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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

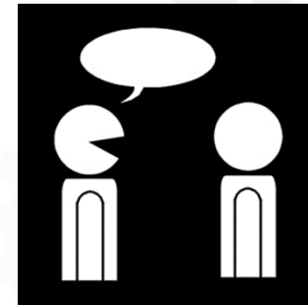
Library Search, 1970



Web Search, 1990



Dialogue System, 2015



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

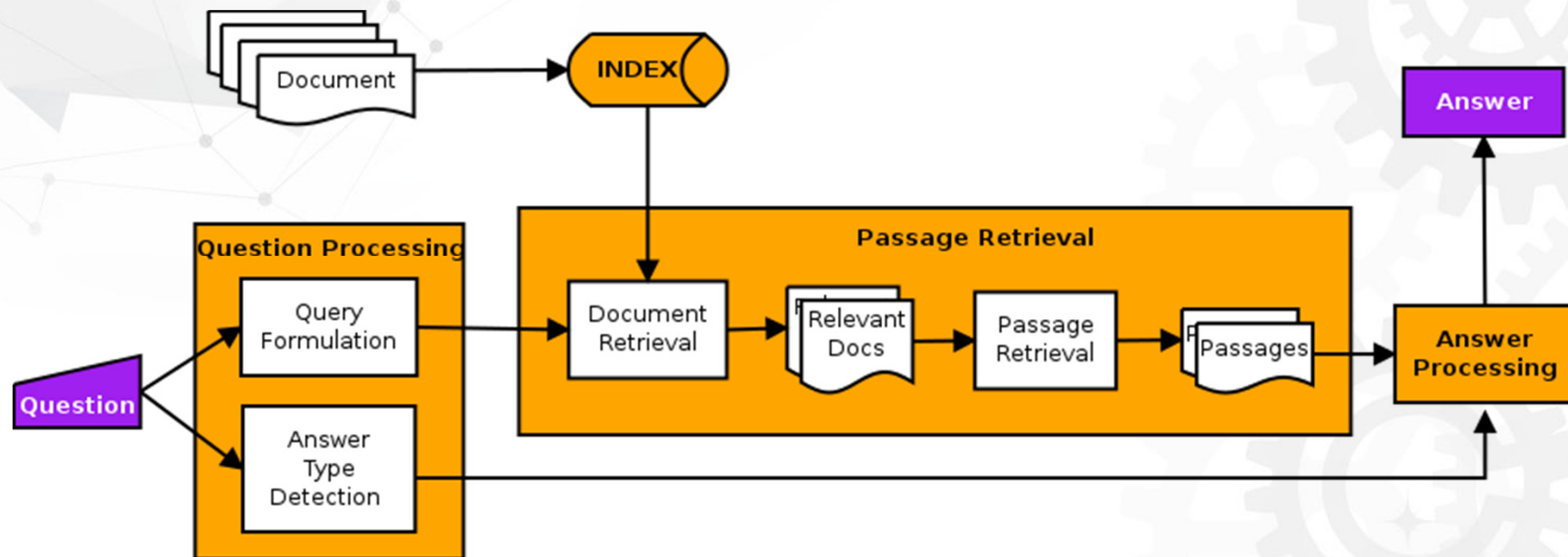
Dialogue System Models

- 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

Question Answering Architecture



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



Baidu 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»

百度为您找到相关结果约29,500,000个  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?... - 百度快照 - 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](#)
2017年3月10日 - 人吃饭要一日三餐,老公做爱为什么不可以做到一日三次呢?buyimeiren ...我想看日本AV,哪位有具体网址? 西片里外国白人、黑人的阴茎那么长,很清楚...
[iask.sina.com.cn/b/203... - 百度快照 - 1925条评价](#)

