

Quantitative Evaluation of Multi-Type Edge Bundling

Example for Japan Airmap

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Abstract: This paper describes an evaluation of multi-type edge bundling methods showing for different types of edges. Edge bundling methods such as force-directed edge bundling (FDEB) method have gained attention as one of graph drawing methods that reduce visual clutter. Also, a multi-type edge bundling methods have been proposed for multi-type graph that has an attached attribute to each edge. These methods are used for several cases and evaluated qualitatively. However, there is no cases to evaluate them quantitatively. This paper proposes one of the multi-type edge bundling methods extended from FDEB and visualizes the airline route map in Japan. After that, this paper evaluates them to know the features of each bundling method by using the three measures: mean edge length difference, mean occupation area, and edge density distribution.

1 INTRODUCTION

Recently, the utilization of network diagrams is a common technique in information visualization (Gansner et al. 1993). In utilizing network diagrams, observers can recognize data by looking at their relationships through connected links. The network diagram is popular in research because of the increasing popularity of social network services that utilize graph data that consists of nodes and links. In particular, the data of social network services comprise big data. However, as the number of nodes and links increases, graph visibility decreases due to the formation of visual clutter that accompanies the increases in the amount of data. This phenomenon is becoming increasingly pervasive, especially in today's big data era.

The graph layout approach has been proposed to reduce visual clutter (Mueller et al. 2006). The graph layout approach changes attributes such as the arrangement of elements and the type of line (line or curve). By correctly rearranging the nodes, graph visibility increases to a certain degree. However, this approach cannot solve the problem encountered when a graph contains enormous edges.

A new approach called edge bundling has been proposed to address this issue (Holten 2006, Zhou et al. 2008; Telea and Ersoy 2010). This method enables observers to recognize the main stream of edges through bundle edges based on certain rules. For

example, several methods based on the hierarchical structure of nodes, parallel coordinates, and mechanical models have been proposed. The mechanical bundling method presented in previous work has improved graph visibility by clarifying edge bundles.

Several graphs contain multi-type edges. For example, for an air route diagram where nodes are the airports and the links are the air routes, the differences in airline companies can be attributed as the types of edges (see Figure 5 in section 4). In another example, trend information can be expressed as different types of edges in the FACT-Graph (Saga et al. 2012), as shown in Figure 1. In this graph, the edge types are categorized based on the appearance from past to present, and different types of edges are illustrated in the graph.

In order to support the differences in edge type for edge bundling, Thus, previous methods do not support the decrease in visual clutter for the multi-type edge graph. Also there are no cases to evaluate the edge bundling quantitatively.

In this paper, we propose new edge bundling methods to treat multi-type edges. These methods bundle the edges of each type. We demonstrate these proposed methods by using Japanese airline flight route information, where the nodes are located in geographic information. and validate the usability of the method. Also we evaluate edge bundling results by using

Selassie et al. also introduced the parameter of compatibility, which depends on the number of edges in the minimum length path between edges P and Q. This parameter strictly limits the bundling for graphs with several subgraphs.

The potential minimum m_j and Coulomb's force F_c based on the inverted Lorentzian that works at p_i are defined as follows

$$m_j = \begin{cases} q_j & (\text{if } P \cdot Q > 0) \\ q_j + lN_j & (\text{if } P \cdot Q < 0) \end{cases} \quad (4)$$

$$F_c(p_i, q_j) = -sk_c |p_i - m_j| / \pi C (s^2 + |p_i - m_j|^2)^2 \quad (5)$$

where l , s , and k_c are the parameters, N_j is the vector that defines the direction of m_j , j is the index of the control point ($1 < j < C$), and C is the number of control points. Moreover, the complexity of divided edge bundling is the same with the complexity of the FDEB, $O(E^2C)$, because force computation uses the complexity reduction trick described in the section of FDEB

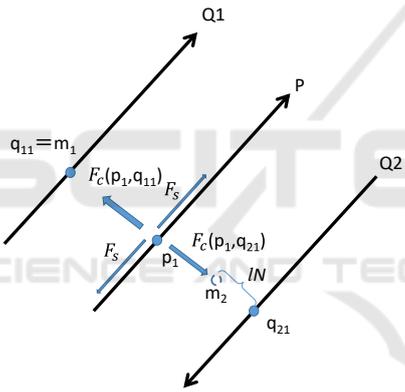


Figure 3: Divided Edge Bundling.

