A Large-Scale Dataset for Known-Item Question Performance Prediction

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Abstract

Searchers who cannot resolve a known-item information need using a search engine (e.g., as the searcher only remembers vague details about a movie from some years ago) might post a respective question on a question answering platform, hoping that the discussion with other people can help to identify the item. To foster research on applying query performance prediction to known-item information needs, especially in the light of the upcoming tip-of-my-tongue known-item retrieval track at TREC 2023, we build a large-scale dataset of 1.28 million known-item questions (47 % have an identified answer) from the r/tipofmytongue subreddit. As the "performance" of a known-item question, we use the time it took the community to solve the question (or the absence of a solution) and evaluate the effectiveness of seven standard pre-retrieval query performance predictors in a pilot study. Not surprisingly, none of the tested predictors can really assess the performance of known-item questions.

1. Introduction

Many queries submitted to search engines are navigational queries in the sense that searchers want to find a specific item (for the early days of web search, Broder [1] stated an amount of more than 20 %). Among the navigational queries, known-item and re-finding queries (i.e., accessing an item known to exist or even previously accessed), can be particularly challenging when a searcher is unable to recall a suitable identifier for the item [2, 3]. Actually, many people who post known-item questions on Reddit state that their previous attempts with traditional search engines were unsuccessful [4]. Still, known-item information needs posted on question answering platforms have hardly been part of retrieval system evaluations (e.g., at TREC tracks); the main focus usually are informational queries. Since also the previous work on query performance prediction (QPP) [5, 6, 7] used the standard retrieval collections, the effectiveness of query performance prediction on difficult known-item information needs (often expressed as rather verbose questions with explanations) has not been analyzed yet.

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Figure 1: Examples of known-item questions: the left question is answered within a minute, the middle question within three days, and the rightmost question has not been answered after a week.
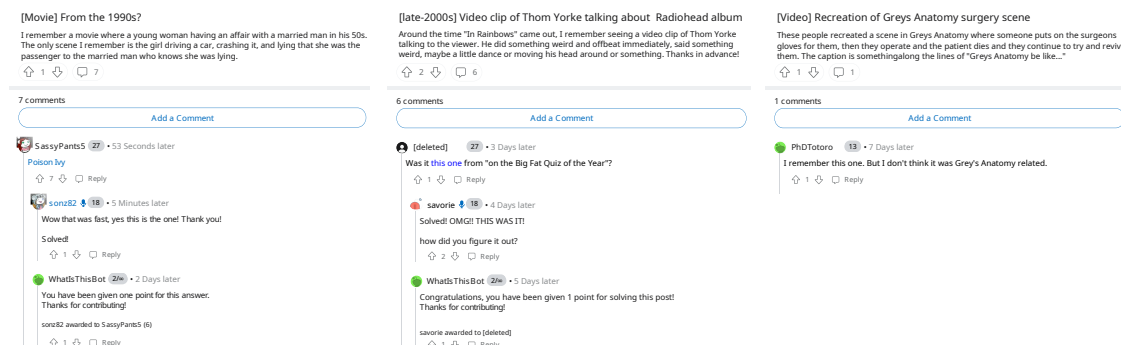
Retrieving answers to known-item questions will be the focus of the upcoming ToT track at TREC 2023.[1] The task can be very challenging since known-item questions often do not mention a suited identifier for the item or only a loosely similar one [8, 9], and since information in the question might even be wrong [10]. As training data, the organizers of the ToT track announced a dataset by Bhargav et al. [8]: 15,000 questions with linked known items (focus: movies and books) extracted from the subreddit r/tipofmytongue/.[2] Having run an adapted version of Bhargav et al.'s approach, our resulting TOMT-KIS dataset (tip-of-my-tongue known-item search) contains 1.28 million known-item questions (47 % with an identified answer) and is freely available under a permissive license.[3]

As an indicator for the "performance" of a known-item question, we simply use the time elapsed to answer the question. For example, the left question in Figure 1 was answered within a minute and thus performed better than the middle question (answered after three days) that again performed better than the rightmost question (not answered after a week). In a pilot study with seven standard pre-retrieval query performance predictors, we find that none of the predictors can really assess the performance of known-item questions.

2. A Large-Scale Dataset of Known-Item Questions

To create our TOMT-KIS dataset, we have crawled all questions and discussions from the tip-of-my-tongue subreddit. Note that answers are not explicitly tagged in the subreddit discussions, but the guidelines state that the asker should reply with "Solved!" to a post with the correct answer. In creating the dataset that the TREC 2023 ToT track organizers suggest as training data, Bhargav et al. [8] focused on questions for which the asker replied exactly with "Solved!" to some post and from these they only kept the questions where the answer contains exactly one link to a Wikipedia, IMDb, or GoodReads page. Using this rather restrictive approach, only 15,000 questions with a specified known-item answer were identified.

