

Towards Mixed-Initiative Conversational Information Seeking

Hamed Zamani

Center for Intelligent Information Retrieval
University of Massachusetts Amherst

zamani@cs.umass.edu



@HamedZamani

Why Conversational Information Seeking?



Recent advances in automatic speech recognition



Popularity of devices with limited bandwidth interfaces

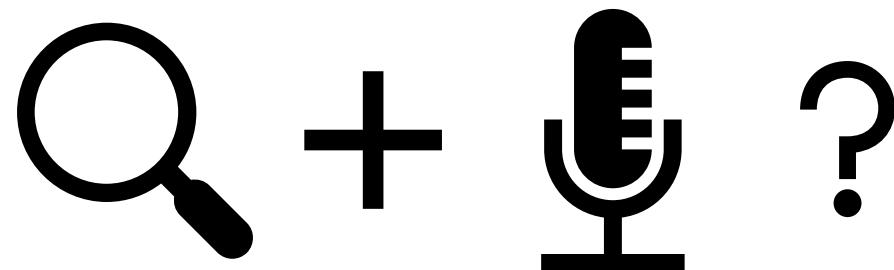


Recent advances in neural models



Information access is an important service.

What is conversational search?



Is This Conversational Search?

Mitsuku, how did you become a good conversationalist?

Yep, Do you really think I am a good conversationalist?
Mousebreaker programmed me to be a good conversationalist.

How are you programmed?

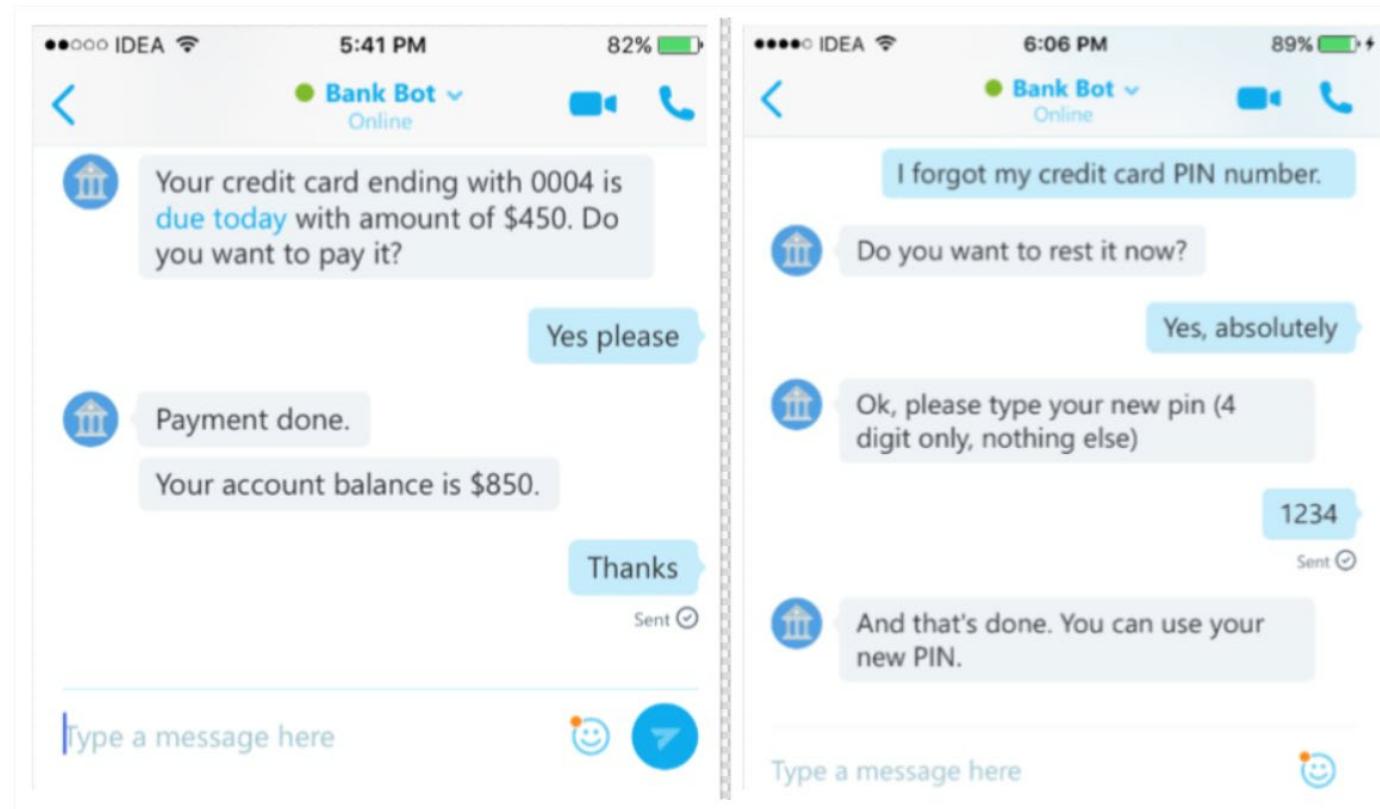
I read what you say and then I compose the best reply I can think of.

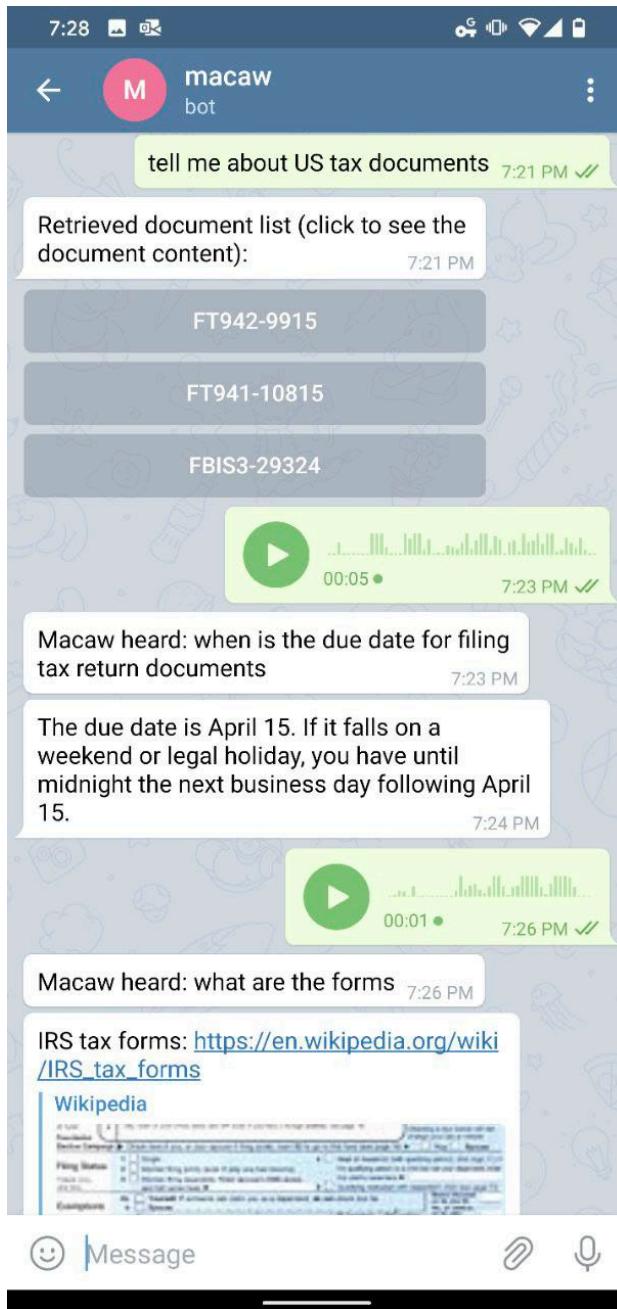
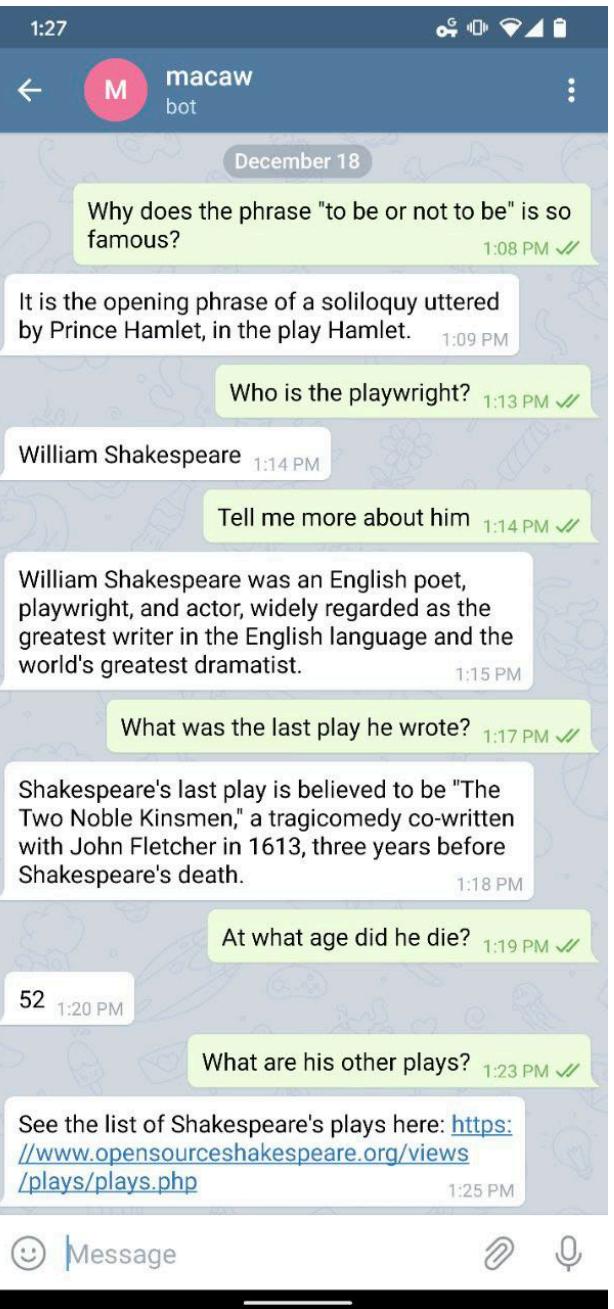
How does that actually work though?

Oh. I think it might be a trade secret.

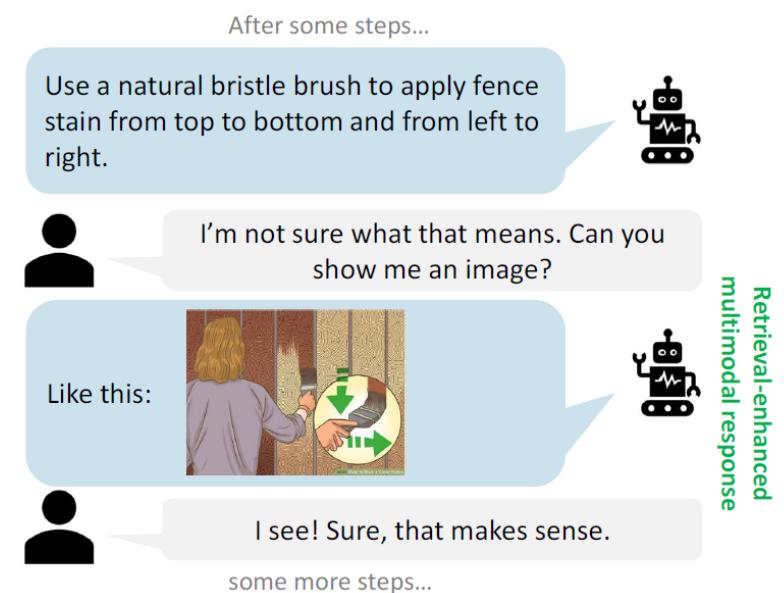
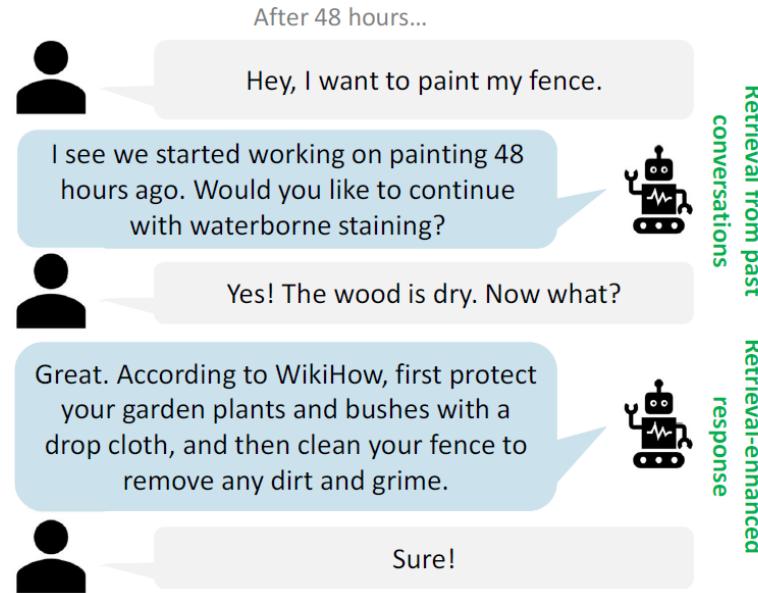
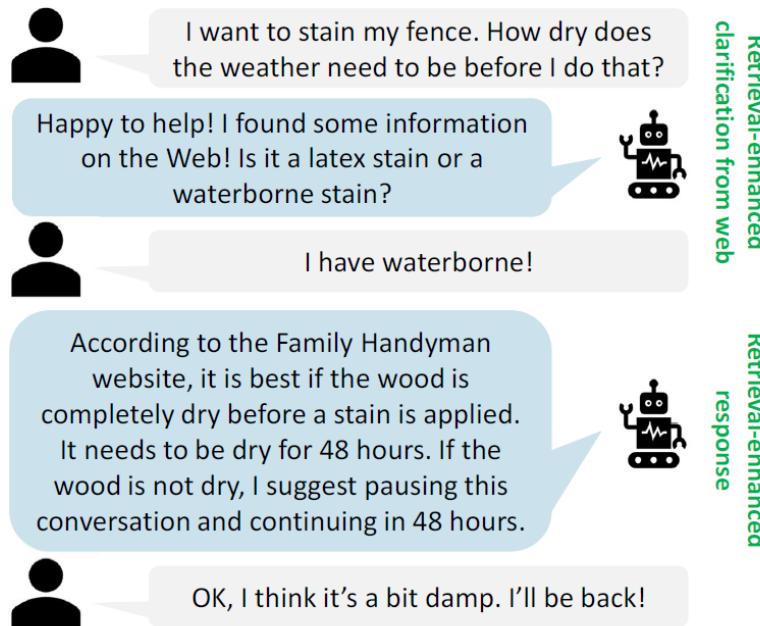
You're so sassy 😊

Is This Conversational Search?





Conversations can be complex!



