Co-evolutionary dynamics in social networks: A case study of Twitter





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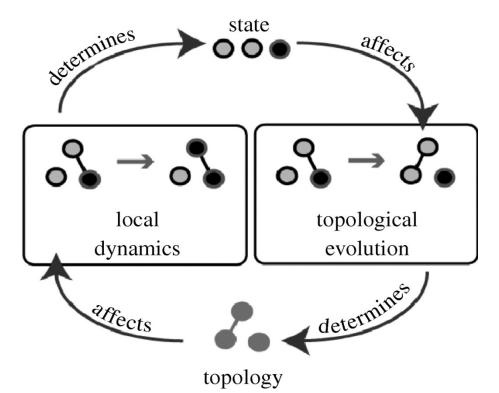
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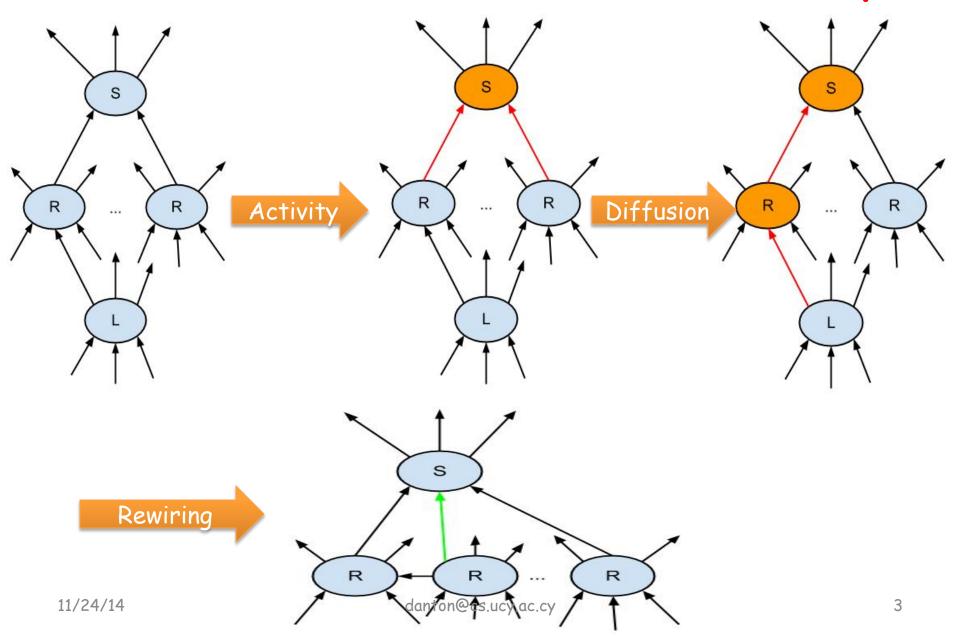
Co-evolutionary dynamics 101

• Coupled dynamics <u>ON and OF</u> networks



"Adaptive Coevolutionary Networks: A Review", Thilo Gross and Bernd Blasius, Journal of the Royal Society: Interface, 2008

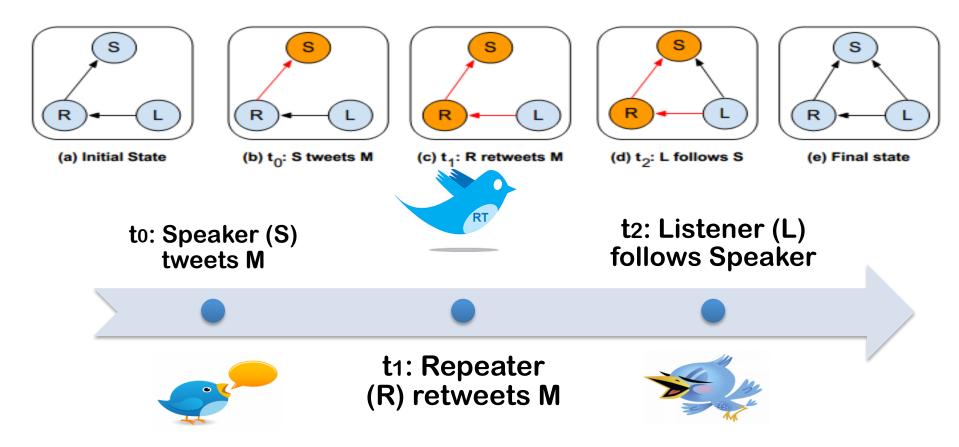
Are social nets co-evolutionary?



