

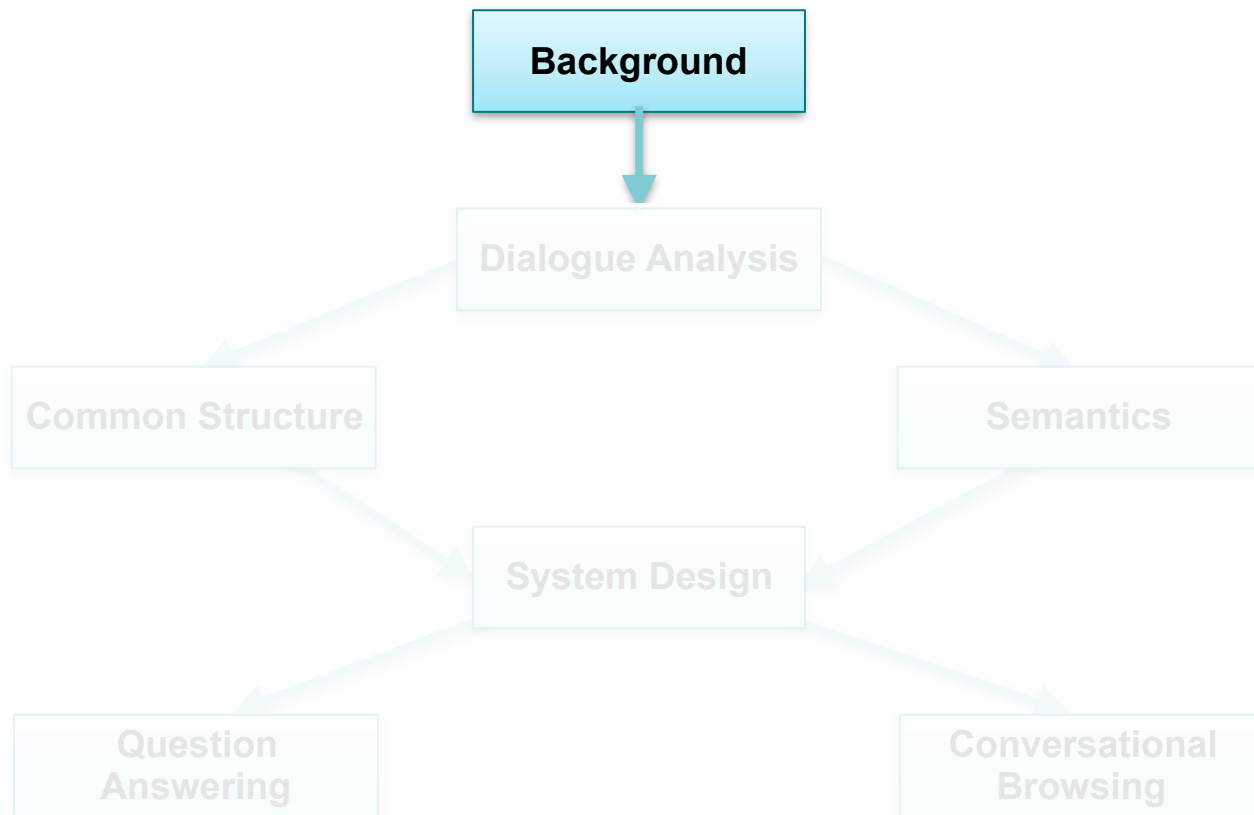


TECHNISCHE
UNIVERSITÄT
WIEN
Vienna | Austria

Knowledge-based Conversational Search

Svitlana Vakulenko

Outline



Conversational Search

Conversational Search?

- Information Seeking **Task**
- Dialogue **Interface**



Conversational Search?

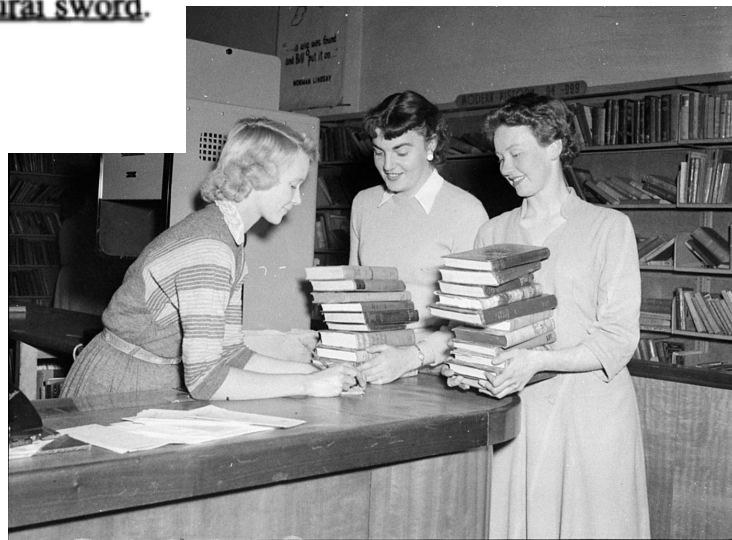
- **Information Seeking Task**
- **Dialogue Interface**

Search



Information-seeking Dialogue

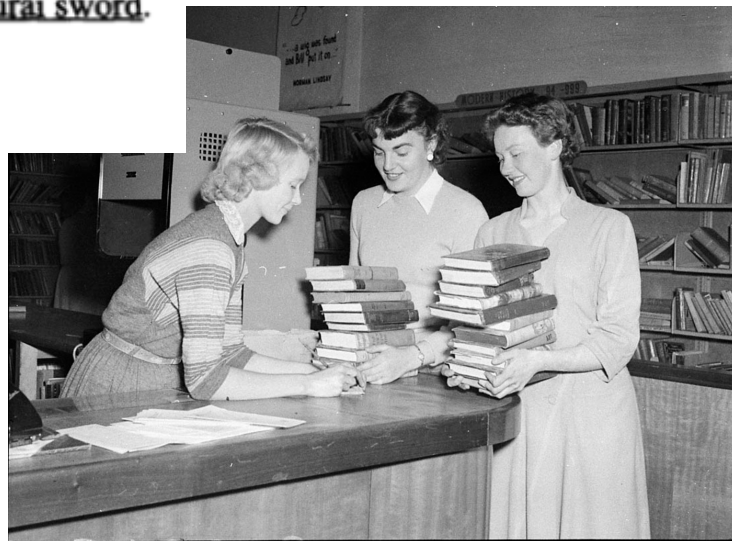
IS: Do you have the book that *they showed* in the filmstrip?
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Information-seeking Dialogue

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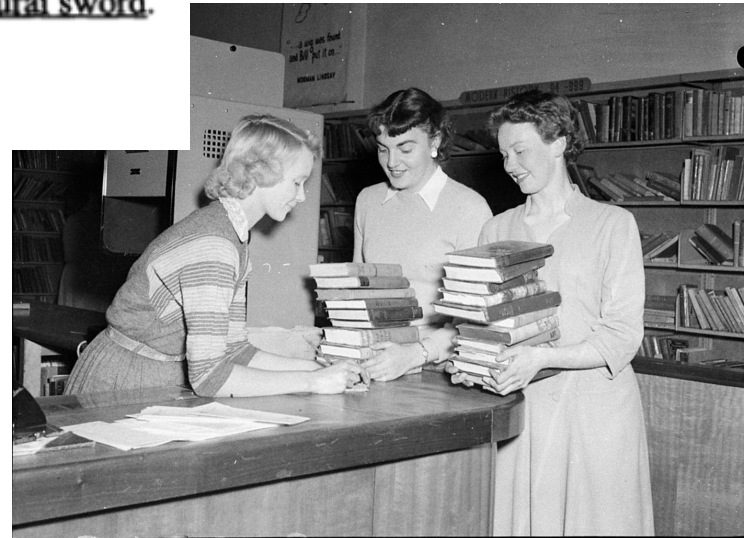
1. Ambiguous (underspecified) questions



Information-seeking Dialogue

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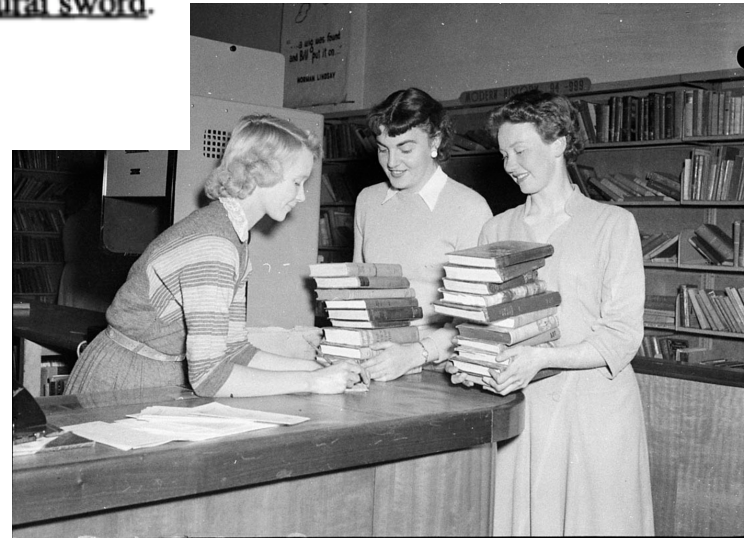
1. Ambiguous (underspecified) questions
2. Guessing



