Google matrix analysis of the World Trade Network (WTN)

Leonardo Ermann



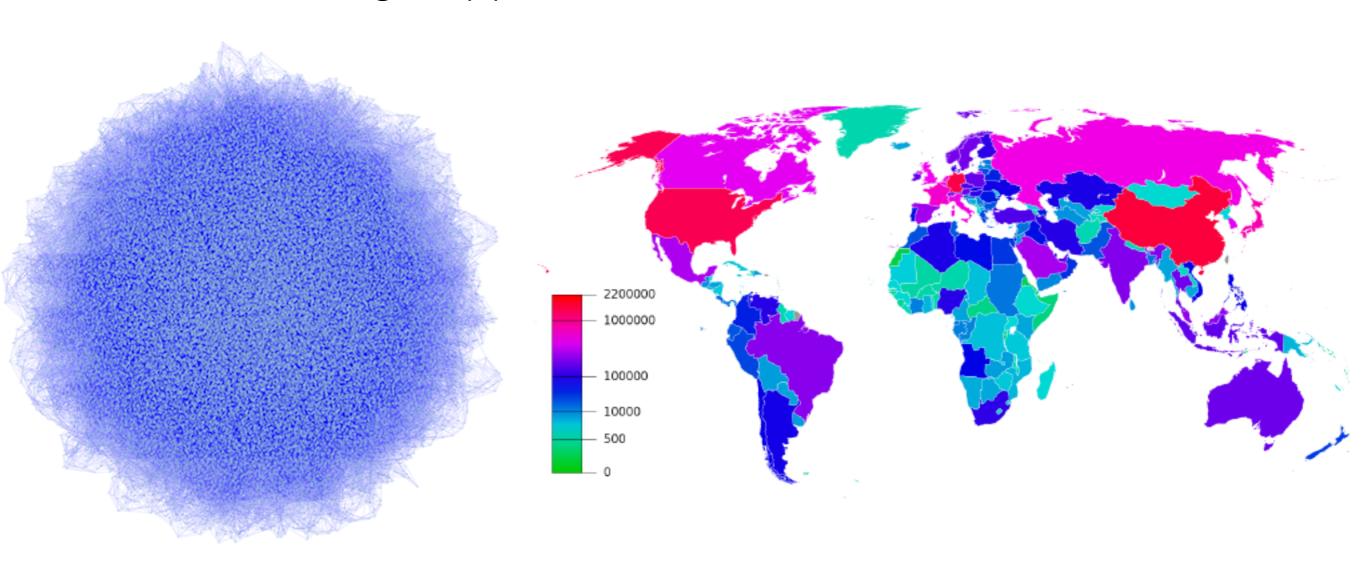
Colab. Dima Shepelyansky

Celestin Coquide, José Lages and Klaus Frahm

Google Matrix: fundamentals, applications and beyonds
October 17th, 2018
IHES, Paris, France

Motivations

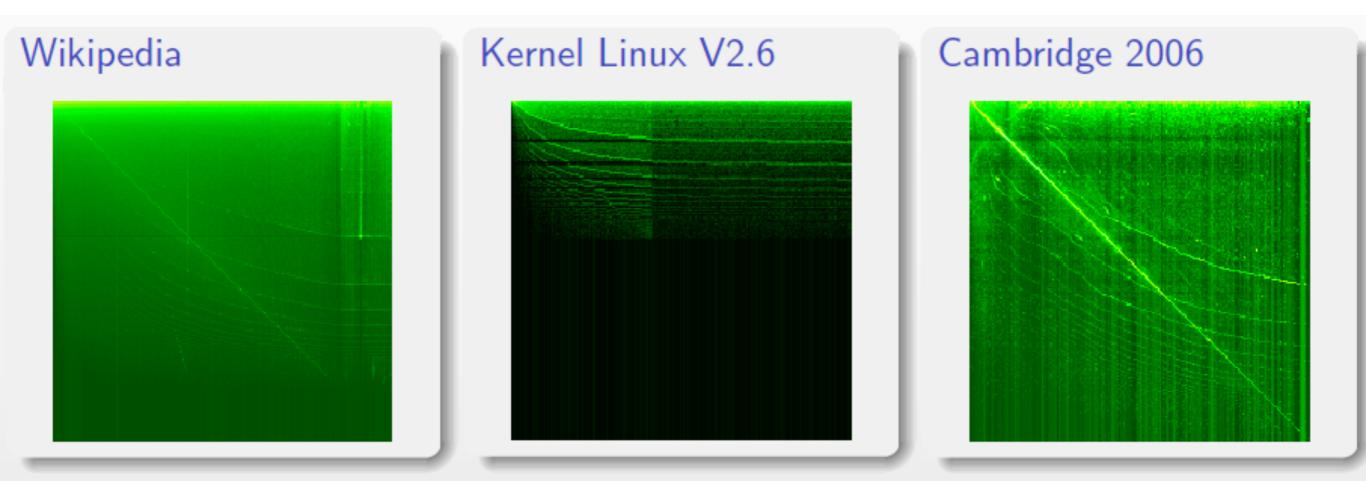
Google approach to the World Trade Network



Analyze, model, simplify, visualize and highlight useful information of large networks using nontrivial and global properties

Google Matrices

- www: Wikipedia (diff. lang.), Universities (UK, France)
- Operating Systems: Kernel Linux; NOMAO (spot recommendation)
- Bitcoin network
- Ulam method for Dynamical Systems: Standard Map



World Trade Network

Import-Export trade database:

United Nation Commodities Trade Network HTTP://COMTRADE.UN.ORG/DB/



- * Each year from 1962 to 2017
- * All UN countries: ~ 230 (Nc=227 in 2008)
- * Product classification (SITC Rev. 1): Np=61
- * Trade volume is given in USD ($N=13847 \times 50$ years)

Money Matrices

$$M_{c,c'} = \$ (c' \to c)$$
 $M_{c,c'}^p = \$ (c' \to c)$

Google Matrix of WTN

Brin and Page (1998)

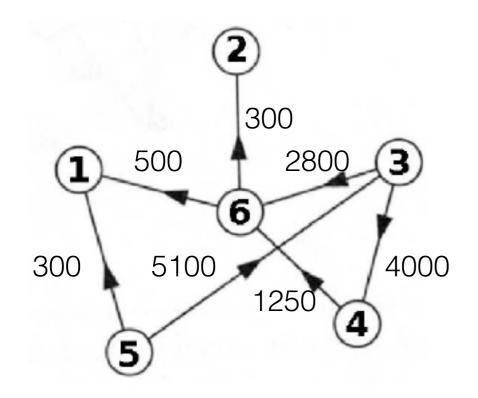
Introduced by Frahm yesterday

centrality measure:

Spectral Indices

- directed networks
- easy to compute
- incoming links
- non-local properties

directed network



Money matrix

$$M = \begin{pmatrix} 0 & 0 & 0 & 0 & 300 & 500 \\ 0 & 0 & 0 & 0 & 0 & 300 \\ 0 & 0 & 0 & 0 & 5100 & 0 \\ 0 & 0 & 4000 & 0 & 0 & 0 \\ 0 & 0 & 2800 & 1250 & 0 & 0 \end{pmatrix}$$

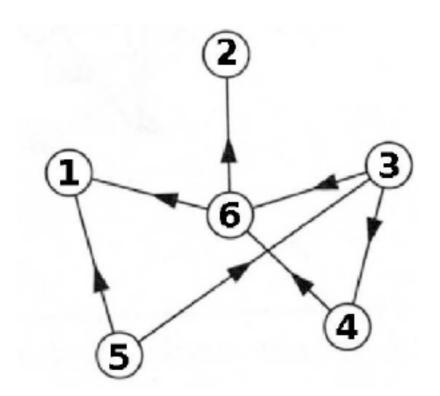
Google Matrix

Brin and Page (1998)

centrality measure:
Spectral Indices

- directed networks
- easy to compute
- incoming links
- non-local properties

directed network



weighted adjacency matrix and dangling nodes

$$S = \begin{pmatrix} 1/6 & 1/6 & 0 & 0 & 0.055 & 0.625 \\ 1/6 & 1/6 & 0 & 0 & 0 & 0.375 \\ 1/6 & 1/6 & 0 & 0 & 0.945 & 0 \\ 1/6 & 1/6 & 0.5888 & 0 & 0 & 0 \\ 1/6 & 1/6 & 0.412 & 1 & 0 & 0 \end{pmatrix}$$

$$-\sum_{i} S_{i,j} = 1$$

- Perron-Frobenius (non-negative)
- $\lambda_1 = 1$ (degeneracy)

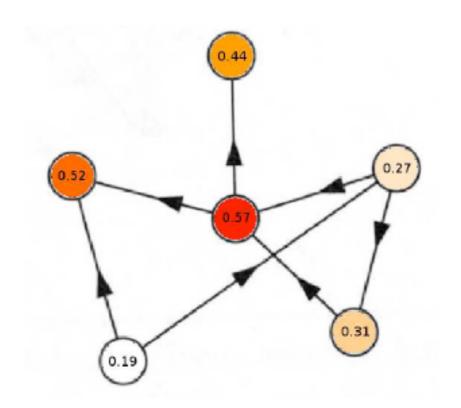
Google Matrix

Brin and Page (1998)

PageRank

$$\mathbf{G}P = P$$

directed network



centrality measure:

Spectral Indices

- directed networks
- easy to compute
- incoming links
- non-local properties

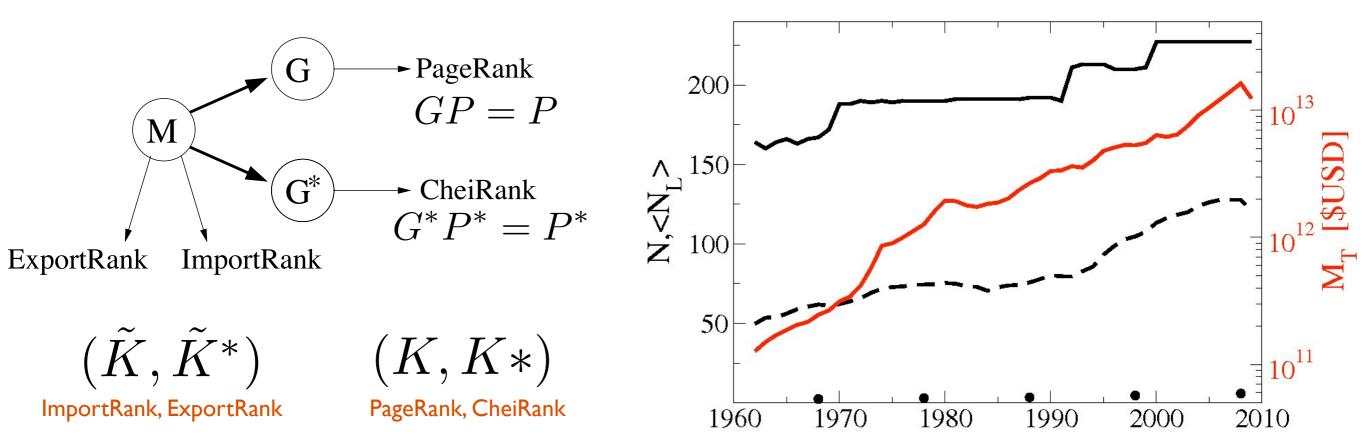
$$G = \alpha S + (1 - \alpha)E/N \ (\alpha = 0.85)$$

Google Matrix

$$G = \begin{pmatrix} 0.166 & 0.166 & 0.025 & 0.025 & 0.07175 & 0.556 \\ 0.166 & 0.166 & 0.025 & 0.025 & 0.025 & 0.344 \\ 0.166 & 0.166 & 0.025 & 0.025 & 0.828 & 0.025 \\ 0.166 & 0.166 & 0.5248 & 0.025 & 0.025 & 0.025 \\ 0.166 & 0.166 & 0.025 & 0.025 & 0.025 & 0.025 \\ 0.166 & 0.166 & 0.375 & 0.875 & 0.025 & 0.025 \end{pmatrix}$$

- $\alpha \to S, (1-\alpha) \to \text{random node}$
- Perron-Frobenius (positive) $\lambda_1 = 1$
- $\Delta \ge (1 \alpha)$ (global convergence)

Google matrix of the WTN



Democracy in countries but not in products

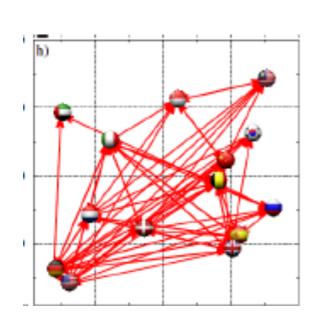
Google matrix of the WTN

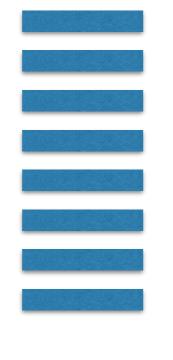
1) WTN (all com. or 1 prod) N=Nc=227

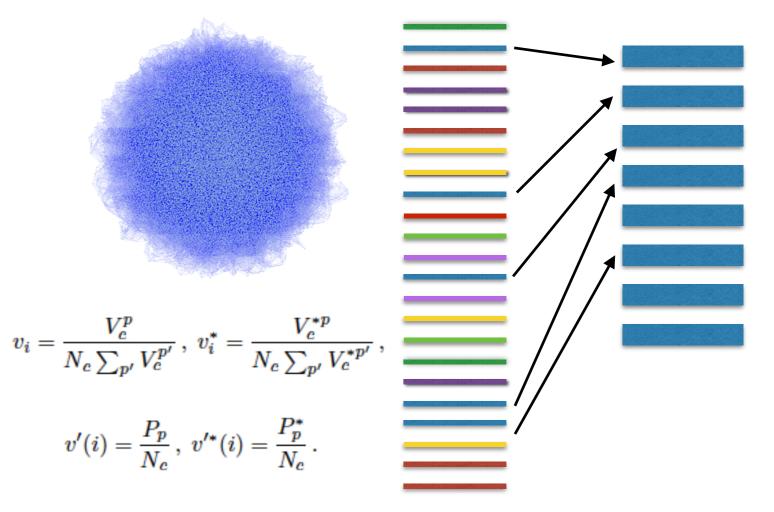
2) WTN (multiprod) Np=61; N=13847

non-interacting products

personalized vector → prop to Vp of each c 2nd iteration → reduced Pp for all c

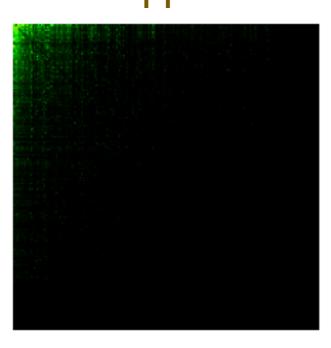




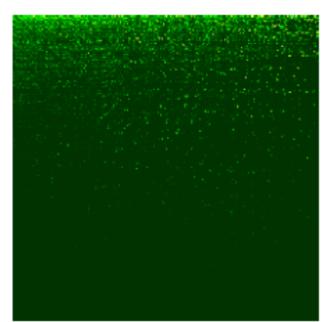


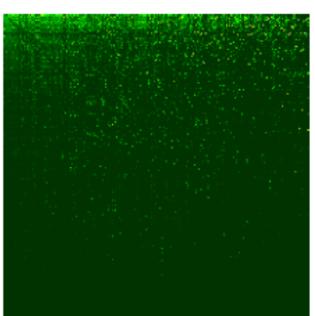
L. Ermann and D.L. Shepelyansky, APPA, Vol. 120, A-158 (2011), http://www.quantware.ups-tlse.fr/QWLIB/tradecheirank

