

### Dialogue System in Practice Suning Bot Platform

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# Agenda

- Overview of Dialogue System
- Natural Language Understanding (NLU)
- Dialogue Management (DM)
- Natural Language Generation (NLG)
- AI + HI = MI Architecture
- Demo









### Information Retrieval Trend

Dialogue System, 2015







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## **Dialogue Structure**

#### Local Structure

- Utterance organized in turns
- Coherence between turns or utterances
- Single-turn and multi-turn

#### Global Structure

- Opening
  - Greetings
- Body
  - Topics
- Closing
  - farewells







## **Dialogue System**

#### Definition

- A computer system intended to converse with a human, with a coherent structure
- Dialogue Characteristics
  - Exchanging ideas
  - Requesting information
  - Sustaining relationships







## **Dialogue System Types**

#### Chat Bot

- Free chat on specific or open topics
- xiaolce, Microsoft
- Duer, Baidu
- Task Bot
  - Task completion on special domains
  - Cortana, Microsoft
  - Siri, Apple







## **Dialogue System Models**

#### Data-driven models

- Large-scale corpus is required
- Retrieval-based methods
- Generation-based methods
- Interaction-driven models
  - Online interaction between users and agents
  - Reinforcement learning-based approaches







### **Question Answering Overview**

#### Definition

Automatically answer a user's question by machine

#### • Types of QA

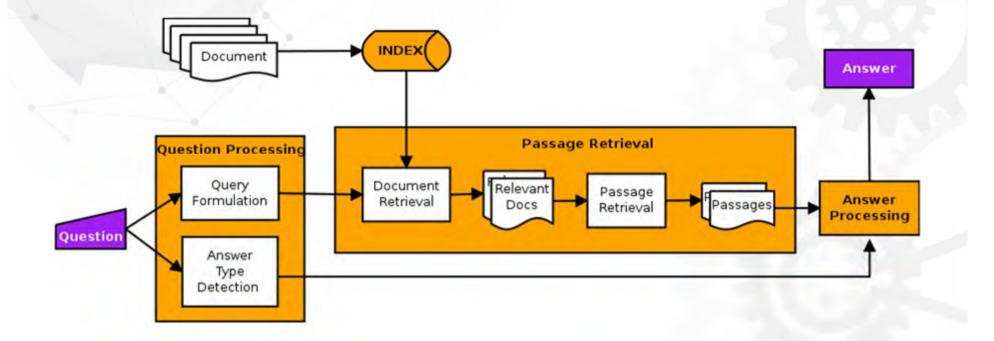
- Types of questions
  - Factoid, who is the wife of Obama
  - Opinion, what is your point to air pollution
  - Yes-no, are you happy?
  - Comparison, what are the differences between mac and windows?
- Types of content
  - Textual QA
  - Visual QA
- Key technology behind Dialogue System





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### **Question Answering Architecture**





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### Difference between QA and IR

#### Input

- QA, utterance, free text
- IR, query word
- Output
  - QA, answers
  - IR, documents
- Key technology
  - QA, provide answers for given questions
  - IR, find documents related to queries







### Example

Baid 百度吃饭,不要日餐

百度一下

网页 新闻 贴吧 知道 音乐 图片 视频 地图 文库 更多»

百度为惩找到相关结果约29,500,000个

了搜索工具

#### 吃饭一定要一日三餐吗? 百度知道

5个回答-提问时间:2011年11月05日 最佳答案:当然不一定,你可少吃多餐,而且科学实验证明这种进食方式对身体很好,一般是一日 三餐的,早晨要吃好(早上吃的像皇帝).中午要吃饱(中午吃的像平民).晚上... 更多关于吃饭,不要日餐的问题>> zhidao.baidu.com/link?...+--直度快照-评价

人吃饭要一日三餐,老公做爱为什么不可以做到一日三次....爱问知识人

人到了吃饭时间点感觉不饿是不是可以不用按时吃饭?\_百度知道

吃日料。你一定用的到的日餐礼仪」-搜狐旅游



2016年4月21日 - 康大泰生嘉禾日餐厅在正式学习日餐礼仪之前,先为 您介绍一家地道的日式料理餐厅,位于青岛康大泰生大酒店一樣的嘉禾 日料,纯日式装修风格的和室,身看日式服装的服务... travel.sohu.com/201604. - - 百度快照 - 61条评价

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### **Dialogue System Components**

