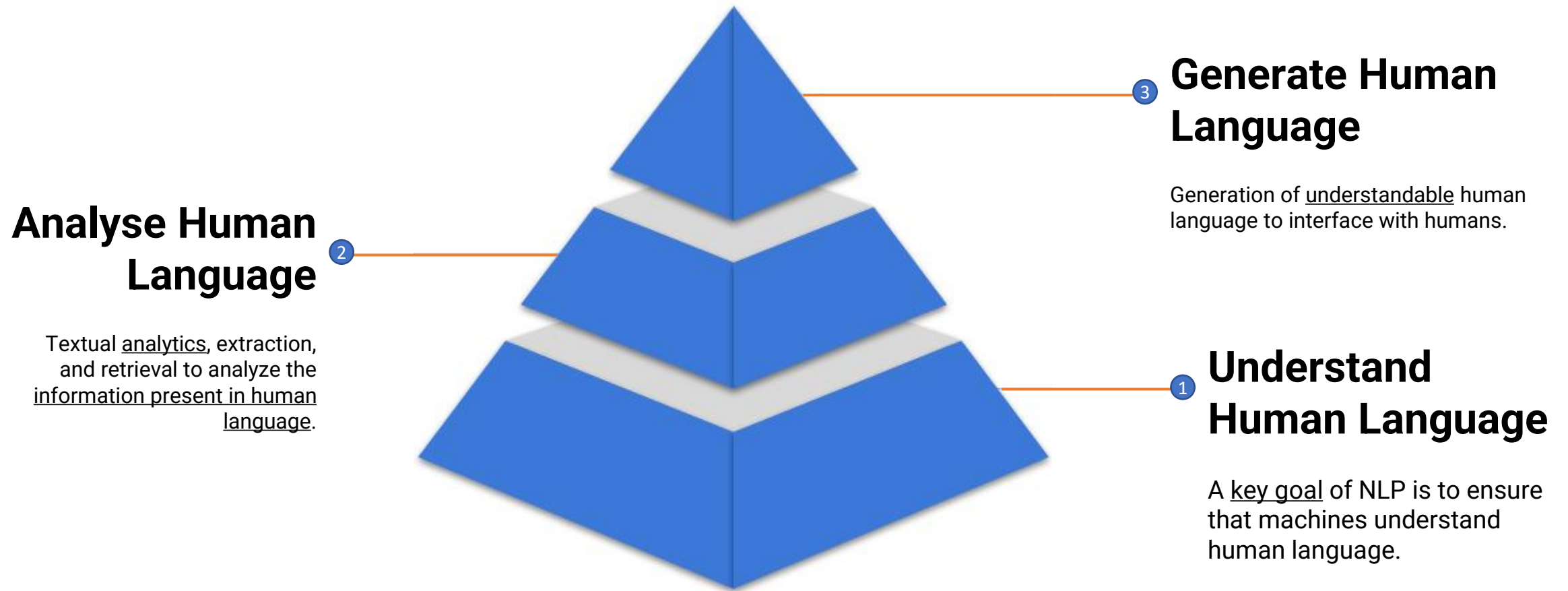


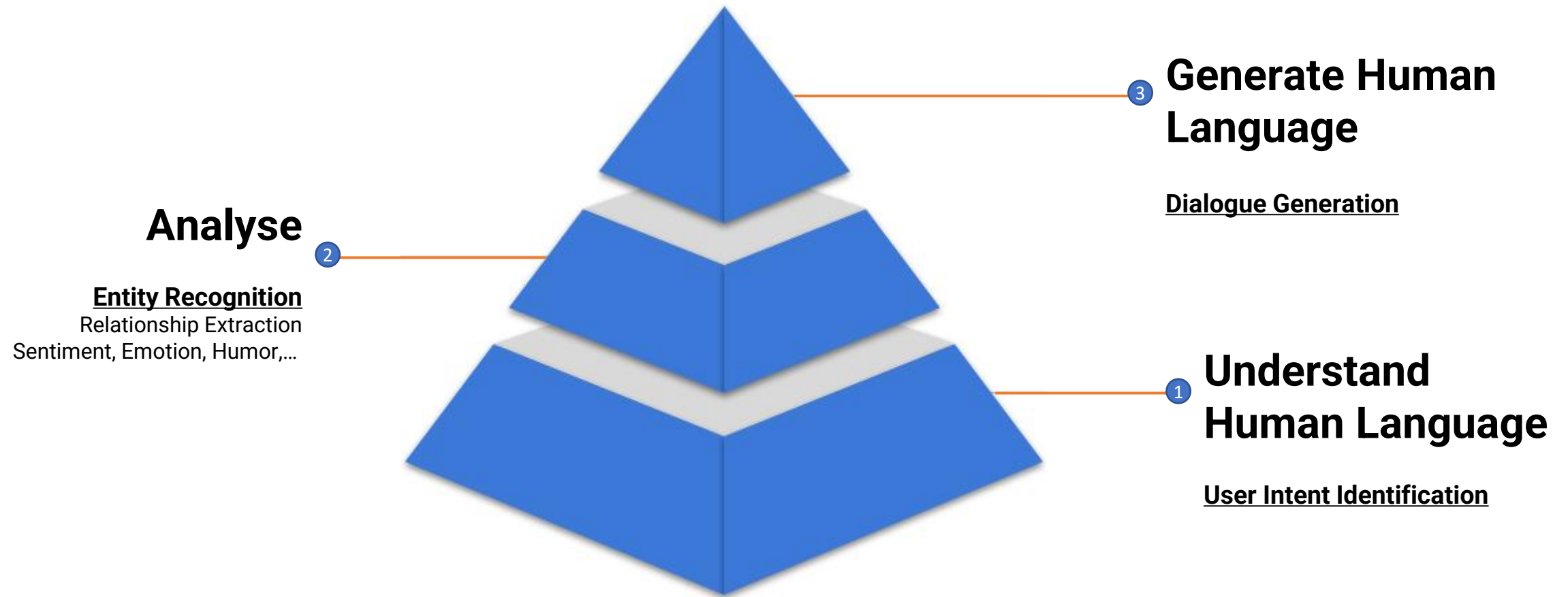
# **Dialogue Processing: The Role of NLP and its Building Blocks**