L. Ermann and D.L. Shepelyansky, EPJB (2015).

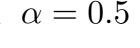


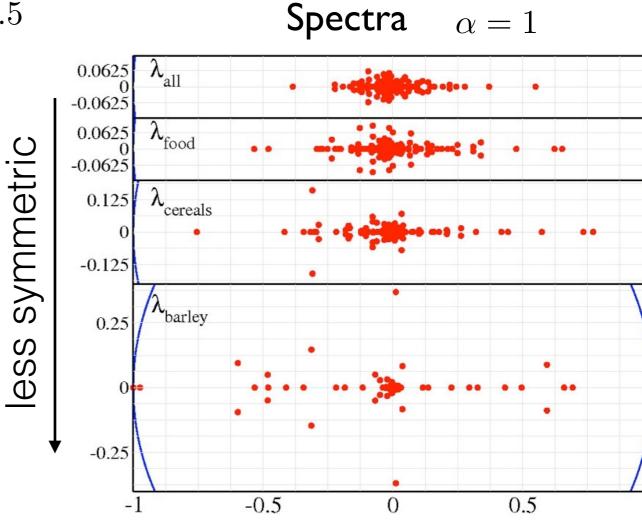
all commodities



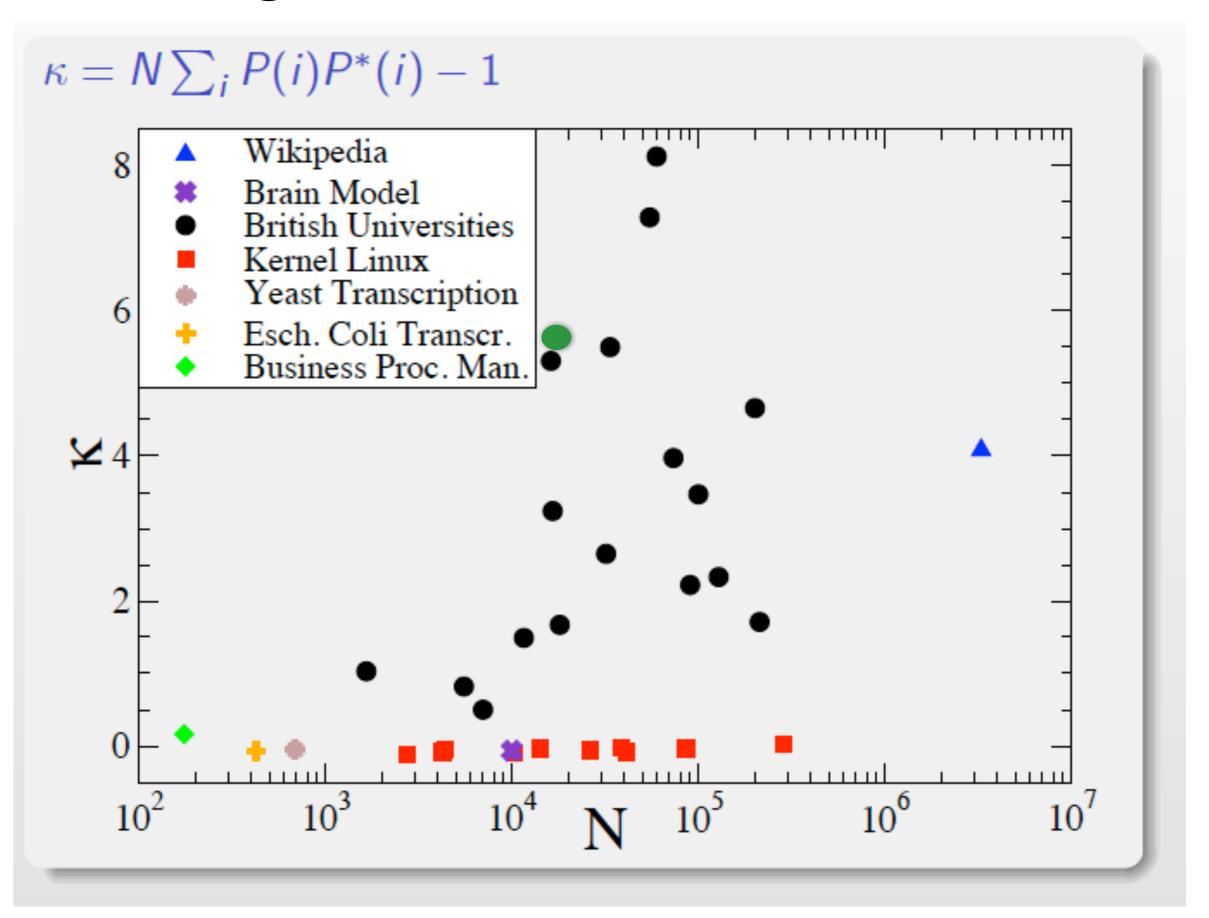


PageRank, CheiRank, ImportRank, ExportRank $~\alpha=0.5$ Zipf law $P \stackrel{\exists}{\sim} I/K$ 10^{-2} 10^{-3} *d, 10-4 10^{-2} 10^{-3} 10^{-4}

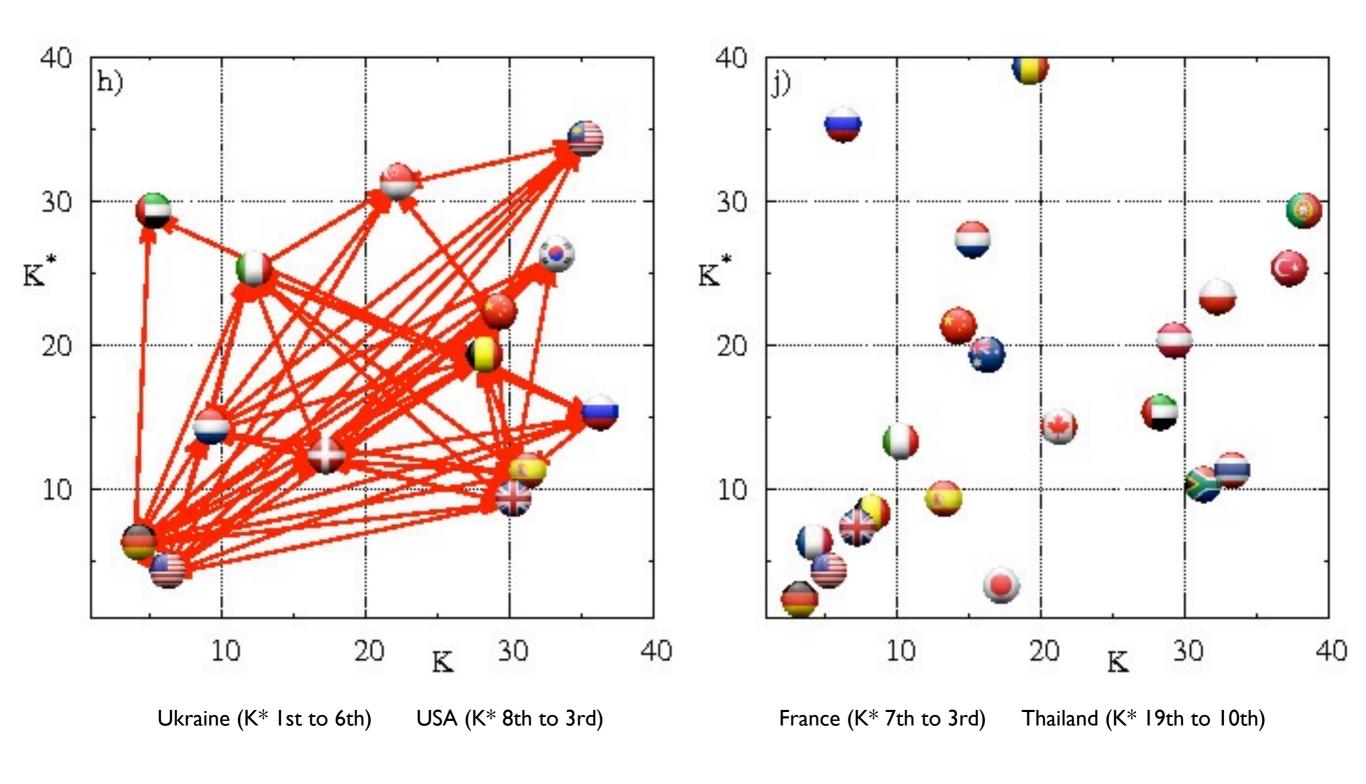




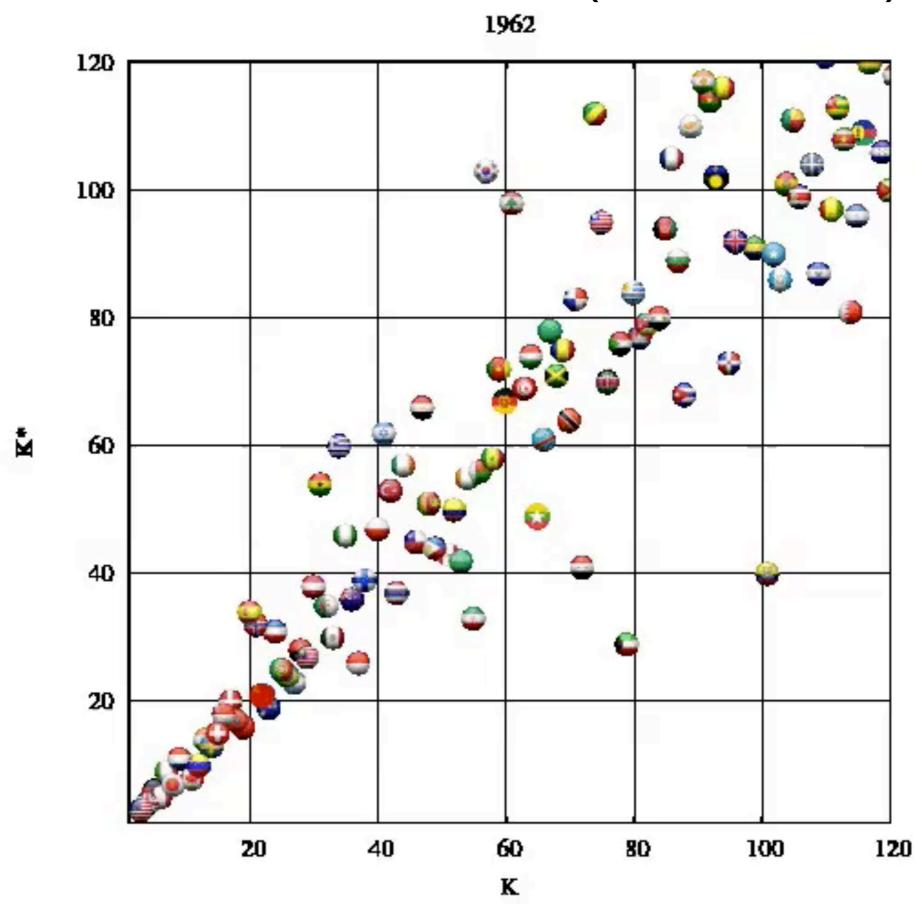
PageRank CheiRank correlator



2D rank examples: barley and cars



2d rank evolution (all comm.)



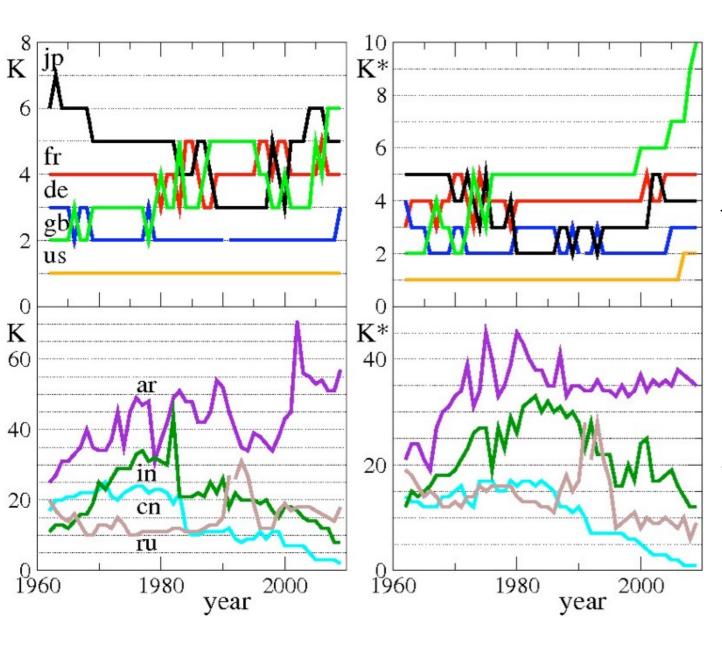
2d rank evolution

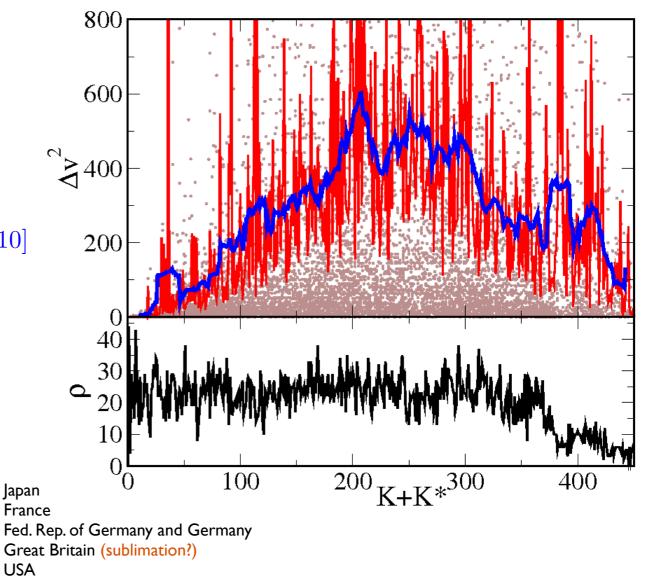
Velocity square vs. K+K*

$$\Delta v^2 = [K(t) - K(t-1)]^2 + [K^*(t) - K^*(t-1)]^2$$

average per $K + K^*$

average in $[K + K^* - 10, K + K^* + 10]$





crisis detection

Argentina India China (deposition) USSR and Russian Fed.