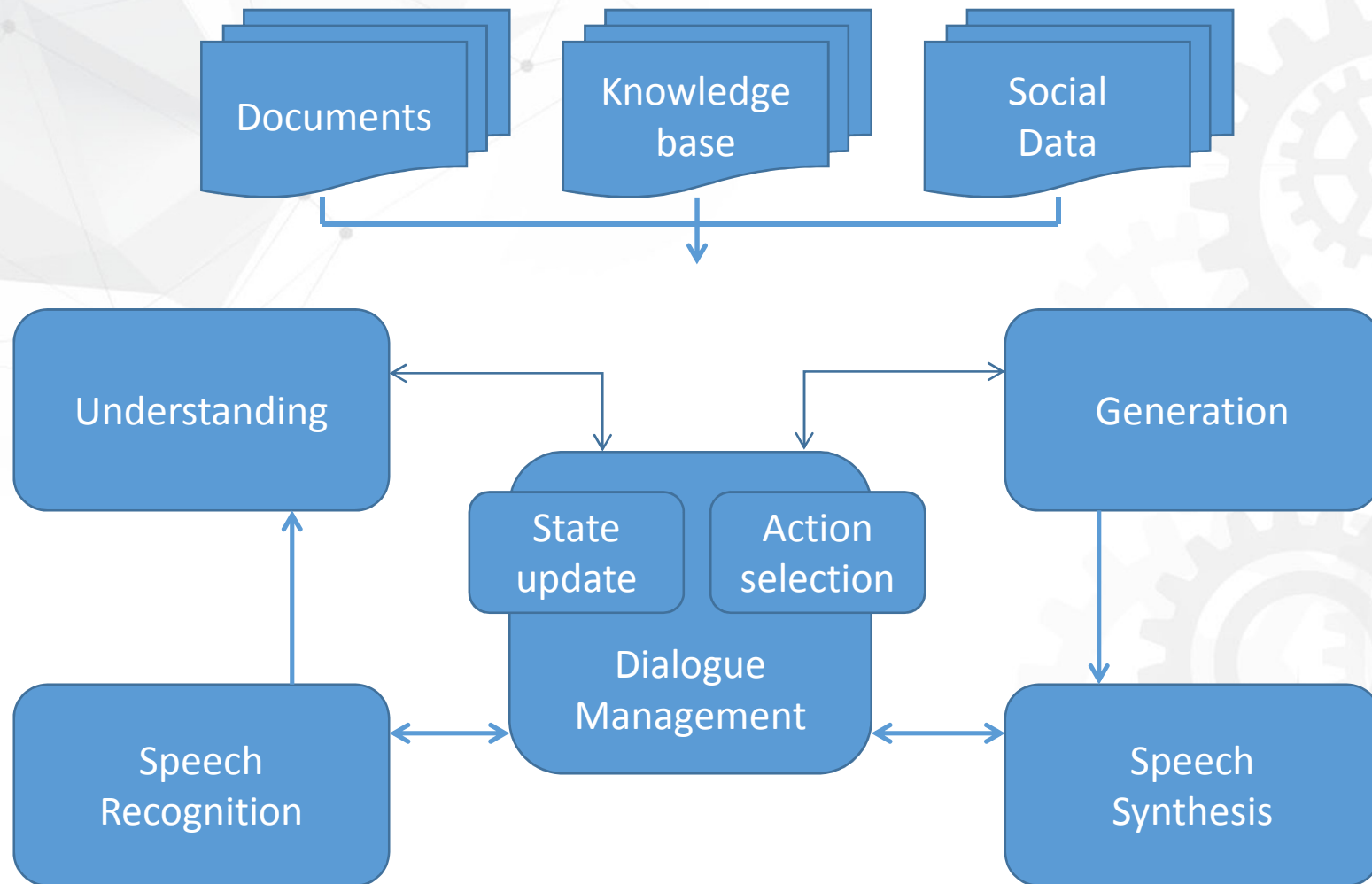
[人到了吃饭时间点感觉不饿是不是可以不用按时吃饭?_百度知道](#)
11个回答 - 最新回答: 2015年09月25日 - 4人觉得有用
要按时吃饭才好 追问 我觉得人应该是感觉饿了才吃 按照一日三餐的习俗太束缚...三餐要规律,身体机能才会正常,不饿也要加减吃,才能维持体力到下一餐之前,不要...
[更多关于吃饭,不要日餐的问题>>](#)
[zhidao.baidu.com/link?... - 百度快照 - 评价](#)

[吃日料,你一定用的到的日餐礼仪!_搜狐旅游](#)
 2016年4月21日 - 康大家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,位于青岛康大家生大酒店一楼的嘉禾日料,纯日式装修风格和室,身着日式服装的服务...
[travel.sohu.com/201604... - 百度快照 - 61条评价](#)

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

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://developer.amazon.com/alexa-skills-kit
WATSON AlchemyAPI	Paid	Yes,	Intent, entity, relation, sentiment	Not yet	http://www.alchemyapi.com/api