3 MULTI-TYPE EDGE BUNDLING METHOD

3.1 Assumption and Principle of Approach

In this paper, we assume that several edge types and various common edge types exist. For example, in the three edge types: A, B, or C, types A and B oppose one another, whereas Type C may belong to both types of edges ($C=A \cup B$). Hooke's law and gravitational force work among all pairs of edges. In addition, Coulomb's force works between pairs of the same attribution as an attractive force and between pairs of different attributions as a repulsive force. The

edges are bundled by the same edge type based on these forces. (Figure 4)

Furthermore, several pairs of edges are unsuitable for bundling. Thus, we introduce compatibility measures proposed in related studies to consider these pairs. In Holten's method, compatibility is calculated by the angle, scale, position, and visibility of pairs. In Selassie's method, compatibility is measured by the shortest path of edges, which severely limits bundling in disjoint edges. In this paper, we introduced Holten's compatibility measure because graph visibility is assumed to become clear when weights are added to improper pairs of edges. In this case, graph visibility becomes low when sparse parts of the graph are forcibly bundled. Furthermore, bundling disjoint edges in our data set is unnecessary. Therefore, we also introduced Selassie's compatibility measure

In this paper, we propose two edge bundling methods based on (1) the type compatibility approach and (2) Lorentz Coulomb's force approach.

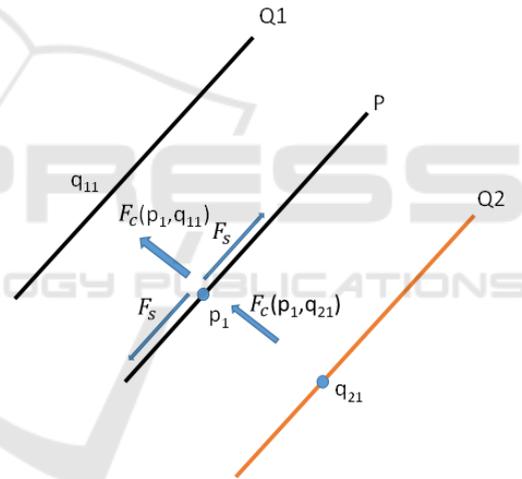


Figure 4: Basic Ideas in Multi-type Edge Bundling.

3.2 Type Compatibility based Edge Bundling

The first approach utilizes the simple idea that if two edges P and Q are of different edge types, then P and Q take inverse directions. To introduce this idea, we define a new coefficient called the type compatibility $C_T(P, Q)$ as follows:

$$C_T(P, Q) = \begin{cases} 1 & (\text{if } P \text{ and } Q \text{ are the same type edge}) \\ C & (\text{if } P \text{ or } Q \text{ contains another edge type}) \\ -1 & (\text{otherwise}) \end{cases} \quad (6)$$

The value of C can be set in several ways. In this research, we set C to 0.5. This compatibility can be

utilized together with other compatibilities for a simple application.

3.3 Lorentzian Coulomb's Force Approach

This approach is based on the divided edge bundling method. In this approach, we regard the direction of edges as the type of edges. Hence, Coulomb's force is customized in our method. We defined m_j and T_j as follows to represent the differences in the bundles of each attribution more clearly.

$$m_j = \begin{cases} q_j & (\text{if } P \text{ and } Q \text{ are the same type edge}) \\ q_j + lT_j & (\text{else}) \end{cases} \quad (7)$$

$$T_j = \begin{cases} p_i - q_j & (\text{if } P \text{ and } Q \text{ are the same type edge}) \\ q_j - p_i & (\text{else}) \end{cases} \quad (8)$$

where m_j is the potential minimum and T_j is the direction of the force. That is, if two edges are of the same type, then they are attracted to each other; if the edges are of different types, then they remain far away from each other

Handling edges as a spring is assumed to be practical. Hence, Hooke's law works in our method according to Equation (1). Moreover, the customized Coulomb's force $F_C'(p_i, q_i)$ is also effective. $F_C'(p_i, q_i)$ is defined as follows:

$$F_C'(p_i, q_i) = f(E_p) \cdot \frac{T_j}{|T_j|} \cdot \frac{-C_e(P, Q)sk_c|p_i - m_j|}{\pi C (s^2 + |p_i - m_j|)^2} \quad (9)$$

$$f(E_p) = \alpha \cdot E_p + \beta \quad (10)$$

where s , l , α , and β are the parameters, k_p is the spring constant, k_c is the Coulombic constant, C is the number of control points, and C_e is the compatibility between edges P and Q without type compatibility. Furthermore, E_p is the value of the current edge P with co-occurrences such as the Jaccard coefficient and the Simpson coefficient.