WTN model

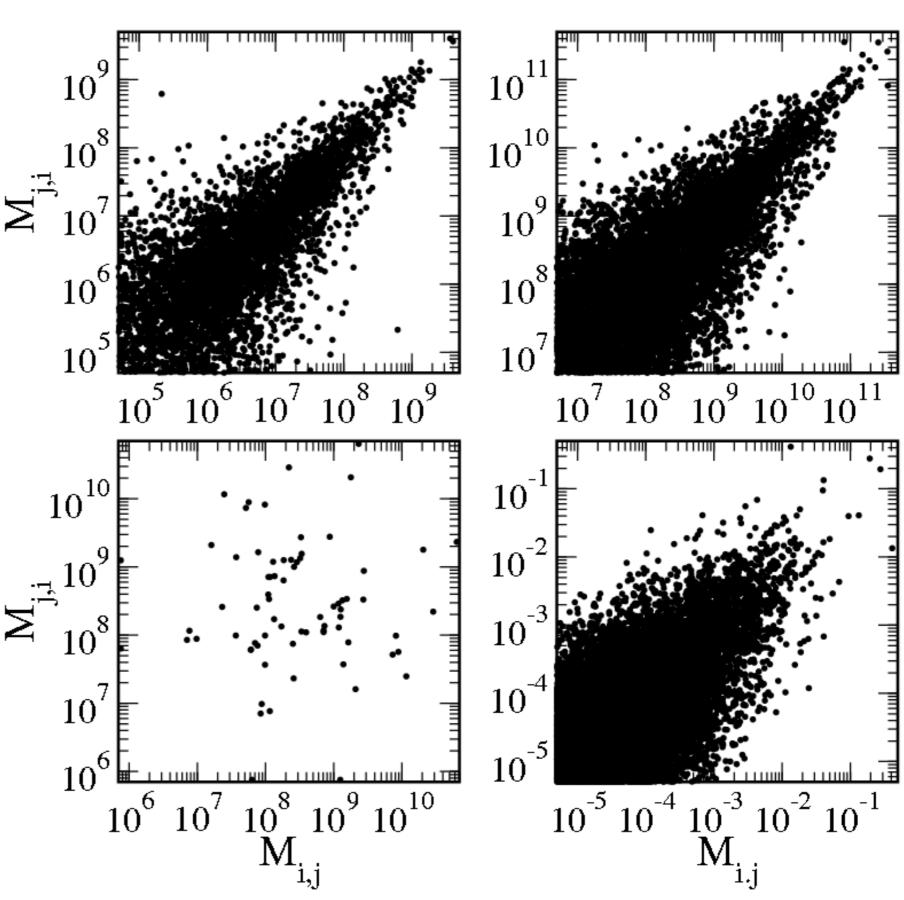
• Gravity model of trade:

$$M_{i,j} = g m_i m_j / D_{i,j}$$
 (symmetric)

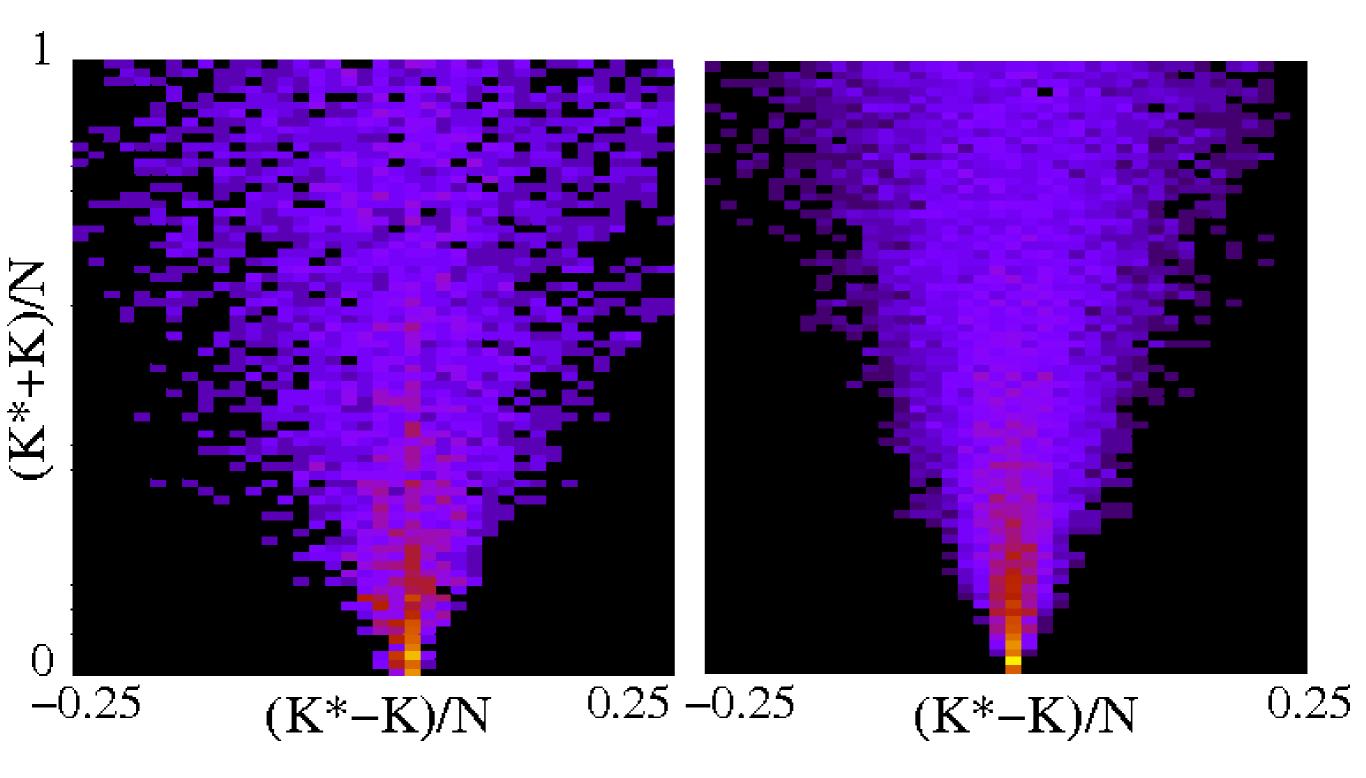
• Random model

$$M_{i,j} = \epsilon_i \epsilon_j / ij$$
 $\epsilon_{i,j} \in [0,1)$ (preserves Zipf law)

t:: all commodities (1962, 2008); b: crude petroleum (2008), random model



model statistics



Crisis

					Rank	country	b_i (negative)	global rank
					1	Greece	0.5131	1
				$P_i^* - P_i$	2	•	0.2505	15
Rank yea	r country	$b_i(t+1) - b_i(t)$	balance	$b_i = \frac{1}{w_i}$		Romania		19
1 197	3 Lebanon	-0.376		-	_	Portugal		23
2 200	1 Argentina	-0.269	weight	$w_i = P(i) + P^*$	(")	Mexico	0.1743	37
3 198	1 Mexico	-0.258			7	Canada USA	0.1633 0.1457	40 45
4 198	3 Nigeria	-0.253			8	UK	0.1397	49
5 200	2 Saudi Arabia	-0.250			9	Poland	0.1326	51
6 198		-0.247			10	France	0.1086	62
							2008; w>0	0.05 (~20%)
7 199		-0.244	0.6₽	 				
8 196	2 Venezuela	-0.236			-]		
9 197	3 Nigeria	-0.230	0.4			J		
10 199	4 Mexico	-0.230	0.4]		
11 199	7 Rep. of Korea	-0.219	02	and the property of the				
12 198	3 U. Arab Emir.	-0.213	<u>Ş</u>					
13 200	5 Iran	-0.210	즉 하		e.			
14 197	8 Iran	-0.210	<u>±</u>	700		·		
15 199	3 Turkey	-0.204	-0.2		1.	· -		
16 197	5 India	-0.202	-			- 4		
17 199	8 Russian Fed.	-0.202	-0.4			· -		

-0.6± -0.6

-0.4

-0.2

b_i(t)-b_i

0.4

18 1976

19 1987

20 1989

Iraq

Argentina

Venezuela

-0.200

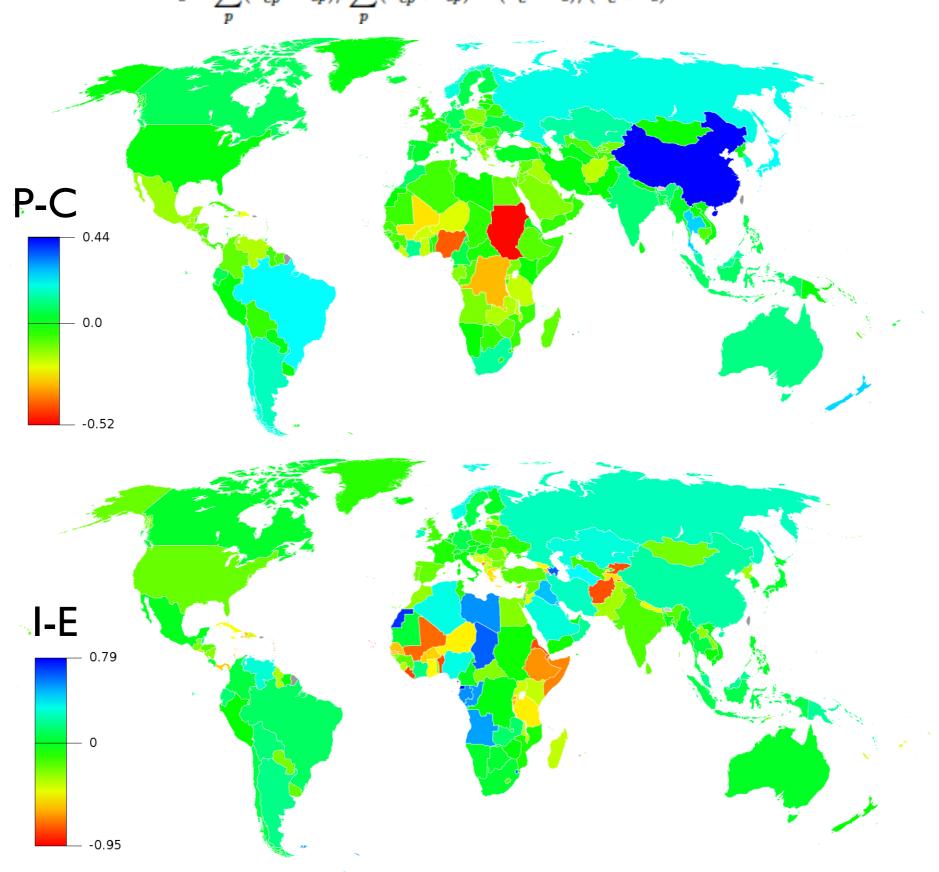
-0.196

-0.192

Multi-product WTN

Country balance

$$B_c = \sum_{p} (P_{cp}^* - P_{cp}) / \sum_{p} (P_{cp}^* + P_{cp}) = (P_c^* - P_c) / (P_c^* + P_c).$$