**Dr Diptesh Kanojia**  
**Surrey Institute for People-Centred AI**  
**Department of Computer Science, University of Surrey**

# Natural Language Processing (NLP)



# Natural Language Processing (NLP): A Dialogue Perspective



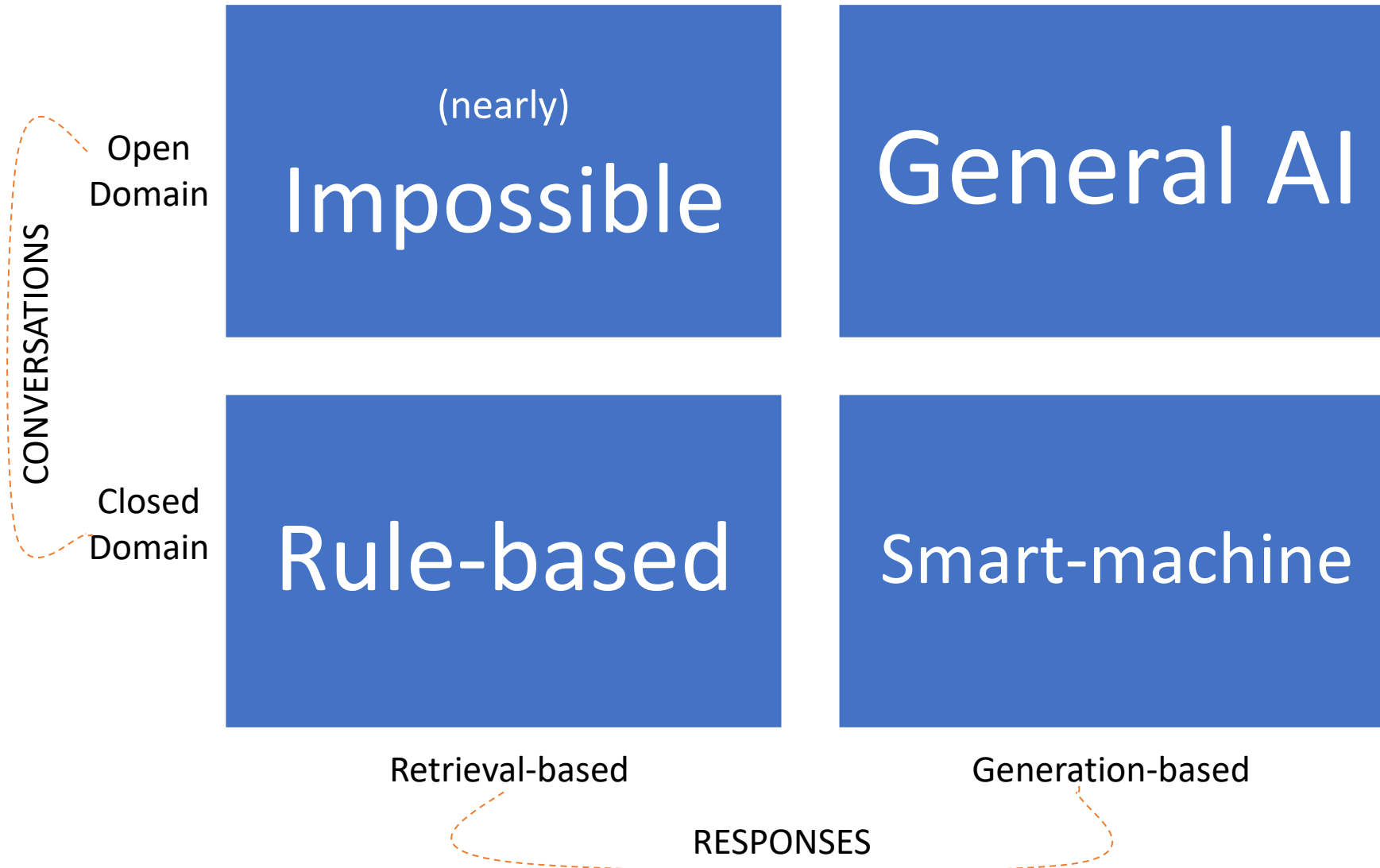
# Why Conversational AI?

