Combating misinformation on social media: from detection to mitigation

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Overview

- Part I: Misinformation and human perception (Jenny)
 - Social media is a double-edged sword
 - Information credibility
 - User perception of information perception
- Part II: Misinformation detection (Jey Han, online)
 - Automated Fact Checking
 - Rumour Detection
 - Challenges
- Part III: Misinformation mitigation. (Jenny)
 - Information propagation models
 - Network-level mitigation
 - Personalised mitigation

Part I: Misinformation and human perception

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Contents

- Social media is a double-edged sword
- Information credibility
- User perception

Social media is more than text-based media

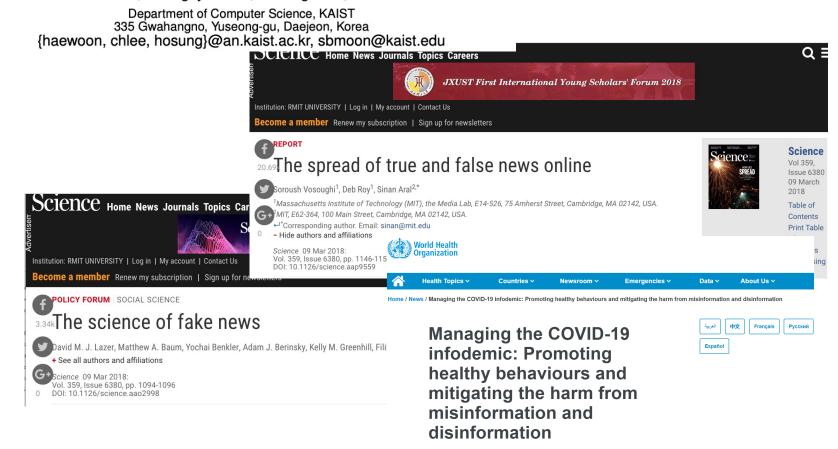
- User generated contents: text posts and comments, photos and videos.
- Social network: connecting users or groups.



Social media is a double-edged sword

What is Twitter, a Social Network or a News Media?

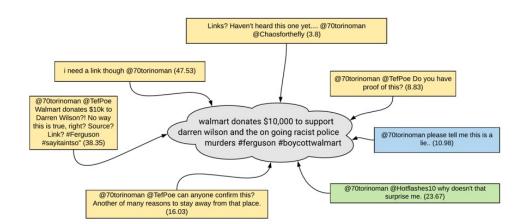
Haewoon Kwak, Changhyun Lee, Hosung Park, and Sue Moon



What is misinformation?

- "Misinformation is incorrect or misleading information" – Merriam-Webster Dictionary.
- Types of misinformation: false claims, rumours, fake news, and disinformation.





Understand user perception of social media information credibility

- "Credibility is the quality of being believed or accepted as true, real, or honest, whether it regards the information or the source". (Tseng &Fogg, 1999).
- Judgement of information credibility is a type of information behaviour.



11 users have different credibility ratings:

- Very credible 6
- Somewhat credible 1
- Not credible 3
- Cannot decide 1

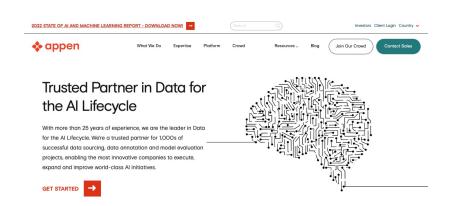
Research questions

- Do users' demographic attributes significantly correlate with their judgement of information credibility?
- What features users use to judge social media information credibility?
- Are there confounding factors for information credibility judgement?
- Are humans credulous (more than machines) about social media information?

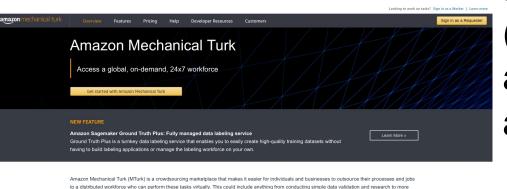
Information credibility for news on Twitter

- Twitter news data: keyword search on Twitter
- User demographic data: questionnaires on crowd sourcing platform
- User credibility ratings: crowd source user study
- Data analysis: multi-way chi-square analysis and association analysis

User study on crowdsource platforms



- Ensuring data quality:
 - Specify qualification level of crowd workers.
 - Gold questions were used to ensure data quality.



subjective tasks like survey participation, content moderation, and more. MTurk enables companies to harness the collective intelligence, skills, and insights from a global workforce to streamline business processes, augment data collection and analysis, and accelerate machine learning development.

