Google matrix: fundamentals and applications



Dima Shepelyansky (Quantware group, LPT CNRS, Toulouse) www.quantware.ups-tlse.fr/dima

with L.Ermann (CNEA TANDAR), **Klaus Frahm (LPT UPS)**, K.Jaffres-Runser (N7 Tlse), J.Lages (U Besancon); A.Chepelianskii (LPS Orsay)



* Markov (1906) \rightarrow Brin and Page (1998) \rightarrow Google matrix and search engines * reduced Google matrix of directed networks (brief introduction)

* Applications: multiproduct world trade networks (UN COMTRADE), Wikipedia Ranking of World Universitis (WRWU), world terror networks, infectious diseases influence, protein cancer networks ...

Support: EC FET Open NADINE (2012-2015), APLIGOOGLE-CNRS (2016-2017), LABEX NEXT (2017-2019) + thanks to UN COMTRADE

(1906) Markov vs Wigner (1955)



1945: Nuclear physics \rightarrow Wigner (1955) \rightarrow Random Matrix Theory 1991: WWW, small world social networks \rightarrow Markov (1906) \rightarrow Google matrix

Despite the importance of large-scale search engines on the web, very little academic research has been done on them.

S.Brin and L.Page, Comp. Networks ISDN Systems 30, 107 (1998)

Google matrix construction rules

Markov chains (1906) and Directed networks



For a directed network with *N* nodes the adjacency matrix **A** is defined as $A_{ij} = 1$ if there is a link from node *j* to node *i* and $A_{ij} = 0$ otherwise. The weighted adjacency matrix is

$$S_{ij} = A_{ij} / \sum_k A_{kj}$$

In addition the elements of columns with only zeros elements are replaced by 1/N.

Google matrix construction rules

Google Matrix and Computation of PageRank $P = SP \Rightarrow P$ = stationary vector of S; can be computed by iteration of S.To remove convergence problems:

• Replace columns of 0 (dangling nodes) by $\frac{1}{N}$:

	(0	1/2	1/3	0	1/5	to 1	(0	1/3	0	0	0 \
	1	0	1/3	1/3	1/5		1/2	0	1/2	0	0
S =	0	1/2	0	1/3	1/5	$S^* =$	1/2	1/3	0	1	0
	0	0	1/3	0	1/5		0	1/3	1/2	0	1
	0	0	0	1/3	1/5		0	0	0	0	0/

• To remove degeneracies of $\lambda = 1$, replace **S** by **Google matrix**

 $\mathbf{G} = \alpha \mathbf{S} + (\mathbf{1} - \alpha) \frac{\mathbf{E}}{N}$; $GP = \lambda P$ => Perron-Frobenius operator

- α models a random surfer with a random jump after approximately 6 clicks (usually α = 0.85); PageRank vector => P at λ = 1 (Σ_i P_j = 1).
- CheiRank vector P^* : $G^* = \alpha S^* + (1 \alpha) \frac{E}{N}$, $G^*P^* = P^*$ (S* with inverted link directions) Chepelianskii arXiv:1003.5455 (2010) ...

Computation algorithms

* PageRank vector by power iteration:

multiplication of initial random vector by *G* matrix; convergence to $\lambda = 1$ eigenvector as α^t , about t = 200 iterations are enough for double precision convergence (all eigenvalues have $|\lambda| \le \alpha < 1$ except $\lambda = 1$); on average there are only about 10-20 nonzero links for each node (about 20 multiplications of vector by a line of matrix)

 \rightarrow small-world structure of real networks or six degrees of separation (Milgram Psychology Today (1967));

* Arnoldi algorithm: eigenvalues with largest $|\lambda|$ and related selected eigenvectors corresponding to quasi-isolated communities.

