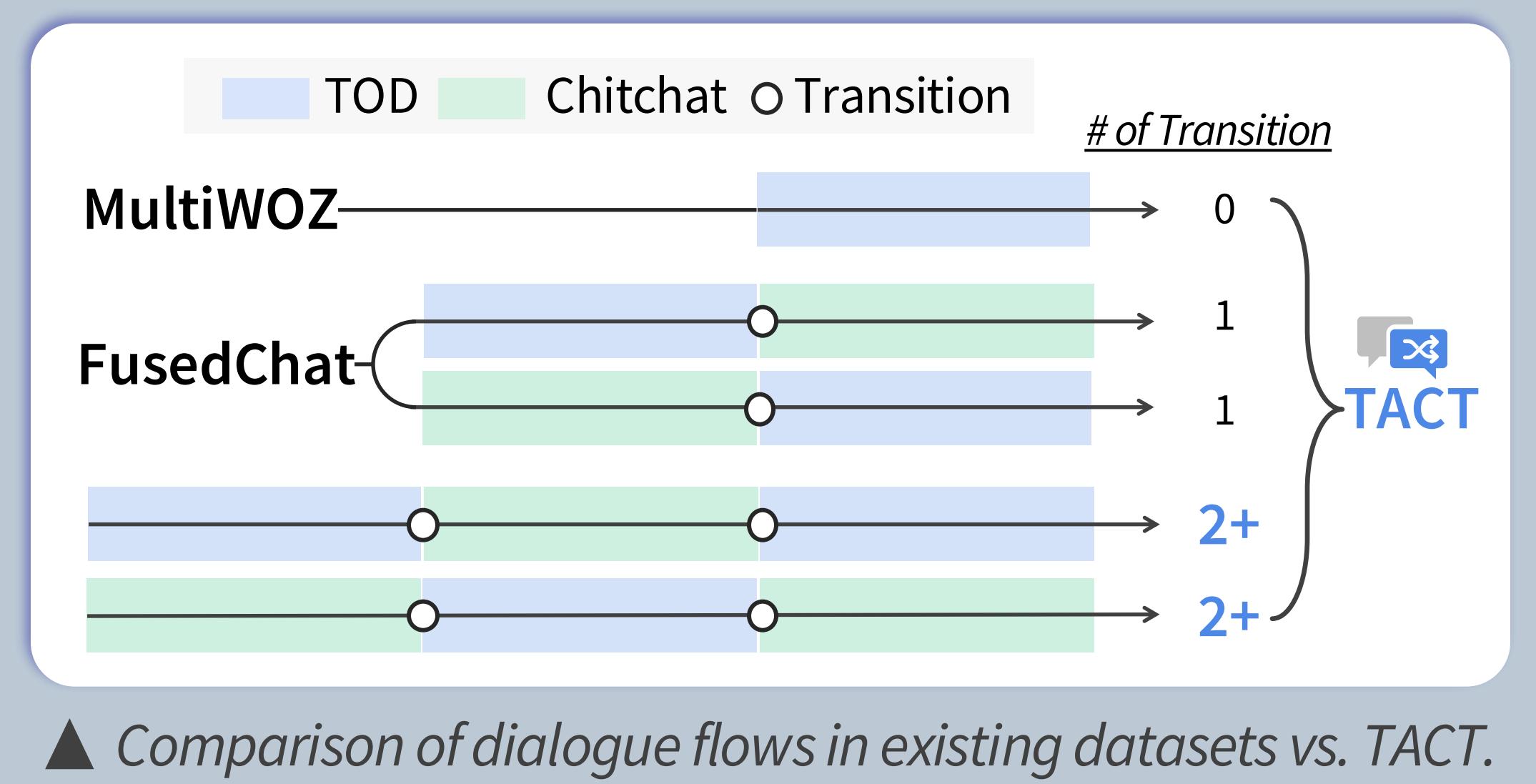
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Motivation

