



# Corrected network measures

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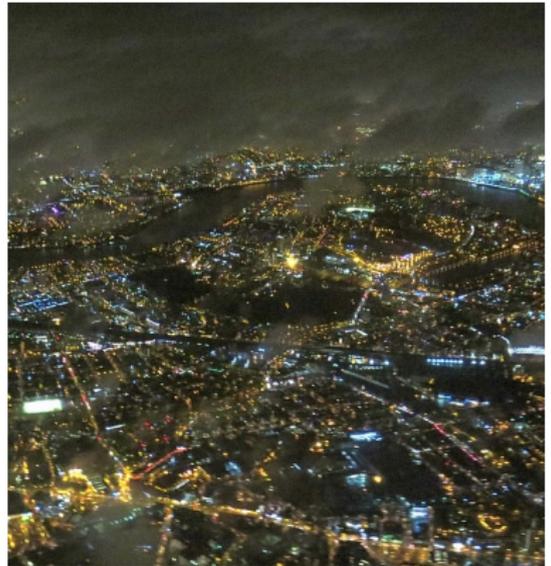
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**Current version of slides (December 16, 2015, 11:05):**

<http://vlado.fmf.uni-lj.si/pub/slides/ercim15.pdf>



# Network element importance measures

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To identify important / interesting elements (nodes, links) in a network we often try to express our intuition about important / interesting element using an appropriate measure (index, weight) following the scheme

*larger is the measure value of an element,  
more important / interesting is this element*

Too often, in analysis of networks, researchers uncritically pick some measure from the literature.



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We discuss two well known network measures: the overlap weight of an edge (Onnela et al., 2007) and the clustering coefficient of a node (Holland and Leinhardt, 1971; Watts and Strogatz, 1998) .

For both of them it turns out that they are not very useful for data analytic task to identify important elements of a given network. The reason for this is that they attain the largest values on "complete" subgraphs of relatively small size – they are more probable to appear in a network than that of larger size.

We show how their definitions can be corrected in such a way that they give the expected results.

The (topological) *overlap weight* of an edge  $e = (u : v) \in \mathcal{E}$  in an undirected simple graph  $\mathbf{G} = (\mathcal{V}, \mathcal{E})$  is defined as

$$o(e) = \frac{t(e)}{(\deg(u) - 1) + (\deg(v) - 1) - t(e)}$$

where  $t(e)$  is the number of triangles (cycles of length 3) to which the edge  $e$  belongs. In the case  $\deg(u) = \deg(v) = 1$  we set  $o(e) = 0$ .

Introducing two auxiliary quantities

$$m(e) = \min(\deg(u), \deg(v)) - 1 \quad \text{and} \quad M(e) = \max(\deg(u), \deg(v)) - 1$$

we can rewrite the definition

$$o(e) = \frac{t(e)}{m(e) + M(e) - t(e)}, \quad M(e) > 0$$

and if  $M(e) = 0$  then  $o(e) = 0$ .

It holds

$$0 \leq t(e) \leq m(e) \leq M(e).$$

Therefore

$$m(e) + M(e) - t(e) \geq t(e) + t(e) - t(e) = t(e)$$

showing that  $0 \leq o(e) \leq 1$ .

The value  $o(e) = 1$  is attained exactly in the case when  $m(e) = M(e) = t(e)$ ; and the value  $o(e) = 0$  exactly when  $t(e) = 0$ .



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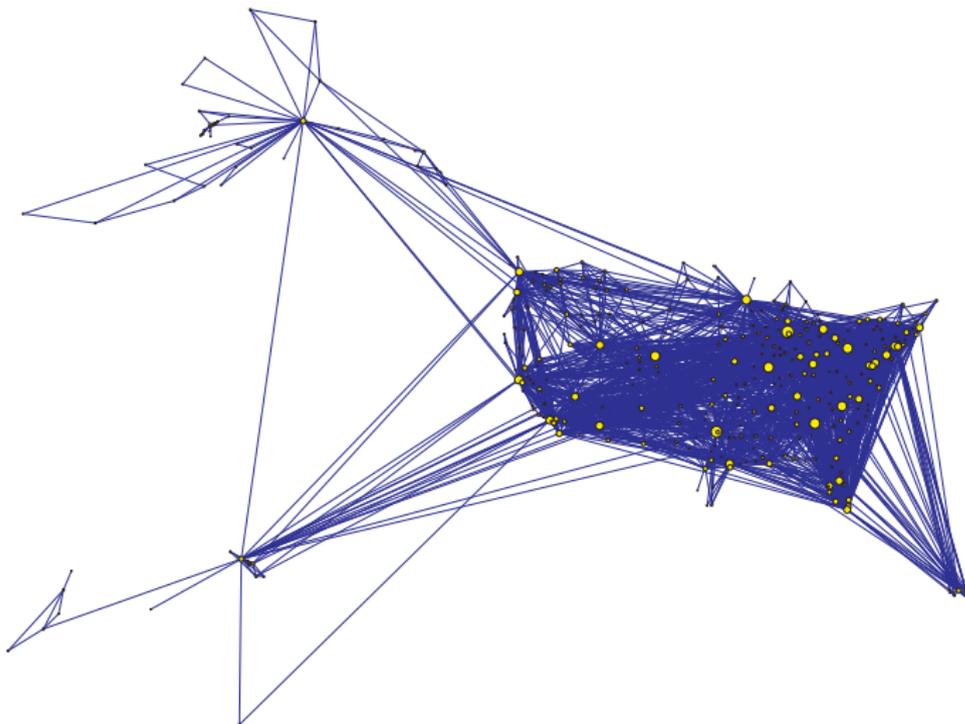
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# Edges with the largest overlap cut at 0.8

