

The Empirical Analysis of Weighted Directed Networks: An Application to the World Trade Web

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Motivations and Goals

- **Illustrating some statistical tools for the analysis of empirically-observed networks**

- Three simple but important methodological points
- Leading example: network of trade among World countries

- **Methodological points**

- 1 If appropriate, empirical analyses should be carried out in the framework of **weighted** networks
- 2 Often, but not always, **directed** network analyses should be preferred to **undirect** ones
- 3 The empirical analysis of **weighted, directed networks** requires new statistical tools

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Background Papers

- Fagiolo, G. (2006a), “Directed or Undirected? A New Index to Check for Directionality of Relations in Socio-Economic Networks”, *Economics Bulletin*, 3: 1-12.
- Fagiolo, G. (2006b), “Clustering in Complex Directed Networks”, Working Paper, arXiv:physics/0612169v2.
- Fagiolo, G., Reyes, J. and Schiavo, S. (2007), “The Evolution of the World Trade Web: A Weighted Network Approach”, LEM Working Paper, Sant’Anna School of Advanced Studies, forthcoming.

From Social to Hard Sciences...

● Networks of interpersonal relations

- Old idea in sociology: “relevant others” (Miller, 1963)
- Explaining patterns of interactions among people of groups
- Friendship (Rapoport & Horvath, 1961; Milgram, 1967)
- Marriage (Padgett & Ansell, 1993)
- Job-market interactions (Granovetter, 1974)

● Statistical analysis of network topology

- Albert & Barabási (2002), Newman (2003), Pastor-Satorras & Vespignani (2004), Dorogovtsev & Mendes (2003), ...
- Properties of real-world technological, biological and information networks
 - WWW and the Internet, peer-to-peer networks, power grids, train routes, airline connections, electronic circuits, metabolism, protein interactions, neural networks

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... And Back to Social Sciences (Econophysics)

● Empirical analysis of social and economic networks

- Socio-economic systems as networks?
- A non-exhaustive list of applications
 - scientific co-authorship (Newman, 2001) and citation (Redner, 1998)
 - telephone calls (Aiello et al., 2000)
 - email exchanges (Kossinets and Watts, 2006)
 - sexual relationships (Liljeros et al., 2001)
 - knowledge spill-overs among firms in industrial clusters (Giuliani and Bell, 2001)
 - market investment (Battiston and Catanzaro, 2004)
 - patent citation and innovation networks (Breschi and Lissoni, 2001; Ahuja, 2000)
 - firm alliance formation (Gulati, 1998; Garcia Pont and Nohria, 2002)
 - R&D teams and other within-firm networks (Reagans and co-authors, 2001, 2004)
 - social capital (Walker, Kogut, and Shan, 1997)
 - company ownership and control (Garlaschelli et al., 2005)
 - financial networks (Kullman et al., 2001)
 - bank-firm relationships (De Masi et al., 2007)
 - and also ...

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The network of international trade flows (1/3)

● Main idea

- Web of trade relations among countries as a networks
- Countries = nodes
- Links = existence of trade relationship (import/export)

● Any value added?

- Standard empirics: imports-exports as country-specific variables
- Network analysis: flows as relational variables
- Topological structure, higher-order trade structure
- Network structure and macroeconomic dynamics

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The network of international trade flows (2/3)

● Old tradition in political sciences

- Relational variables are more important than country characteristics to explain international trade patterns
- Focus on core-periphery and world dependency theories
- Snyder & Kick (1979), Breiger (1981), Nemeth & Smith (1985), Schott (1986), Smith & White (1992), Sacks et al. (2001), Kim & Shin (2002), Kastelle et al. (2006), Mahutga (2006)

● Econophysics enters the stage

- Serrano and Boguña (2003), *Physical Review E*
- Li, Jin, and Chen (2003), *Physica A*
- Garlaschelli and Loffredo (2004, 2005), *Physical Review Letters*

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The network of international trade flows (3/3)