Information-seeking Dialogue

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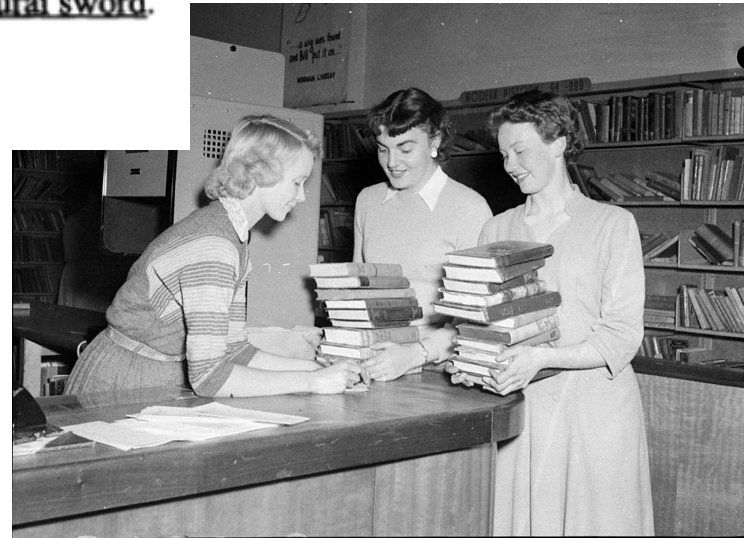
1. Ambiguous (underspecified) questions
2. Guessing
3. **Additional evidence**



Information-seeking Dialogue

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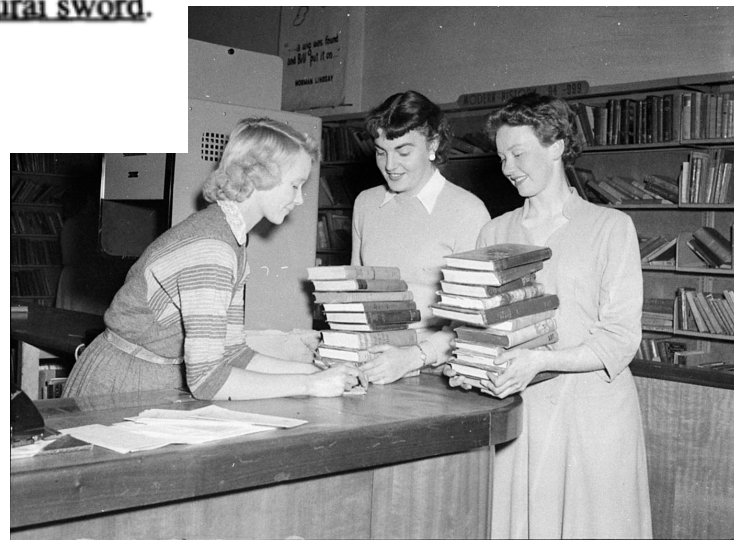
1. Ambiguous (underspecified) questions
2. Guessing
3. Additional evidence
4. **Common knowledge**



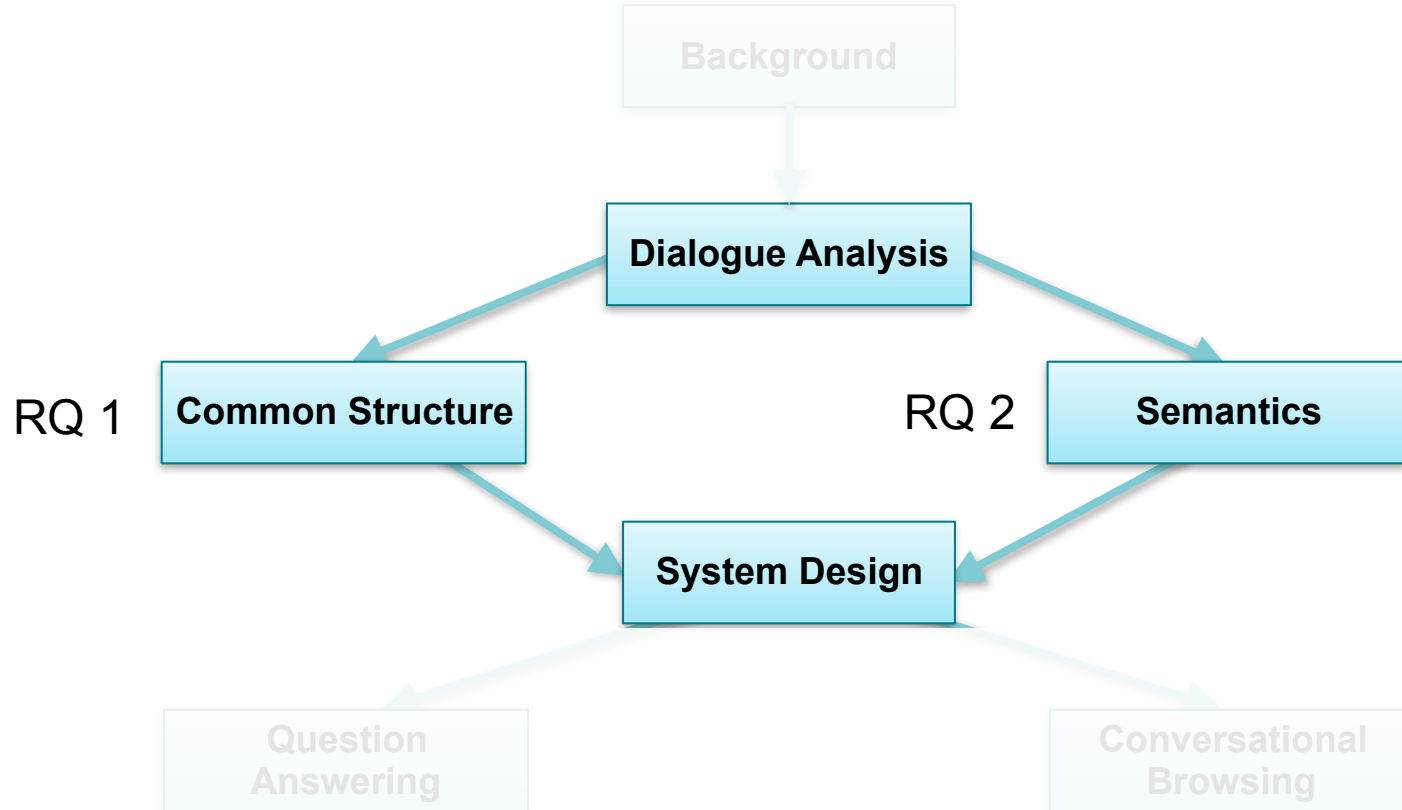
Information-seeking Dialogue

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1. Ambiguous (underspecified) questions
2. Guessing
3. Additional evidence
4. Common knowledge
- 5. Feedback**