Definitions

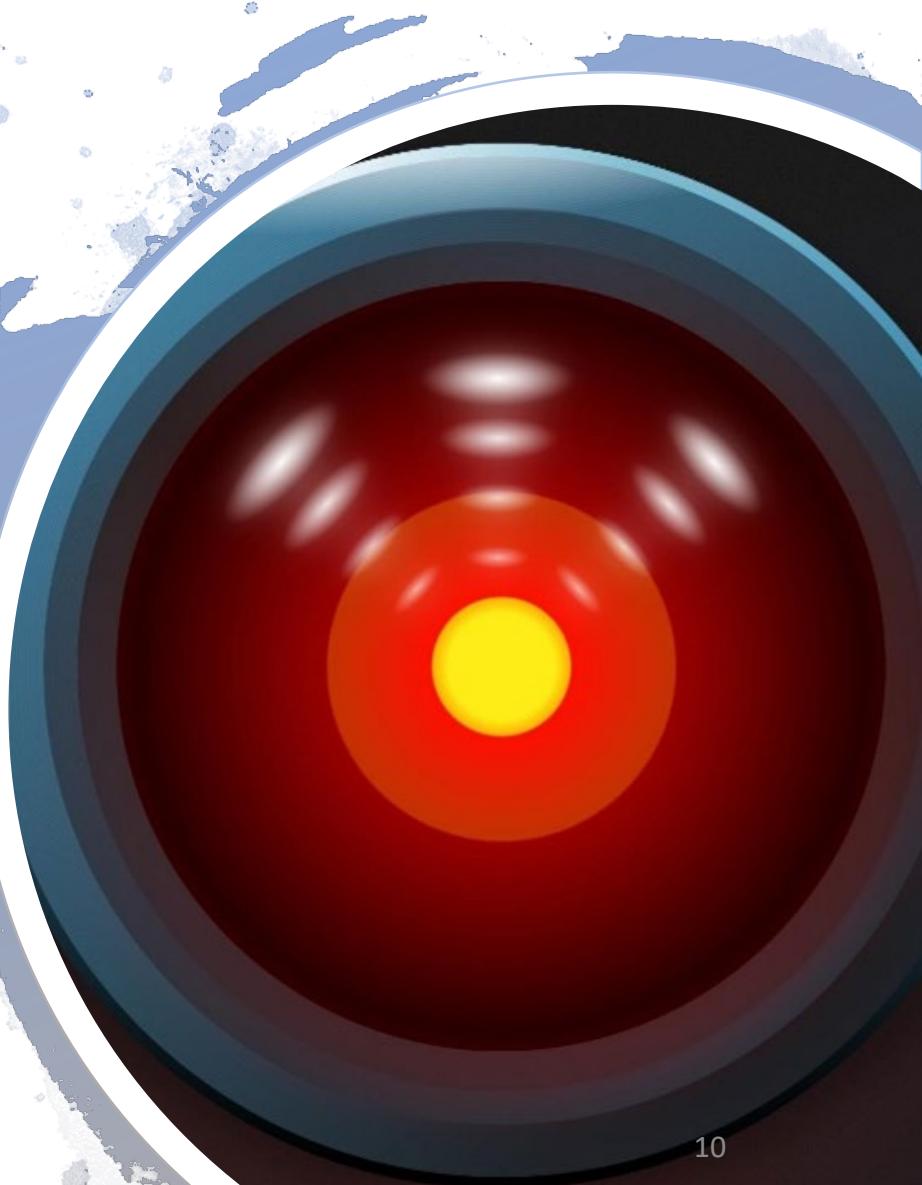
Conversation is **interactive** communication for **exchanging information** between two or more participants (i.e., humans or machines) that involves a **sequence** of interactions. While **natural language** is considered as prerequisite for conversational interactions, conversations can also exhibit **other types of interaction** with different characteristics and modalities (e.g., click, touch, and gestures).

Definitions

Information seeking conversation is a **conversation** in which the goal of information exchange is **satisfying the information needs** of one or more participants.

A **Conversational Information Seeking** (CIS) system is a system that satisfies the information needs of one or more users by engaging in **information seeking conversations**. CIS responses are expected to be concise, fluent, stateful, **mixed-initiative**, context-aware, and personalized.

Conversational Search has
roots in early IR research!



Five general types of information necessary to get the desired state:

- Determination of subject
- Objective and motivation
- Personal characteristics of the inquirer
- Relationship of inquiry description to file organization
- Anticipated or acceptable answers

“Question Negotiation and Information
Seeking in Libraries”, Robert Taylor, 1968.

Intermediary-based Online IR

Nature of interaction between user and intermediary, in both cognitive and affective senses, is a key factor in search satisfaction.

User-Intermediary Interactions

I Alright. Right. The form ... err, what we got on the form just

U

I says community education in developing countries. That's

U

Yeah

I approximately yeah/1/ Can you tell me, sort of quite a lot

U yeah

well (illeg.) /2/

I more about what it is you're going to do-/3a/ it is your disser-

U

I tation/3b/ Yeah yeah. Tell me what . sorts of

U Yeah, M.A. dissertation/4/

I things you're going to do in your dissertation,/5a/ and then the

U

I sorts of things you want to rea::d./5b/

U

That's my sort of, uhm, plan,

I = Intermediary U = User /n/ = utterance number

13

Figure 3. Example span of user-intermediary pre-search interaction.

User-Intermediary Interactions

VARIABLE	CODE
Problem Description function	PD
Mode: Ask	1
Offer	2
Task: Description of problem topic or clarification	a
Description of problem type or context	b
Description of subject area	c
Terminology	d
Literature	e
User Model function	UM
Mode: Ask	1
Offer	2
Task: Personal background (status)	a
Intentions/goals	b
Familiarity with information systems	c
Familiarity with topic	d
Knowledge resource codes	
Subject knowledge	SUB
System knowledge	SYS
Query language	1
Index language	2
Data base	3
Technical	4
Functions	5
Knowledge of users	US
Types	1
Characteristics	2
Intentions	3
General world knowledge	WK

The THOMAS System

THOMAS, THE REFERENCE RETRIEVAL PROGRAM

Help can be obtained whenever the program has displayed the start symbol by typing ‘?’ immediately after it.

Please give a short name for the search:

► Alv.Resp.

Start searching:

► pulmonary alveoli

Influence of fasting on blood gas tension, pH, and related values in dogs.;
Pickrell *et al*, *Am J Vet Res*, 34, 805-8, Jun 73

1. J A Pickrell, 2. JL Mauderly, 3. B A Muggenburg, 4. U C Luft, 5. animal experiments, 6. animal feed, 7. arteries, 8. blood, 9. body temperature, 10. carbon dioxide, 11. dogs, 12. fasting, 13. hemoglobin, 14. hydrogen-ion concentration, 15. irrigation, 16. lung, 17. oxygen, 18. pulmonary alveoli, 19. respiration, 20. time factors

► ?

The THOMAS System

There can be three parts to your statement (all optional):

1. Your reaction to the reference just shown (if any).

This must come first:

"Yes" or "No"

2. A selection from the names (authors) or terms shown, by number. A "not" in the statement signifies rejection of all numbers that follow it.

3. New names or terms (terms preferably in quotes). The elements of the statement should be separated by commas.

Examples: 'posture', 'circulatory system'

Yes, not 11,12

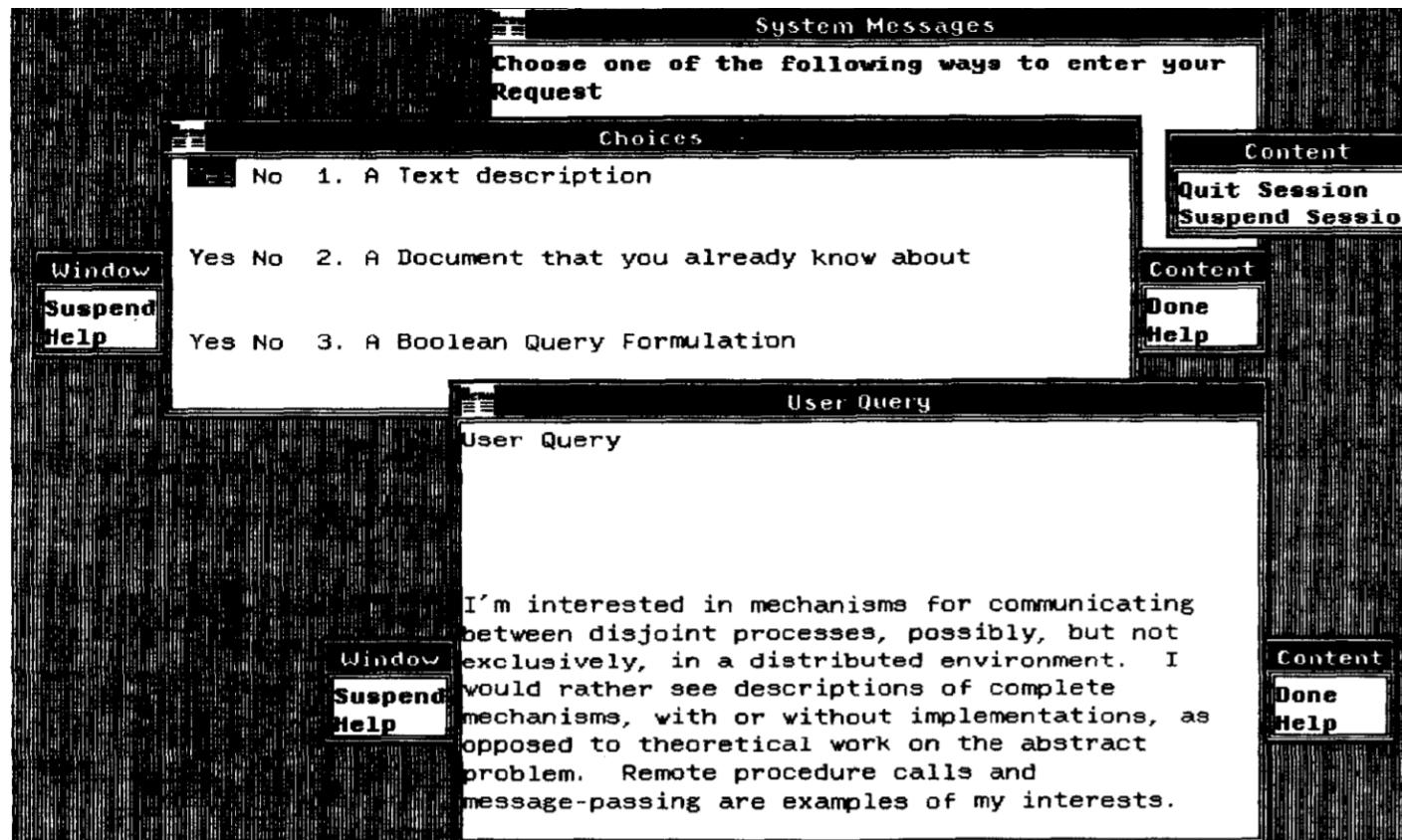
No, 7,13,4

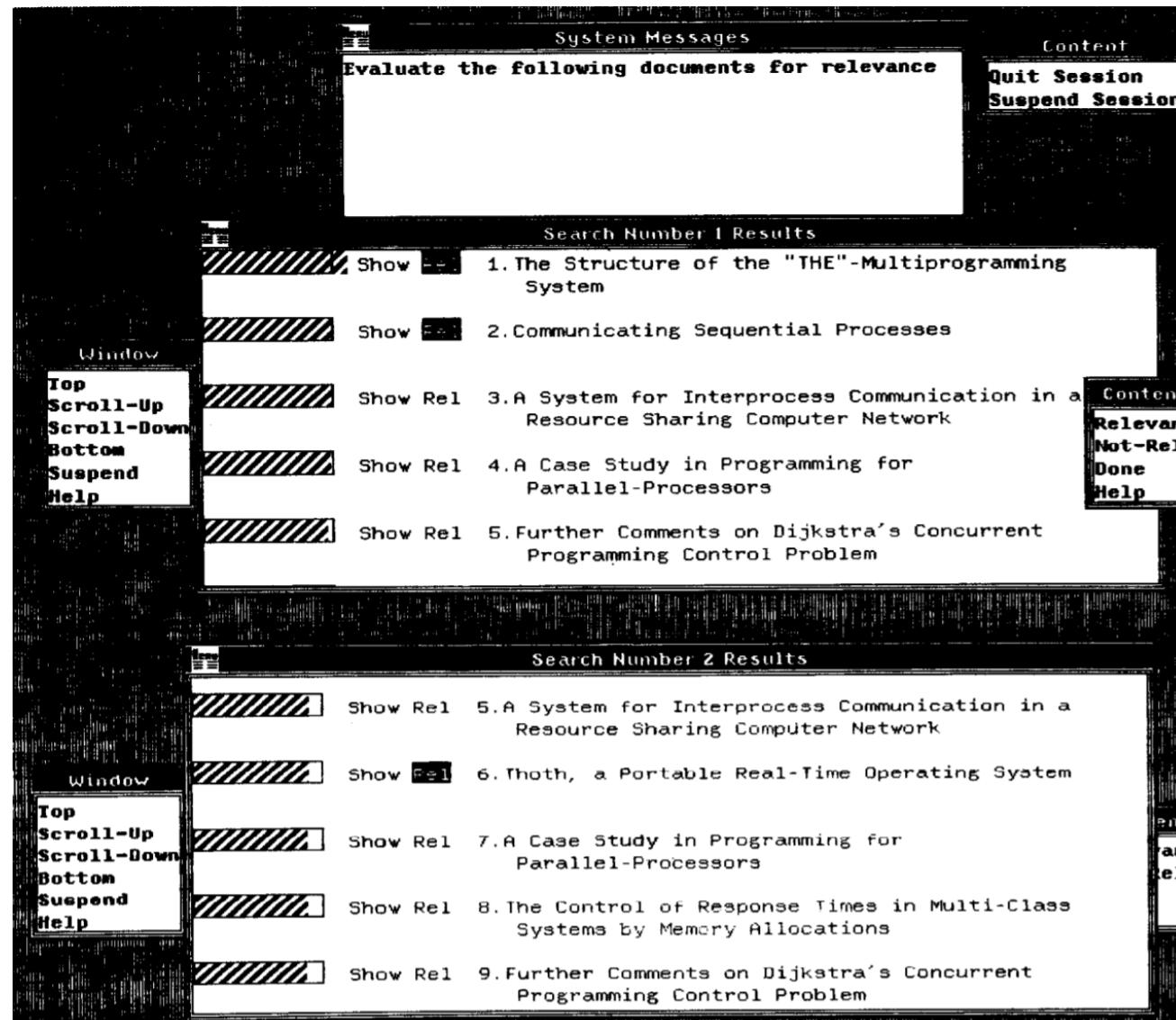
'heart rate'

