# A Novel Metric for Assessing User Influence based on User Behaviour

ANTONELA TOMMASEL – DANIELA GODOY



#### Table of Contents

- 1. Introduction
- 2. Information-based User Characterisation
- 3. Quantitatively Assessing User Influence
- 4. Data Analysis
- 5. Summary

- The potential or ability of an individual to:
  - Engage others in a certain act.
  - Induce others to behave in a particular manner.
- In micro-blogging sites, the definition relies on status attributes. (e.g. followers, re-tweets, mentions)
- The behaviour of users regarding the information diffusion might be related to their influence.

- The potential or ability of an individual to:
  - Engage others in a certain act.
  - Induce others to behave in a particular manner.
- In micro-blogging sites, the definition relies on status attributes. (e.g. followers, re-tweets, mentions)
- The behaviour of users regarding the information diffusion might be related to their influence.

However, a high level of popularity is NOT sufficient for also being influential in terms of triggering social responses.

Commercial Metrics

They do not publicly state how scores are computed.



- They have received several critics.
  - How measurements are computed.
  - Effect of spam-bots on the algorithms.

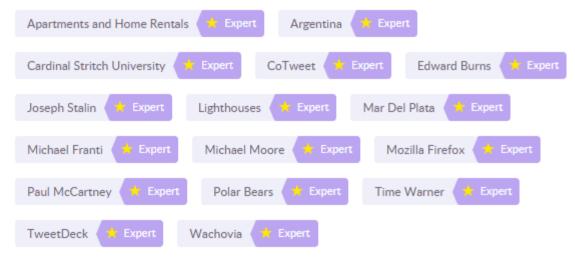


#### Comparing Commercial Metrics



#### Cristina Fernandez

Cristina Fernández de Kirchner, Presidenta de la Nación Argentina. En facebook: http://www.facebook.com/CFKArgentina



#### My Kred Communities Score CFKArgentina Follow Global Kred Yoga Outreach Level Marketing Travel

Data Analysis

My Kred Communities Score



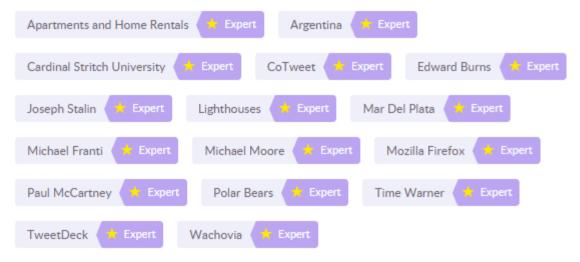


Comparing Commercial Metrics



#### Cristina Fernandez

Cristina Fernández de Kirchner, Presidenta de la Nación Argentina. En facebook: http://www.facebook.com/CFKArgentina





There is no consensus on what it means to be an influential user!!

#### User Influence

A novel metric of user influence based on analysing user behaviour regarding the patterns of information diffusion.

#### User Influence

# A novel metric of user influence based on analysing user behaviour regarding the patterns of information diffusion.

- It considers both content-based and topological factors.
- It aims at assessing the interactions with his/her neighbourhood.

#### Table of Contents

- 1. Introduction
- 2. Information-based User Characterisation
- 3. Quantitatively Assessing User Influence
- 4. Data Analysis
- 5. Summary

#### Behavioural-based User Characterisation

Information Source

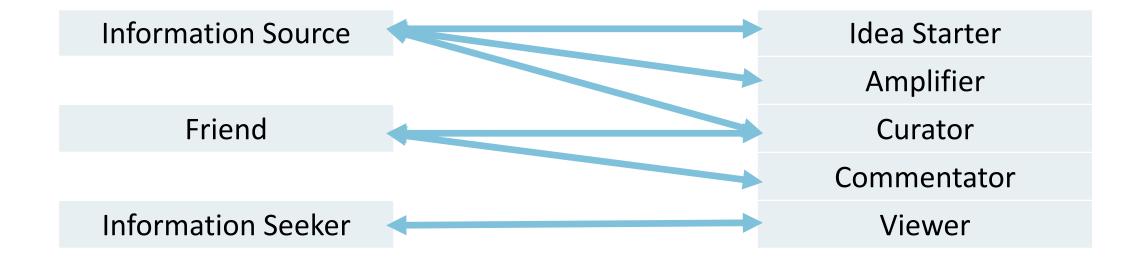
Friend

Information Seeker

Idea Starter **Amplifier** Curator Commentator Viewer

#### Behavioural-based User Characterisation

Association Between Characterisations



#### Table of Contents

- 1. Introduction
- 2. Information-based User Characterisation
- 3. Quantitatively Assessing User Influence
- 4. Data Analysis
- 5. Summary

