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# **Controlling Information Flow in Online Social Networks**

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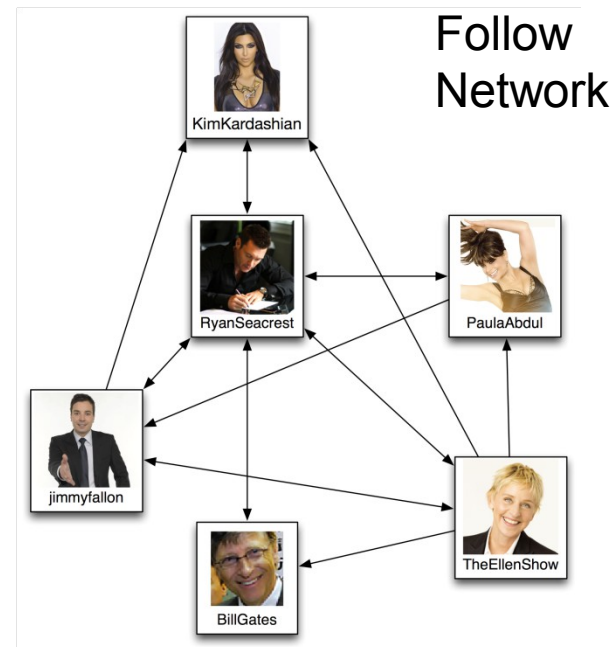
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# Introduction



## ■ Follow Links in Twitter

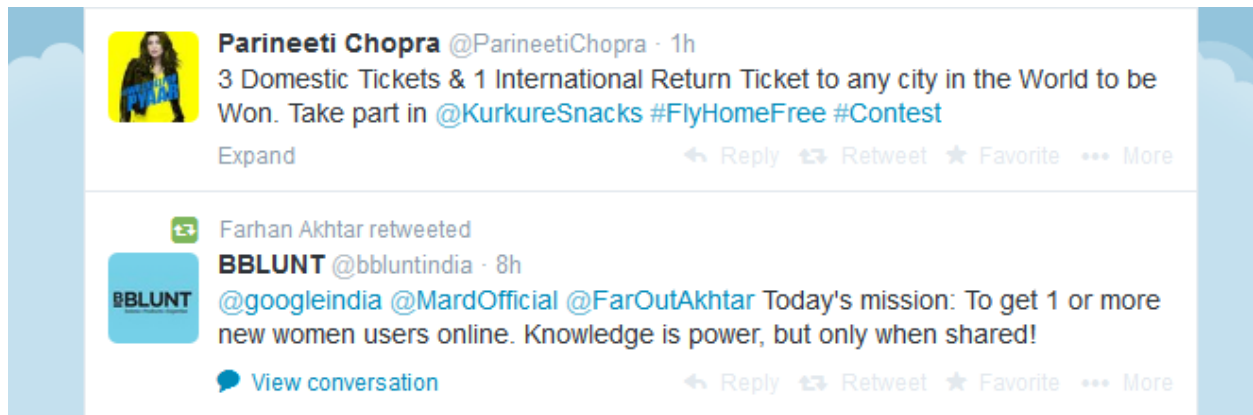
- A user can be followed by any number of users
- All tweets by the user are shown in the timeline of her followers.



# Introduction (Continued)

## ■ Mention Links in Twitter

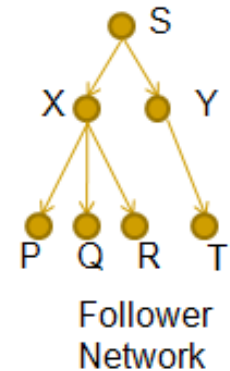
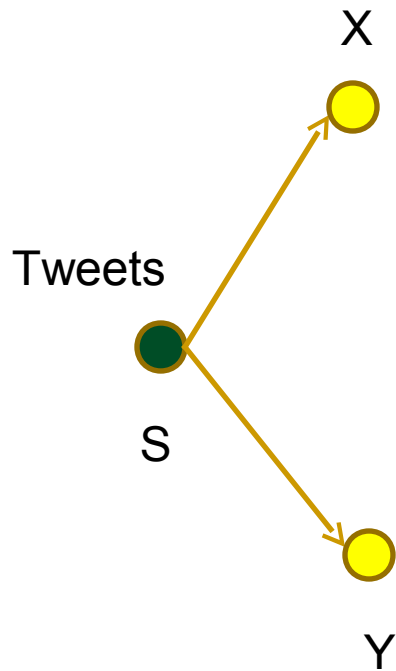
- A mention is any Twitter update that contains "@username" anywhere in the body of the Tweet.
- Non-Followers can also be mentioned



# Introduction (Continued)

## Information Diffusion via Follow Links

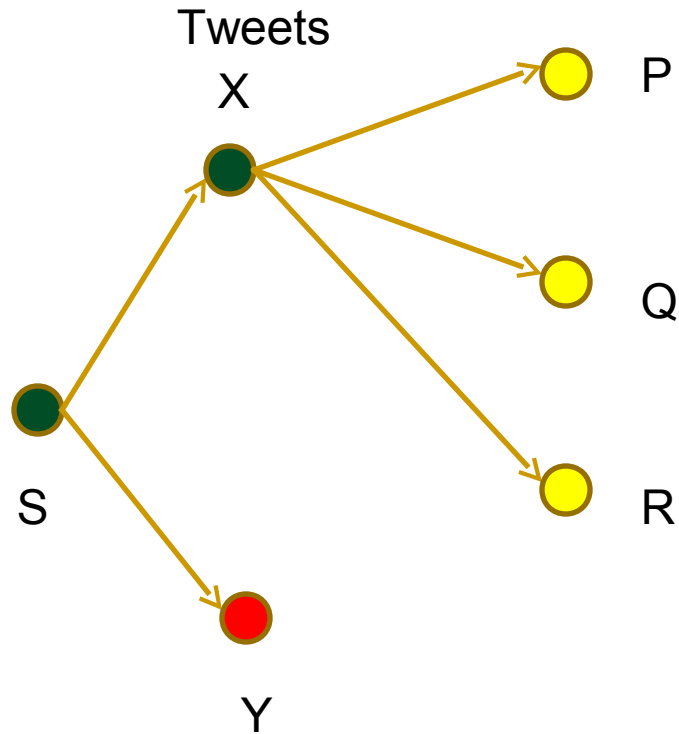
S has two followers X & Y



# Introduction (Continued)

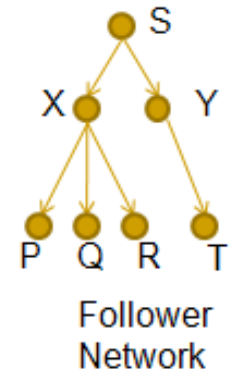
## Information Diffusion via Follow Links

X has 3 followers- P, Q & R



X decides to re-tweet

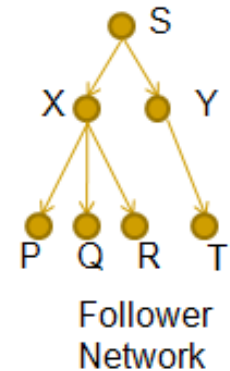
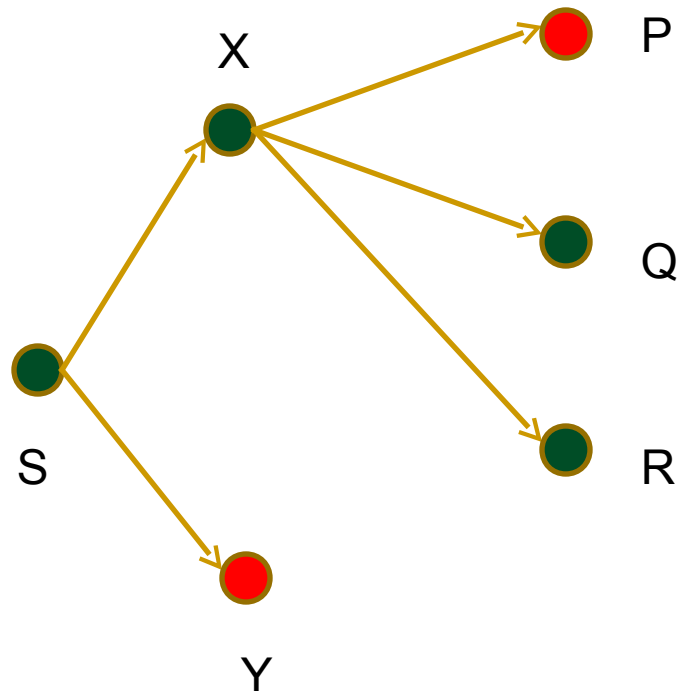
Y decides not to re-tweet



# Introduction (Continued)

## Information Diffusion via Follow Links

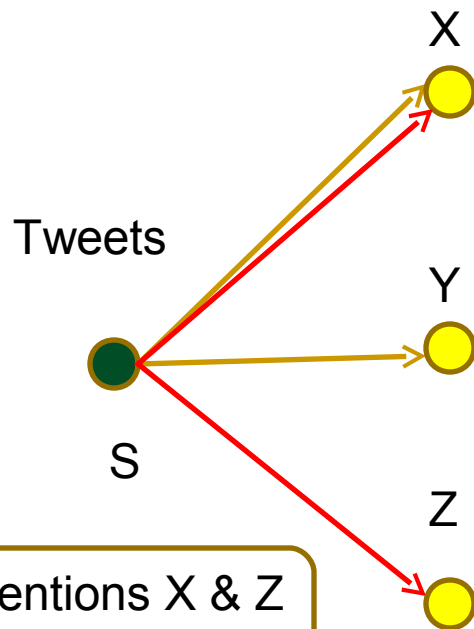
P decides not to re-tweet  
Q & R decide to re-tweet



And in this way the information propagates....

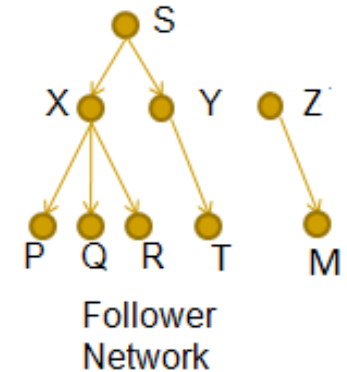
# Introduction (Continued)