Dialogue Management

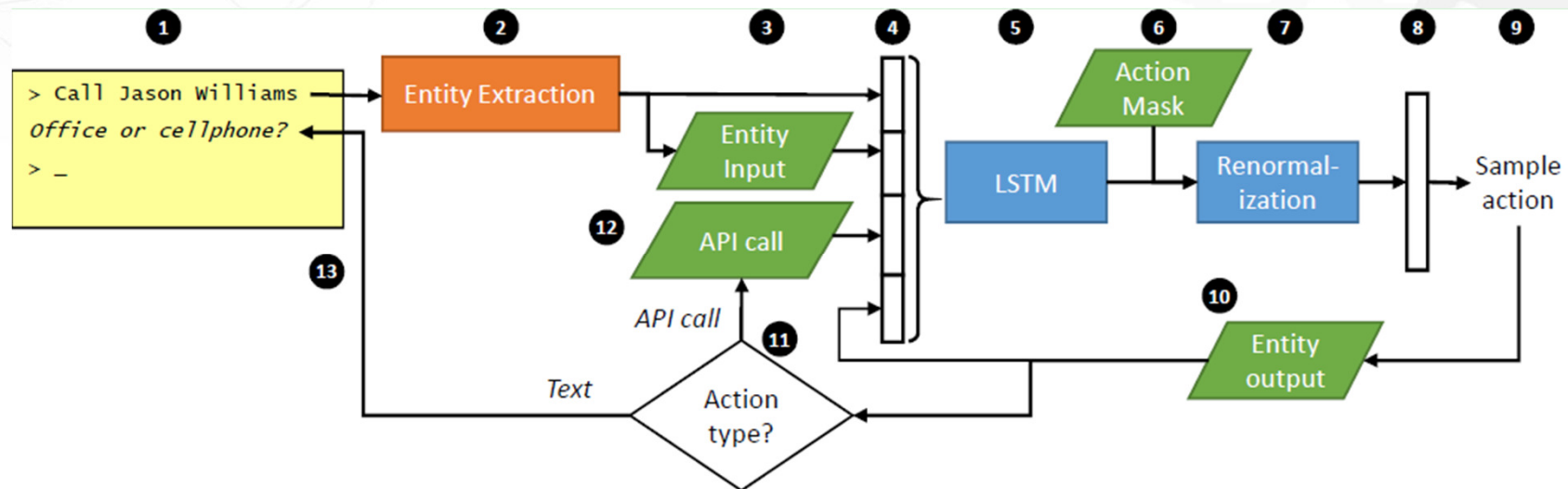
- Brain of the Dialogue System
- Dialogue State Update
 - DM maintains current dialogue state
 - DM updates the state as new input arrives
- Action Selection
 - Make decisions based on current state and input

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)