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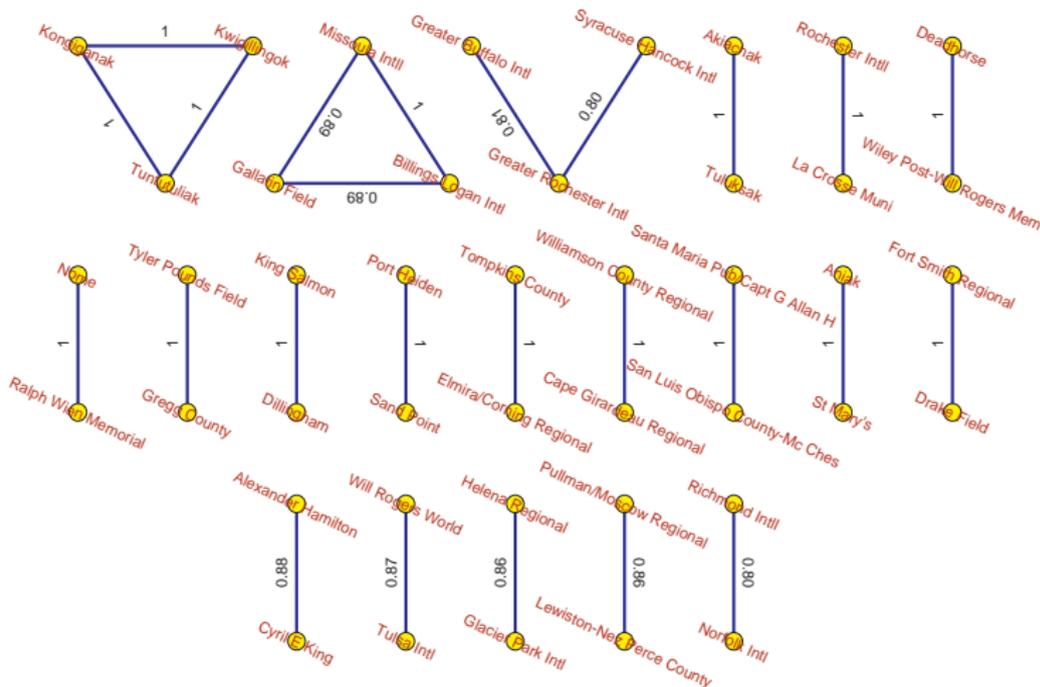
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# Zoom in

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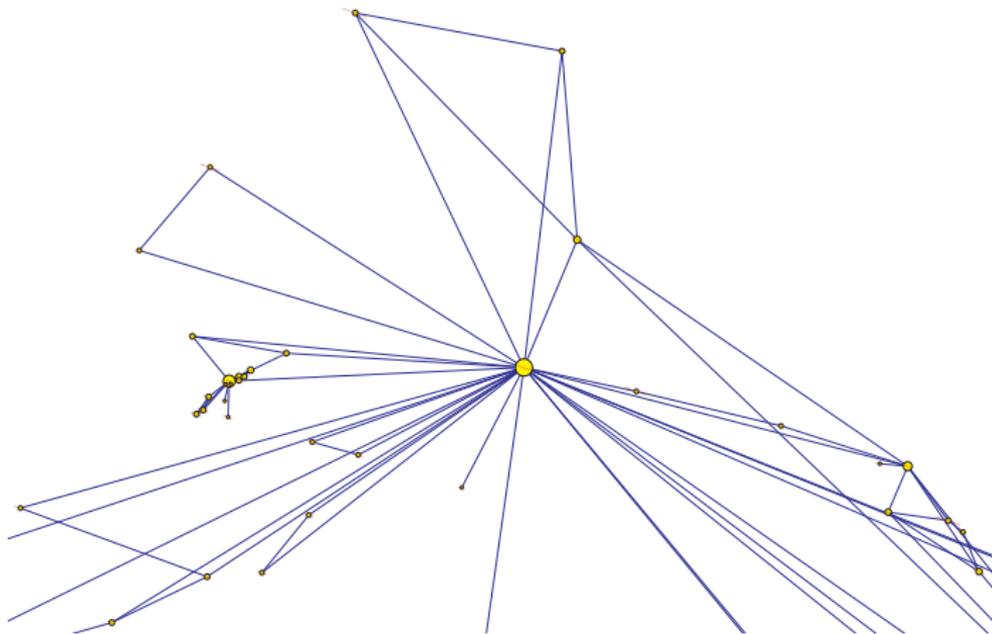
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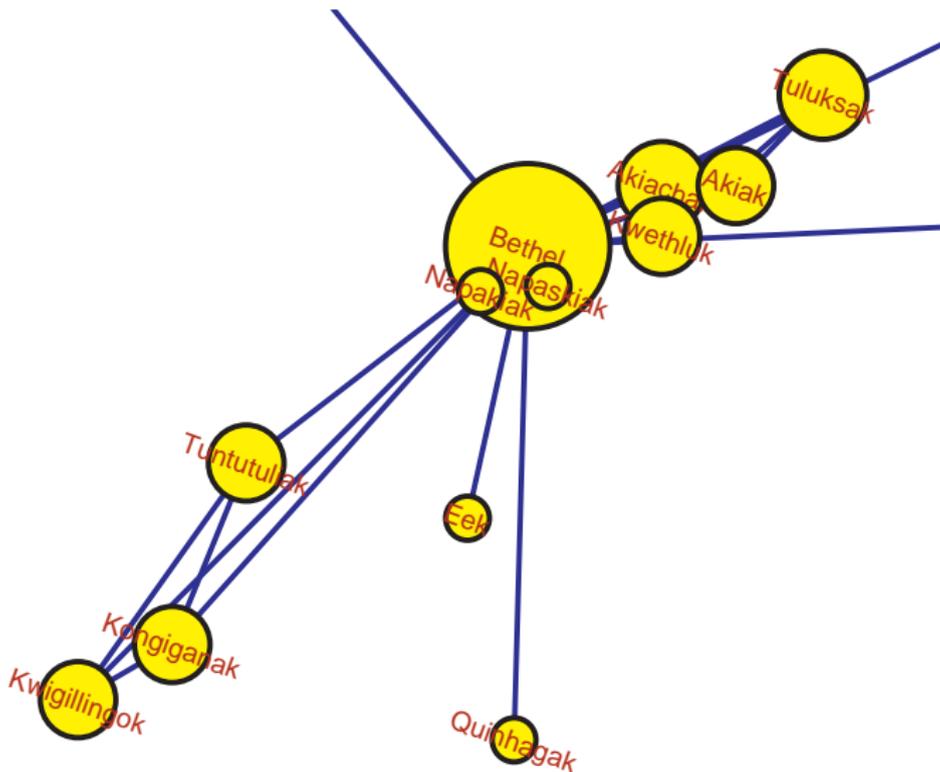
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# Observation