• Filter crowd workers. 40 (2*20) Gold questions are randomly inserted as gold questions.

User demographics: gender, age, education, country.

754 users(crowd workers) from 76 countries. Demographic attributes grouped into binary and categorical levels.

- Age: binary (young, old); categorical (Boomer, Gen X, Gen Y, Gen Z).
- Education: binary (below and above university); categorical (school, non-degree tertiary, undergrad and postgraduate)
- Location: binary (Eastern and Western); categorical (Asia-pacific, America, Europe, Africa)

Table 1 Demographic profiles distribution.

| Demographic | Value | # | % |
|-------------|----------------------------|-----|------|
| Gender | Male | 521 | 69.2 |
| | Female | 233 | 30.8 |
| Age | 16-19 years old | 58 | 7.7 |
| | 20-29 years old | 327 | 43.4 |
| | 30-39 years old | 243 | 32,2 |
| | 40-49 years old | 89 | 11.8 |
| | 50 years and older | 37 | 4.9 |
| Education | High school | 127 | 16.8 |
| | Technical training | 58 | 7.7 |
| | Diploma | 81 | 10.7 |
| | Bachelor's degree | 287 | 38.1 |
| | Master's degree | 137 | 18.2 |
| | Doctorate degree | 14 | 1.9 |
| | Professional certification | 50 | 6.6 |
| Location | Asia | 275 | 36.5 |
| | Europe | 247 | 32.8 |
| | South America | 130 | 17.2 |
| | North America | 65 | 8.6 |
| | Africa | 37 | 4.9 |

Twitter posts

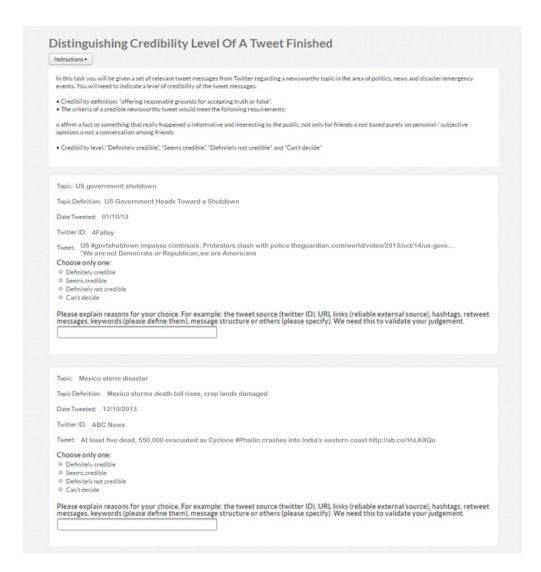
- 1510 news tweets for 20 news topics from major online newswires.
 - BBC, Reuters, CNN, Guardian and New York Times.
 - Topic keywords are used to search for relevant Twitter posts. Examples: "US government shutdown", "Navy Yard shooting", "Earthquake in Pakistan".
- Three types of news -- breaking news, political news and natural disaster news -in 4 years.
- Trending and non-trending news topics.
- Two writing styles: opinion/emotion-bearing and factual.

Table 2
Tweets news attributes distribution.

| News attribute | Value | # | % |
|----------------|------------------|-----|------|
| News type | Breaking news | 509 | 33.8 |
| | Natural disaster | 500 | 33.2 |
| | Politic | 499 | 33.0 |
| Year | 2011 | 374 | 24.8 |
| | 2012 | 375 | 24.9 |
| | 2013 | 377 | 25.0 |
| | 2014 | 382 | 25.3 |
| Trending | Trending | 781 | 51.8 |
| | Not trending | 727 | 48.2 |

Credibility ratings

- Four credibility rating levels:
 - Definitely credible
 - Somewhat credible
 - Definitely Not credible
 - Cannot decide
- Exclude the middle level "maybe credible" option to avoid "lazy" judgements.
- Free text data to collect features.



