

# Towards Unified Conversational Recommender Systems via Knowledge-Enhanced Prompt Learning

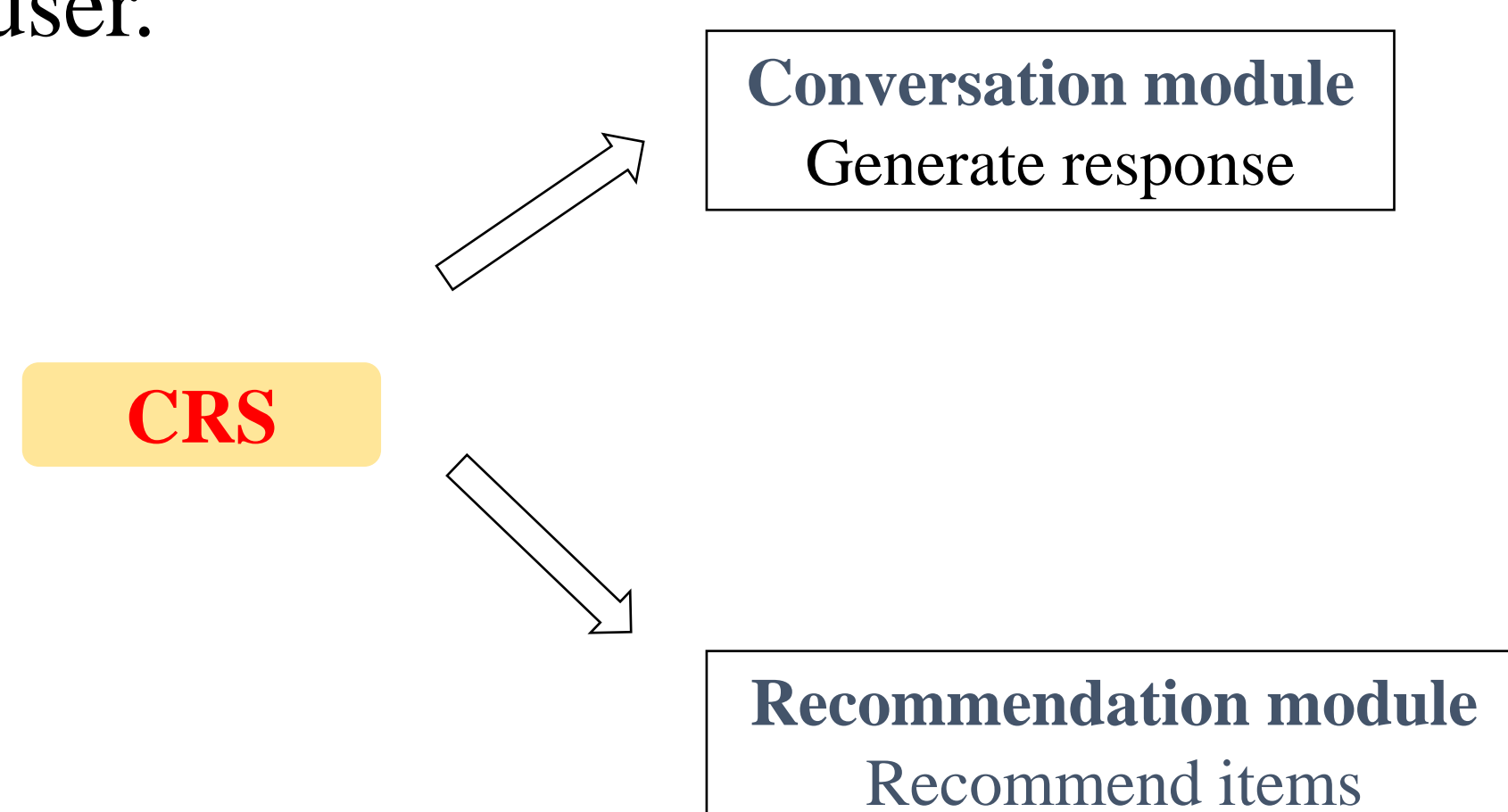
Xiaolei Wang\*, Kun Zhou\*, Ji-Rong Wen, Wayne Xin Zhao  
Renmin University of China

## Background

### Conversational recommender system (CRS)

A system aims to provide high-quality recommendations to users via natural language conversations. Typically, it needs to solve two tasks:

- **Recommendation:** predict user-preferred items;
- **Conversation:** generate a proper response for conversing with the user.