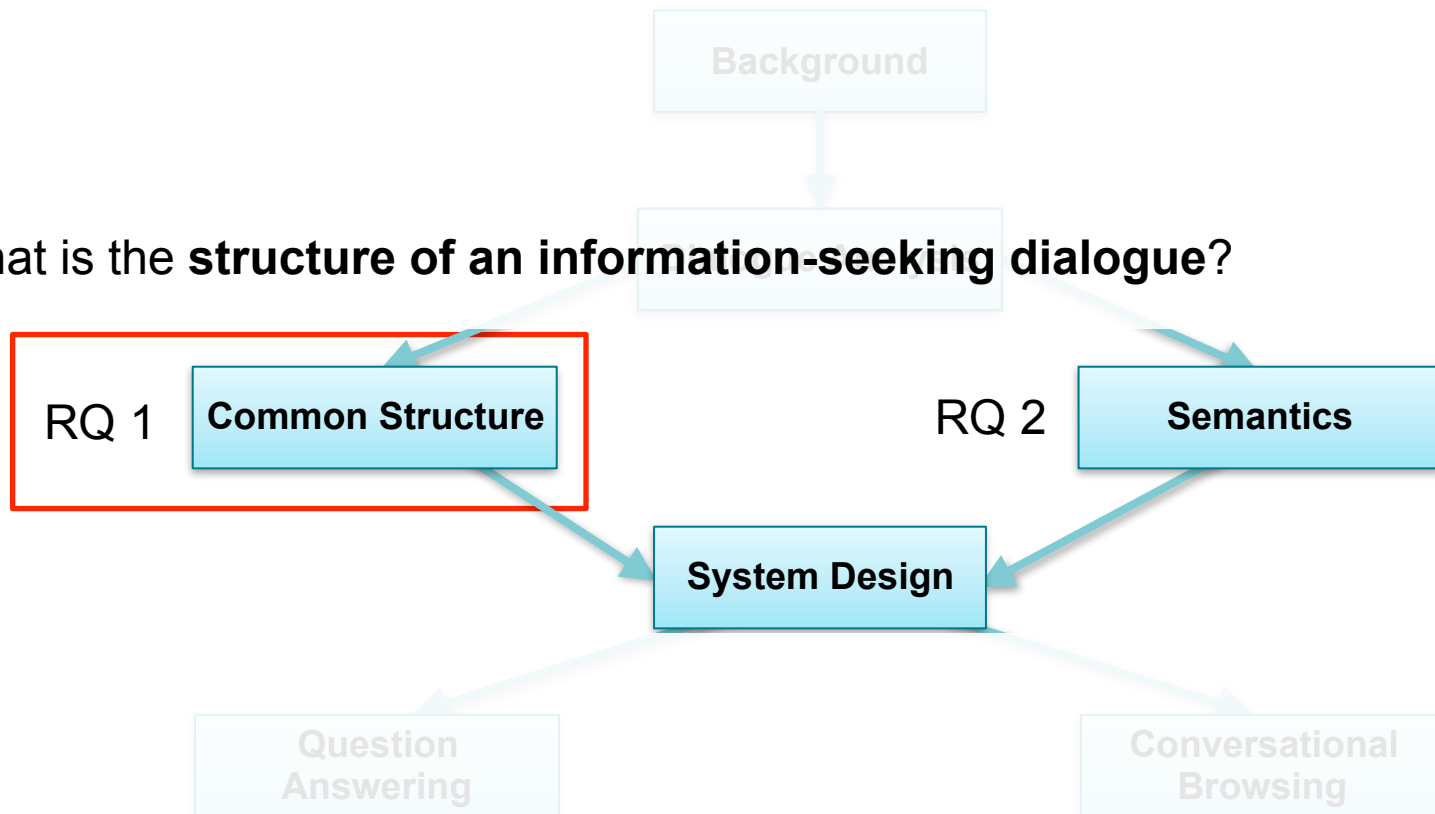


Outline

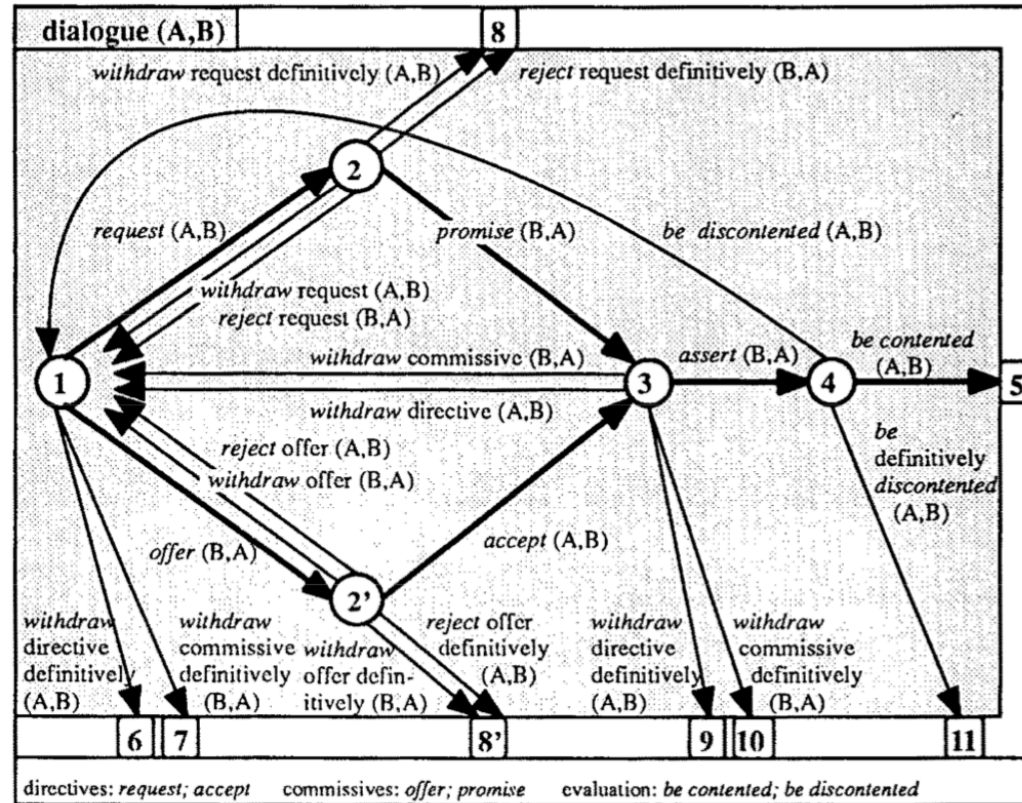


Outline

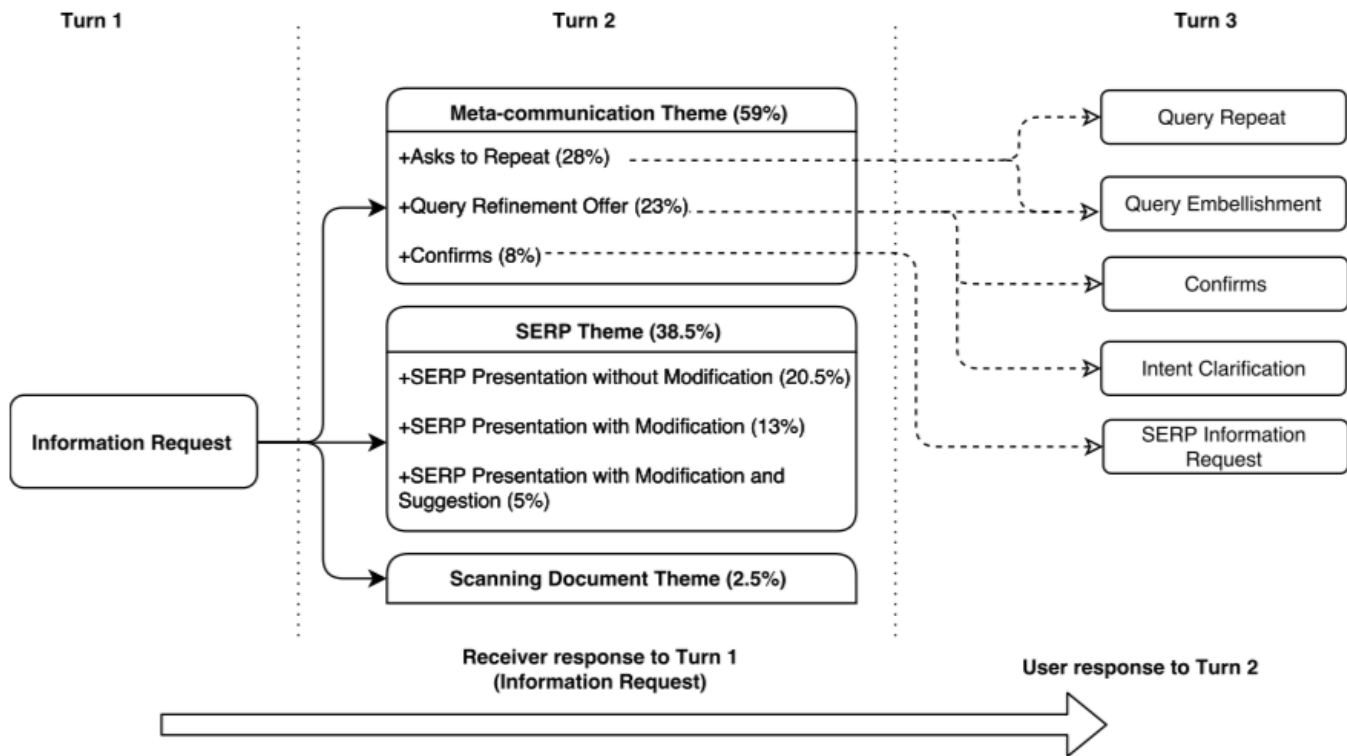
What is the **structure of an information-seeking dialogue**?



COnversational Roles model



Spoken Conversational Search model



Conversation Mining

Process Mining

- system log (workflow) analysis
 - P - process
 - T - trace, where $T \in P$
 - $T = \langle e_1, e_2, e_3 \dots \rangle$, where e - event and $e \in E$

Process Mining

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P_1

$T_1 : e_1 e_2 e_3$

$T_2 : e_2 e_1 e_3$

$T_3 : e_1 e_3$

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P_1

$T_1 : e_1 \ e_2 \ e_3$ e_1 - place order

$T_2 : e_2 \ e_1 \ e_3$ e_2 - send request

$T_3 : e_1 \ e_3$ e_3 - deliver order

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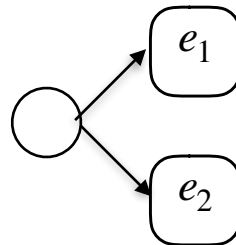
$T_2 : e_2 e_1 e_3$

$T_3 : e_1 e_3$

e_1 - place order

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$T_1 : e_1 e_2 e_3$

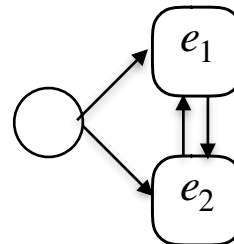
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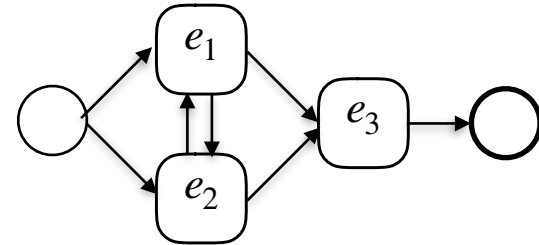
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Conversation Mining

