OBJECT SEARCH PATTERN APPLICATION USING QUERIES

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Abstract– This survey concern with the problem of determinizing probabilistic data which accept only deterministic input by accepting this type of input data to be stored in legacy systems. The automated data analysis/enrichment techniques Probabilistic data may be generated. These techniques are entity resolution, information extraction, and speech processing. The legacy application system may belong to the pre-existing web applications like Flicker, Picasa, etc. The big aim is to generate a deterministic representation of probabilistic data which optimizes the deterministic data quality on the end-application. Those type of determinization problem should be deployed in the context of two different data processing tasks— triggers and selection queries. Thresholding or top-1 selection techniques are traditionally used for determinization which lead to suboptimal performance for many applications. Instead this survey proposes a query-aware strategy and explains its advantages over existing solutions through a comprehensive empirical study over real and synthetic datasets.

Keywords– Determination, Uncertain Data, Data Quality, Query Workload, Branch and Bound Algorithm.

I. INTRODUCTION

Due to the advancement of cloud computing and the propagation of web-based applications, users need to store their data in various existing web applications. Frequently, user data is generated frequently through a variety of process namely signals processing, data analysis/enrichment techniques. This process could be done before being stored in the web applications. For example recent cameras support vision analysis to create tags such as indoors/outdoors, scenery, landscape/portrait, etc. Also these cameras often have microphones for users to speak out a descriptive sentence. This type of output should be processed by a speech recognizer to generate a set of tags to be associated with the photo [1].

The photo along with the set of tags/sub words should be streamed in real-time using wireless connectivity to Web applications. This method could be used in application like Flicker. Inserting those data into web applications introduces a new challenge. The big drawback is such type of automatically generated content is often confusing the people and also produces objects with probabilistic attributes. For instance

vision analysis may produce the result with tags [2], [3].Likewise Automatic Speech Recognizer (ASR) may create an N-best list or a confusion network of utterances [1], [4]. Those types of probabilistic data must be "determinized" before being stored in legacy web applications. The determinization problem should be referred as the problem of mapping probabilistic data into the corresponding deterministic representation.

Several approaches should be followed for a determinization problem. Two basic strategies namely most probable value / all the possible values of the attribute with non-zero probability could be used respectively. These are called the Top-1 techniques. For example a speech recognition system that generates a single answer/tag for each utterance can be viewed as using a top-1 strategy. Another method might be to choose a threshold τ and include all the attribute values with a probability higher than τ. Those types of approaches are being doubter to the end application because they often lead to suboptimal results.

A better approach among those is design customized determinization strategies which select a determinate representation which should optimize the quality of the end-application. For instance some type of end application supports triggers/alerts on a automatically generated content. Examples of those type of end-application includes publish/subscribe systems such as Google Alert, wherein users specify their subscriptions in the form of keywords (e.g., "California earthquake") and predicates over metadata (e.g., data type is video). Google Alert should forward all matching data items to the user based on the subscriptions details.

While considering a video about California Earthquake that is to be published on Youtube . The video contains a more number of tags that should be extracted using either automated vision processing and/or information extraction techniques applied over transcribed speech. Those type of automated tools may produce tags with probabilities like "California": 0.9, "earthquake":0.6, "election": 0.7, while the true tags of the video could be "California" and "earthquake".

The determinization process useful to associate the video with appropriate tags. The subscribers who are really interested in the particular video Example "California Earthquake" is notified while the other is not frustrated by irrelevant data. Thus, in the example above, the determinization process should be used to minimize metrics such as false positives and false negatives that result from a determinized representation of data.

In a Flicker application photos are uploaded automatically from cameras along with tags which may be generated based on speech annotation or image analysis. Flicker supports effective retrieval based on a photo tags. In that type of application users may be interested in choosing determinate representation which optimizes set-based quality metrics such as F-measure instead of minimizing the false positives/negatives rates.

II.   EXISTING SYSTEM

In 2012 R. Nuray-Turaen et.al proposes [5] and J. Li proposes [6] which explain giving deterministic answers to a query e.g. conjunctive selection query [5] over a probabilistic database. Unlike the problem of determinizing an answer to a query the big goal of this paper is to determine the data to enable and it should be stored in legacy deterministic databases such that the determinized representation optimizes the expected performance of queries in the future.

Advantage:-
- They explore how to determinate answers to a query over a *probabilistic* database.