Main objectives of this work

- In the context of Twitter:
 - Examine presence of co-evolutionary effects
 - Quantify their likelihood and analyze them statistically
 - Develop a simple probabilistic model, based on empirical results
 - Examine how co-evo dynamics may affect a social network in long-term

Tweet-Retweet-Follow (TRF) events

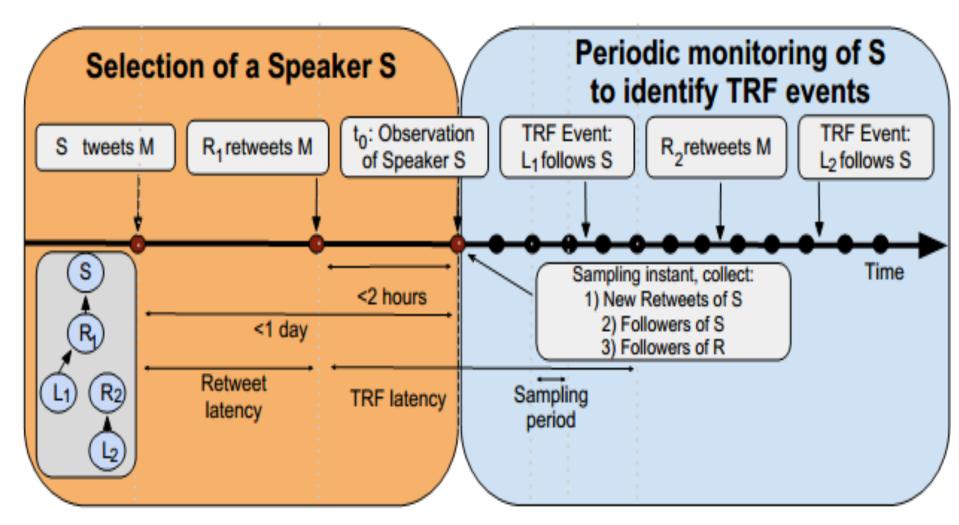


 TRF events: clear case of co-evolutionary dynamics

Definition of TRF event

- A Tweet-Retweet-Follow event
 - Speaker S,
 - Repeater R,
 - Listener L
- Occurs when:
 - a) S tweets a message M at time t_0
 - b) R retweets M at some time $t_1 > t_0$
 - c) A follower L of R follows S within Δ hours from t_1

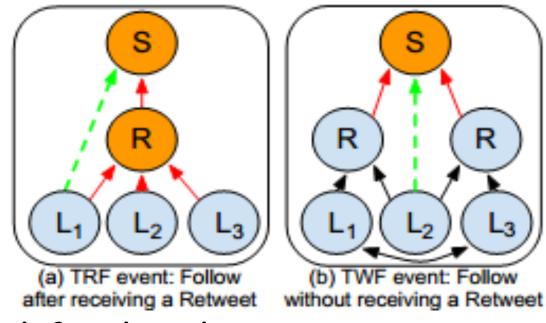
Data collection methodology



Collected data

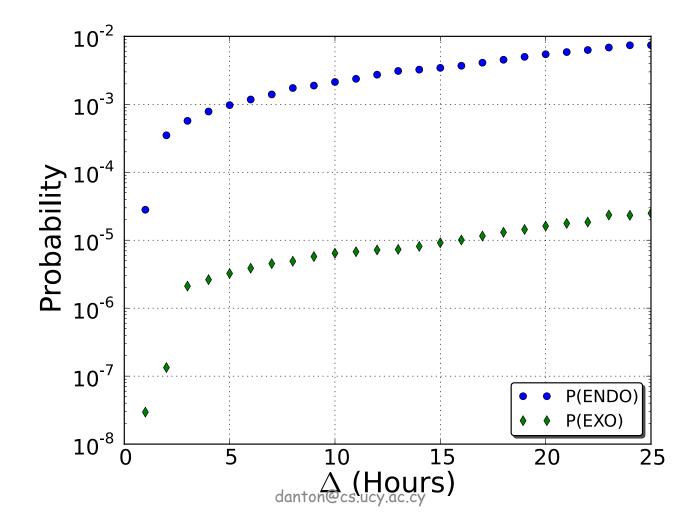
- September 19-25 2012
 - 4746 Speakers monitored
 - Posted 386,980 tweets
 - 83860 Repeaters
 - 146,867 Retweets
 - 120 milion RT events
 - 7451 TRF events (17% of observed new followers)
- Bot-filtering
 - Remove bot accounts (accounts suspended by Twitter)
 - 1% of collected accounts
 - 10% of identified TRF events

Does receiving a retweet increase probability of a new follower link? (compared to not receiving a retweet)