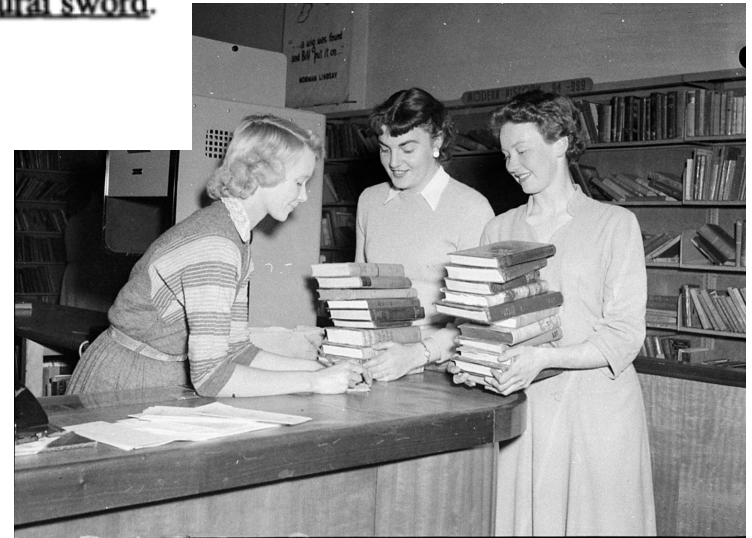
- conversation analysis
 - P - communication process
 - T - dialogue transcript, where $T \in P$
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Conversation Mining

- conversation analysis
 - P - communication process
 - T - dialogue transcript, where $T \in P$
 - $T = \langle u_1, u_2, u_3 \dots \rangle$, where u - utterance and $u \in U$
 - utterance labeling: $U \rightarrow L$, where $|U| > |L|$
 - $T = \langle l_1, l_2, l_3 \dots \rangle$, where l - label and $l \in L$

Example

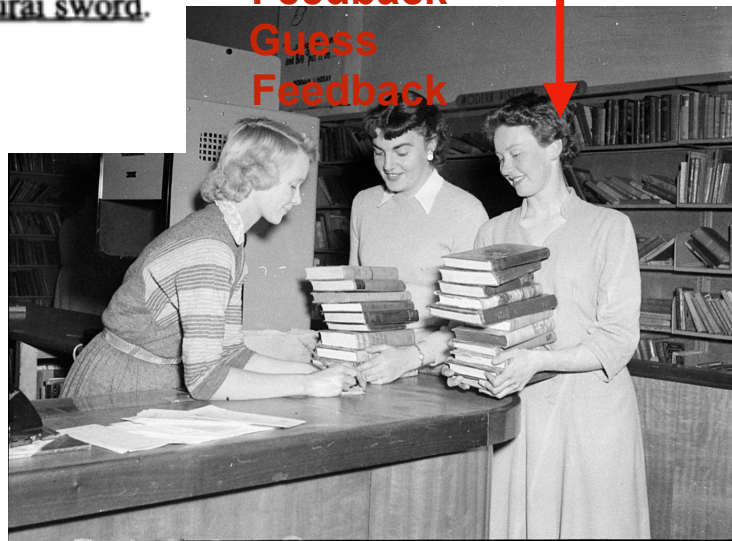
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Question
Guess
Hint
Guess
Feedback
Guess
Feedback



Datasets

Seeker (User) - Intermediary (Agent) - [Information Source]

- **Spoken Conversation Search** 39 dialogues 13 labels
- **Open Data Exploration** 26 dialogues 20 labels
- **DSTC1** bus schedules 15,866 dialogues 37 labels
- **DSTC2** restaurant reservation 2,118 dialogues 21 labels

Datasets

Seeker (User) - Intermediary (Agent) - [Information Source]

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human-human
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human-machine
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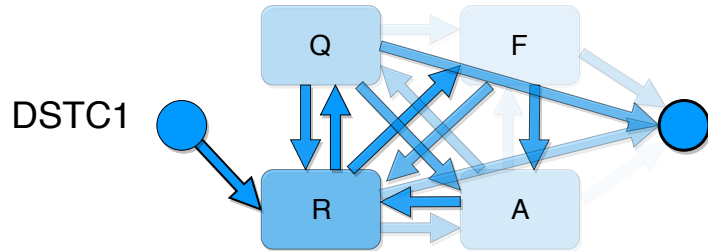
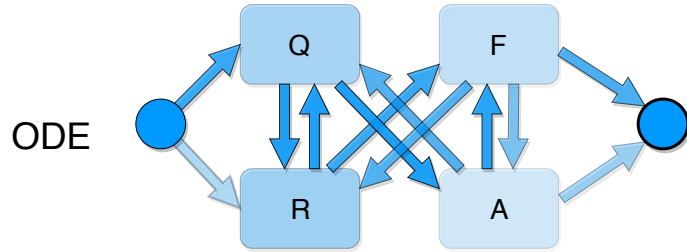
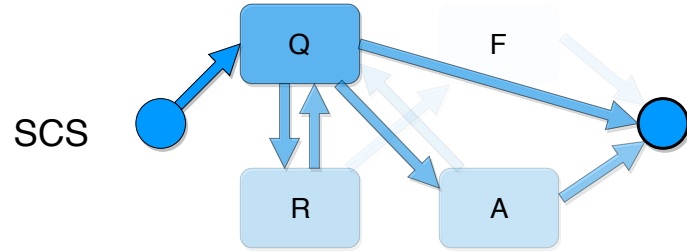
QRFA Annotation Schema

	Proactive	
<i>User</i>	Query	Information Prompt
<i>Agent</i>	Request	Offer Understand

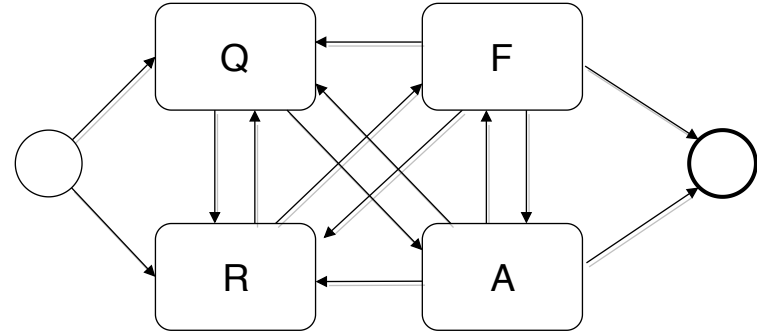
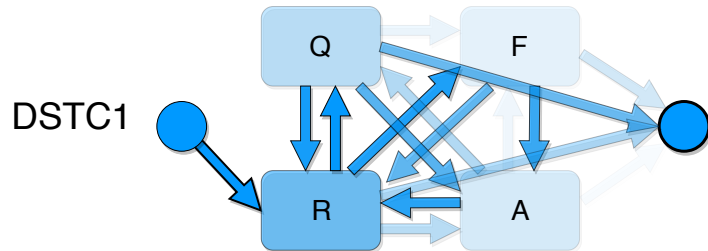
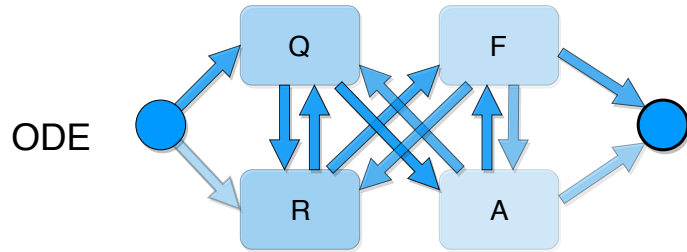
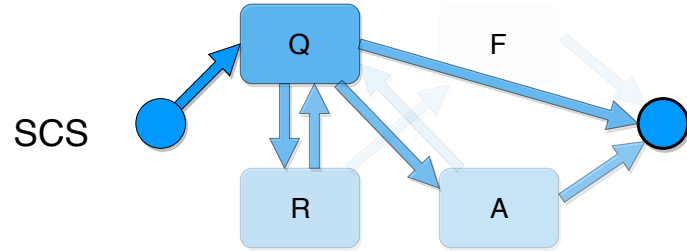
QRFA Annotation Schema

	Proactive		Reactive	
<i>User</i>	Query	Information Prompt	Feedback	Positive Negative
<i>Agent</i>	Request	Offer Understand	Answer	Results Backchannel Empty

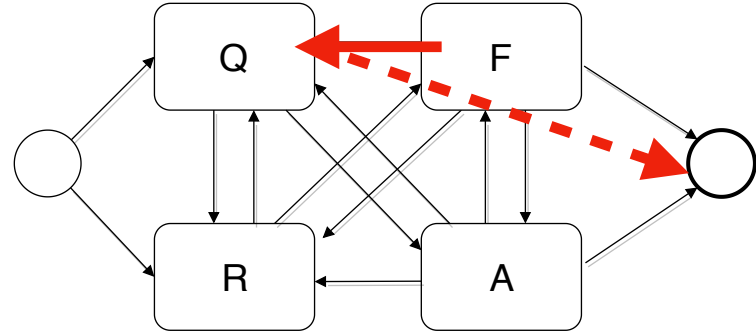
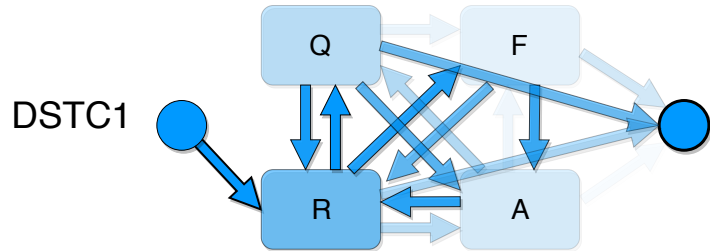
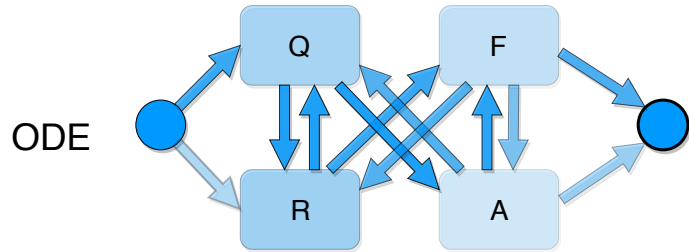
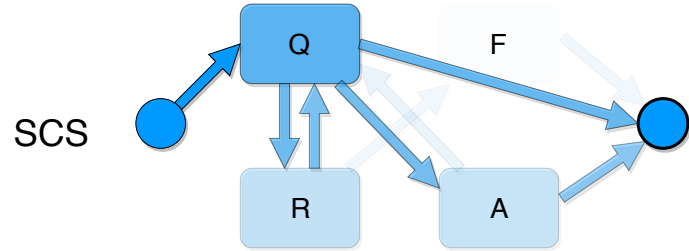
QRFA Model