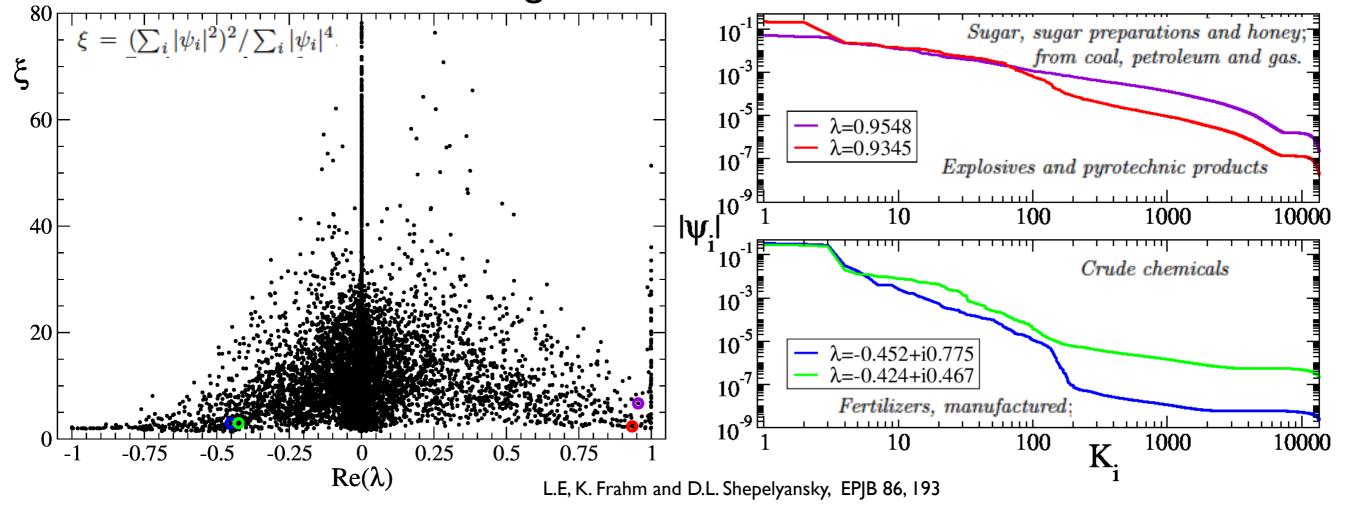


2d ranking of products 60 Meat and meat preparations 1963 1978 Dairy products and eggs ish and fish preparations 50 50 Coffee, tea, cocoa, spices & man Feed. Stuff for animals Miscellaneous food preparations 40 40 Beverages **K*** **K*** Tobacco and tobacco manuf. Hides, skins and fur skins, undress. p Oil seeds, oil nuts and oil kernels Crude rubber incl. synth & recl. 30 30 Wood, lumber and cork Pulp and paper Textile fibres, not manuf., & waste Crude fertilizers & crude minerals 20 Metalliferous ores and metal scrap 20 Crude animal & vegetable mat. Coal, coke and briquettes Petroleum and petroleum products Gas, natural and manufactured 10 Electric energy Chemical elements & compounds 50 60 Crude chem. from coal, petr. & gas 50 60 Dyeing, tanning & colouring mat. Medicinal & pharmaceutical prod. 60 60 Perfume mat., toilet & clean. prep. Fertilizers, manufactured 2008 Explosives and pyrotechnic prod. 1993 Plastic materials, etc. Chemical materials and products Leather. Manuf. & dressed fur skin Rubber manufactures, nes Wood & cork manuf. exc. furniture Paper, paperboard and manuf. 40 40 Textile yarn, fabrics, etc. Non metallic mineral manuf., nes **K*** **K*** Iron and steel Non ferrous metals p 30 Machinery, other than electric 30 Electrical machinery, apparatus and appliances Transport equipment 20 20 Travel goods, handbags and similar articles Scientif & control instrum, photogr gds, clocks 10 10 Postal packages not class. According to kind ■ Special transact. Not class. According to kind ♦ Animals, nes, incl. Zoo animals, dogs and cats ▲ Firearms of war and ammunition therefor Coin, other than gold coin, not legal tender K_p^{40} 60 50 10 20 30 50 60

multi-prod WTN spectrum

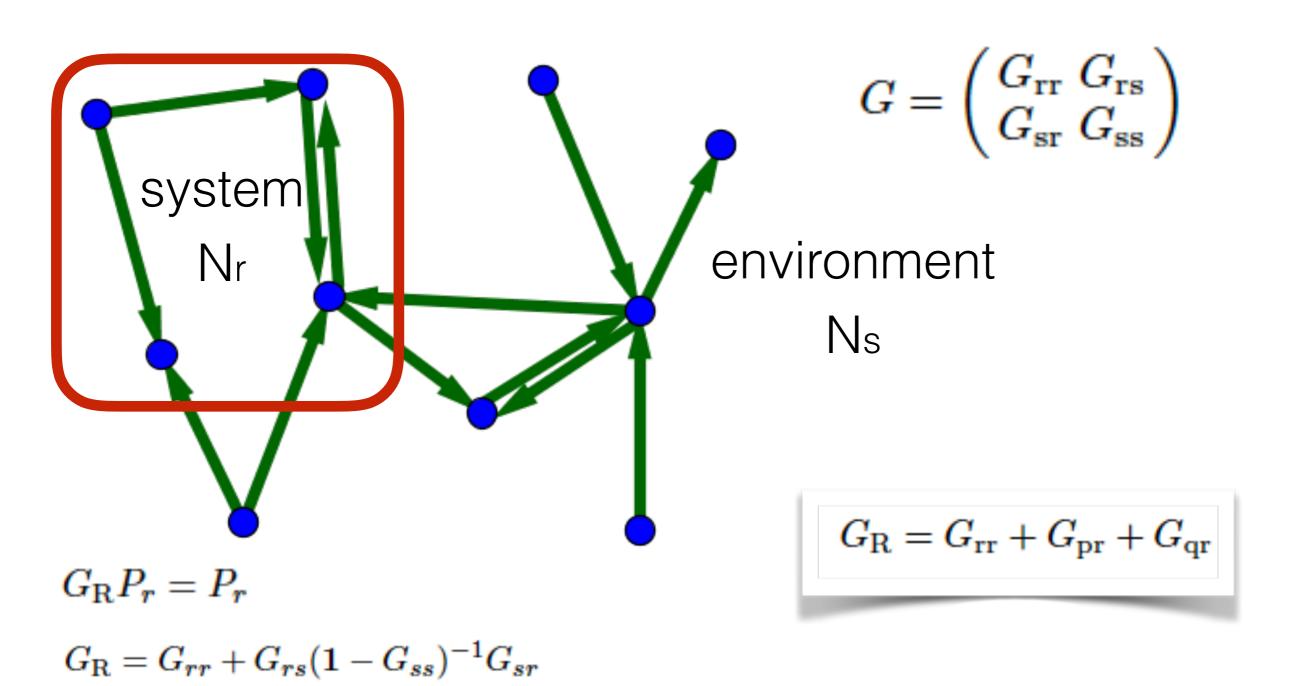
K_i	$ \psi_i $	country	$ \psi_i $	country	$ \psi_i $	country	$ \psi_i $	country	
		prod: 57		prod:06		prod:56		prod:52	
1	0.052	USA \	0.216	/Mali	0.332	Brazil	0.288	Japan \	
2	0.044	Tajikistan	0.201	Guinea	0.304	Bolivia	0.279	Rep. of Korea	
3	0.042	Kyrgyzstan	0.059	USA	0.274	Paraguay	0.245	China	
4	0.022	France	0.023	Germany	0.031	Argentina	0.020	Australia	
5	0.021	Mexico	0.021	Mexico	0.017	Uruguay	0.013	USA	
6	0.018	Italy	0.021	Canada	0.009	Chile	0.012	U Arab Em	
7	0.018	Canada	0.018	UK	0.004	Portugal	0.010	Canada	
8	0.015	Germany	0.015	Israel	0.004	Angola	0.010	Singapore	
9	0.013	U Arab Em	0.015	C d'Ivoire	0.004	Spain	0.009	Germany	
10	0.012	Qatar	0.014	Japan	0.003	France	0.008	New Zealand	

eigenstate communities



Reduced G matrix of WTN

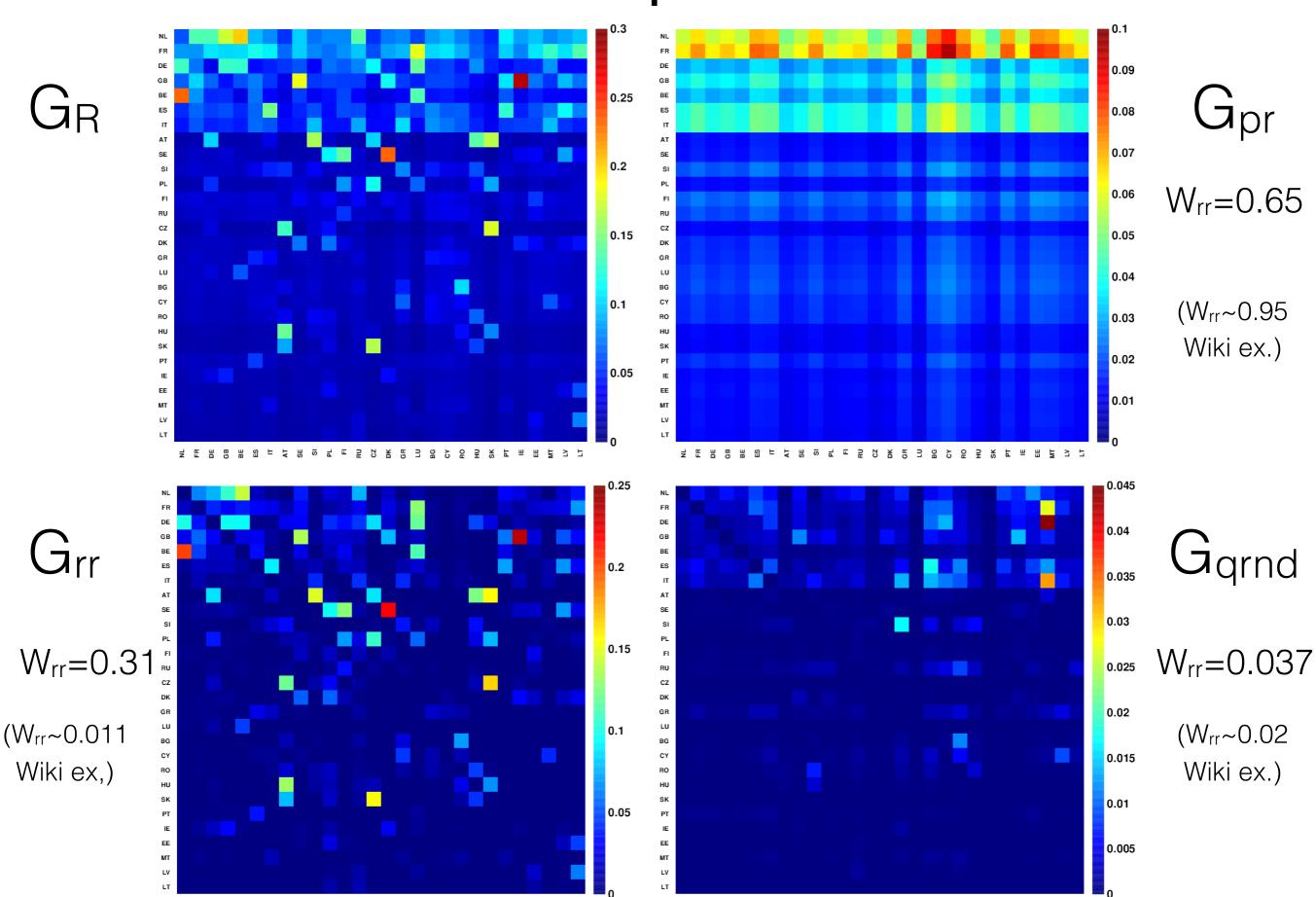
reduced Google matrix



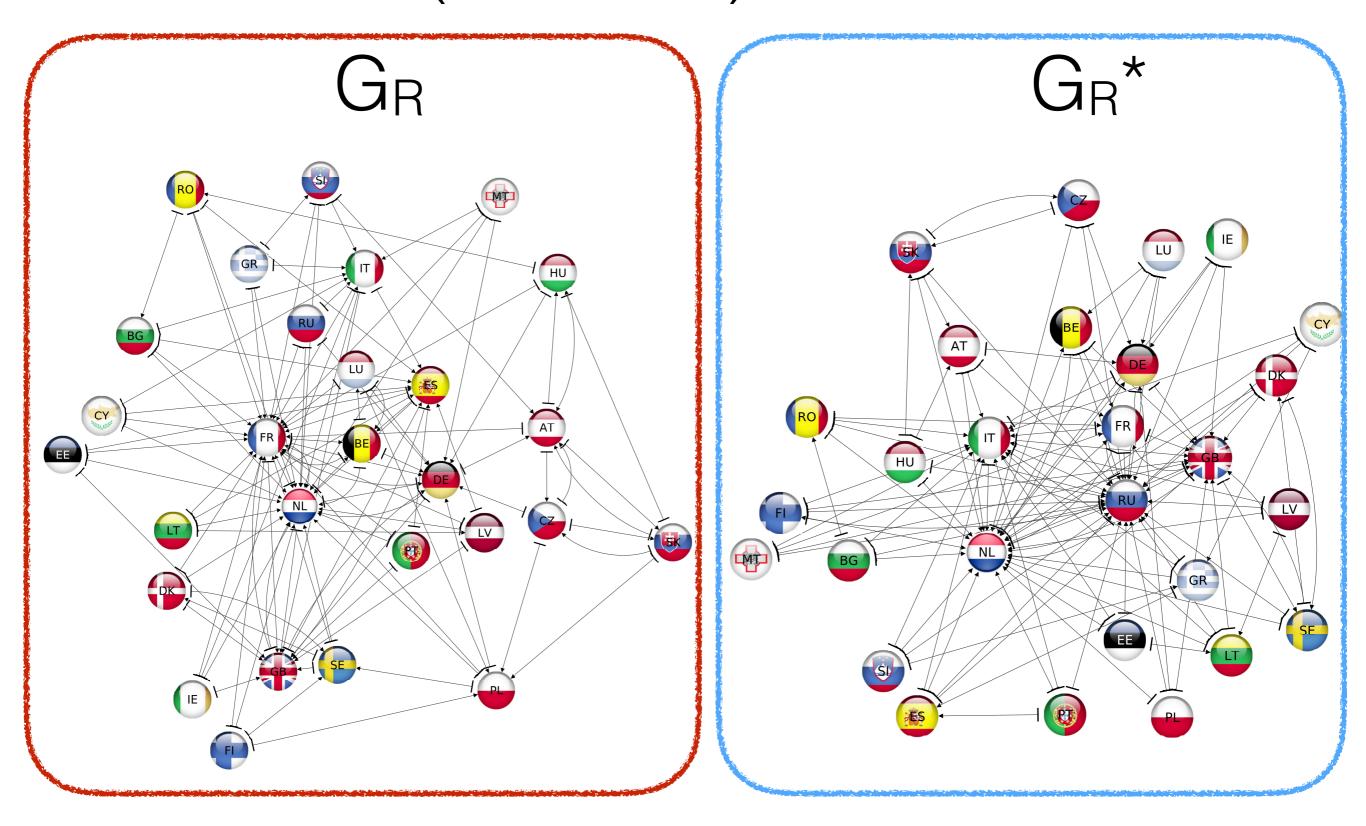
K. Frahm and D.L. Shepelyansky, arxiv: 1602.02394 (yesterday Frahm's introduction)

