# The Anatomy of a Scientific Rumor

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## The announcement

- On 4th July 2012 the ATLAS and CMS collaborations announced the results of the discovery of a new particle with the features of the Higgs boson.
- The world discussed the news and updates through traditional media and online social media. This represents an interesting large-scale phenomenon that can be analyzed in time and space.



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## Why interesting?



# Why interesting? **Understanding human behavior at large-scale.**



## Again, why interesting?

- If you know how a piece of information spreads on a network, maybe you can study how to **control its diffusion**, range, velocity, etc.
  - Marketing. You want to maximize the diffusion of an advertisement.
  - Disaster response (disease outbreaks, earthquakes, terrorist attacks, ...).
    - In some scenarios, you might actually want to demote the info...
- On the other hand, you can **detect if something is happening** and what (trending topics), and perhaps take appropriate action.
  - Systems architecture. Systems could adapt to the context.



## Twitter in a nutshell

- A user can post 140-chars messages (called *tweets*). A message can contain one or more *#hashtags* and @*mentions*, it can be a
  @*reply*. A user can *retweet* messages that they find interesting.
- A user can *follow* other users. Their messages show up in the *home timeline*, in descending chronological order.





## The data

- We crawled from Twitter all the messages sent between July 1-7 with at least one the keywords: 1hc, cern, higgs.
  985,590 tweets in total.
- We also crawled the social network of users 456,631 users connected by 14,855,875 edges.
- Locations were geocoded from users' profile info.









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@ColinEberhardtPossibly the biggest scientific discovery of our time, the #Higgs Boson,7141 retweetsannounced in glorious MS Comic Sans Font <a href="http://t.co/jDxLkqNx">http://t.co/jDxLkqNx</a>

@CERN#CMS: "we have observed a new boson with a mass of  $125.3 \pm 0.6$  GeV at4504 retweets4.9 sigma significance." Thun- derous applause. #Higgs #ICHEP2012

@CERNCERN Press Release: CERN experiments observe particle consistent with2507 retweetslong-sought #Higgs boson <a href="http://t.co/MBjIwytL">http://t.co/MBjIwytL</a> #ICHEP2012

@timscottAt the end of the #higgs announcement, one of the CERN team will pause,2062 retweetsnonchalantly say "oh, one more thing", then calmly teleport away.

**@ProfBrianCox**Not only have taxpayers spent more on banks than we've spent on science2033 retweetssince Jesus. Bob Diamond is keeping the Higgs out of the headlines!









Before: July 1-3 During: July 4 After: July 5-7





Intertime and interspace distributions Temporal delay and spatial distance between two consecutive tweets between *any* user in the network.





Intertimes,  $\tau$  [sec]

Intertimes,  $\tau$  [sec]

Intertimes,  $\tau$  [sec]

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Intertimes, interspaces

Temporal delay and spatial distance between two consecutive tweets between *any* user in the network.





Retweets by day

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 $log(\tau) = N(\mu = 5.627 \pm 0.008, \sigma = 1.742 \pm 0.006)$ 

User activity intertimes

Temporal delay between two consecutive tweets between *the same* user in the network.





Degree, k

Degree distribution

UNIVERSITY<sup>OF</sup> BIRMINGHAM The distributions have been shifted along the y axis to show their structure. Dashed lines are shown for guidance only.



Replies, RTs intertimes

UNIVERSITY<sup>OF</sup> BIRMINGHAM Power-law scaling relationships with exponent < 1 occur in nature only with cut-off. Even in those cases, they are very rare.

## Rumor spreading

- We investigate how the "Higgs boson rumor" spreads in the social network.
- Different from disease epidemics, we cannot use the unstructured SI model.
- We distinguish between two different states for users in the network. A user is either *active* or *non-active*.
- A(t), D(t) represent the number of active and non-active users, respectively.
- a(t), d(t) represent the fraction of active and non-active users, respectively.

$$A(t) + D(t) = N$$



- As a first step, we **ignore social ties**.
- Once a user has tweeted, he/she is considered active. Hence, the number of active users is monotonically non-decreasing.
- We model the number of new activations as proportional to the number of users who haven't been activated yet. More formally:

$$A(t + \Delta t) = A(t) + \lambda [N - A(t)]\Delta t$$

$$\frac{da(t)}{dt} = \lambda[1 - a(t)]$$

constant activation rate

$$a(t) = 1 - [1 - a(t_k)]e^{-\lambda(t - t_k)}$$

starting time of period k





Model w/out social ties, W/out de-activation Red dots represent real data, black lines represent model data.

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- We now build a model that takes into consideration **social cascading**.
- In this refined model, a user that tweets about the Higgs boson is active, then can become non-active after a certain amount of time.
- Activation: an active user can make each of their followers active with probability rate  $\lambda(t)$ .
- **Deactivation**: an active user can become non-active with probability rate  $\beta(t)$ . This models the time-limited visibility of tweets (newer tweets replace old ones, users read only most recent tweets). Non-active users cannot activate their followers.







• At time t, the probability that a non-active user, connected to  $j_a$  active users, becomes active (i.e. is activated by at least one of them) is:

$$p_{\lambda}(t;j_A) = 1 - [1 - \lambda(t)]^{j_A}$$

- In general, the probability that a non-active user is connected to a certain number of active users depends on degree correlations, i.e., the probability of observing a vertex with out-going degree k\_out connected to a vertex with in-going degree k\_in.
- However, it has been shown [Boguná et al., PRL 2003] that for *pure* scalefree networks with exponent between 2 and 3 degree correlations do not affect the spreading dynamics. We use this simplifying assumption for this network, exhibiting scale-free degree of 2.5 for k>200.



• At time *t*, the probability that a non-active user *with a certain in-degree* is connected to j<sub>A</sub> active users (with *any* out-going degree) is:

$$\tilde{p}(t; j_A, k^{in}) = \frac{\binom{A(t)}{j_A} \binom{N - A(t) - 1}{k^{in} - j_A}}{\binom{N - 1}{k^{in}}},$$

which accounts for all the ways you can arrange  $j_A$  users in  $k^{in}$  places.

• Hence, the probability of activation of a user with a certain in-degree is:

$$P_{\lambda,k^{in}}(D \to A) = \sum_{j_A=1}^{k^{in}} \tilde{p}(t; j_A, k^{in}) p_{\lambda}(t; j_A).$$



• By summing this probability over the in-degree distribution, we obtain the probability of activation at time t.

$$\Theta_{\lambda}(t) = \sum_{k^{in}} \mathcal{P}(k^{in}) P_{\lambda, k^{in}}(D \to A)$$

• Finally, we use this to derive a discrete activation model:

$$A(t+1) = (1 - \beta(t))A(t) + (N - A(t))\Theta_{\lambda(t)}(t),$$



• In order to account for the decreasing interest on a topic over time, we model the activation rate as time-varying, with an exponential decay with time scale  $\tau=1/\xi$ 

$$\lambda(t+1) = (1-\xi)\lambda(t)$$

• Finally, we perform large-scale Monte-Carlo simulations over the parameters, to fit our model with the observed data.





Time [hours]

#### Real data and model



Red dots represent real data, solid lines represent model data. Parameters were chosen so that chi-squared is minimized.

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## Thanks! Questions?

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