### **Reputation Risk Contagion**

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# **Overview of aims and methodology**

Overall aim: to measure to what extent the reputation of one organisation is affected by the reputation of other similar organisations

- We measure reputation by data mining targeted content, followed by sentiment analysis of that content. Result: a single-number measurement of reputation on a per-day basis.
- Use the reputation measure to elucidate a network structure, using a Bayesian methodology. (Nothing is assumed about such a network a priori.)
- Use the *de Groot* method to measure consensus, and hence the proportion of reputation due to systemic factors.



# What is reputation?

"Reputation"

A perception of an organisation on the part of stakeholders that can affect, **positively or negatively**, the business relationship between the stakeholder and the organisation

"Reputation Event" - An occurrence or action that affects Reputation

"Reputation Risk" - The difference between stakeholder expectation and organisation performance<sup>(1)</sup>

"Reputation Risk Measurement" - Numerical assessment of Reputation

(1) Federal Reserve Boston (1995) Supervisory Letter SR 95-51 (SUP): Rating the Adequacy of Risk Management Processes



#### www.alva-group.com

Measurement





### **Example content scoring**

1. On Twitter, @blognewcastle (203 followers) wrote: "*I'm a big fan of @santanderuk* (11 Dec 2015)

Category	Sentiment	Score, s
Sentiment	Positive, qualified by 'big'	8.0
Influence	Few followers: not influential	1.0
Prominence	Neutral	5.5
Relevance	No references to other organisations	10.0

Content Score = 24.5/4 = 6.125



### **Example index compilation**

Content		Score, m	Weight, w	m×w
C1	"I'm a big fan of @XYZ-Bank"	6.125	0.12	0.735
C2	"XYZ-Bank does hardly provides good service" (Local TV consumer feature)	4.7	0.6	2.82
C3	"XYZ-Bank's mortgage interest rates is the best available" (Sunday Times 'Best Buy' tables)	8.62	0.9	7.758
Sum			1.62	11.313
	Weights reflect importance of content and source	Index	/alue = 11.31	3/1.62 = 6.9



Useful view: cumulative sentiment – used later to assess 'network drag'





#### **Sentiment Analysis references**

Comprehensive review and analysis: "Sentiment Analysis", Bing Liu 2015

Preliminary work: (e.g.) Wiebe 1990 and 1994, Hearst 1992

Early work: (e.g.) Wiebe (2000), Das and Chen (2001), Tong (2001), Nasukawa & Lee (2003) – "Sentiment Analysis" Dave et al (2003) – "Opinion Mining"



# De Groot model for opinion formation <sup>(1)</sup>

- Described by a network of arbitrary complexity, with an influence matrix, *T*. In this case its structure is not known a priori
- *T<sub>ij</sub>* is represents the weight that agent *i* places on the current belief of agent *j* in forming agent *i*'s opinion
- Agents start with an initial opinion p(r=0), interact with other agents, and at the next time step (r=1), update their own opinion to p(r=1)based on *T*. Further iterations produce p(r=2),  $p(r=3)^{(2,3)}$ ...
- Assumption: full accessibility of information<sup>(4)</sup>
- (1) DeGroot, M.H. (1974) Reaching a Consensus. Jnl. American Statistical Association (69). 118-121
- (2) DeMarzo, P., Vayanos, D. and Zwiebel, J. (2003) Persuasion Bias, Social Influence and Unidimentional Opinions. Quarterly Journal of Economics (118) 909-968
- (3) Golub, B. and Jackson. M.O. (2010) *Naive Learning in Social Networks and the Wisdom of Crowds*. American Economic Journal Microeconomics. 112-149
- (4) Pan, Z (2012) Opinions and Networks: How Do They Effect Each Other. Comput Econ 39,157–171



# **De Groot model for opinion formation**

p(1) = Tp(0)

In general: p(r) = Tp(r-1)which implies  $p(r) = T^rp(0), r = 1, 2, ...$ 

There may be a limiting case that represents converged opinion <sup>(1)</sup>:

$$p(\infty) = \lim_{r \to \infty} \left( T^r p(0) \right)$$

(1) Chatterjee, S. and Seneta, E. (1977) Towards Consensus: Some Convergence Theorems on Repeated Averaging. J. Appl. Prob. 14, 89-97



We have to discover a network based on agents' sentiment with respect to banks, and then derive the corresponding influence matrix T. In many other cases it's the other way round: the network is given and T is derived from it.

Let S(i, t) be the sentiment of Agent *i* on day *t*. Then the sentiment movement is M(i, t) = S(i, t) - S(i, t-1).

We count all movements greater than or equal to a 'high' threshold  $\lambda_H$  and all movements greater than or equal to a 'very high' threshold  $\lambda_{VH}$ .

