Semantic Mining of Social Networks

Synthesis Lectures on the Semantic Web: Theory and **Technology**

Editors

Ying Ding, Indiana University Paul Groth, Elsevier Labs

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Synthesis Lectures on the Semantic Web: Theory and Application is edited by Ying Ding of Indiana University and Paul Groth of Elsevier Labs. Whether you call it the Semantic Web, Linked Data, or Web 3.0, a new generation of Web technologies is offering major advances in the evolution of the World Wide Web. As the first generation of this technology transitions out of the laboratory, new research is exploring how the growing Web of Data will change our world. While topics such as ontology-building and logics remain vital, new areas such as the use of semantics in Web search, the linking and use of open data on the Web, and future applications that will be supported by these technologies are becoming important research areas in their own right. Whether they be scientists, engineers or practitioners, Web users increasingly need to understand not just the new technologies of the Semantic Web, but to understand the principles by which those technologies work, and the best practices for assembling systems that integrate the different languages, resources, and functionalities that will be important in keeping the Web the rapidly expanding, and constantly changing, information space that has changed our lives. Topics to be included:

- Semantic Web Principles from linked-data to ontology design
- · Key Semantic Web technologies and algorithms
- Semantic Search and language technologies
- The Emerging "Web of Data" and its use in industry, government and university applications
- Trust, Social networking and collaboration technologies for the Semantic Web
- The economics of Semantic Web application adoption and use
- Publishing and Science on the Semantic Web
- · Semantic Web in health care and life sciences

Semantic Mining of Social Networks

Jie Tang and Juanzi Li 2015

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Semantic Mining of Social Networks

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ABSTRACT

Online social networks have already become a bridge connecting our physical daily life with the (web-based) information space. This connection produces a huge volume of data, not only about the information itself, but also about user behavior. The ubiquity of the social Web and the wealth of social data offer us unprecedented opportunities for studying the interaction patterns among users so as to understand the dynamic mechanisms underlying different networks, something that was previously difficult to explore due to the lack of available data.

In this book, we present the architecture of the research for social network mining, from a microscopic point of view. We focus on investigating several key issues in social networks. Specifically, we begin with analytics of social interactions between users. The first kinds of questions we try to answer are: What are the fundamental factors that form the different categories of social ties? How have reciprocal relationships been developed from parasocial relationships? How do connected users further form groups?

Another theme addressed in this book is the study of social influence. Social influence occurs when one's opinions, emotions, or behaviors are affected by others, intentionally or unintentionally. Considerable research has been conducted to verify the existence of social influence in various networks. However, few literature studies address how to quantify the strength of influence between users from different aspects. In Chapter 4 and in [138], we have studied how to model and predict user behaviors. One fundamental problem is distinguishing the effects of different social factors such as social influence, homophily, and individual's characteristics. We introduce a probabilistic model to address this problem.

Finally, we use an academic social network, ArnetMiner, as an example to demonstrate how we apply the introduced technologies for mining real social networks. In this system, we try to mine knowledge from both the informative (publication) network and the social (collaboration) network, and to understand the interaction mechanisms between the two networks. The system has been in operation since 2006 and has already attracted millions of users from more than 220 countries/regions.

KEYWORDS

social tie, strong/weak ties, parasocial interactions, reciprocity, social influence, collective classification, graphical model, social network analysis, social relationship, relationship mining, link prediction, influence maximization, network centrality, user modeling, social action, social theories, social balance, social status, triadic closure, factor graph, influence propagation, conservative influence propagation, non-conservative influence propagation, user behaviour prediction, profile extraction, expert finding, name disambiguation, ArnetMiner

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