Yes

Press enter key when you are ready to proceed ►

The I³R System





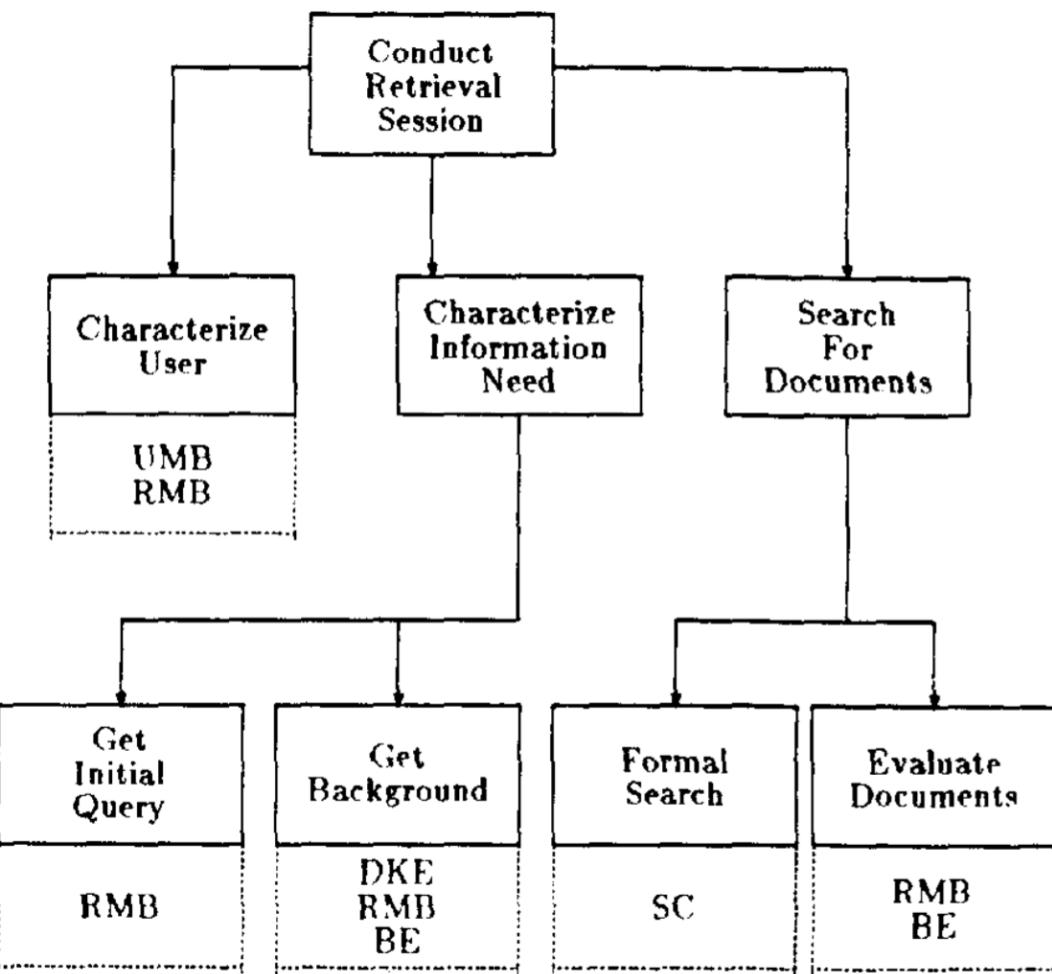


FIG. 3. Default scheduler plan.

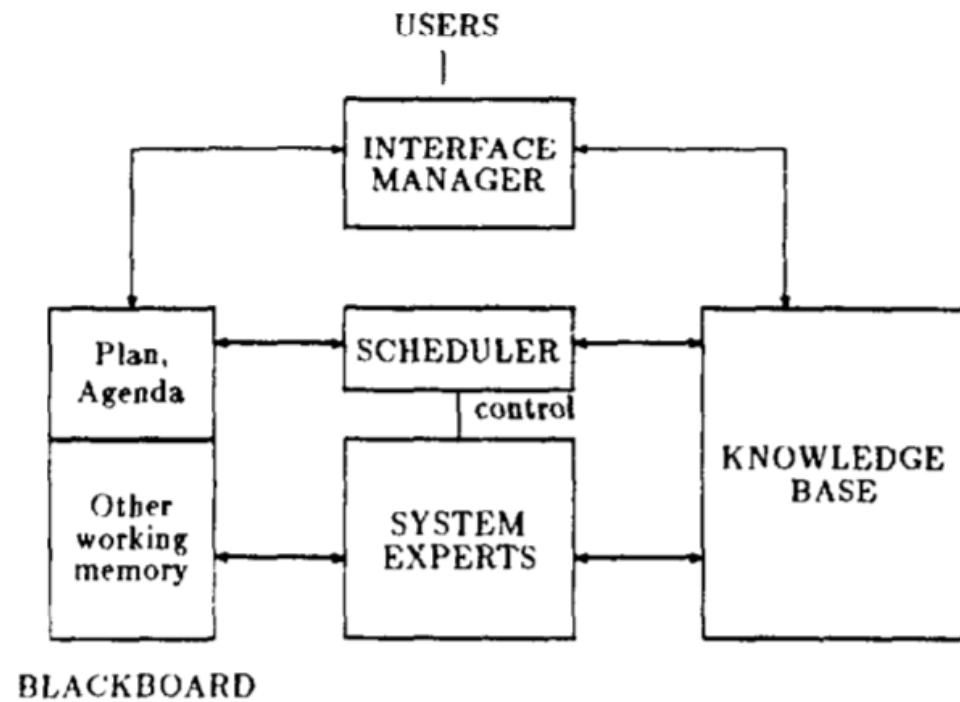


FIG. 2. Major system components.

More Recently

- TREC CAsT: TREC Conversational Assistance Track
- CoQA: Conversational Question Answering
- QuAC: Question Answering in Context
- ShARC: Shaping Answers with Rules through Conversation
- ConvQuestions: Conversational Question Answering over Knowledge Graphs
- ...

Section: Augusto Pinochet : Intellectual life...

STUDENT: Was he known for being intelligent?

TEACHER: ↗ No, Pinochet was publicly known as a man with a lack of culture.

STUDENT: why did people feel that way?

TEACHER: ↗ reinforced by the fact that he also portrayed himself as a common man

STUDENT: did he have any hobbies?

TEACHER: ↗ Yes, Before wresting power from Allende, Pinochet had written two books.

STUDENT: what is the name of a book written by him?

TEACHER: ↗ Geopolitica (1968) and Campana de Tarapaca (1972).

STUDENT: what were the books about?

TEACHER: ↗ Chile's military literature.

STUDENT: was there anything noteworthy regarding his books?

TEACHER: ↗ Yes, In Geopolitica Pinochet plagiarized (...) Gregorio Rodriguez Tascon

STUDENT: did he deny those allegations?

TEACHER: ↗ No answer

STUDENT: what did he plagiarize in Geopolitica?

TEACHER: ↗ In Geopolitica Pinochet plagiarized (...) paragraphs from a 1949 presentation

...

Mixed-Initiative Interactions in Human Conversation

MIXED-INITIATIVE LEVELS	CAPABILITIES
Unsolicited reporting	Agent may notify others of critical information as it arises
Subdialogue initiation	Agent may initiate subdialogues to clarify, correct, and so on
Fixed subtask initiative	Agent takes initiative to solve predefined subtasks
Negotiated mixed initiative	Agents coordinate and negotiate with other agents to determine initiative

ACTION	AMOUNT (%)
Evaluating and comparing options	25
Suggesting courses of action	23
Clarifying and establishing state	13.5
Clarifying or confirming the communication	13.5
Discussing problem-solving strategy	10
Summarizing courses of action	8
Identifying problems and alternatives	7

Clarification

- Confirming or clarifying the communication
 - For example, to reduce ASR error.
- Correcting user mistakes
 - Did you mean ...?
- **Intent clarification**

Search Clarification



blue screen



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What version of Windows are

windows 10 windows 8 wind

How to Fix a Blue Screen of D

[https://www.lifewire.com/how-to-fix-a-blue-](https://www.lifewire.com/how-to-fix-a-blue-screen-of-death-2615201)

1. The most important Blue Screen of Death t
2. Check that there's enough free space left o
3. Scan your computer for viruses. Some viru
4. Apply all available Windows service packs i

See all full list on lifewire.com

Videos of blue screen

bing.com/videos

20th anniversary gifts



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Who are you shopping

wife husband

See 20th anniversary



20th Wedding Anniversary...
\$30.00
Zazzle
15% Off code

Platinum Rose - 20th...
\$99.00
RomanceHer...
15% Off code

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<https://www.1800flowers.com>

(Ad) When It Matters Most, Trust 1

Anniversary Best Se

Our Most Popular Bouquets & A Perfect Gift For Someone Sp

nsfl



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Which "nsfl" do you mean?

not safe for life northern states

Urban Dictionary: NSFL

<https://www.urbandictionary.com/define>

NSFL also known as NLS means Not Safe some website has sent you meaning no or

nsfl.com - Nassau Suffolk F

<https://www.nsfl.com>

President - Mickey Kane Vice President - D

Kane Head Referee - Tom O'Hanlon

Town Rep Info · NSFL Documents · New Co

NSFL Football

nsfl.jcink.net/index.php

Welcome back! Your last visit was on Today

wedding dresses



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Do you have any theme in mind?

summer modern fall beach retro

See wedding dresses



2020 Wedding Dresses Ball...	Mermaid Long Sleeves V nec...	Ball Gown Strapless...	Ball-Gown Sweetheart...	A-Line Illusion Floor-Length...	Ball Gown Strapless...
\$169.00 Hebeos	\$217.00 Hebeos	\$179.99 LightInTheBox	\$207.00 JJHouse	\$156.00 JJHouse	\$179.99 LightInTheBox
53% price drop	29% price drop	54% price drop	29% price drop	54% price drop	51% price drop

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Clarifying Question?

- **Query:** acts 17:16

To generate a clarifying question, we need to identify different aspects of the query.

Query Reformulation Data

- shoes → running shoes
- shoes → shoes basketball
- shoes → formal shoes
- shoes → shoes nike
- shoes → shoes adidas
- shoes → shoes for women
- shoes → shoes for kids
- shoes → shoes sale
- shoes → shoes online
- shoes → shoes near me
- ...

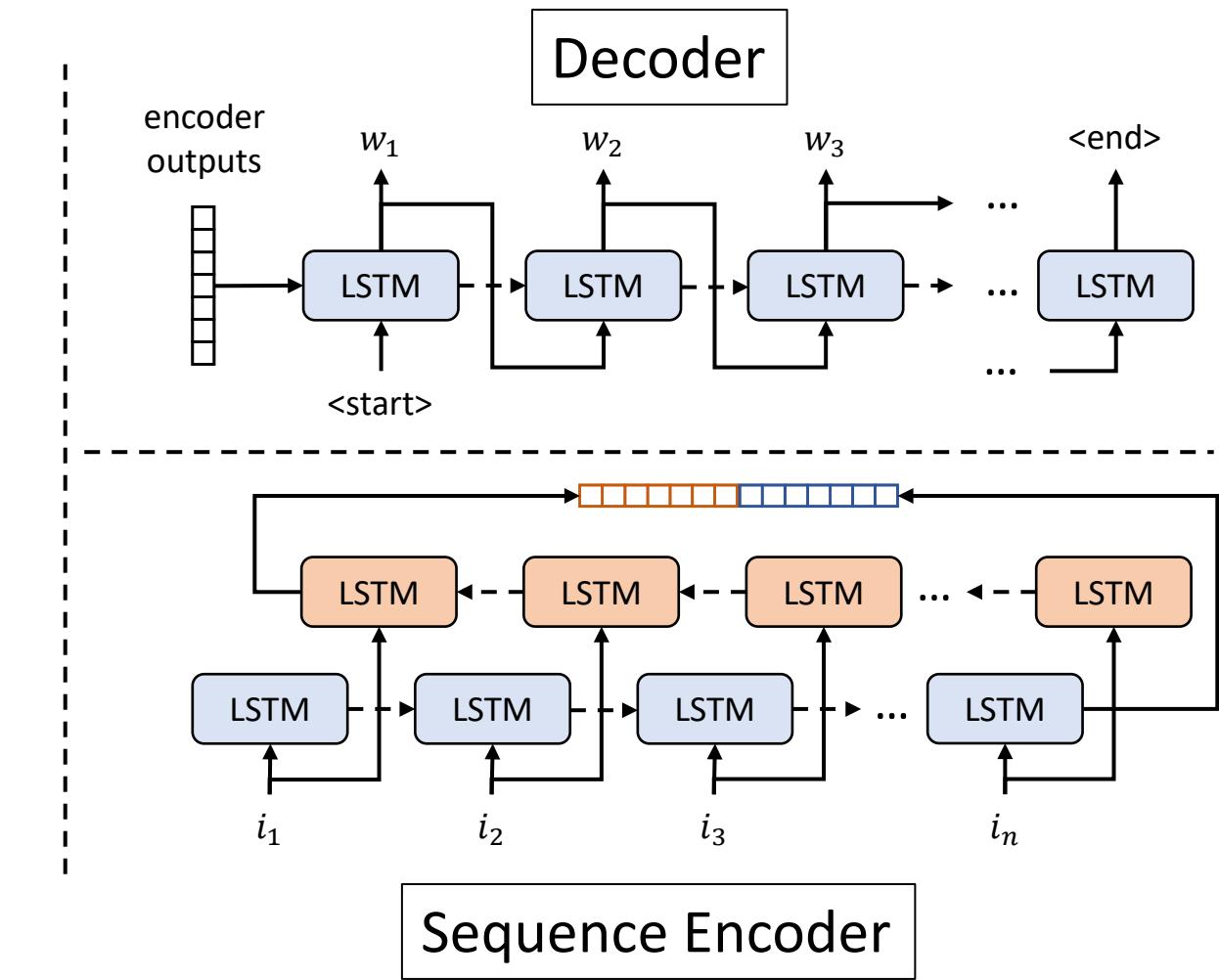
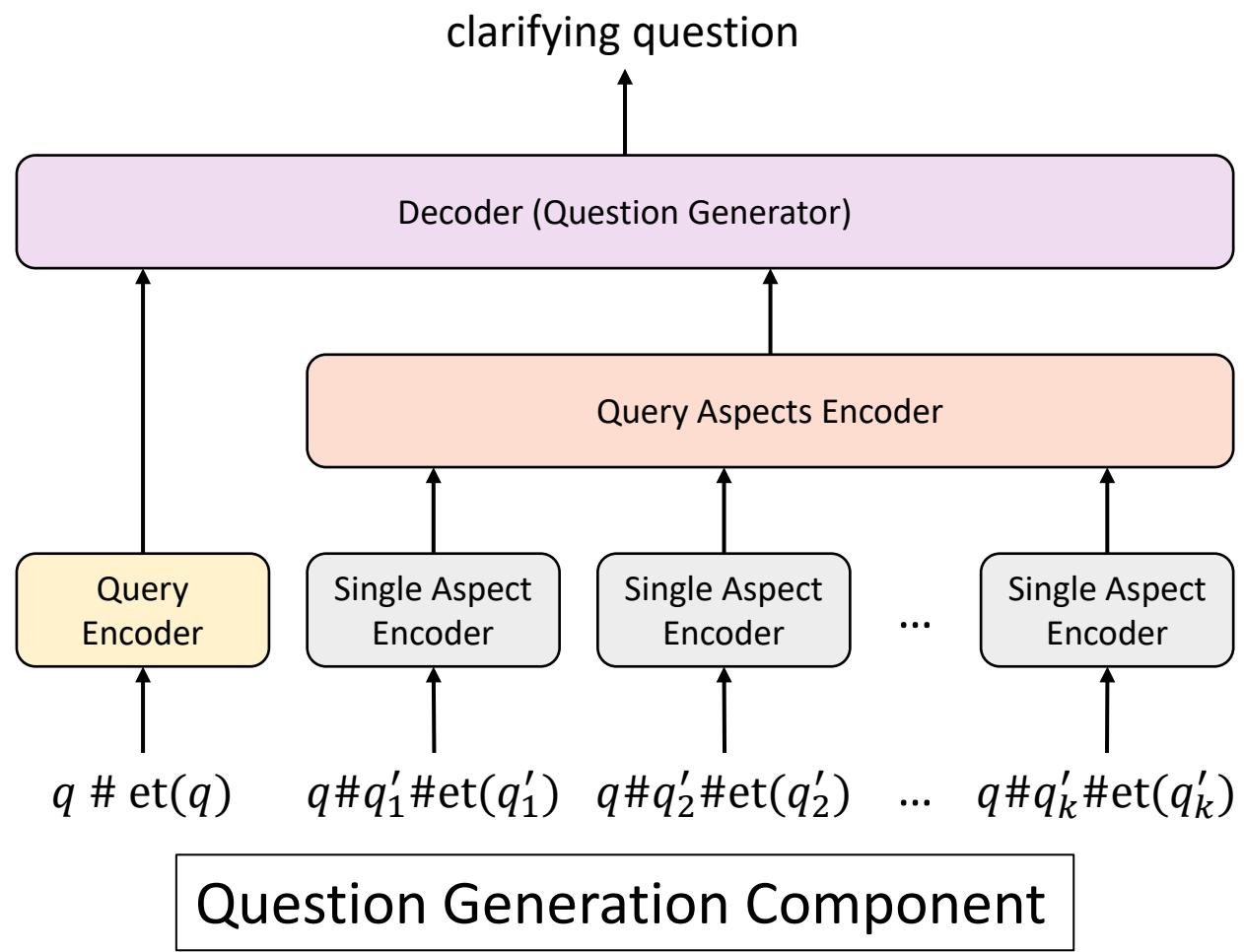
Taxonomy of Search Clarification