### Prompt Learning

Most of PLMs are pre-trained with the objective of language modeling but are fine-tuned on downstream tasks with quite different objectives. To overcome the gap between pre-training and fine-tuning, prompt learning has been proposed, which relies on carefully designed prompts to reformulate the downstream tasks as the pre-training task.

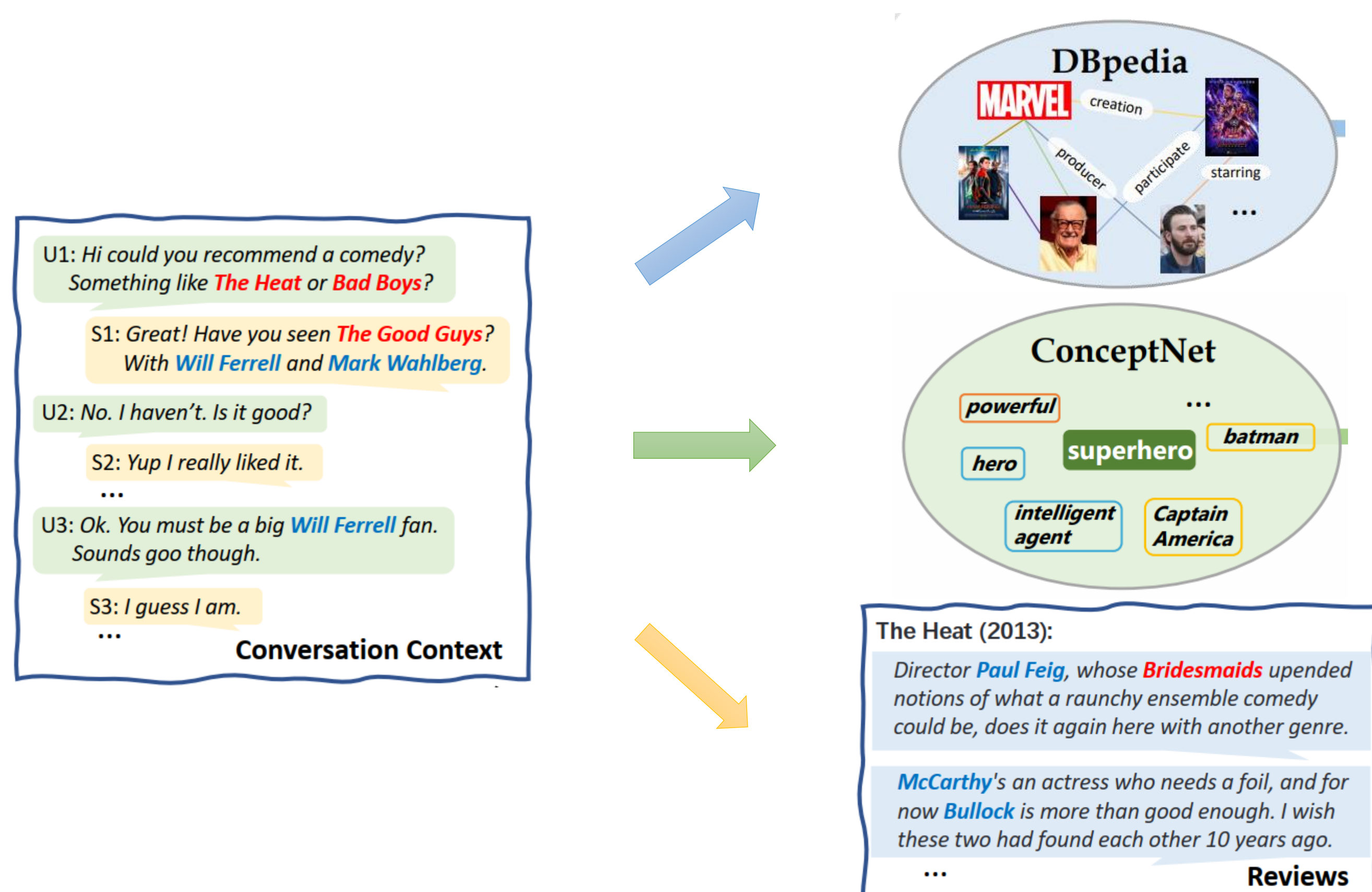
- **Discrete Prompt:** actual text string
- **Continuous Prompt:** described directly in the embedding space of the underlying PLM

## Motivation

### Existing works

A capable CRS should be able to **seamlessly integrate the recommendation module and the conversation module**, because they are highly coupled.

- One line of work incorporates shared knowledge resources and their representations (e.g., knowledge graphs and reviews)



- Another line of work designs special representation alignment strategies, such as pre-training tasks and regularization terms (e.g., mutual information maximization and contrastive learning)

The fundamental issue of **semantic inconsistency** between the recommendation and conversation modules has not been well addressed.