* Reduced Google matrix: description of interactions of subset of selected nodes in a huge network

* GPU codes for reduced Google matrix: 100 times acceleration compared to one-processor computer; collaboration with

Denis Demidov (Russian Academy of Sciences, Kazan; see https://github.com/ddemidov for GPU oriented codes)

* IT companies in Russia:

http://www.gradient-geo.com/en/index.php (Kazan, geo-oil); http://ledas.com/ (Novosibirsk)

Computation algorithms: spectrum of G

example: spectrum of G for ENWikipedia 2009 (3.28M nodes)



Ermann, Frahm, DS EPJB (2013)

Directed networks analyzed

* Wikipedia editions: EN (2009) *N* = 3282257;

24 editions Wiki2013: *N* = 4212493 EN, *N* = 1532978 DE, *N* = 1352825 FR 24 editions Wiki2017: *N* = 5416537 EN, *N* = 2057898 DE, *N* = 1866546 FR

- * Entier Twitter (2009): N = 41 millions
- * Entier Phys. Rev. citation network(1893-2009): N = 460422

* World Trade Network (WTN) from UN COMTRADE about 50 years: N = 227 for all commodities; multiproduct trade with 61 products N = 13847; available with 5000 products and $N \approx 1$ million

- * Bitcion network transactions (beginning 2009 till April 2013): N = 6297009
- * Linux Kernel network: N = 285509
- * UK university networks till 2006: U Oxford, Cambridge $N \approx 200000$
- * Network of protein-protein interactions for cancer: $N \approx 3000$

see Ermann, Frahm, DS Rev Mod Phys (2015) http://www.quantware.ups-tlse.fr/dima/subjgoogle.html

Top historical figures of 24 Wikipedia editions

2DRanking of Wikipedia articles; top 100 historical figures; comparison with historical studies of M.Hart (37 and 43 percent overlap) 35 centures and all countries by birth place; 17 millions wiki-articles



A.Zhirov, O.Zhirov, DS EPJB (2010); Y.-H.Eom,P.Aragon, D.Laniado, A.Kaltenbrunner, S.Vigna, DS PLoS ONE (2015)

Top historical figures of 24 Wikipedia editions

Top global PageRank historical figures: Carl Linnaeus, Jesus, Aristotle ...



theguardian

News Sport Comment Culture Business Money Life & style Jobs

4

8

Pi

in

Arti

Comment is free

And the winner of Wikipedia's influence list is ... an 18th century botanist. Hear hear

Carl Linnaeus is hardly a household name, but the Swedish doctor who created a global naming system for species deserves this accolade



Patrick Barkham theguardian.com, Friday 13 June 2014 09.00 BST Jump to comments (51)

Media highlights: The Guardian, The Independent, The Washington Post, France24, EC CORDIS ==>Uppsala Universitet: "Carl Linnaeus ranked most influential person of all time" ... (about 20 countries) Competitors: MIT Pantheon project http://pantheon.media.mit.edu (2014); Stony-Brook NY http://www.whoisbigger.com/ (2014)

Reduced Google matrix

A selected network of interest with $N_r < N$ nodes called reduced network. Block structure of *G* matrix:

$$G = \left(egin{array}{cc} G_{rr} & G_{rs} \ G_{sr} & G_{ss} \end{array}
ight)$$

with *s* index for scattering network $N_s = N - N_r$. Reduced G_R matrix

(analogy with quantum chaotic scattering, nuclear and mesoscopic physics)

$$G_{\rm R}P_r = P_r$$
 , $G_{\rm R} = G_{rr} + G_{rs}(1 - G_{ss})^{-1}G_{sr} = G_{rr} + G_{pr} + G_{qr}$

Useful expansion

$$(\mathbf{1}-G_{ss})^{-1} = \mathcal{P}_c \frac{1}{1-\lambda_c} + \mathcal{Q}_c \sum_{l=0}^{\infty} \bar{G}_{ss}^l$$

with projector $\mathcal{P}_{c} = \psi_{B}\psi_{L}^{T}$ on eigenstate of maximal eigenvalue λ_{c} of G_{ss} , the complementary projector $\mathcal{Q}_{c} = \mathbf{1} - \mathcal{P}_{c}$ and $\overline{G}_{ss} = \mathcal{Q}_{c}G_{ss}\mathcal{Q}_{c}$. K.Frahm, DS arxiv:1602.02394 (2016)