Summary

- Novel definitions for each dimension are presented.
- A quantitative assessment of user influence is presented.
- Scores are constrained to the interval [0,1].
- Corrections are applied in order to avoid undetermined values.

Idea Starter

- Highly engaged with the media.
- The majority of their posts correspond to original content.
- They interact with a limited and selected group of users.

$$\frac{|Tweets_{ORIGINAL}\{Tweets_{ORIGINAL}RT \geq \mu - \sigma\}|}{|Tweets|} * \frac{\sum Tweets_{ORIGINAL}RT}{|ReTweets|}$$

- Assesses the proportion of original posts.
- Assesses the distribution of received retweets over the published tweets.
- Assesses the impact of posts in the neighbourhood of the user.
- The higher the score, the more the user behaves as an Idea Starter, and thus as an Information Source.

Amplifier

- Share ideas and opinions posted by other users.
- The majority of their posts correspond to retweets of Idea Starters.
- Posts are also highly retweeted by their followers.

```
\frac{|Tweets_{ORIGINAL}| + |Tweets_{REPLY}| + |Tweets_{MENTION}|}{|Tweets|} * \\ \frac{|Tweets|}{\sum Tweets_{ORIGINAL}RT + \sum Tweets_{REPLY}RT + \sum Tweets_{MENTION}RT}}{|ReTweets|} * \\
```

- Assesses the interaction between a user and his/her social network.
- Assesses the impact of posts in the neighbourhood.
- The higher the score, the more a user behaves as an Amplifier, and thus as an Information Source.

Curator

- Interact with both Idea Starters and Amplifiers.
- Share a lot of content and interact with a large number of users.

$$\frac{|Interactions \in \{Followees \cup Followers\}|}{|Interactions|} * \frac{|Interactions \in \{Followees \cup Followers\}|}{|Followees| + |Followers|}$$

- Assesses the number of interactions with other users.
- Assesses to what extend a user interacts only with his/her neighbourhood.
- The higher the number of interactions, the higher the score, and the less the user behaves as an Information Source.

Summary

# Quantitative Assessment of Personality

Follower/Followee Ratio (FF-Ratio)

$$\frac{|Followers|}{|Followers| + |Followees|}$$

- The content-related dimensions do not consider the size of the neighbourhood of a user.
- Two users might achieve the same score but have a totally different neighbourhood.
- Leverages the importance of the neighbourhood size.

Information Source Index

$$\mu_{IDAC} = \frac{Idea\ Starter + Amplifier + (1 - Curator)}{3}$$

$$IS = \frac{2 * \mu_{IDAC} * FF_{Ratio}}{\mu_{IDAC} + FF_{Ratio}}$$

Data Analysis

- High values of IS denote users behaving as Information Sources, whereas low values of IS denote users behaving as Information Seekers.
- Information Sources meet all the requirements for being regarded as influential users.
- The higher the IS of a user, the higher the influence of the user.

#### Table of Contents

- 1. Introduction
- 2. Information-based User Characterisation
- 3. Quantitatively Assessing User Influence
- 4. Data Analysis
- 5. Summary

**Data Analysis** 

## Data Analysis Settings

**Data Collections** 

Argentina

(13 users)

Miscellaneous

(8 users)

Music

(8 users)

**Politics** 

(7 users)

**Sports** 

(7 users)

Technology

(7 users)

