Extracting Query Facets from Search Results: A Survey

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ABSTRACT: Web search queries are often ambiguous or multi-faceted, which makes a simple ranked list of results inadequate. To assist information finding for such faceted queries, we explore a technique that explicitly represents interesting facets of a query using groups of semantically related terms extracted from search results. As an example, for the query “baggage allowance”, these groups might be different airlines, different flight types (domestic, international), or different travel classes (first, business, economy). We name these groups query facets and the terms in these groups facet terms. We develop a supervised approach based on a graphical model to recognize query facets from the noisy candidates found. The graphical model learns how likely a candidate term is to be a facet term as well as how likely two terms are to be grouped together in a query facet, and captures the dependencies between the two factors. We propose two algorithms for approximate inference on the graphical model since exact inference is intractable. Our evaluation combines recall and precision of the facet terms with the grouping quality. Experimental results on a sample of web queries show that the supervised method significantly outperforms existing approaches, which are mostly unsupervised, suggesting that query facet extraction can be effectively learned.

KEYWORDS: Query Facet, Semantic Class Extraction, Multi-faceted Query

I. INTRODUCTION

An increasing amount of information consists of a combination of both structured and unstructured data. For example, patent documents contain structured properties such as inventors, assignees, class codes, and filing date, as well as a body of unstructured text. Helpdesk tickets store not only structured data such as the ticket’s originator, responsible party, and status, but also text describing the problem and its origin. Increasingly, enterprises want to run analytics [5, 27] on text to extract valuable structured information such as chemical compounds used in a patent and products mentioned in a helpdesk ticket. As a result, the number of unique structured properties in those data sets can be fairly large (from mid to high tens or even hundreds). Performing discovery-driven analysis on such data sets becomes challenging since a user may not know which properties to focus on. Ideally, a user would like to just type in some keywords into a system which would then guide him to areas of interest. A promising query interface for such mixed data is Faceted search, which is widely used by e-commerce sites such as amazon.com and shopping.com for querying their catalogs. For example, a user might enter “digital camera” in the keyword window of shopping.com for querying their catalogs. There are potentially thousands of matches, but only a few popular ones can be displayed on the screen. To assist navigation, the system also shows in a separate panel summaries of search results, such as a count of digital cameras in each range of price and resolution (we refer to properties such as price and resolution as facets).

II. LITERATURE SURVEY

1. Zhicheng Dou, Member, IEEE, Zhengbao Jiang, Sha Hu, Ji-Rong Wen, and Ruihua Song, February 2016, “Automatically Mining Facets for Queries from Their Search Results”.

In this paper, we study the problem of finding query facets. We propose a systematic solution, which we refer to as QD Miner, to automatically mine query facets by aggregating frequent lists from free text, HTML tags, and repeat regions within top search results. We create two human annotated data sets and apply existing metrics and two new combined metrics to evaluate the quality of query facets. Experimental results show that useful query facets are mined by the approach. We further analyze the problem of duplicated lists, and find that facets can be improved by modeling...
finegrained similarities between lists within a facet by comparing their similarities. We have provided query facets as candidate subtopics in the NTCIR-11 IMine Task.


In this paper, we propose an effective harvesting framework for deep-web interfaces, namely Smart Crawler. We have shown that our approach achieves both wide coverage for deep web interfaces and maintains highly efficient crawling. Smart Crawler is a focused crawler consisting of two stages: efficient site locating and balanced in-site exploring. Smart Crawler performs site-based locating by reversely searching the known deep web sites for center pages, which can effectively and many data sources for sparse domains. By ranking collected sites and by focusing the crawling on a topic, Smart Crawler achieves more accurate results. The in-site exploring stage uses adaptive link-ranking to search within a site; and we design a link tree for eliminating bias toward certain directories of a website for wider coverage of web directories. Our experimental results on a representative set of domains show the effectiveness of the proposed two stage crawler, which achieves higher harvest rates than other crawlers. In future work, we plan to combine pre-query and post-query approaches for classifying deep web forms to further improve the accuracy of the form classifier.


This paper presented a client-side privacy protection framework called UPS for personalized web search. UPS could potentially be adopted by any PWS that captures user profiles in a hierarchical taxonomy. The framework allows users to specify customized privacy requirements via the hierarchical profiles. In addition, UPS also performed online generalization on user profiles to protect the personal privacy without compromising the search quality. We proposed two greedy algorithms, namely Greedy DP and Greedy IL, for the online generalization. Our experimental results revealed that UPS could achieve quality search results while preserving users customized privacy requirements. The results also confirmed the effectiveness and efficiency of our solution. For future work, we will try to resist adversaries with broader background knowledge, such as richer relationship among topics (e.g., exclusiveness, sequentiality, and so on), or capability to capture a series of queries from the victim. We will also seek more sophisticated method to build the user profile, and better metrics to predict the performance (especially the utility) of UPS.