Disadvantage:-
- Solutions in [5], [6] cannot be straightforwardly applied to such a determinization problem.
- The people should not aware of any prior work that directly addresses the issue of determinizing probabilistic data
- Most people are interested in best deterministic representation of data and not that of an answer to a query
- Existing end-applications that take only deterministic input which may not be possible in [5][6].
- These problems settings lead to different challenges.

Determinizing Probabilistic Data.

In 2008 R. Cheng et.al [13] address a problem which choose the set of uncertain objects to be cleaned, in order to achieve the best improvement in the quality of query answers. However the goal is to improve quality of single query and optimize quality of overall query workload. Also this survey focuses on how to select the best set of objects and each selected object is cleaned by using human clarification. In this all objects should be determinate automatically. These differences essentially lead to different optimization challenges.

In 2010 V. Jojic et.al proposes [14] and In 2012 D. Sontag proposes [15] .Both concentrate in the area MAP inference in graphical model which aims to find the assignment to each variable which should be jointly maximizes the probability defined by the model. The cost-based metric determinization problem can be potentially viewed as an instance of MAP inference problem. In this way the challenge leads to the developing fast and high-quality approximate algorithm for solving the corresponding NP-hard problem.

Probabilistic Data Models

In 2008 L. Antova proposes [16] this should be a variety of advanced probabilistic data models. This survey focuses on determinizing probabilistic objects, such as image tags and speech output, for which the probabilistic attribute model suffices. The determining probabilistic data stored in more advanced probabilistic models such.

Disadvantage:-
- Extending our work to deal with data of such complexity remains difficult.

Index term selection

In 2001 D. Carmel *et al* proposes [19]"Static index pruning for information retrieval systems," which deal with the problem of selecting terms to index document for document retrieval. A term-centric pruning method contains top postings for each item according to the individual score. Each item should have posting term appeared in an ad hoc search query.

In 2007 J. Li and M. Sun et.al proposes [19] "Scalable term selection for text categorization," propose a scalable item selection for text categorization. It should be based on coverage of the items. This paper concentrate on getting the right set of terms which is most relevant to document. In this problem, a set of possibly relevant terms and their relationship to the document is already given by other data processing techniques. The main goal is does not explore the relevance of terms to documents, but to select keywords from the given set of terms

to represent the document, such that the quality of answers to *triggers/queries* is optimized

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Query intent disambiguation.

Reference from [**17**]–[19] describe History of data in which a query workload and click graph have been used. In the research area of query intent disambiguation Query logs are used to predict more relevant terms for queries, or in other words more accurate intent of queries. The main goal is not to predict appropriate terms. In order to get the right keywords from the terms those are already created by automated data generation tool.Query workload in this setting is not a source of feature for classification but most useful to drive the optimization of the end performance of application. Thus techniques in these works are not directly applicable in the context of the problem domain we are addressing.

Query and tag recommendation.

[19] In 2010 I. Bordino et.al proposes query recommendation and tag recommendation]. Based on a query-flow graph representation of query logs a measure of semantic similarity between queries models should be developed which is used for the task of producing diverse and useful recommendations.

In 2010 Rae *et al.* [19] propose an extendable framework of tag recommendation, using co-occurrence analysis of tags used in both user specific context (personal, social contact, social group) and non user specific context (collective). The focus of these works is on how to model similarities and correlations between queries/tags and recommend queries/tags based on those information. However, our goal is not to measure similarity between object tags and queries, but to select tags from a given set of uncertain tags to optimize certain quality metric of answers to multiple queries.

III. PROPOSED SYSTEM

This paper studies the problem of deteminizing datasets with probabilistic attributes which should be generated by automated data analyses/enrichment techniques. This approach explodes a workload of triggers/queries to choose the "best" deterministic representation for two types of application. First type supports triggers on generated content and another that supports effective retrieval. This survey introduces the solution for the problem of *determinizing* probabilistic data. Given a workload of triggers/queries, the main challenge is to find the deterministic representation of the data which would optimize certain quality metrics of the answer to these triggers/queries

IV. ADVANTAGES OF PROPOSED SYSTEM

- This framework solves the problem of determinization by minimizing the expected cost of the answer to queries.
- The developed a branch and-bound algorithm that finds an approximate near-optimal solution to the resulting NP-hard problem.
- Produce a collection of objects to optimize set-based quality metrics, such as F-measure.
- We develop an efficient algorithm that reaches near-optimal quality
- The solutions should be extended to handle a data model where mutual exclusion exists among tags.
- The correlations among tags can be leveraged in these solutions to get better results.
- This proposes systems are designed to handle various types of queries.
- The proposed techniques are very efficient and reach high-quality results that are very close to optimal solution.
- They are robust to small changes in the original query workload.