Colab. C Coquide, J. Lages & D. Shepelyansky

Petroleum for 27 European countries + Russia



Petroleum (27 Eu + Ru) 4 friend networks



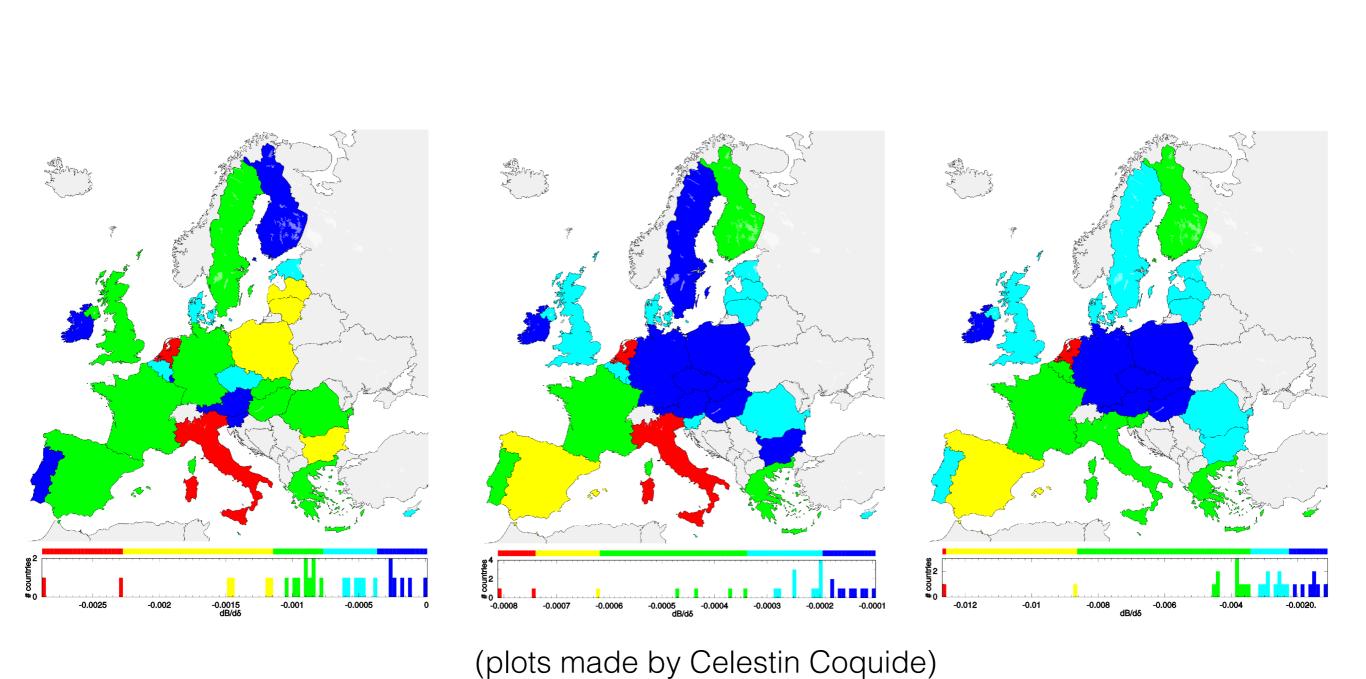
(plots made by Celestin Coquide)

Balance sensitivity to petroleum from:

Saudi Arabia

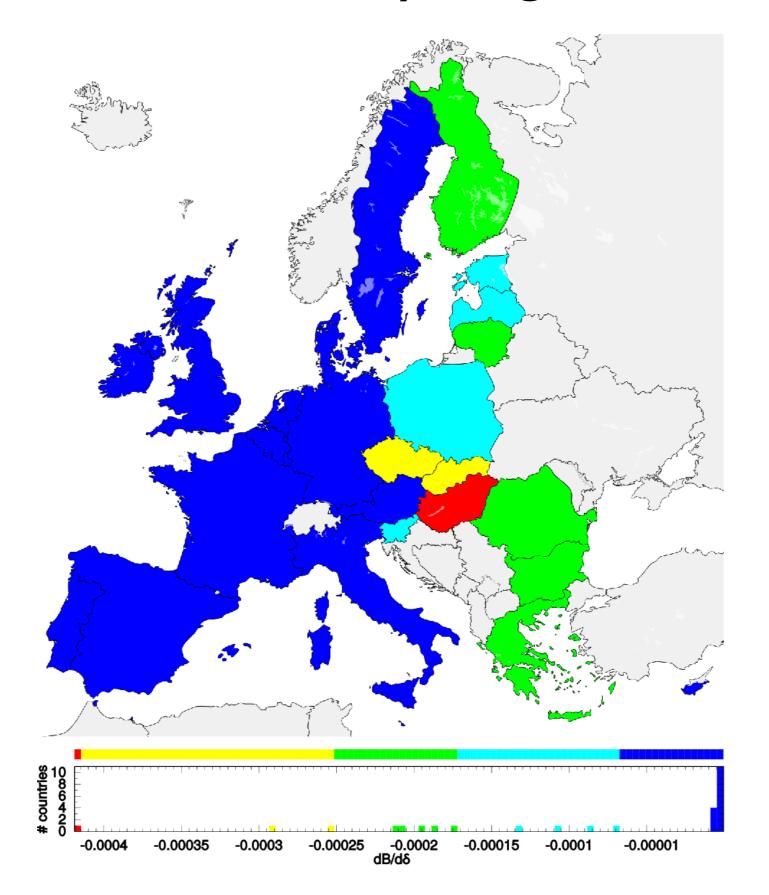
Russia

USA



Balance sensitivity to gas from:

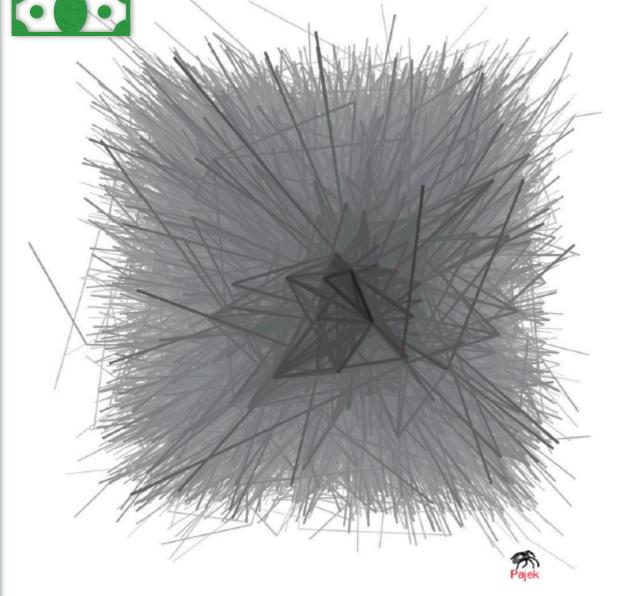




Bitcoin network



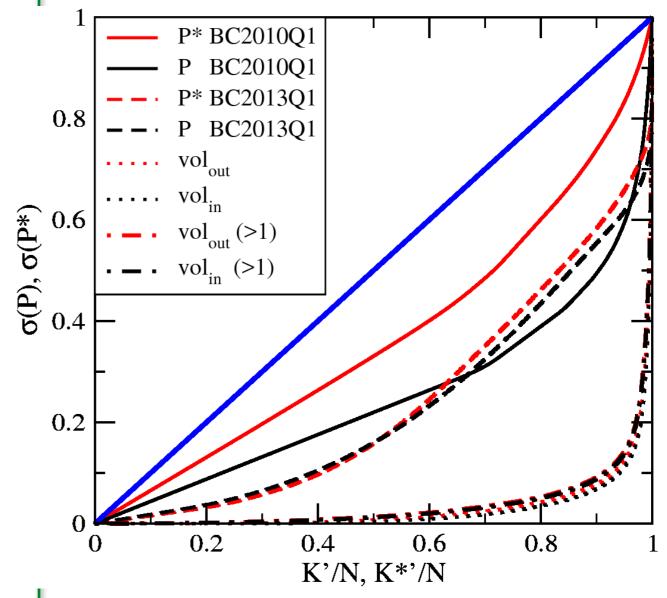
Interbank payment network



Fedwire interbank payment network. First day of Sample. 6600 nodes and over 70,000 undirected links [39].

Soramaaki et al, Physica A 379, 317 (2007)

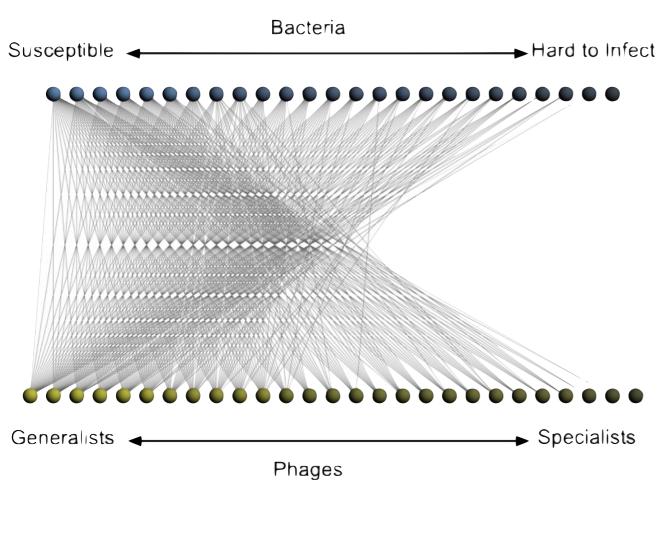




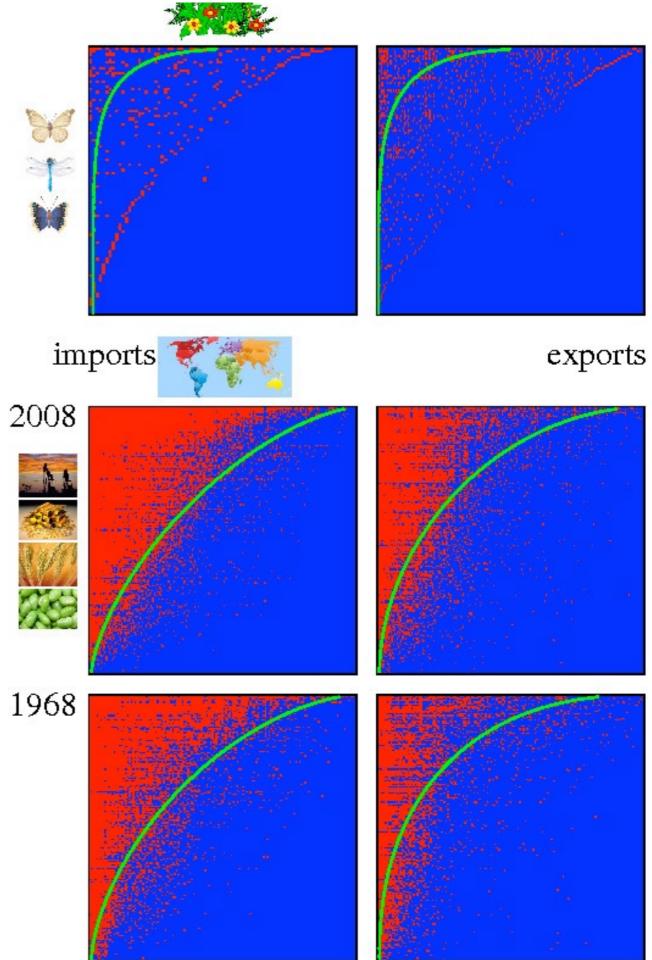
LE, K Frahm and D.L. Shepelyansky, Eur. Phys. J. B 91 (2018)

Ecological ranking for WTN

Nestedness



$$Q_{c,p}^{(i,e)} = \begin{cases} 1 & \text{if } m_{c,p}^{(i,e)} \ge \mu \\ 0 & \text{if } m_{c,p}^{(i,e)} < \mu \end{cases}$$



Nestedness

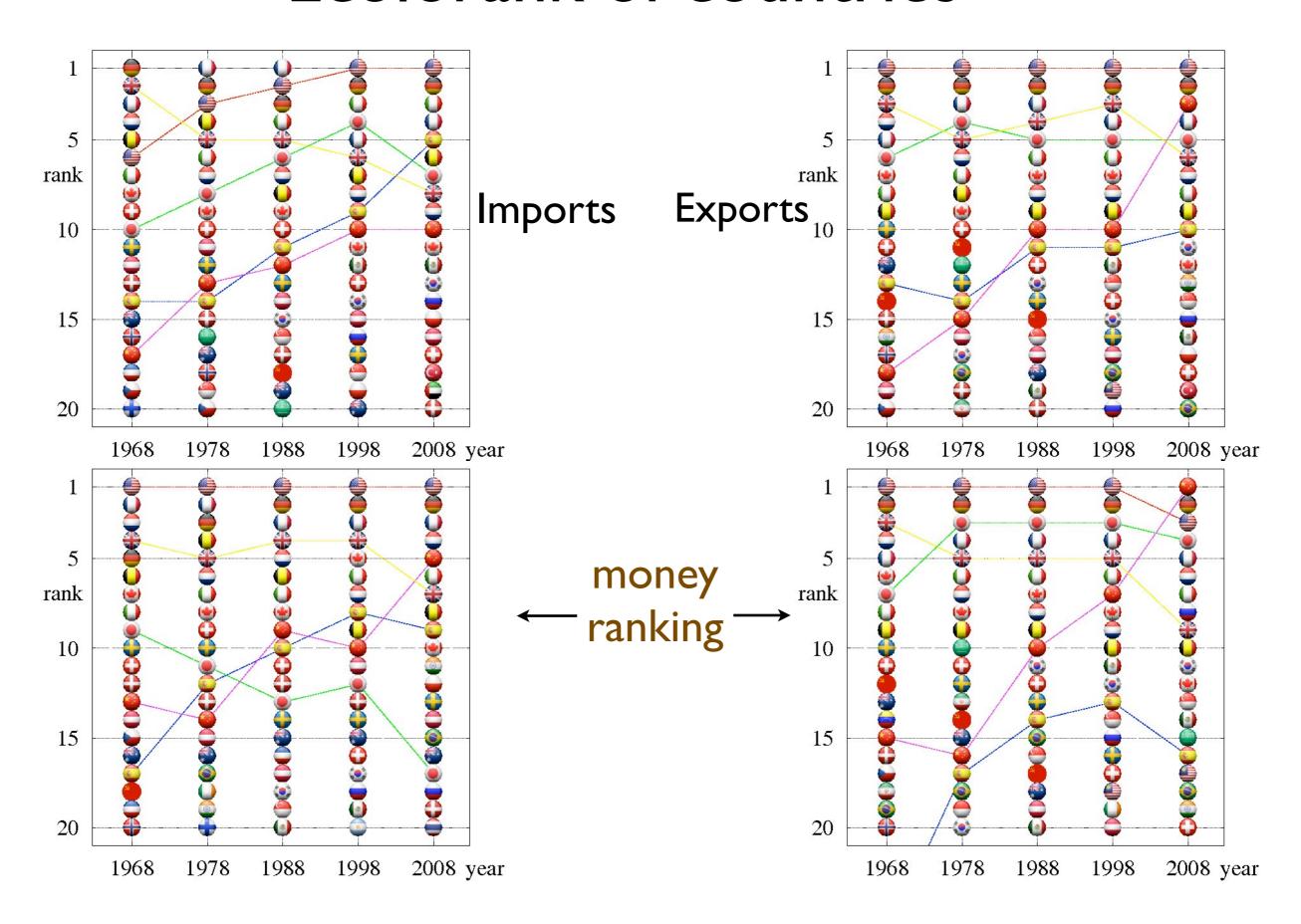
$$m^{(i,e)} = M^{(i,e)}/M_{\text{max}}$$

$$M_{p,c}^{(i)} = \sum_{c'=1}^{N_c} M_{c,c'}^p \qquad M_{p,c}^{(e)} = \sum_{c'=1}^{N_c} M_{c',c}^p$$