## Information Diffusion via Follow & Mention Links



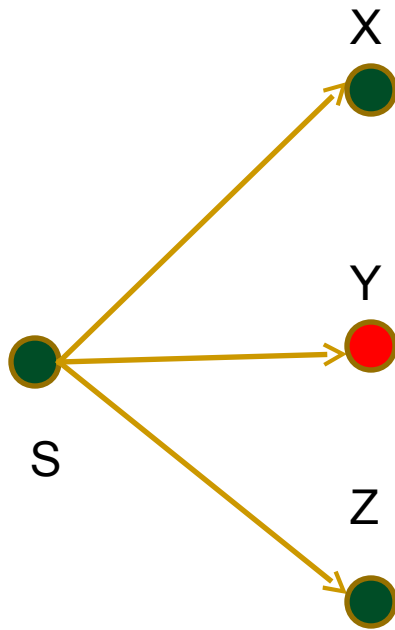
S mentions X & Z  
S: ...@X @Z

S has two followers X & Y

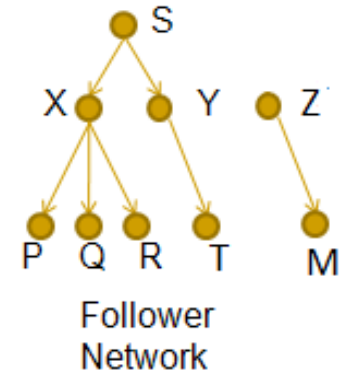


# Introduction (Continued)

## Information Diffusion via Follow & Mention Links



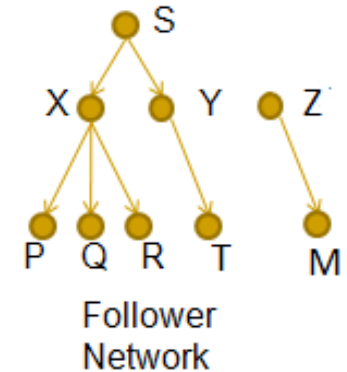
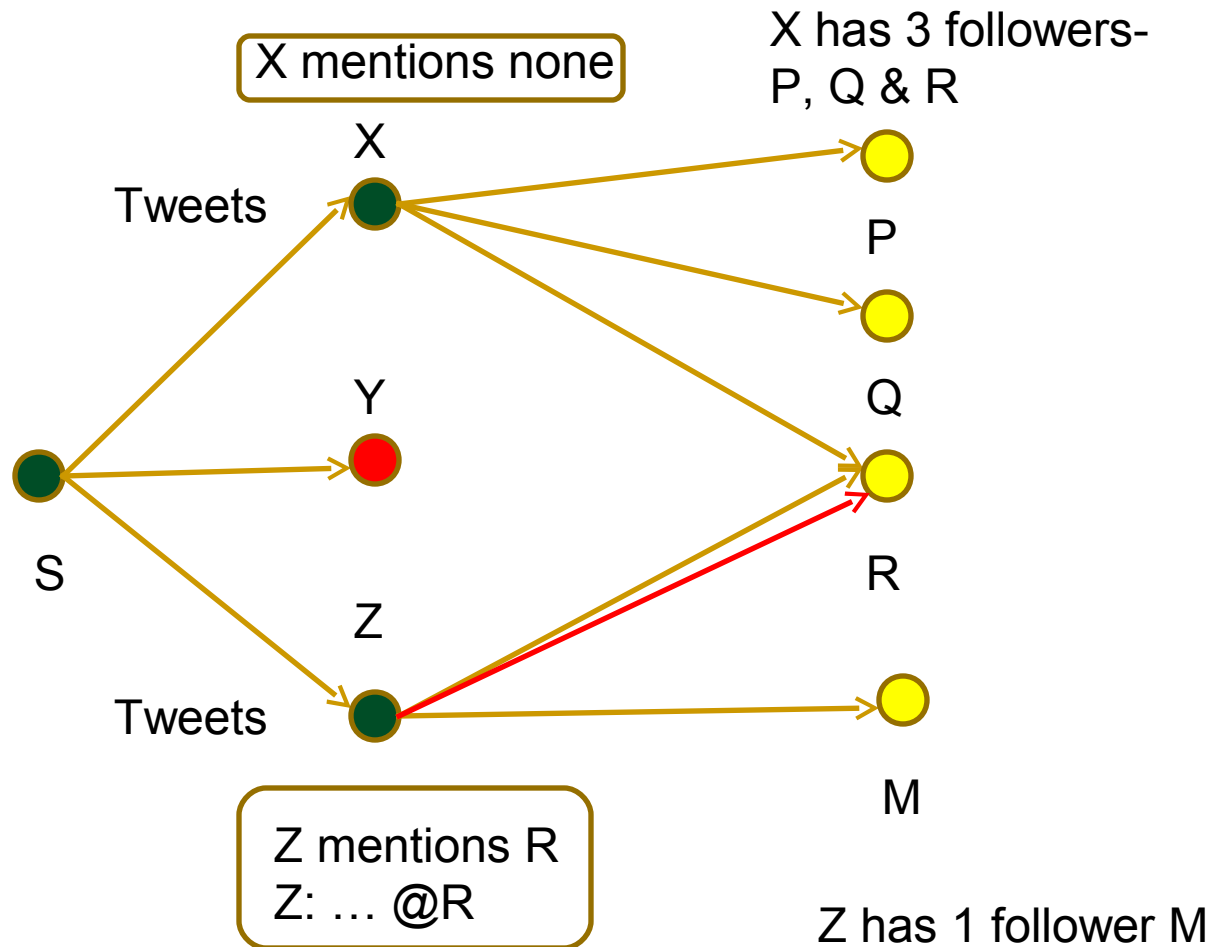
X, Z decide to re-tweet  
Y decides not to





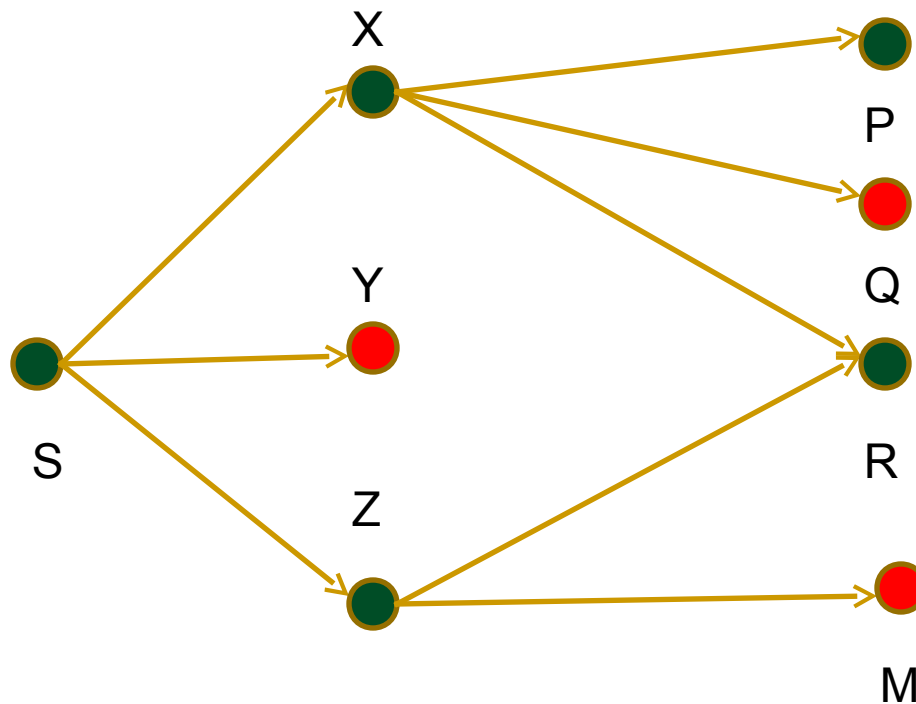
# Introduction (Continued)

## Information Diffusion via Follow & Mention Links

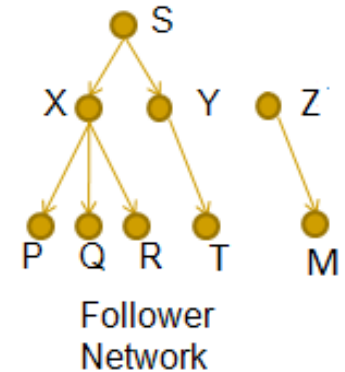


# Introduction (Continued)

## Information Diffusion via Follow & Mention Links



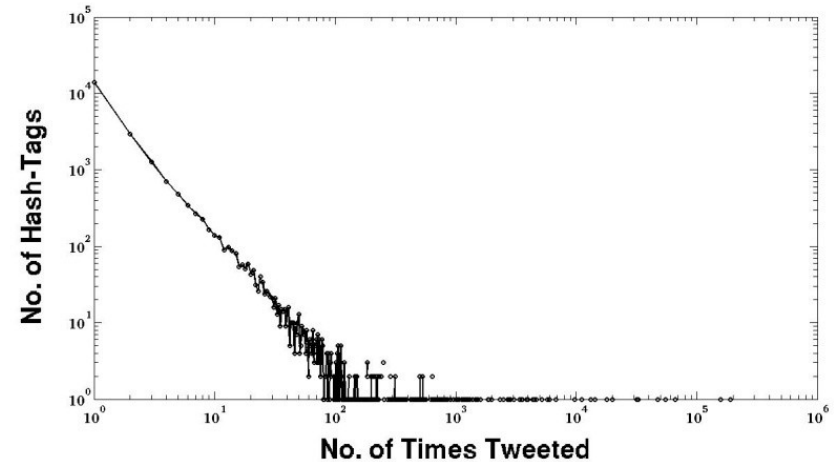
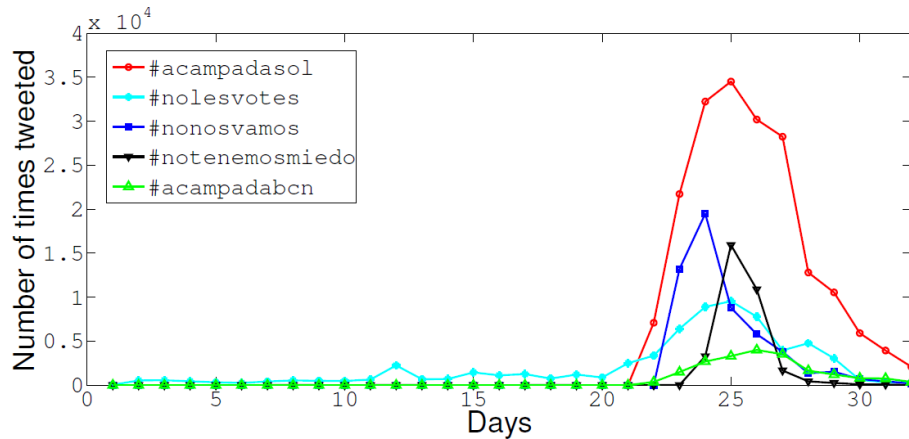
P, R decide to propagate  
Q, M decide not to propagate



And in this way the information propagates...

# Introduction (Continued)

## Hashtag Popularity (Number of users tweeted)



(15m Dataset)

### Observation:

1. Different hashtags have different temporal pattern of popularity
2. Few hash-tags are highly popular , but most are not

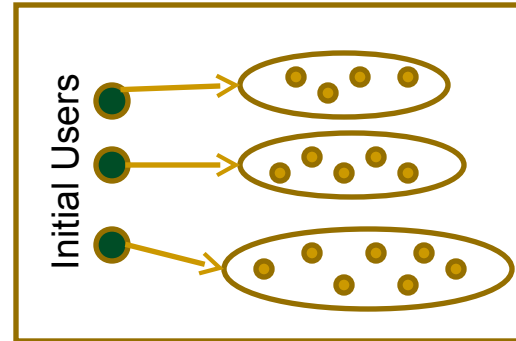
### Research Question:

Investigate the Key-factors controlling the popularity of a hashtag

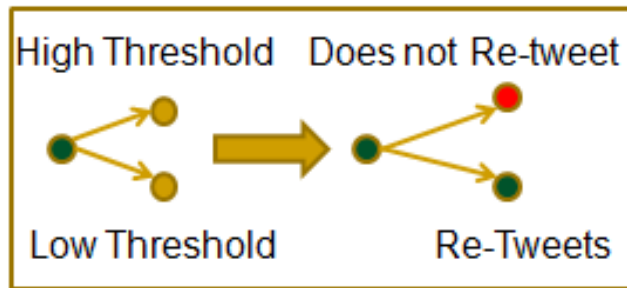
# Introduction (Continued)

## Factors Influencing Popularity of a hashtag

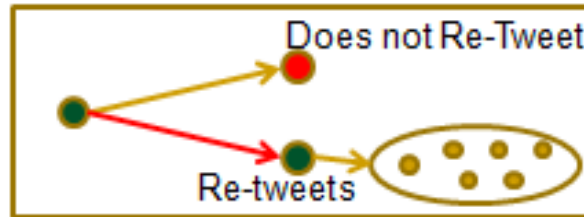
- Number of Initial Users (gets information from external sources)



- Passivity-Threshold of Users

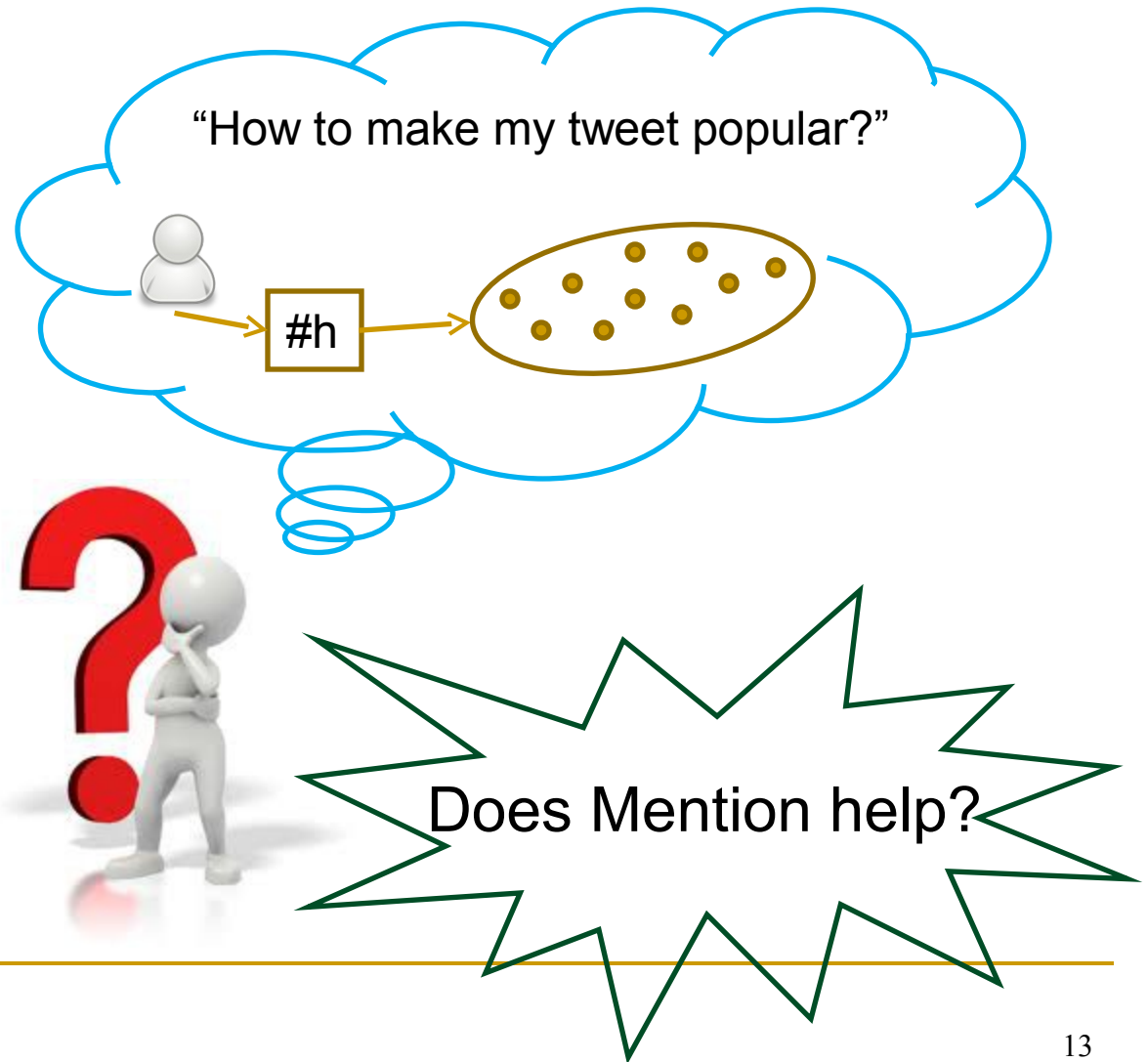


- Mention Usage



# Problem Statement

How to increase the popularity of a Hashtag?