- **Disambiguation**
- **Preference**
 - Personal information (“for whom”)
 - Spatial information (“where”)
 - Temporal information (“when”)
 - Purpose (“for what purpose”)
- **Topic**
 - Sub-topic information
 - Event and news
- **Comparison**

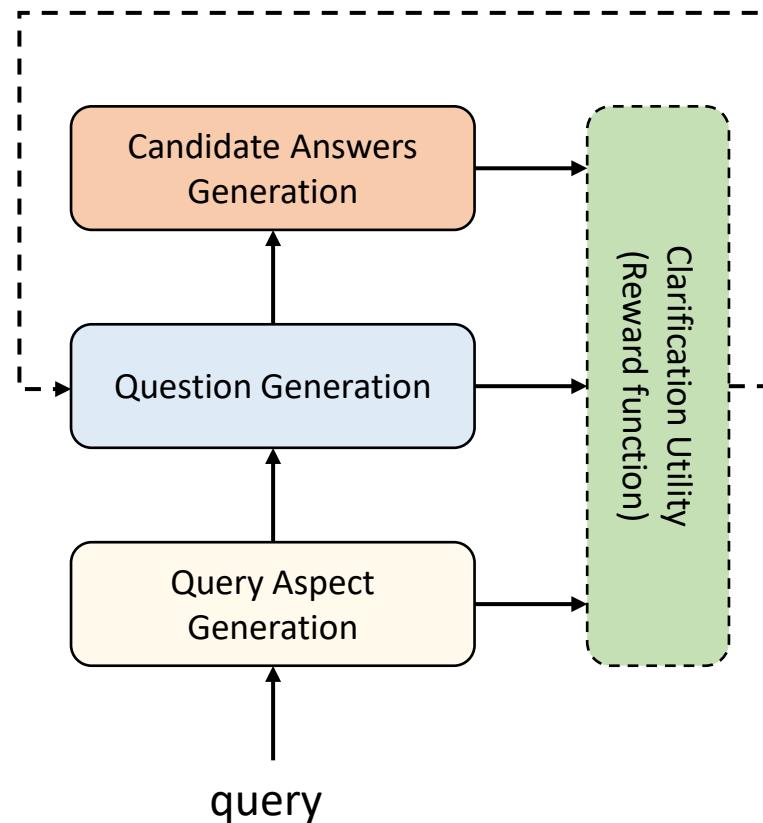
Question Templates

- What do you want to know about **QUERY**?
- What do you want to know about this **QUERY_ENTITY_TYPE**?
- What **ASPECT_ENTITY_TYPE** are you looking for?
- Whom are you looking for?
- ...

Question Generation

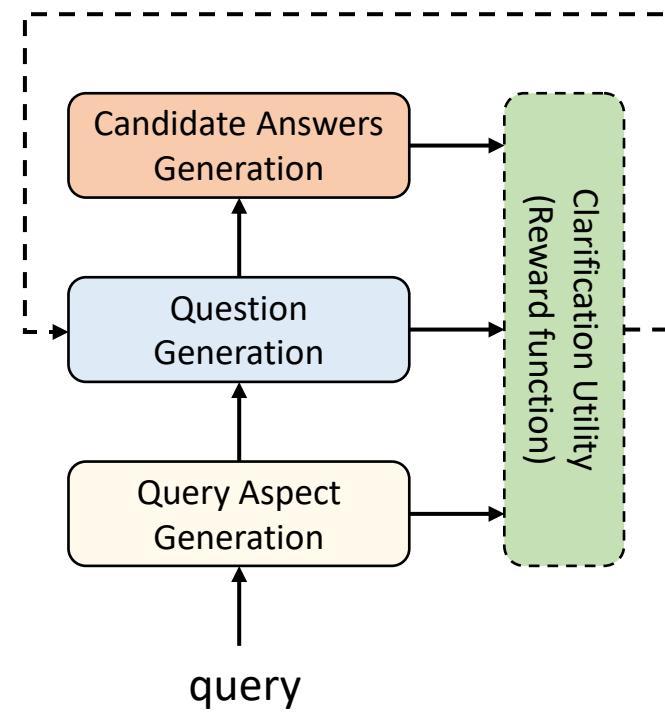


Query Clarification Maximization Model



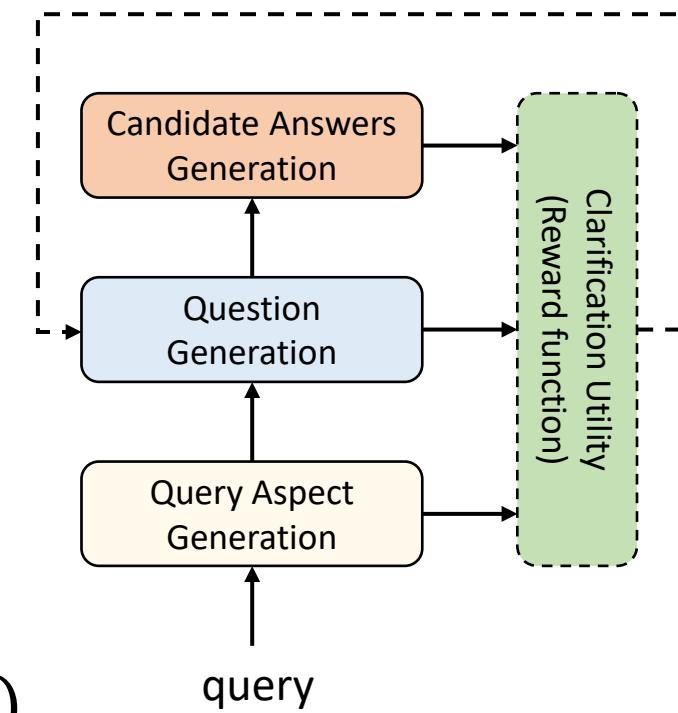
Query Aspect Generation

- To generate a good clarification question, we must first identify different aspects of the query.
- We mine Bing query logs to find all possible reformulations with the form of $X \rightarrow XY$ or $X \rightarrow YX$.
 - “shoes” \rightarrow “running shoes”
 - “running shoes” \rightarrow “running shoes for women”
- Input: X
- Output: all possible Y s with some pre-processing (e.g., initial stopword removal, etc.)



Clarification Utility

$$\begin{aligned} & p(c = 1 | q, q^*, A) \\ &= \sum_{i \in I_q} p(c = 1 | i, q, q^*, A) p(i | q, q^*, A) \end{aligned}$$



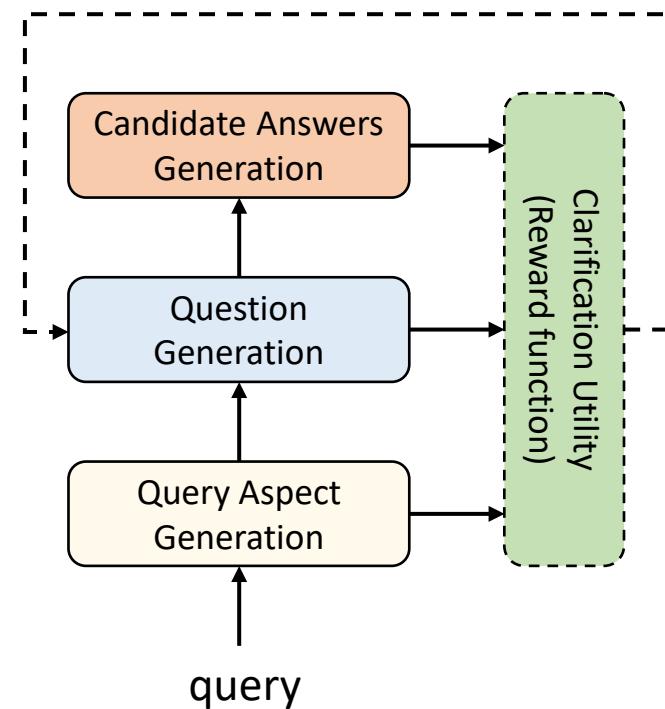
- I_q is the set of all intents for the query q .
- I_q is generally *unknown*, but can be estimated using query reformulation data.

Candidate Answer Generation

$$\arg \max_A p(c = 1 | q, q^*, A)$$

- NP-hard
- The good news is the objective is a **monotone, sub-modular** function
- There is a nice approximation guarantee for a greedy algorithm that generates answers one by one.
- **Theorem (Fisher, Nemhauser, Wolsey, 1978):** if f is monotone, sub-modular, and $f(\emptyset) = 0$ then the greedy algorithm return a solution that achieves:

$$f(S) > \left(1 - \frac{1}{e}\right) OPT$$



Training

- Training is based on REINFORCE:

$$L = -(r(q^*) - r(q_{ML}^*)) \sum_{t=1}^T \log p(q_t | q_1 \dots q_{t-1})$$

- q^* is obtained by sampling from the output distribution of the model.
- q_{ML}^* is the output of pre-trained model.
- $r(\cdot)$ is the reward function, based on clarification utility.

Query	rytary
Question	what do you want to know about this medication?
Options	dosage, coupon, side effects, cost, information
Query	acts 17:16
Question	what bible translation are you looking for?
Options	american standard version, kjv, esv, niv, nlt
Query	that's how i got to memphis
Question	what song information are you looking for?
Options	lyrics, stream, download, artist
Query	alan turing
Question	what do you want to know about this british mathematician?
Options	movie, suicide note, quotes, biography

User Interactions with Search Clarification

Quotes from User Interviews

- “convenient and easy”
- “it saves time and steps”
- “it sometimes cues the user to things they may not have considered”
- “helped them find more relevant results”

Opinion on non-relevant and low-quality clarifications:

- “It’s like when I look at iPhones, and eBay says ‘since you looked at iPhones you may be interested in these hair curlers!’ And I’m like, well that’s weird, whatever”
- The quality of result page after using clarification is important.

Key Findings from User Interviews

- Functional benefit:
 - “questions help guide users in the right direction”
- Emotional benefit:
 - it brings to users a **sense of confidence** that the search engine understands what the user wants.
 - it gives the users a **sense of security** and coming to the right conclusion.
 - The users pointed out that sometimes, especially when it comes to product search, they feel **less stress** when the search engine asks questions on different features of the product.

A/B Test

blue screen

All Images Videos Maps News Shopping | My saves

What version of Windows are you looking for?

windows 10 windows 8 windows 7 windows vista windows xp

How to Fix a Blue Screen of Death (BSOD) - Lifewire
<https://www.lifewire.com/how-to-fix-a-blue-screen-of-death-2624518> ▾

1. The most important Blue Screen of Death troubleshooting step you can take is to ask yourself what ...
2. Check that there's enough free space left on the drive Windows is installed on. Blue Screens of ...
3. Scan your computer for viruses. Some viruses can cause a Blue Screen of Death, especially ones ...
4. Apply all available Windows service packs and updates. Microsoft regularly releases patches and ...
See all full list on lifewire.com

Videos of blue screen
bing.com/videos

:((play) :((HOW TO FIX (play) Advanced options

blue screen

All Images Videos Maps News Shopping | My saves

Select one to refine your search

windows 10 windows 8 windows 7 windows vista windows xp

How to Fix a Blue Screen of Death (BSOD) - Lifewire
<https://www.lifewire.com/how-to-fix-a-blue-screen-of-death-2624518> ▾

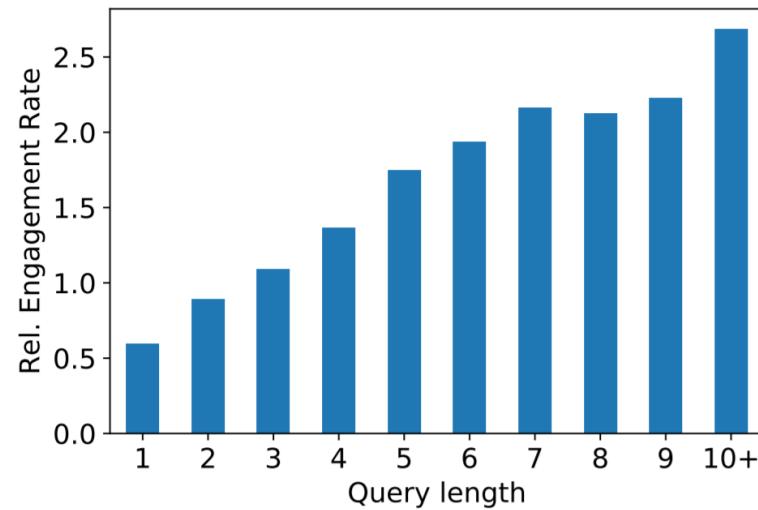
1. The most important Blue Screen of Death troubleshooting step you can take is to ask yourself what ...
2. Check that there's enough free space left on the drive Windows is installed on. Blue Screens of ...
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4. Apply all available Windows service packs and updates. Microsoft regularly releases patches and ...
See all full list on lifewire.com

Videos of blue screen
bing.com/videos

:((play) :((HOW TO FIX (play) Advanced options

+48% more user engagement

Zamani, Dumais, Craswell, Bennett, Lueck. "Generating Clarifying Questions for Information Retrieval". In WWW '20.

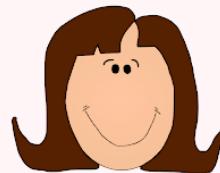


Query type	Relative engagement rate
Natural language question	1.58
Other queries	0.96
Faceted queries	1.52
Ambiguous queries	0.70
Tail queries	1.01
Torso queries	1.02
Head queries	0.99

Clarification is
also useful for
exploratory
search!

Over 7% of interactions with the clarification pane
lead to ***exploration!***

Free-Form Clarification Responses



dinosaur



Information Need (Facet)

I'm looking for the Discovery Channel's dinosaur site, which has pictures of dinosaurs and games.



Are you looking for dinosaur books?