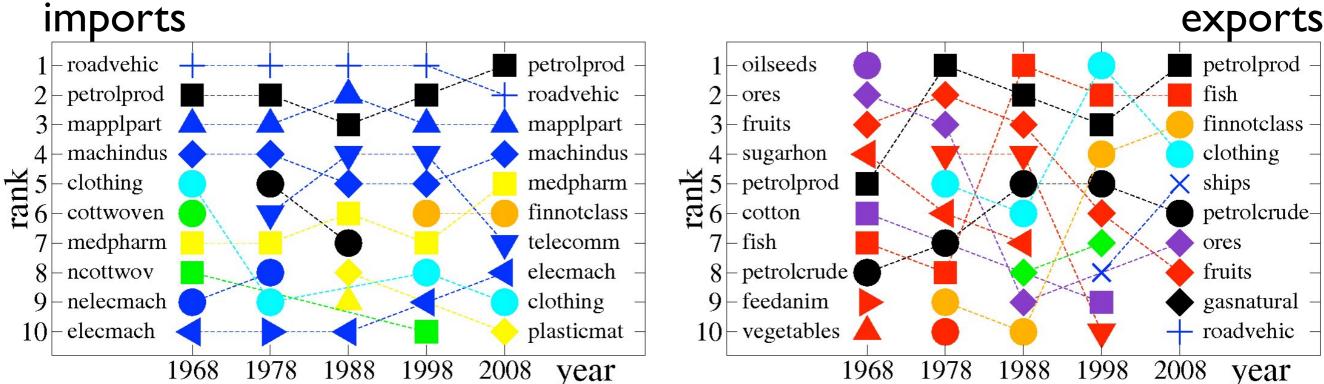
$$M_{p,c}^{(i)} = \sum_{c'=1}^{N_c} M_{c,c'}^p \qquad M_{p,c}^{(e)} = \sum_{c'=1}^{N_c} M_{c',c}^p$$

$$M_{p,c}^{(e)} = \sum_{c'=1}^{N_c} M_{c',c}^p$$

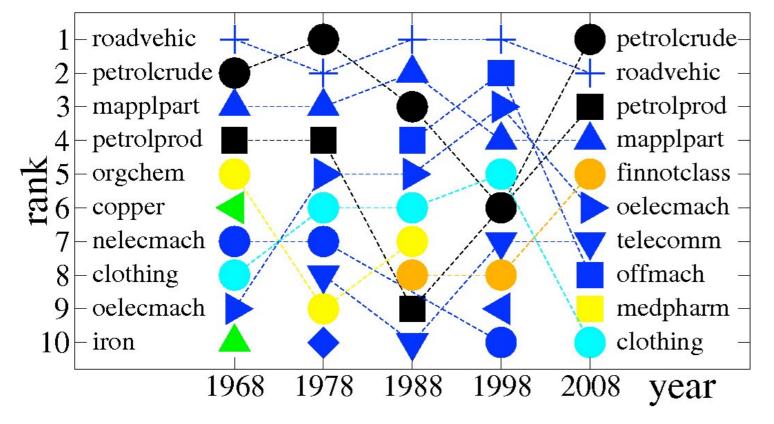
Ecolorank of countries

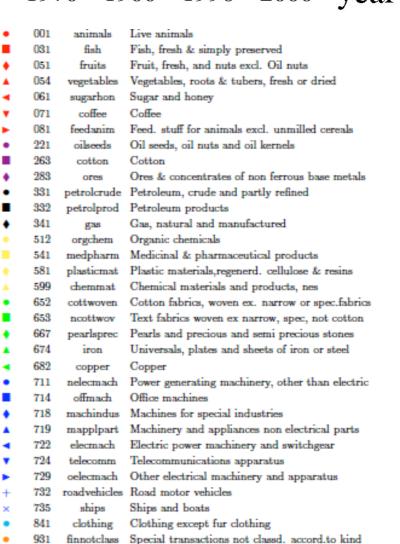


Ecolorank of products



money rank





Conclusions

- Google matrix of the WTN (democratic in countries, global network properties):
 - 1) one product of all comm. (Nc)
 - 2) multiprod (Nc x Np)
 - 2d-ranking, spectrum, communities in eigenstates, correlation between P-C, comparison with I-E, new tool for trade analysis
- Dynamics, crisis, sensitivity to price variation
- reduced Google matrix on WTN
- Nestedness based ranking of WTN
- Bitcoin network

Thank you



Fractal Weyl law

0.5

0

-0.5

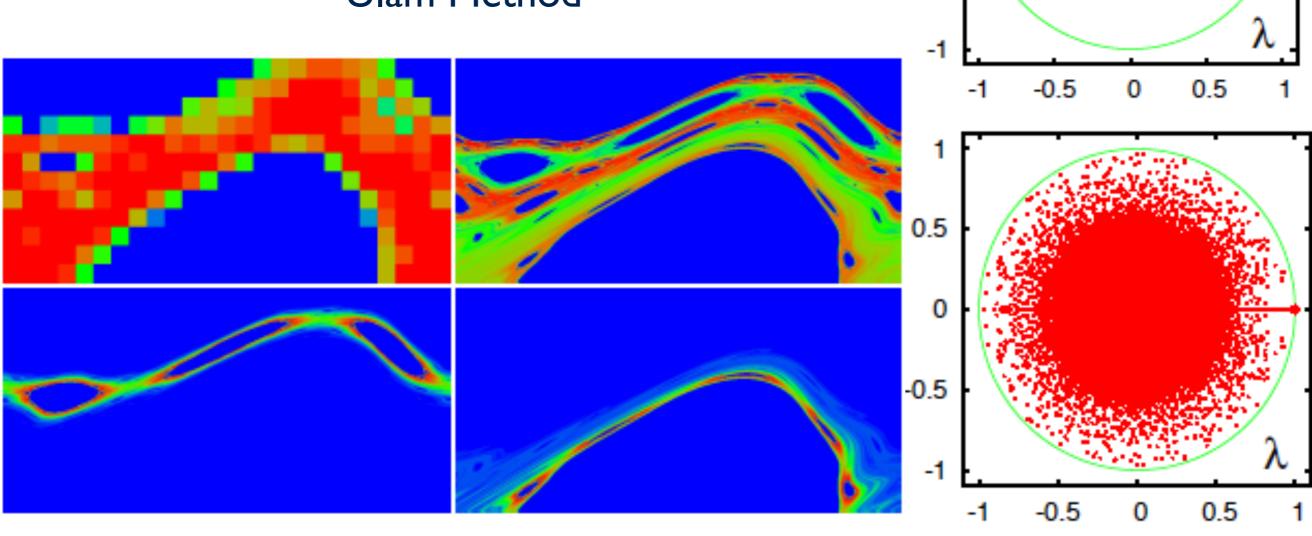
(Introduction by Nonnenmacher yesterday)

Chirikov standard map

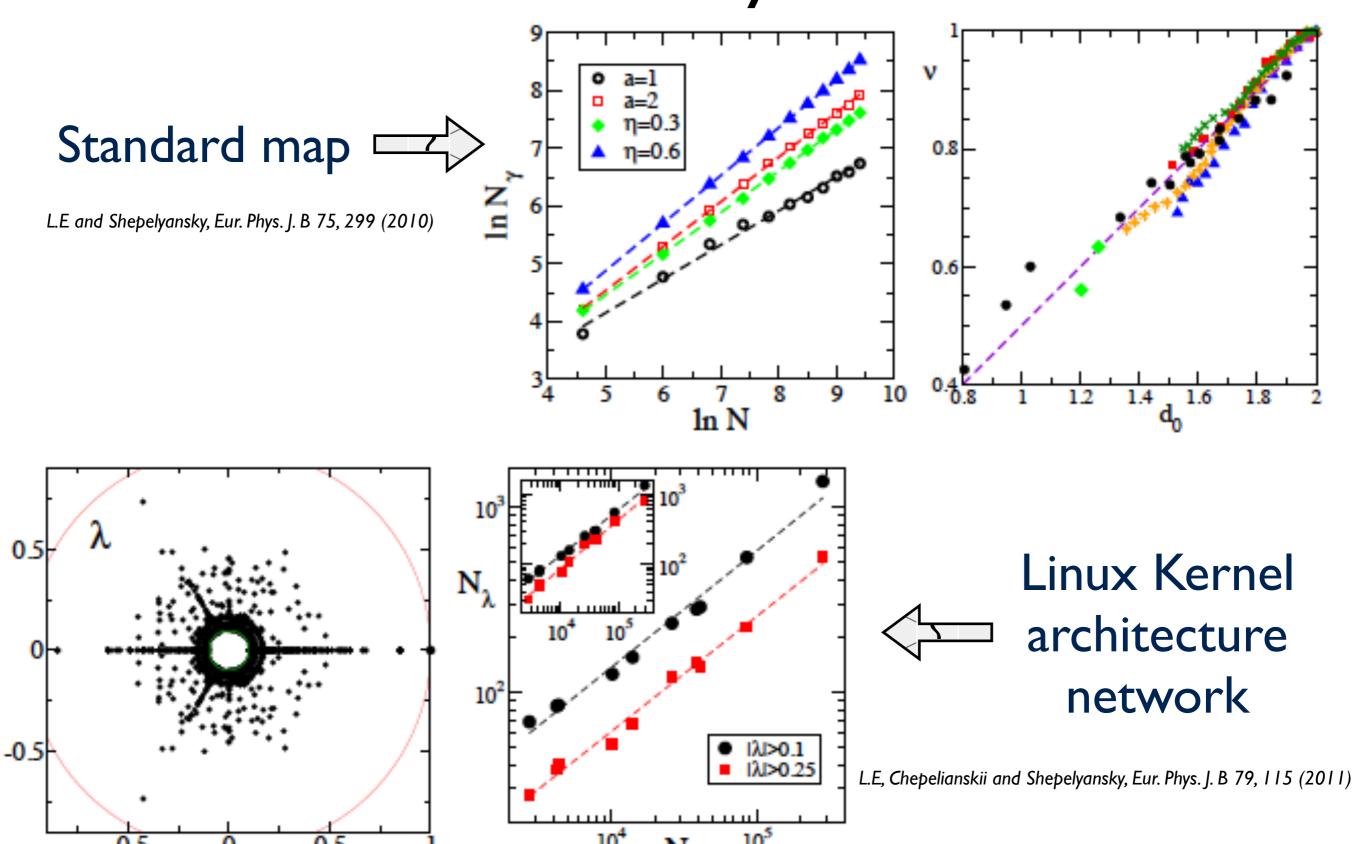
$$\bar{p} = p + K \sin x$$

$$\bar{x} = x + \bar{p}$$

Ulam Method



Fractal Weyl law



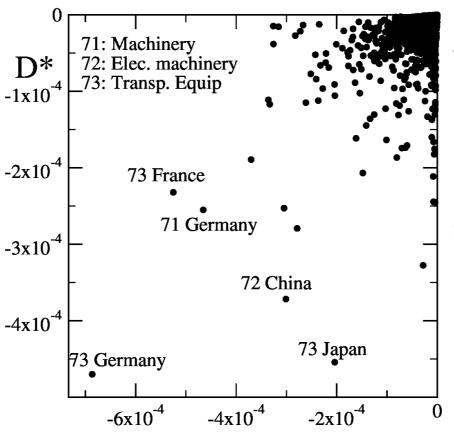
product names

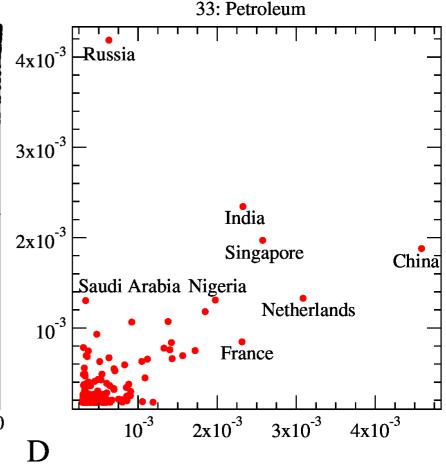
code	name	code	name
00	Live animals	54	Medicinal and pharmaceutical products
01	Meat and meat preparations	55	Perfume materials, toilet & cleaning preptions
02	Dairy products and eggs	56	Fertilizers, manufactured
03	Fish and fish preparations	57	Explosives and pyrotechnic products
04	Cereals and cereal preparations	58	Plastic materials, etc.
05	Fruit and vegetables	59	Chemical materials and products, nes
06	Sugar, sugar preparations and honey	61	Leather, lthr. Manufs., nes & dressed fur skins
07	Coffee, tea, cocoa, spices & manufacs. Thereof	62	Rubber manufactures, nes
08	Feed. Stuff for animals excl. Unmilled cereals	63	Wood and cork manufactures excluding furniture
09	Miscellaneous food preparations	64	Paper, paperboard and manufactures thereof
11	Beverages	65	Textile yarn, fabrics, made up articles, etc.
12	Tobacco and tobacco manufactures	66	Non metallic mineral manufactures, nes
21	Hides, skins and fur skins, undressed	67	Iron and steel
22	Oil seeds, oil nuts and oil kernels	68	Non ferrous metals
23	Crude rubber including synthetic and reclaimed	69	Manufactures of metal, nes
24	Wood, lumber and cork	71	Machinery, other than electric
25	Pulp and paper	72	Electrical machinery, apparatus and appliances
26	Textile fibres, not manufactured, and waste	73	Transport equipment
27	Crude fertilizers and crude minerals, nes	81	Sanitary, plumbing, heating and lighting fixt.
28	Metalliferous ores and metal scrap	82	Furniture
29	Crude animal and vegetable materials, nes	83	Travel goods, handbags and similar articles
32	Coal, coke and briquettes	84	Clothing
33	Petroleum and petroleum products	85	Footwear
34	Gas, natural and manufactured	86	Scientif & control instrum, photogr gds, clocks
35	Electric energy	89	Miscellaneous manufactured articles, nes
41	Animal oils and fats	91	Postal packages not class. According to kind
42	Fixed vegetable oils and fats	93	Special transact. Not class. According to kind
43	Animal and vegetable oils and fats, processed	94	Animals, nes, incl. Zoo animals, dogs and cats
51	Chemical elements and compounds	95	Firearms of war and ammunition therefor
52	Crude chemicals from coal, petroleum and gas	96	Coin, other than gold coin, not legal tender
53	Dyeing, tanning and colouring materials		