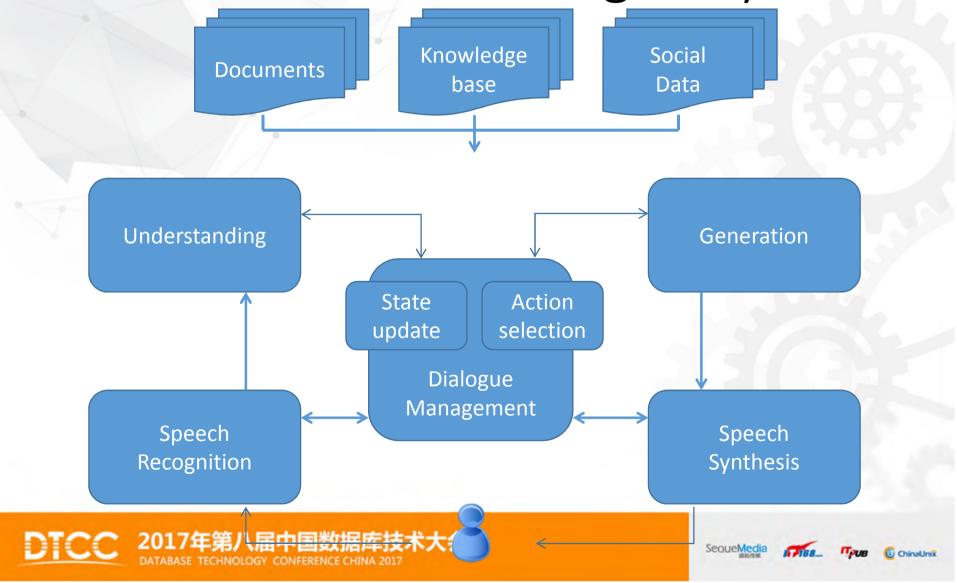
- Automatic Speech Recognition (ASR)
  - Recognize voice signal to textual representation
- Natural Language Understanding (NLU)
  - Convert text to semantic representation
- Dialogue Management (DM)
  - Update dialogue state and perform the right actions
- Natural Language Generation (NLG)
  - Select the best answer based on current (state, action)
- Text-to-Speech Synthesis (TTS)
  - Synthesize natural voices using the generated text





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### Architecture for Dialogue System



### Natural Language Understanding

#### Definition

• Convert text to semantic representation

#### Challenges

• Semantic gap between literal text and semantics

#### Examples

- 我想买手机
- 推荐一款女性手机,不要苹果
- 有IPhone么?
- 水果机怎么卖?







## NLU Challenges

- Speech Recognition Error
  - Word error rate 10-20%
- Short-text processing
  - Less context
  - Sparsity
- Semantically similar expressions
  - Polysemy
  - Synonyms
- Colloquial transcription
  - Incorrect syntactic parsing







# NLU Approaches

#### Rule-based methods

• Expert rules to extract the semantics from text

Pros

- No labeled training data is required
- Easy to explain
- Cons
  - Expert knowledge to define general rules
  - Time consuming, over-specified and incomplete







# NLU Approaches

#### Statistical Methods

• Train statistical models based on labeled data

Pros

- Less expert involvement
- Data + feature engineering + algorithm = desired model
- Cons
  - Labeled training data is required
  - Difficult to explain







### NLU as a Service

#### Definition

- Intention identification
- Entity extraction
- Existing NLU services
  - WIT.AI
  - API.AI
  - LUIS.AI
  - Watson
  - Alexa







### Comparison of NLU services

Services	Free or paid	Multi- language	Basic functionality	User-defined	Reference
WIT.AI	Free	Yes, 11	Intent, entity, context, action	Intent, entity, context, action	https://wit.ai
API.AI	Paid	Yes, 14	Intent, entity, context, action, text2speech, speech2text, cross- platform	Intent, entity	https://api.ai/
LUIS.AI	Free	Yes,	Intent, entity, cross- platform	Intent, entity	https://www.luis. ai/
ALEXA skill set	Free	Yes,	Intent, entity, household platform	Intent, entity, action	https://develope r.amazon.com/al exa-skills-kit
WATSON AlchemyAPI	Paid	Yes,	Intent, entity, relation, sentiment	Not yet	http://www.alch emyapi.com/api
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### Dialogue Management

- Brain of the Dialogue System
- Dialogue State Update
  - DM maintains current dialogue state
  - DM updates the state as new input arrives
- Action Selection
  - Make decisions based on current state and input





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## DM Approaches

#### Frame-based method

- Frame is a set of slot-value pairs
- Frame is gradually filled by the multi-turn request-response pairs
- Pros
  - Easy to implement
- Cons
  - Difficult to extend to other domains