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[1]https://trec-tot.github.io/
[2]https://www.reddit.com/r/tipofmytongue/
[3]Data and code for TOMT-KIS: https://github.com/webis-de/QPP-23

Table 1

Excerpt from TOMT-KIS for the examples from Figure 1. For questions a moderator tagged as solved, our precision-oriented heuristic extracted the post with the answer. The data is stored in JSONL format (127 attributes; excerpt of 6 attributes shown in tabular form for readability).

Question				Answer	
Title	Content	UTC	Solved	Content	UTC
[Movie] From the 1990s?	I remember a movie where a young woman having an affair with a married man […]	1576446820	✓	[Poison Ivy] (https://www.imdb.com/ title/tt0105156/)]	1576446873
[late-2000s] Video clip of Thom Yorke talking about Radiohead album	Around the time "In Rainbows" came out, I remember seeing a video clip of Thom Yorke […]	1598985699	✓	Was it [this one] (https:// m.youtube.com/watch? v=rrnNn0mt8h4) from […]	1599242625
[Video] Recreation of Greys Anatomy surgery scene	These people recreated a scene in Greys Anatomy […]	1479737186	✗	—	—

Answer Identification   Analyzing different questions from the subreddit, we observed that askers often do not reply to the correct answer with "Solved!" but with other posts (e.g., "This was it!"), and we found a further metadata field (maintained by moderators but not used by Bhargav et al.) that indicates whether a question is solved. We thus adapted the answer identification method of Bhargav et al. to recall more questions with answers. For questions that the moderator metadata indicates as solved, we use four manually created lexical patterns as a rule-based answer identification method: if a post from the asker in reply to a post a contains 'solved', 'thank', 'yes', or 'amazing', we view the post a as the answer. On 50 random questions and 50 questions for which this heuristic identified an answer (none of them used to develop the rules), the achieved precision is 92 % at a recall of 78 %.

Dataset Description   Our TOMT-KIS dataset is available in JSONL format. For each question, we include all the attributes available in the crawled data and add the chosen answer when our heuristic could extract it. Table 1 shows an excerpt for the questions from Figure 1.

Overall, TOMT-KIS includes 127 attributes for each question, such as `title`, `content`, `created_utc` (indicating the posted question's timestamp), and `link_flair_text` (indicates whether the question is solved; set by moderators). The complete tree of the discussion on each question is stored in the `comments` field. To simplify subsequent processing steps, we run our precision-oriented answer identification heuristic on questions tagged as solved by a moderator and add four "new" attributes when the heuristic could identify an answer: (1) `answer_detected` is a Boolean flag indicating whether our heuristic could extract an answer, (2) `solved_utc` specifies the timestamp when the identified answer was posted, (3) `chosen_answer` contains the extracted answer, and (4) `links_on_answer_path` contains all links to Reddit-external pages that were found in posts between the question and the post with the answer (this can be used for future retrieval experiments like retrieving known-item candidates from a web crawl [8]).

Dataset Analysis   Many questions in the tip-of-the-tongue subreddit have assigned tags that roughly describe an assumed category of the known item. Table 2 shows the 60 most frequent tags in our TOMT-KIS dataset, excluding the [tomt] tag itself (all questions have it).

Table 2
The 60 most frequent tags in our TOMT-KIS dataset; we later merged similar ones to larger categories.