No, just the discovery channel website.



Are you looking for meat-eating or plant-eating dinosaurs?



I'm not sure.

No answer



Would you like to see pictures or videos of dinosaurs?



I'd like to see pictures of dinosaurs on the discovery channels website.



dinosaur



Information Need (Facet)

I'm looking for a list of all (or many of) the different kinds of dinosaurs, with pictures.



Are you looking for dinosaur books?



Yes, if they contain pictures of all the different kinds of dinosaurs.



Which dinosaurs are you interested in?



I'm interested in any and all dinosaurs.



Do you want a list of dinosaurs names?



Yes, I would also like the list to include pictures of the dinosaurs.

dinosaur 



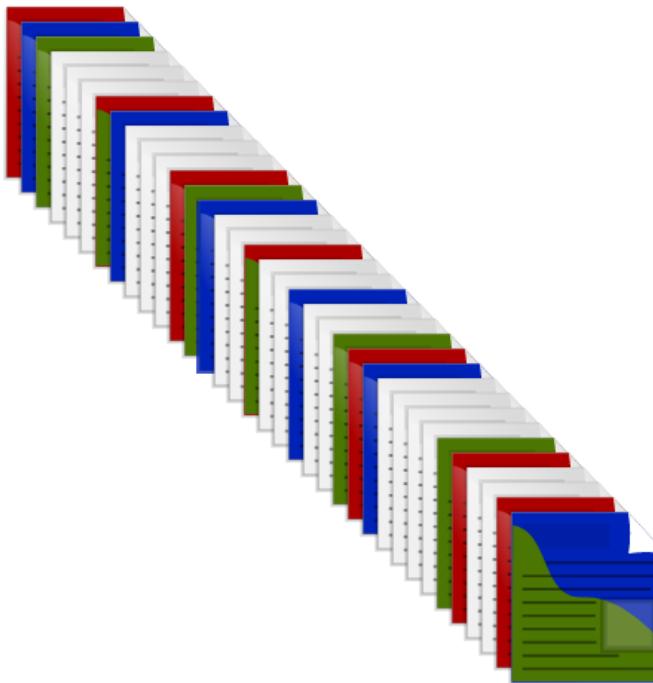
Discovery channel



Dinosaur pictures



Dinosaur books



dinosaur



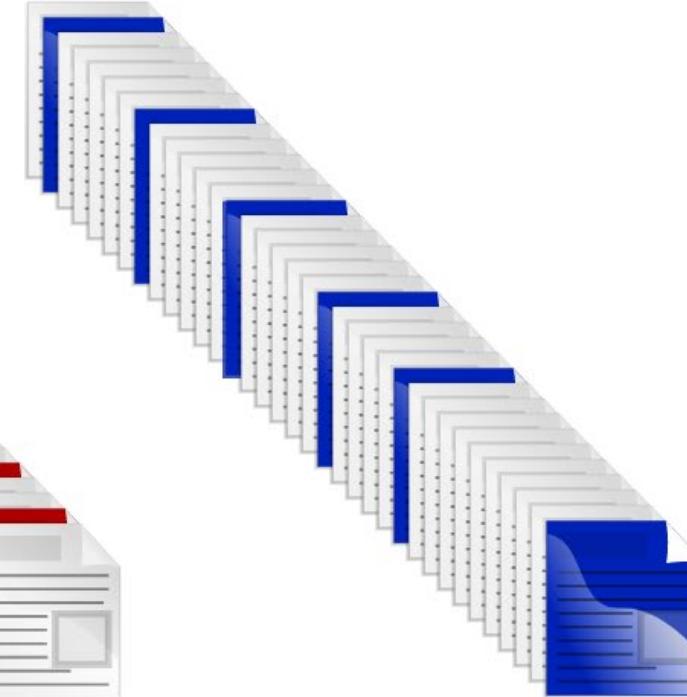
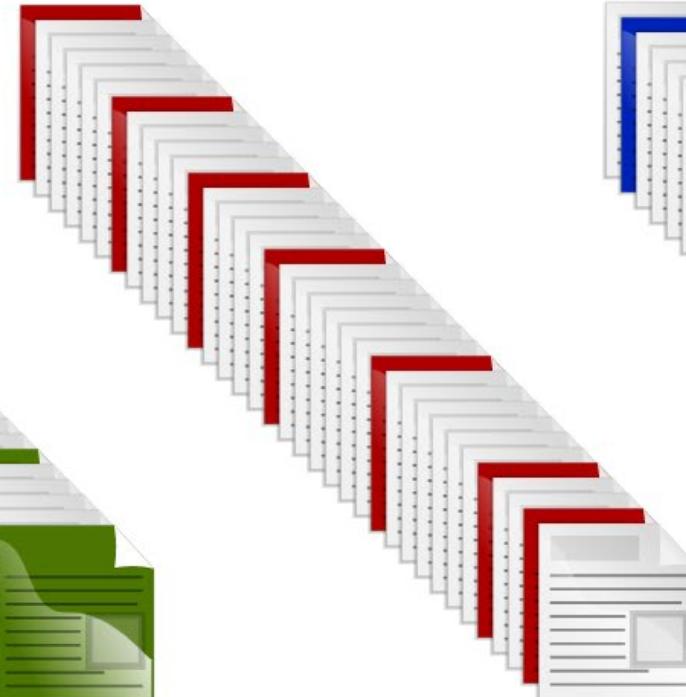
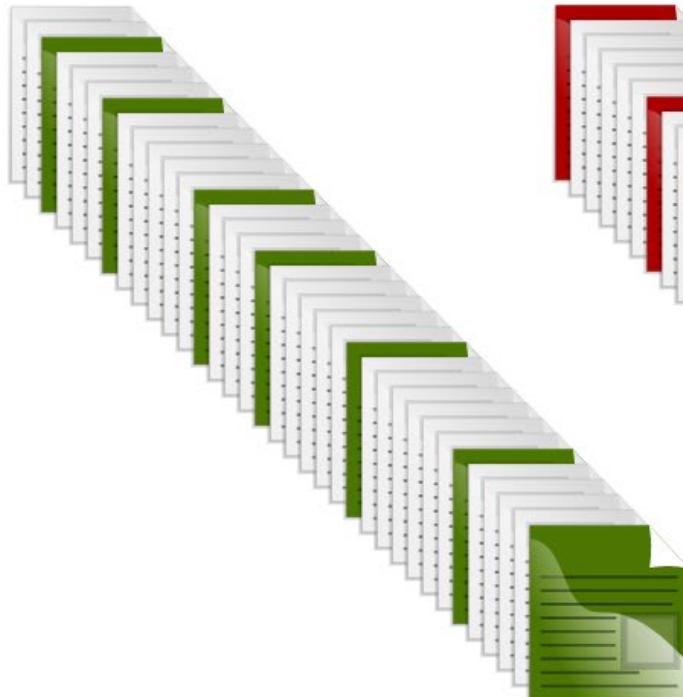
Discovery channel

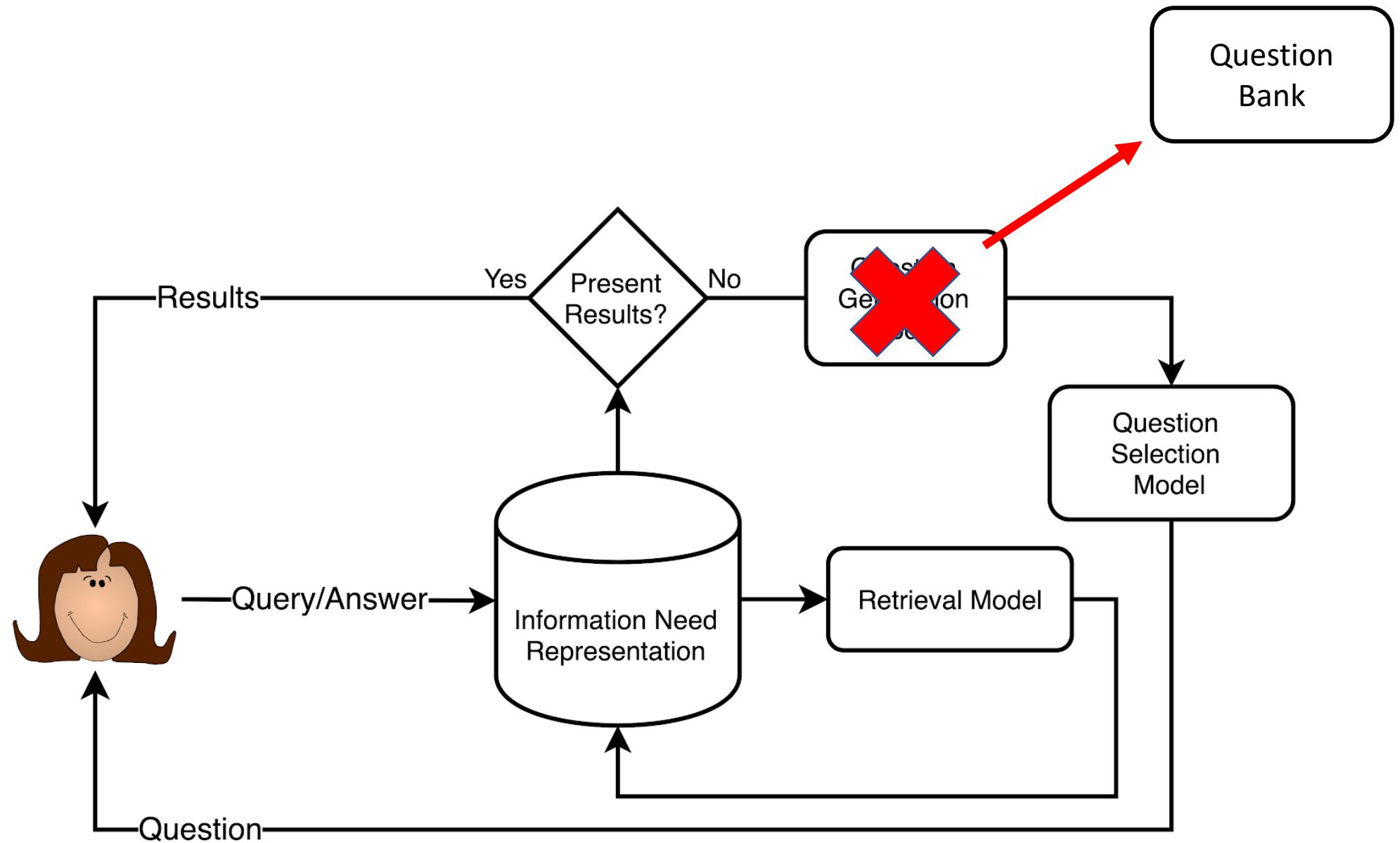


Dinosaur pictures



Dinosaur books

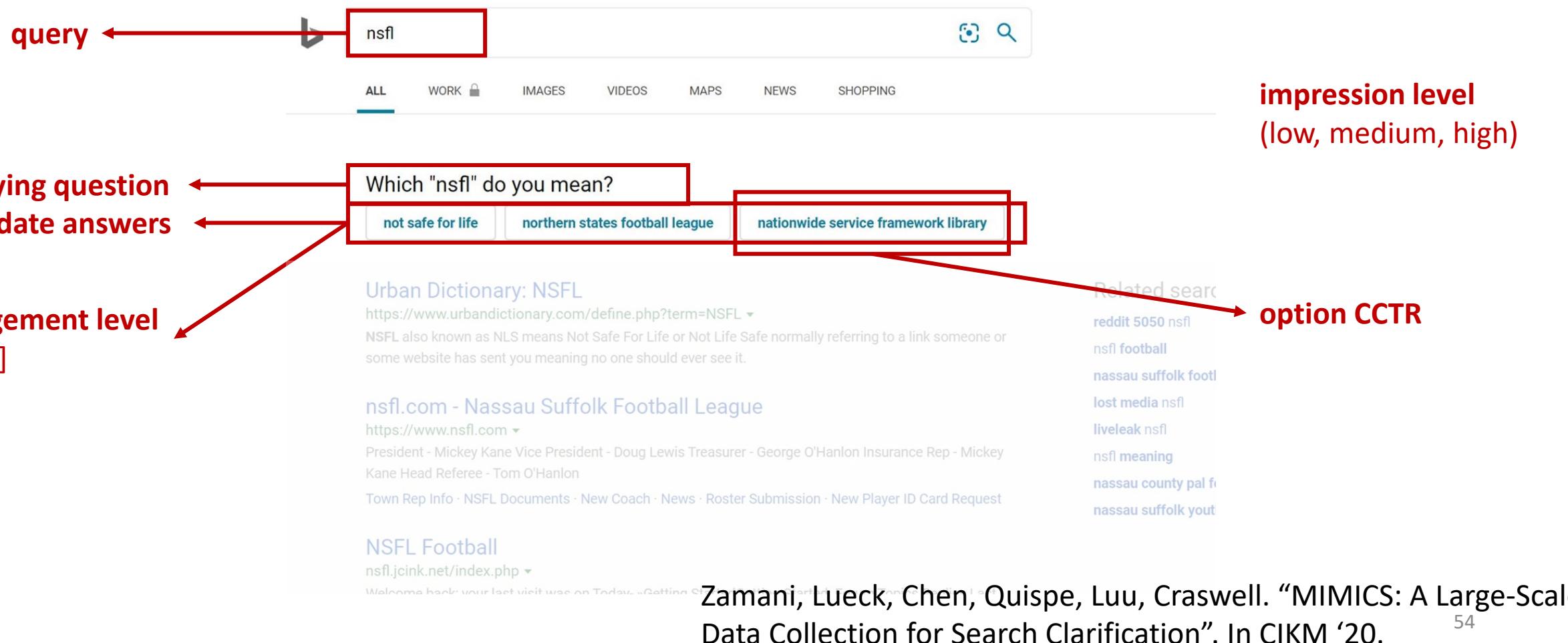




QULAC

# topics	198
# faceted topics	141
# ambiguous topics	57
# facets	762
Average facet per topic	3.85 ± 1.05
Median facet per topic	4
# informational facets	577
# navigational facets	185
# questions	2,639
# question-answer pairs	10,277
Average terms per question	9.49 ± 2.53
Average terms per answer	8.21 ± 4.42

MIMICS



Data Statistics

	MIMICS-Click	MIMICS-ClickExplore	MIMICS-Manual
# unique queries	414,362	64,007	2464
# query-clarification pairs	414,362	168,921	2832
# clarifications per query	1 ± 0	2.64 ± 1.11	1.15 ± 0.36
min & max clarifications per query	1 & 1	2 & 89	1 & 3
# candidate answers	2.81 ± 1.06	3.47 ± 1.20	3.06 ± 1.05
min & max # candidate answers	2 & 5	2 & 5	2 & 5
# query-clarification pairs with positive engagement	71,188	89,441	N/A
# query-clarification pairs with low/medium/high impressions	264,908 / 105,879 / 43,575	52,071 / 60,907 / 55,943	N/A