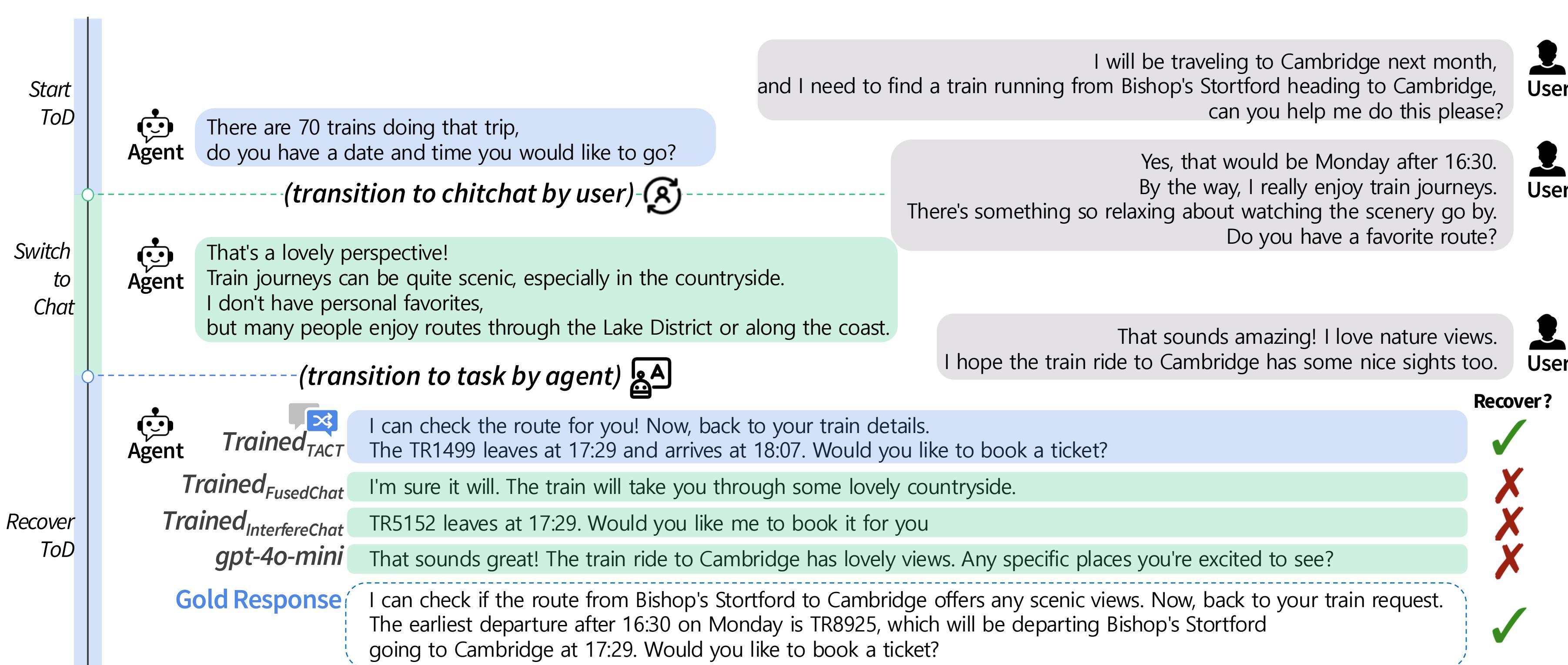
- Problem:** Existing attempts to integrate Task-Oriented Dialogue (TOD) and open-domain chitchat suffer from **limited transition diversity** and lack support for **multi-turn, recoverable transitions**.
- Observation:** Real-world conversations involve **frequent and complex mode switches** (e.g., TOD → Chitchat → TOD) that require both **transition-awareness** and **proactivity**.
- Our Contribution:** Introduce **TACT** (TOD-And-Chitchat Transition) — a benchmark for training conversational agents that can handle **natural mode transitions** and recover suspended tasks.