Rank	Tag	Count	Rank	Tag	Count	Rank	Tag	Count
1	[song]	161,385	21	[meme]	5,789	41	[early 2000s]	2,695
2	[movie]	158,543	22	[reddit post]	5,748	42	[2010s]	2,646
3	[video]	67,642	23	[comic]	5,503	43	[flash game]	2,584
4	[music]	47,591	24	[gif]	5,383	44	[animation]	2,189
5	[book]	47,578	25	[image]	4,894	45	[band]	2,175
6	[2000s]	38,200	26	[film]	4,320	46	[reddit]	2,088
7	[game]	32,961	27	[short story]	4,287	47	[tv series]	2,069
8	[2010s]	22,122	28	[2000s]	4,282	48	[album]	2,022
9	[tv show]	17,731	29	[quote]	4,269	49	[computer game]	2,012
10	[music video]	17,434	30	[pc game]	4,115	50	[2010s?]	1,998
11	[website]	14,140	31	[picture]	4,095	51	[2000's]	1,977
12	[video game]	13,007	32	[commercial]	4,012	52	[manga]	1,939
13	[cartoon]	10,774	33	[90s]	3,560	53	[movie/tv]	1,934
14	[tv]	9,395	34	[videogame]	3,371	54	[short film]	1,917
15	[youtube video]	8,107	35	[2000s?]	3,355	55	[2000s-2010s]	1,890
16	[1990s]	7,784	36	[1980s]	3,028	56	[youtube channel]	1,860
17	[youtube]	7,636	37	[webcomic]	2,955	57	[documentary]	1,758
18	[anime]	7,284	38	[story]	2,872	58	[tiktok]	1,713
19	[show]	7,215	39	[subreddit]	2,782	59	[movies]	1,655
20	[word]	6,394	40	[toy]	2,705	60	[2020]	1,623

Besides [tomt], a question has between 0 and 14 tags (average: 0.89), some of which directly express uncertainty (e.g., [2010s?] vs. [2010s]).

For further analyses, we manually merged the original tags to form larger categories (e.g., combining [song], [music], etc. to a "Music" category). Table 3 shows general statistics for all questions from our new TOMT-KIS dataset and for the four most popular merged categories. As a proxy for the "performance" of a known-item question, we use the time elapsed until the solution was posted (columns Δ_T; lower is better). Obviously, the performance varies between years and between categories. For instance, relatively more known-item questions in the movies category are solved than in the music category (52 % for movies vs. 46% for music) and the average time until an answer is posted is lower (9 hours for movies vs. 14 hours for music). Since query performance prediction is evaluated via measuring the correlation of a predictor's query ranking (by predicted performance) to the ground truth ranking [5, 7, 11], we build training, validation, and tests sets for the set of all questions and for the shown four most popular categories by each time sampling 100 questions for validation and 100 questions for test while keeping all other questions as training data.

Table 4 shows length statistics of the titles and contents of the questions, and of the identified answers in our TOMT-KIS dataset as the average number of characters and words (whitespace tokenization). Overall, the titles of questions are much shorter (14 words on average across all categories) than the explanations in the content field (81 words on average), while the identified answers again are rather short (12 words on average). There are some notable differences between categories like the titles for movie questions being longer (15 words on average) or the answers in the book category being longer (16 words on average).

Table 3
Overview of our TOMT-KIS dataset by year: number of questions (#), proportion of questions that are solved as identified by our heuristic (Solv.), and average time in hours to solve a question (Δ_T); for all questions combined and for the four most frequent categories (Movies, Music, Books, Games).

Year	All Questions			Movies			Music			Books			Games		
	#	Solv.	Δ_T	#	Solv.	Δ_T	#	Solv.	Δ_T	#	Solv.	Δ_T	#	Solv.	Δ_T
2009	1,045	0.01	2.11	47	0.00	—	36	0.03	0.22	18	0.00	—	9	0.00	—
2010	3,861	0.02	96.34	268	0.03	671.28	97	0.02	0.49	97	0.00	—	25	0.08	0.09
2011	24,544	0.02	10.88	2,702	0.02	2.95	1,591	0.02	2.01	834	0.03	99.47	246	0.03	1.07
2012	52,356	0.34	22.58	7,672	0.38	13.25	6,231	0.35	72.53	2,294	0.33	15.81	1,491	0.42	18.05
2013	84,675	0.48	9.04	11,332	0.53	23.26	12,564	0.46	6.83	2,894	0.50	14.41	2,326	0.56	5.92
2014	95,949	0.42	6.30	13,747	0.47	4.30	14,961	0.42	7.25	2,878	0.44	9.33	2,422	0.48	13.76
2015	110,609	0.43	10.07	16,775	0.47	5.21	18,462	0.43	12.81	3,392	0.42	10.07	2,798	0.46	9.78
2016	97,984	0.47	6.17	16,021	0.50	4.78	19,143	0.46	5.53	2,996	0.44	5.54	2,380	0.49	4.44
2017	100,888	0.46	3.76	17,081	0.51	3.40	19,532	0.44	4.24	3,215	0.45	5.79	2,272	0.48	4.25
2018	124,126	0.48	4.47	21,058	0.53	3.75	21,400	0.45	6.78	4,530	0.49	5.13	2,736	0.47	5.17
2019	132,977	0.52	11.13	26,491	0.58	6.31	25,029	0.48	13.90	5,670	0.57	9.96	3,772	0.49	10.67
2020	176,154	0.50	17.30	34,962	0.55	11.91	28,091	0.49	22.03	7,079	0.54	22.81	4,547	0.47	18.37
2021	143,675	0.52	24.78	30,008	0.57	17.72	22,126	0.52	26.81	5,837	0.54	23.59	4,104	0.49	49.64
2022	130,582	0.50	8.82	27,907	0.56	5.73	19,437	0.50	11.61	5,840	0.51	9.12	3,831	0.46	15.12
Total	1,279,425	0.47	11.78	226,071	0.52	8.99	208,700	0.46	14.38	47,574	0.49	13.43	32,959	0.48	16.33