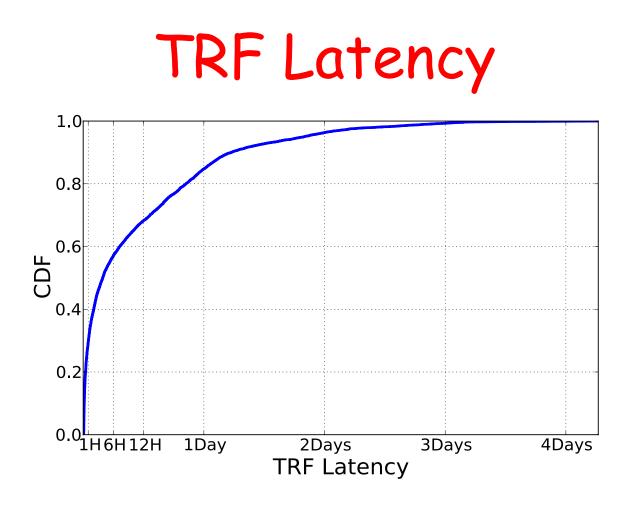


- Control for local structure
- Examine the probability for a new follower in a time window Δ

Effect of receiving (or not receiving) a retweet



11/24/14



 A new Listener follows Speaker typically within first 24-48 hours from last received retweet

Which factors affect TRF probability?

Factor	Description		
Structural Features			
F(S)	Number of followers of <i>S</i>		
F'(S)	Number of followees of <i>S</i>		
AGE(S)	Number of days since S joined Twitter		
$S \rightarrow L$	Reciprocity: whether the Speaker was		
	following the Listener at the time of the		
	TR event		
Informational Features			
ST(S)	Total number of tweets of S		
$A_{rate}(S)$	Rate of S tweets per day		
$Tweets(S,L,\Delta)$	Number of distinct tweets of S received		
	by L during period Δ		
$Retweets(S,L,\Delta)$	Number of distinct retweets of S re-		
	ceived by L during period Δ		
$Repeaters(S,L,\Delta)$	Number of Repeaters R that L received		
	tweets of S from during period Δ		

Logistic Regression

Examine which of the previous features affect TRF probability significantly:

$$ln\left(\frac{P_{TRF}}{1-P_{TRF}}\right) = \kappa_0 + \sum_{i=1}^n \kappa_i X_i$$

κ_i denotes the effect of feature X_i on the odds of TRF events

Factor	Description	Odds ratio	95% CI	
Structural Features				
F(S)	Number of followers of <i>S</i>	1.000***	[1.000, 1.000]	
F'(S)	Number of followees of <i>S</i>	0.999***	[0.999,0.999]	
AGE(S)	Number of days since S joined Twitter	0.998***	[0.998, 0.998]	
$S \rightarrow L$	Reciprocity: whether the Speaker was following the Lis-	27.344***	[25.663,29.136]	
	tener at the time of the TR event			
Informational Features				
ST(S)	Total number of tweets of S	1.000***	[1.000, 1.000]	
$A_{rate}(S)$	Rate of S tweets per day	0.989***	[0.988,0.991]	
$Tweets(S,L,\Delta)$	Number of distinct tweets of S received by L during pe-	2.010***	[1.781,2.270]	
	$\operatorname{riod}\Delta$			
Retweets (S, L, Δ)	Number of distinct retweets of S received by L during	1.603***	[1.371, 1.873]	
	period Δ			
$Repeaters(S, L, \Delta)$	Number of Repeaters R that L received tweets of S from	2.076***	[1.889, 2.282]	
	during period $\overline{\Delta}$			

- 1. Reciprocity: Speaker already follows Listener (about half of TRF events)
- 2. Number of retweets of S received by L: how many times does S appear in L's timeline?

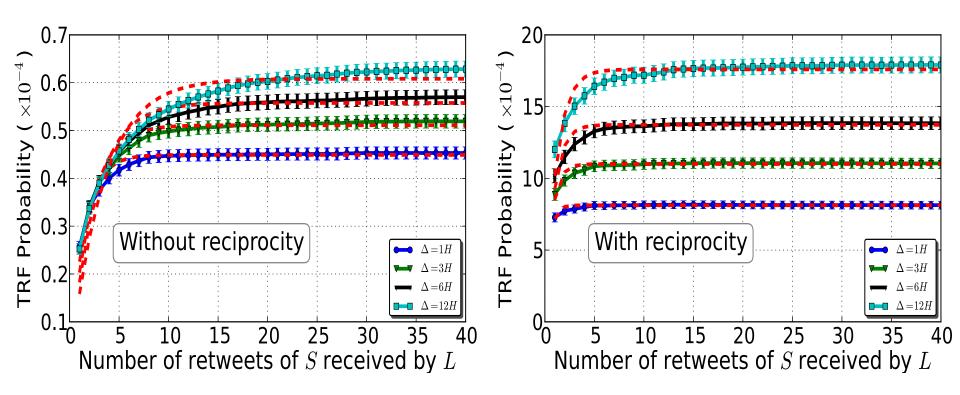
A simple model of TRF events

- Suppose each retweet leads to TRF event independently with probability q
- After receiving *n* retweets, probability of TRF = $1 (1 q)^n$
- But, Listener does not read all tweets/retweets
 "Observation" probability p

