# CheiRank

The **CheiRank** is an eigenvector with a maximal real eigenvalue of the Google matrix G^* constructed for a directed network with the inverted directions of links. It is similar to the PageRank vector, which ranks the network nodes in average proportionally to a number of incoming links being the maximal eigenvector of the Google matrix G with a given initial direction of links. Due to inversion of link directions the **CheiRank** ranks the network nodes in average proportionally to a number of outgoing links. Since each node belongs both to **CheiRank** and PageRank vectors the ranking of information flow on a directed network becomes **two-dimensional**.

Nodes with links in the plane of PageRank and CheiRank
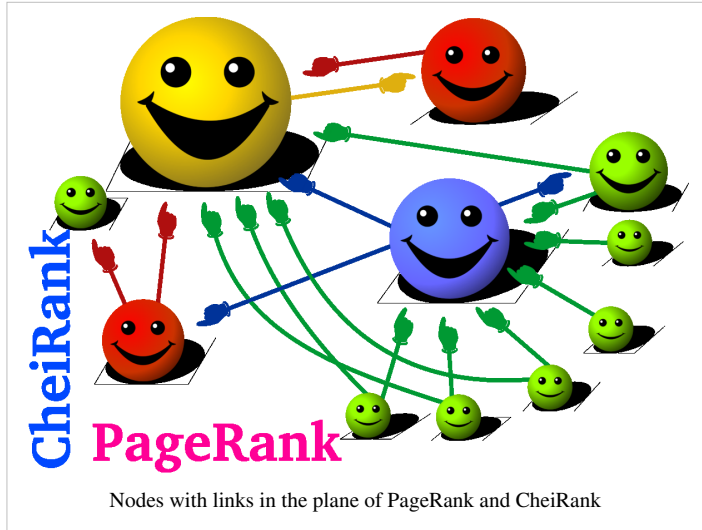
Definition

For a given directed network the Google matrix is constructed in the way described in the article Google matrix. The PageRank vector is the eigenvector with the maximal real eigenvalue $\lambda = 1$. It was introduce in [1] and is discussed in the article PageRank. In a similar way the CheiRank is the eigenvector with the maximal real eigenvalue of the matrix G^* built in the same way as G **but** using inverted direction of links in the initially given adjacency matrix. Both matrices G and G^* belong to the class of Perron–Frobenius operators and according to the Perron–Frobenius theorem the CheiRank P_i^* and PageRank P_i eigenvectors have nonnegative components which can be interpreted as probabilities [2,3]. Thus all N nodes i of the network can be ordered in a decreasing probability order with ranks K_i^*, K_i for CheiRank and PageRank P_i^*, P_i respectively. In average the PageRank probability P_i is proportional to the number of ingoing links with $P_i \propto 1/K_i^\beta$ [4,5,6]. For
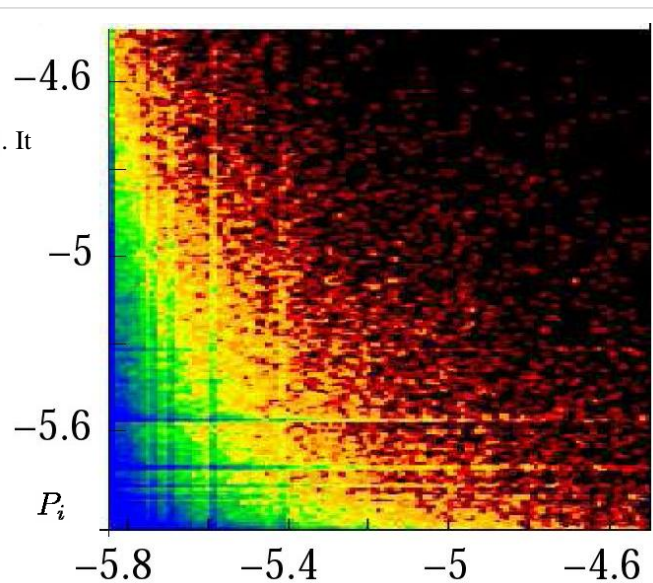

Fig1. Distribution of procedure calls of Linux Kernel network in the plane of PageRank probability $x = \log_{10} P_i$ and CheiRank probability $y = \log_{10} P_i^*$ for Linux version 2.6.32 with matrix size $N = 285509$ at $\alpha = 0.85$, color shows the density of nodes with white for maximum and blue for minimum, black space has no nodes (from [7])

the World Wide Web (WWW) network the exponent $\beta = 1/(\nu - 1) \approx 0.9$ where $\nu \approx 2.1$ is the exponent for ingoing links distribution [4,5]. In a similar way the CheiRank average proportional to the number of outgoing links with $P_i^* \propto 1/K_i^{*\beta^*}$ with $\beta^* = 1/(\nu^* - 1) \approx 0.6$ where $\nu^* \approx 2.7$ is the exponent for outgoing links distribution of the WWW [4,5]. The CheiRank procedure call network of Linux Kernel software in [6], the term itself was used in [7]. While the PageRank highlights very well known and popular nodes, the CheiRank highlights very communicative nodes. Top PageRank

and CheiRank nodes have certain analogy to authorities and hubs appearing in the HITS algorithm [9] but the HITS is query dependent while the rank probabilities P_i and P_i^* classify all nodes of the network. Since each node belongs both to CheiRank and PageRank we obtain a two-dimensional ranking of network nodes. It should be noted that there had been early studies of PageRank in networks with inverted direction of links [10,11] but the properties of two-dimensional ranking had not be analyzed in detail.