TV Shows – Movies

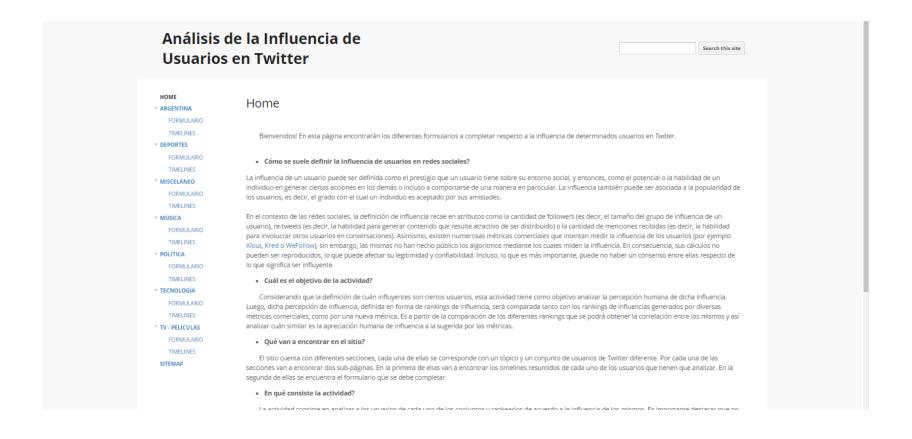
(10 users)

# Data Analysis Settings

#### Human Assessment

- Considering that there is no consensus on what means to be an influential user the human perception of influence was analysed.
- 31 Undergraduate and graduate students from an Artificial Intelligence course at UNICEN University (Argentina) ranked Twitter users according to their perceived influence.
- To combine the rankings provided by the students, Twitter users were assigned the mode of the provided rankings.