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

NLG Pipeline

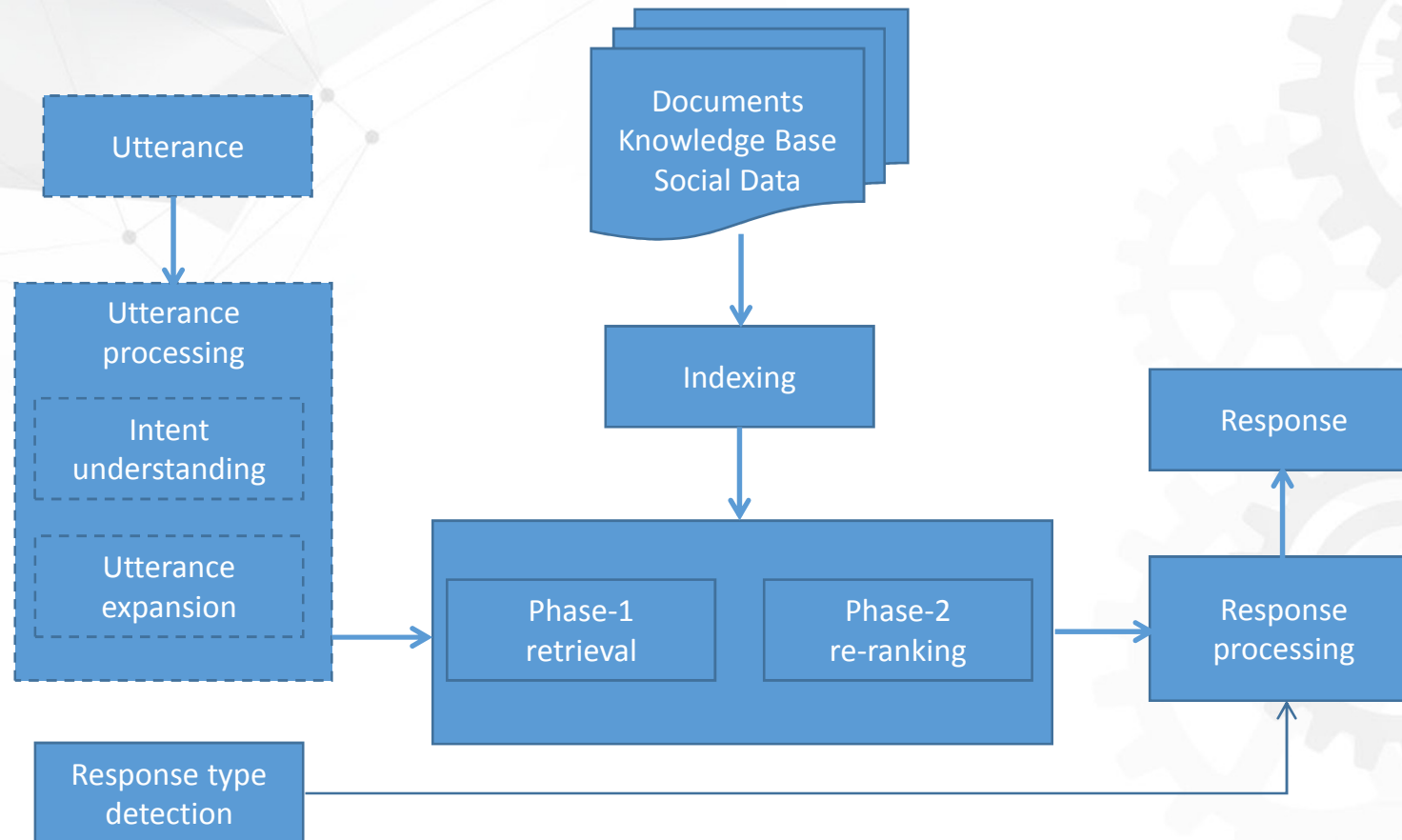
- [Input]
 - Content/Text planning (what to say)
 - Intent, entity, context, goal, etc
- [Text plan]
 - Sentence plan/realization (how to say)
 - Sentence selection, re-ordering and ranking
- [Rendering]
 - Surface plan/realization (how to present)
 - Response decoration

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

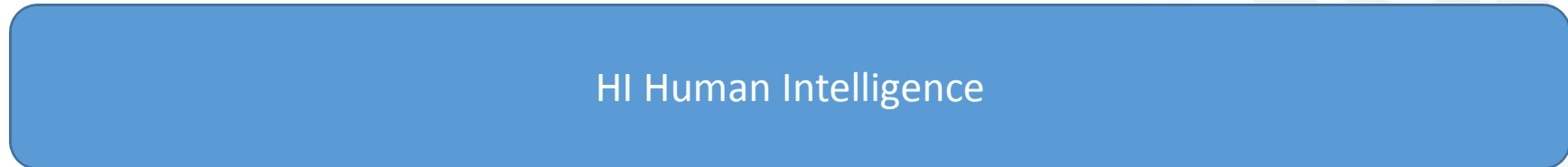


AI+HI in Dialogue System

- Current working model
 - Completely HI
 - AI + HI
 - Completely AI
- 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