Table 4
Avg. lengths (characters / words) of the title and content of a question, and of the identified answer.

Length	All Questions			Movies			Music			Books			Games		
	title	cont.	answ.	title	cont.	answ.	title	cont.	answ.	title	cont.	answ.	title	cont.	answ.
# Char.s	83.42	420.70	41.58	92.67	483.90	43.77	77.17	365.92	37.13	92.54	614.41	58.01	80.86	520.73	37.37
# Words	13.86	81.67	11.83	15.14	95.08	11.11	12.65	67.16	11.75	15.11	118.41	16.05	13.07	100.10	10.59

3. Known-Item Question Performance Prediction on TOMT-KIS

We conduct a pilot study on TOMT-KIS with the seven pre-retrieval query performance predictors implemented in the qpptk toolkit[4]—including the state-of-the-art max-var approach [6, 7]. In the experiments, we use the Robust04 CIFF index[5] from the Open-Source IR Replicability Challenge [12] to compute corpus-relative values that the predictors need.

Following previous practice on evaluating query performance prediction [6], Table 5 shows the rank correlations of the predictors to the ground truth in form of Kendall's τ, Spearman's ρ, and Pearson's r (a score of 1 indicates perfect correlation, 0 indicates random correlation, and -1 indicates perfect inverse correlation). In all scenarios, the existing predictors only achieve correlation scores close to 0 which indicates that they are not suited for known-item question performance prediction. For the development of more effective known-item question performance prediction, our new TOMT-KIS dataset can form an ideal starting point.

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[4]https://github.com/Zendelo/QPP-EnhancedEval/tree/main/code/qpptk
[5]https://github.com/osirrc/ciff/blob/master/README.md

Table 5
Effectiveness of the seven pre-retrieval predictors implemented in qpptk (max-idf, avg-idf, scq, avg-scq, var, max-var, avg-var) on our five known-item question test sets (all questions, movies, music, books, games). Correlation to the ground truth rankings given as Kendall's τ, Spearman's ρ, and Pearson's r.

Category	max-idf			avg-idf			scq			avg-scq			var			max-var			avg-var		
	τ	ρ	r	τ	ρ	r	τ	ρ	r	τ	ρ	r	τ	ρ	r	τ	ρ	r	τ	ρ	r
All	.06	.06	.07	.02	.02	.07	.11	.16	.09	-.05	-.06	.02	.12	.18	.10	.08	.10	.03	-.06	-.07	.00
Movies	-.06	-.08	-.20	-.21	-.31	-.16	.03	.05	-.02	-.03	-.04	.04	.04	.06	-.00	.04	.06	.04	-.06	-.07	.01
Music	.04	.05	.02	.05	.07	.09	.07	.09	.02	-.06	-.08	-.01	.07	.09	.02	.01	.01	-.02	-.06	-.07	-.05
Books	.16	.18	.10	-.09	-.13	-.13	.04	.06	.01	-.11	-.16	-.15	.02	.03	.01	.00	-.00	-.03	-.11	-.16	-.16
Games	.15	.17	.18	.04	.06	.03	.05	.06	.03	-.06	-.09	-.07	.05	.07	.04	.05	.06	.05	-.05	-.05	.05

4. Conclusion

We have constructed the new large-scale TOMT-KIS dataset[6] for known-item question performance prediction by crawling the complete tip-of-my-tongue subreddit (1.28 million questions; 47 % with heuristically identified answers). As a proxy for the performance of a question, we use the time elapsed until the solving answer was posted. In a pilot study, none of the existing pre-retrieval query performance predictors implemented in the qpptk toolkit could really predict a known-item question's performance. Known-item question performance prediction thus is still not "solved" and forms an interesting subject for future research—with our dataset as a possible starting point. Other interesting directions could be to use TOMT-KIS as an enrichment of the training data provided by the organizers of the upcoming TREC 2023 ToT track.

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