User reported features for credibility judgements

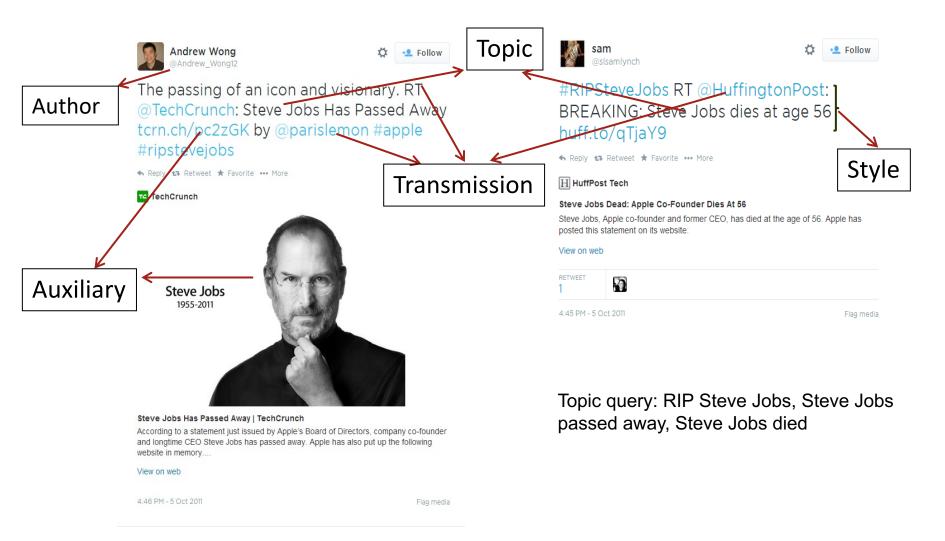
Summative content analysis of textual data (Hsieh, 2005)

Table 3 Features reported by readers to judge credibility for news tweets.

| Category | Feature | Description |
|--------------|------------------|---|
| Author | Tweet author | Twitter ID or display name |
| | | e.g. Sydneynewsnow |
| Transmission | User mention | Other Twitter user's Twitter ID |
| | | mentioned in the tweet starting |
| | | with the @ symbol e.g. @thestormreports |
| | Hashtag | The # symbol used to categorise |
| | | keywords in a tweet e.g. #Pray4Boston |
| | Retweet | Contain the letters RT (retweet) |
| | | in the tweet and the retweet count |
| Auxiliary | Link | Link to outside source - URLs, |
| | | URL shortener |
| | Media | Picture or video from other sources |
| | | embedded within the tweet |
| Topic | Alert phrase | Phrase that indicates new or |
| | | information update regarding |
| | | a news topic - e.g. Update |
| | Topic keyword | The search keyword regarding |
| | | a news topic e.g. Hurricane Sandy |
| Style | Language | The language construction of the tweet |
| | | (formal or informal English) |
| | Author's opinion | Tweet that conveys the author's |
| | _ | emotion or feeling towards the news topic |
| | Fact | Factual information on the tweet |
| | | regarding the news topic |

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A user study on features for users to judge credibility of news on Twitter



Correlation analysis of user demographics and credibility ratings

- Correlation analysis (chi-square) of user demographics at original, binary and categorical levels.
- Only location is significantly correlated with credibility ratings at all levels. The African-user –cannot-decide combination gives the highest dependence.
- Education is strongly correlated with credibility judgement. Further analysis
 (correlation rule, Brin et al, 1997) reveals that strong dependence originates from the
 group of users with a professional certification giving "not credible" rating.