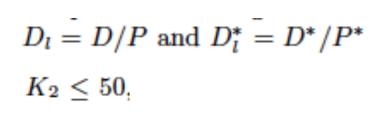
Sensitivity to price variation

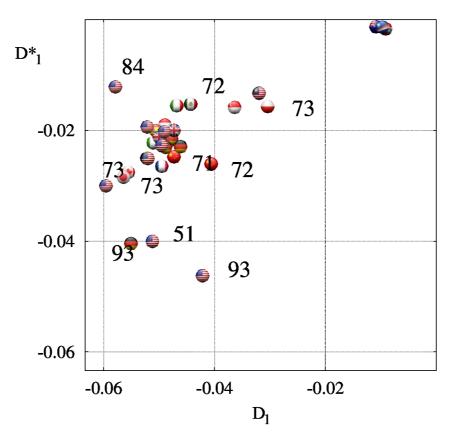
$$D = dP/d\delta = \Delta P/\delta$$

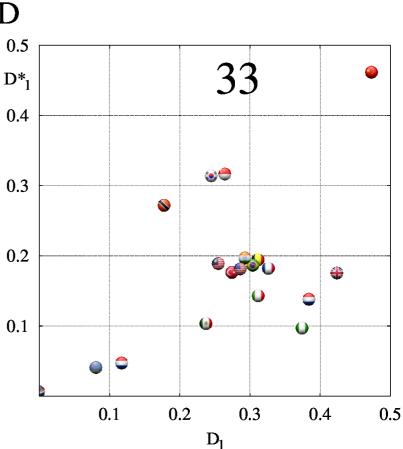
 $D^* = dP^*/d\delta = \Delta P^*/\delta$.
 $\delta' = 0.01, 0.03, 0.05$



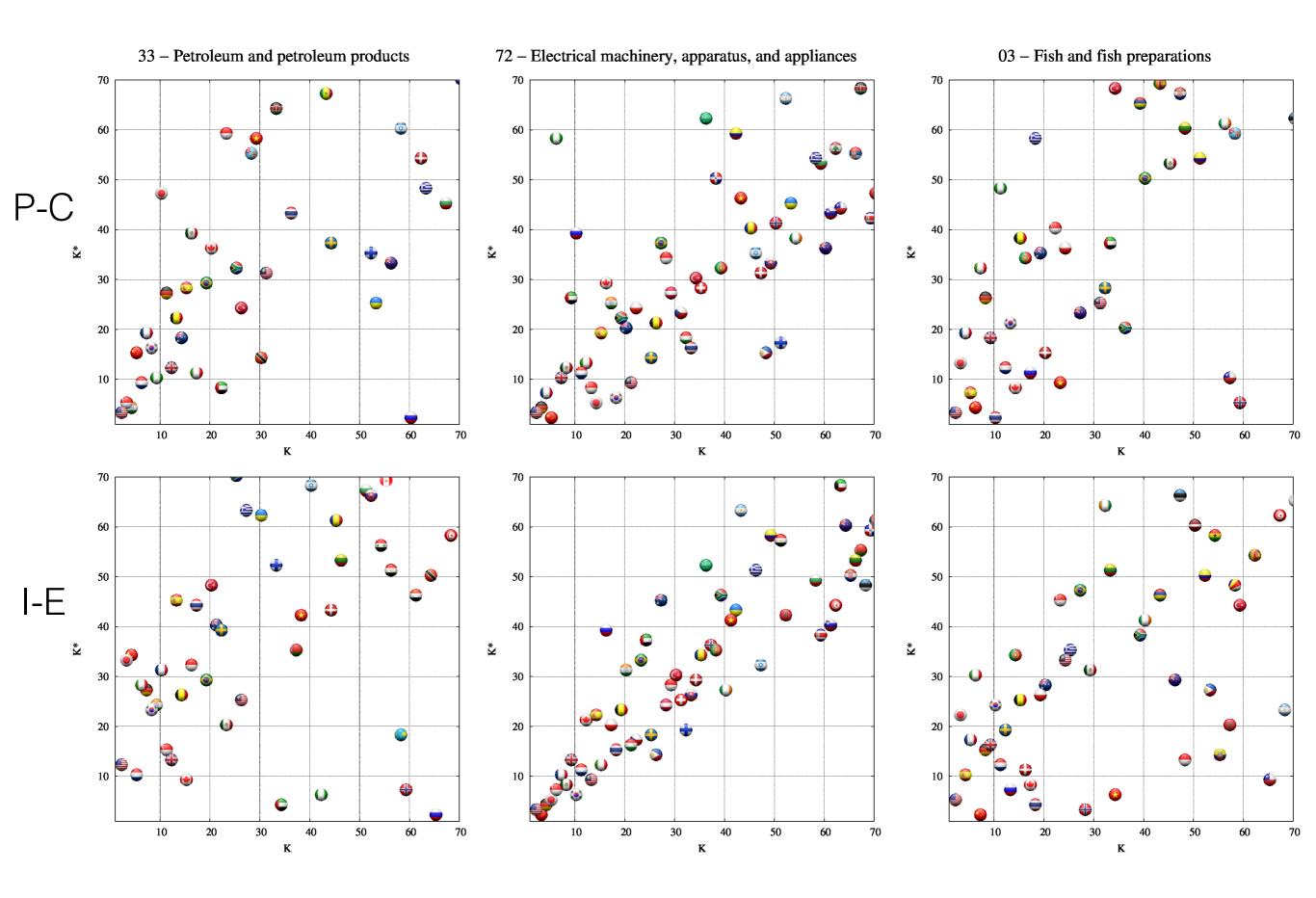




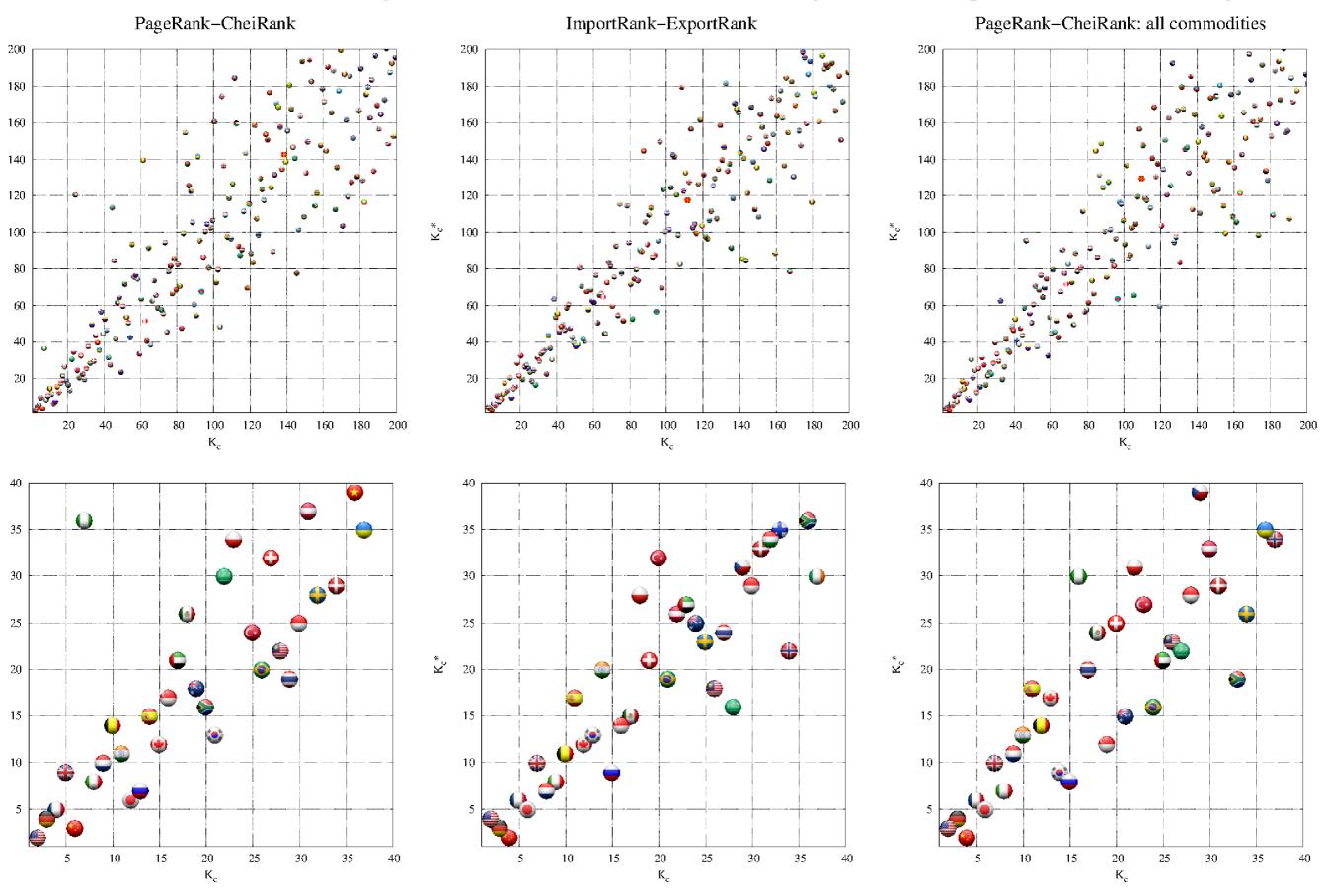




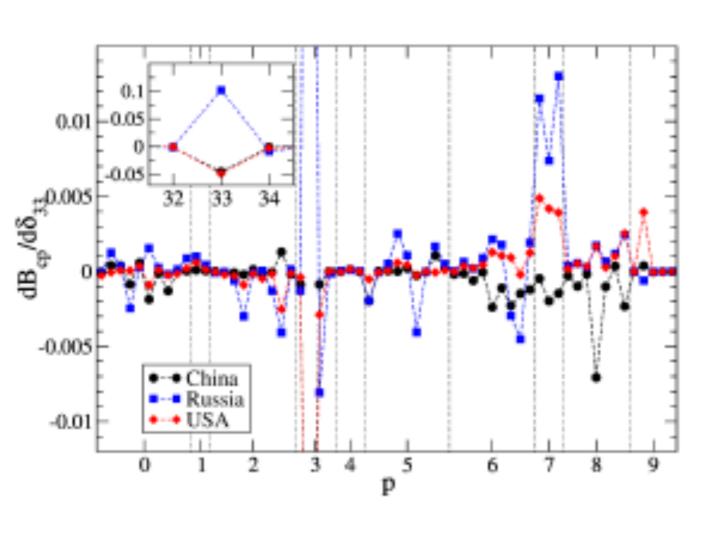
2d reduced ranks

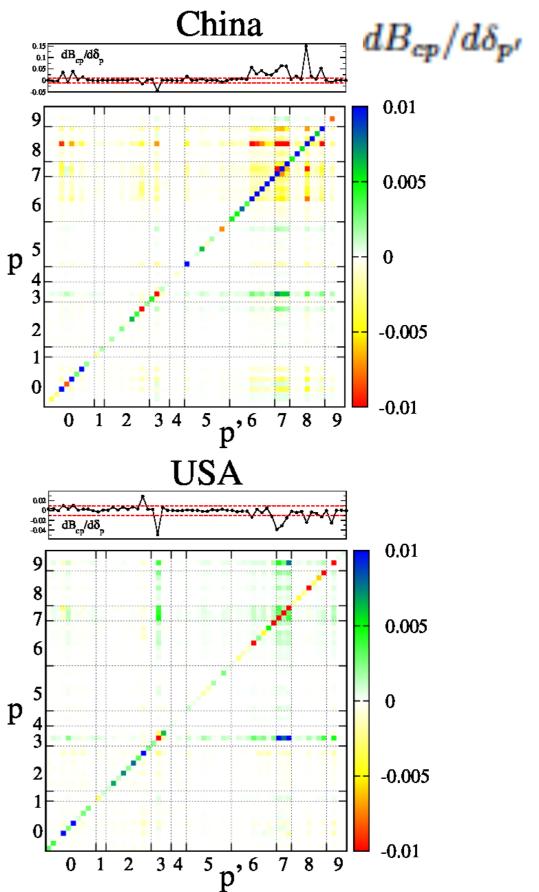


2d ranking of countries (multiproducts)