# Problem Statement (Continued)

Available Options:

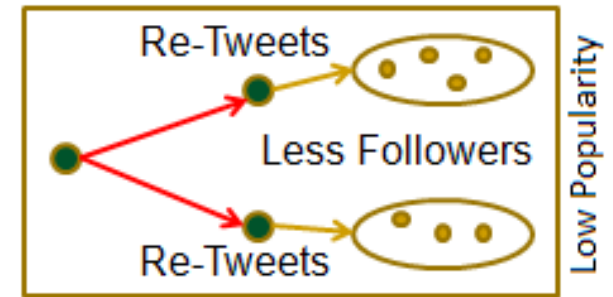
## ■ Mention Friends:

□ Pro:

- High probability of re-tweet

□ Con:

- Low popularity if friends are not popular (less number of followers).



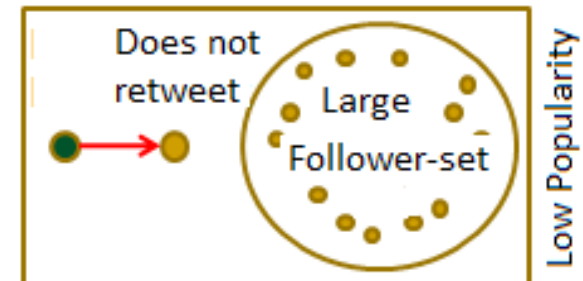
## ■ Mention Celebrities

□ Pro:

- High popularity if they re-tweet

□ Con:

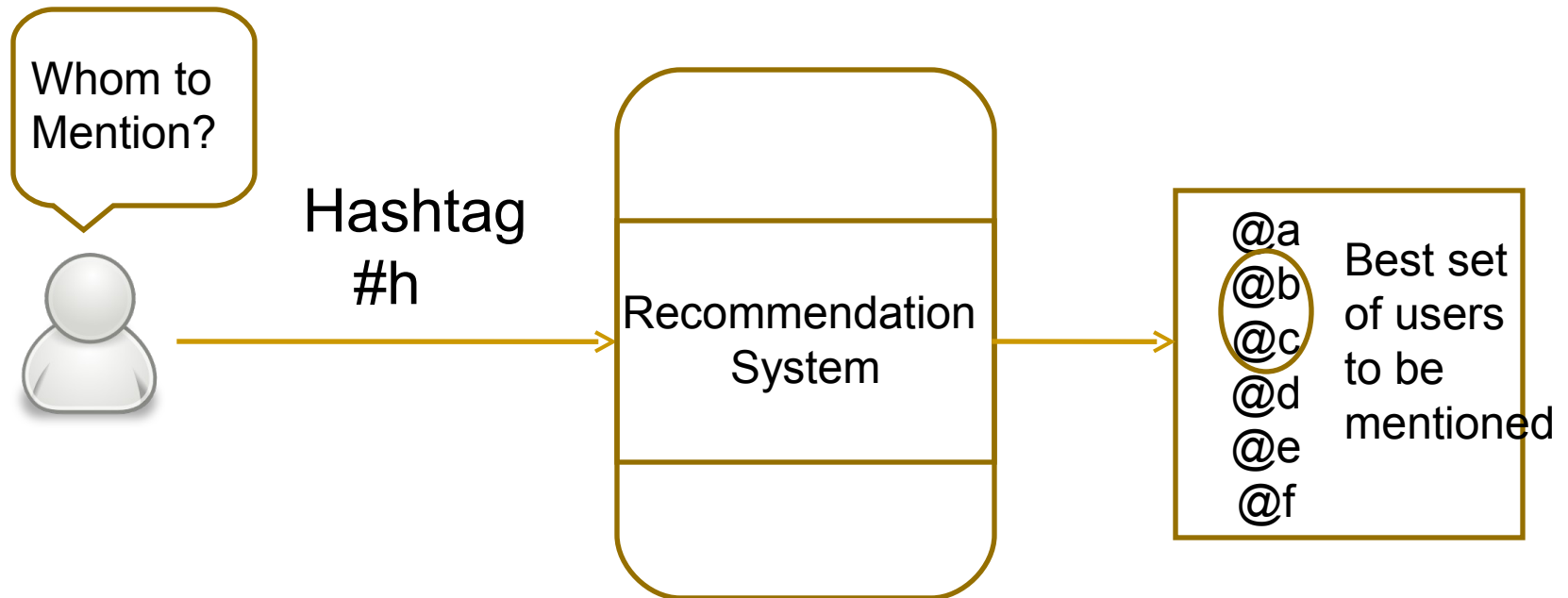
- Low probability of re-tweet (high number of followers)



Need to maintain a **BALANCE**

# Objective

## Recommendation System



**Parineeti Chopra** @ParineetiChopra · 1h

3 Domestic Tickets & 1 International Return Ticket to any city in the World to be Won. Take part in @KurkureSnacks #FlyHomeFree #Contest

Expand

Reply Retweet Favorite More

# Outline

- Data-study & Measurements
  - Dataset
  - Representation
  - Dependency on Mention
- Model for hashtag Propagation
  - Description
  - Set Parameters from Dataset
  - Validation
  - Insights
- Conclusion



# Data-study & Measurements

## Dataset

- 15m Dataset → Contains information about tweets posted during the revolutionary movements in Spain during May 2011



- Data:
  - user\_id ; timestamp ; hashtag\_list ; mention\_list
  - Follow Links
- Statistics:
  - Total users: 87569
  - Total Tweets: 529393
  - Hash-Tags: 22376

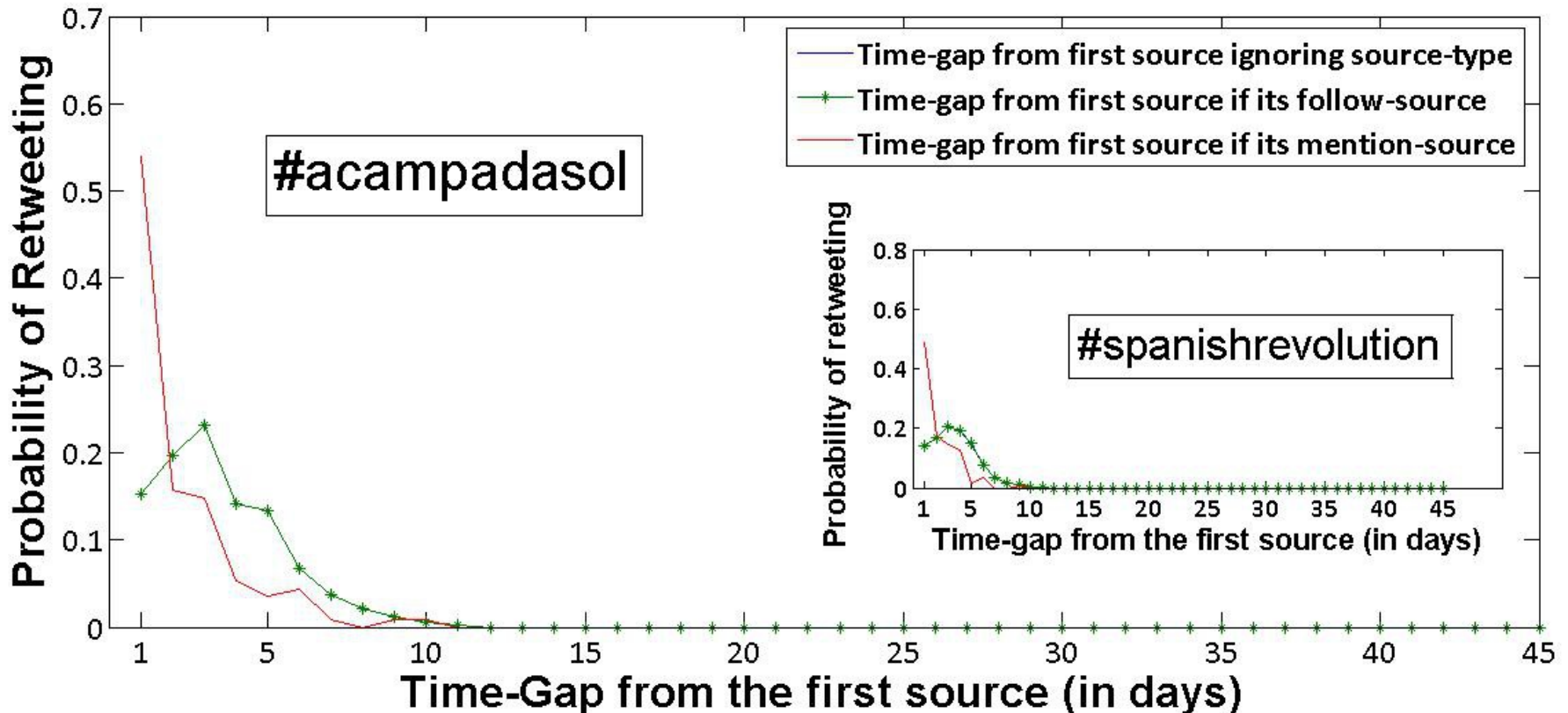
**Filtered** relationships only to those who sent at least a message in that **topic**; or were mentioned by someone who did.

Source: <http://cosnet.bifi.es/research-lines/online-social-systems/15m-dataset>

Y. Moreno et al. The dynamics of protest recruitment through an online network. Scientific reports, 1, 2011

Other datasets: Arab-spring datasets

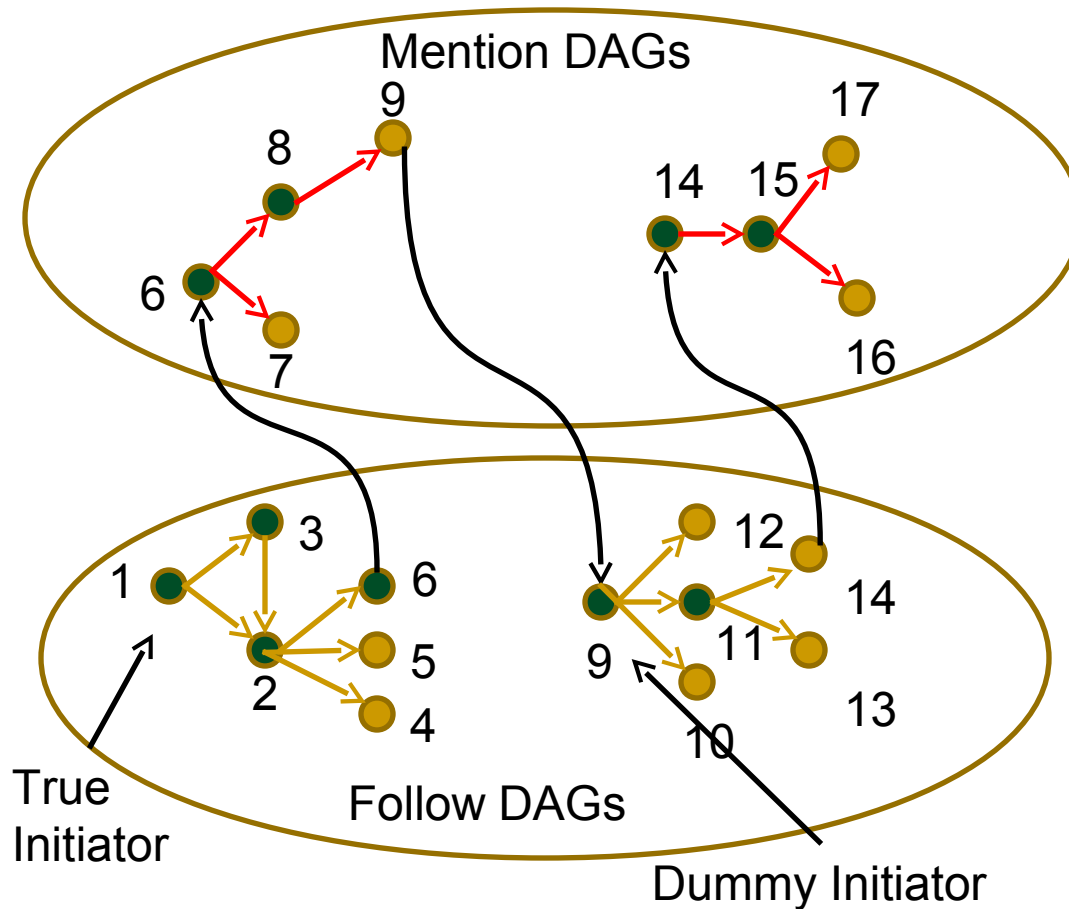
# Data-study & Measurements



Retweet comes faster if the source is “Mention”