● Basic ingredients of the analysis

- Take N countries and T time-periods (years)
- Collect statistics on e_{ij}^t = exports from country i to country j in year $t = 1, \dots, T$
- In each t , build a $N \times N$ adjacency matrix A^t , where $a_{ij}^t = 1$ iff $e_{ij}^t > \underline{e}$
- **Critical point:** thresholds and GDP scaling

● Problems

- Directed or undirected analysis?
- Using appropriate tools for directed analysis
- Disregarding **heterogeneity of link importance**

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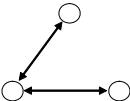
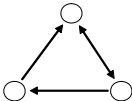
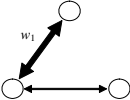
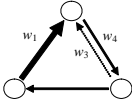
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A Taxonomy of Network Classes

- **Two Dimensions: Links can be**
 - **binary** or **weighted**
 - **undirected** or **directed**

Links	Undirected	Directed
Binary	 <p>BUN</p>	 <p>BDN</p>
Weighted	 <p>WUN</p>	 <p>WDN</p>

A Taxonomy of Network Classes

• Two Dimensions: Links can be

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Links	Undirected	Directed
Binary	Adjacency Matrix $A = \{a_{ij}\}$ Symmetric: $a_{ij} = a_{ji}$	Adjacency Matrix $A = \{a_{ij}\}$ Asymmetric
Weighted	Weight Matrix $W = \{w_{ij}\}$ Symmetric: $w_{ij} = w_{ji}$	Weight Matrix $W = \{w_{ij}\}$ Asymmetric

Binary Undirected Networks (BUNs)

● Standard BUN statistics

- Node degree (d_i)
- Average nearest-neighbor degree (*ANND*)
- Clustering coefficient
- Average shortest-distance path
- Betweenness centrality

● When is a BUN analysis appropriate?

- Suppose we can disregard link directionality
- Nature of relationships must be binary
- No heterogeneity among links
- Example: Marriage (Padgett and Ansell, 1993)
- What about airline traffic, the Internet, scientific citations, the WTW?
- Intensity or importance of links may strongly differ!

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What if these conditions are not met?

- **Employing a weighted undirected network (WUN) approach**

- From (symmetric) adjacency matrix to (symmetric) weight matrix
- Need for a generalization of BUN statistics
- Barrat et al. (2004, *PNAS*); Barthélemy et al. (2005, *Physica A*)

- **Two crucial necessary conditions**

- 1 WUN analysis should bring some value added: heterogeneity must be relevant
- 2 Results must not depend on the weighting setup

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WUN Indicators: A brief tutorial (1/2)

● Node Strength

$$s_i = \sum_j w_{ij} = W_{(i)} \mathbf{1}.$$

● Average Nearest-Neighbors Strength (ANNS)

$$anns_i = d_i^{-1} \sum_j a_{ij} s_j = d_i^{-1} \sum_j \sum_h a_{ij} w_{jh} = \frac{A_{(i)} W \mathbf{1}}{A_{(i)} \mathbf{1}}.$$

● Weighted Average of Nearest-Neighbors Degree (WANND)

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WUN Indicators: A brief tutorial (2/2)

● Node Disparity (Herfindahl Concentration Index)

$$h_i = \frac{(N-1) \sum_j \left(\frac{w_{ij}}{s_i} \right)^2 - 1}{N-2} = \frac{(N-1) \frac{1}{s_i^2} \sum_j w_{ij}^2 - 1}{N-2} = \frac{(N-1) \frac{w_{(i)}^{[2]} 1}{(w_{(i)} 1)^2} - 1}{N-2}$$

● Binary Clustering Coefficient (CC)

$$C_i(A) = \frac{\frac{1}{2} \sum_{j \neq i} \sum_{h \neq (i,j)} a_{ij} a_{ih} a_{jh}}{\frac{1}{2} d_i (d_i - 1)} = \frac{(A^3)_{ii}}{d_i (d_i - 1)}.$$