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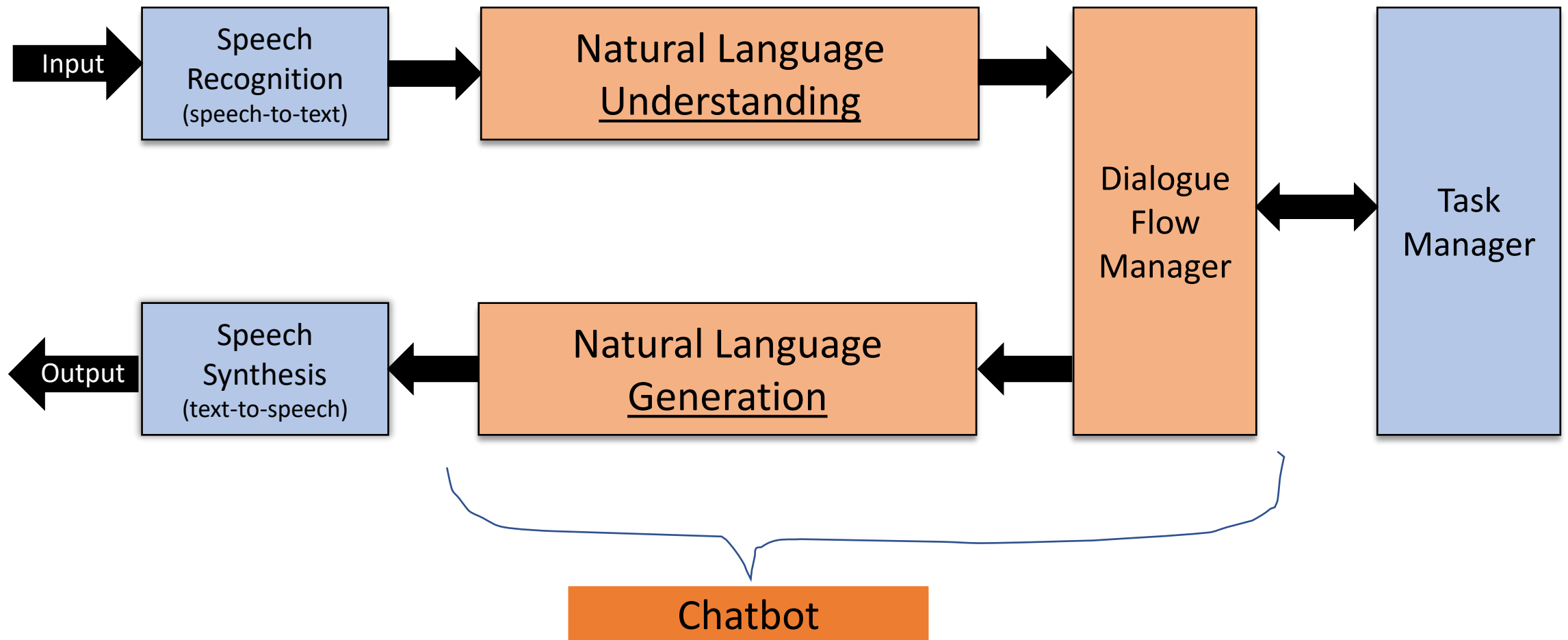
- Text Messaging is a popular form of interaction and chatbots are able to streamline interaction between people and 'services'.
- Chatbots are scalable
- Always available – less dependence of human resources.
- Helpful for organization in multiple geographies.