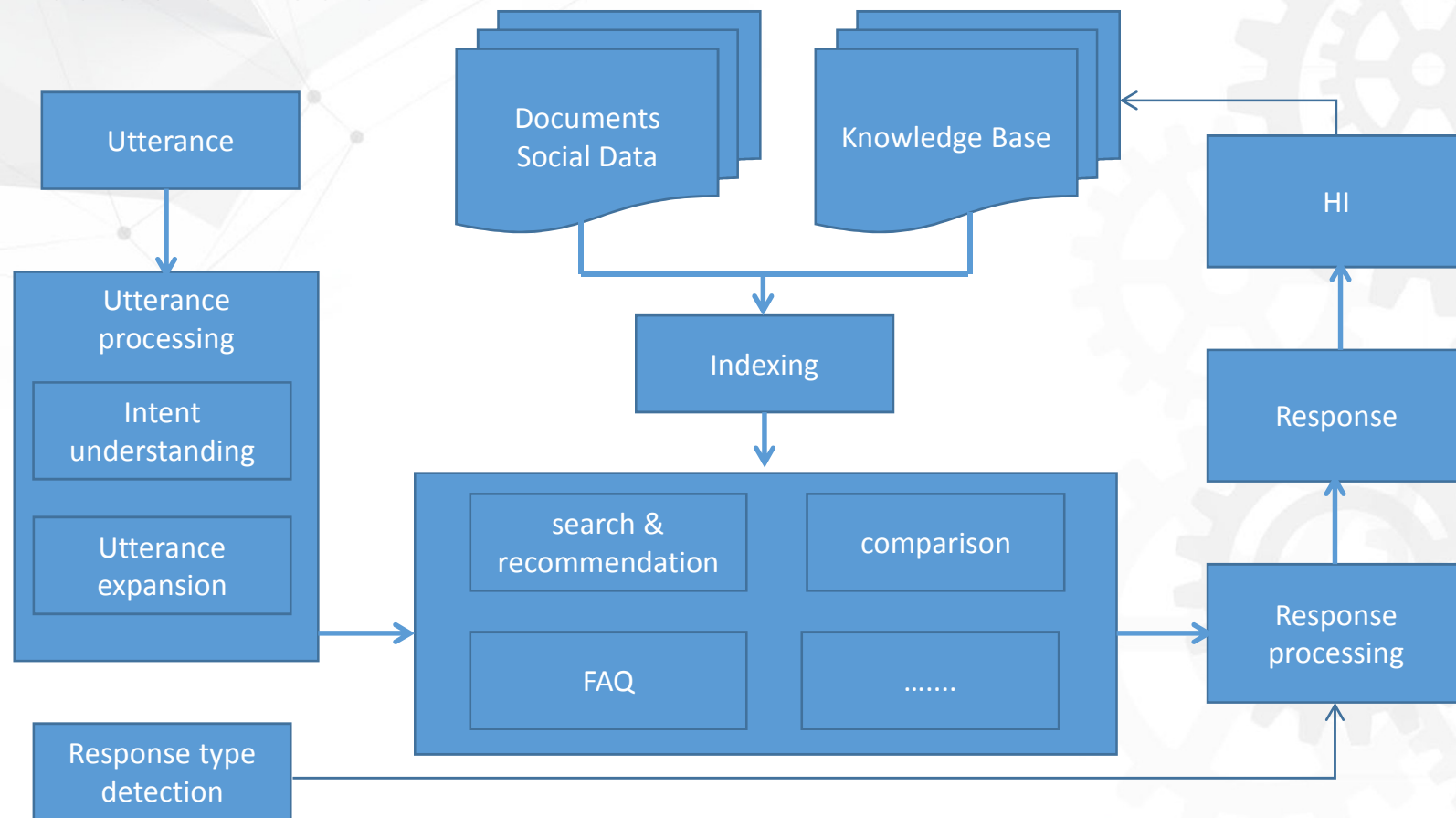
Mixed Intelligence Architecture

- $MI = AI + HI$



Retrieval-based Dialogue System

- Basic architecture



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

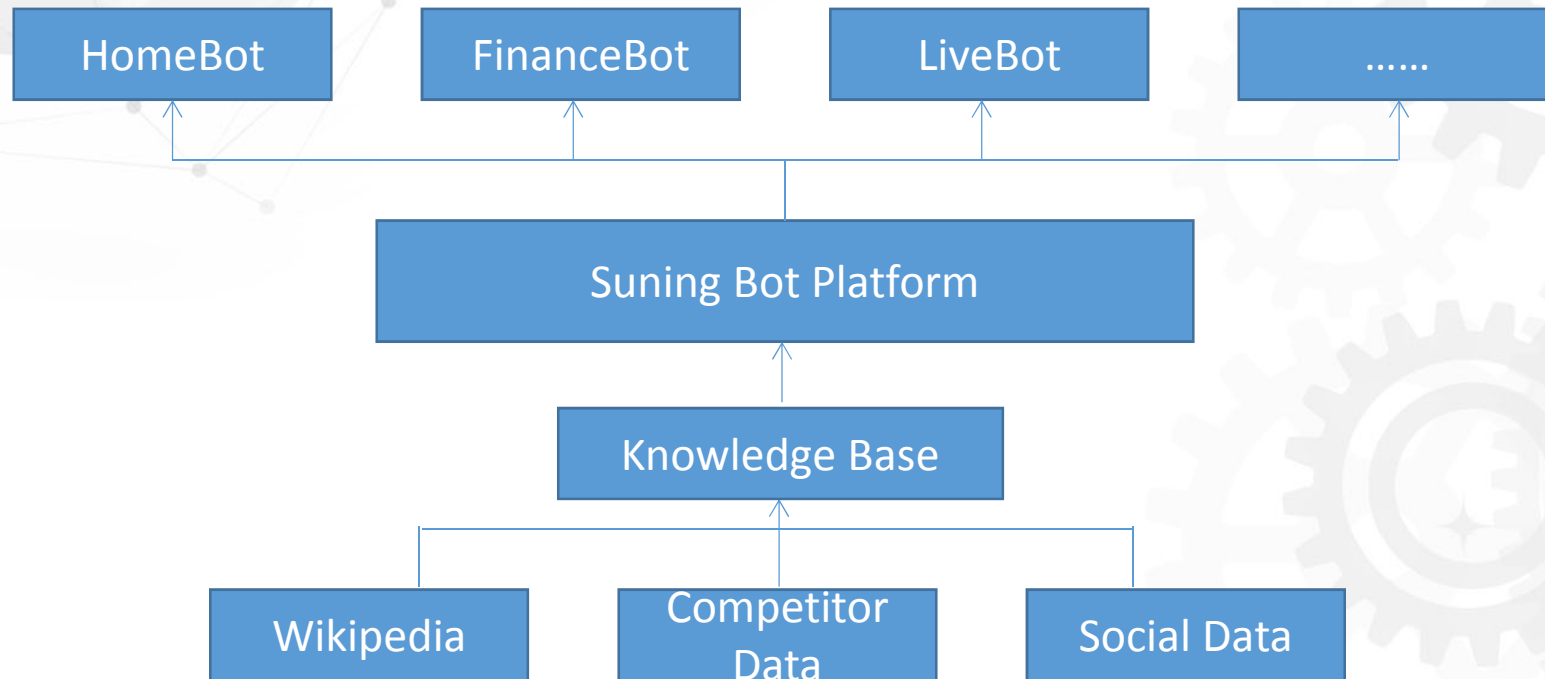
Application Scenario