QRFA Model

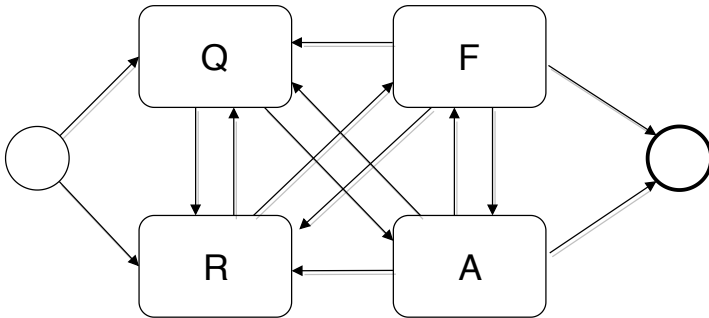


QRFA Model



Model Evaluation

- **Simplicity**, fitness, generalisation

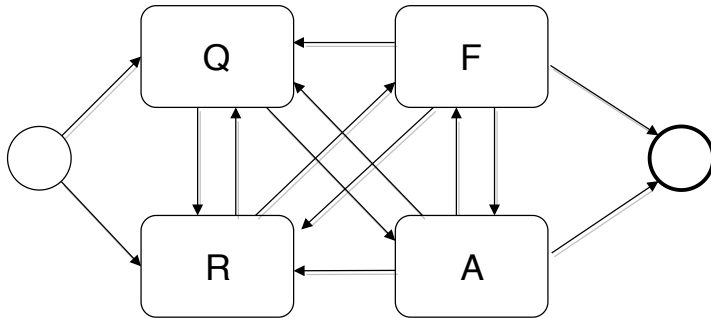


Model Fitness

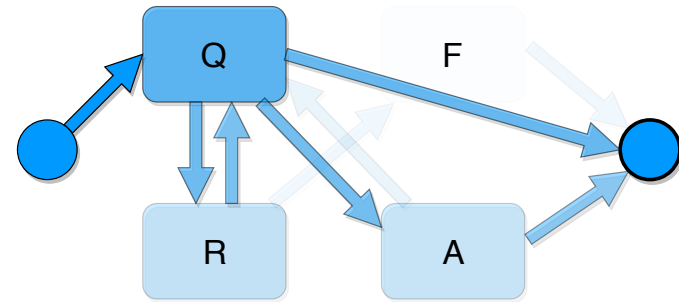
- **Fitness**/dialogue: QRFA Min 0.65 Max 1 Average 0.96 SD 0.04

COR Min 0.38 Max 0.93 Average 0.67 SD 0.1

- Dialogues with **fitness=1**: QRFA 0.61 COR 0.01



SCS

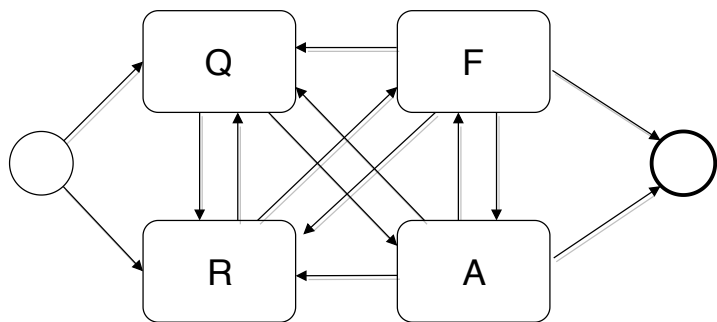


Model Generalisation

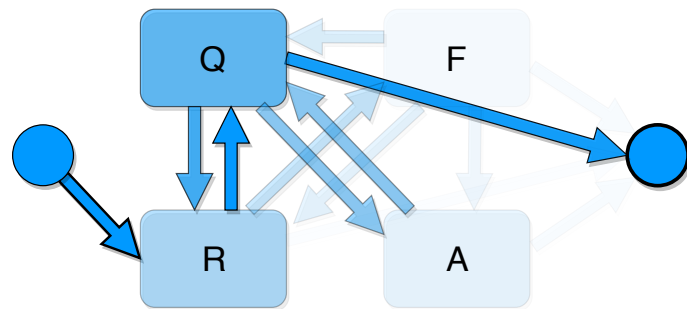
- **Fitness**/dialogue: QRFA Min 0.8 Max 1 Average 0.99 SD 0.02

COR Min 0.53 Max 0.91 Average 0.7 SD 0.05

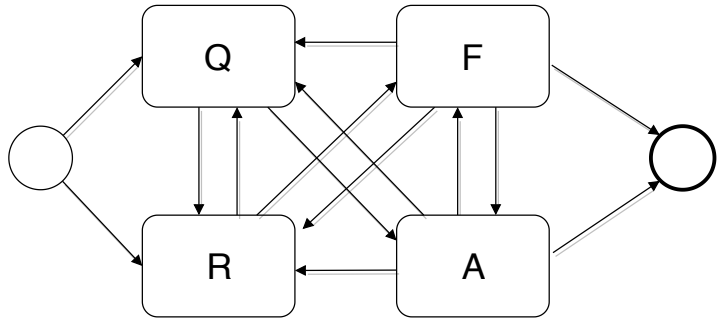
- **Break-down detection**: P 0.92 R 0.55