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From this example we see that in real-life networks edges with the largest overlap weight tend to be edges with relatively small degrees in their end-nodes. Because of this the overlap weight is not very useful for data analytic tasks in searching for important elements of a given network. We can try to improve the overlap weight definition to better suit the data analytic goals.

For this we introduce a quantity

$$\mu = \max_{e \in \mathcal{E}} t(e)$$

We define a *corrected overlap weight* as

$$o'(e) = \frac{t(e)}{\mu + M(e) - t(e)}$$

By the definition of  $\mu$  for every  $e \in \mathcal{E}$  it holds  $t(e) \leq \mu$ . Since  $M(e) - t(e) \geq 0$  also  $\mu + M(e) - t(e) \geq \mu$  and therefore  $0 \leq o'(e) \leq 1$ . Also  $o'(e) = 0$  exactly when  $t(e) = 0$ . But,  $o'(e) = 1$  exactly when  $\mu = M(e) = t(e)$ .



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with the largest corrected overlap weight, cut at 0.5

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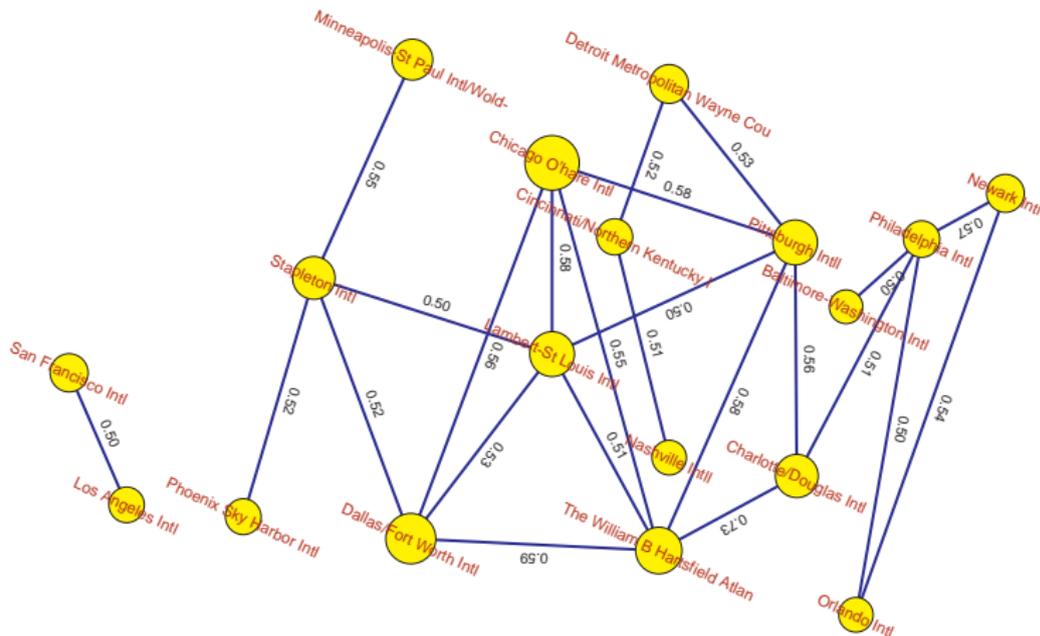
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$$\mu = 80$$



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with the largest corrected overlap weight