V. CONCLUSION

In this survey we have considered the problem of determinizing uncertain objects to enable such data to be stored in pre-existing systems, such as Flicker, that take only deterministic input. The goal is to generate a deterministic representation that optimizes the quality of answers to queries/triggers that execute over the deterministic data representation. We have proposed efficient determinization algorithms that are orders of magnitude faster than the enumeration based optimal solution but achieve almost the same quality as the optimal solution. As future work, we plan to explore determinization techniques in the context of applications, wherein users are also interested in retrieving objects in a ranked order.

References

[1] D. V. Kalashnikov, S. Mehrotra, J. Xu, and N. Venkatasubramanian, "A semantics-based approach for speech annotation of images," *IEEE Trans. Knowl. Data Eng.*, vol. 23, no. 9, pp. 1373–1387, Sept. 2011.

[2] J. Li and J. Wang, "Automatic linguistic indexing of pictures by a statistical modeling approach," IEEE Trans. Pattern Anal. Mach. Intell., vol. 25, no. 9, pp. 1075–1088, Sept. 2003.

[3] C. Wangand, F. Jing, L. Zhang, and H. Zhang, "Image annotation refinement using random walk with restarts," in Proc. 14th Annu. ACM Int. Conf. Multimedia, New York, NY, USA, 2006.

[4]   B. Minescu, G. Damnati, F. Bechet, and R. de Mori, "Conditional use of word lattices, confusion networks and 1-best string hypotheses in a sequential interpretation strategy," in Proc. ICASSP, 2007.

[5]   R. Nuray-Turan, D. V. Kalashnikov, S. Mehrotra, and Y. Yu, "Attribute and object selection queries on objects with probabilistic attributes," ACM Trans. Database Syst., vol. 37, no. 1, Article 3,Feb. 2012.

[6]   J. Li and A. Deshpande, "Consensus answers for queries over probabilistic databases," in Proc. 28th ACM SIGMOD-SIGACTSIGART Symp. PODS, New York, NY, USA, 2009.

[7]   R. Cheng, J. Chen, and X. Xie, "Cleaning uncertain data with quality guarantees," in Proc. VLDB, Auckland, New Zealand, 2008.

[8]   V. Jojic, S. Gould, and D. Koller, "Accelerated dual decomposition for MAP inference," in Proc. 27th ICML, Haifa, Israel, 2010.

[9]   D. Sontag, D. K. Choe, and Y. Li, "Efficiently searching for frustrated cycles in map inference," in Proc. 28th Conf. UAI,2012.

[10] P. Andritsos, A. Fuxman, and R. J. Miller, "Clean answers over dirty databases: A probabilistic approach," in Proc. 22nd ICDE,Washington,

[11] D. Carmel et al., "Static index pruning for information retrieval systems," in Proc. 24th Annu. Int. ACM SIGIR, New Orleans, LA, USA, 2001.

[12] J. Li and M. Sun, "Scalable term selection for text categorization," in Proc. EMNLP-CoNLL, Prague, Czech Republic, 2007.

[13] X. Li, Y. Wang, and A. Acero, "Learning query intent from regularized click graphs," in Proc. 31st Annu. Int. ACM SIGIR,Singapore, 2008.

[14] E. Pitler and K. Church, "Using word-sense disambiguation methods to classify web queries by intent," in Proc. Conf. EMNLP, Singapore, 2009.

[15] A. Ashkan, C. L. Clarke, E. Agichtein, and Q. Guo, "Classifying and characterizing query intent," in Proc. 31th ECIR, Toulouse, France, 2009.

[16] J. Teevan, S. Dumais, and D. Liebling, "To personalize or not to personalize: Modeling queries with variation in user intent," in Proc. SIGIR, Singapore, 2008.

[17] I. Bordino, C. Castillo, D. Donato, and A. Gionis, "Query similarity by projecting the query-flow graph," in Proc. 33rd Int. ACM SIGIR, Geneva, Switzerland, 2010.

[18] A. Anagnostopoulos, L. Becchetti, C. Castillo, and A. Gionis, "An optimization framework for query recommendation," in Proc. 3rd ACM Int. Conf. WSDM, New York, NY, USA, 2010.

[19] A. Rae, B. Sigurbjörnsson, and R. V. Zwol, "Improving tag recommendation using social networks," in Proc. RIAO, Paris,France, 2010.