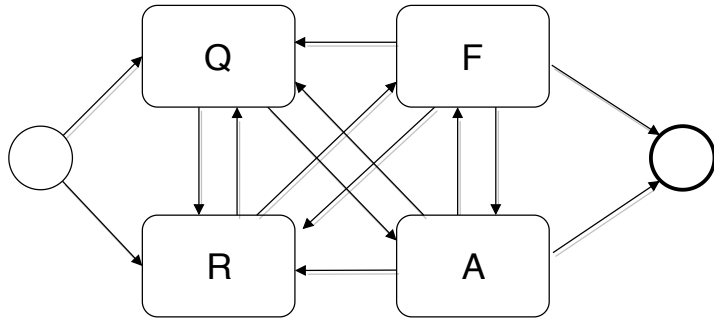
DSTC2



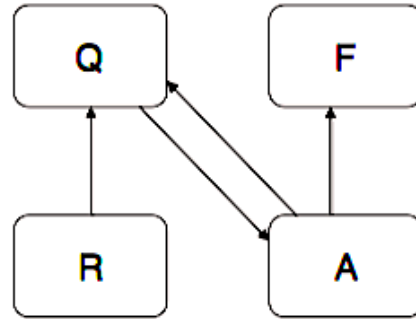
Interaction Modes



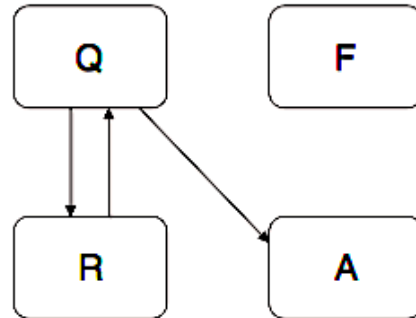
Interaction Modes



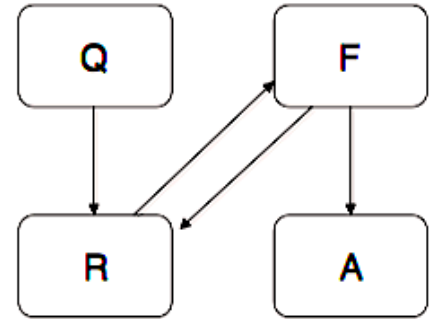
(a) Question Answering



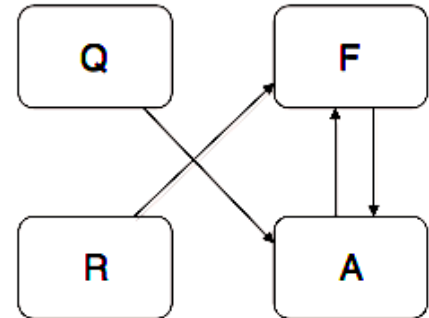
(b) Query Refinement



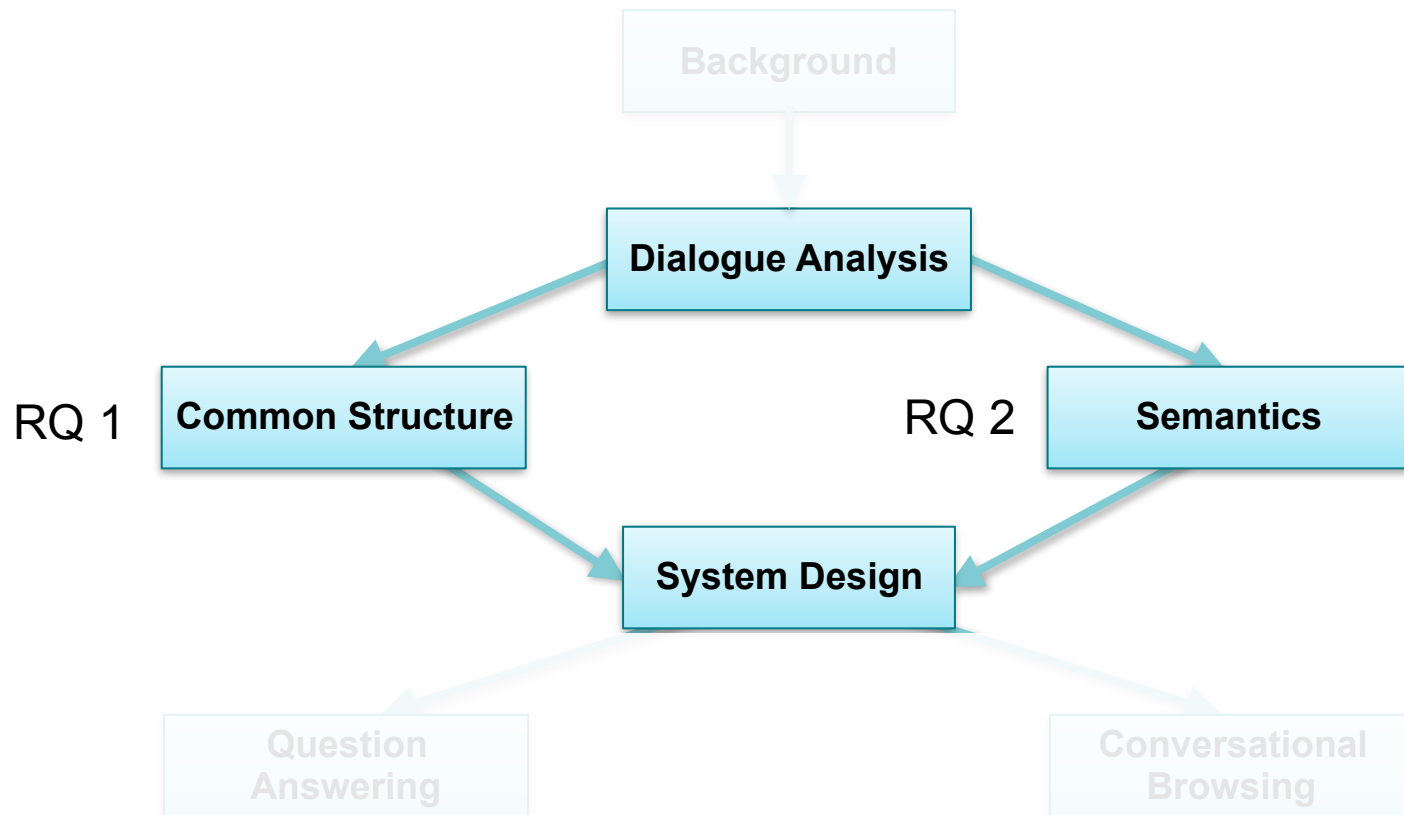
(c) Offer Refinement



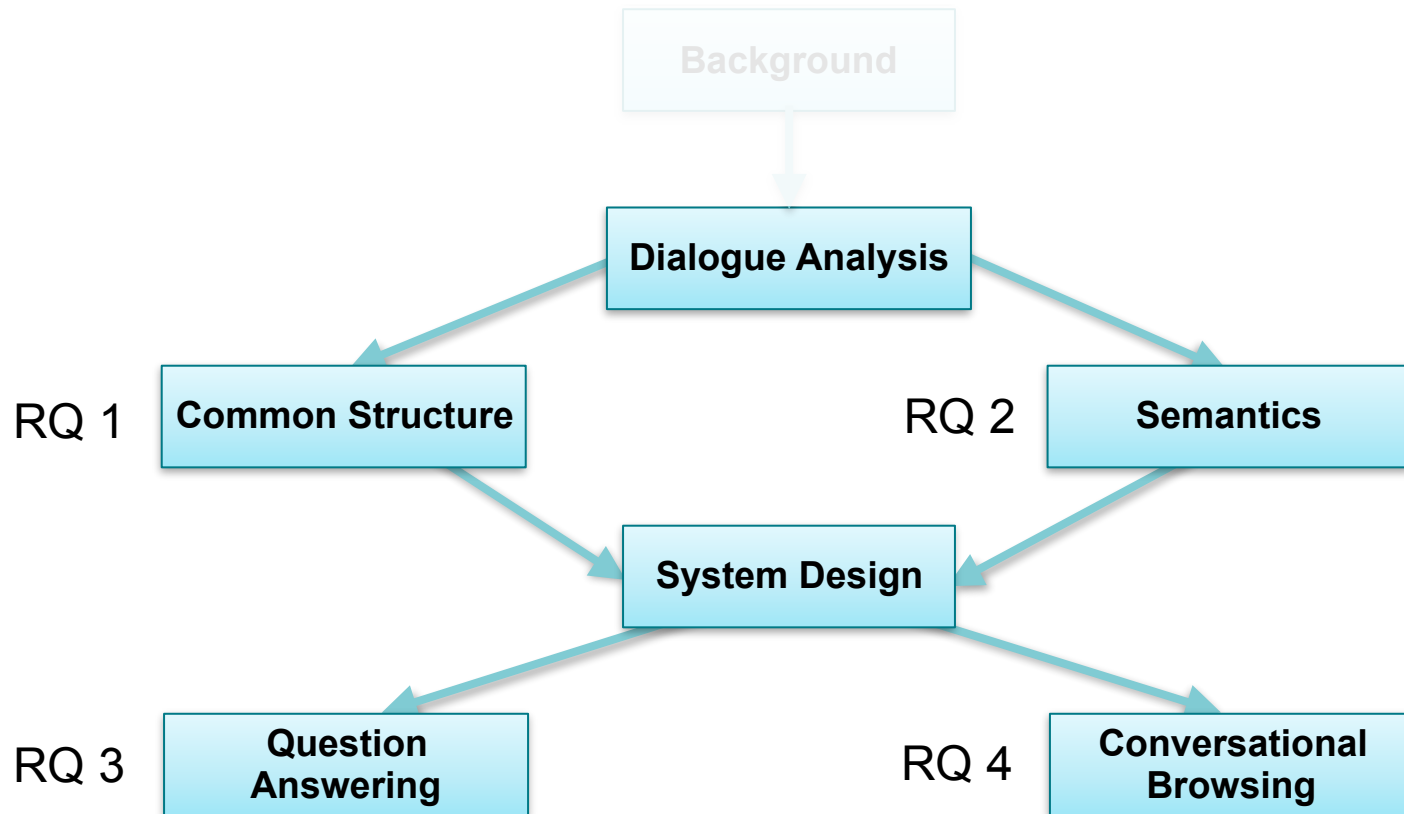
(d) Answer Refinement



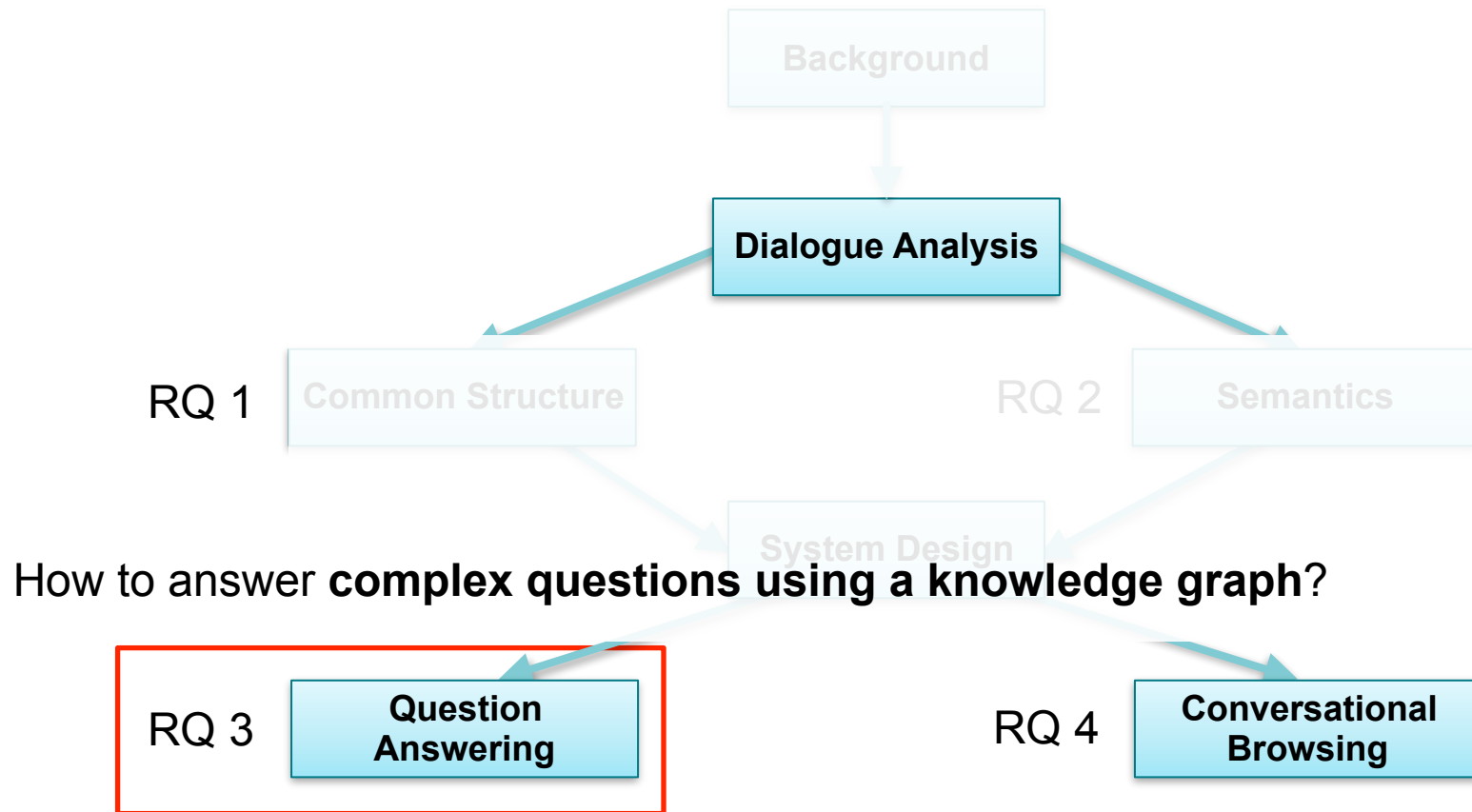
Outline



Outline

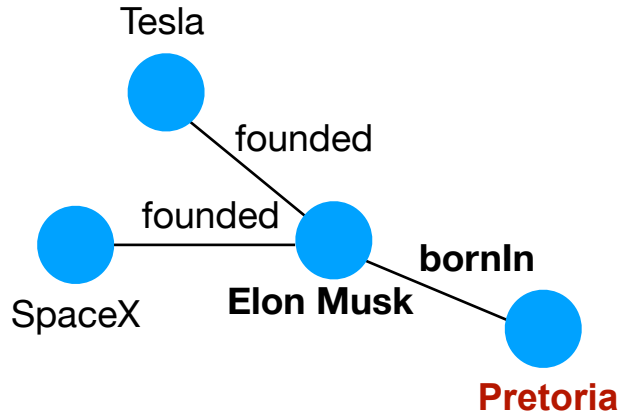


Outline



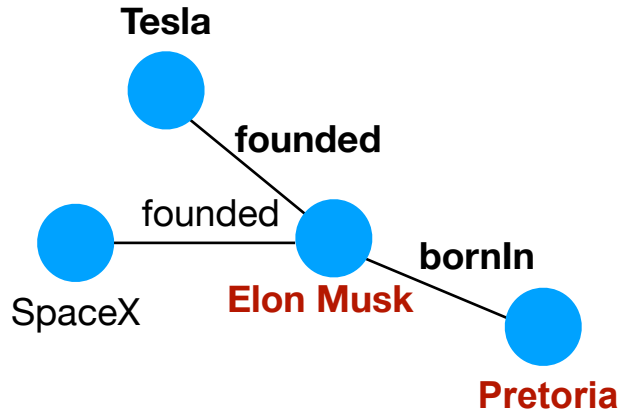
Simple Question Answering

