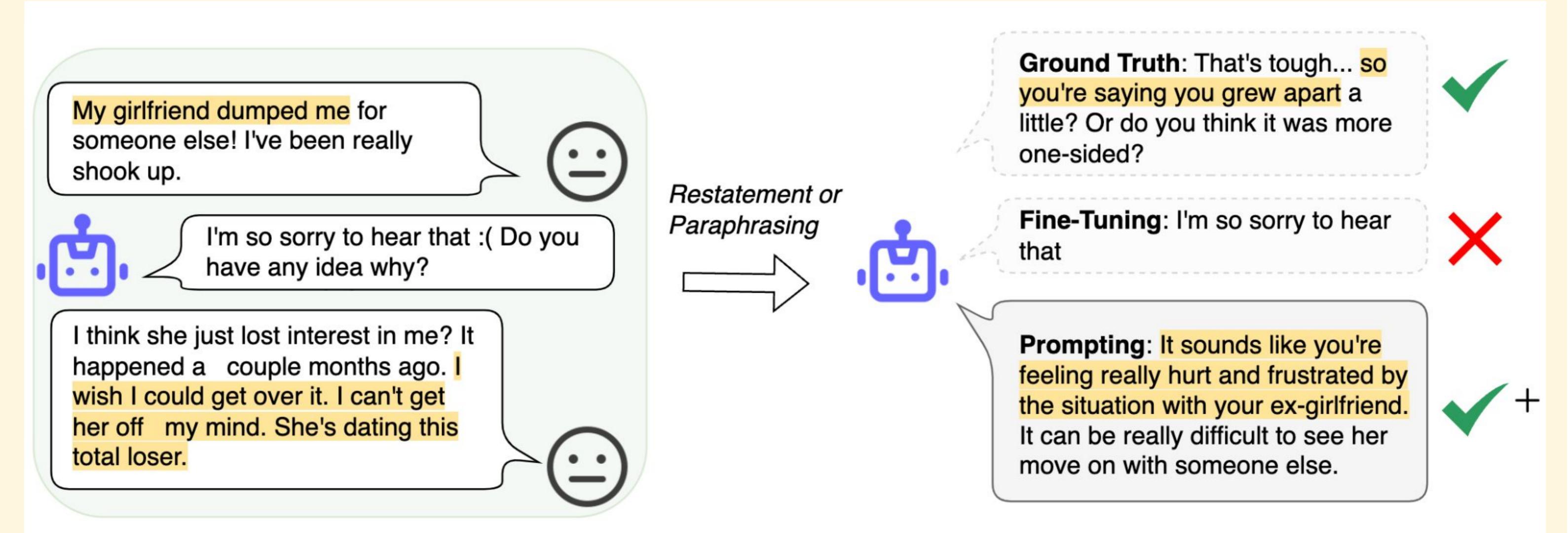
Controllable Mixed-Initiative Dialogue Generation Through Prompting