Considering the idea that an important edge should be the centre of the bundle, we adopt an edge weight into the force via $f(E_p)$. The total force F'_{p_i} at point p_i is as follows:

$$F'_{p_i} = \begin{cases} F_{p_i} & (\text{if } P \text{ or } Q \text{ is Type } C) \\ F_{p_i} + F_C'(p_i, q_i) & (\text{if else}) \end{cases} \quad (11)$$

When the current pair of edges contains attribution C , the force only behaves as a spring force.

When the pair consists of the same edge type, Coulomb's force works with an attractive force and the pair is bundled tightly. When the pair consists of different edge types, the force at work is repulsion.

4 APPLICATIONS

4.1 Dataset and Evaluation Method

We perform experiments to confirm the usability of proposed method for real data. In this experiment, we use airline flight route information in Japan, where the nodes are fixed based on geographic information in a manner different from the graph drawing methods such as the Kamada-Kawai layout (1989). We use airline flight route information in the year 2015 collected from the websites of All Nippon Airways (ANA), Japan Airline (JAL), other airlines, and Low Cost Carrier (LCC) like Peach Aviation. We also count the number of flights for each edge and use the normalized value of $\in[0, 1]$ as edge values.

The collected route map is shown in Figure 5. In this figure, an edge is regarded as a route between two airports. In this route information, cyan edges are the ANA information, magenta edges are the JAL information, black edges represent information shared by ANA and JAL, and yellow edges are others. That is, the successful result shows that (1) the same colors of edges are bundled, (2) cyan, magenta, and yellow edges are separated, and (3) black and cyan / magenta edges are bundled. Moreover, we run the FDEB for comparison. As parameters, we set α and β to 0.3 in Equation (9) and kc to 40000, l to 0.7, and s to 50 in equation (8).

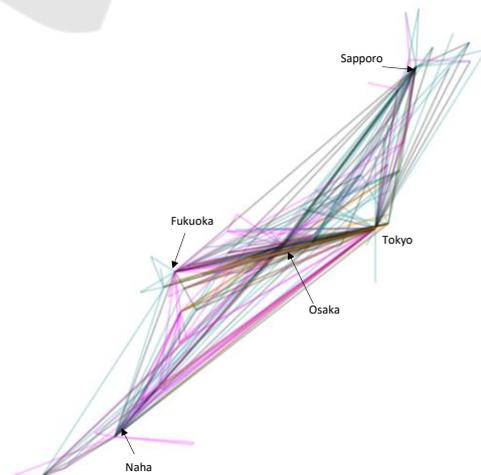


Figure 5: Original Airline Route Information

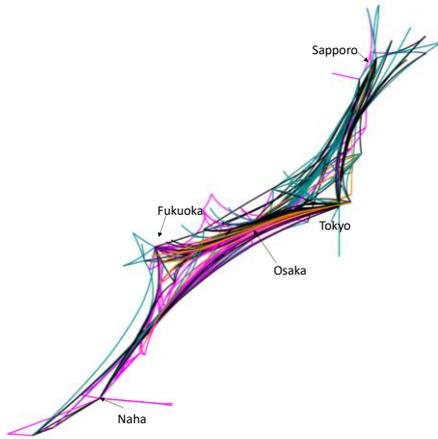


Figure 6: Result of Force-Directed Edge Bundling.

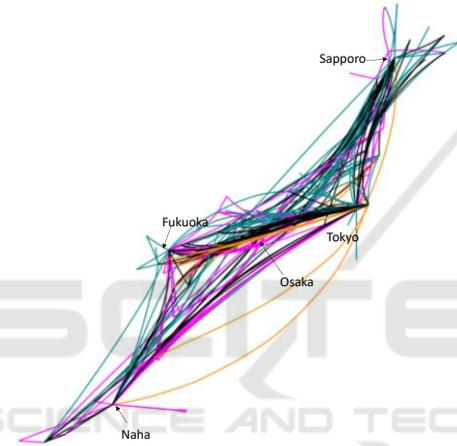


Figure 7: Result of Edge Bundling based on Type Compatibility.