● Weighted Clustering Coefficient (WCC)

$$\tilde{C}_i(W) = \frac{\frac{1}{2} \sum_{j \neq i} \sum_{h \neq (i,j)} w_{ij}^{\frac{1}{3}} w_{ih}^{\frac{1}{3}} w_{jh}^{\frac{1}{3}}}{\frac{1}{2} d_i (d_i - 1)} = \frac{(W^{[\frac{1}{3}]})_{ii}^3}{d_i (d_i - 1)},$$

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Data

● International trade data

- Gleditsch (2002) database
- See <http://ibs.colorado.edu/~ksg/trade/>

● Data structure

- We employ a panel of 159 countries
- Time period: 1981-2000 ($T = 20$ years)
- Baseline observation
 - $\{e_{ij}^t\}$: Exports from country i to country j in year t
 - GDP_i^t and $pcGDP_i^t$ of country i in year t
- Data in current US\$ (deflated)

● Important remark

- Suppose export flows are sufficiently symmetric. . .

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Links and Weights

● Adjacency matrix A_i^t

- We follow commodity flow (rows: exporters)
- A link ij exists if $e_{ij}^t = e_{ji}^t > 0$

● Weight matrix W_i^t

- We use the baseline definition
 - Exports from i to j divided by **exporter's GDP (i)**
- But we experiment with many alternatives:
 - Exports from i to j divided by **importer's GDP (j)**
 - Exports from i to j (not scaled)
 - Same as above but now divided by total exports
- Weights are renormalized s.t. $w_i^t \in [0, 1]$

Links and Weights

- **Adjacency matrix A_i^t**

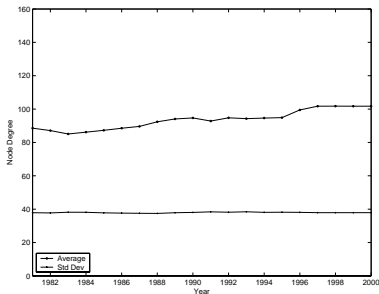
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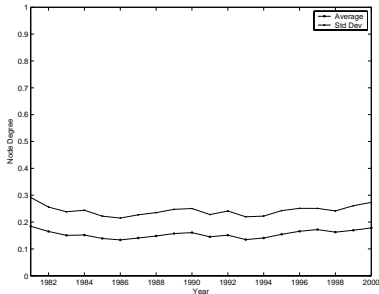
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WTW Connectivity: Average and Standard Deviation

Highly connected BUN vs. Weakly connected WUN



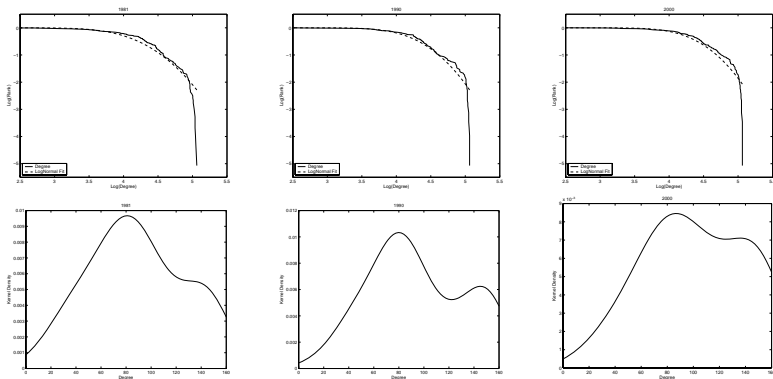
Node Degree (Ave/StDev)



Node Strength (Ave/StDev)

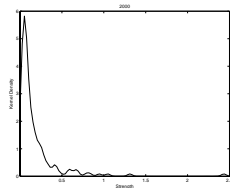
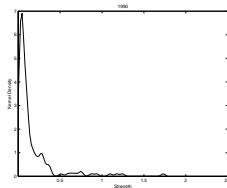
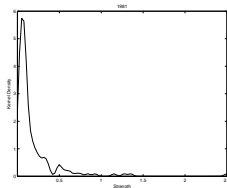
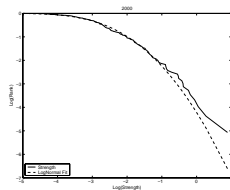
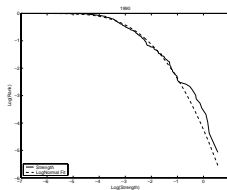
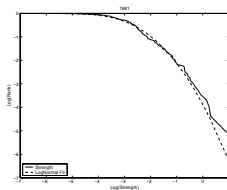
WTW Connectivity: Shape of Degree Distributions