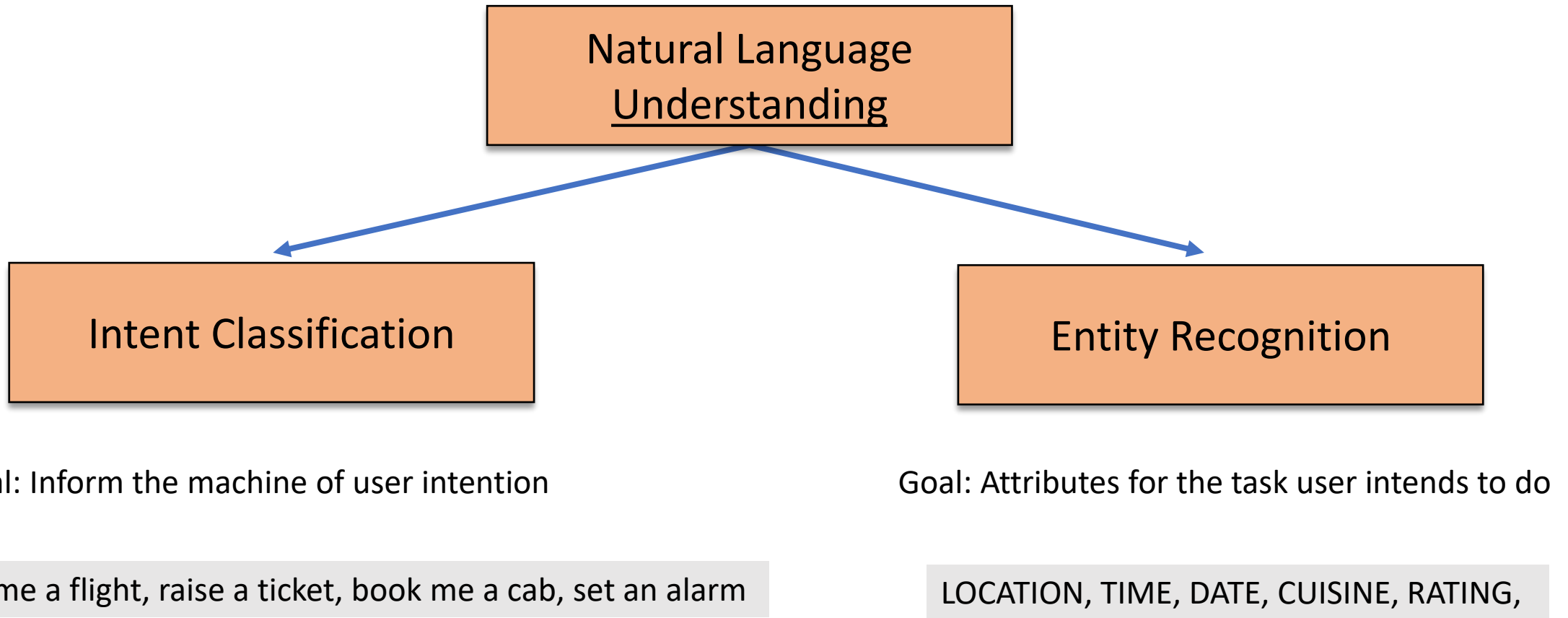
# Conversational Framework



# Dialogue System Architecture: Research Domains

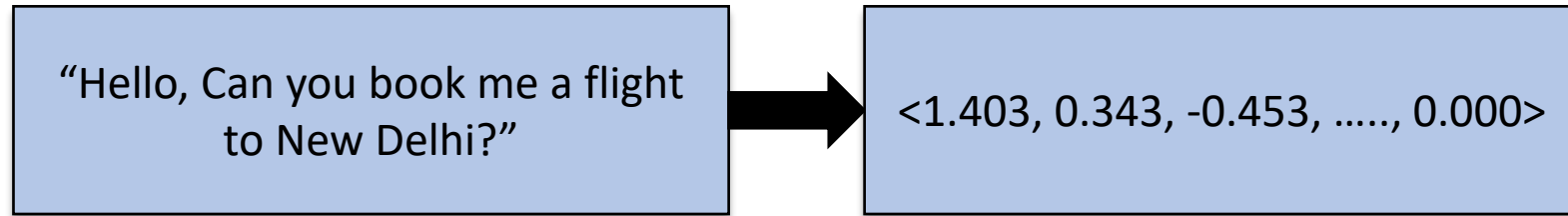


# NLU Engine Components



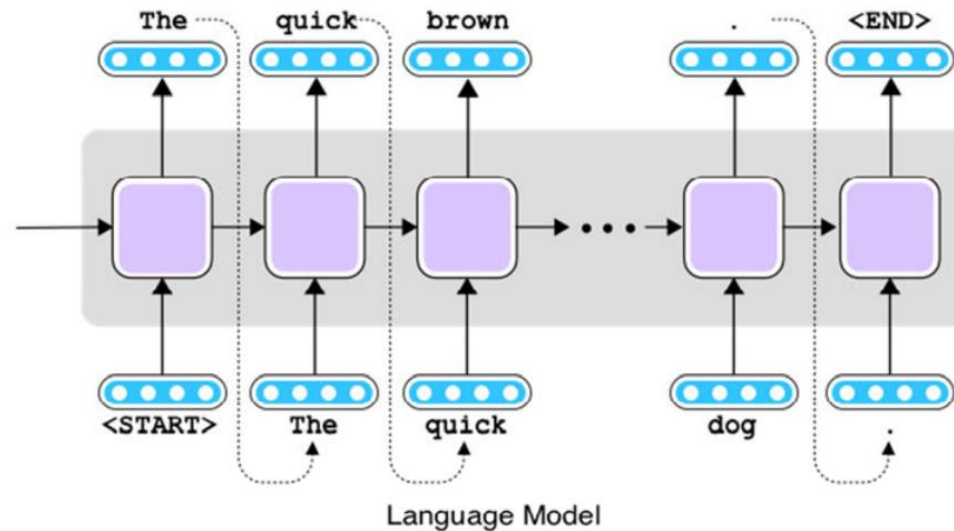
# Language Understanding

- Machine Understanding

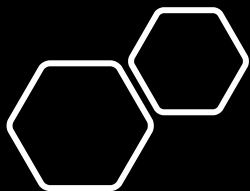


- Language Models

- Tokenization
- Vectorization

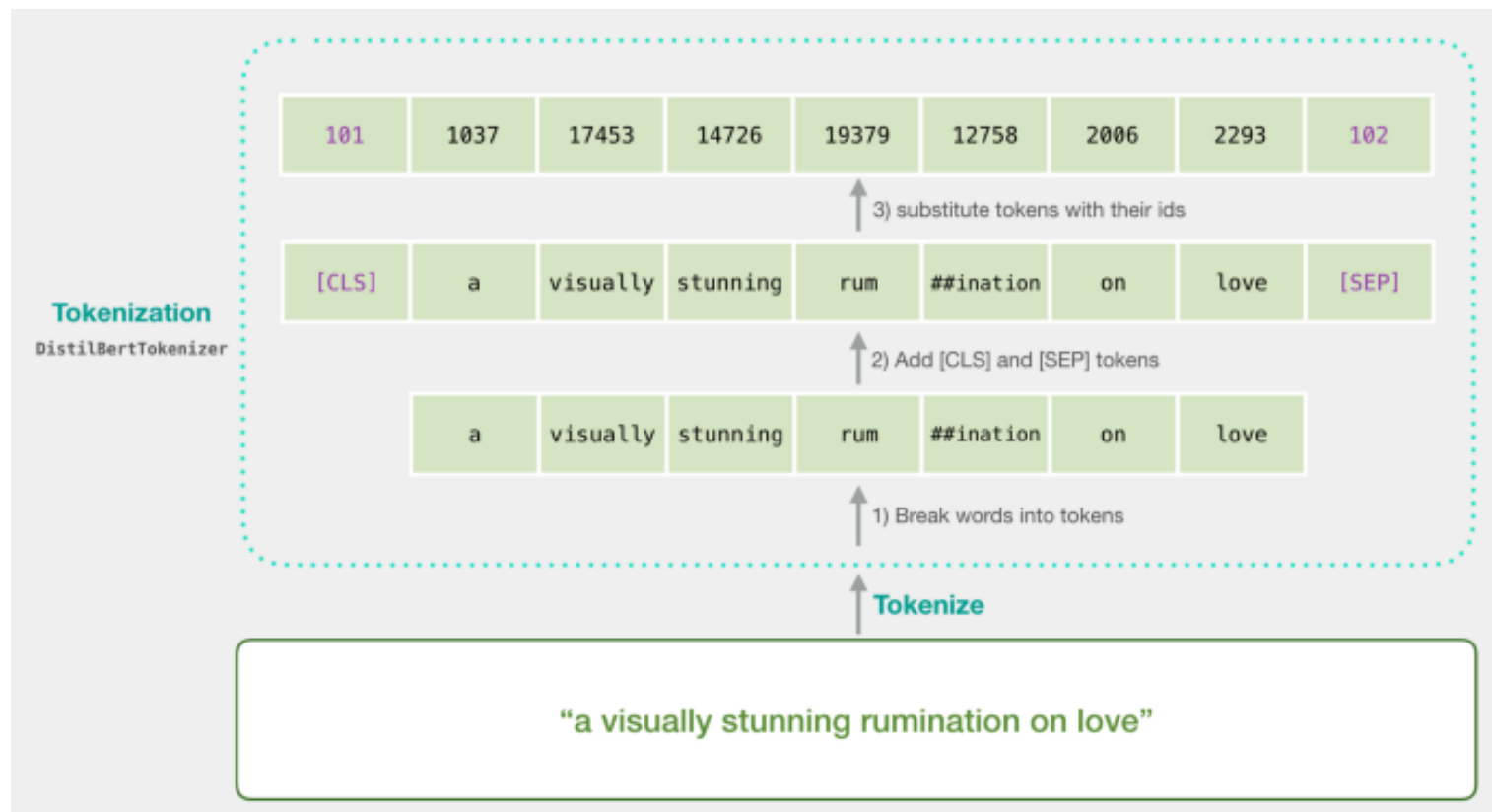






# Tokenization

- Allows the model to understand known tokens
- Breaks unknown tokens into sub-words
- Sub-words based on partial known words