$$P_{TRF}(n) = p \times (1 - (1 - q)^n)$$

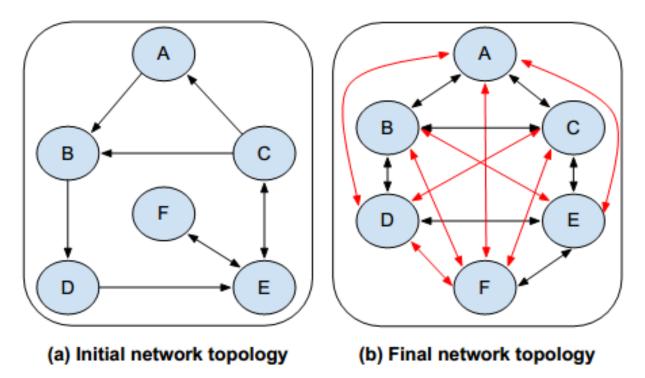
- Reciprocity increases product p×q by a factor of 100
- Time window Δ affects mostly probability p
 - With reciprocity, $p \approx 25 \times 10^{-4}$ and $p \times q \approx 10^{-3}$ ($\Delta = 24$ hours)

TRF model evaluation



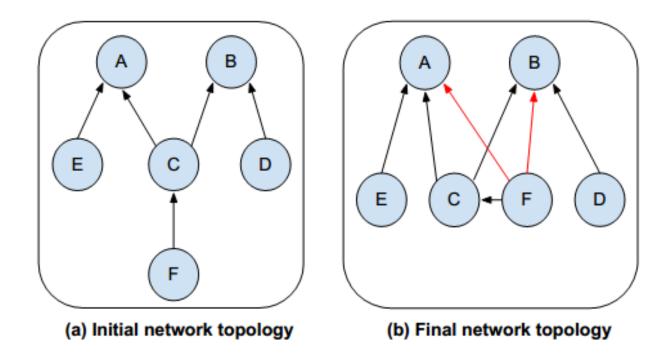
What is the effect of TRF events in the long-term evolution of a social (sub-)network?

Does sub-network form a Strongly-Connected Component?



- It will evolve to fully connected network
- TRF events create cliques (strong communities)

Does sub-network have hierarchical structure (no directed cycles)?

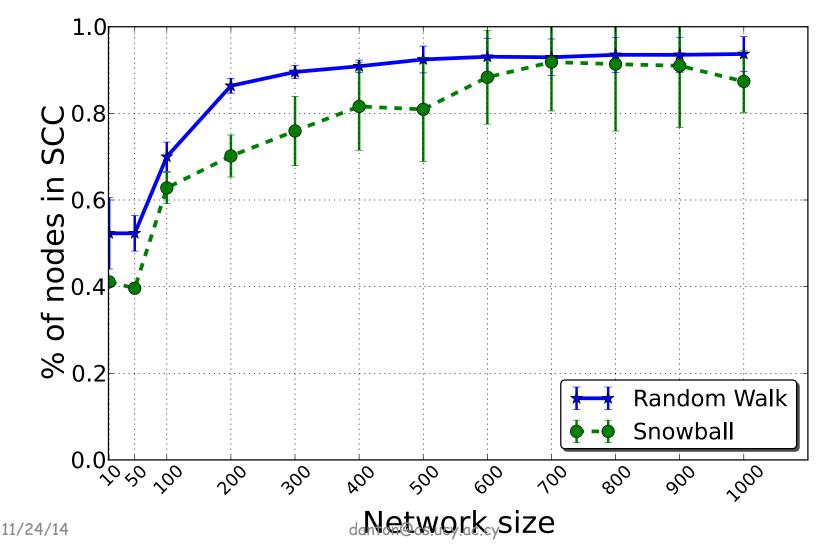


- Network evolves to a two-level hierarchy
- In each "sphere of influence", an influencer is directly connected to her followers

How common are directed cycles in connected sub-graphs of the Twitter topology?

- Analyzed an older measured Twitter topology (41.7M nodes)
 - Sampling using "forest-fire" and "snowball" methods
 - Each sampled sub-network is weakly connected
 - Samples of different sizes
- Use Tarjan's algorithm to identify longest cycle (largest SCC) in sampled sub-network

For sub-graphs with more than 500 nodes, about 90% of nodes belong in SCC component



Conclusions

- Observed co-evolutionary dynamics on Twitter
 - Tweet-Retweet-Follow events
- TRF events are responsible for 20% of the new edges in Twitter's network

 80% occur within 1 day after the retweet
- Proposed a probabilistic model for TRF events (simple enough for analytical studies?)
- TRF events tend to transform Twitter sub-nets to cliques