# Data-study & Measurements

## Representation



### True Initiators

Those who get the information from external Source and spread in the network  
e.g. 1

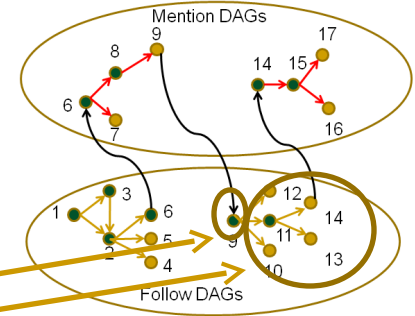
### Dummy Initiators

Those who get the Information only through Mention links and spread in the network  
e.g. 9

# Data-study & Measurement

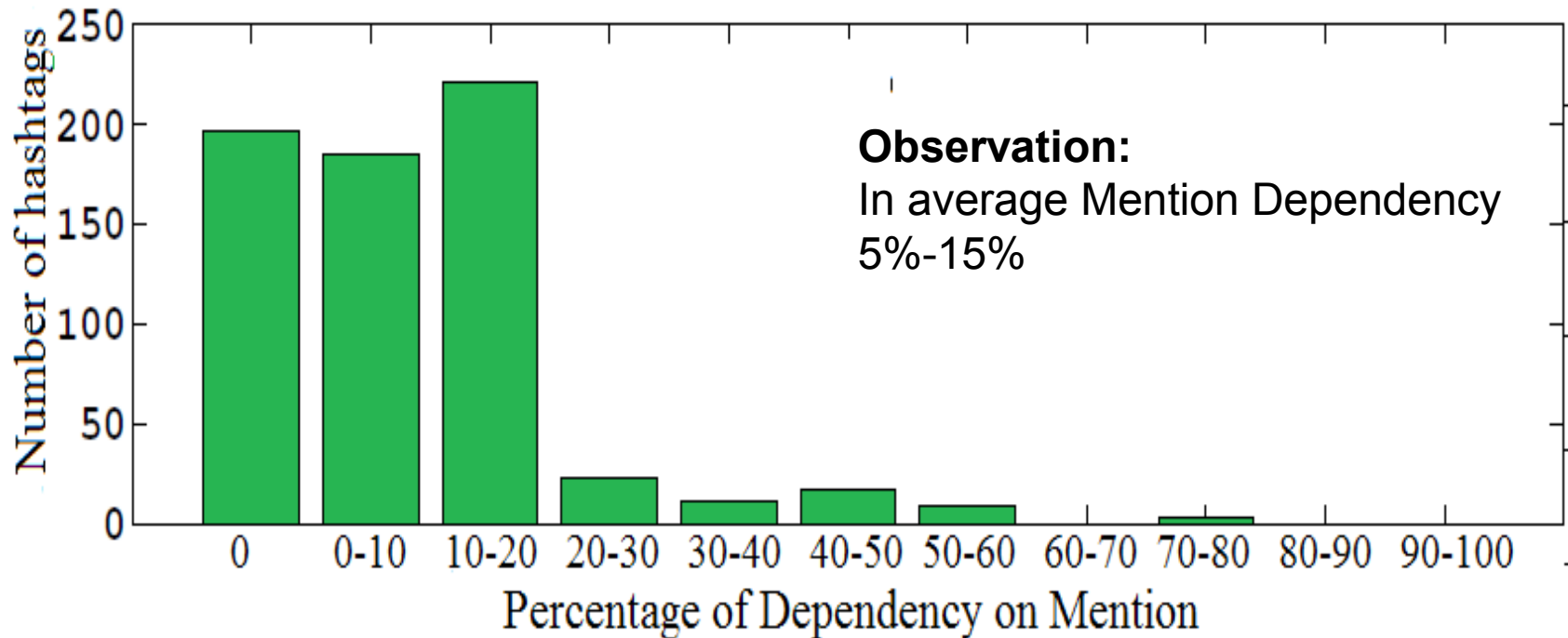
## Dependency on Mention

- For each hashtag '#h',
  - Set of dummy initiators (set A)
  - Set of users who belong to only the DAGs rooted by dummy initiators (set B)
  - Set of users who have tweeted #h (set C)
- The users in sets A & B would not have got the information ('#h') without "Mention"
- Dependency of '#h' on Mention =  
Fraction of tweeting users who would not have received the hashtag without Mention  
$$= (|A|+|B|)/|C|$$



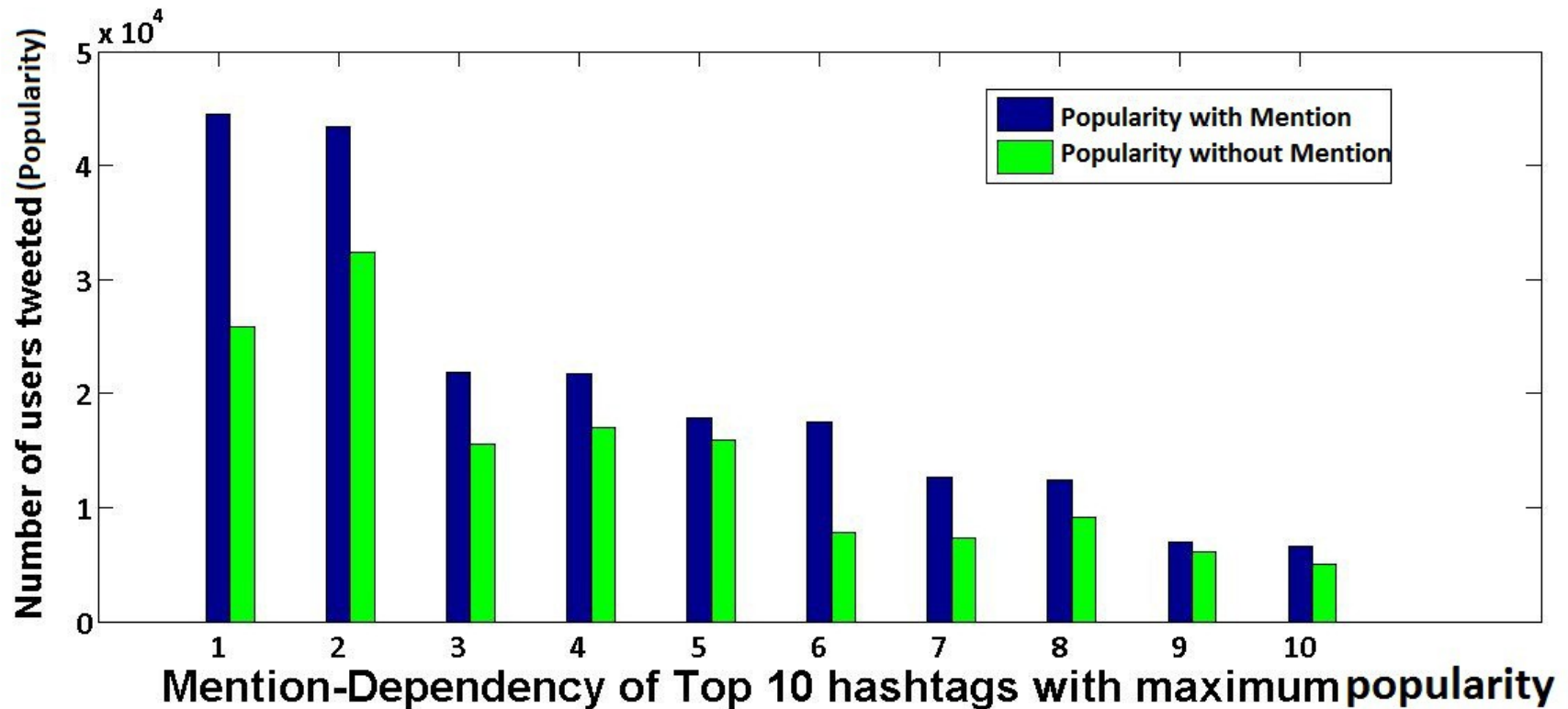
# Data-study & Measurements

## Dependency on Mention



# Data-study & Measurements

## Dependency on Mention

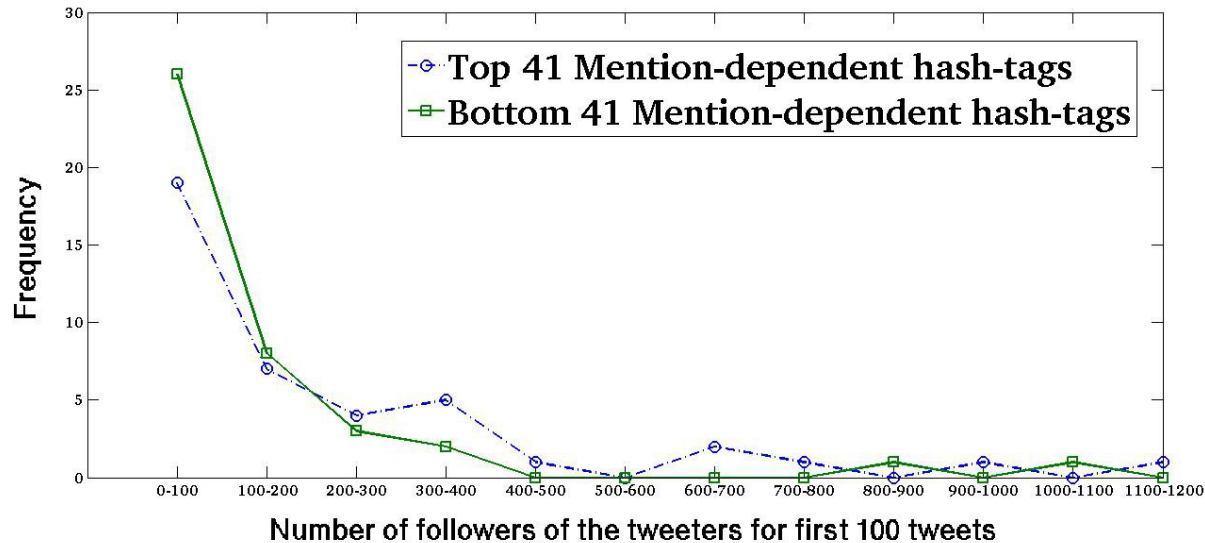


# Data-study & Measurements

## Group Hashtags based on Dependency on Mention

- Grouped the hash-tags based on their dependency on Mention
  - Group 1: Consists of top 41 Mention-dependent hashtags  
Average popularity: 8092.7  
Average number of mentioned users: 484.83
  - Group 2: Consists of bottom 41 Mention-dependent hashtags  
Average popularity: 82.21  
Average number of mentioned users: 4.48
- Observation: Highly popular hashtags are generally more dependent on “Mention”

# What did the tweeters of Group 1 hashtags do differently from the tweeters of Group 2 hashtags in the first 100 tweets?