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	u	v	t(e)	d(u)	d(v)	o'(e)
	The WB Hartsfield Atlan	Charlotte/Douglas Intl	= 76	101	87	0.73077
	The WB Hartsfield Atlan	Dallas/Fort Worth Intl	= 73	101	118	0.58871
	Chicago O'hare Intl	Pittsburgh Intl	= 80	139	94	0.57971
	Chicago O'hare Intl	Lambert-St Louis Intl	= 80	139	94	0.57971
	Dallas/Fort Worth Intl	Chicago O'hare Intl	= 78	118	139	0.55714
	The WB Hartsfield Atlan	Chicago O'hare Intl	= 77	101	139	0.54610



# US Airports links

$\rho'( \text{WB Hartsfield Atlanta, Charlotte/Douglas Intl} ) = 0.7308$

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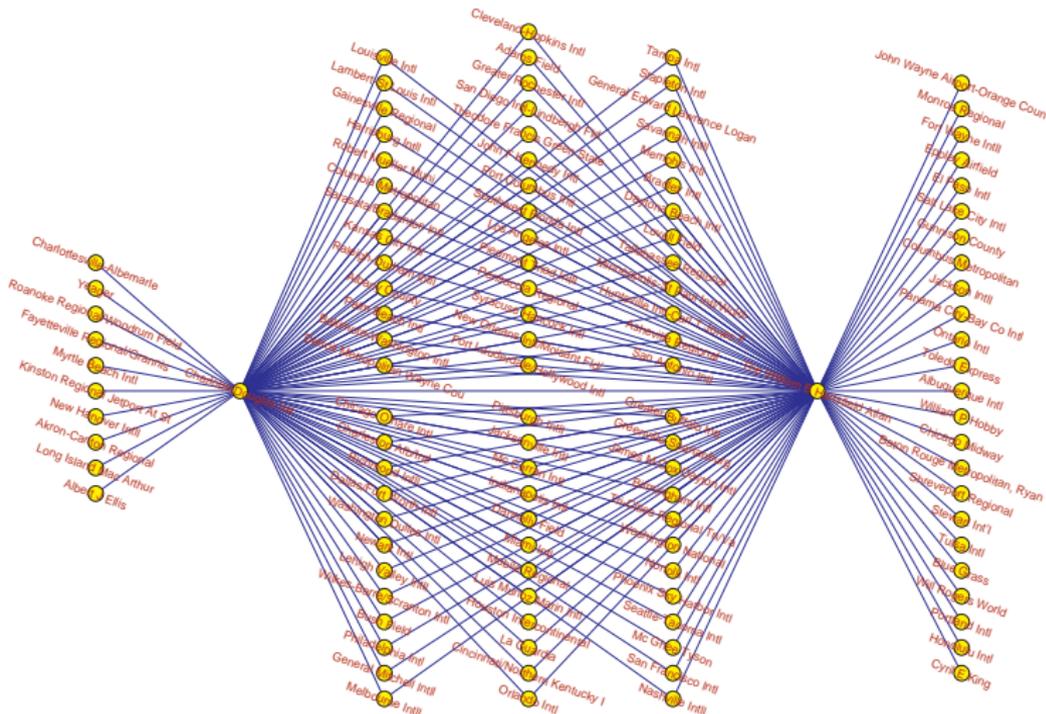
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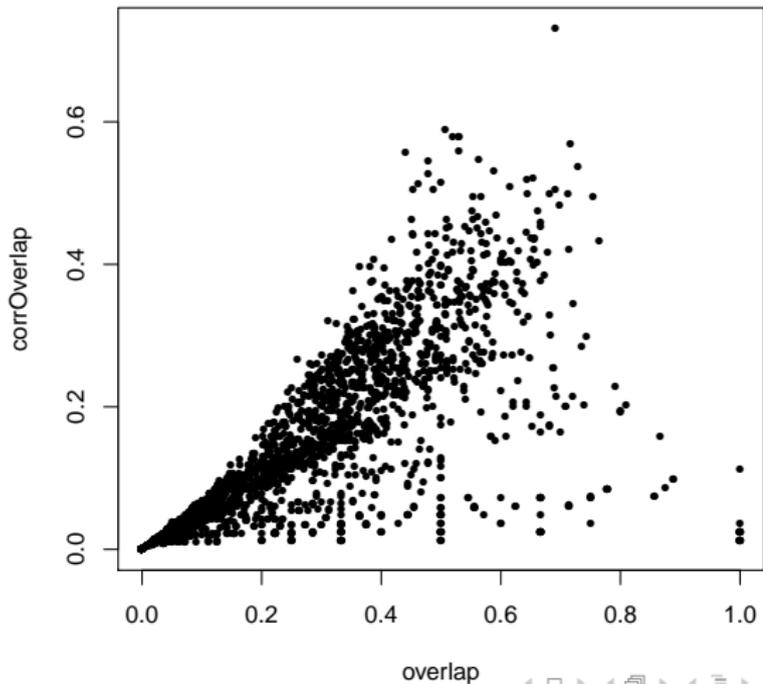
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### Overlap weights