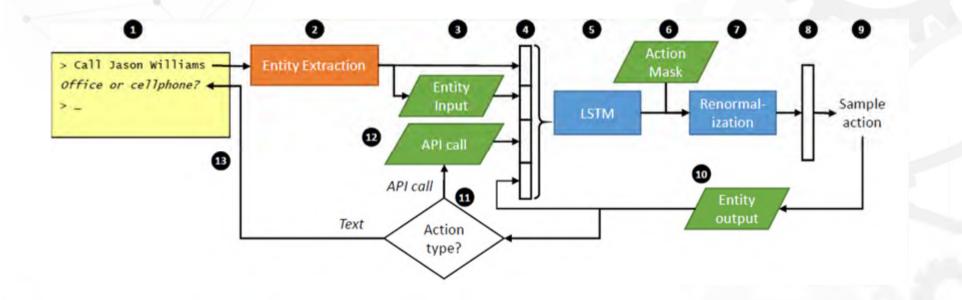




## DM Approaches

#### Neural network learning approach

- Formulated as a sequence labeling problem
- Predict the next action based on (input, state, etc)



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### Natural Language Generation

#### Definition

• Given input, create a natural language expression that is well-formed and human-readable

#### Possible input

- User utterance
- User profile
- Dialogue state
- Communication goal
- Generated output
  - Passage or sentence
  - Answer in knowledge base
  - Automatic summarization



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## NLG Pipeline

#### [Input]

- Content/Text planning (what to say)
  - Intent, entity, context, goal, etc

#### [Text plan]

- Sentence plan/realization (how to say)
  - Sentence selection, re-ordering and ranking
- [Rendering]
  - Surface plan/realization (how to present)
    - Response decoration







## NLG Approaches

- Retrieval-based methods
  - Fixed response set
  - Closed-domain
  - Professional assistant
- Generation-based methods
  - Dynamic responses
  - Open-domain
  - Social assistant

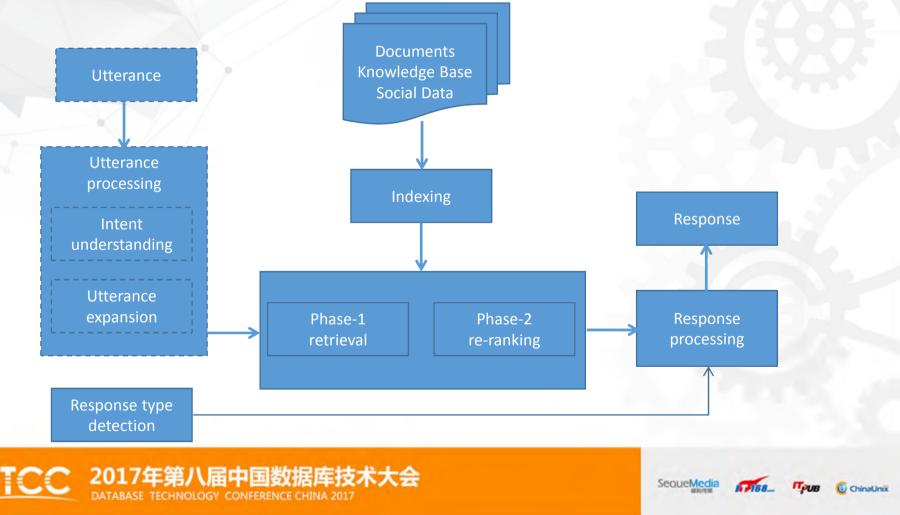






## NLG Approaches

#### Architecture for retrieval-based methods



### AI+HI in Dialogue System

- Current working model
  - Completely HI
  - AI + HI
  - Completely Al
- Proposed working model
  - Group chat involving user, bot and human agents
  - Bot activates the human agent involvement
    - Based on context, user profile and agent expertise
    - AI + HI = MI, Mixed Intelligence