Where is the Elon Musk born?



Complex Question Answering

Where is the founder of Tesla born?

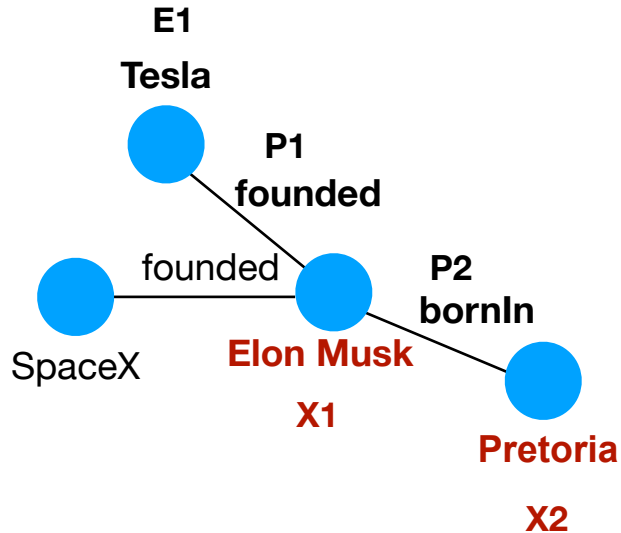


Message Passing

Parsing

Where is the founder of Tesla born?

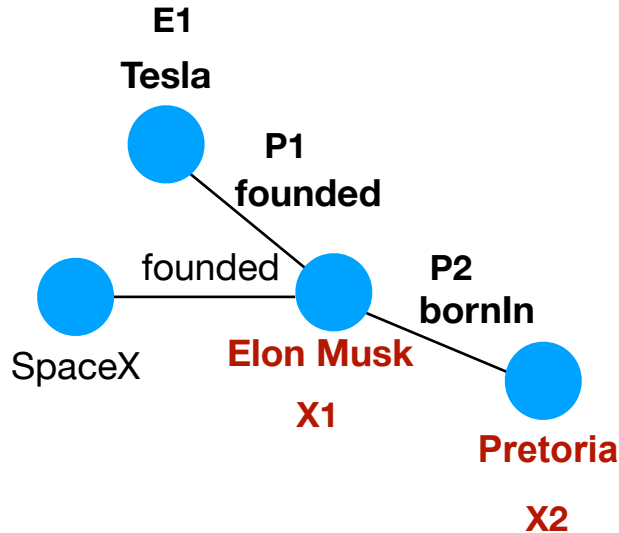
P1 E1 P2



Matching

Where is the founder of Tesla born?

P1 E1 P2



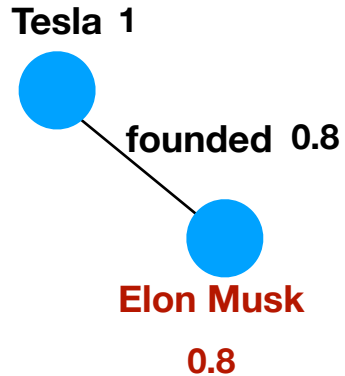
P1		E1	
founder	1	Tesla	1
founded	0.8	Nicola Tesla	0.6
		Tesla coil	0.5
P2			
bornIn	0.8		

Message Passing I

Where is the founder of Tesla born?

P1 E1 P2

P1		E1	
founder	1	Tesla	1
founded	0.8	Nicola Tesla	0.6
		Tesla coil	0.5



Message Passing II

Where is the founder of Tesla born?