#### Data Analysis Settings Web Site



https://sites.google.com/site/influenciatwitterusers/

# Data Analysis Settings Web Site

#### **POLÍTICA**

**FORMULARIO** 

TIMELINES

@barackObama

@algore

@tonyblairoffice

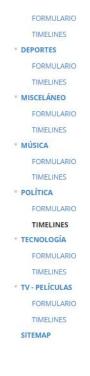
@PrensaMichelle

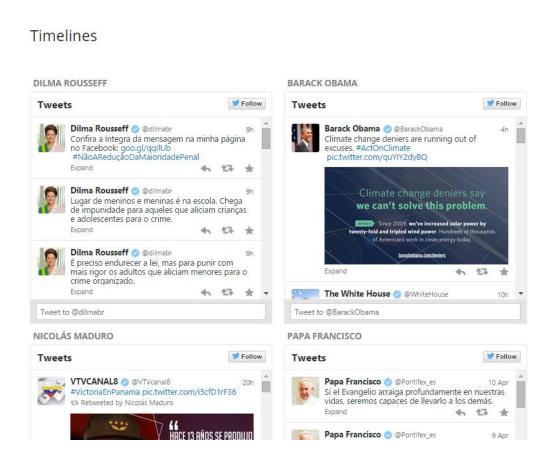
@PutinRF\_Eng

@dilmabr

@NicolasMaduro

@Pontifex\_es

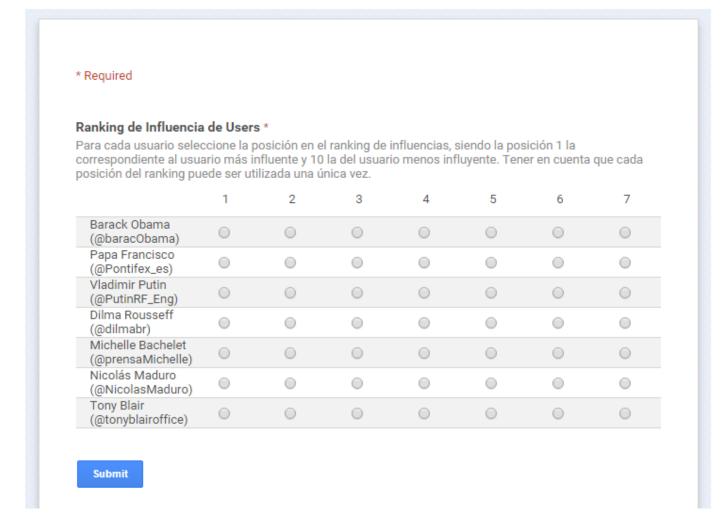




https://sites.google.com/site/influenciatwitterusers/

## Data Analysis Settings

Web Site



https://sites.google.com/site/influenciatwitterusers/

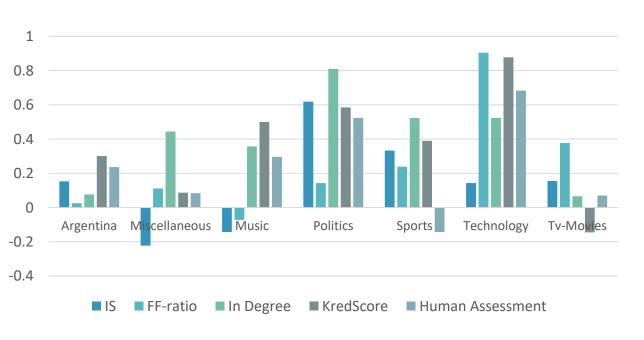
## Metrics Used for Comparison

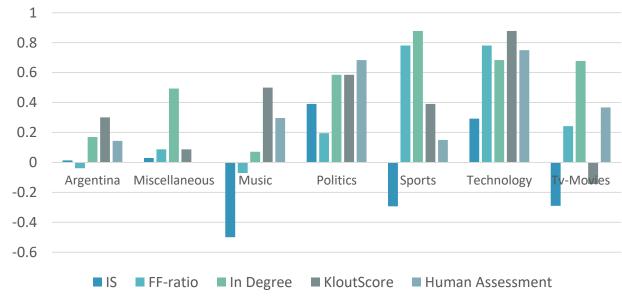
Klout Score Kred Score

In Degree FF-Ratio

Klout Vs. All

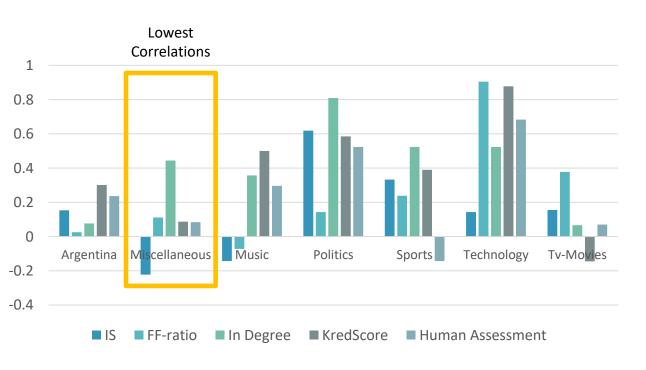
Kred Vs. All

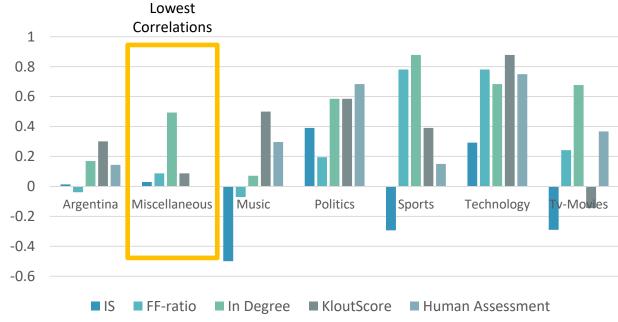




Klout Vs. All

Kred Vs. All

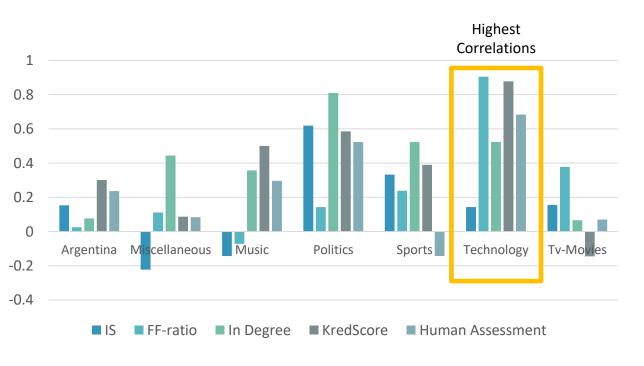


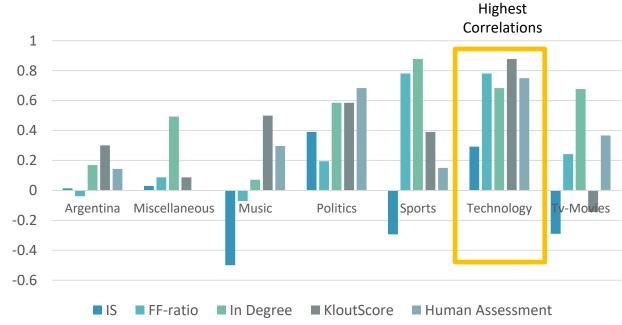