Weak skewness; not lognormal/power-law; bimodality



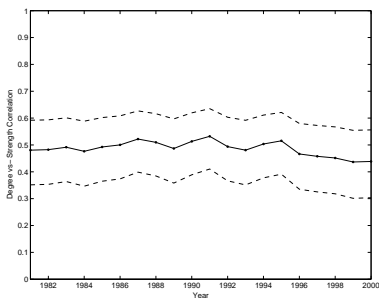
WTW Connectivity: Shape of **Strength** Distributions

High skewness; more lognormal/power-law; no bimodality

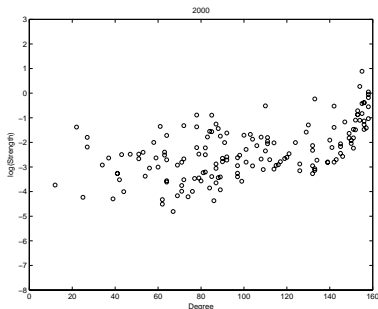


WTW Connectivity: Degree-Strength Correlation

Positive but not very strong correlation



Deg-Str Correlation



Deg vs. Str in $t=2000$

Remark: WUN analysis, a **first** value added

● WUN Connectivity

- A picture substantially different from BUN
- Trade link heterogeneity matters

● Degree-Strength Distributions

- Degree: Bimodality
- Strength: Skewed distributions, quasi scale-free, core-periphery structure

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WTW Assortativity

- Do **strongly**-connected countries trade with **strongly**-connected partners?
 - In terms of node degree (BUN) and ANND
 - In terms of node strength (WUN) and ANNS-WANND
- Networks can be
 - Assortative: Positive correlation
 - Disassortative: Negative correlation

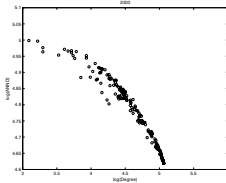
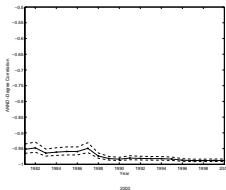
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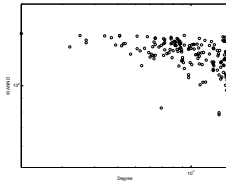
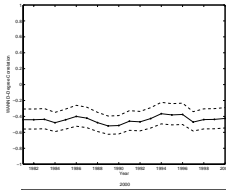
WTW Assortativity: Correlation patterns

BUN: Strongly disassortative; WUN: Weakly disassortative

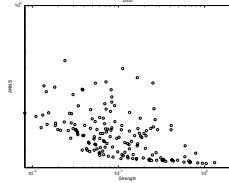
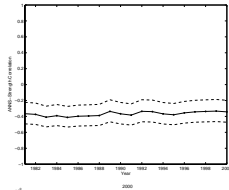
ANND-DEG



WANND-DEG



ANNS-STR



Remark: WUN analysis, a **second** value added

- **WUN Connectivity**

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- But: Emergence of intermediate periphery
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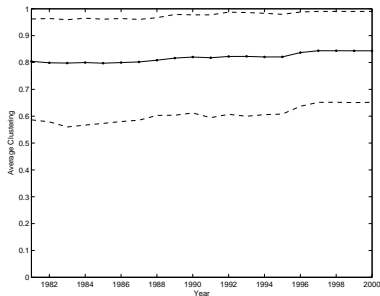
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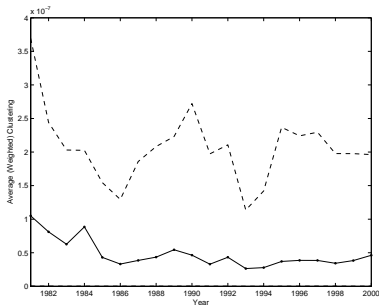
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WTW Clustering: Average Levels

Do countries hold many/intense trade relationships with countries that intensively trade with each other?