# Word Embeddings: Vectorization

- Word embeddings are a powerful concept that can give your application a boost in the performance, but they cannot take context into account (words can have many meanings)
- Contextualised embeddings or simply contextualisation, is a way to transform the entire sentence into a series of vectors that take into the context. Notable attempts in contextualised embeddings include CoVe3 and ELMo, although the biggest breakthrough was achieved by BERT, a Transformer-based pretrained language model.
- BERT contextualises the input through a series of Transformer encoder layers, so it inherits all the strengths of the Transformer. Its self-attention mechanism enables it to "random access" over the input and capture long-term dependencies among input tokens. So unlike traditional language models, the Transformer can take into account the context in both directions

## Semi-supervised Learning Step

**Model:**



**Dataset:**

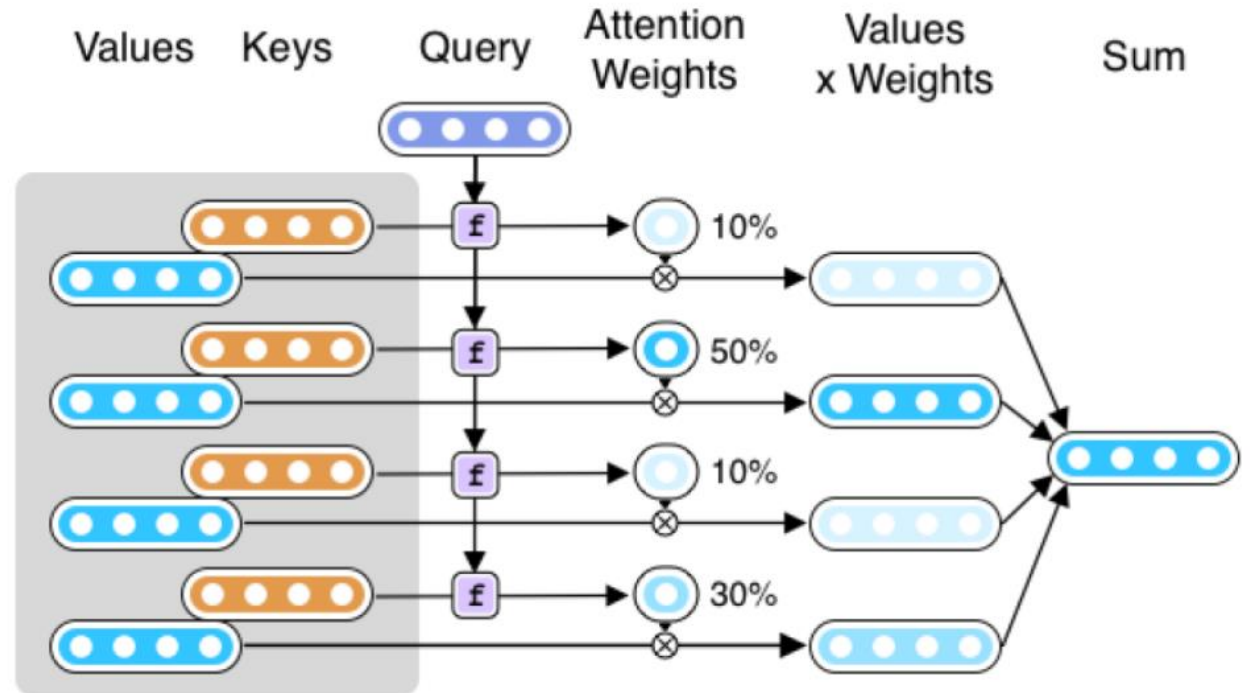


**Objective:**

Predict the masked word  
(language modeling)

# Attention!

- It is like having some sort of key-value store that contains all of the input's information and looking it up with a query (in the context).
- The stored values are not just a single vector, but a list of vectors for each token associated with its corresponding keys.



# Intent Classification: Task Perspective

- In essence, an intent classifier analyses texts automatically and categorizes them into intents such as booking, spam, complaint, query\_reservation, query\_service, query\_complaint, and so on.
- It enables the organization to be more customer-centric, particularly in areas like customer service and sales.
- Leads to faster dealing with massive number of queries while providing individualized service.

