

A Centrality Measure in Currency Exchange Rate (Ukuran Pemusatan Bagi Kadar Tukaran Matawang)

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ABSTRACT

We propose the average of weights of all links adjacent to each currency as a centrality measure. This measure, together with the existing centrality measures will be helpful in interpreting the network topology of currencies. A case study will be presented and discussed to illustrate the advantage of the proposed measure.

Keywords: Correlation matrix, Kruskal algorithm, minimum spanning tree, sub-dominant ultrametric

ABSTRAK

Kami mengusulkan purata pemberat bagi semua pautan bersebelahan dengan setiap mata wang sebagai ukuran memusat. Ukuran ini, bersama-sama dengan ukuran memusat sedia ada akan membantu dalam mentafsirkan topologi jaringan bagi kadar pertukaran matawang. Satu kajian kes akan dibentangkan dan dibincangkan untuk menggambarkan kelebihan dari ukuran yang diusulkan.

Katakunci : Matrik korelasi, algoritma Kruskal, pokok perentangan minimum, ultrametrik subdominan

INTRODUCTION

A currency such as Malaysian ringgit, US dollar, British pound sterling, etc., can be used for payments transaction within country. In any country whose residents conduct business abroad or engage in financial transactions with residents in other countries, they need for an exchanges of one currency to another currency, so that payments can be made in a form acceptable to foreigners. The exchange rates have direct influence on all other markets because the price of any asset is expressed in terms of a currency (Gorski *et al.*, 2008).

The role of importance of each particular currency can be express by the use of centrality measure such as degree centrality, closeness centrality and betweenness centrality. However, the problem with those measures is that they take into account only the number of direct and indirect links between two currencies. For that reason, we acknowledge the weight between two different currencies to propose a new measure of centrality which represents the “strength” of currency.

To illustrate the role of the propose measure, in this paper we conduct a study on the 32 currencies of highest global currency market turnover (Bank for International Settlements, 2010). For that purpose, those currency will be viewed as a complex system consisting of 32 currencies as nodes connected by $[32-1] * [32]/2 = 496$ links each of which corresponds to the correlation coefficient between two different nodes.

The nodes and links will be considered as a network or, more specifically, a weighted undirected graph (Jayawant & Glavin, 2009). This point of view is useful in order to visualize, simplify, and summarize the most important information contained in that complex system. The rest of paper is organized as follows. In Section 2 we discuss the analysis of that complex system by using the propose centrality measure as well as the current existing measures. To illustrate the advantage of the proposed measure, a case study on currency exchange rate will be presented and discussed in Section 3. At the end of this paper, we will draw attention to a conclusion.

PROPOSED CENTRALITY MEASURE

In this section we discuss the analysis of a complex system by using network topology approach. This allows us to visualize and simplify a complex system, and to summarize the information contained therein. We show that the centrality measures usually used as the principal tools to summarize the information are not sufficient. This motivate us to propose another measure.

Network Topology

The essence of a network is its nodes (currencies) and the way how they are linked. Network analysis was originally developed in computer science (De Nooy *et al.*, 2004). Nowadays, it has been used in various fields of study. See, for example, Krichel & Bakalbasi (2006) in sociology, Mantegna (1999) and Miccichè *et al.* (2003) in finance, and Park & Yilmaz (2010) in transportation.

In financial industry, network analysis starts with correlation matrix followed by transforming it into a distance matrix (Mantegna & Stanley, 2000). From this matrix we construct a minimum spanning tree (MST) and the corresponding sub-dominant ultrametric (SDU) distance matrix. For this purpose we use Kruskal algorithm (Kruskal, 1956) as suggested in Mantegna & Stanley (2000) and Jayawant & Glavin (2009). MST will then be used to construct network topology of currencies. This is a simplification of the complex system of currencies and their correlation structure which will be used to summarize the most important information. The visualization of MST can be made possible by using the open source called 'Pajek' (Batagelj & Mrvar, 2003; Batagelj & Mrvar, 2011 and De Nooy *et al.*, 2004). Furthermore, to interpret the MST we use the standard tools, i.e., centrality measures. To make the network topology more attractive and easy to interpret, we use the Kamada Kawai procedure provided in Pajek (Kamada & Kawai, 1989).

Centrality Measures

From network analysis view point, the role or degree of importance of each particular node can be analyzed by using its centrality measures such as degree, betweenness, and closeness centralities. These will help us to find the most important nodes in the network structure (Xu *et al.*, 2009; Abbasi & Altmann, 2010; Monárrez-Espino & Caballero-Hoyos, 2010).

Degree centrality indicates the connectivity of nodes. It provides information on how many other nodes are connected with a particular node. On the other hand, betweenness centrality is reflects the extent to which a node lies in relative position with respect to the others (Freeman, 1977). This measure indicates the potentiality of node to influence the others. Closeness centrality measures how close a node is to all other nodes in terms of correlations. Closeness can also be regarded as a measure of how long the information is to spread from a given node to other reachable nodes.