Examples

An example of nodes distribution in the plane of PageRank and CheiRank is shown in Fig.1 for the procedure call network of Linux Kernel software taken from [7].
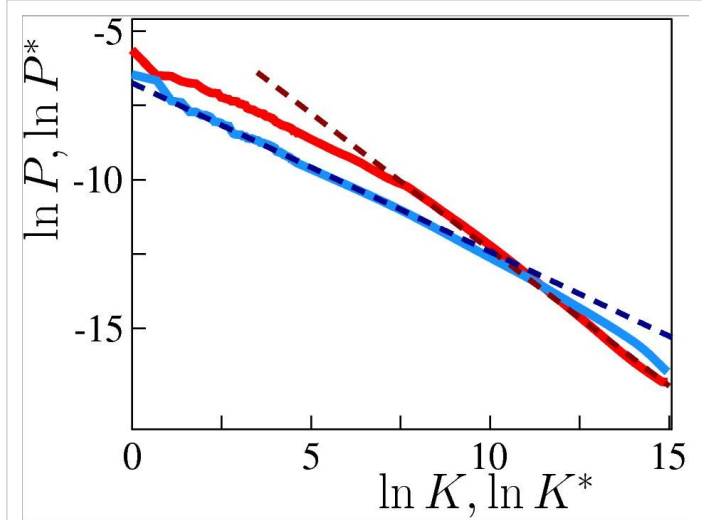
Fig2. Dependence of probability of PageRank P (red curve) and CheiRank P^* (blue curve) on the corresponding rank indexes K and K^*. The straight dashed lines show the power law dependence with the slope $\beta = 0.92; 0.57$ respectively, corresponding to $\beta = 1/(\nu - 1)$

(from [8])

The dependence of P, P^* on K, K^* for the network of hyperlink network of Wikipedia English articles is shown in Fig.2 from [8]. The distribution of these articles in the plane of PageRank and CheiRank is shown in Fig.3 from [8]. The difference between PageRank and CheiRank is clearly seen from the names of Wikipedia articles (2009) with highest rank. At the top of PageRank we have 1.United States, 2.United Kingdom, 3.France while for CheiRank we find 1.Portal:Contents/Outline of knowledge/Geography and places, 2.List of state leaders by year, 3.Portal:Contents/Index/Geography and places. Clearly PageRank selects first articles on a broadly known subject with a large number of ingoing links while CheiRank selects first highly communicative articles with many outgoing links. Since the articles are distributed in 2D they can be ranked in various ways corresponding to projection of 2D set on a line. The horizontal and vertical lines correspond to PageRank and CheiRank, 2DRank combines
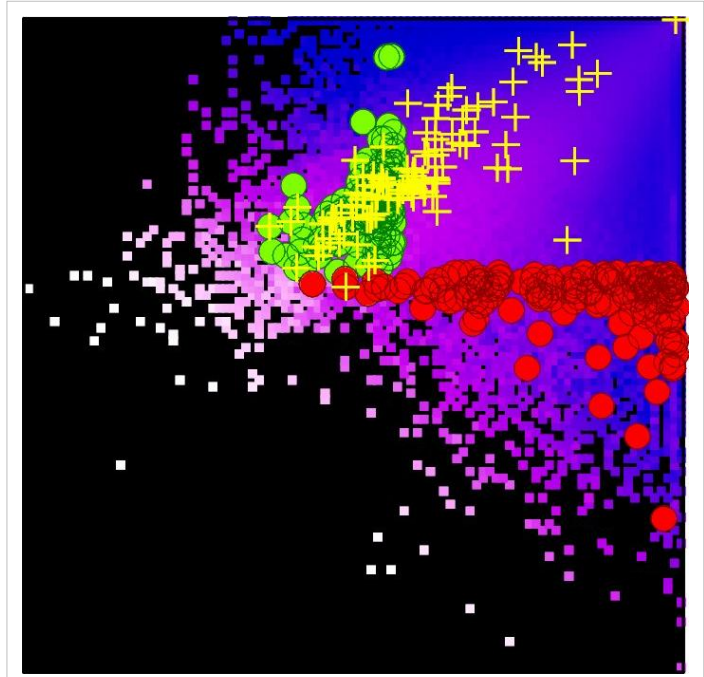

Fig3. Density distribution of Wikipedia English articles (2009) in the plane of PageRank and CheiRank indexes $0 < \ln K, \ln K^* < \ln N$ shown by color with blue for minimum and white for maximum (black for zero); green/red points show top 100 personalities from PageRank/CheiRank, yellow pluses show top 100 personalities from Hart's book, number of articles $N = 3282257$ (from [8])

properties of CheiRank and PageRank as it is discussed in [8]. It gives top Wikipedia articles 1.India, 2.Singapore, 3.Pakistan.