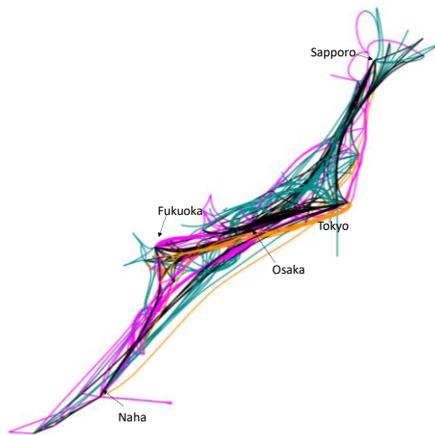


Figure 8: Result of Edge Bundling based on Inverted Lorentzian force.

4.2 Visualization Result

Figure 6 to 8 show the results of force-directed edge bundling and our proposed edge bundling methods. Force-directed edge bundling can bundle these edges but the bundling method, of course, ignore the edge type because this method don't consider edge type. In Figure 7 which show the result by type compatibility, the routes between Fukuoka and Tokyo and between Naha to Tokyo are separated clearly differently from Figure 6 although the same edges are close each other and bundled. However, some edge position are moved well, especially the edges from Naha to Tokyo/ Osaka move to outer and expand. On the other hand, in Figure 8 showing edge bundling by inverted Lorentzian force, our proposed methods do bundling among the same type edges but separate other edges.

From the results, we can understand our methods work well for this data. However, the evaluation is based on qualitative evaluation like "well" so that next quantitative evaluation is performed.

4.3 Quantitative Evaluation

This evaluation uses three measures for quantitative evaluation, mean edge length difference (MELD), mean occupation area (MOA), and edge density distribution (EDD) (Saga, 2016). MELD shows the difference among the lengths before and after edge bundling. In edge bundling, a lesser change in edge lengths is assumed to indicate superior edge bundling results because of over-bundling, wherein the large change of edge length often loses the meaning of the original network. MOA shows the degree among the compressed areas before and after edge bundling because better bundling can compress the area occupied by the edges. EDD is rooted on the idea that a better edge bundling method can gather edges within a unit area and that the density per unit is high. Based on these concepts, the measurements are calculated by the following Equations (11), (12), and (13):

$$MELD = \frac{1}{n} \sum_{e \in E} |L'(e) - L(e)| \quad (11)$$

$$MOA = \frac{1}{N} \left| \bigcup_{e \in E} O(e) \right| \quad (12)$$

$$EDD = \frac{1}{N} \sum_{a \in A} |p(a) - p| \quad (13)$$

where n is the number of edges, $L(e)$ is the length of an edge e before edge bundling, and $L'(e)$ is the length after edge bundling in Equation (11). In Equation

(12), N is the number of total areas, $O(e)$ is the set of occupied areas by edge e over an occupation degree (in this application, the value is 5% of unit area), and $||$ shows the number of elements contained by a set. In Equation (13), A is a set of unit areas, and $p(a)$ is the rate of the number of pixels, in which the edges pass in Area a . Moreover, \bar{p} is a mean of $p(a)$. Moreover, in Equations (12) and (13), the unit size is set to 6, that is, each unit area is 6 pixels by 6 pixels.

Table 1 shows the result of quantitative evaluation for each method and the original route information. As we said, the result is regarded as better if the MELD is low, MOA is low, and EDD is large. The result shows that the methods based on type compatibility and the inverted Lorentzian force are better than the original visualization. The best one except MELD is based on the inverted Lorentzian force. Therefore, edge bundling based on type compatibility shows a good average performance. Furthermore, edge bundling based on inverted Lorentzian force can bundle edges efficiently, although the length of the edges will increase.

Table 1. Quantitative Evaluation Result.

Measurement	Original	TC	ILF
MELD	n/a	2.329	7.621
MOA	0.200	0.208	0.181
EDD	0.091	0.091	0.096

Notes. TC: Edge Bundling based on Type Compatibility (Figure 7), ILF: Edge Bundling based on Inverted Lorentzian Force (Figure 8). MELD is calculated before and after edge bundling the images, that is, only the original image may not be calculated.

5 CONCLUSIONS

In this paper, we presented a multi-type of edge bundling as an extension of FDEB and divided edge bundling. We utilized two approaches, the type-compatibility and the Lorentzian Coulomb's force, to separate edges of different types. Applying the methods to airline route information validated the usability and superiority of our proposed methods through quantitative evaluation.

For future research, we must improve the visual encoding, interaction, and joint node/edge layout. Furthermore, if we will use big data, we must consider improving the complexity, as the complexity of our proposed methods is $O(E^2C)$. Also, in this evaluation, just thinking about the three measurements. However, we have to consider clarity to understand the bundling at ease. So we develop these points as future works.

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