

Can Scholarly Literature and Patents be Represented in a Hierarchy of Topics Structured to Contain 20 Topics per Level? Balancing Technical Feasibility with Human Usability

Michael Edwards¹, Mahadev Dovre Wudali², James Callahan³, Paul Worner⁴, Jeffrey Maudal⁵, Patricia Brennan⁶, Julia Laurin⁷ and Joshua Schnell⁸

¹michael.edwards@thomsonreuters.com

²mahadev.wudali@thomsonreuters.com

³jim.callahan@thomsonreuters.com

⁴paul.worner@thomsonreuters.com

⁵jeff.maudal@thomsonreuters.com

Data Center Operations, Thomson Reuters, 610 Opperman Drive, Eagan, Minnesota 55123

⁶patricia.brennan@thomsonreuters.com

⁷julia.laurin@thomsonreuters.com

⁸joshua.schnell@thomsonreuters.com

Intellectual Property & Science, Thomson Reuters,
1500 Spring Garden St, Philadelphia, Pennsylvania 19130

Introduction

The Intellectual Property & Science division of Thomson Reuters curates millions of records a year covering scholarly literature (Web of Science®), patents and intellectual property (Derwent World Patent Index®) and life sciences discovery (Cortellis®). These millions of records could be connected through billions of potential relationships, such as that represented by a citing relationship between literature and patents, or by different documents that pertain to similar topics. By building these relationships using machine learning techniques we hope to unite information from different data sources to enable extraction of knowledge such that the whole is greater than the sum of the parts, with minimal human effort required.

However, connecting these documents in a meaningful way is challenging from both a technological perspective as well as a usability perspective. As shown in Figure 1, studying citation patterns among approximately 250,000 articles from the Web of Science, or 1/200 of the full data set, generates a citation graph that, while rich with information, is extremely difficult to use to understand knowledge flows.

This challenge is the focus of our presentation. For this research project, we have created a graph of the topics represented in a subset of the scholarly literature and granted patents, in order to explore ways to constrain the visualization of this topic graph to emphasize usability. While many additional research areas remain, our initial findings suggest that such constraint enables users to easily explore the knowledge graph in way that maximizes understanding while minimizing user effort.

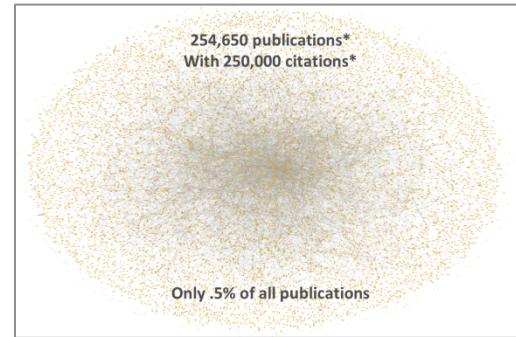


Figure 1. Ball and stick diagram of the citing relationships among a select set of publications from Web of Science®.

Generation of the Topic Graph

We chose to use topic modelling based on the latent dirichlet allocation (LDA) algorithm (Blei, Ng & Jordan, 2003) to generate connections between documents that reflect the shared knowledge among scholarly articles and granted patents. From Web of Science, we selected 27 million publications published since 1990 that had abstracts in English. Our past experience with LDA topic modelling led us to take a hierarchical approach to clustering the documents based on topics. We created a tree of over 1 million topics for the corpus, parceling out the topics into manageable chunks (20 at a glance) which were a better fit for human perception. We also created our own algorithm for applying these topics to patents, demonstrating a flexible, unsupervised technique for combining two distinct content sets. We found that the hierarchy we produced generally exhibited 4 to 5 levels of depth to the terminal nodes or documents.

Understanding the Knowledge Graph

We created the Epiphany tool to more effectively navigate the corpus of scholarly articles, using both browse and search interactions. As shown in Figure 2, the tool supports drill-down (e.g. 2.6 million articles assigned to an algorithm-focused topic; left side green), as well as search, (e.g. 8 topics strongly related to “genetic programming”; right side orange). This allows users to interact with topics and the relevant documents to understand the underlying data.

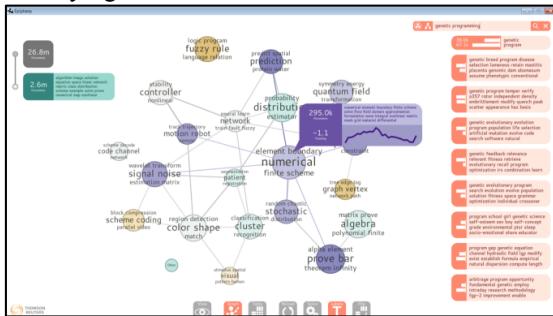


Figure 2. Screenshot of Epiphany tool showing topic clusters matching “genetic programming” search criteria.

Drilling down into the topic details is shown in Figure 3. At the top in purple are statistics on the topic itself including the number of documents closely associated with the topic, the most frequent terms and the Trending metric score for the topic.

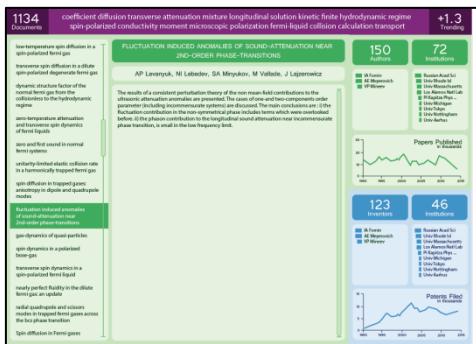


Figure 3. Screenshot of Epiphany tool Topic Details screen.

The right side of the panel contains two statistics sections, one in green for scientific papers and one in blue for patents. The header for each of the sections includes counts of the unique number of authors (or inventors) and unique number of institutions (or assignees) responsible for creation of the documents associated with the topic. Below these counts are a breakdown of the most commonly mentioned authors (inventors) and institutions (assignees). Finally, the bottom part of the statistics section is a graph of the proportion of documents assigned to this topic out of all documents published for each year.

Project Outcomes

The purpose of this research project is to test the application of scalable machine learning techniques to generate a knowledge graph that is accessible to the analyst. Now that we have developed the Epiphany tool, we have begun using it to gather feedback on this approach from a cross section of potential users. We expect to present that feedback at the ISSI2015 conference specifically to answer the question of whether a topic graph of millions of records of scholarly literature and granted patents can indeed be represented in hierarchical structure with a maximum of 20 topics at each level.

Acknowledgments

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References

Blei, D.M., Ng, A.Y., & Jordan, M.I. (2003). Latent Dirichlet Allocation. *The Journal of Machine Learning Research*, 3, 993-1022.