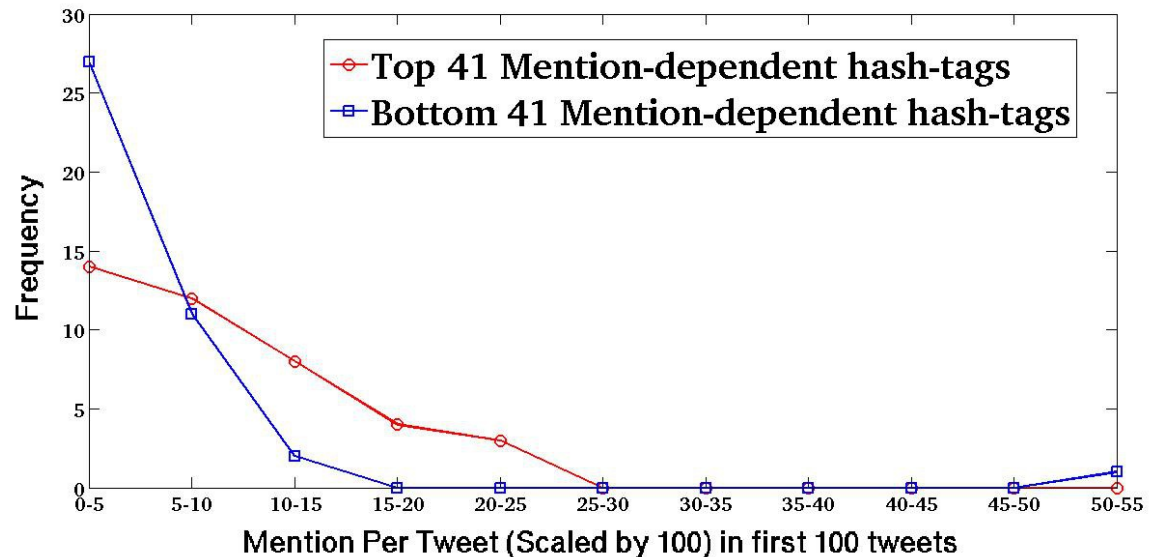


## Observation:

Almost similar popularity of initial users

## Observation:

Group 1 hashtag tweeters used mention at almost double of the rate of Group 2 tweeters



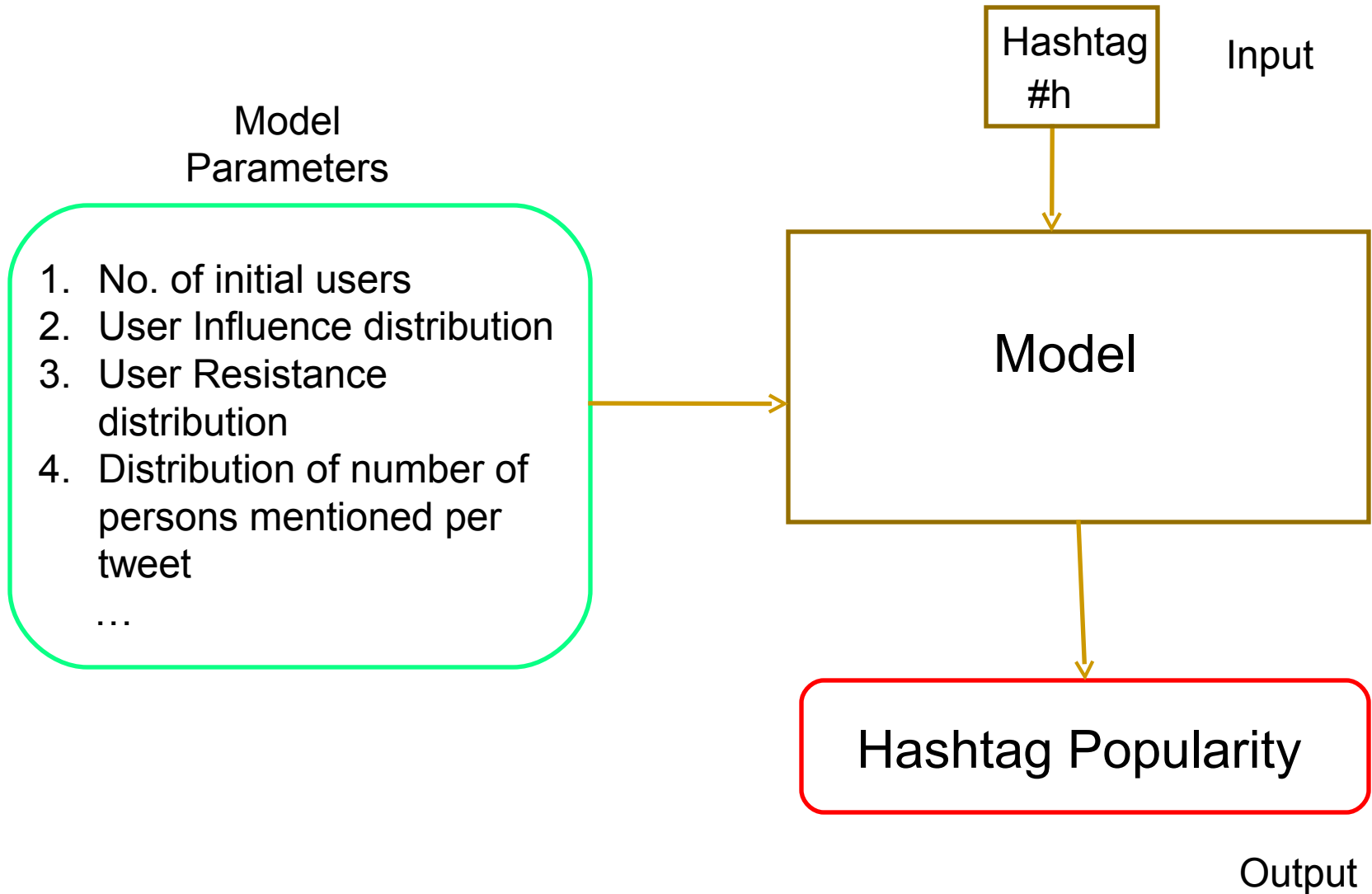


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# Outline

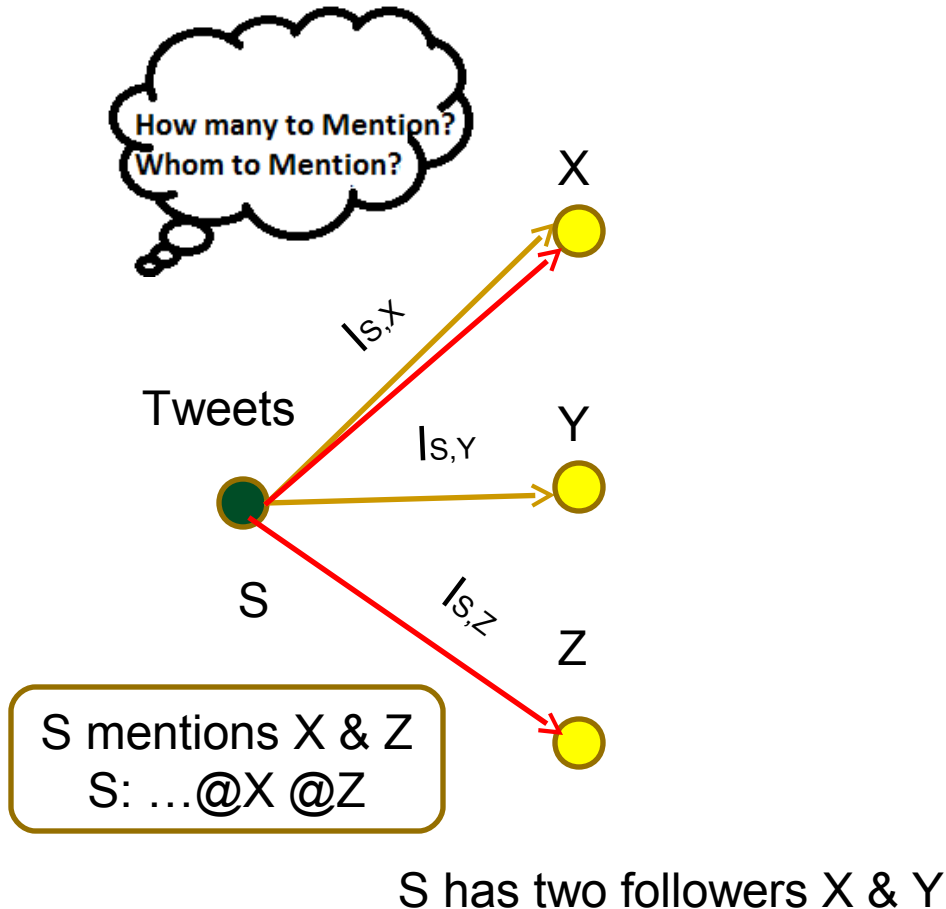
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# Model for Hashtag Propagation