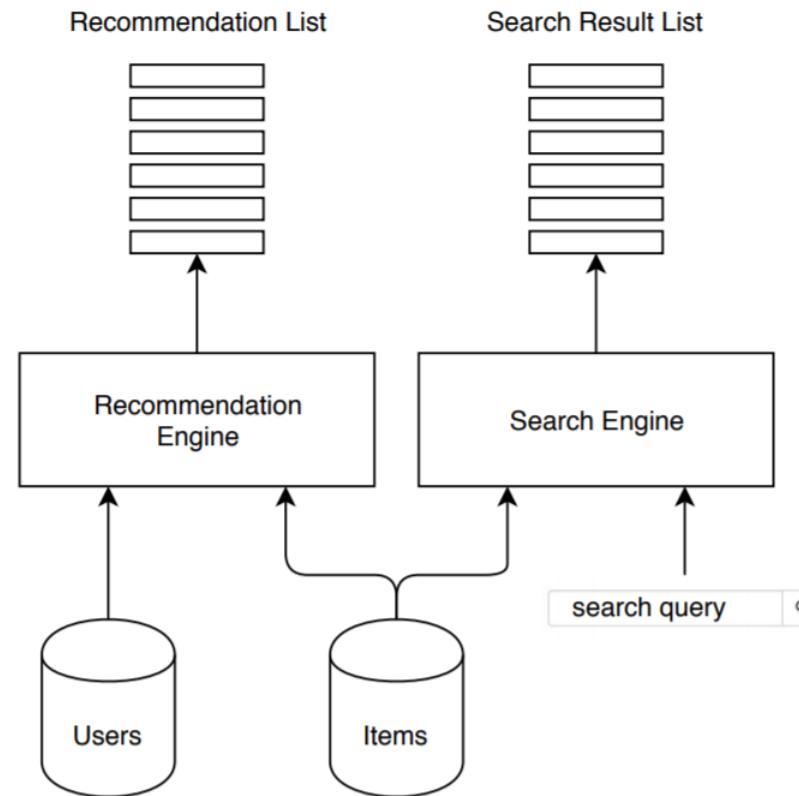
Qulac vs. MIMICS

	Qulac [SIGIR 2019]	MIMICS [CIKM 2020]
# queries	198	>450,000
Document type	webpages	webpages
Clarifying questions	Generated through crowdsourcing	Generated using a machine learning model
User responses to clarification	Generated through crowdsourcing	Real user interaction signals

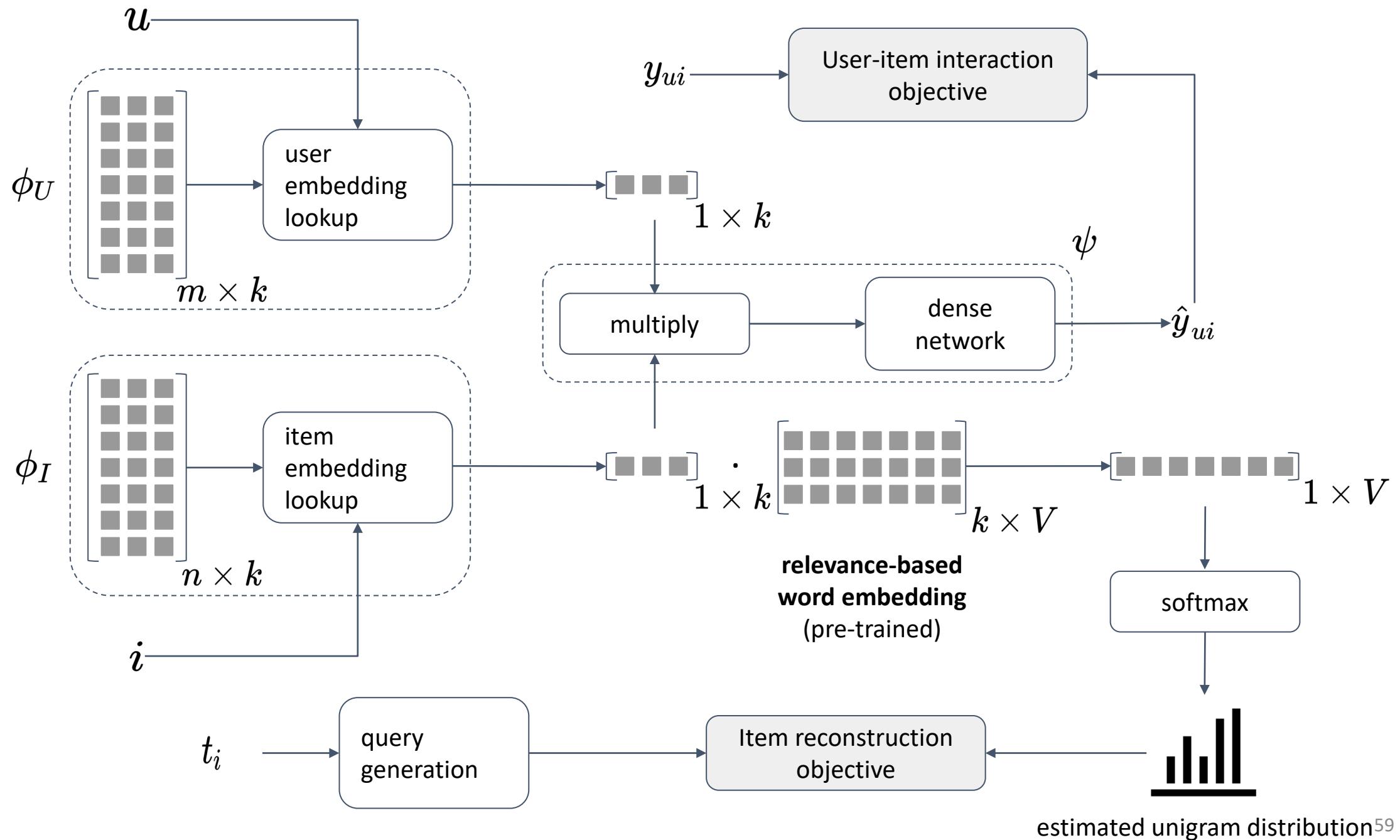
MIMICS-Duo [Tavakoli et al., SIGIR 2022] enables both offline and online evaluation for search clarification.

Bridging the Gap between IR and RecSys

Joint Modeling of Search and Recommendation



Zamani and Croft. "Joint Modeling and Optimization of Search and Recommendation". In DESIRES '18.⁵⁸



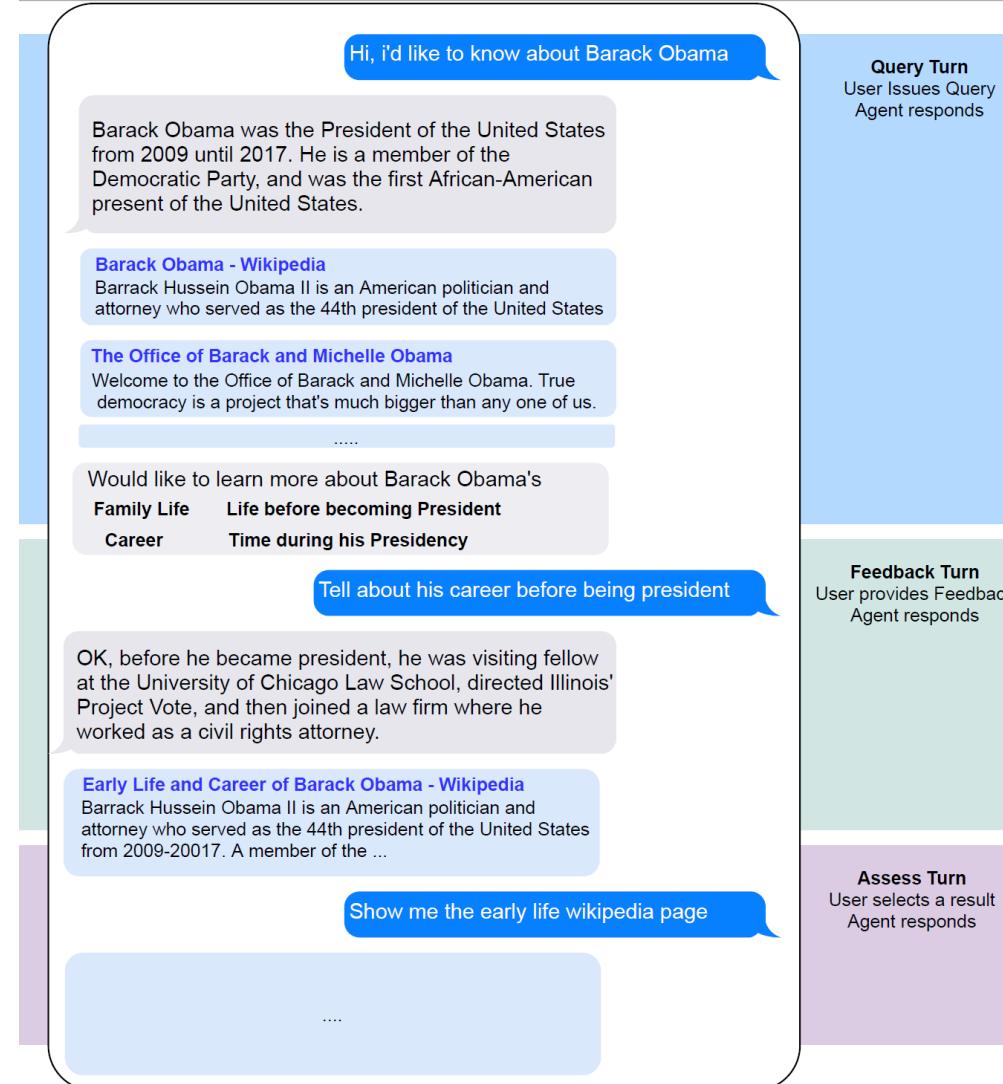
The Lord of the Rings (1978)	Batman Returns (1992)	Gandhi (1982)	The Mask (1994)
fantasy	batman	documentary	cartoon
magic	character	film	parody
movies	superhero	directed	movie
wizard	horror	prize	black ⁵
animation	thriller	award	comic
potter	starring	supporting	comedy
cartoon	fantasy	films	film
fiction	movie	movie	monster
classic	joker	fiction	thriller
novel	comedy	drama	shows

Applications of JSR

- Improving both search and recommendation performance
- Interpretable, transparent, and explainable recommendation
- User profiling
- Universal representation across domain and modality
- Conversational recommendation

Modeling and Measuring Conversational Search

With a focus on mixed-initiative interactions

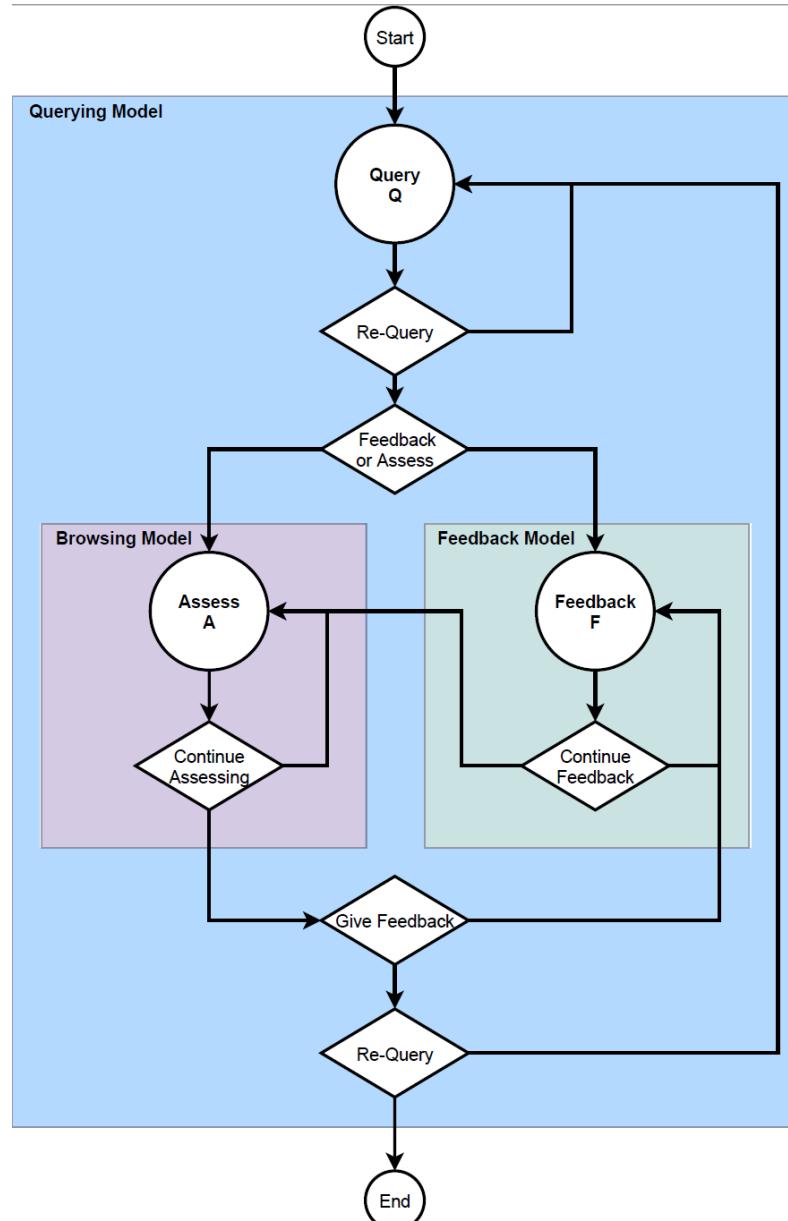


Query Turn
User Issues Query
Agent responds

Feedback Turn
User provides Feedback
Agent responds

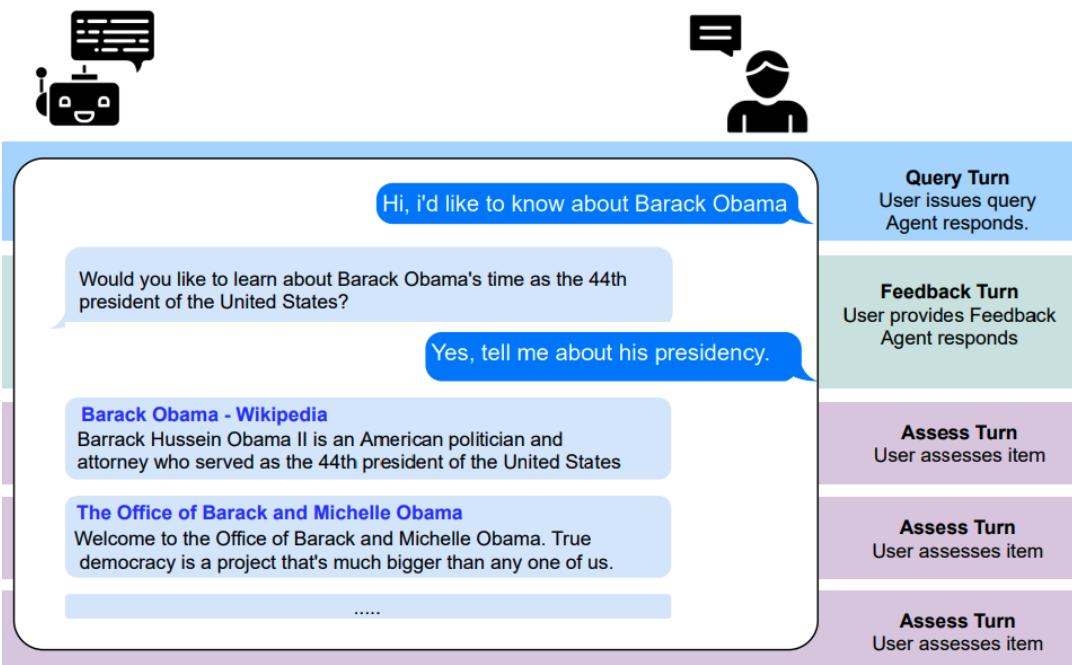
Assess Turn
User selects a result
Agent responds

Aliannejadi, Azzopardi, Zamani, Kanoulas, Thomas, and Craswell.
 "Analyzing Mixed Initiatives and Search Strategies during Conversational Search". *CIKM '21*.