Protein-protein interactions for cancer networks

PLOS ONE

Reduced Google matrix for directed biological networks



Fig 1. Using reduced Google matrix approach for inferring hidden causal relations in signaling pathways. Here the structure of the contextdependent global regulatory network is symbolically shown as consisting of two layers: the upper (nodes A-K) is the global signaling network whose structure does not depend on the context and the lower is a symbolic view of the contextual transcriptional regulatory network (TRN) whose structure can change between a "normal" and a "cancer" cell. Thick node borders denote a pathway embedded into the global signaling network. Black arrows denote direct physical interactions. Pink arrows denote inferred hidden directed regulations through the global regulatory network (bath kayers). In the final representation of the pathway (on the right), one can show those hidden regulations through the global regulatory network (bath kayers). In the structure. Also, the color of the pathway nodes can show those hidden regulations through the effect of the pageRank decreased in the cancer network while red corresponds to the opposite.

https://doi.org/10.1371/journal.pone.0190812.g001

Lages, DS, Zinovyev (PLOS ONE 2018)

G-reduced: G20 political leaders 2012

G_R example: G20 political leaders 2012 from EnWiki2013



G-reduced: G20 political leaders 2012

G_R example: G20 political leaders 2012 direct links of G_{rr}



G-reduced: G20 political leaders 2012

 G_R example: G20 political leaders 2012 indirect links of G_{qr} (non-diagterms)



World trade network (WTN) of United Nations COMTRADE 1962-2010



Number of countries (black), links (dashed/points) and mass volume in USD (red)

Leonardo Ermann, DS arxiv:1103.5027 (2011); EPJB (2015)

PageRank, CheiRank of World Trade



Year 2008: Probabilities of PageRank P(K) (red), CheiRank $P^*(K^*)$ (blue) for all commodities (top) and crude petroleum (bottom), $\alpha = 0.5$; 0.85 (full/dotted); (dashed curves are for ImportRank, ExportRank); dashed line Zipf law $P \sim 1/K$; 227 countries

Ranking of World Trade



Ranking of World Trade



Rank table 2008 (74% of countries of G20)

Ran	K	K^*	K_2	$ ilde{K}$	\tilde{K}^*
1	USA	China	USA	USA	China
2	Germany	USA	China	Germany	Germany
3	China	Germany	Germany	China	USA
4	France	Japan	Japan	France	Japan
5	Japan	France	France	Japan	France
6	UK	Italy	Italy	UK	Netherlands
7	Italy	Russian Fed.	UK	Netherlands	Italy
8	Netherlands	 Rep. of Korea 	Netherlands	Italy	Russian Fed.
9	India	UK	India	Belgium	UK
10	Spain	Netherlands	Rep. of Korea	Canada	Belgium
11	Belgium	• Singapore	Belgium	Spain	🥚 Canada
12	Canada	• India	Russian Fed.	Rep. of Korea	•Rep. of Korea
13	Rep. of Korea	Belgium	Canada	Russian Fed.	Mexico
14	Russian Fed.	Australia	Spain	Mexico	Saudi Arabia
15	Nigeria	Brazil	Singapore	Singapore	• Singapore
16	Thailand	🔴 Canada	Thailand	India	Spain
17	Mexico	Spain	Australia	Poland	Malaysia
18	Singapore	South Africa	Brazil	Switzerland	Brazil
19	Switzerland	Thailand	Mexico	Turkey	 India
20	Australia	U. Arab Emir.	U. Arab Emir.	Brazil	Switzerland

Multiproduct WTN

Example: year 2008, $N_c = 227$, $N_p = 61$, N = 13847, import (top) - export (bottom) in millions USD



Multiproduct WTN: ranking of products

Democracy in countires, volume fraction in products => personalized vector in *G*. Left: 1993, right: 2008



CheiRank-PageRank balance (2008)

 $B_c = (P_c^* - P_c)/(P_c^* + P_c)$ (top - CheiRank-PageRank; bottom -Export-Import volume; multiproduct world trade $N_c = 227 countries, N_p = 61 products, N = 13847 ==> UN COMTRADE)$



Sensitivity to petrolem price (2008)

 $B_c = (P_c^* - P_c)/(P_c^* + P_c)$, color => $dB_c/d\delta_{petroleum}$)