Those measures are computed based on the MST as follows (Borgatti, 1995; Sieczka & Holyst, 2009; and Park & Yilmaz, 2010):

- i. Degree centrality of node i is $d = \sum_{j=1}^n \alpha_{ij}$ where $\alpha_{ij} = 1$ if the i -th and j -th nodes are linked and 0 otherwise.

- ii. Betweenness centrality of node i , b_i , is the ratio of the number of path passing through i between two different nodes and the number of all possible paths from j to k for all j and k where $j \neq i$ and $k \neq i$.
- iii. Closeness centrality of node i , c_i , is the ratio of the number of links in the MST, which is equal to $(n-1)$, and the number of links in the path from i to j for all $j \neq i$.

Degree centrality is the simplest of the node centrality measures by using the local structure around nodes only. In order to identify the role of importance, degree centrality is no longer appropriate to be the best measure. The higher the degree centrality does not reflect to the strength of each particular node.

Due to that limitation of degree centrality, in this subsection we introduce “average of weights” as another measure. It is the average of weights of all links adjacent to each node. This measure reflects the strength of influence of a particular node to the others. The node that has larger scores in all measures is considered to be more central in terms of its influence to the others.

A CASE STUDY

As a case study, 32 currency exchange rate of highest global market turnover (Bank for International Settlements, 2010) will be analyzed. The sampling period of the data collection is taken in daily basis from February 1, 2000 to April 28, 2011 from Sauder School of Business (2011). In general there are 5 trading days per week and about 20 trading days per month. In this analysis, Gold Ounces (XAU) is used as the currency base. Based on the MST issued from Matlab version 7.8.0 (R2009a), see Figure 1, in Appendix we present the centrality measure discussed previously.

In Figure 1 (a) – (d) we present the MST together with the results in Appendix. The size and colour of the node represent the score of centrality measure and the rank of importance for degree centrality, betweenness centrality, closeness centrality and average of weight centrality. From that Figure, we learn that:

1. Based on degree centrality, see Figure 1 (a), SGD and MYR (red points) have the highest number of links (6) in the network. HKD (yellow point) and EUR (blue point) are 5 and 4 links respectively. Each of the followings has 3 links: MXN, DKK, and TWD (green points). The rests are of 1 and 2 links only. The higher the number of links, the higher the influence of a particular currency.
2. In terms of betweenness, see Figure 1 (b), SGD (red point) has an excellent position compared to the others where the information flow in the network can easily reach others. This node is the closest node to the others. The second (third, fourth, and fifth, respectively) closest node to the others are HKD (yellow point) (CNY (blue point), MYR (green point) and all purple points (USD, GBP and EUR), respectively).
3. According to the closeness centrality, see Figure 1 (c), the two most important nodes are SGD and CNY (red point) plays the most important role in the network followed by, in order of importance: HKD (yellow points or the second most important), GBP and USD (blue point or the third most important). This means that those currencies strongly influence the others.
4. Average of weight centrality can be used to indicate the average correlations between a particular node and the other nodes adjacent to it. In terms of degree, SGD and MYR are the most influential variables while in terms of sum of weights the most influential is TRY (red point), followed by ZAR, (yellow points), IDR and MXN (blue points) as well as BRL (green point). See Figure 1 (d).

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APPENDIX
Centrality measure

Node	Currency	Degree	Betwenness	Closeness	Average
1	USD	2	0.396	0.263	0.142
2	EUR	4	0.391	0.235	0.342
3	JPY	1	0	0.235	0.547
4	GBP	2	0.396	0.267	0.474
5	AUD	2	0.065	0.199	0.512
6	CHF	1	0	0.191	0.318
7	CAD	2	0.125	0.242	0.547
8	HKD	5	0.624	0.298	0.223
9	SEK	2	0.065	0.168	0.598
10	NZD	1	0	0.167	0.429
11	KRW	1	0	0.194	0.531
12	SGD	6	0.658	0.304	0.445
13	NOK	1	0	0.167	0.407
14	MXN	3	0.127	0.194	0.643
15	INR	1	0	0.189	0.301
16	RUB	1	0	0.189	0.539
17	PLN	1	0	0.163	0.487
18	TRY	1	0	0.163	0.806
19	ZAR	1	0	0.145	0.786
20	BRL	1	0	0.163	0.617
21	DKK	3	0.185	0.199	0.294
22	TWD	3	0.127	0.238	0.406
23	HUF	2	0.065	0.194	0.507
24	CNY	2	0.516	0.304	0.170
25	MYR	6	0.400	0.231	0.461
26	THB	1	0	0.235	0.348
27	PHP	1	0	0.194	0.427
28	CLP	1	0	0.235	0.519
29	IDR	1	0	0.189	0.644
30	ILS	1	0	0.231	0.414
31	COP	1	0	0.189	0.541
32	SAR	1	0	0.231	0.317