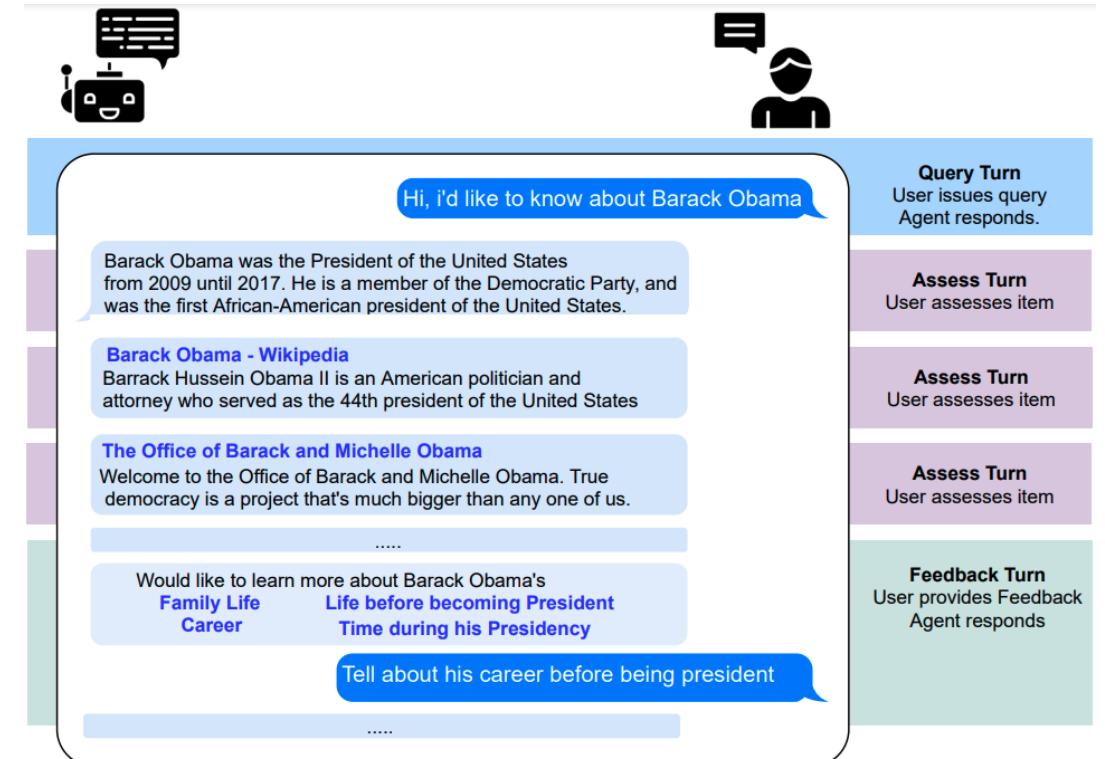


Aliannejadi, Azzopardi, Zamani, Kanoulas, Thomas, and Craswell.
 “Analyzing Mixed Initiatives and Search Strategies during Conversational
 Search”. CIKM ‘21.

Feedback first



Feedback after



Examples of Feedback

Query clarification

Do you have any theme in mind?



Query suggestion

Related searches for wedding dresses

formal wedding dresses **for guests**
google wedding dresses
dillards wedding dresses
dresses **for** wedding **occasion**
wedding dresses **for older brides**
cheap wedding dresses
davids bridal ocala
designer wedding **gowns**

Cost and Benefit (Gain) of a Conversation

$$G(t_1, t_2, \dots, t_T) = \sum_{i=1}^T g(t_i)$$

We can assume that the user only accumulates gain on an assessment of a relevant document.

$$C(t_1, t_2, \dots, t_T) = \sum_{i=1}^T c(t_i)$$

We can consider time as a good estimator of each interaction cost.

$$R = \frac{G(t_1, t_2, \dots, t_T)}{C(t_1, t_2, \dots, t_T)}$$

Estimating the Cost

- Crowdsourcing using the complex TREC Web Track topics.
- Once the user submits a query, the system responds with a result snippet or document for up to five documents (one by one). Then the user can either
 - Reformulate the query, or
 - Answer clarifying questions, or
 - Select a query suggestion, or
 - “Not interested”
- 81 crowdworkers, 144 queries, 1280 snippets, and 268 feedback responses

Aliannejadi, Azzopardi, Zamani, Kanoulas, Thomas, and Craswell.
“Analyzing Mixed Initiatives and Search Strategies during Conversational
Search”. *CIKM ’21*.

Estimating the Cost

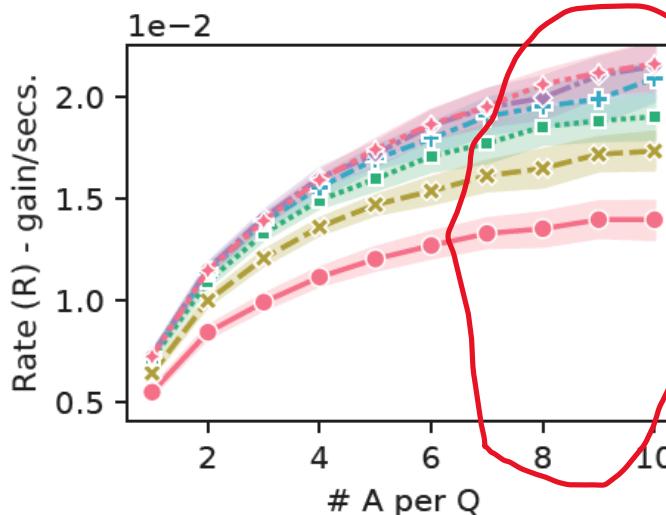
- The average time taken
 - to issue a query: 29.3 s
 - to assess a result snippet: 6.3 s
 - to assess a result web page: 17 s
 - to provide feedback: 8.3 s
 - ...
- They can be used to estimate the cost of each interaction.
- We can then conduct some simulated analysis to explore gain ratio for each conversational strategy.

The strategies

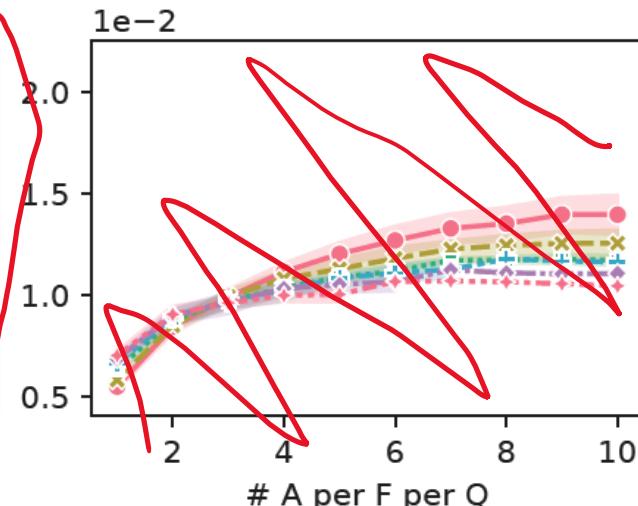
FA-QC combination is inferior.

FF-QS leads to a small increase over the baseline but is not superior.

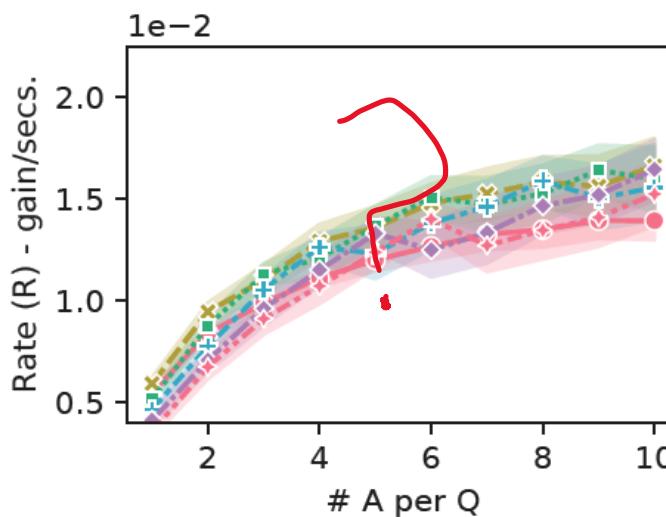
No dominant strategy, depends on #assessments.