G-reduced for EU sensitivity to RU-US gas

 $B_c = (P_c^* - P_c)/(P_c^* + P_c)$, color => $dB_c/d\delta_{gas}$) in 2016 RU/USA -left/right, 27 EU countries of 2008 (Coquide, Ermann, Lages, DS 2019)



Shaghai vs. Wikipedia World University Ranking

Top 100 (Shanghai ARWU2017 - left, wiki WRWU2017 - right), about 20 millions wiki-articles of 24 language editions



ARWU2017: 1.Harvard, 2.Stanford, 3.Cambridge, 4.MIT, 5.Berkeley (...74.Rice; USA: 37) WRWU2017: 1.Oxford, 2.Cambridge, 3.Harvard, 4.Columbia, 5.Yale (...357.Rice; USA: 56) 60percent overlap for top 100 Coquide, Lages, DS (arxiv 2018)

Overlap ARWU, WRWU, EnWiki, FrWiki, DeWiki

Left: overlap ARWU2017 with WRWU2017 (black), ARWU2013 with WRWU2013 (red); ARWU2103 with ARWU2017 (green), WRWU2013 with WRWU2017 (blue); Right: ARWU2017 with EnWiki2017 (black), FrWiki2017 (red), DeWiki2017 (blue)



Different cultural views on university ranking Coquide, Lages, DS (arxiv 2018)

World influence of Harvard University/EnWiki2017



Coquide, Lages, DS (arxiv 2018)

Reduced network of U-friends over centuries



Century reduced friendship network constructed for universities of PageRank top 100 list of WRWU computed from \bar{G}_{rr} + \bar{G}_{arnd} averaged over 24

Wikipedia editions. Coquide, Lages, DS (arxiv 2018)

(Quantware group, CNRS, Toulouse)

Э

Analysis of world terror networks (ENWiki2017)

reduced Google matrix for 64 world countries and 95 terrorist groups (from ENWiki2017 with 5.4 millions articles)



Friendship network structure from G_{qr} + G_{rr} with the top terrorist groups (marked by their respective colors) and countries (marked by cyan color); it is

shown with 2 friends for top terrorist groups of each category and top friend 2 countries for each group. El Zant, Frahm, Jaffres-Runser, DS (EPJB 2018) 2.

Analysis of world terror networks (ENWiki2017)

reduced Google matrix for 64 world countries and 95 terrorist groups (from ENWiki2017 with 5.4 millions articles)



PageRang sensitivity *D* to weight link variation ISIS - country $(dP_c/(P_c d\delta))$ El Zant, Frahm, Jaffres-Runser, DS (EPJB 2018)

HIV(AIDS) influence of world countries (ENWiki2017)

reduced Google matrix for 195 world countries and 230 infectious diseases (from ENWiki2017 with 5.4 millions articles)



Rollin, Lages, DS (bioRxiv 2018)

World impact of lung cancer (ENWiki2017)

reduced Google matrix for 195 world countries, 37 cancers and 203 drugs (from ENWiki2017 with 5.4 millions articles)



Towards bank financial network ranking



K.Soramäki *et al.*, *The topology of interbank payment flows*, Physica A **379**, 317 (2007) (only about 6000 nodes for all Federal Reserve USA) 2000 bank units for all Germany

* World network of flights:

network of flights between world cities (undirected ?); weighted with aviation fuel of flight (directed network); $N \sim 10^4$ (proposal of Nicolas Martin)

- * Industrial networks of control:
- e.g. interactions between sensors ?
- * Reduced Google matrix for Twitter network:

determine leading nodes, keywords and their interactions ...

* Financial flows and bank transactions: ?

* Network of protein-protein interactions in biological networks: SIGNOR https://signor.uniroma2.it/ open public REACTOME https://reactome.org/ open public METACORE https://clarivate.com/products/metacore/ private (request of millions of USD for the whole network of about 20000 nodes)

see available tools at Ermann, Frahm, DS Rev Mod Phys (2015) http://www.quantware.ups-tlse.fr/dima/subjgoogle.html

Further applications of Markov chains and Google matrix for ALTRAN? Artificial intelligence at Toulouse?



Google matrix: fundamentals, applications and beyond, IHES workshop 15-18 Oct 2018 (see www.ihes.fr)