As shown in the following figure, although the recommendation module predicts the movie “Frozen 2 (2019)”, the conversation module generates a mismatched response that contains another movie “Pretty Woman (1990)”.

USER:	Hello! I am looking for some movies.
HUMAN:	What kinds of movie do you like? I like <b>animated</b> movies such as <i>Frozen (2013)</i> .
USER:	I do not like <b>animated</b> films. I would love to see a movie like <i>Pretty Woman (1990)</i> starring <b>Julia Roberts</b> . Know any that are similar?
KG5F:	<b>Recommendation:</b> Frozen 2 (2019) <b>Response:</b> <i>Pretty Woman (1990)</i> is a great movie.
OURS:	<b>Recommendation:</b> My Best Friend's Wedding (1997) <b>Response:</b> Have you seen <i>My Best Friend's Wedding (1997)</i> ? <b>Julia Roberts</b> also stars in it.
HUMAN:	<i>Pretty Woman (1990)</i> was a good one. If you are in it for <b>Julia Roberts</b> you can try <i>Runaway Bride (1999)</i> .

Two Reasons:

- Most of these methods develop the two modules with different architectures or techniques;
- Results from one module can-not be perceived and utilized by the other.

## Approach

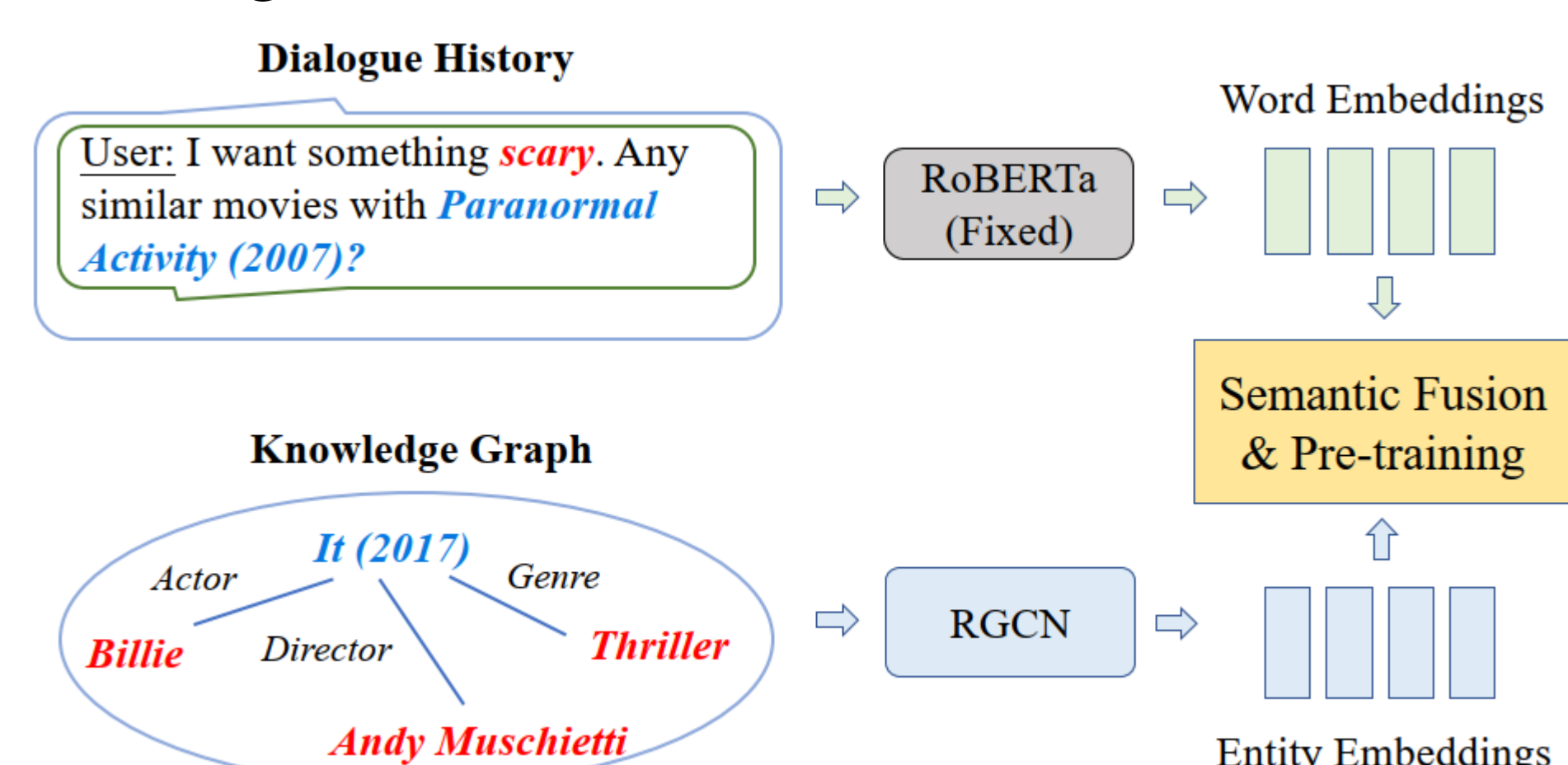
### Overview

Base PLM: DialoGPT (pre-trained with dialogue corpus)

Prompt-augmented Dialogue Context:  $\tilde{C} \rightarrow \underbrace{p_1, \dots, p_{n_p}}_{\text{prompt tokens}}, \underbrace{w_1 \dots w_{n_w}}_{\text{word tokens}}$

### Semantic Fusion

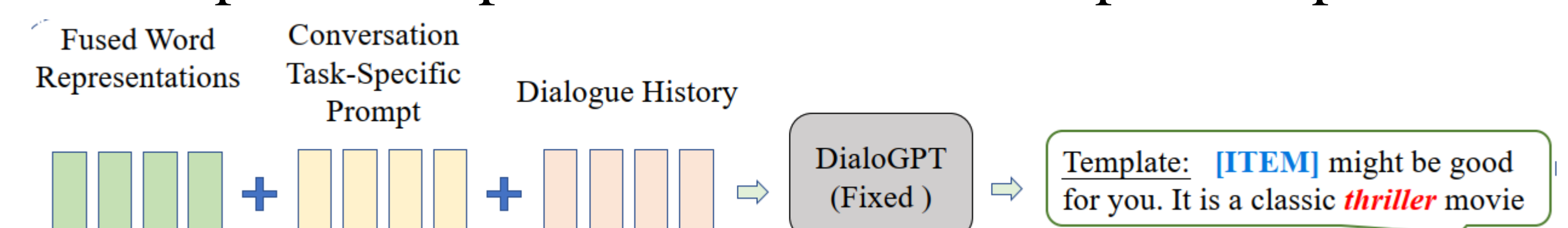
Due to pre-trained on general dialogue corpus, DialoGPT lacks domain knowledge needed for CRS.



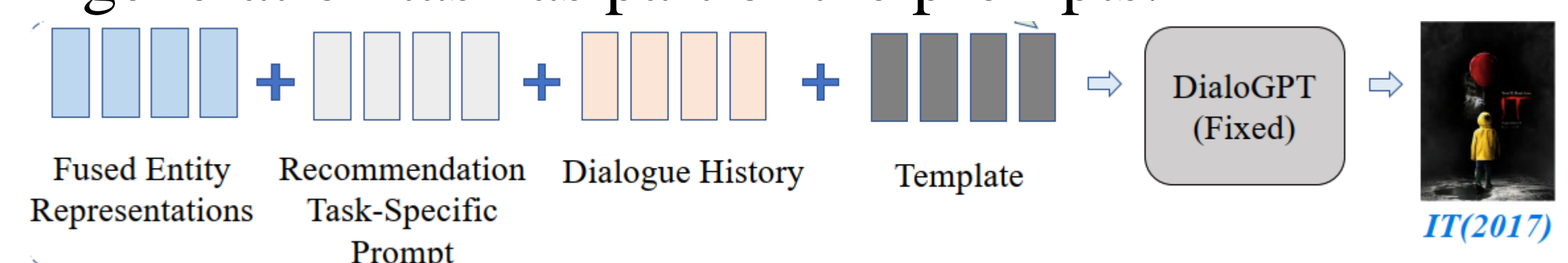
- Encoding: RoBERTa for word tokens, R-GCN for KG entities
- Semantic Fusion: Bi-attention
- Pre-Training: Self-supervised learning

### Subtask Prompt Design

- Response Generation  $\tilde{C}_{gen} \rightarrow [\tilde{T}; P_{gen}; C]$ 
  - To share intermediate results, DialoGPT generates the response template instead of the complete response.



- Item Recommendation  $\tilde{C}_{rec} \rightarrow [\tilde{E}; P_{rec}; C; S]$ 
  - Include the response templates from the response generation task as part of the prompts.



- Finally, the items will be filled into the template as complete responses.