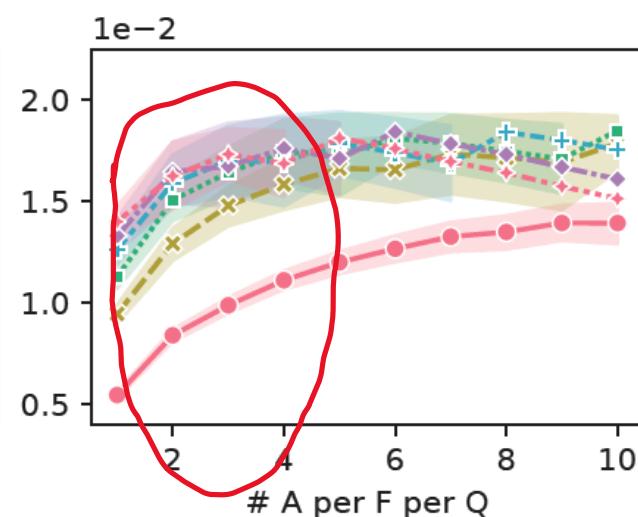
(a) FF-QC



(b) FA-QC



(c) FF-QS



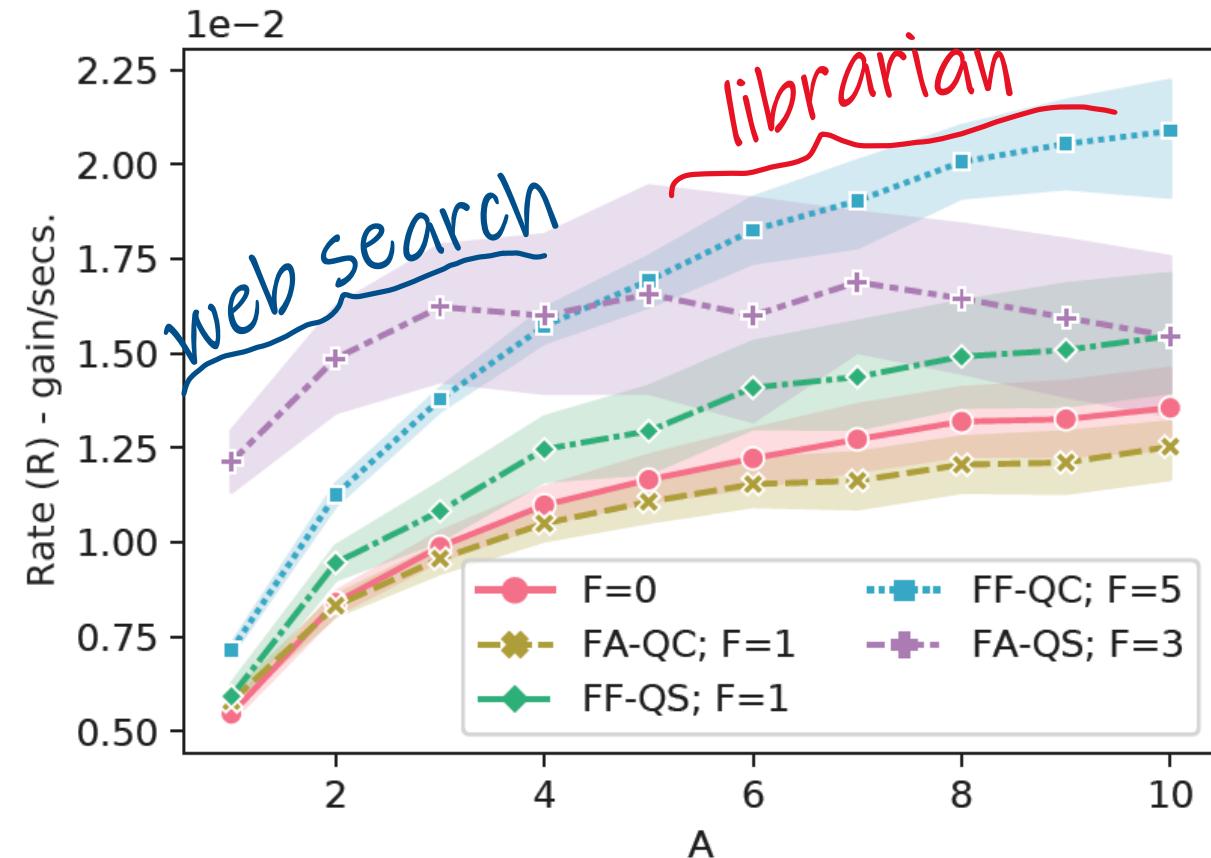
(d) FA-QS



Best strategies

Searcher is only willing to examine a few items: FA-QS

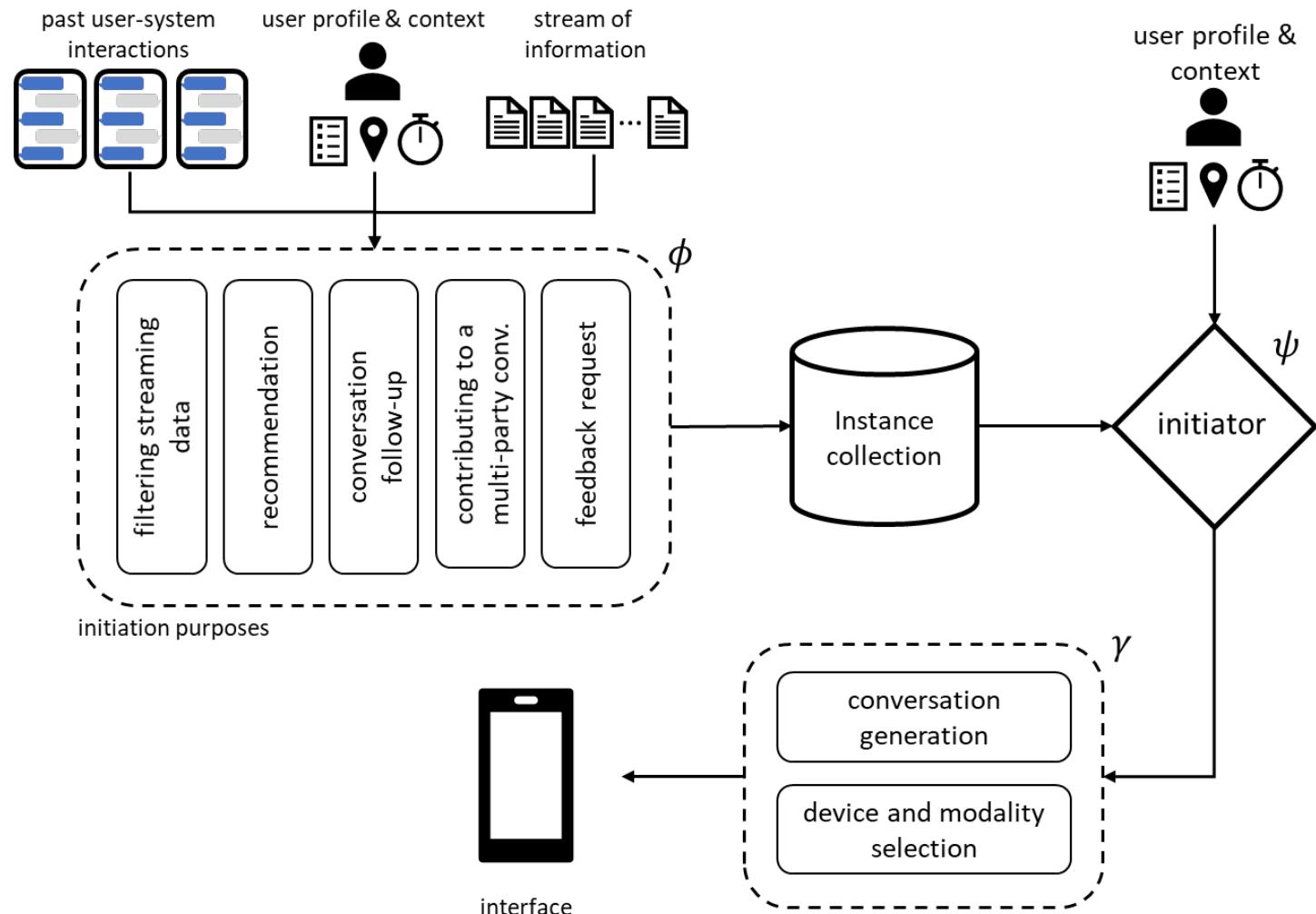
Searcher is willing to go deeper: FF-QC



Agent-Initiative Interactions

Three Major Dimensions in Agent-Initiative Interactions

- Initiation Moment (**when** to initiate a conversation?)
 - Instant initiation
 - Opportune moment initiation
- Initiation purpose (**why** to initiate a conversation?)
 - Filtering streaming information based on user profile
 - Recommendation
 - Following up a past conversation
 - Contributing to a multi-party human conversation
 - Feedback request
- Initiation means (**how** to initiate a conversation?)
 - Device
 - Interaction Modality



Initiation Purposes: Contributing to a multi-party human conversation

- Example:
 - Monitoring factual accuracy of human conversations.
 - Introducing opposing views on the topic being discussed.

Initiation Purposes: Following up a Past User-System Conversation

- Based on new information or new deployment of models
- Example:
 - CIS systems are not by any means perfect and they make mistakes in responding to user's requests. Based on new information or new models deployed in the system, a CIS system may initiate a conversation at an opportune moment to accept and correct its mistakes that was made in the past.

Open Problems and Future Directions

- Evaluating mixed-initiative conversational search systems
- Studying other types of mixed-initiative interactions
- Search result list explanation
- Integrating chit-chat into conversational search and recommendation
- Generating clarifying questions with no access to query logs
- Preference elicitation
- Joint conversational search and recommendation

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References

Thank you!



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Information Retrieval

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and Filip Radlinski



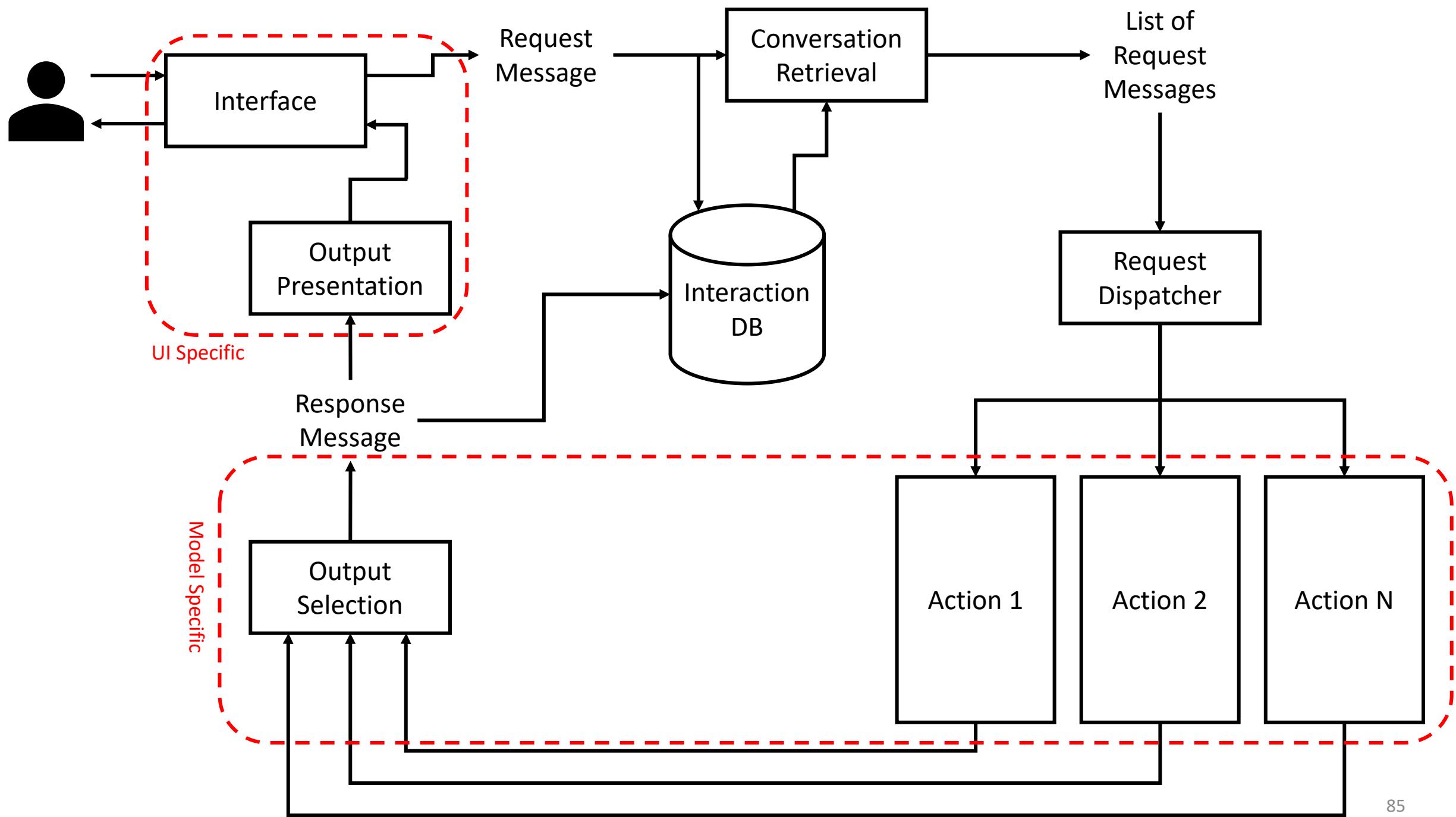
@HamedZamani

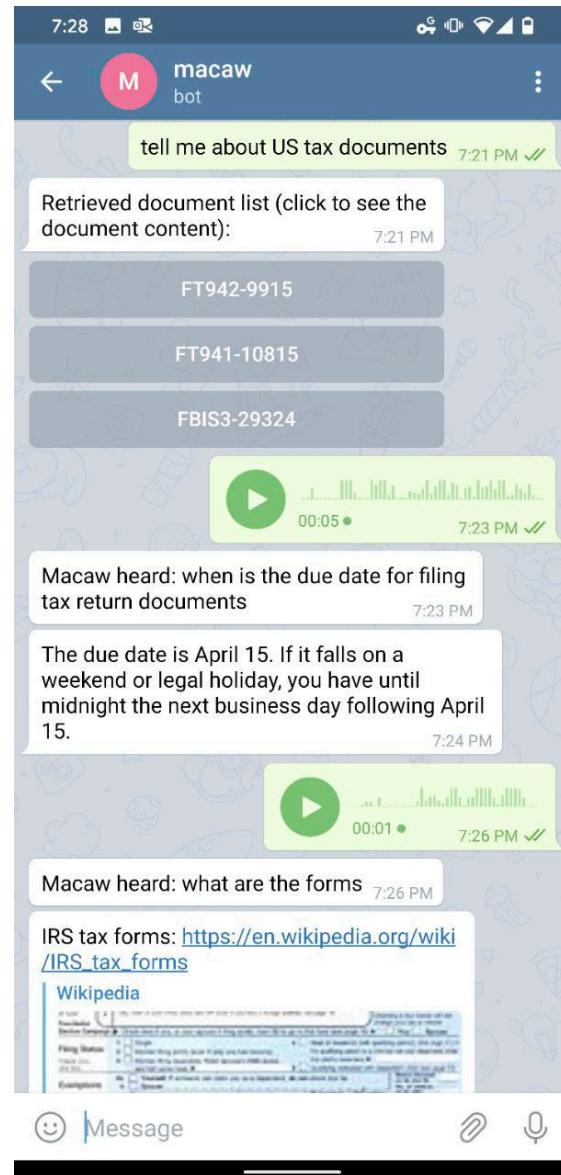
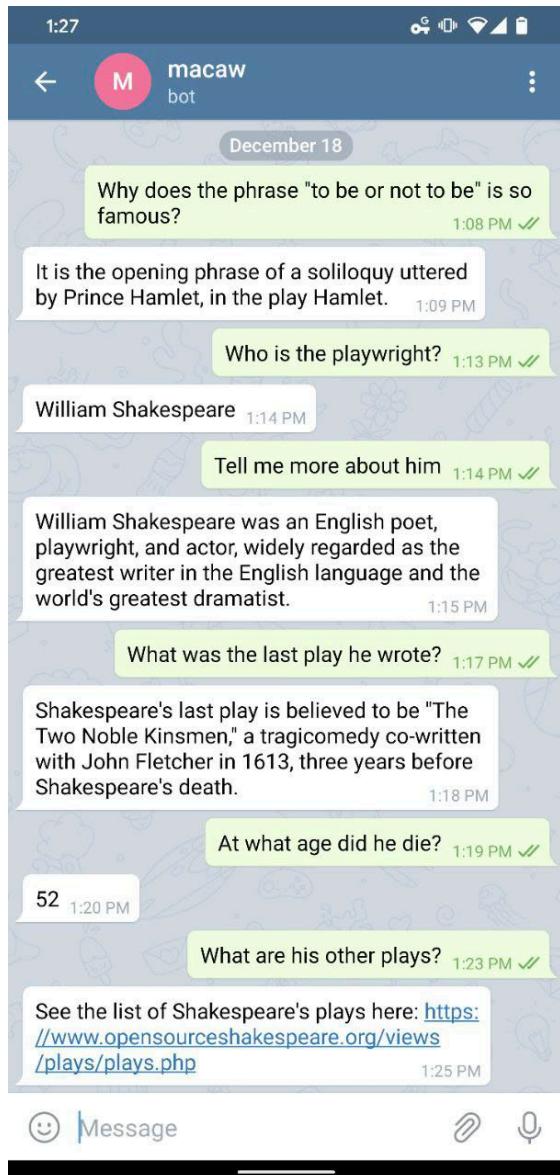
<https://arxiv.org/pdf/2201.08808.pdf>



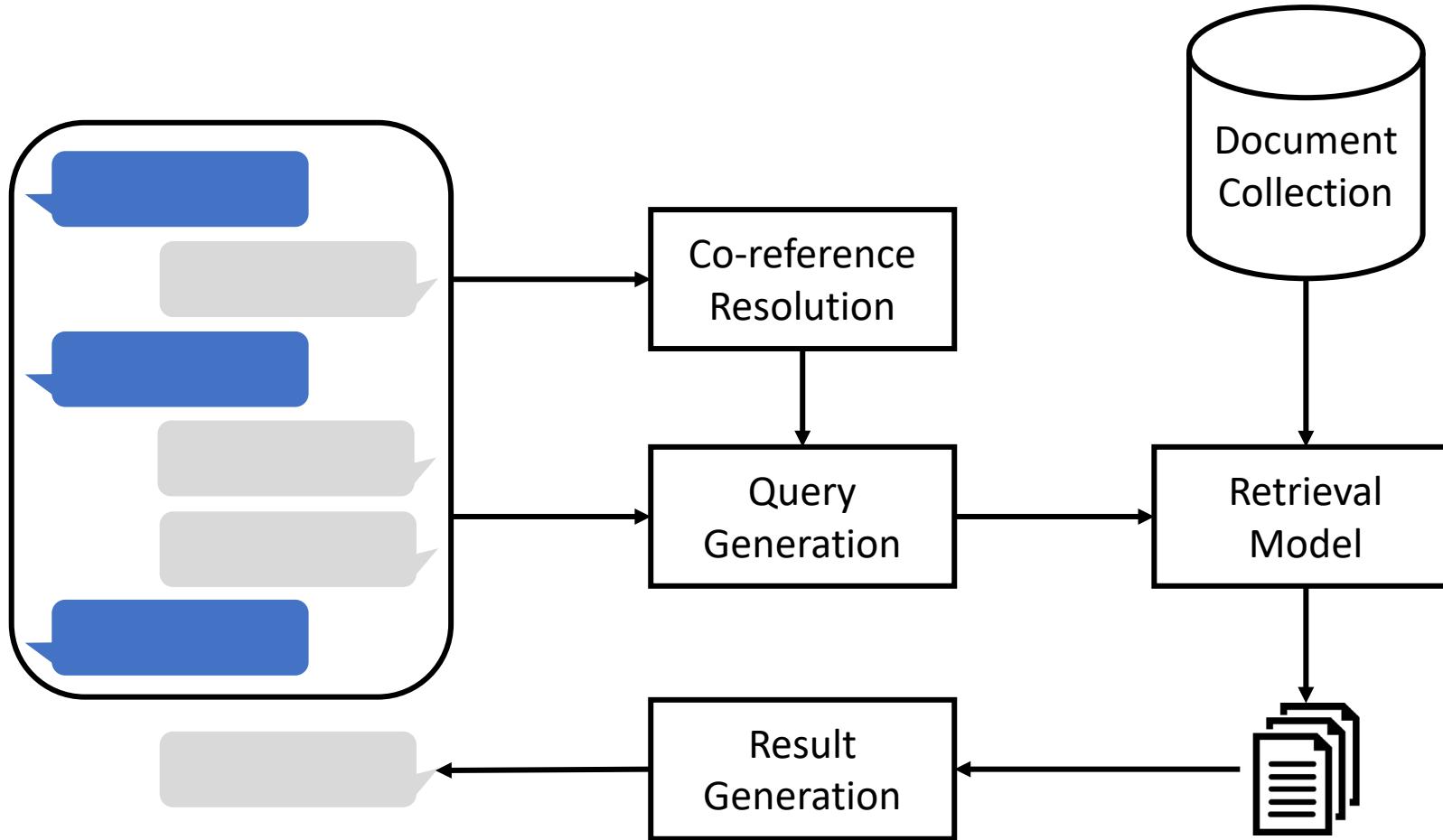
Macaw

<https://github.com/microsoft/macaw>





Action 1 & 2: Search and QA



Other Actions

- Action 4: Clarification
- Action 5: Recommendation
- Action 6: ...

Wizard of Oz setting