# Comparison – $\min\text{Deg}(e)/\max\text{Deg}(e)$

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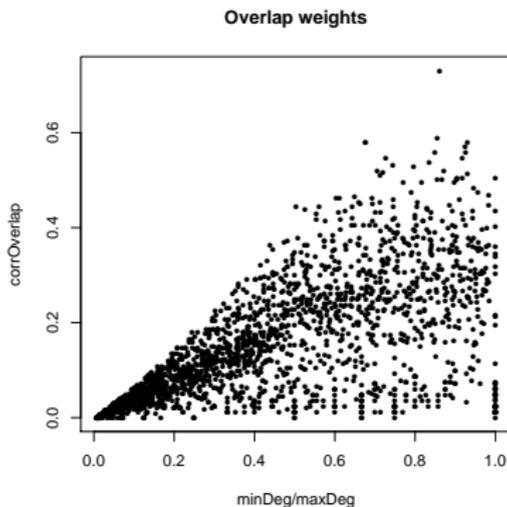
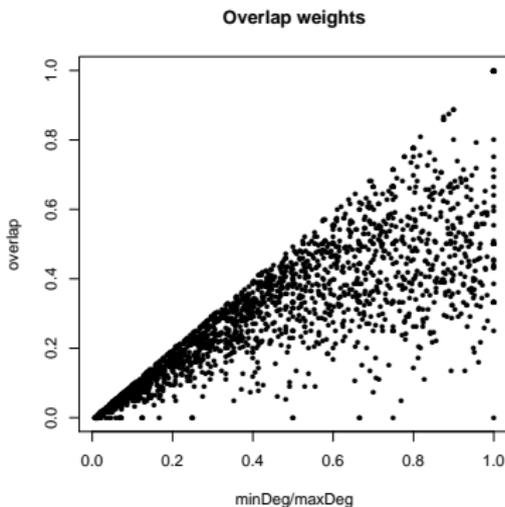
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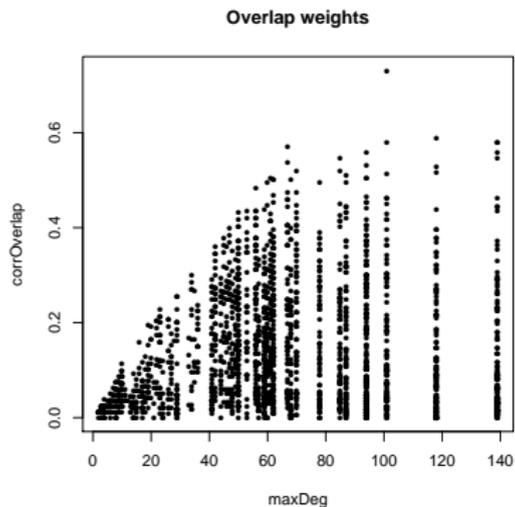
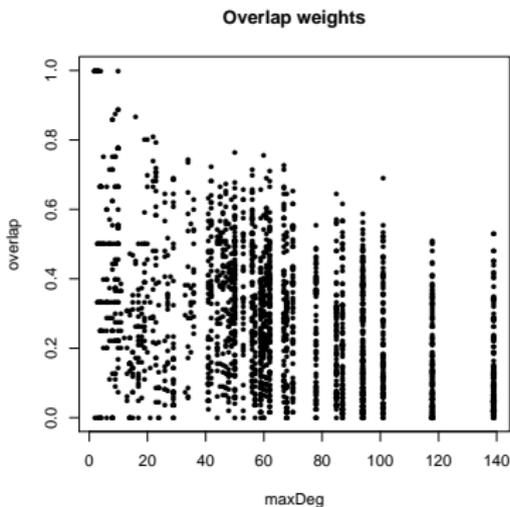
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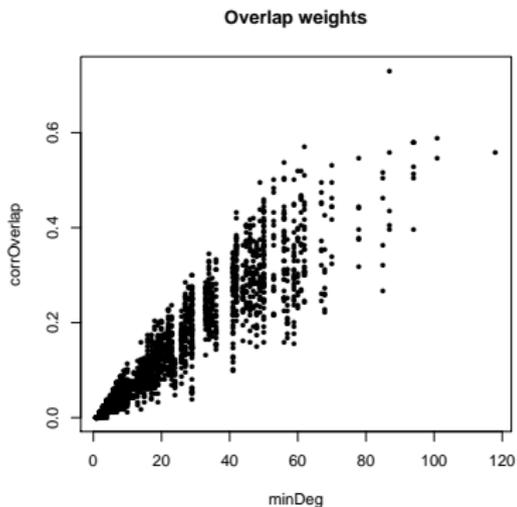
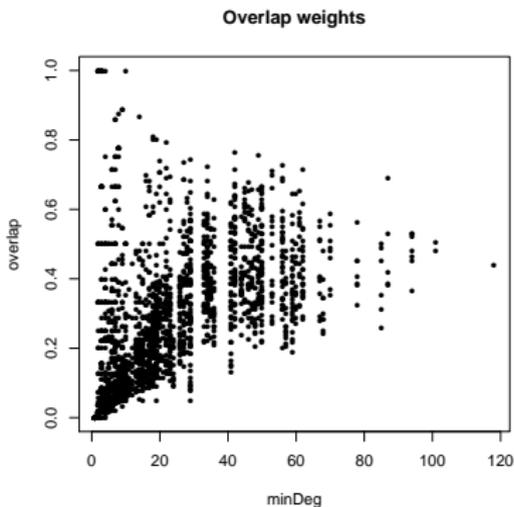
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# Comparison – # of triangles