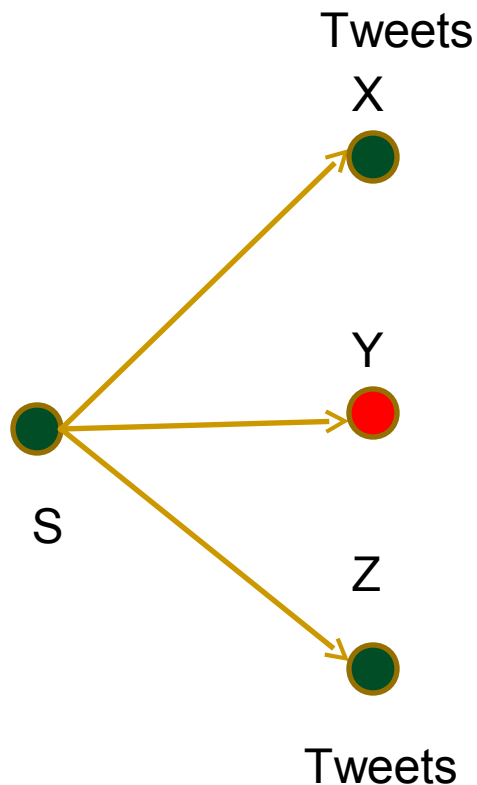
# Model for Hashtag Propagation

## Intuition Behind the Model



# Model for Hashtag Propagation

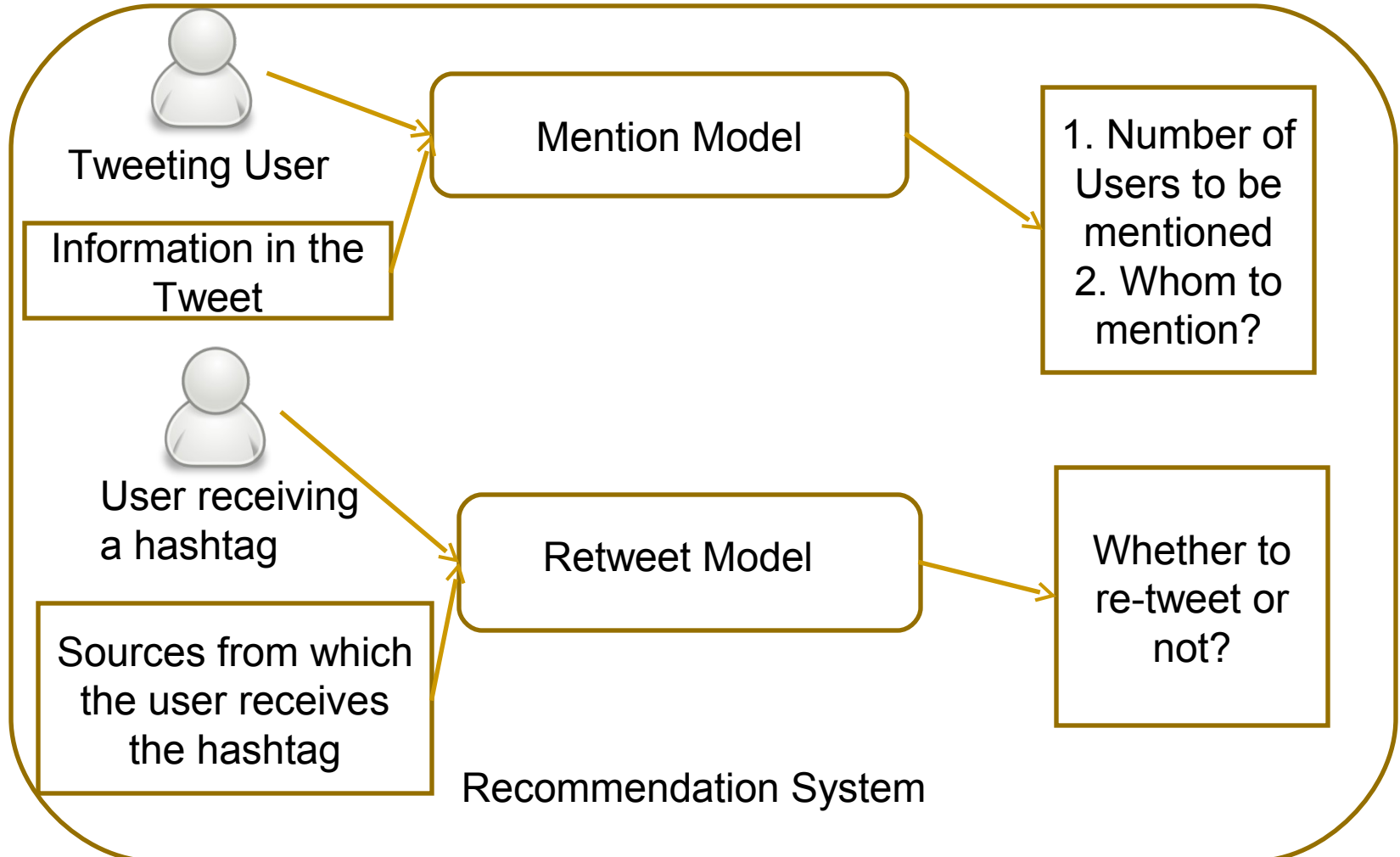
## Intuition Behind the Model



X, Z decide to propagate  
Y decides not to propagate

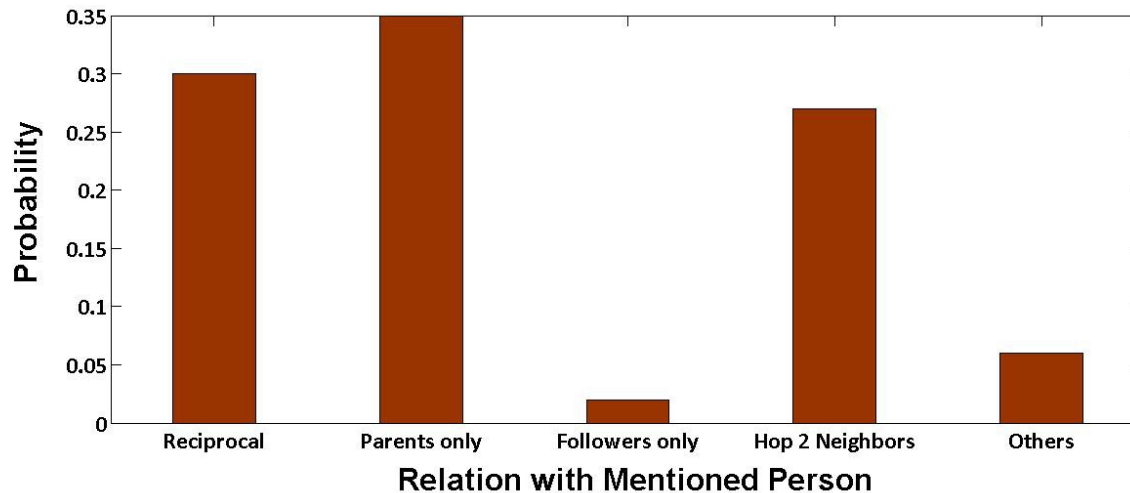
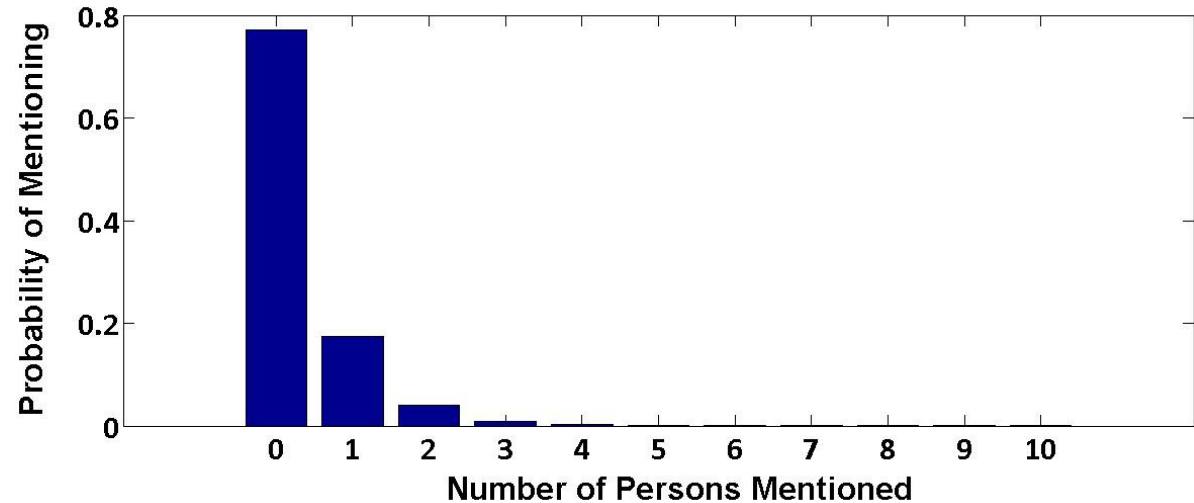
# Model for Hashtag Propagation

## Model Components



# Mention Model

- Distributions calculated from the Dataset



# Retweet Model (Linear Threshold Based)

- We calculate the weightage of each link a user (say, u1) gets
  - Factors:

- Influence of the user (say, u2) from whom the link is coming (Calculated using PageRank)
- Importance of the link based on the
  - Type of Link (Follow/Mention/Mixed)
  - Social Tie between u1 & u2 (Reciprocity) (calculated from dataset)
- Time-gap between u2's tweet and the current time

- We also calculate the passivity/resistance of each user

$$1 - \frac{\text{Number of times tweeted}}{\text{Number of times received any hashtag}}$$

# Retweet Model (M/C Learning Based)

- For each Retweet,
  - Collect the following features of each of the sources
    - Influence using PageRank
    - Number of followers
    - Time-Gap
    - Type of Link (Follow/Mention)
    - Relation with retweeting user (follower/parent/reciprocal)
  - Combine the features by taking “Average” and “Standard Deviation”
  - Collect the following features of the retweeting user
    - Influence using PageRank
    - Number of followers
    - Number of times tweeted
    - Number of times tweeted the current hashtag
    - Number of times got exposed to the current hashtag
- Total 16 features per retweet (including “Number of sources”)



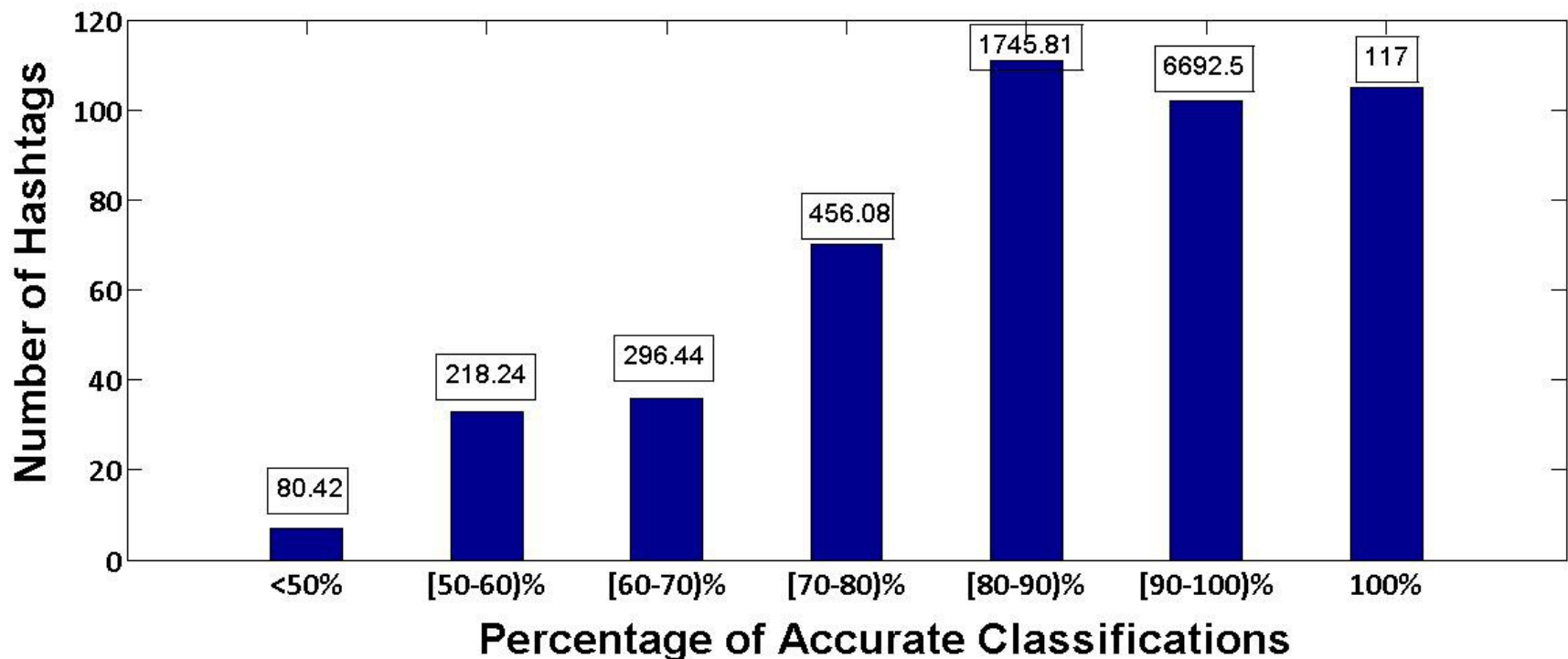
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# Retweet Model (M/C Learning Based)

- Collect the features for
  - Positive Class: Users who received the information and retweeted
  - Negative Class: Users who received the information but did not retweet
- Train a Support Vector Machine using those feature-vectors

# Retweet Model (M/C Learning Based)

$$\text{Accuracy} = \frac{|\text{Correctly classified samples}|}{|\text{Test Set}|}$$



# Validation ( Using threshold based Retweet Model)

1. Simulate model with a fixed number of initial users with different parameter values and get the popularity values
2. From dataset, collect hashtags with almost same number of initial users as the simulation
3. Check whether popularities of those real hashtags from the dataset fall within the range of simulated popularity values

Hashtag	Real Original Tweeters	Real Final Popularity	Predicted Popularity for 1100 initial users
#worldrevolution	1091	1866	[1341-3733]
#acampadasol	1125	3449	[1341-3733]

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# Conclusion

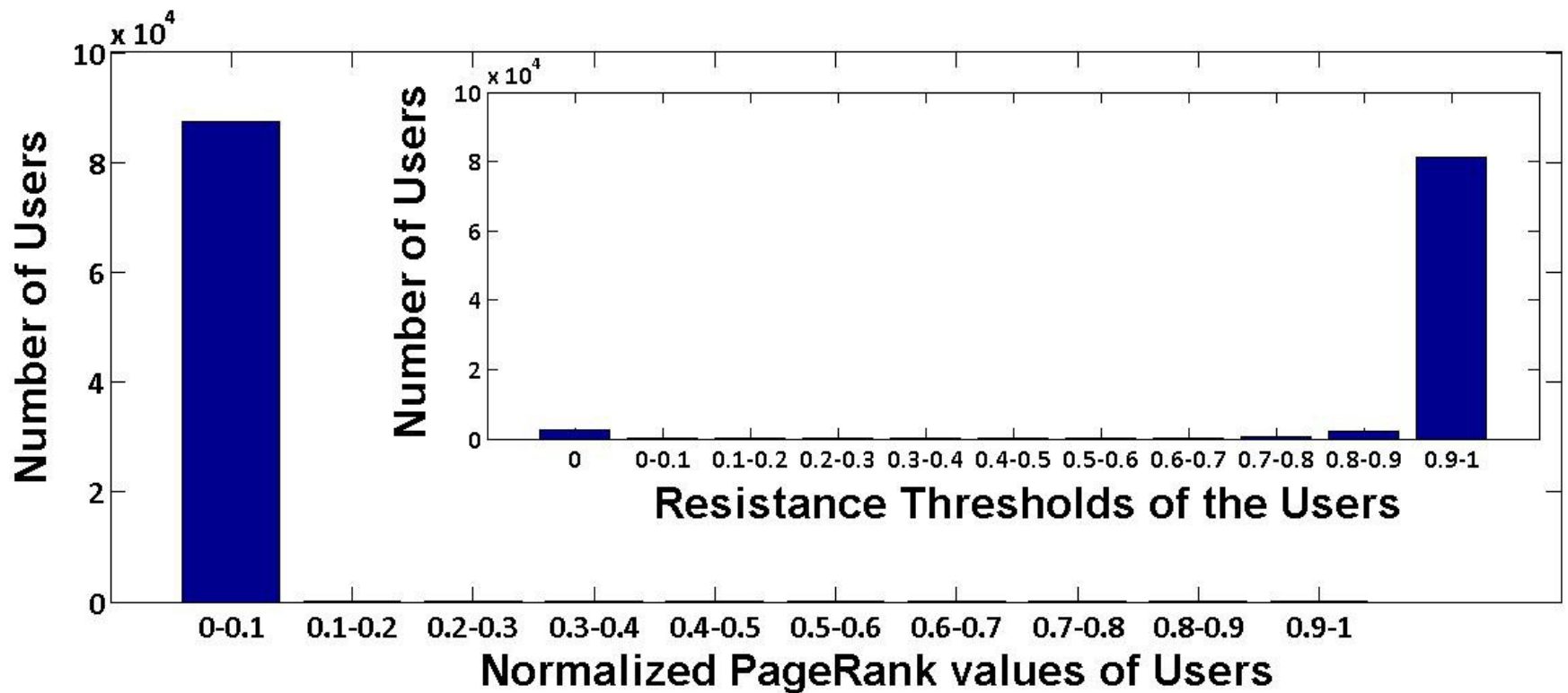
- “Mention” definitely plays a key-role in deciding the
  - Popularity of hashtags
  - Speed at which the hashtag propagates
- Using insights from simulations of our model, our recommendation System should try to
  - Suggest minimum number of users (due to character limitation of tweets)
  - So that maximum popularity can be achieved within less time