The 2D ranking highlights the properties of Wikipedia articles in a new rich and fruitful manner. According to the PageRank the top 100 personalities described in Wikipedia articles have in 5 main category activities: 58 (politics), 10 (religion),17 (arts), 15 (science), 0 (sport) and thus the importance of politicians is strongly overestimated. The CheiRank gives respectively 15, 1, 52, 16, 16 while for 2DRank one finds 24, 5, 62, 7, 2. Such type of 2D ranking can find useful applications for various complex directed networks including the WWW.

CheiRank and PageRank naturally appear for the world trade network, or international trade, where they and linked with export and import flows for a given country respectively [12].

Possibilities of development of two-dimensional search engines based on PageRank and CheiRank are considered in [13]. Directed networks can be characterized by the correlator between PageRank and CheiRank vectors: in certain networks this correlator is close to zero (e.g. Linux Kernel network) while other networks have large correlator values (e.g. Wikipedia or university networks) [7,13].

Simple network example

A simple example of the construction of the Google matrices G and G^*, used for determination of the related PageRank and CheiRank vectors, is given below. The directed network example with 7 nodes is shown in Fig.4. The matrix S, built with the rules described in the article Google matrix, is shown in Fig.5; the related Google matrix is $G = \alpha S + (1-\alpha)ee^T/N$ and the PageRank vector is the right eigenvector of G with the unit eigenvalue ($GP = P$). In a similar way, to determine the CheiRank eigenvector all directions of links in Fig.4 are inverted, then the matrix S^* is built, according to the same rules applied for the network with inverted link directions, as shown in Fig.6. The related Google matrix is $G^* = \alpha S^* + (1-\alpha)ee^T/N$ and the CheiRank vector is the right eigenvector of G^* with the unit eigenvalue ($G^*P^* = P^*$). Here $\alpha \approx 0.85$ is the damping factor taken at its usual value.
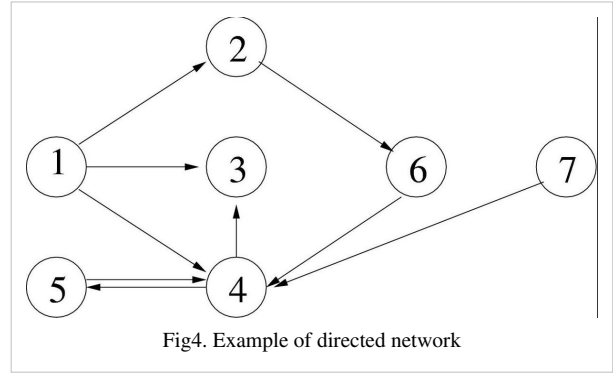

Fig4. Example of directed network

$$\begin{pmatrix} 0 & 0 & \frac{1}{7} & 0 & 0 & 0 & 0 \\ \frac{1}{3} & 0 & \frac{1}{7} & 0 & 0 & 0 & 0 \\ \frac{1}{3} & 0 & \frac{1}{7} & \frac{1}{2} & 0 & 0 & 0 \\ \frac{1}{3} & 0 & \frac{1}{7} & 0 & 1 & 1 & 1 \\ 0 & 0 & \frac{1}{7} & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 1 & \frac{1}{7} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{7} & 0 & 0 & 0 & 0 \end{pmatrix}$$

Fig5. Related matrix S

$$\begin{pmatrix} \frac{1}{7} & 1 & \frac{1}{2} & \frac{1}{4} & 0 & 0 & \frac{1}{7} \\ \frac{1}{7} & 0 & 0 & 0 & 0 & 1 & \frac{1}{7} \\ \frac{1}{7} & 0 & 0 & 0 & 0 & 0 & \frac{1}{7} \\ \frac{1}{7} & 0 & \frac{1}{2} & 0 & 1 & 0 & \frac{1}{7} \\ \frac{1}{7} & 0 & 0 & \frac{1}{4} & 0 & 0 & \frac{1}{7} \\ \frac{1}{7} & 0 & 0 & \frac{1}{4} & 0 & 0 & \frac{1}{7} \\ \frac{1}{7} & 0 & 0 & \frac{1}{4} & 0 & 0 & \frac{1}{7} \end{pmatrix}$$

Fig6. Related matrix S^*

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External links

- Two-dimensional ranking of Wikipedia articles (http://www.quantware.ups-tlse.fr/QWLIB/2drankwikipedia/)
- World trade: CheiRank versus PageRank (http://www.quantware.ups-tlse.fr/QWLIB/tradecheirank/)
- Towards two-dimensional search engines (http://www.quantware.ups-tlse.fr/QWLIB/dvvadi/)

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