# Intent Classification: Dataset

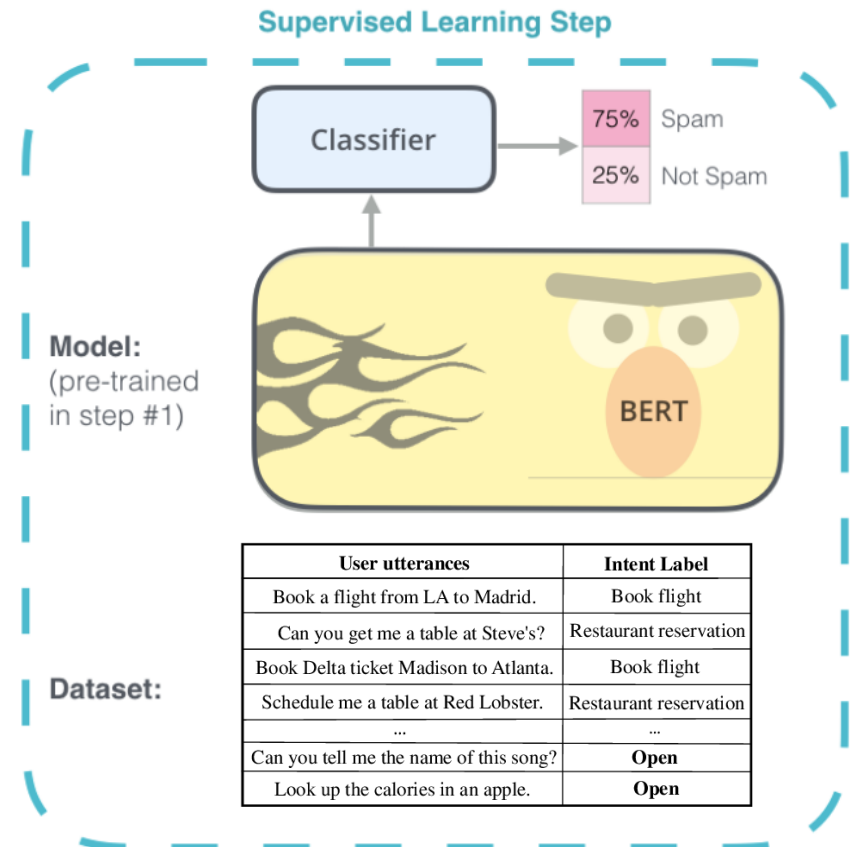
- Potential text coupled with intent as a label.

User utterances	Intent Label
Book a flight from LA to Madrid.	Book flight
Can you get me a table at Steve's?	Restaurant reservation
Book Delta ticket Madison to Atlanta.	Book flight
Schedule me a table at Red Lobster.	Restaurant reservation
...	...
Can you tell me the name of this song?	<b>Open</b>
Look up the calories in an apple.	<b>Open</b>

# Classification Task

- Each data instance is converted into a vector
- The machine learns to map the label to this vector
- The complete training data is mapped to its labels – instance by instance (or batch by batch)
- Machine understands which tokens correspond to what label

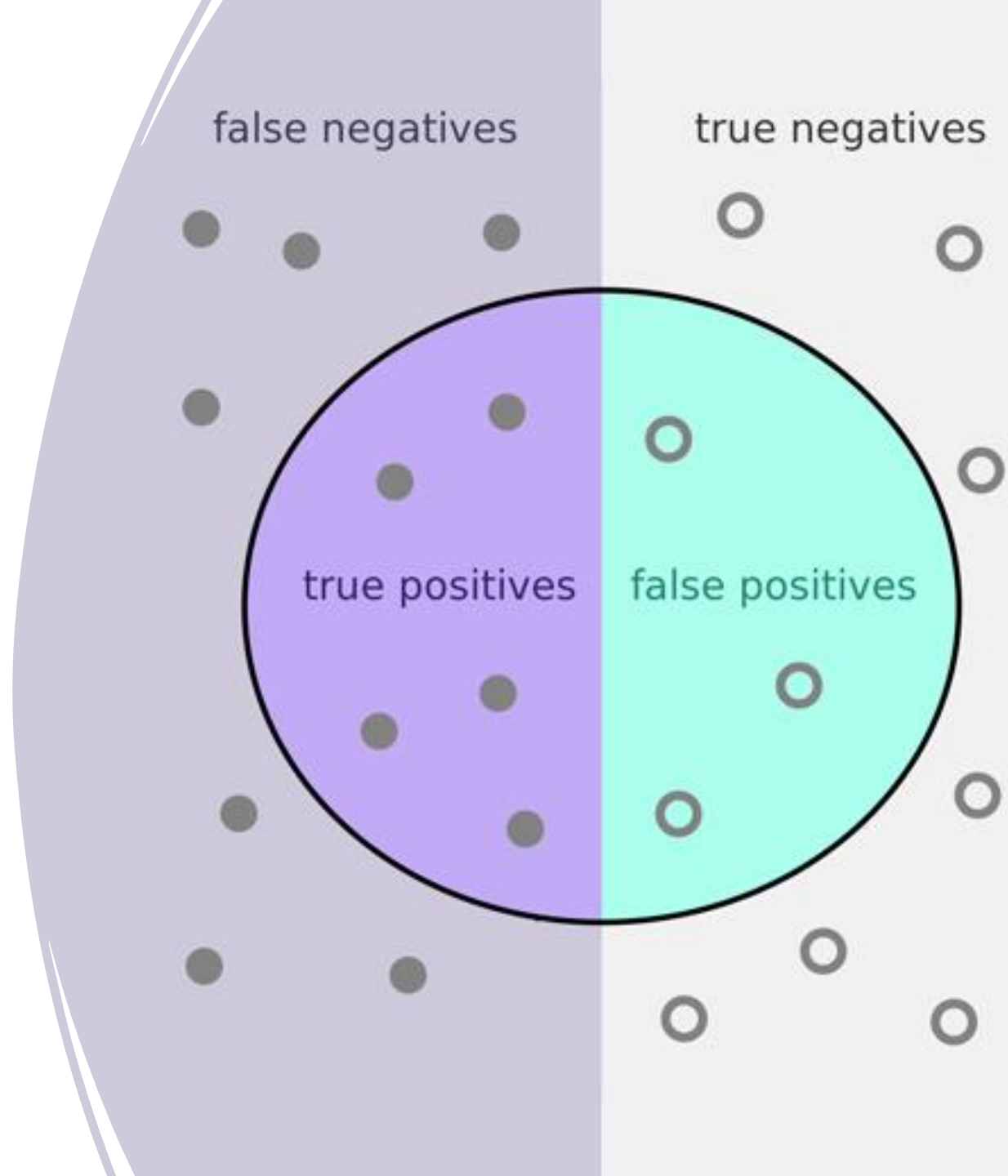
2 - **Supervised** training on a specific task with a labeled dataset.



# Testing / Evaluation

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- Held-out set from the data known as Test set.
- Evaluation using statistical measure known as F1-score which is based on Precision and Recall scores.
- Precision – quality of the prediction of intent
  - What proportion of intent identifications was actually correct?
- Recall – how many of the intents were found?