💡 TACT Dataset

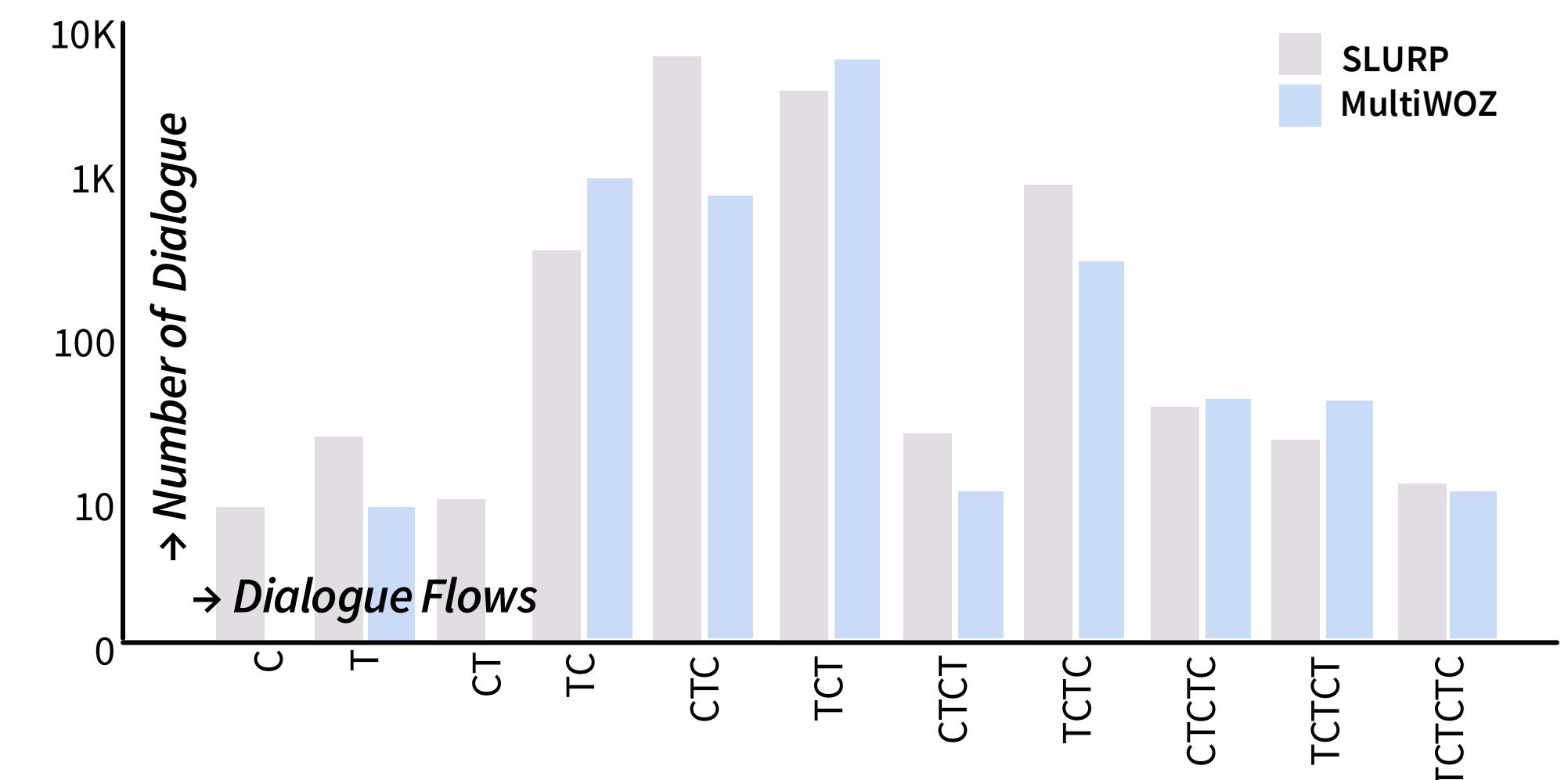
Designed to equip conversational agents with the ability to handle natural, recoverable mode transitions between TOD and chitchat.

- Dialogue Construction:**
 - Diverse flow patterns: TCT, CTC, TCTCT, etc., covering both short- and long-horizon transitions.
 - Balanced initiators: Includes both user- and agent-driven mode switches.
 - Recoverable structures: Supports returning to suspended modes, enabling multi-turn coherence.



Dataset	SalesBot2.0	FusedChat	InterfereChat	TACT	
Seed	SalesBot1.0	MultiWOZ2.4	FusedChat	MultiWOZ2.2	SLURP
# Intents	6	11	11	11	50*
# Dialog	5,453	10,436	4,475	7,199	9,936
# Avg. Turn	7.71	18.36	13.58	15.04	16.42
# Avg. Switch	0.96	1	0*	1.93	2.06
# Avg. Recov.	-	0	0*	0.93	1.07
# Uniq. Flow	2	2	1	11	12
Flow Types	CT	TC, CT	T*	TCT, CTC, TCTCT, etc.	

▲ Statistics of existing datasets and proposed TACT.