P1

E1

P2

P2

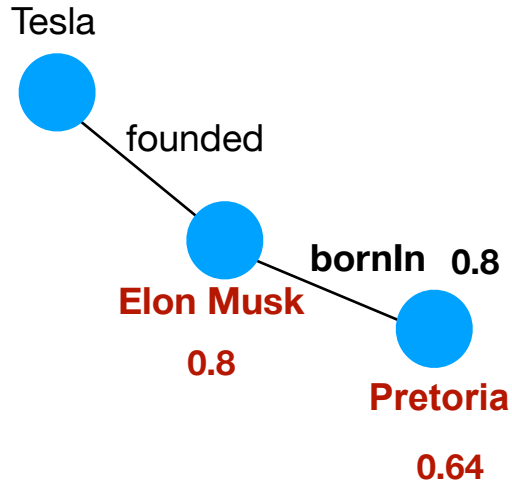
bornIn

0.8

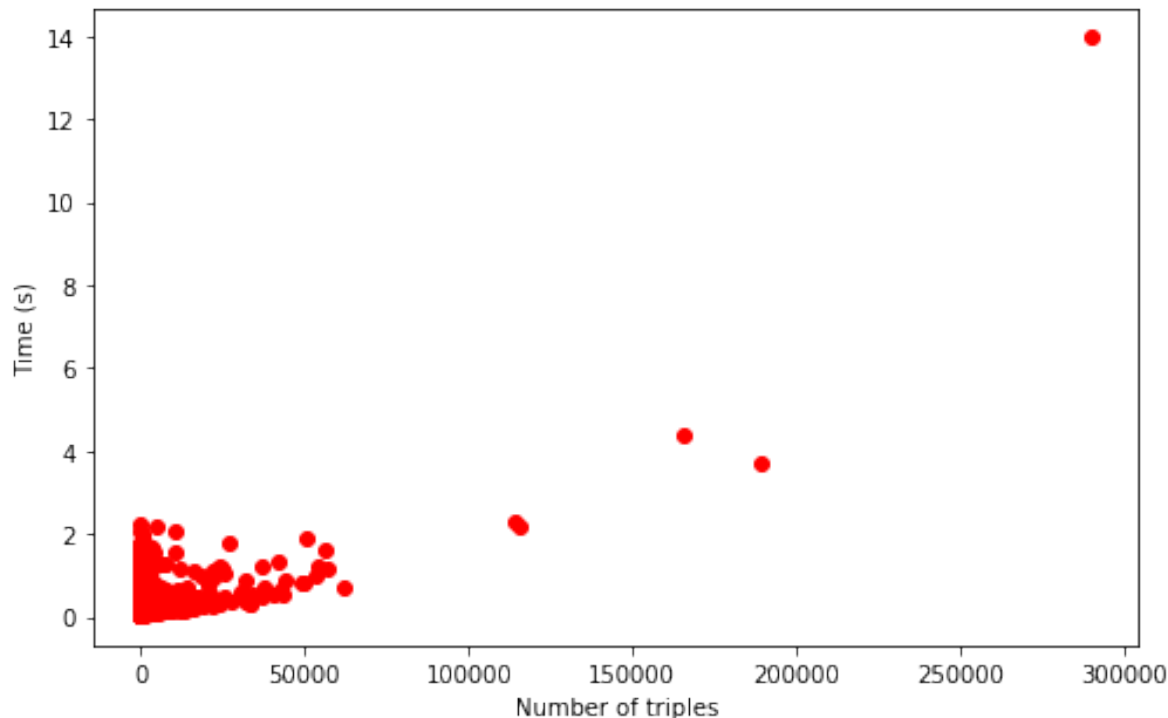
E2

Elon Musk

0.8



Results: Scalability



Processing times per question from the LC-QuAD test split
(Min: 0.01s. Median: 0.68s. Mean: 0.72s. Max: 13.97s)

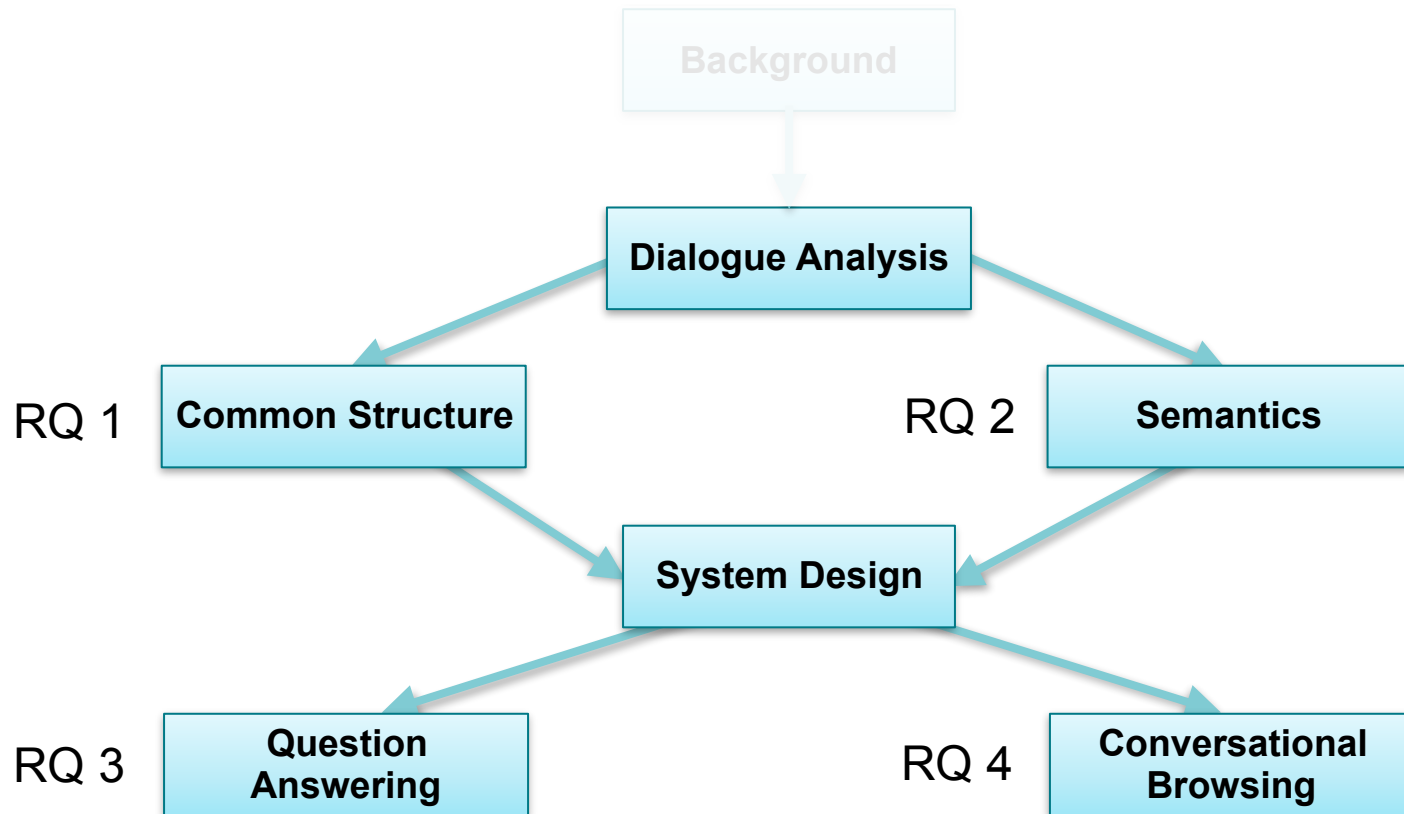
<http://www.rdfhdt.org>

Results: Performance

Approach	<i>P</i>	<i>R</i>	<i>F</i>	<i>Runtime</i>
<i>Baseline</i>	0.22	0.38	0.28	1.50s/q
<i>Us</i>	0.25	0.50	0.33	0.72s/q

Baseline: WDAqua, template-based SPARQL generator

Outline



List of Publications

Vakulenko, Svitlana, et al. "QRFA: A Data-Driven Model of Information Seeking Dialogues. *ECIR*. 2019. Best User Paper Award

Vakulenko, Svitlana, et al. "Measuring semantic coherence of a conversation." *ISWC*. 2018. Spotlight paper

Vakulenko, Svitlana, et al. "Message Passing for Complex Question Answering over Knowledge Graphs." *CIKM*. 2019.

Vakulenko, Svitlana, et al. "Conversational browsing: Dialog-based Access to Structured Information Sources." *Under review*. 2019.

List of Publications

1. Svitlana Vakulenko, Javier D. Fernandez, Axel Polleres, Maarten de Rijke and Michael Cochez. Message Passing for Complex Question Answering over Knowledge Graphs. CIKM. 2019.
2. Svitlana Vakulenko, Vadim Savenkov and Maarten de Rijke. Conversational browsing: Dialog-based Access to Structured Information Sources. (Under review)
3. Sophia Keyner, Vadim Savenkov and Svitlana Vakulenko. Open Data Chatbot. ESWC 2019.
4. Svitlana Vakulenko, Kate Revoreda, Claudio Di Ciccio and Maarten de Rijke. QRFA: A Data-Driven Model of Information Seeking Dialogues. ECIR. 2019. *Best paper (User track)*.
5. Svitlana Vakulenko, Maarten de Rijke, Michael Cochez, Vadim Savenkov and Axel Polleres. Measuring Semantic Coherence of a Conversation. ISWC. 2018. *Spotlight paper*.
6. Svitlana Vakulenko, Ilya Markov and Maarten de Rijke. Conversational Exploratory Search via Interactive Storytelling. SCAI@ICTIR. 2017.
7. Svitlana Vakulenko and Vadim Savenkov. TableQA: Question Answering on Tabular Data. SEMANTICS. 2017. *Best poster and demo nomination*.
8. Sebastian Neumaier, Vadim Savenkov and Svitlana Vakulenko. Talking Open Data. ESWC. 2017.
9. Kathrin Figl, Samuel Kielsing, Christiane Rank, Svitlana Vakulenko. The Effect of Fake News Flags on the Believability of Social Media Posts. ICIS. 2019.
10. Svitlana Vakulenko, Lyndon Nixon, and Mihai Lupu. Character-based Neural Embeddings for Tweet Clustering. SocialNLP@EACL. 2017.

11. Albert Weichselbraun, Stefan Gindl, Fabian Fischer, Svitlana Vakulenko, and Arno Scharl. Aspect-based extraction and analysis of affective knowledge from social media streams. In IEEE Intelligent Systems. 2017.
12. Svitlana Vakulenko, Albert Weichselbraun, and Arno Scharl. Detection of valid sentiment-target pairs in online product reviews and news media articles. WI. 2016.
13. Svitlana Vakulenko, Max Gobel, Arno Scharl and Lyndon Nixon. Know which way the Wind Blows: Visualising the Propagation of News on the Web. NewsIR@ECIR. 2016.
14. Gregory Katsios, Svitlana Vakulenko, Anastasia Krithara and Georgios Paliouras. Towards Open Domain Event Extraction from Twitter: REVEALing Entity Relations. DERIVE@ESWC. 2015.
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Contributions

- **Conversation mining** *methodology* + **QRFA** *data-driven model*
- **Semantic coherence** as a classification *task* + *experimental evaluation*
- **Message-passing** *algorithm* for complex question answering from KGs
- *Task, dataset, prototype* + *user studies* for **conversational browsing**

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