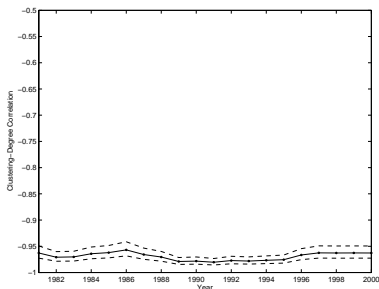
Average Node CC (Expected ~ 0.6)



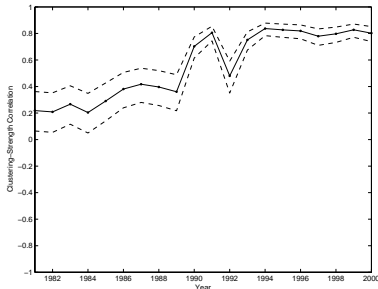
Average Node WCC (Expected ~ 0.4)

WTW Clustering: Correlation with Degree/Strength

Are better connected countries more clustered?



BUN, CC-Degree: Strongly Negative



BUN, WCC-Strength: Positive

Remark: WUN analysis, a **third** value added

● BUN Clustering

- Highly clustered on average
- Highly-connected countries hold trade partners that do not trade with each other
- Poorly-connected countries do not trade among them but are connected to the hubs

● WUN Clustering

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- Countries holding intense trade relationships are typically involved in highly-interconnected triples
- A sort of “rich club phenomenon”?

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Binary vs. Weighted Analysis: Summary

- **We have shown that**

- Link heterogeneity matters a lot in studying WTW
- If link heterogeneity is not taken into account we are disregarding a lot of information: a very different picture may emerge
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- Not at all!
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What happens if networks are directed?

● Maintained Assumption: Undirected Networks

- Majority of real-world networks are intrinsically directed
- Example: WTW
- Directed or undirected analysis?
- General Rule
 - Undirected: If network is intrinsically symmetric (marriage)
 - Directed: Must statistically detect if empirically-observed network is **sufficiently** asymmetric

● Why can't we simply employ a directed analysis?

- If network is directed (binary or weighted) appropriate tools and indicators must be employed
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WDN Indicators: Strength and ANNS

● In, Out, Total Strength

- $s_i^{in} = \sum_j w_{ji} = W_{(i)}^T \mathbf{1}$
- $s_i^{out} = \sum_j w_{ij} = W_{(i)} \mathbf{1}$
- $s_i^{tot} = s_i^{in} + s_i^{out} = (W^T + W)_{(i)} \mathbf{1}$

● Average Nearest-Neighbor Strength

- $anns_i^{out-out}$: Average out-strength of i 's out-neighbors
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WDN Indicators: Strength and ANNS

• More formally

- $anns_i^{out-out} = (d_i^{out})^{-1} \sum_j a_{ij} s_j^{out} = \frac{A_{(i)} W^T \mathbf{1}}{A_i \mathbf{1}}$
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$$anns^{tot} = \frac{(A^T + A)_{(i)} (W^T + W) \mathbf{1}}{(A^T + A)_{(i)} \mathbf{1}}$$

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Clustering in WDNs (Fagiolo, 2006b)