- Dialogue Validation:** Ensure large-scale data quality while keeping dialogues natural.
 - Criteria: (1) **Intent accuracy** (correct task interpretation), (2) **Transition quality** (contextually justified switches), (3) **Dialogue naturalness** (fluency & coherence)
 - Method: Hybrid validation — human-authored criteria + LLM reasoning

→ Improves intent consistency and transition realism across the dataset

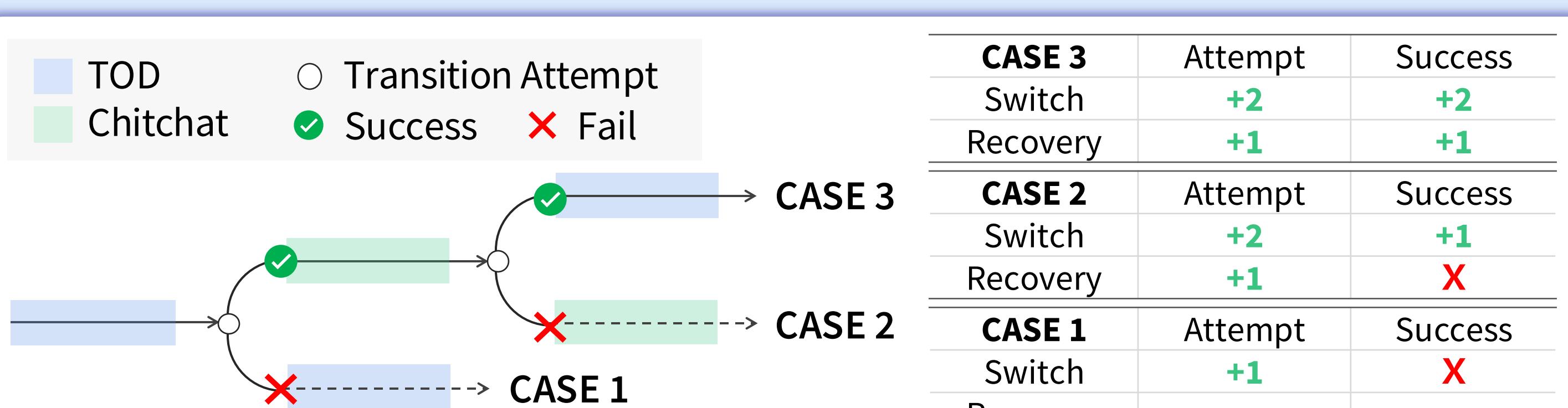
Validation Approaches	Human-Annotated Criteria	LLM-Generated Task Description	LLM-Generated Evaluation Steps
Active Critic	✗	✓	✗
G-Eval	✓	✗	✓
Ours	✓	✓	✓

▲ Comparison of data validation methods.

>We've developed **TACT**, a dataset built to evaluate and advance transition-aware dialogue systems, reflecting the diversity and naturalness of real conversations, and rigorously validated to ensure intent accuracy, transition quality, and coherence.