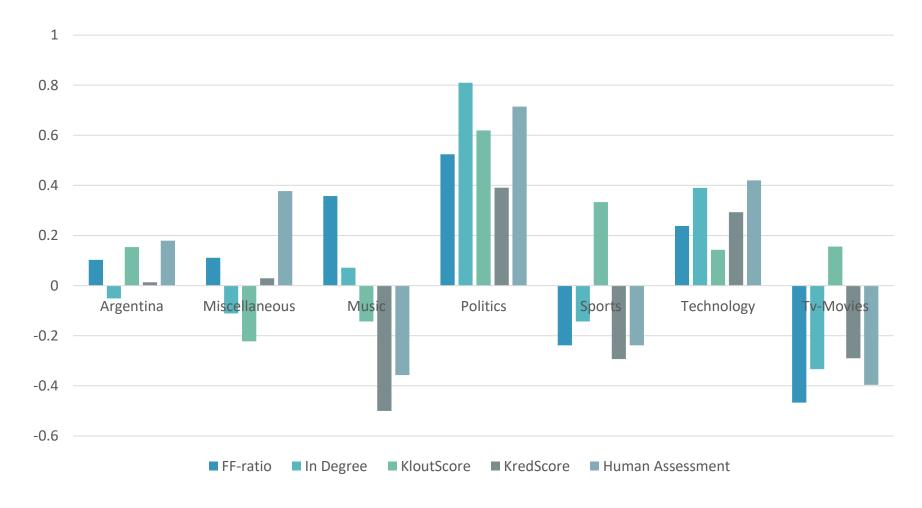
Klout Vs. All

Kred Vs. All

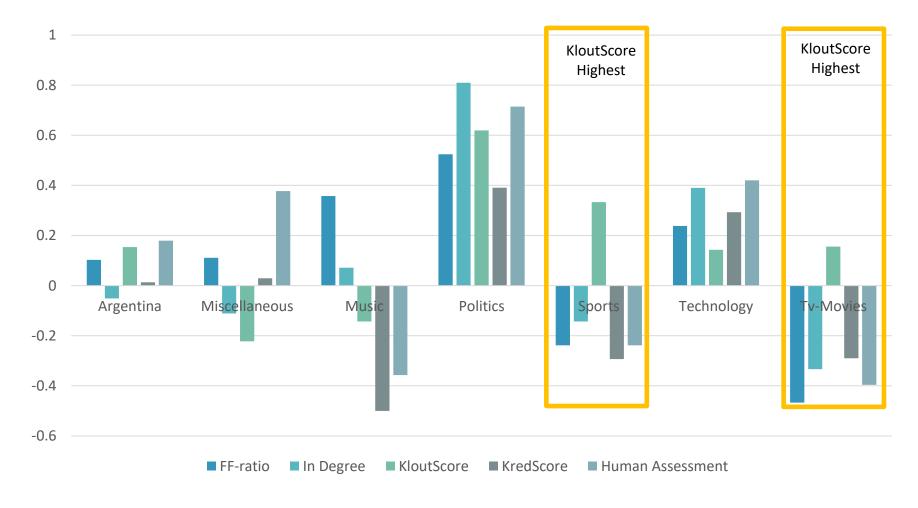
**Data Analysis** 





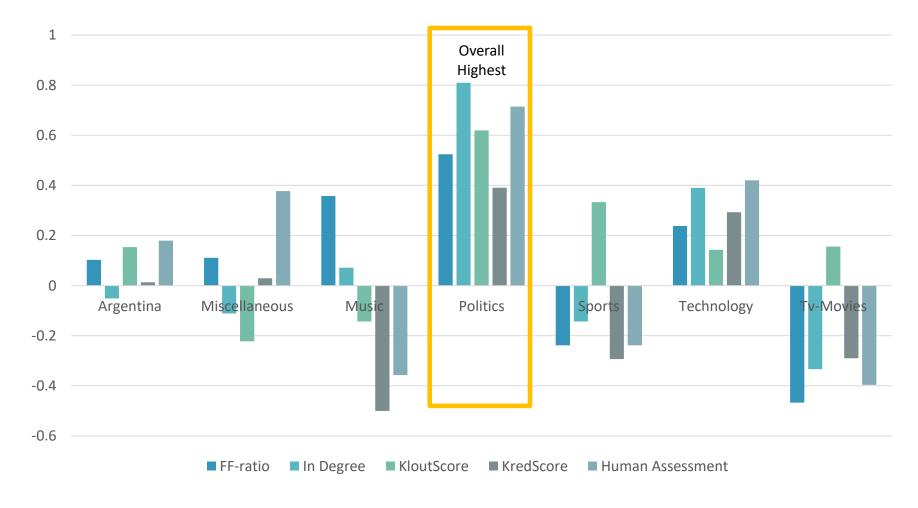




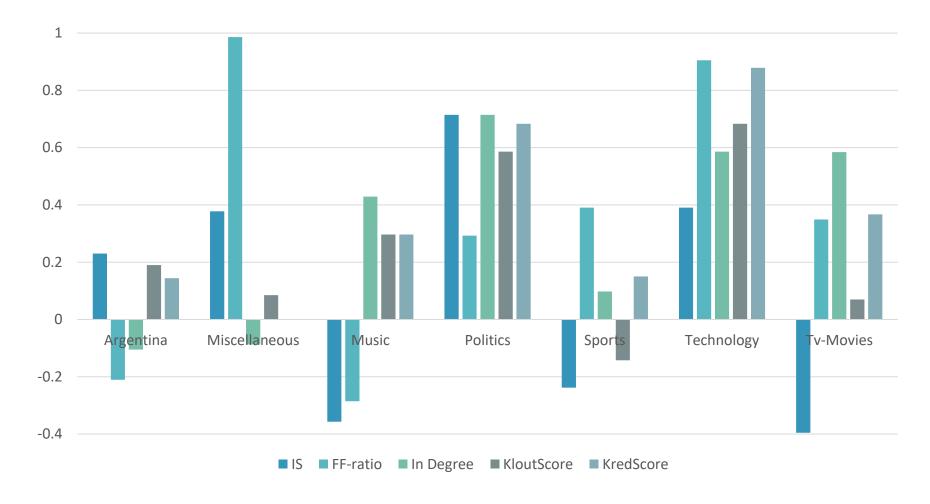


**Data Analysis** 

## Analysis of Influence Metrics



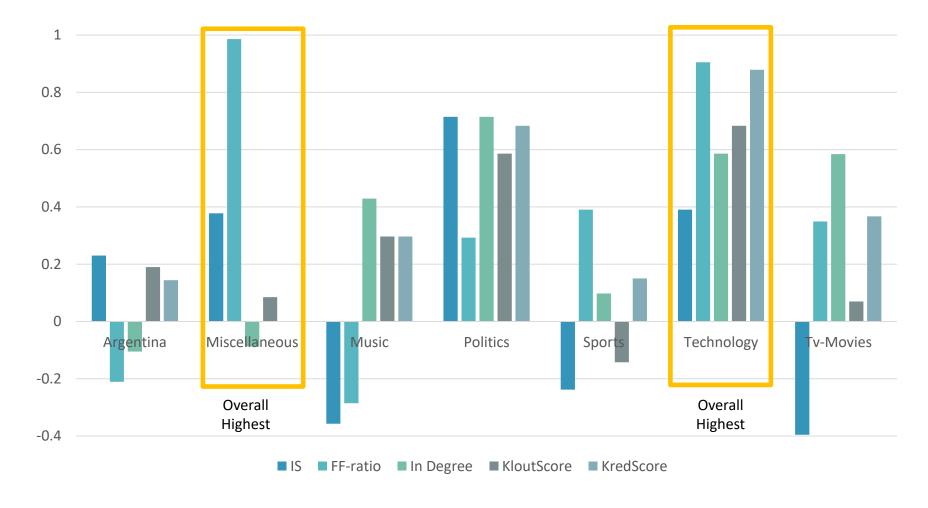
Human Assessment Vs. All



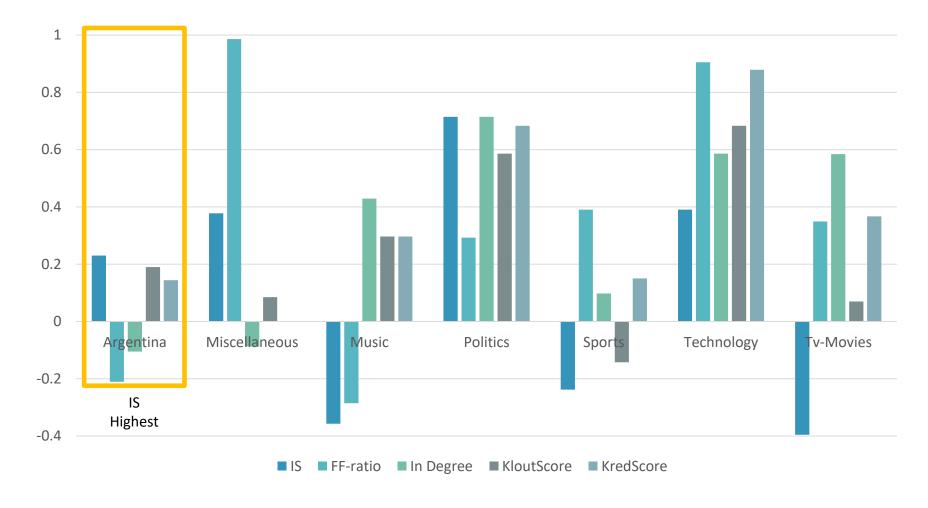
Summary

## Analysis of Influence Metrics

Human Assessment Vs. All



Human Assessment Vs. All



#### Table of Contents

- 1. Introduction
- 2. Information-based User Characterisation
- 3. Quantitatively Assessing User Influence
- 4. Data Analysis
- 5. Summary

## Summary

- This work proposes a novel metric for measuring user influence.
- The metric is based on analysing user behaviour in the diffusion of information.
- The metric considers both content-based and topological factors.