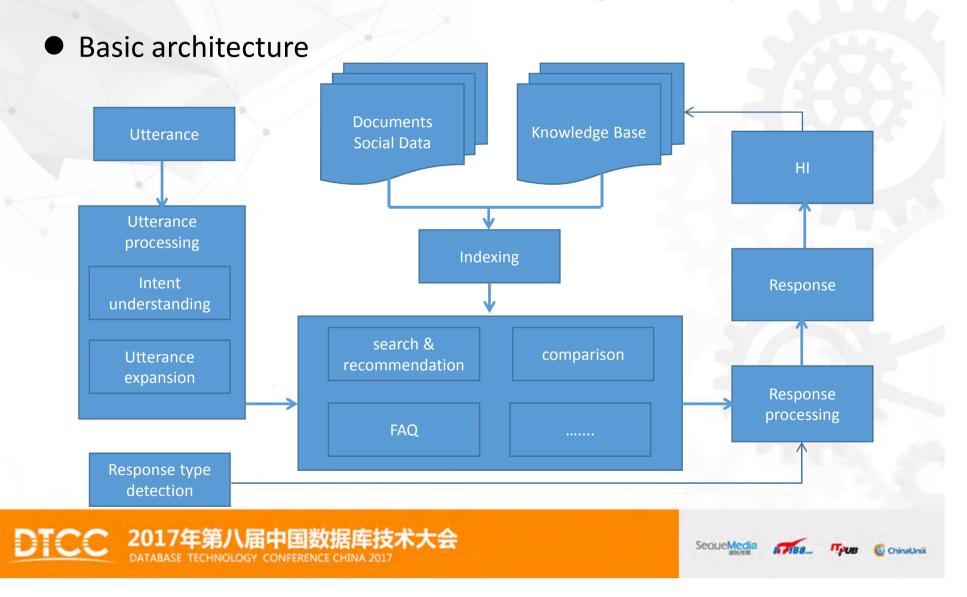


### Mixed Intelligence Architecture

#### • MI = AI + HI



### Retrieval-based Dialogue System



### **Evaluation of Dialogue System**

#### Data

- SQuAD, Stanford Question Answering Dataset
- TREC QA, Text REtrieval Conference QA dataset
- bAbi, Facebook QA dataset
- Offline evaluation
  - Model evaluation metrics
- Online evaluation
  - User engagement
  - Conversion rate







### **Application Scenario**

#### • Characteristics of AI

- Work anytime, 7/24
- Big memory
- Consistency
- Suitable tasks
  - Simple, routine work
  - High concurrency, large traffic



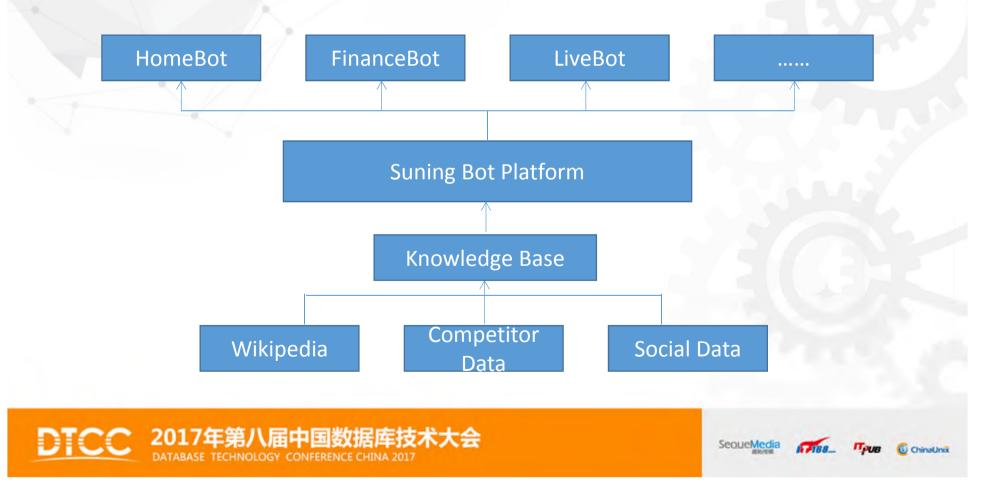




## Suning Bot Platform

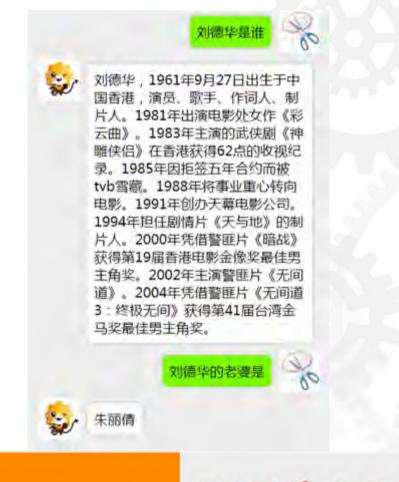
#### XBot Platform

HomeBot, FinanceBot, LiveBot and AdsBot, etc



### Simple Demos

### ● 小苏-智能机器人 00 哈哈哈哈哈哈,宝宝什么事这么高 兴!? 我想和你去看电影 我喜欢电视,但妈妈说看多了对眼 睛不好呢 and and 不要抛弃我好吗 感冒怎么办 999感冒灵,暖暖的,很贴心



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