Patterns	Graphs	t_i^*	T_i^*	CCs for BDNs	CCs for WDNs
Cycle		$(A)_{ii}^3$	$d_i^{in} d_i^{out} - d_i^{*}$	$C_i^{cyc} = \frac{(A)_{ii}^3}{d_i^{in} d_i^{out} - d_i^{*}}$	$\tilde{C}_i^{cyc} = \frac{(\tilde{W})_{ii}^3}{d_i^{in} d_i^{out} - d_i^{*}}$
Middleman		$(AA^T A)_{ii}$	$d_i^{in} d_i^{out} - d_i^{*}$	$C_i^{mid} = \frac{(AA^T A)_{ii}}{d_i^{in} d_i^{out} - d_i^{*}}$	$\tilde{C}_i^{mid} = \frac{(\tilde{W}\tilde{W}^T \tilde{W})_{ii}}{d_i^{in} d_i^{out} - d_i^{*}}$
In		$(A^T A^2)_{ii}$	$d_i^{in} (d_i^{in} - 1)$	$C_i^{in} = \frac{(A^T A^2)_{ii}}{d_i^{in} (d_i^{in} - 1)}$	$\tilde{C}_i^{in} = \frac{(\tilde{W}^T \tilde{W}^2)_{ii}}{d_i^{in} (d_i^{in} - 1)}$
Out		$(A^2 A^T)_{ii}$	$d_i^{out} (d_i^{out} - 1)$	$C_i^{out} = \frac{(A^2 A^T)_{ii}}{d_i^{out} (d_i^{out} - 1)}$	$\tilde{C}_i^{out} = \frac{(\tilde{W}^2 \tilde{W}^T)_{ii}}{d_i^{out} (d_i^{out} - 1)}$
All (D)	All 8 graphs above	$\frac{(A+A^T)_{ii}^3}{2}$	$d_i^{tot} (d_i^{tot} - 1) - 2d_i^{*}$	$C_i^D = \frac{(A+A^T)_{ii}^3}{2T_i^D}$	$\tilde{C}_i^D = \frac{(\tilde{W}+\tilde{W}^T)_{ii}^3}{2T_i^D}$

Two crucial issues

- **How can we decide whether a directed analysis should be preferred?**
 - Computing percentage of reciprocated links
 - Correlation between upper and lower diagonal entries (see Garlaschelli and Loffredo, 2004, *Physical Review Letters*)
 - More robust statistical checks?
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A Simple Index (Fagiolo, 2006a)

● Idea

- The more the network is undirected, the smaller $\|\tilde{W} - \tilde{W}^T\|$ (appropriately normalized)

● Technical assumption

$$Q = \{q_{ij}\} = \tilde{W} - (1 - \tilde{W})I_N$$

● Define

$$\tilde{S}(Q) = \frac{\|Q - Q^T\|_F^2}{\|Q\|_F^2 + \|Q^T\|_F^2} = \frac{\|Q - Q^T\|_F^2}{2\|Q\|_F^2} = \frac{1}{2} \left[\frac{\|Q - Q^T\|_F}{\|Q\|_F} \right]^2$$

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- Expanding the sums...

$$\tilde{S}(Q) = 1 - \frac{\sum_i \sum_j q_{ij} q_{ji}}{\sum_i \sum_j q_{ij}^2}.$$

- To get an index in $[0, 1]$, define:

$$S(Q) = \frac{N+1}{N-1} \tilde{S}(Q),$$

- We can find $(m_W(N), s_W(N))$ such that:

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● Using the index

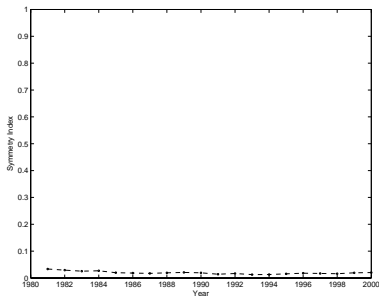
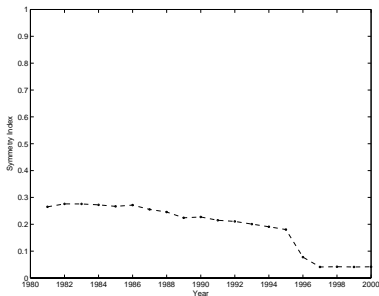
- Define $Q = \{q_{ij}\} = \tilde{W} - (1 - \tilde{W})I_N$
- Compute the index $S_W(Q)$
- Fix a threshold k (in term of standard deviations)
- If $S_W(Q) > k$ the graph is **asymmetric**

TABLE I: The index S and its standardized version $S_{\{*\}}, \{*\} = \{B(inary), W(eighed)\}$ for social networks studied in [3], cf. Chapter 2.5.