Table 5Demographic profiles and credibility perception chi-square results.

| Demographic | Data setting | Credibility | dibility | |
|-------------|--------------|-----------------|----------|--|
| | | \mathscr{Z}^2 | p-value | |
| Gender | Original | 1.51 | 0.68 | |
| | Binary | 1.51 | 0.68 | |
| | Categorical | 1.51 | 0.68 | |
| Age | Original | 14.87 | 0.25 | |
| | Binary | 4.68 | 0.20 | |
| | Categorical | 9.84 | 0.13 | |
| Education | Original | 49.43 | 9.20E-5 | |
| | Binary | 4.78 | 0.19 | |
| | Categorical | 12.29 | 0.20 | |
| Location | Original | 80.79 | 2.92E-12 | |
| | Binary | 39.62 | 1.29E-8 | |
| | Categorical | 80.33 | 1.39E-13 | |

Correlation analysis of news attributes and credibility

- News-type correlates with Credibility, and strong dependence for "breaking news" and "very credible".
- Trending correlates with Credibility, and strong dependence for "trending" and "very credible".

Table 7
News attribute correlation with reader's credibility perception.

| News attribute | Credibility | | |
|----------------|-----------------|----------|--|
| | \mathscr{X}^2 | p-value | |
| News type | 93.75 | 5.04E-18 | |
| Year | 61.89 | 5.78E-10 | |
| Trending | 8.09 | 0.04 | |

Correlation analysis of user demographics and news attributes

- User-age correlates with news-year.
- User-location correlates with news-type.
- User-location correlates with news-year.

Table 9Correlation between combination of reader's demographics and news attributes with credibility perception.

| Demographic | News attribute | Credibility | |
|-------------|----------------|-----------------|---------|
| | | \mathscr{X}^2 | p-value |
| Gender | News type | 6.94 | 0.33 |
| | Year | 8.43 | 0.49 |
| | Trending | 7.38 | 0.06 |
| Age | News type | 35.53 | 0.06 |
| | Year | 53.06 | 0.03 |
| | Trending | 18.59 | 0.10 |
| Education | News type | 47.81 | 0.09 |
| | Year | 64.56 | 0.15 |
| | Trending | 16.92 | 0.53 |
| Location | News type | 38.35 | 0.03 |
| | Year | 55.16 | 0.02 |
| | Trending | 17.17 | 0.14 |

User perception versus machine prediction for credibility



Gupta, Aditi, et al. "Tweetcred: Real-time credibility assessment of content on twitter." SocInf 14.

Human perception vs. machine prediction for credibility

Humans are more credulous than machines on the credibility of news tweets.

Table 4.4: The agreement matrix between reader's credibility perception and automated credibility prediction

| | | TweetCred | | | | |
|---|-------------------|-----------|-----|-----|-------|--|
| Very credible Somewhat credible Not cre | | | | | Total | |
| LS | Very Credible | 256 | 654 | 67 | 977 | |
| \mathfrak{R} eaders | Somewhat credible | 51 | 230 | 50 | 331 | |
| \mathbf{Re} | Not credible | 1 | 4 | 3 | 8 | |
| | Total | 308 | 888 | 120 | 1316 | |

User demographics and credibility features

- All demographic attributes are somewhat correlated with credibility features.
- Demographic attributes are most correlated with topic and style features while least correlated with transmission features.
- Association analysis shows that Author and Auxiliary and Transmission are frequently used together.

Table 11The chi-square correlation between demographics and features used in credibility perception.

| Demographic | Feature categories | | | | |
|-------------|----------------------------|------------|-------------------------|-----------------------------|--------------------|
| | Author (\mathscr{X}^2) | Topic (ℋ²) | Style (\mathcal{X}^2) | Auxiliary (\mathcal{X}^2) | Transmission (p) |
| Gender | 0.01 | ***18.15 | ***23.27 | 1.59 | 0.59 ^a |
| Age | ***16.63 | ***26.65 | ***41.99 | 8.65 | 1.00 ^a |
| Education | 11.12 | ****31.87 | ***50.12 | **16.53 | *0.03 ^a |
| Location | ***46.87 | ****83.81 | ****67.35 | ***13.60 | 1.00 ^a |

p < 0.05, p < 0.01, p < 0.001, p < 0.001.

^a Calculated using Fisher's Exact Test.

Looking ahead

- Strategies for publishing effective, truthful information to convince end users.
 - Official news media, fact-checking services
- Insights for improving machine learning systems for better information service.
 - Location-based information service.
- What about machine generated misinformation?
 - Can machines generate seem-credible information to deceive social media users?
 - Do the features in machine generated texts affect human perception in the same way?

Acknowledgements





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