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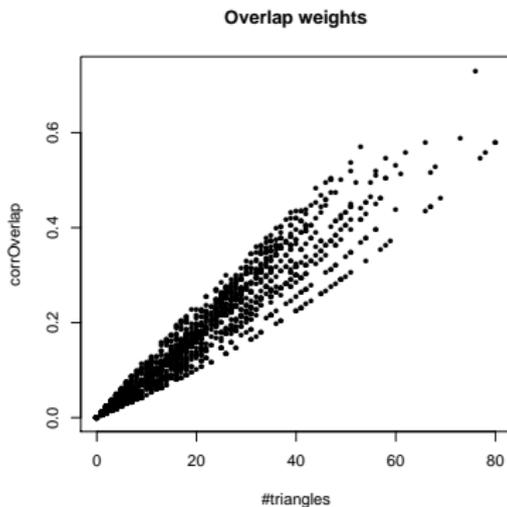
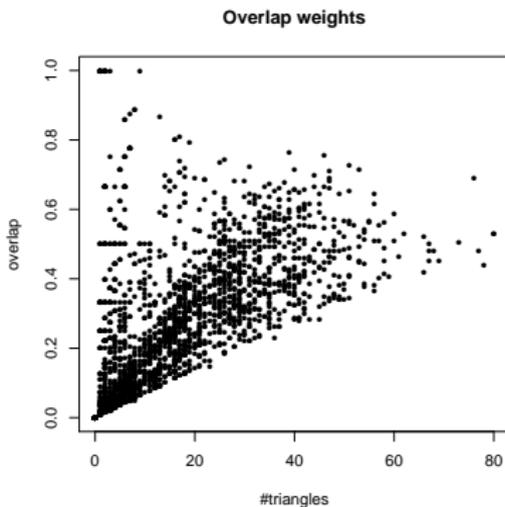
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For a node  $u \in \mathcal{V}$  in an undirected simple graph  $\mathbf{G} = (\mathcal{V}, \mathcal{E})$  its clustering coefficient is measuring a local density in node  $u$  and is defined as

$$cc(u) = \frac{|\mathcal{E}(N(u))|}{|\mathcal{E}(K_{\deg(u)})|} = \frac{2 \cdot E(u)}{\deg(u) \cdot (\deg(u) - 1)}, \quad \deg(u) > 1$$

where  $N(u)$  is the set of neighbors of node  $u$ . If  $\deg(u) \leq 1$  then  $cc(u) = 0$ .

It is easy to see that

$$E(u) = \frac{1}{2} \sum_{e \in S(u)} t(e)$$

where  $S(u)$  is the star in node  $u$ .

It holds  $0 \leq cc(u) \leq 1$ .  $cc(u) = 1$  exactly when  $\mathcal{E}(N(u))$  is isomorphic to  $K_{\deg(u)}$ .



# US Airports links with clustering coefficient = 1

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1	Wiley Post-Will Rogers Mem	28	Kwethluk	55	Kongiganak
2	Ralph Wien Memorial	29	Hector Intll	56	Bellingham Intl
3	Aniak	30	Tompkins County	57	La Crosse Muni
4	Toledo Express	31	Cape Girardeau Regional	58	Hilo Intll
5	Myrtle Beach Intl	32	Merced Municipal/Macready Fie	59	Rochester Intl
6	Rota Intl	33	King Salmon	60	Kapalua
7	Jack Mc Namara Field	34	Modesto City-County--Harry Sh	61	Lihue
8	Port Heiden	35	Natrona County Intl	62	Mc Allen Miller Intl
9	New Hanover Intll	36	Williamson County Regional	63	Rio Grande Valley Intl
10	Santa Maria Pub/Capt G Allan	37	Deadhorse	64	Eareckson As
11	Fayetteville Regional/Grannis	38	Nome	65	Corpus Christi Intl
12	Lovell Field	39	Akiak	66	St Petersburg/Clearwater In
13	St Paul Island	40	Dillingham	67	Lehigh Valley Intll
14	Elmira/Corning Regional	41	Evansville Regional	68	Gainesville Regional
15	San Luis Obispo County-Mc Che	42	Charlottesville-Albemarle	69	Burlington Regional
16	Binghamton Regional/Edwin A L	43	Bishop Intll	70	Lafayette Regional
17	Fort Smith Regional	44	Gunnison County	71	Tuntutuliak
18	St Mary's	45	Friedman Memorial	72	Tallahassee Regional
19	Asheville Regional	46	Aspen-Pitkin Co/Sardy Field	73	University Park
20	Molokai	47	Mbs Intll	74	Sand Point
21	Worcester Muni	48	Kwigillingok	75	Tyler Pounds Field
22	Drake Field	49	Minot Intl	76	Tweed-New Haven
23	Dubuque Regional	50	Pago Pago Intl	77	Gregg County
24	Tri-Cities Regional Tn/Va	51	Babelthuap/Koror	78	Wilkes-Barre/Scranton Intl
25	Monterey Peninsula	52	Decatur	79	Eastern Oregon Regional At
26	Detroit City	53	Quincy Muni Baldwin Field	80	Stewart Intl
27	Joplin Regional	54	Rafael Hernandez		

Again we see that the clustering coefficient attains its largest value in nodes with relatively small degree. The probability that we get a complete subgraph on  $N(u)$  is decreasing fast with increasing of  $\text{deg}(u)$ .