4. Anju G R1 and Karthik M2, Year 2016, “Mining Queries From Search Results: A Survey”. As the primary methodology of discovering question features, can be enhanced in numerous angles.

For instance, some semi administered bootstrapping list extraction calculations can be utilized to iteratively extricate more records from the top results. Particular site wrappers can likewise be utilized to concentrate on top notch records from legitimate sites. Including these rundowns may enhance both precision and review of inquiry features. Grammatical feature data can be utilized to further check the homogeneity of records and enhance the nature of inquiry aspects. We will investigate these points to refine aspects later on. We will likewise research some other related themes to discovering inquiry aspects. Great portrayals of question aspects might be useful for clients to better comprehend the features.

III. METHODOLOGY

We do not find any publicly available dataset for evaluating query facets. Therefore, we build two datasets from scratch. First, we build a service for finding facets, and invite human subjects to issue queries on topics they know well. We collect 89 queries issued by the subjects, and name them as “UserQ”. As this approach might induce a bias towards topics in which lists are more useful than general web queries, we further randomly sample another set of 105 English queries from a query log of a commercial search engine, and name this set of queries as “RandQ”. For each query, we first ask a subject to manually create facets and add items that are covered by the query, based on his/her knowledge after a deep survey on any related resources (such as Wikipedia, Freebase, or official web sites related to the query). We then aggregate the qualified items in the facets returned by all algorithms we want to evaluate, and ask the subject to assign unlabeled items into the created facets. New facets will be created for the items that are not covered by the existing facets. A facet named “misc” is automatically created for each query by the labeling system. Subjects can add the noisy or irrelevant items into this facet. The main purpose of creating this “misc” facet is to help subjects to distinguish between bad and unjudged items. During evaluation, “misc” facets are discarded before mapping generated facets to manually labelled facets.
We address the problem of finding query facets. A query facet is a set of items which describe and summarize query one important aspect of a query. Here a facet item is typically a word or a phrase. A query may have multiple facets that summarize the information about the query from different perspectives. In sample facets are for some queries. Facets are for the query “watches” cover the knowledge about watches in five unique aspects, including brands, gender categories, supporting features, styles, and colors. The query “visit Beijing” has a query facet about popular resorts in Beijing (Tiananmen square, forbidden city, summer palace, . . .) and a facet on travel related topics (attractions, shopping, dining, . . .). Query facets provide interesting and useful knowledge about a query and thus can be used to improve search experiences in many ways. First, we can display query facets together with the original search results in an appropriate way. Thus, users can understand some important aspects of a query without browsing tens of pages. Second, query facets may provide direct information or instant answers that users are seeking. For example, for the query “lost season”, all episode titles are shown in one facet and main actors are shown in another. In this case, displaying query facets could save browsing time. Third query facets may also be used to improve the diversity of the ten blue links. We can re-rank search results to avoid showing the pages that are near-duplicated in query facets at the top. Query facets also contain structured knowledge covered by the query, and thus they can be used in other fields besides traditional web search, such as semantic search or entity search.

V. PROPOSED SYSTEM

We propose aggregating frequent lists within the top search results to mine query facets and implement a system called QDMiner. More specifically, QDMiner extracts lists from free text, HTML tags, and repeat regions contained in the top search results, groups them into clusters based on the items they contain, then ranks the clusters and items based on how the lists and items appear in the top results. We propose two models, the Unique Website Model and the Context Similarity Model, to rank query facets. In the Unique Website Model, we assume that lists from the same website might contain duplicated information, whereas different websites are independent and each can contribute a separated vote for weighting facets. However, we find that sometimes two lists can be duplicated, even if they are from different websites. For example, mirror websites are using different domain names but they are publishing duplicated content and contain the same lists. Some content originally created by a website might be re-published by other websites; hence the same lists contained in the content might appear multiple times in different websites.

VI. CONCLUSION AND FUTURE WORK

We assume that the important aspects of a query are usually presented and repeated in the query’s top retrieved documents in the style of lists, and query facets can be mined out by aggregating these significant lists. We propose a systematic solution, which we refer to as QDMiner, to automatically mine query facets by extracting and grouping frequent lists from free text, HTML tags, and repeat regions within top search results.

REFERENCES