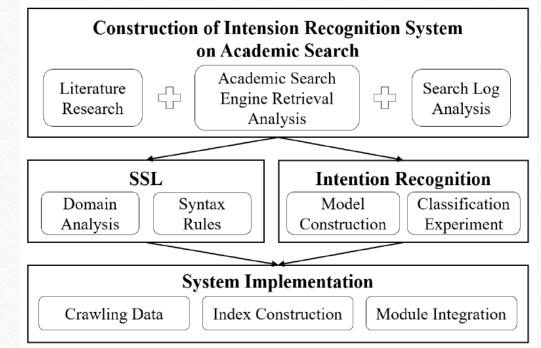
Design and Implementation of an Academic Search System Based on a General Query Language and Automatic Question Answering

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## INTRODUCTION

- Design of a general query language for academic search.
- Understanding the search intentions of users' questions.
- Implemented system of QA-oriented academic search engine .



#### SSL: SCHOLAR SPECIFIC LANGUAGE

#### • An expansion of traditional DSL.

#### 3 main modules:

Type module: Represent the type of purpose information or intention.
Field module: Which expresses the query mode which is the specific combination of destination information attributes.

3. **Refinement module**: Which is used to represent the refined query semantics of the result information, includes both the post-filtering semantics and the secondary retrieval semantics.

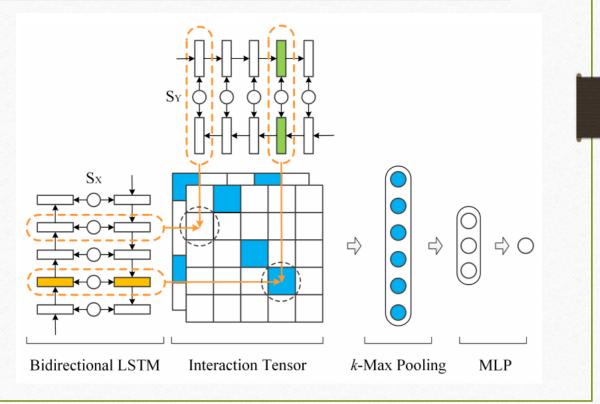
#### SSL: SCHOLAR SPECIFIC LANGUAGE

- intent = "" module ""
- module = type "," field ["," refinement]
- type = DQUOTE "type" DQUOTE ":" intent-category
- field = DQUOTE "field" DQUOTE ":" fields
- refinement = DQUOTE "refinement" DQUOTE ":" refinements
- intent-category = DQUOTE ("paper"/"citation"/"entity"/"concept"/"qa") DQUOTE
- name = "subject", "author", "time", "title", "keywords", "abstract", "source", "institution", "foundation", "doi", "classification\_code", "content", "paper\_type", "journal", "conference", "question", "concept"
- item = DQUOTE [bool] 1\*char DQUOTE

- item = DQUOTE [bool] 1\*char DQUOTE
- bool = "+" / "|" / "-"
- refinements = "" quantity "," rank ["," field] ""
- quantity = DQUOTE "quantity" DQUOTE ":" quantity-item
- quantity-item = "-1" / number
- rank = DQUOTE "rank" DQUOTE ":" rank-item
- rank-item = DQUOTE ("relevence" / "citations" / "down-load\_num" / "time") DQUOTE
- number = D \*DIGIT
- D = "1" / "2" / "3" / "4" / "5" / "6" / "7" / "8" / "9"
- char = unescaped / escaped
- unescaped = %x20-21 / %x23-5B / %x5D-FF
- escaped = %x5C (""" / "/ "b" / "f" / "n" / "r" / "t" / ("u"4(ALPHA / DIGIT)))

### UNDERSTANDING USER INTENTION

Intention Recognition
Use DL-based text classifiers to infer
user's search intention through dialog
text(MV-LSTM, 2016).



### UNDERSTANDING USER INTENTION

#### Experiment:

- Dataset Baidu Academic and Luojia Academic search
- Labelled with 4 labels:

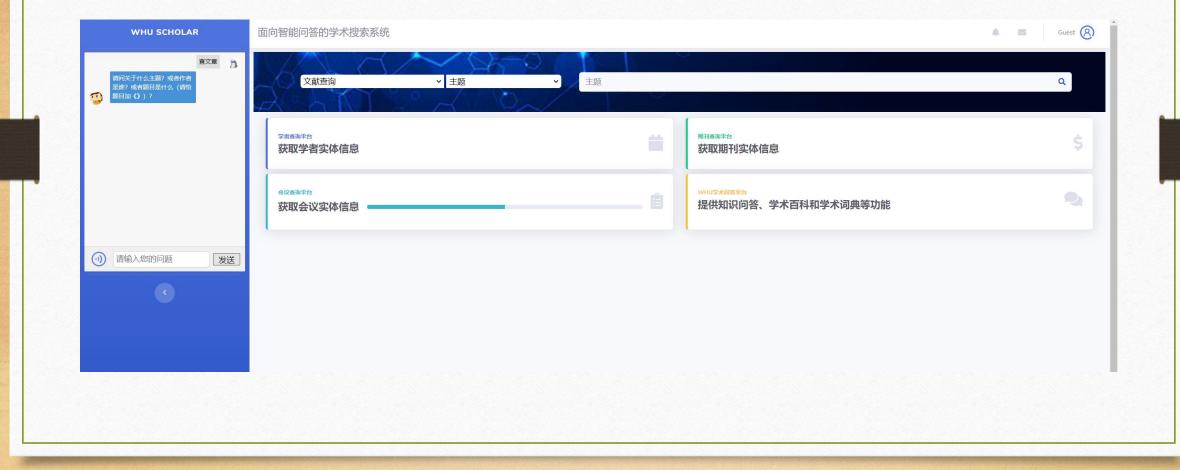
0 literature query1 academic entities query3 academic concept query4 free question-answering

Methods	<b>Precision</b> <sub>macro</sub>	<b>Recall</b> <sub>macro</sub>	F1 <sub>macro</sub>
Naive Bayes	0.6969	0.6952	0.6855
Logistic Regression	0.9506	0.9549	0.9505
SVM	0.9136	0.9124	0.9126
MV-LSTM(Ours)	0.9519	0.9512	0.9513

#### **QA-ORIENTED ACADEMIC SEARCH ENGINE**

Application Layer	QA-Oriented Academic Search Engine
	Interactive Question and Academic search
Technique Layer	Question Answering Module
	Speech Recognition     Intention Recognition     Dialog
	Question Parsing Speech Synthesis
	Retrieval Module
	Search Language Parsing Information Retrieval
Data Layer	Wiki Data     Literature Data     Indexing Data
	Vocabulary Resources Search Log Data
Physical Layer	Server Storage Network Backup

### System Demo



# Thanks For Listening