To get a corrected version of the clustering coefficient we proposed in Pajek to replace  $\deg(u)$  in the denominator with  $\Delta = \max_{v \in \mathcal{V}} \deg(v)$ . In this paper we propose another solution – we replace  $\deg(u) - 1$  with  $\mu$ :

$$cc'(u) = \frac{2 \cdot E(u)}{\mu \cdot \deg(u)}, \quad \deg(u) > 0$$

To show that  $0 \leq cc'(u) \leq 1$  we have to consider two cases:

a.  $\deg(u) \geq \mu$ : then for  $v \in N(u)$  we have  $\deg_{N(u)}(v) \leq \mu$  and therefore

$$2 \cdot E(u) = \sum_{v \in N(u)} \deg_{N(u)}(v) \leq \sum_{v \in N(u)} \mu = \mu \cdot \deg(u)$$

b.  $\deg(u) < \mu$ : then  $\deg(u) - 1 \leq \mu$  and therefore

$$2 \cdot E(u) \leq \deg(u) \cdot (\deg(u) - 1) \leq \mu \cdot \deg(u)$$

The value  $cc'(u) = 1$  is attained in the case a on a  $\mu$ -core, and in the case b on  $K_{\mu+1}$ .



# US Airports links

with the largest corrected clustering coefficient

Corrected  
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Rank	Value	Id	Rank	Value	Id
1	0.3739	Cleveland-Hopkins Intl	26	0.2990	Minneapolis-St Paul Intl/Wold-
2	0.3700	General Edward Lawrence Logan	27	0.2956	General Mitchell Intl
3	0.3688	Orlando Intl	28	0.2942	Phoenix Sky Harbor Intl
4	0.3595	Tampa Intl	29	0.2935	Palm Beach Intl
5	0.3488	Cincinnati/Northern Kentucky I	30	0.2914	Charlotte/Douglas Intl
6	0.3457	Detroit Metropolitan Wayne Cou	31	0.2881	Memphis Intl
7	0.3455	Newark Intl	32	0.2859	Lambert-St Louis Intl
8	0.3429	Baltimore-Washington Intl	33	0.2847	San Diego Intl-Lindbergh Fld
9	0.3415	Miami Intl	34	0.2824	Pittsburgh Intl
10	0.3405	Washington National	35	0.2762	Stapleton Intl
11	0.3379	Nashville Intl	36	0.2724	Washington Dulles Intl
12	0.3359	John F Kennedy Intl	37	0.2661	Dallas/Fort Worth Intl
13	0.3347	Philadelphia Intl	38	0.2595	Raleigh-Durham Intl
14	0.3335	Indianapolis Intl	39	0.2541	Chicago O'hare Intl
15	0.3335	La Guardia	40	0.2489	San Francisco Intl
16	0.3311	Mc Carran Intl	41	0.2386	Greater Buffalo Intl
17	0.3301	Fort Lauderdale/Hollywood Intl	42	0.2295	John Wayne Airport-Orange Coun
18	0.3106	New Orleans Intl/Moisant Fld/	43	0.2241	Seattle-Tacoma Intl
19	0.3095	Bradley Intl	44	0.2211	Sarasota/Bradenton Intl
20	0.3045	Port Columbus Intl	45	0.2207	Ontario Intl
21	0.3038	Los Angeles Intl	46	0.2175	Syracuse Hancock Intl
22	0.3036	Houston Intercontinental	47	0.2163	San Jose Intl
23	0.3036	Kansas City Intl	48	0.2158	Norfolk Intl
24	0.3017	Southwest Florida Intl	49	0.2144	Salt Lake City Intl
25	0.3002	The William B Hartsfield Atlan	50	0.2056	Greater Rochester Intl