 $C(i, \lambda) = \{ M(i, t): abs(M(i, t)) \ge \lambda, 1 \le t \le n \}, \text{ where } \lambda = \lambda_H \text{ or } \lambda_{VH} \}$ 



Distribution of movements M(i, t) $\lambda_H$  marks the extreme 5% of movements  $\lambda_{VH}$  marks the extreme 1% of movements



Index first difference



Drive the influence matrix *T* using a Bayesian approach:

Given an Agent *i*, and a *different* Agent *j*, count the number of very large movements in the sentiment of Agent *j* ( $i \neq j$ ), given that there was a *large* movement in the sentiment of Agent *i*.

$$T_{ij} = C(j, \lambda_{VH}) | C(i, \lambda_{H}) = (C(j, \lambda_{VH}) \text{ and } C(i, \lambda_{H})) / C(i, \lambda_{H})$$

(a large movement in the sentiment of Agent i, associated with a very large movement in the sentiment of Agent j implies that Agent i has influenced Agent j)



In the case *i* = *j* there is a different interpretation.

it's a measure of the extent to which agent *i* values its own opinion, where 'agent' means all those who comment.

From the equation for  $T_{ij}$ 

 $C(j, \lambda_{VH}) = C(i, \lambda_{H})), \text{ so}$  $T_{ii} = C(i, T_{VH})/C(i, T_{H}).$ 



	(0.459	0.084	0	0	0	0.079	0.115	0.189	0.074	0 )
-	0.088	0.237	0	0	0	0.131	0.177	0.192	0.062	0.113
	0	0	0.238	0	0.133	0.122	0.133	0.179	0.194	0
	0	0	0.109	0.437	0	0.160	0.140	0.154	0	0
T	0	0	0.103	0.099	0.375	0	0.146	0.181	0.096	0
	0.050	0.040	0	0	0	0 470	0 00 1	0 1 4 0	0 072	
	0.052	0.049	0	0	0	0.4/9	0.204	0.142	0.0/3	0
	0.052 0.058	0.049 0.078	0	0.038	0	0.479	0.204 0.518	0.142 0.084	0.073	0 0.041
	0.052 0.058 0.049	0.049 0.078 0.095	0 0 0.050	0 0.038 0	0 0 0.033	0.479 0.086 0.078	0.204 0.518 0.120	0.142 0.084 0.483	0.073 0.097 0.060	0 0.041 0.033
	0.052 0.058 0.049 0	0.049 0.078 0.095 0	0 0 0.050 0	0 0.038 0 0.064	0 0 0.033 0.109	0.479 0.086 0.078 0.162	0.204 0.518 0.120 0.084	0.142 0.084 0.483 0.117	0.073 0.097 0.060 0.465	0 0.041 0.033 0

Zero entries indicate that the corresponding network is not fully connected: not all agents can influence all others directly.



### Network corresponding to T

thin = non-influential thick = influential

Not all edges are bidirectional







In practice we observe convergence for  $T^t$  for r > 6The network corresponding to  $T_{\infty}$  is fully connected







Network corresponding to  $T_{\infty}$ 

Surprising results!

- Agents 6, 7, 8 and 9 are most influential: they do not attract extreme negative comment.
- (Lloyds, NatWest, TSB, Virgin)
- 'Bad banks' (2 RBS, 5 HSBC) are not influential.
- 'Best bank' (3 Nationwide) is not influential



#### **Consensus view**

The normalised cumulative excess reputation index values ( $\sum (S(t)-5.5)$ ) gives an initial perception vector of sentiment with respect to banks:

p(0) = (0.128, 0.027, 0.154, 0.180, 0.031, 0.135, 0.073, 0.117, 0.141, 0.085)

Then the consensus view is:  $p(\infty) = T_{\infty} p(0) = (0.104, 0.104, 0.104, ..., 0.104)$ 

This consensus value is an effective 'smoothing' of the initial perception vector. The value 0.104 corresponds to a cumulative excess 33.0: slightly positive. So as a group, banks are slightly good!

(There is an interesting view that p(0) could be arbitrary or normally distributed from Pan (2010 and 2012))





#### Variation of Bayesian Thresholds $\lambda_H$ and $\lambda_{VH}$ Generally insensitive





### Impact

#### Super-stressed effect of sentiment on product sales.

Product	Positive sentiment (%)	Negative sentiment (%)
Sales volume	3.4	7.9
Income	1.3	2.9
Profit after tax	1.3	3.6

#### Expected values of the effect of sentiment on product sales.

Product	Positive sentiment	Negative sentiment (%)
	(70)	
Sales volume	1.6	2.3
Income	0.6	0.9
Profit after tax	0.7	0.9



The initial perception vector of sentiment with respect to banks: p(0) = (0.128, 0.027, 0.154, 0.180, 0.031, 0.135, 0.073, 0.117, 0.141, 0.085)Was calculated from the cumulative excess vector C = ( $\sum (S(t)-5.5)$ ):

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Impact
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C = (111.3, -212.3, 195.7, 47.4, -198.7, 133.2, -64.0, 76.3, 156.1, -27.9)

Let *J* be a vector whose entries are the column values of  $T_{\infty}$ . Define the total influence of the system,  $\tau$ , by the scalar product  $\tau = C.J \sim 33.1$ 

Each bank experiences an 'network drag' of value  $\tau$  over 24 months, or  $\tau/2$  annually. These are the % components of reputation attributable to the 'network' ( $\tau$  as a % of each member of *C*):

(14.9, -7.8, 8.4, 34.9, -8.3, 12.4, -25.8, 21.7, 10.6, -59.3)