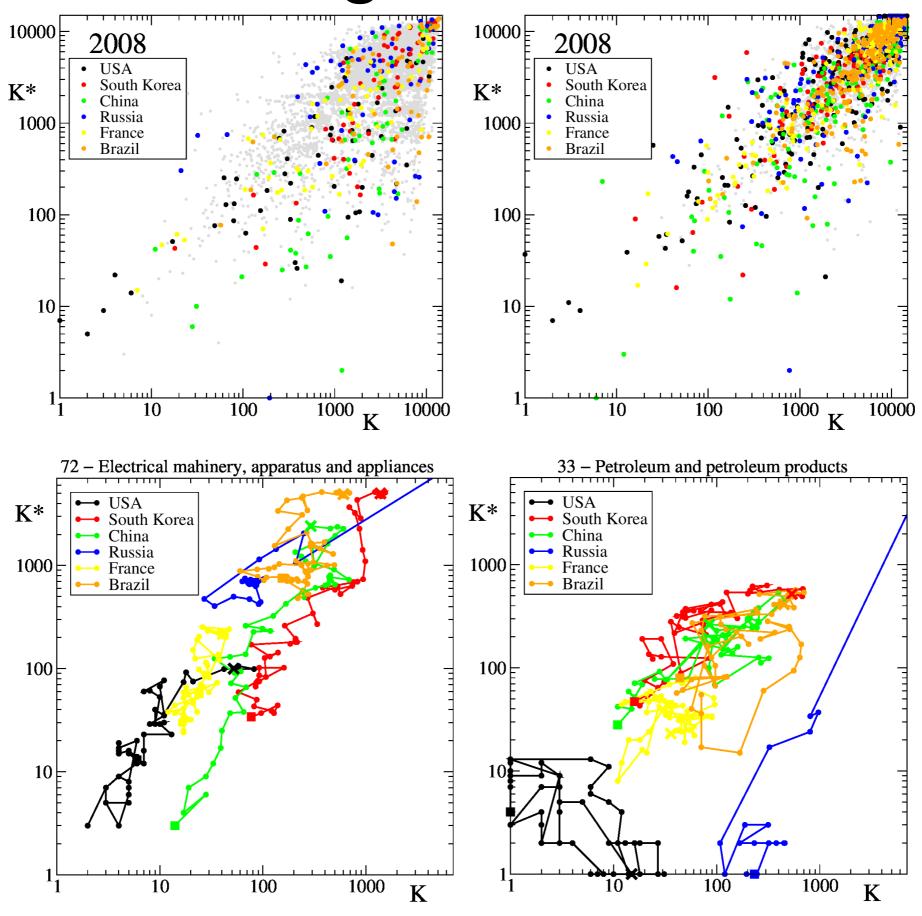


Sensitivity to price variation II

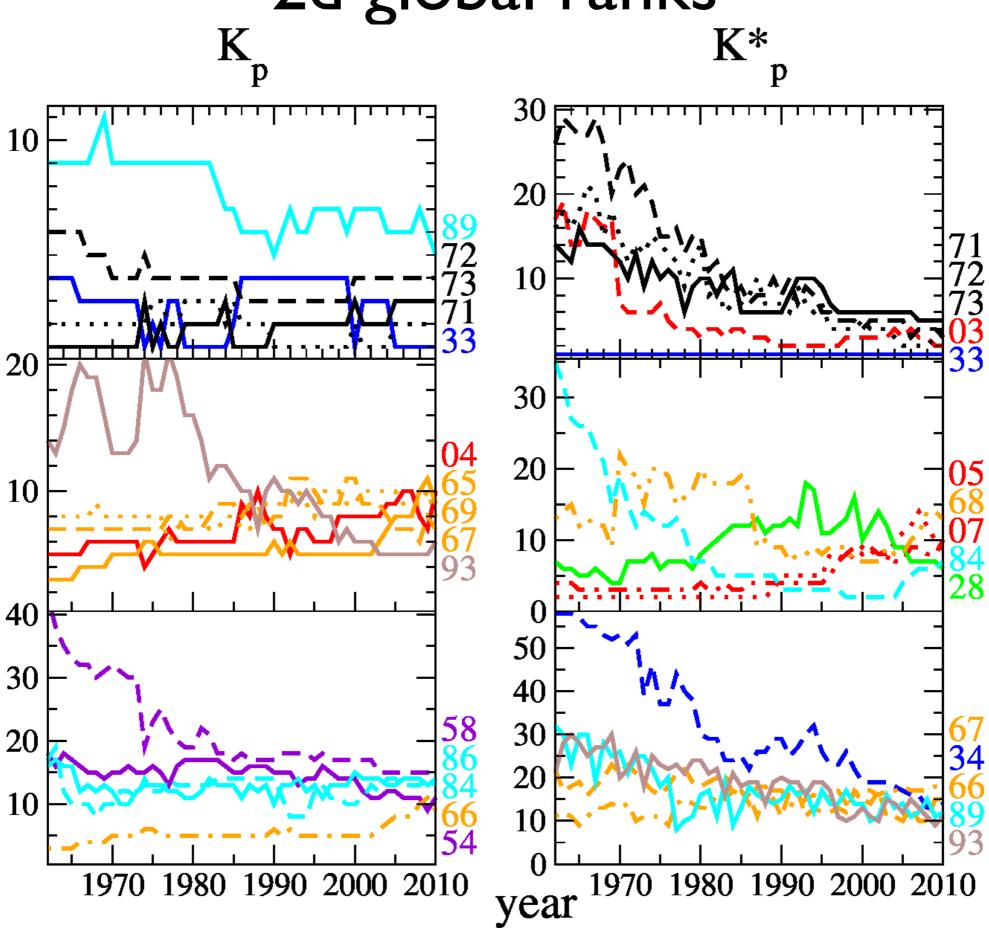




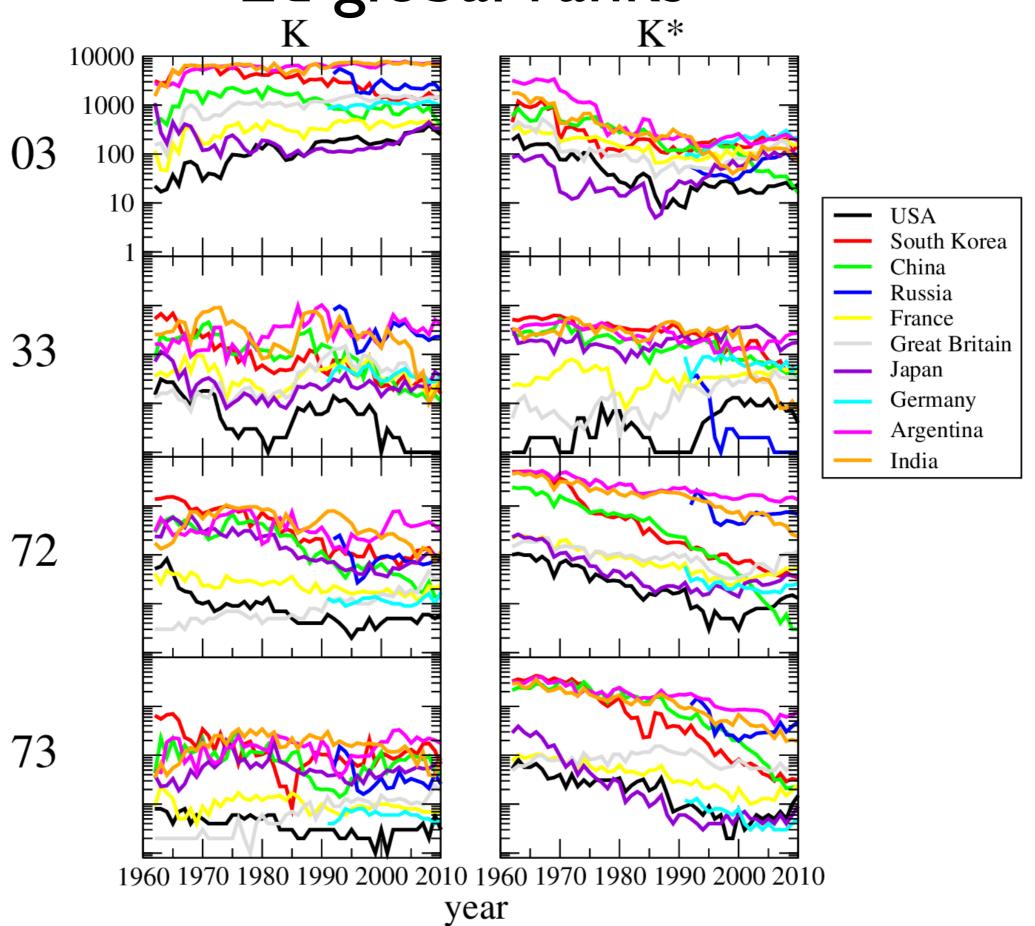
2d global ranks



2d global ranks



2d global ranks

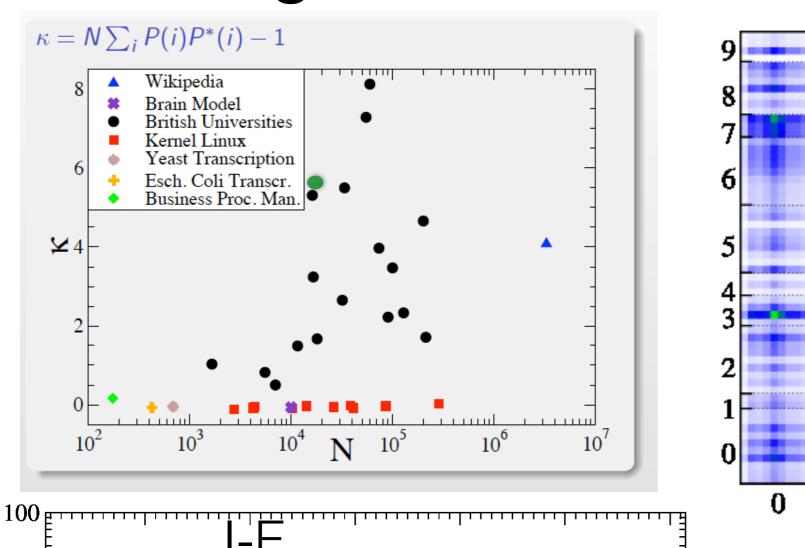


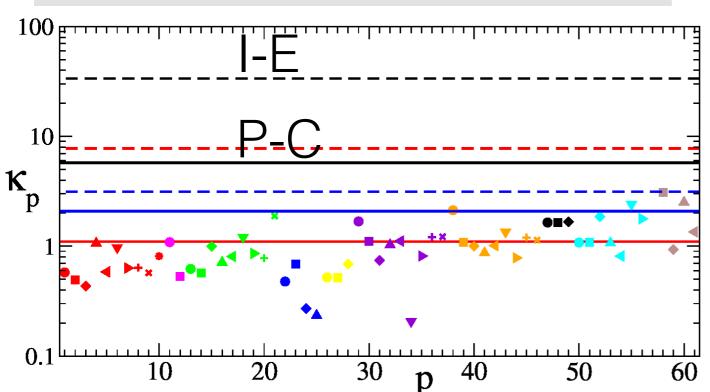
Tab. 1: List of countries: 27 from EU (blue) and the 10 best oil exporters (red) ordered by decreasing PageRank left (CheiRank right) value for petroleum product.

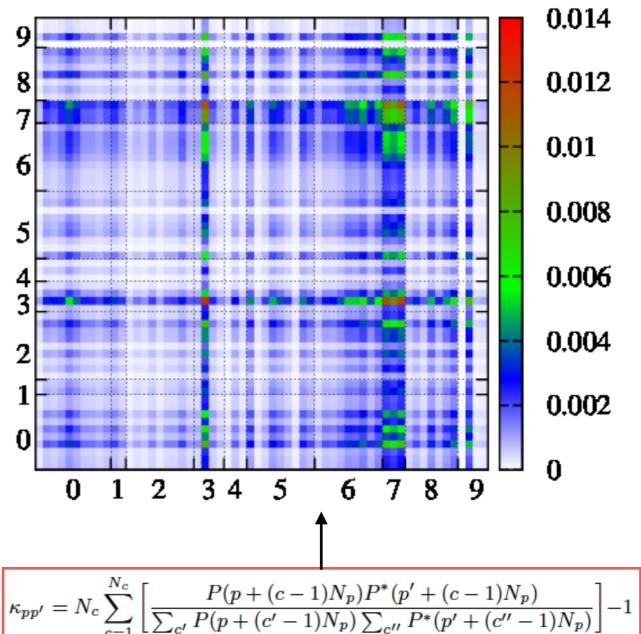
100001110 1 00001001111 1010 (0.		-00)	,
COUNTRY NAME	ISO-2		С
Netherlands	NL		R
France	FR		S
Nigeria	NG		K
Germany	DE		U
United Kingdom	GB		N
Belgium	BE		N
Spain	ES		It
Italy	IT		U
Canada	CA		V
United Arab Emirates	AE		F
Austria	AT		K
Sweden	SE		В
Slovenia	SI		G
Poland	PL		S
Finland	FI		A
Russia	RU		F
Czech Republic	CZ		C
Denmark	DK		S
Greece	GR		Iı
Luxembourg	LU		R
Bulgaria	BG		В
Cyprus	CY		G
Romania	RO		D
Angola	AO		$\mid L \mid$
Hungary	HU		P
Slovakia	SK		P
Portugal	PT		E
Republic of Ireland	IE		Н
Saudi Arabia	SA		A
Estonia	EE		\mathbf{N}
Malta	MT		$\mid L$
Latvia	LV		\mathbf{S}
Venezuela	VE		C
Kazakhstan	KZ		R
Lithuania	LT		C
Iraq	IQ		S
Kuwait	KW		L

t) value for petroleum product.					
COUNTRY NAME	ISO-2				
Russia	RU				
Saudi Arabia	SA				
Kazakhstan	KZ				
United Arab Emirates	AE				
Netherlands	NL				
Nigeria	NG				
Italy	IT				
United Kingdom	GB				
Venezuela	VE				
France	FR				
Kuwait	KW				
Belgium	BE				
Germany	DE				
Spain	ES				
Angola	AO				
Finland	FI				
Canada	CA				
Sweden	SE				
Iraq	IQ				
Romania	RO				
Bulgaria	BG				
Greece	GR				
Denmark	DK				
Lithuania	LT				
Portugal	PT				
Poland	PL				
Estonia	EE				
Hungary	HU				
Austria	AT				
Malta	MT				
Latvia	LV				
Slovakia	SK				
Czech Republic	\mathbf{CZ}				
Republic of Ireland	IE				
Cyprus	CY				
Slovenia	SI				
Luxembourg	LU				

PageRank CheiRank correlator







$$\kappa_p = \kappa_{pp'} \delta_{p,p'}$$