# Cleveland-Hopkins Intl neighbors

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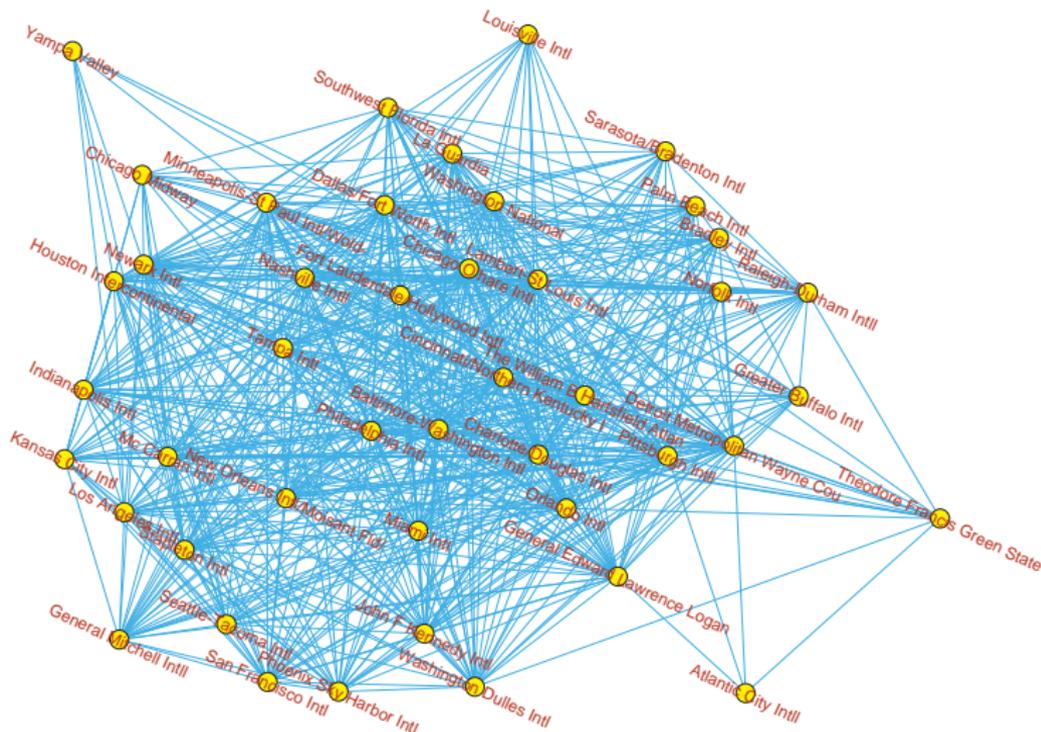
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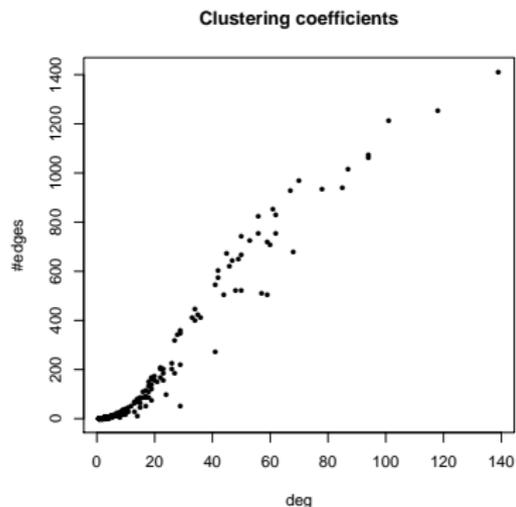
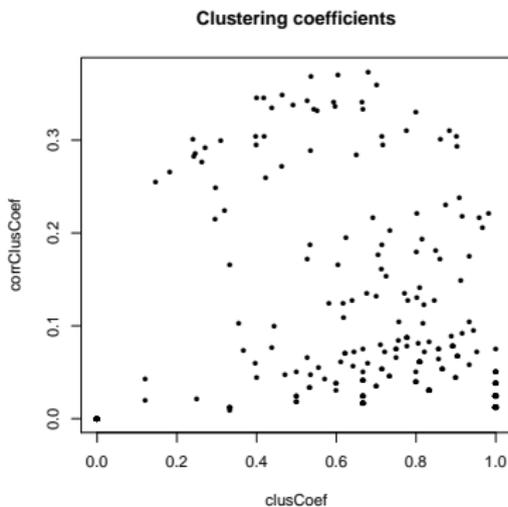
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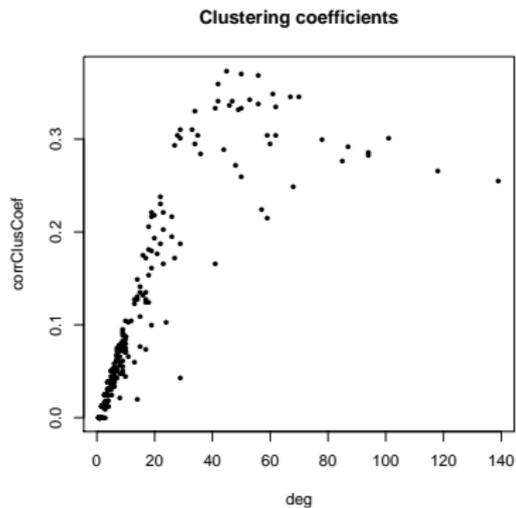
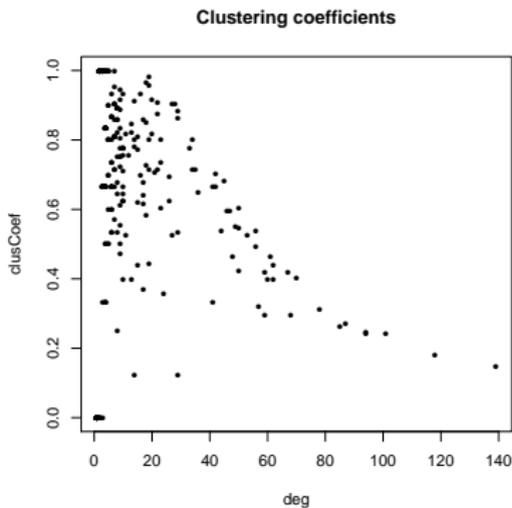
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In the corrected measures we can replace  $\mu$  with  $\Delta$ . Its advantage is that it can be easier computed; but the corresponding measure is less 'sensitive'.



P. W. Holland and S. Leinhardt (1971). "Transitivity in structural models of small groups". *Comparative Group Studies* 2: 107–124.



Onnela, J.P., Saramaki, J., Hyvonen, J., Szabo, G., Lazer, D., Kaski, K., Kertesz, J., Barabasi, A.L.: Structure and tie strengths in mobile communication networks. *Proceedings of the National Academy of Sciences* 104(18), 7332 (2007) [paper](#)



D. J. Watts and Steven Strogatz (June 1998). "Collective dynamics of 'small-world' networks". *Nature* 393 (6684): 440–442.



Wikipedia: [Clustering coefficient](#)



Wikipedia: [Overlap coefficient](#)



# Acknowledgments

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