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

Suning Bot Platform

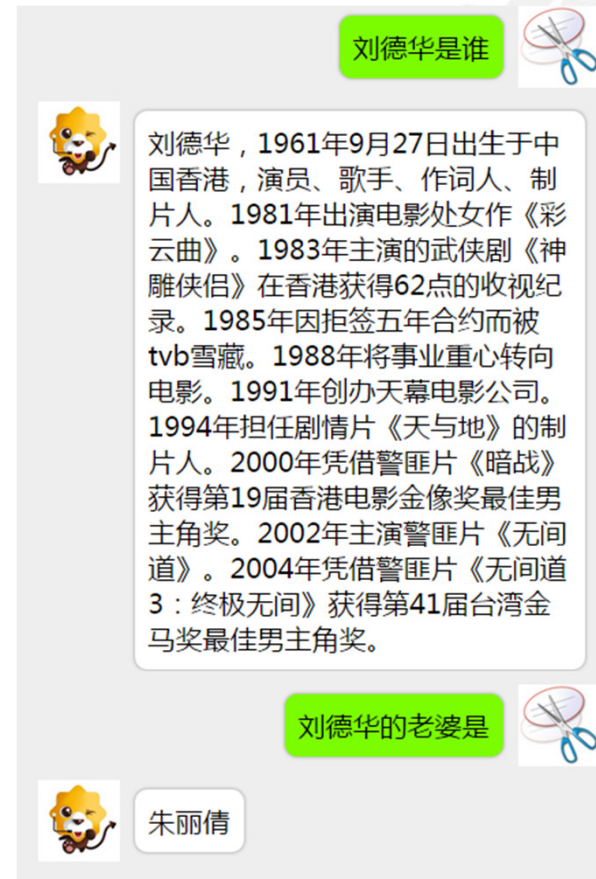
- XBot Platform

- HomeBot, FinanceBot, LiveBot and AdsBot, etc



Simple Demos

- 小苏-智能机器人





THANKS

SequeMedia
盛拓传媒

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