

VISUALISATION AND ANALYSIS OF RELATIONNAL DATA BY CONSIDERING TEMPORAL DIMENSION

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Keywords: Graph drawing, relational analysis, actor network, semantic network, morphing, alliance, evolutionary graph, space/time analogy.

Abstract: Visualization based on graph drawing allows the identification, the evaluation of passed and present structures between actors and concepts. It also allows the deduction of future ones. VisuGraph is developed in order to offer to the users the visualization and the interactive classification of relational data. We propose to complete this prototype with a morphing algorithm which animates with fluidity the representation between different time periods, emphasizing major elements and significant tendencies.

1 INTRODUCTION

Within strategic information framework, graphs appear as a tool of synthetic and intuitive representation of actor networks or semantic networks. Remarkable topologies are thus identified, revealing relationships between the various actors (authors, laboratories, companies, country) and the terms and/or concepts. Moreover, the study of the evolution of a network structure in time (collaborations, co-quotations, co-signatures, co-occurrences, alliances, fusions, acquisitions, licences,...) allows the evaluation of their last and current organizations. It makes it possible to deduce from them their future organizations and their implications in order to make a decision.

In this context, our research institute proposes the Tetralogie platform for the visualisation of relational data. Dedicated to macro analyses, it makes it possible (remotely and with several users) to carry out strategic analyses starting from heterogeneous textual data, by the means of methods of traditional or innovative data analysis. The prototype VisuGraph adds to this platform visualization and interactive classification of relational data.

In this article, we will focus on analysing the dynamicity of relational data networks considering their evolution.

Firstly, we present works on the visualisation of relational information in order to allow the analysis of their evolution. Secondly, within the framework

of the platform Tetralogie, we propose new functionalities for VisuGraph for the visualisation of the evolution of the various networks and to analyze the dynamics of their relations from a strategic point of view.

We implement a graph algorithm, which reveals the successive structures by animating the graph representation between various periods, significant changes and determining actors and/or concepts. The representation is based on spaces/time analogy used for a clock. The objective is to obtain an intuitive reading of the evolution by sequentially distributing the periods on a dial. The representation is animated in a similar way as how we play a video in a fast-forward mode. The strategic placement of the nodes allows then, not only, to locate them in time but also to evaluate their persistence and to deduce the evolution. We develop this approach while insisting on data structures, the optimization of graph drawing and its animation.

2 DATA VISUALISATION

The analysis of the evolution of relational information is typically based on techniques of dynamic graphs visualization.

Researchers have developed numerous network visualisation systems, (DiBattista et al., 1999), including internet connectivity maps, large networks of telephone calls, the structure of research shown as

The VisuGraph prototype, a Tetralogie module adds visualization and interactive classification of relational data, in a comprehensive way and by provides to the user a maximum of synthetic information.

3.2 Representation of Evolutionary Data

The relational data used result from information treatments under Tetralogie. Data are represented in matrices forms by crossed entities over several temporal homogeneous segments (or time periods).

Then our work consists in transforming these data into a networks representation, where the nodes represent entities and the links define the relations between them. It is possible to define a graph for each value of the temporal dimension.

This solution makes it possible to only analyze separately the period time compared to each other and never combined. Another approach consists in building the “total” graph G_{1-n} , like the combination of the graphs G_1, G_2, \dots, G_n of the n periods. The total graph is associated to a matrix resulting from the addition of the matrices of all the periods. The advantage of this representation is to dispose of a total sight of each data for each time period of the analysis: total positioning.

4 VISUGRAPH PROTOTYPE

According to Tufte (Tufte, 1983), “an excellent graph provides to the reader the maximum number of ideas in shortest lapse of time by using less ink and smallest space”. Based on this principle and on Karouach’s works (Karouach and Dousset, 2004) we propose to extend VisuGraph functionalities. Relations are represented using a graph whose nodes are the objects and the edges are the links comparable to springs.

4.1 Graph drawing

4.1.1 Force_Directed Placement

In order to place the nodes as well as possible, we have decided to use force directed placement functions applied on the nodes. These functions follow generally accepted aesthetic criteria for graph rendering, including evenly distributed vertices, minimized edges crossings, and uniform edge lengths.