	Social Network	N	S	S_*
1	Advice relations btw Krackhardt's hi-tech managers	21	0.521327	0.491228
2	Friendship relations btw Krackhardt's hi-tech managers	21	0.500813	0.004610
3	"Reports-to" relations btw Krackhardt's hi-tech managers	21	0.536585	0.860033
4	Business relationships btw Padgett's Florentine families	16	0.000000	-9.232823
5	Marital relationships btw Padgett's Florentine families	16	0.000000	-9.232823
6	Acquaintanceship among Freeman's EIES researchers (Time 1)	32	0.109849	-10.025880
7	Acquaintanceship among Freeman's EIES researchers (Time 2)	32	0.094968	-11.143250
8	Messages sent among Freeman's EIES researchers	32	0.014548	-17.181580
9	Country Trade Flows: Basic Manufactured Goods	24	0.260349	-6.643695
10	Country Trade Flows: Food and Live Animals	24	0.311966	-5.217508
11	Country Trade Flows: Crude Materials (excl. Food)	24	0.272560	-6.306300
12	Country Trade Flows: Minerals, Fuels, Petroleum	24	0.403336	-2.692973
13	Country Trade Flows: Exchange of Diplomats	24	0.080208	-11.620970

Results for the WTW

- **The WTW is extremely symmetric**
 - Binary vs. weighted: no differences
 - Symmetric under all weighting schemes
 - Procedure employed above was appropriate



WTW: A Weighted Directed Network Analysis

- **Is it worthwhile anyway?**

- Despite strong symmetry, does a WDN bring any value added?

- **Clustering associated to different triangle patterns**

- Triangles and their meaning in terms of export/import
- Heterogeneity: CC ranges from 0.0004 to 0.0013
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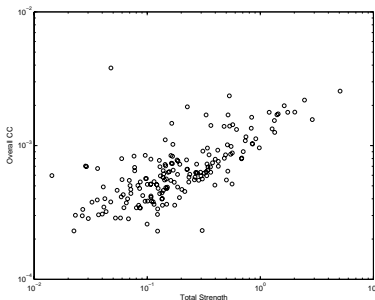
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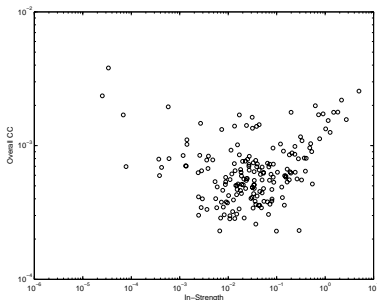
WTW: Clustering-Strength Correlation (1/2)

● How do Directed WCC correlate with strength?

- WCC vs. tot-strength: positive correlation
- WCC vs. in-strength: U-shaped
- Low clustering level of weakly connected countries mainly depends on their weakly exporting relationships



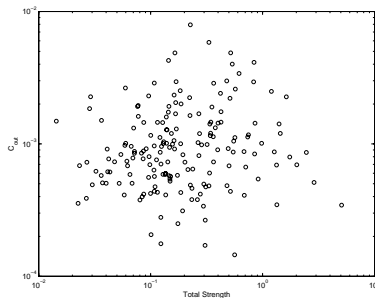
Tot-Str



In-Str

WTW: Clustering-Strength Correlation (2/2)

- **WCC for different triangles vs. strength**
 - WCC for **cyc**, **mid**, **in** are positively correlated with strength
 - WCC for **out** not correlated with strength
 - Countries hold exporting relationships with connected pairs of countries independently of total strength



Directed vs. Undirected Analysis

- **We have shown that**

- Apart from extreme cases, deciding whether to employ a directed or an undirected analysis is an empirical issue
- It is possible to introduce an index to check for network symmetry/asymmetry
- This index is not a hypothesis test but has nice statistical properties

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