

Introduction to Dialogue Systems



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dialogue systems

How can I help you?

I want to book a table at Da Marco

How many people?

Four people

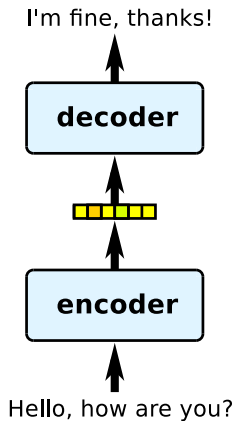
At what time?

- “chatbots”, “conversational agents”, ...
- computational systems that we can **interact** with
- general challenges:
 - representing **state**
 - interacting with **external resources**
- no discussion of speech technologies here! ASR/TTS at endpoints

types of dialogue systems

- **chit-chat / open-domain:**
 - mostly for entertainment value
 - some use in therapy, education
 - often unrestricted in terms of content
- **task-oriented:**
 - searching, booking, troubleshooting, ...
 - often more restricted/“controlled”

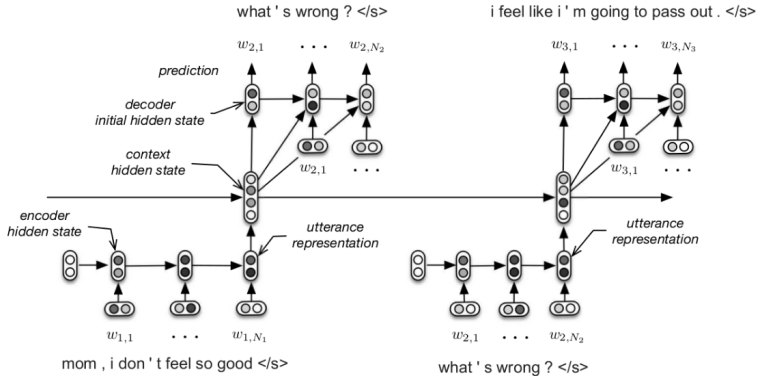
open-domain response generation: basic idea



- basic idea introduced by [Vinyals and Le \(2015\)](#)

how do we represent the **state** (the **history**) of the conversation?

architecture with turn representations



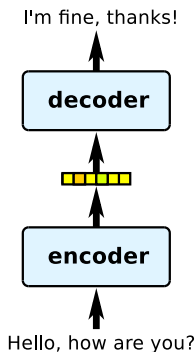
(Serban et al., 2016)

example: DIALOGPT

- the **DIALOGPT** model (Zhang et al., 2020) is a GPT-2 transformer tuned on Reddit conversations
- like the original GPT models, the response generation task is framed as **language modeling**: a Transformer decoder
- the dialogue history is used as a **prompt** (start of sequence)

examples: Meena and Blender Bot

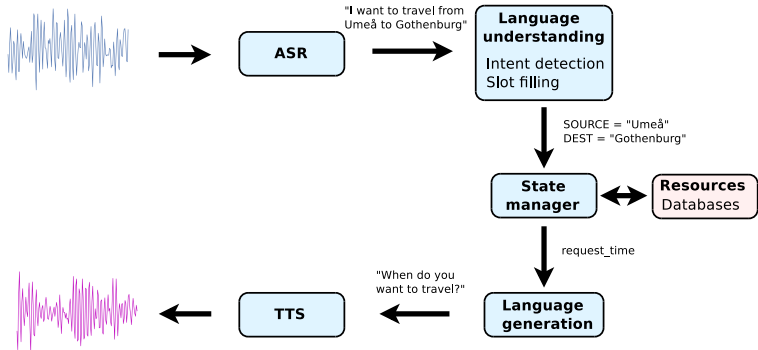
- Google's **Meena** (Adiwardana et al., 2020) and Facebook's **Blender Bot** (Roller et al., 2021) use encoder/decoder Transformers: the dialogue history is the input



open research problems in open-domain dialogue

- **decoding** from the language model: avoiding bland answers
- maintaining **consistency over a long conversation**: “goldfish memory” (Xu et al., 2021)
- **consistency with the outside world** (Komeili et al., 2021)
- **avoiding toxicity** in generated outputs
- Huang et al. (2020) gives an overview of challenges

traditional task-oriented dialogue systems

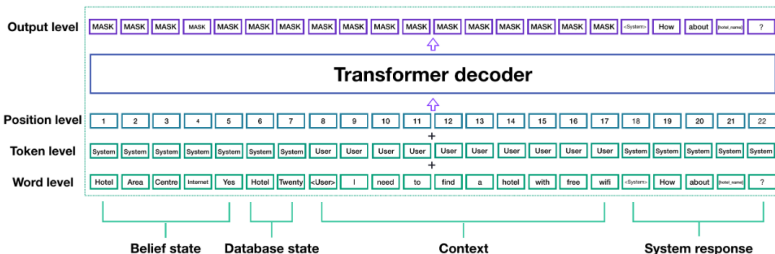


task-oriented dialogue: challenges

- domain-specific design requires **modularity**
 - research has often focused on a module and less on end-to-end performance
 - pipeline design \Rightarrow risk of error propagation
- need to interact with **external resources**
- often highly specialized, training data often scarce

adapting pre-trained models for task-oriented dialogue

- Budzianowski and Vulic (2019) adapt GPT-2 for task-oriented response generation



- **TOD-BERT** (Wu et al., 2020) is a BERT-like model that could be used for several steps in task-oriented dialogue

dialogue datasets

- Reddit, Twitter, ...
- subtitles (<http://opus.nlpl.eu/OpenSubtitles-v2018.php>)
- technical support e.g. Ubuntu (Vinyals and Le, 2015)
- ConvAI2 (Dinan et al., 2020) <http://convai.io/>
- Persona-Chat (Zhang et al., 2018)
<https://github.com/facebookresearch/ParlAI/tree/main/parlai/tasks/personachat>
- overview: *A Survey of Available Corpora for Building Data-Driven Dialogue Systems* (Serban et al., 2018)

evaluating dialogue systems

- **word**-based: overlap, perplexity, ...
- **human**-oriented evaluations: user satisfaction, fluency, ...
- there are conflicting opinions on how well they correlate:
 - [Dinan et al. \(2020\)](#) report weak correlations in the ConvAI2 challenge
 - [Adiwardana et al. \(2020\)](#) report good correlations between perplexity and human scores for the Meena chatbot

references

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