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*Thank  
You*

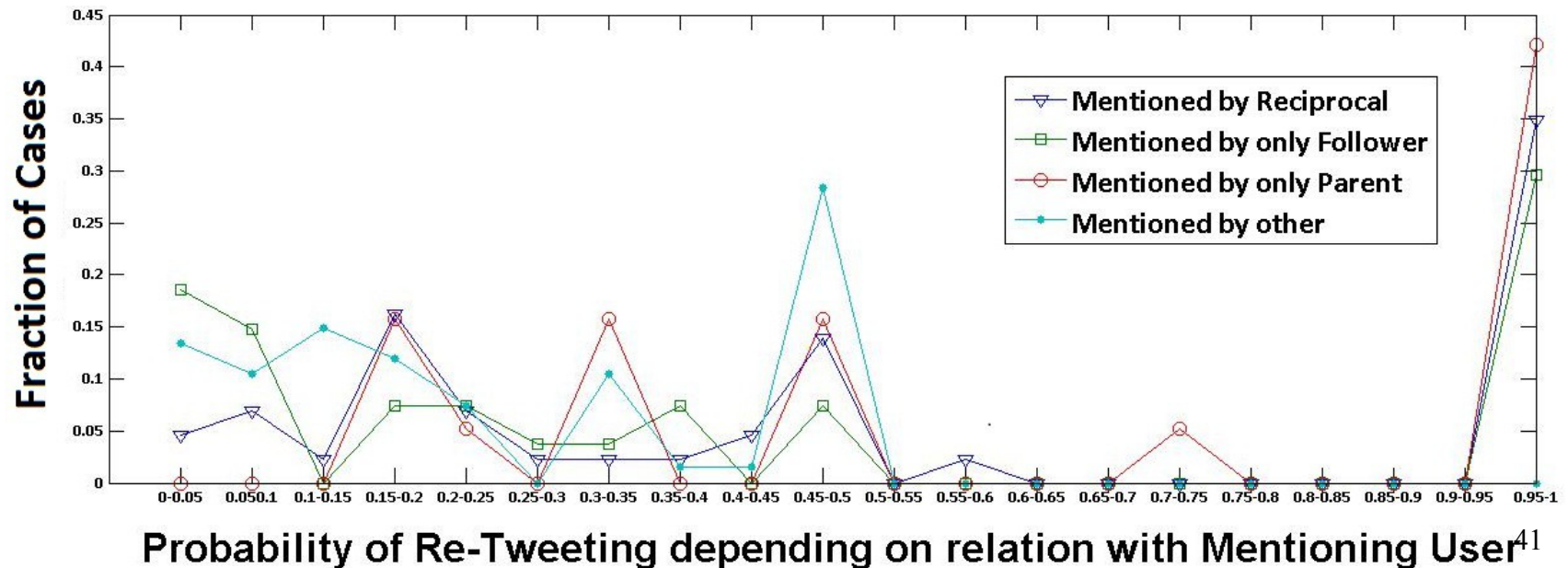
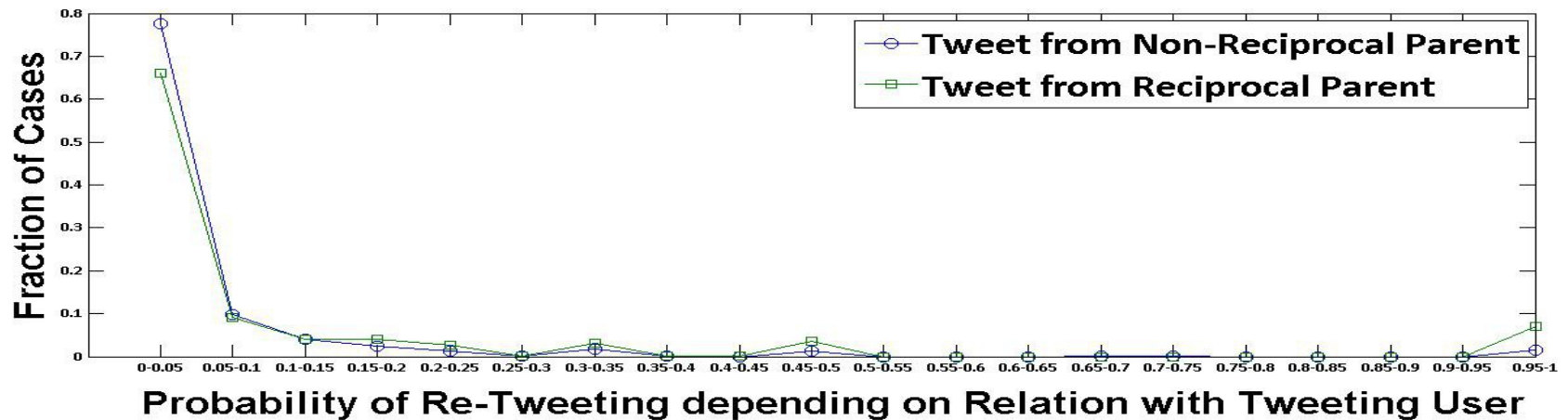
## ■ BACK-UP

# User Influence & User Resistance Distribution

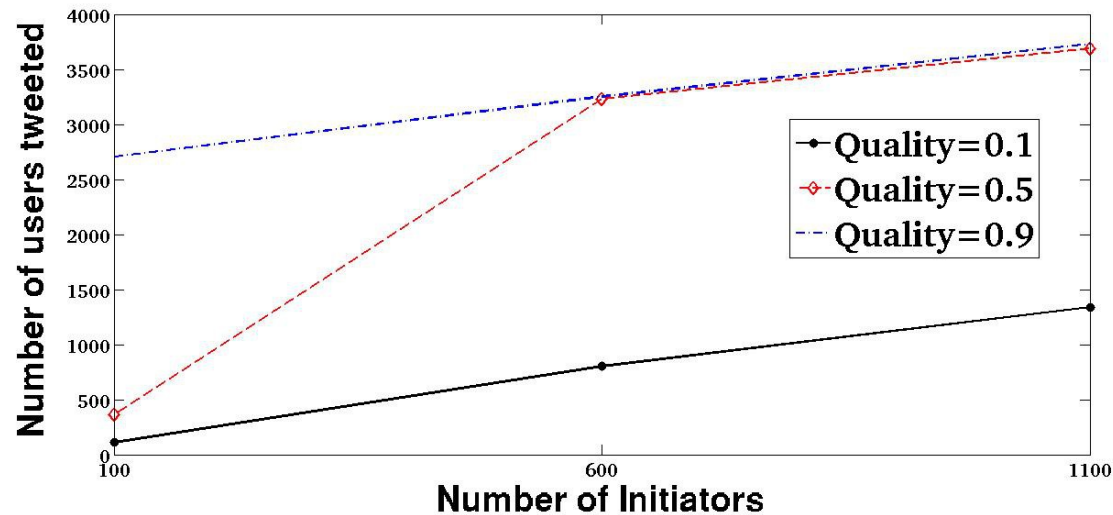




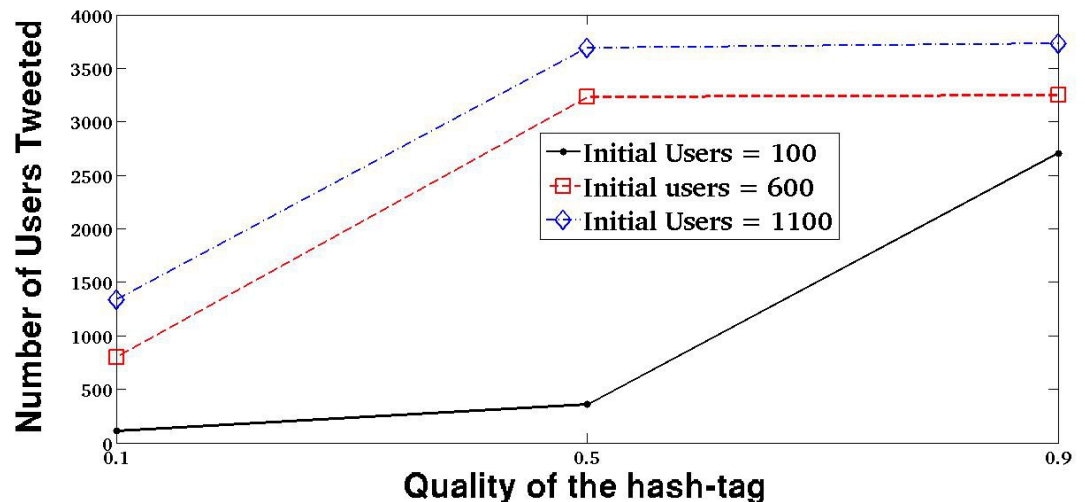
# Importance of Type of Links



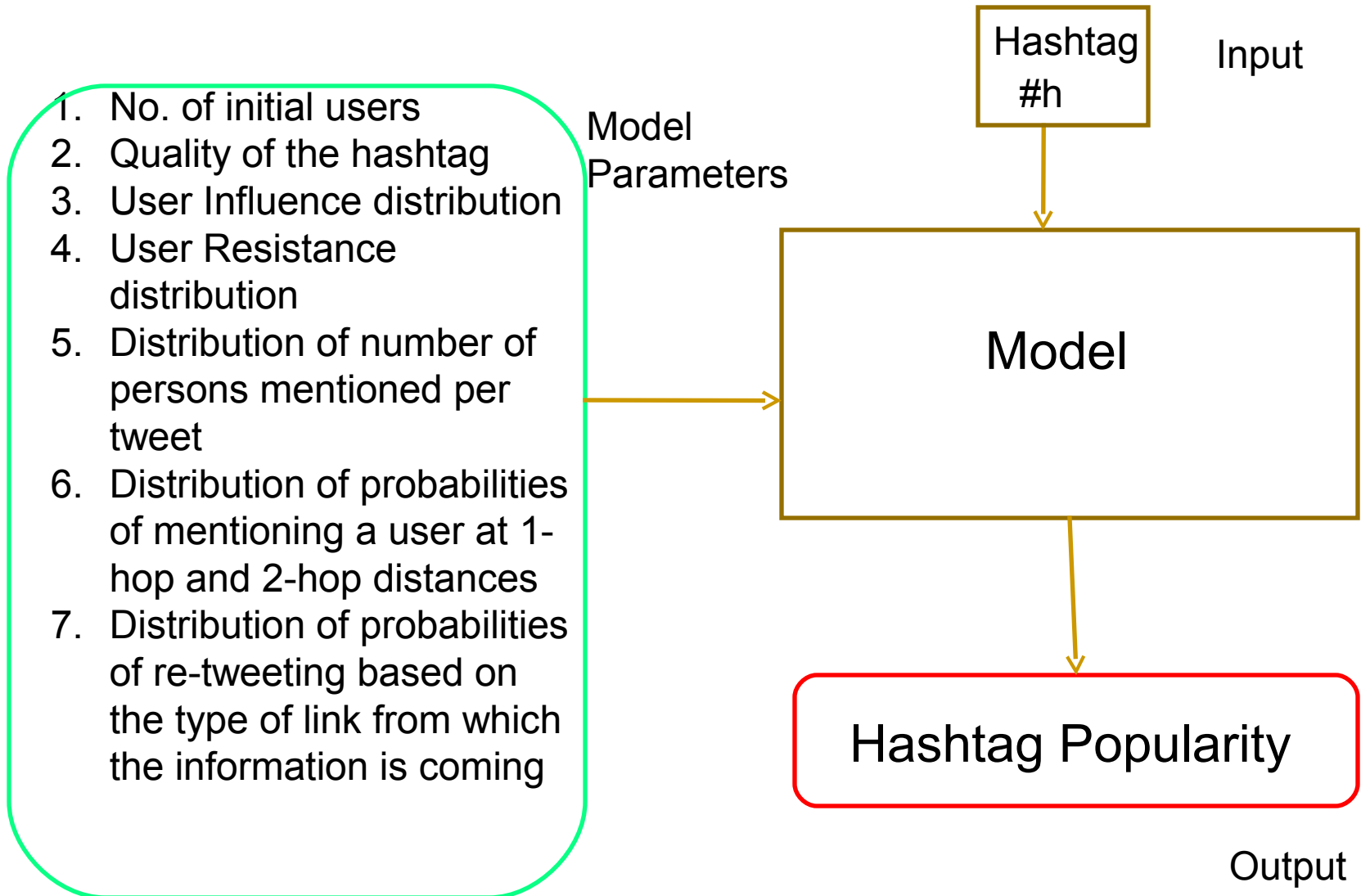
- Simulation : Vary the number of initial users  
Vary the quality of the hashtag  
(set  $\alpha=0.5$ ,  $\beta=0.5$ . Calculate other parameters from dataset)