- The metric was compared to traditional means for assessing user influence, commercial metrics and a human assessment of user influence.

#### Conclusions

- There is no unique definition of user influence.
- User influence might differ according to the analysed topic.
- In most cases, the commercial metrics proved to be independent from each other.
- The highest correlations for the IS were achieved regarding the human assessment.
- Only analysing the topological factors is not sufficient for accurately assessing the influence of users.
- An accurate assessment of user influence might also consider the published content and its influence over the neighbourhood of users.

#### Conclusions

- There is no unique definition of user influence.
- User influence might differ according to the analysed topic.
- In most cases, the commercial metrics proved to be independent from each other.
- The highest correlations for the IS were achieved regarding the human assessment.
- Only analysing the topological factors is not sufficient for accurately assessing the influence of users.
- An accurate assessment of user influence might also consider the published content and its influence over the neighbourhood of users.

#### Conclusions

- There is no unique definition of user influence.
- User influence might differ according to the analysed topic.
- In most cases, the commercial metrics proved to be independent from each other.
- The highest correlations for the IS were achieved regarding the human assessment.
- Only analysing the topological factors is not sufficient for accurately assessing the influence of users.
- An accurate assessment of user influence might also consider the published content and its influence over the neighbourhood of users.

## Questions?

Introduction



# A Novel Metric for Assessing User Influence based on User Behaviour

ANTONELA TOMMASEL – DANIELA GODOY