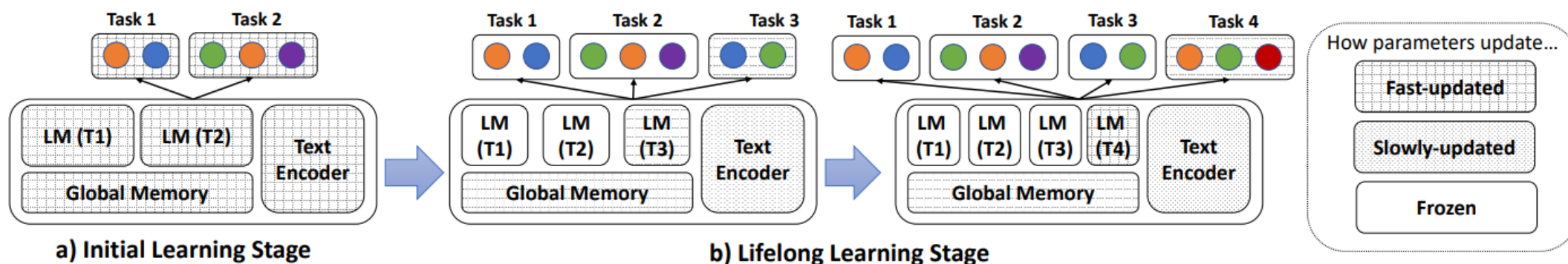


# Lifelong Learning framework

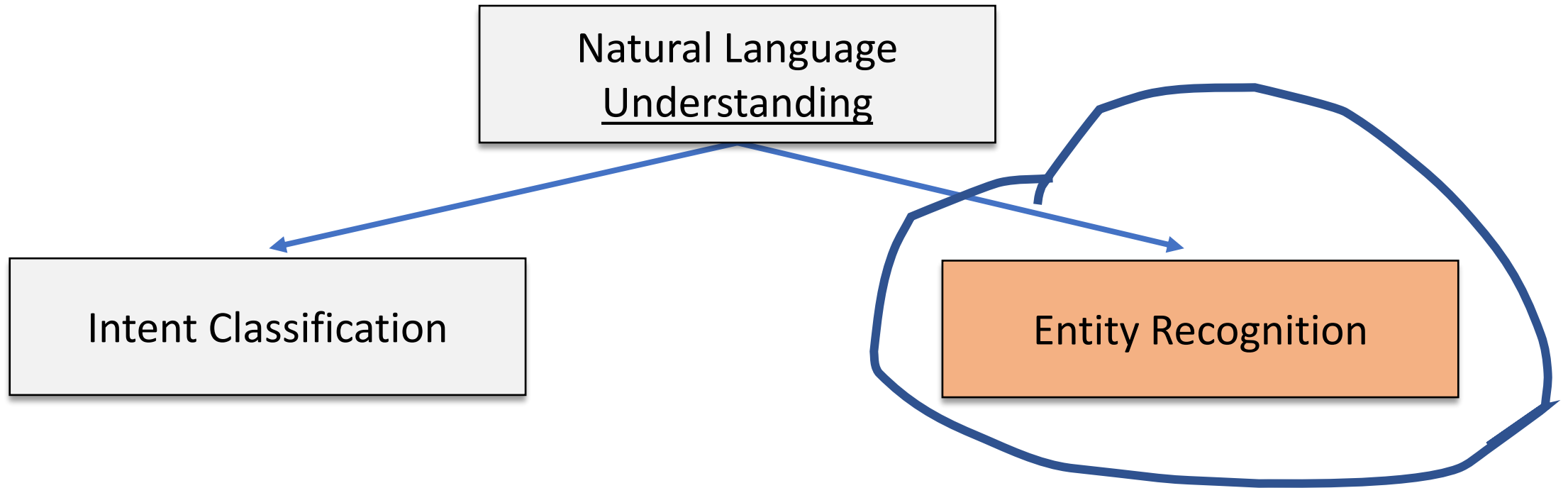
## MeLL: Large-scale Extensible User Intent Classification for Dialogue Systems with Meta Lifelong Learning

Chengyu Wang<sup>1\*</sup>, Haojie Pan<sup>1\*</sup>, Yuan Liu<sup>1</sup>, Kehan Chen<sup>1</sup>, Minghui Qiu<sup>1</sup>, Wei Zhou<sup>1</sup>, Jun Huang<sup>1</sup>, Haiqing Chen<sup>1</sup>, Wei Lin<sup>1</sup>, Deng Cai<sup>2</sup>

<sup>1</sup> Alibaba Group <sup>2</sup> State Key Lab of CAD & CG, Zhejiang University







Entity recognition is the task of identifying and extracting structured information from the unstructured text.

# (Named) Entity Recognition

In fact, the **Chinese** NORP market has the **three** CARDINAL most influential names of the retail and tech space – **Alibaba** GPE, **Baidu** ORG, and **Tencent** PERSON (collectively touted as **BAT** ORG), and is betting big in the global **AI** GPE in retail industry space. The **three** CARDINAL giants which are claimed to have a cut-throat competition with the **U.S.** GPE (in terms of resources and capital) are positioning themselves to become the ‘future **AI** PERSON platforms’. The trio is also expanding in other **Asian** NORP countries and investing heavily in the **U.S.** GPE based **AI** GPE startups to leverage the power of **AI** GPE. Backed by such powerful initiatives and presence of these conglomerates, the market in APAC AI is forecast to be the fastest-growing **one** CARDINAL, with an anticipated **CAGR** PERSON of **45%** PERCENT over **2018 - 2024** DATE.