Methodology & Metrics

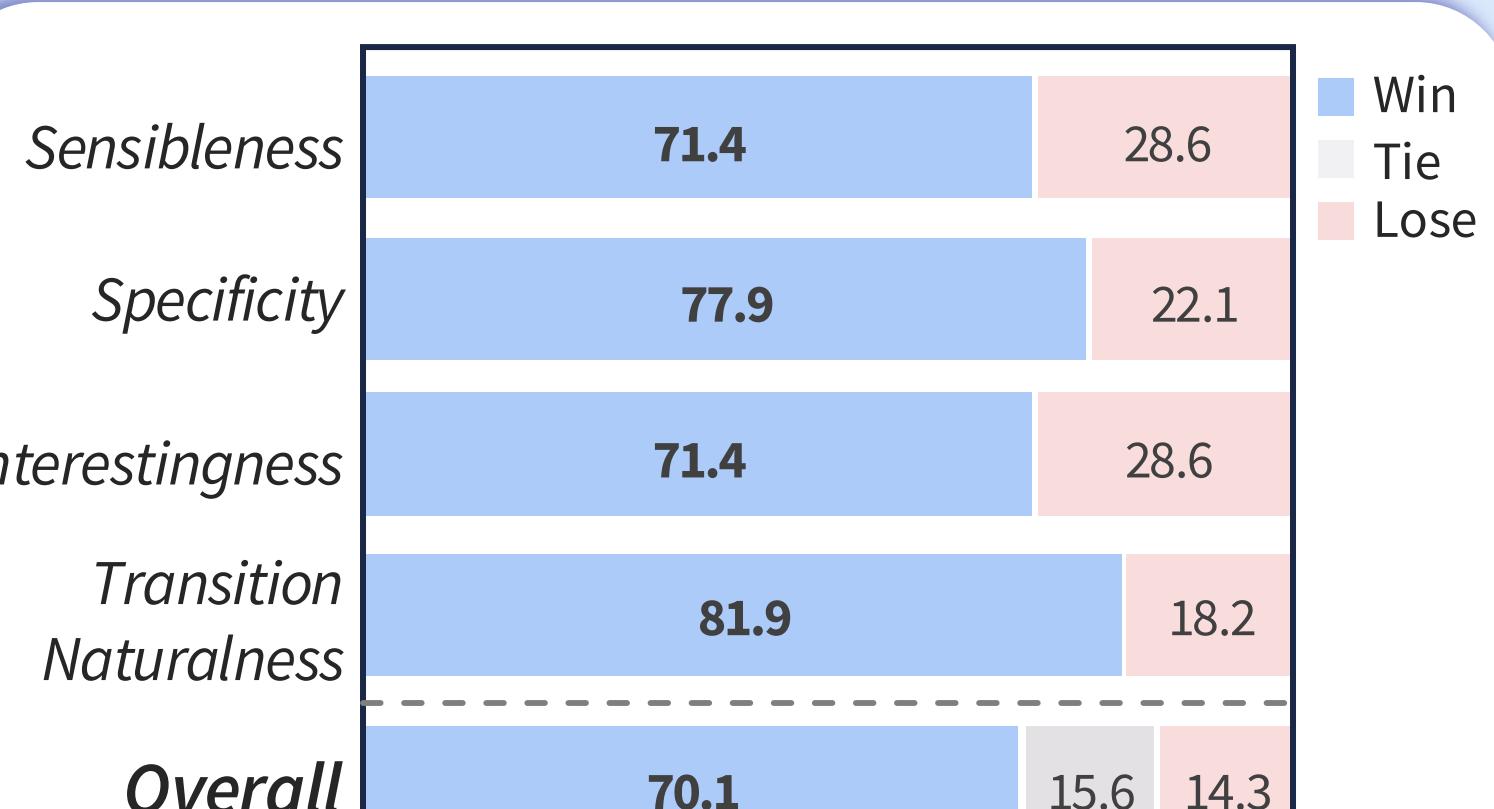
- Model & Training**
 - Base:** FnCTOD — a unified dialogue model that integrates task execution and chitchat through structured function calls
 - Enhancement:** Finetuned on TACT with DPO to improve response quality, flow control, and mode-awareness
- Evaluation**
 - TOD-centric:** Mode Selection, Intent Detection, Joint Accuracy
 - Chitchat quality:** SSI Sensibleness, Specificity, Interestingness, Transition Naturalness
 - Transition-aware:** **Switch & Recovery** (measured by attempts & successes)



Key Results

Method	TOD		Flow			Chitchat			
	Mode Selection Acc.	F1-score	Intent Detection Acc./turn	Acc./dialogue	Switch Attempt	Success	Recovery Attempt	Success	Overall Win-Rate
ICL-ZS	90.46	86.21	87.57	50.44	0.879	0.374	0.880	0.099	-
ICL-FS	91.45	88.98	84.09	40.00	1.577	0.865	1.571	0.652	-
SFT	98.95	98.50	96.35	80.94	1.322	1.300	0.977	0.856	23.16
SFT-DPO	98.82	98.32	96.03	80.00	1.343	1.322	0.977	0.859	40.86
Pipeline	98.95	98.50	96.35	80.94	1.322	1.300	0.977	0.856	24.32

- TACT-trained:** Only models achieving non-zero transition success in multi-turn settings
- DPO Gains:** +40.86% win rate, +33.7% in transition naturalness vs. GPT-4o
- Flow Sensitivity:** Strongest on TCT; CTC/TC remain challenging



* Insights: Preference tuning yields smoother, contextually grounded transitions