According to Eades (Eades, 1984), a graph is comparable with a spring model while taking as a starting point the physical laws of graph drawing. This system generates forces between the nodes that involves their displacement. Attraction forces are calculated only for neighboring nodes and repulsive force are calculated for all pairs of nodes.

The attraction force between the nodes can be proportional to the force of the bond between them. The attraction force between two nodes v_i and v_j is given by:

$$f_a(v_i, v_j) = \beta_{ij} \times d_{ij}^{\alpha a} / K \quad (1)$$

β_{ij} is a function of the edge weight (v_i, v_j) and of the nodes weight v_i and v_j . The factor K is calculated according to the surface of the representation and the number of graph nodes. d_{ij} is the distance between v_i, v_j in the representation.

If the nodes v_i, v_j are not connected by an edge then $f_a(v_i, v_j) = 0$.

The repulsion force between two nodes v_i and v_j is defined by:

$$f_r(v_i, v_j) = -K^2 / d_{ij}^{\alpha r} \times \beta_{ij} \quad (2)$$

The variable αr (resp. αa) is a constant which is used to define the attraction degree (resp. repulsion) between two nodes.

Starting from an initial state of strong energy, we release the system until the nodes are harmoniously positioned without superimposing themselves. On the level of each node, the associated metric value is represented by one or several histogram bar. The static analysis can be at the origin of serious errors in the interpretation over a long period, especially if the visualisation is about non cumulative phenomena.

As a consequence, it appeared necessary to add to VisuGraph dynamicity for faithful and rigorous analyses.

4.1.2 Dynamic Case

In order to add to our prototype a dynamic aspect, we designed a morphing graph algorithm, which reveals the successive structures, by animating the graph between various time periods. It reveals the significant changes and it determines actors and/or concepts. The graph morphing allows to detect, include/understand and even to predict the significant tendencies, through the visualization of data evolution, while being based on the spaces/time analogy. In our case, nonvisible temporal references represent the various time periods. They are fixed in a chronological order and in an equidistant way on

the circumference of the display window (like the hours on a dial).

Graph is influenced by the attribution of new bonds connecting each node to the temporal reference marks, which are related to the time period considered. It generates a displacement, locating each node next to the marks of the time periods that it belongs. After stabilization of the graph, each peripheral sector of the window corresponds to a typology of particular evolution, only the center can contain several types of persistence (continuous presence or over a few spaced periods).

The graphs of various periods can be represented individually, by simply hiding nodes and bonds not concerned by the selected period. It is then possible to detect, for example, an emergent structure or an organisational change and to check the relevance of the following period level.

The representation of the nodes as evolution histogram makes it possible to locate them in time; For example, if the upper-left part (last reference) contains a majority of recent nodes, it is here where we must seek the famous weak signals and try to envisage their evolution.

The distribution of the other nodes is carried out randomly. In the dynamic approach, the drawing of the same total graph makes it possible to locate the nodes according to their specific periods.

5 CONCLUSION

VisuGraph appears as an ergonomic and powerful tool for dynamic data analysis which makes it possible to reveal, include/understand and anticipate the subjacent structures in order to identify their strategic implications. We have demonstrated the potential of an integrative approach to the visualization and analysis of a research field evolution. In particular, we have focused on various practical issues concerning detecting emerging trends and abrupt changes in transient research fronts. The encouraging results indicate that this is a promising line of research with the potentially wide-ranging benefits to users from different disciplines.

This prototype is on its first steps and requires some improvements. The nodes are strongly attracted by the temporal references, changing their first position which took care of the relations with the neighboring nodes. Thus, we would find a compromise for a more flexible animation of the movement between two time periods, then an adapted cinematic for each time period.

Moreover, this morphing is conditioned by the user point of view. It can be, for example, directed towards the detection of strong signals (important or

persistent) or weak signals (appearances, disappearances, reorganizations of actors which can be potentially interesting). Thus, we must locate the problems of each one precisely and draw the graph while taking the user interests into account.

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