To further elaborate on the geographical trends, **North America** LOC has procured **more than 50%** PERCENT of the global share in **2017** DATE and has been leading the regional landscape of **AI** GPE in the retail market. The **U.S.** GPE has a significant credit in the regional trends with **over 65%** PERCENT of investments (including M&As, private equity, and venture capital) in artificial intelligence technology. Additionally, the region is a huge hub for startups in tandem with the presence of tech titans, such as **Google** ORG, **IBM** ORG, and **Microsoft** ORG.

# Token Classification vs. Text Classification

In fact, the Chinese market has the three most influential names of the retail and tech space – Alibaba, Baidu, and Tencent (collectively touted as BAT), and is betting big in the global AI in retail industry space. The three giants which are claimed to have a cut-throat competition with the U.S. (in terms of resources and capital) are positioning themselves to become the 'future AI platforms'. The trio is also expanding in other Asian countries and investing heavily in the U.S. based AI startups to leverage the power of AI. Backed by such powerful initiatives and presence of these conglomerates, the market in APAC AI is forecast to be the fastest-growing one, with an anticipated CAGR of 45% over 2018 - 2024.

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Tokens are provided labels.

Token labels are predicted by machine

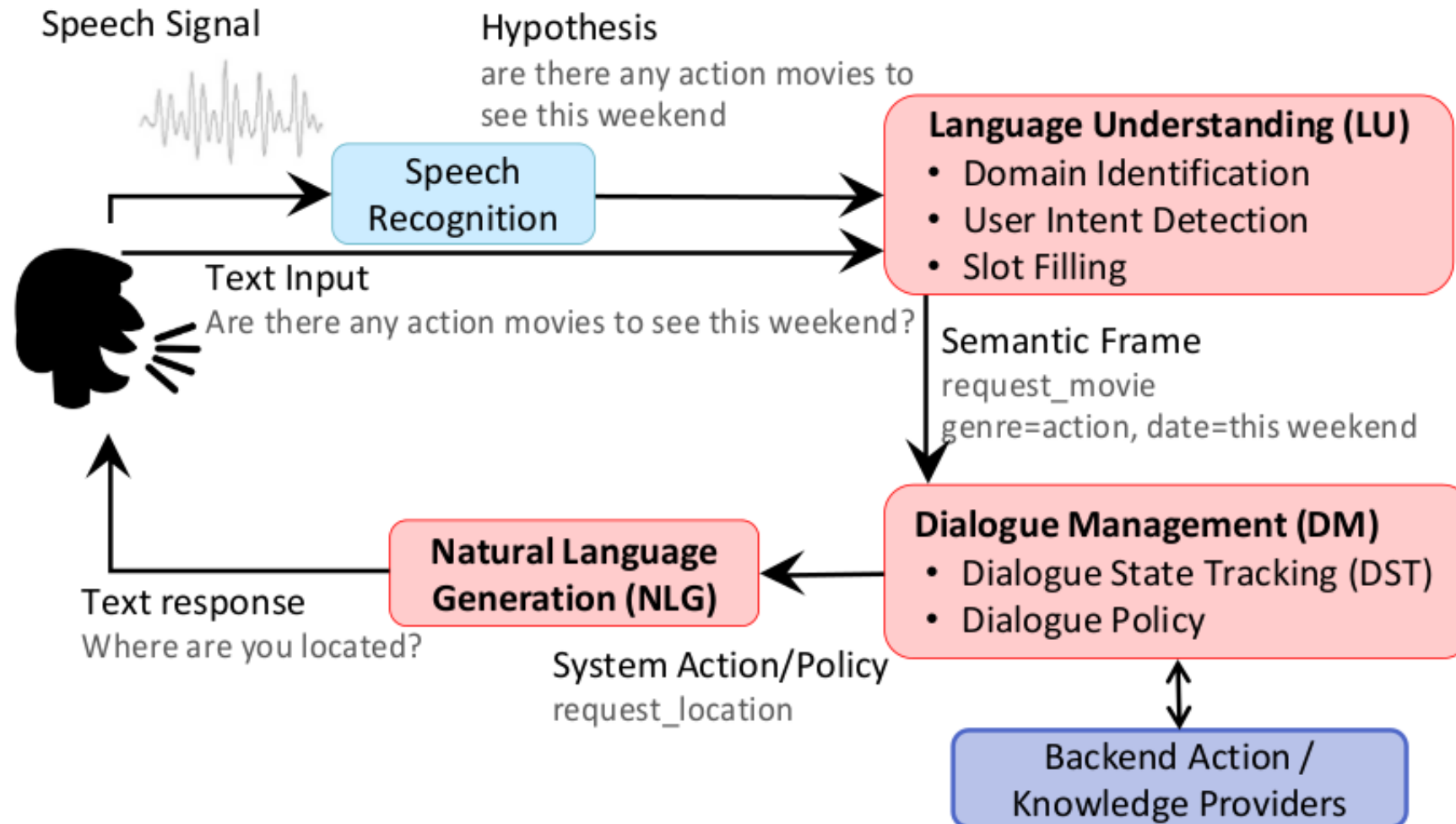
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Schedule me a table at Red Lobster.	Restaurant reservation
...	...
Can you tell me the name of this song?	Open
Look up the calories in an apple.	Open

Text sequence is provided a label.

Hence, text label is predicted by machine

Same evaluation technique used for both tasks.

# Dialogue Flow



# Challenges with Dialogue Flow

- Task oriented chat vs. chit-chatting.
- Change in user intent.
- Dialogue state management – continuation of a leftover conversation.
- Start with heuristics or rules – then go for generation.
- Language Generation!

# Other NLP Investigations

- Abbreviation Detection (LREC 2022)
  - We contribute a large resource for detection of abbreviations and acronyms in the scientific domain.
- Named Entity Recognition (LREC 2022)
  - We contribution the largest known manually annotated Hindi NER corpus.
- Offensive Language Identification for Indic languages (ongoing for Malayalam)
- Using Abstractive Summarization to automate the fact-checking pipeline (Submitted to COLING 2022)
- Investigation on use of Cognitive features to improve NLP task performance.

and many more...

# Thank you!

- Questions?

