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Pavan Kapanipathi
Wright State University - Main Campus

Prateek Jain
Wright State University - Main Campus

Chitra Venkataramani

Amit P. Sheth
Wright State University - Main Campus, amit@sc.edu

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Hierarchical Interest Graph from Tweets

Pavan Kapanipathi¹, Prateek Jain², Chitra Venkataramani², Amit Sheth¹
¹Kno.e.sis Center, Wright State University, Dayton, OH.
²IBM TJ Watson Research Center, Yorktown Heights, NY.
{pavan, amit}@knoesis.org
{jainpr, chitrav}@us.ibm.com

ABSTRACT

Industry and researchers have identified numerous ways to monetize microblogs for personalization and recommendation. A common challenge across these different works is the identification of user interests. Although techniques have been developed to address this challenge, a flexible approach that spans multiple levels of granularity in user interests has not been forthcoming. In this work, we focus on exploiting hierarchical semantics of concepts to infer richer user interests expressed as a Hierarchical Interest Graph. To create such graphs, we utilize users' tweets to first ground potential user interests to structured background knowledge such as Wikipedia Category Graph. We then adapt spreading activation theory to assign user interest score to each category in the hierarchy. The Hierarchical Interest Graph not only comprises of users' explicitly mentioned interests determined from Twitter, but also their implicit interest categories inferred from the background knowledge source.

Categories and Subject Descriptors
H.0 [Information Systems]: General

Keywords
Hierarchical Interest Graph; Personalization; Social Semantic Web; Twitter; Wikipedia; User Profiles

1. INTRODUCTION

Twitter has emerged as a prominent medium for people to communicate opinions and interests regarding events and services. Automatically determining these interests from user’s topical discussions in tweets involve understanding the content of messages and finding topics and/or entities expressed in them¹. For example, preponderance posting of messages such as “Important to note, in current standoff, that Senate-passed plan IS a compromise, accepting lower spending levels passed by House.” indicate that the user is interested in US Government and American Politics.

Techniques such as Bag of Words and topic models do not perform so well on small, informal text as has been argued in [5]. Also, representing user interests as Bag of Concepts [3, 4] has been experimented by citing the advantages of knowledge-bases. However, exploiting these knowledge sources to generate user interests from tweets is an active area of interest. In this work, we present an approach that recognizes entities from tweets and exploits structured background knowledge to represent user interests as Hierarchical Interest Graph (HIG). The structured background knowledge in our case is the Wikipedia category hierarchy. Wikipedia category hierarchy provides an ability to infer user interests which are not explicitly mentioned in tweets. Consequently, from the above example tweet, it is possible to infer that the user is interested in the categories “US Government”, and “American Politics” through hierarchical relationships from “US Senate” and “US House of Representatives” that are mentioned via their popular names.

The HIG generated using our approach extends the existing personalization and recommendation systems by providing flexibility in selecting content with varying level of abstractness. Considering the above example, the HIG includes specific interests such as "US Senate" and "US House of Representatives" and also provides flexibility to leverage the semantically inferred broader topics of interests such as “US Government”, and “American Politics”.

2. APPROACH

The goal of our approach is to construct a Hierarchical Interest Graph (HIG) for a Twitter user. The two primary inputs for our system are: (1) Tweets of a user to determine the basic interests of the user, and (2) Hierarchical background knowledge that can be mapped with users’ basic interests to infer the HIG. We utilize the Wikipedia Hierarchy as the source of structured background knowledge. We opted for Wikipedia because of its vast domain coverage and timely updates.

The system performs the following steps as illustrated in Figure 1: (1) User Interests Generator spots and scores the Wikipedia Entities² from tweets of a user. (2) Interest Hierarchy Generator maps the scored Wikipedia Entities of interest to the Wikipedia Hierarchy to infer user’s HIG.

¹In this work, we consider users’ tweets as a representation of their interests. The common issue is that a user might tweet only about a limited set of his interests.

²Entities are the most specific (leaf) nodes in the hierarchy.
**User Interests Generator:** This module determines the most specific interests of the user from his/her tweets. The process includes: (1) recognizing Wikipedia entities from users’ tweets, and (2) scoring them to represent the extent of users’ interests for the entity. Entity recognition from tweets is performed using Zemanta web service\(^3\). We opted for Zemanta due to its superior performance as evaluated in [2] and their web service’s higher rate limit\(^4\). The scores of the recognized entities is determined based on its normalized frequency using the following equation:

\[ n_{fi} = \frac{\text{frequency}(e_i)}{\text{frequency}(e_{max})} \]

where \( e_{max} \) is the entity that is mentioned most number of times by the user.

**Interest Hierarchy Generator:** The final step is to generate the HIG for a user, given the Wikipedia Hierarchy and the user’s scored specific interests from User Interest Generator. In order to accomplish this task, (1) the specific interests are linked to their appropriate categories in the Wikipedia Hierarchy, and (2) an adaptation to Spreading Activation theory [1] is used spread the scores of the specific interests to the higher nodes in the hierarchy.

A naive spreading of scores (with empirical decays) up the Wikipedia hierarchy infers the HIG of the user. However, we discovered that it does not determine appropriate scores for categories in the hierarchy. Analyzing the scores determined by the naive approach, we found that the issues were an impact of the structure of the Wikipedia Hierarchy. To nullify the impact of the structure:

- **Normalizers:** In order to normalize the scores and hence reduce the propagation of scores up the hierarchy, we utilized the distribution of the nodes in the hierarchy. We experimented with the following two parameters:

  - **Bell:** Dividing by the no. of nodes in the child level.
    
    \[ F_i = \frac{1}{\text{nodes}(n_i+1)} \]

  - **Bell Log:** Log value of the distribution, to reduce the impact of high number of categories.
    
    \[ FL_i = \frac{1}{\log_{10}\text{nodes}(n_i+1)} \]

3[	ext{http://developer.zemanta.com/}]

4[We thank Zemanta for their support for our research.]

3. **EVALUATION**

We experimented the system with two activation functions that use the parameters explained in Section 2. (1) Bell (2) Bell Log. The evaluation was done by a user study with 37 participants. The participants were provided with the top-50 categories of interests (based on the scores) obtained by employing each of the above functions. The participants were asked to mark Yes/No/Maybe reflecting the relevance of the interest category. The Maybe option was provided due to the abstractness of some categories such as Category:Technology, Category:Sports that might not completely be a user’s interest. We then calculated the Mean Average Precision for the results obtained from each activation function as shown in Figure 2. We can then conclude that the system employed with Bell Log activation function performs better in scoring each category in the HIG.

4. **CONCLUSION**

In this work, we have introduced a novel representation of user interests as a Hierarchical Interest Graph and an approach to generate HIG for a twitter user. The approach leverages users’ tweet and introduces new parameters to adapt the spreading activation theory to score the interest categories. A user study has shown that the approach is practical and useful in determining hierarchical interests.

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5. **REFERENCES**