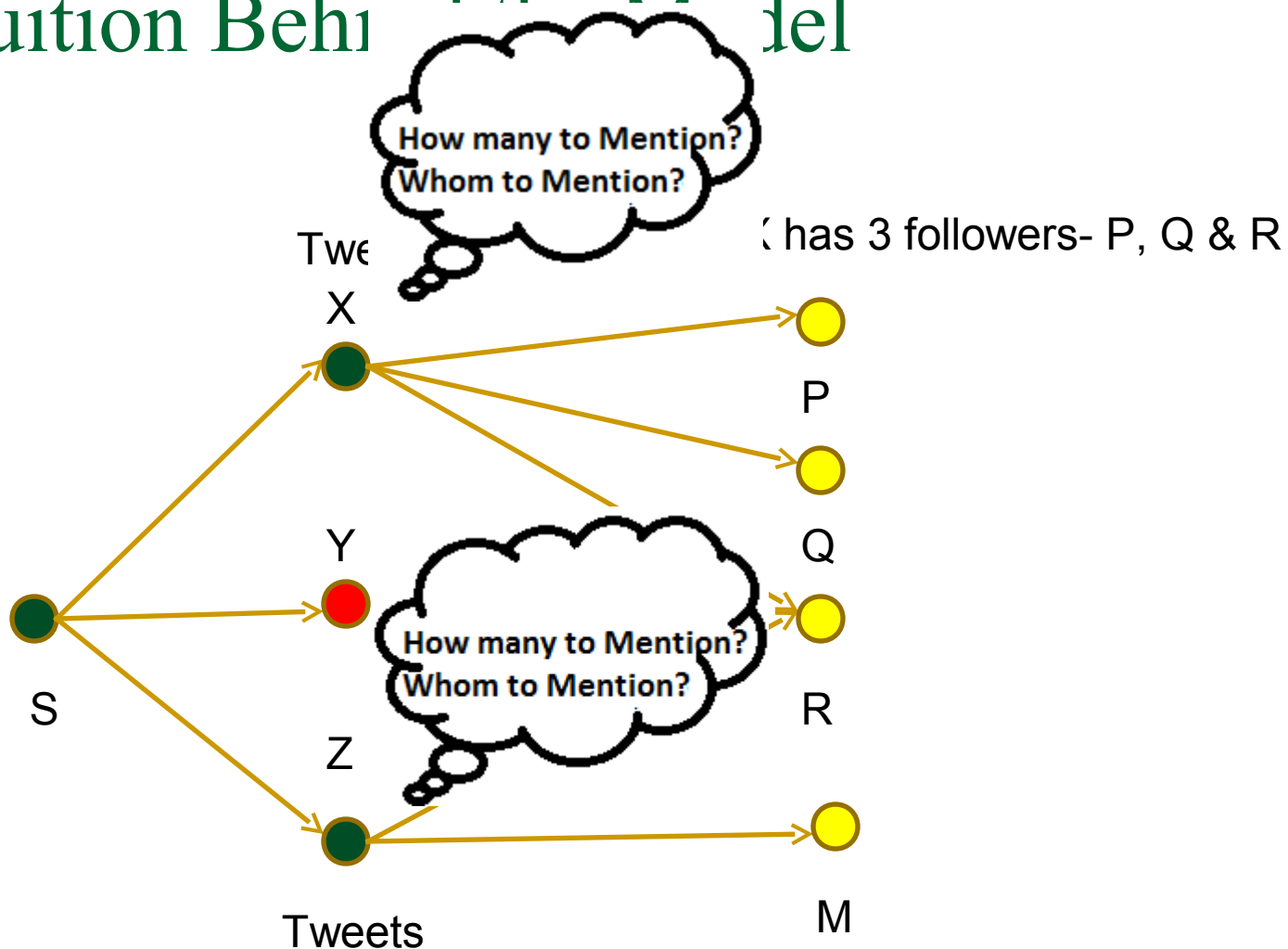
Next, we will validate whether we are getting similar reach for **real hashtags** with similar number of original tweeters.



# Model for Hashtag Propagation



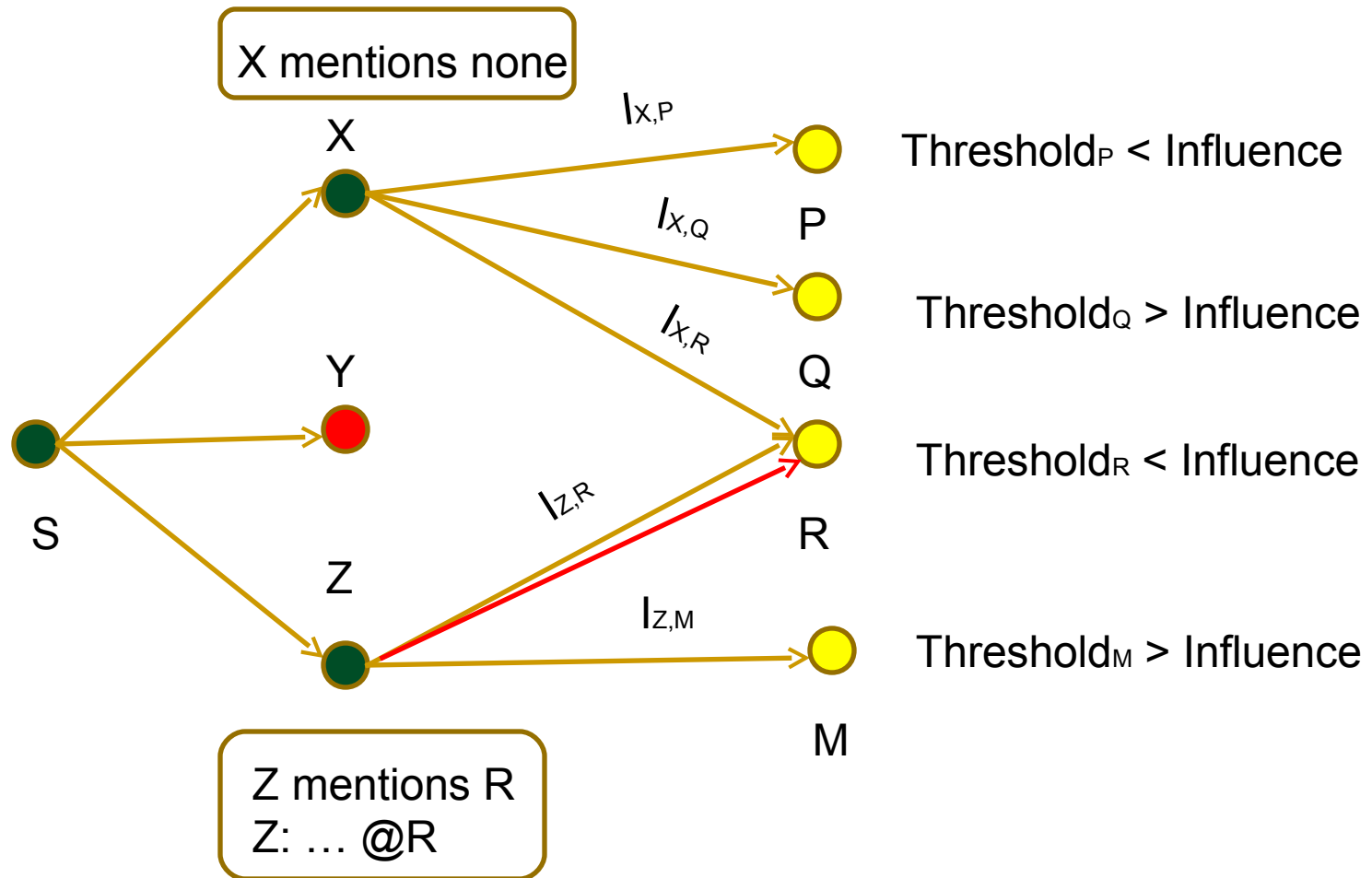
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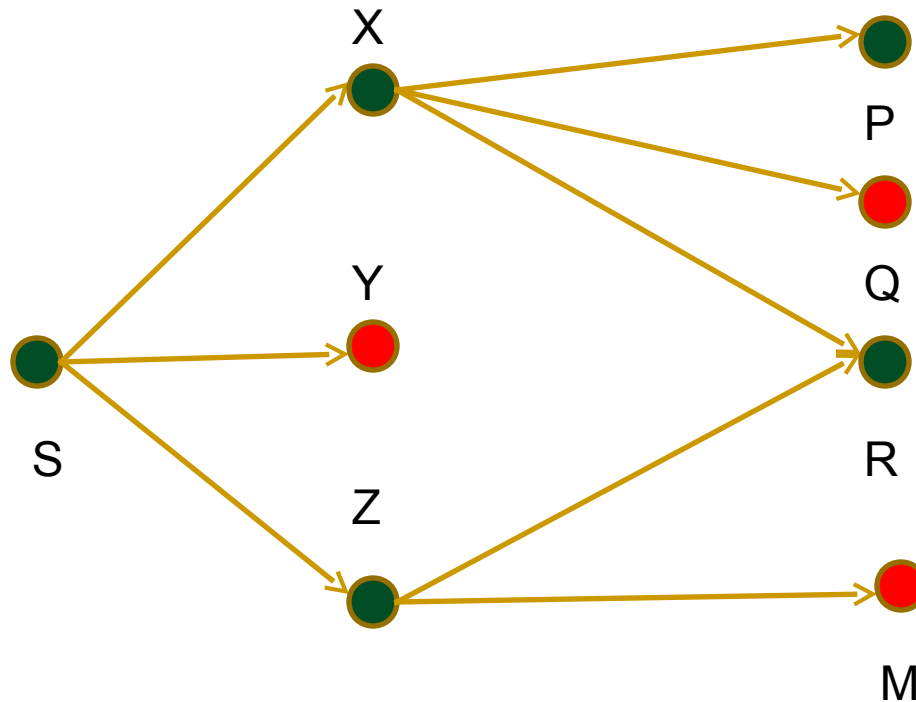
X, Z decide to propagate  
Y decides not to propagate

Z has 1 follower M

# Intuition Behind the Model



# Intuition Behind the Model



And in this way the information propagates...

P, R decide to propagate  
Q, M decide not to propagate

# Validation

Simulation results for, Initial Users= 1100

1340.77	3032.6	3689.89	3729.53	3733.26
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Popularity for different set of parameters

Hash-tag	Original Tweeters	Final Reach
#worldrevolution	1091	1866
#acampadasol	1125	3449

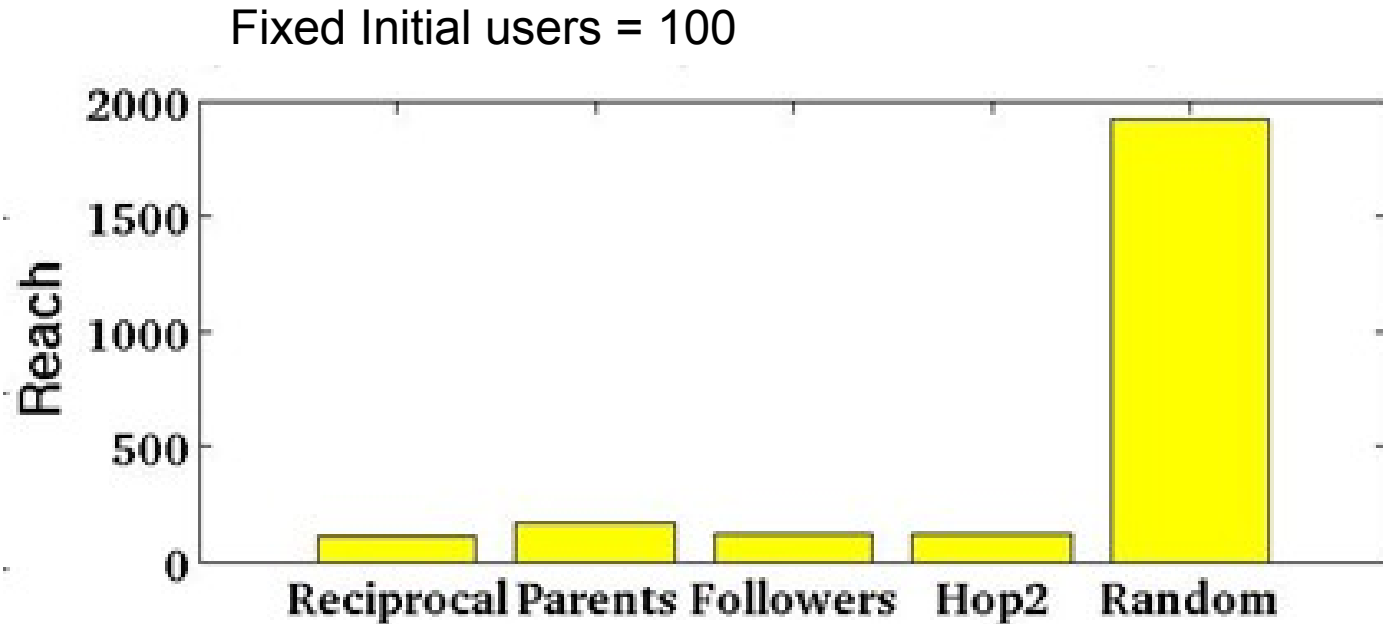
Simulation results for, Initial Users= 100

111.23	158.266	359.43	2179.03	2705.466
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Popularity for different set of parameters

Hash-tag	Original Tweeters	Final Reach
#upyd22m	102	113
#acampadalondres	111	206

# Insights



- If we keep everything fixed and
  - vary the probabilities of choosing whom to mention, randomly mentioning gives the best result