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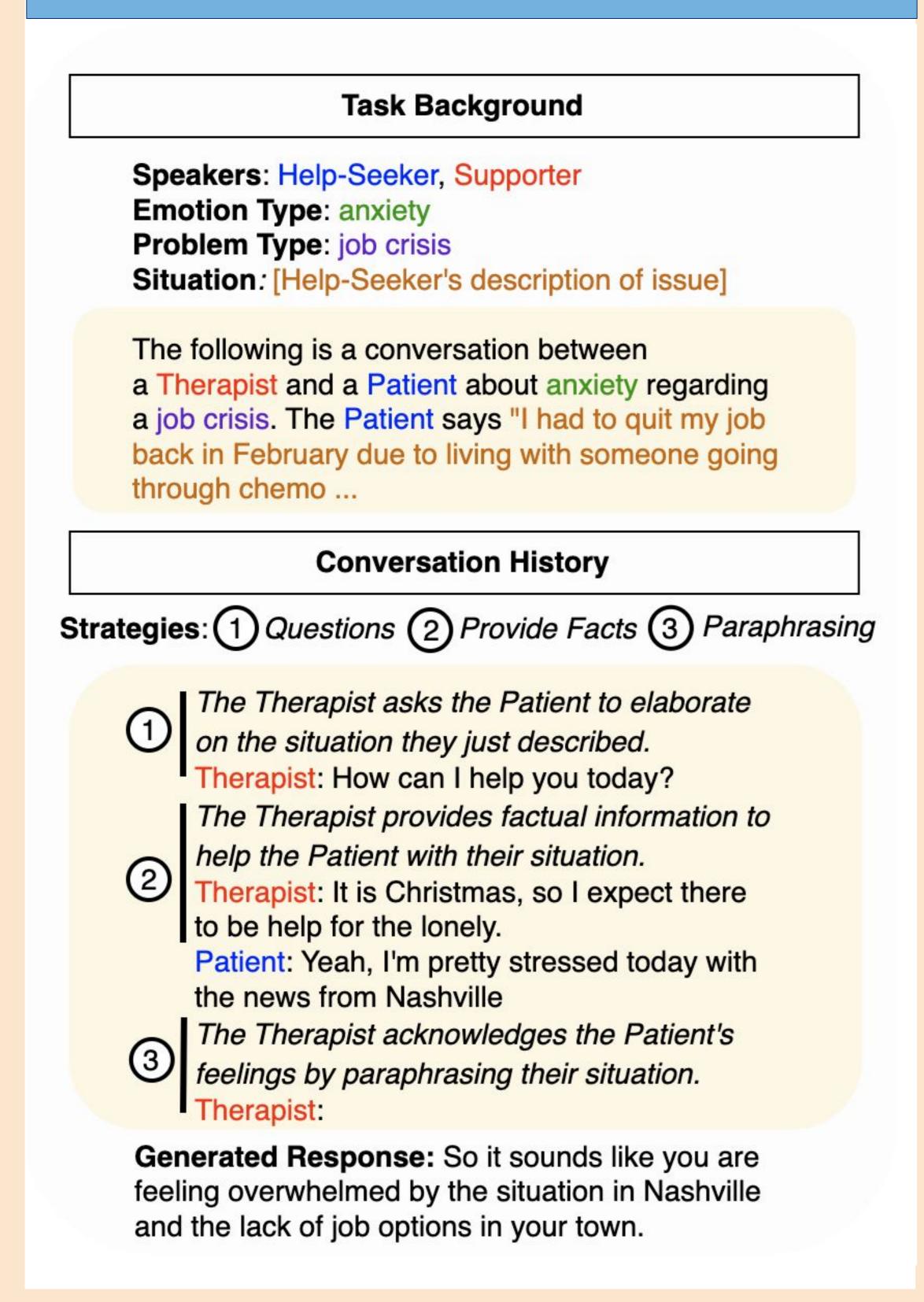


Replace Smaller Fine-Tuned Conditional Generation Models with Prompting



- **Dialogue policy planning is hard, but that's only half the battle.** There's still no guarantee that the planned conversational strategy will actually get executed! It depends on the quality of a conditional generation model.
- Fine-tuning conditional dialogue generation models is also really hard. There's often tradeoff between learning to generate coherent dialogue responses and learning strong semantic control.
- Fine-tuning on crowdsourced data is a bottleneck by definition. Many complex dialogue tasks require expert-level domain knowledge (e.g., persuasive strategies, emotional support) which results in various annotation errors that can get propagated through the training process.
- **Is few-shot prompting the solution?** It doesn't depend on supervised data, and the capacity of LLMs is much greater than prior fine-tuned models. This can mean strong semantic control and coherence.

Example Prompt Construction



Prompting versus Fine-tuning and Ground Truth

Corpus	Metric	FT	GT	Prompt
ESC	Accuracy	0.81	0.85	0.88*
	Coherence	3.57	3.57	3.72
	Consistency	3.63	3.60	3.80^{+*}
	Engagingness	3.55	3.61	3.81 ^{+*}
	Distinct-3	0.89	0.90	0.90
	Distinct-4	0.87	0.90^{*}	0.91^{+*}
	QuantiDCE	3.25	3.03	3.19
	Win Rates			
	v. FT		0.56	0.52
	v. GT	0.44		0.64*
	v. Prompt	0.48	0.36	
P4G	Accuracy	0.88	0.83	0.89
	Coherence	3.66	3.58	3.83^{+*}
	Consistency	3.69	3.56	3.71^{+}
	Engagingness	3.62	3.52	3.69^{+}
	Distinct-3	0.87	0.88	0.89
	Distinct-4	0.88	0.88	0.88
	QuantiDCE	3.16	3.09	3.24^{+}
	Win Rates			
	v. FT		0.56	0.59*
	v. GT	0.48		0.55
	v. Prompt	0.41	0.45	

Prompting
 InstructGPT
 consistently
 outperforms
 fine-tuning and the
 ground truth
 responses on both
 Persuasion4Good
 and Emotional
 Support
 Conversations

