Media Attention to Science

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1. ABSTRACT

How has media attention to science changed over the past decade? Does media attention bring scientific attention too? To answer these questions, we collected media attention statistics from Altmetric.com for over 40,000 papers published in PNAS journal in the last 13 years. Our analysis reveals that (i) Media attention to science has slowly, but steadily increased over time, (ii) Media attention doesn't necessarily translate into scientific attention (citations) and, (iii) It is non-trivial to predict the amount of media attention a paper gets from attributes such as authors, author affiliations, abstract, and title.

2. RELATED WORK

Darling et al. [2] studied the role of social media in the life cycle of a scientific publication and conclude that Twitter has been playing an ever increasing role in the dissemination and publicity of scientific literature. Haustein et al. and others [3, 1] propose and validate the use of new metrics (alt-metrics) that can capture the real world impact of science, including social media buzz created and number of mainstream media mentions gathered by a scientific article. Many popular journals and publishers have already embedded altmetric scores on their online webpages. Closest to our study is the work by Zoller et al [4], who compare citations and altmetric scores from a social bookmarking system. Their analysis on more than 250,000 publications shows that citations and altmetrics are weakly correlated. To the best of our knowledge, our study is the first to look at media related metrics for scientific studies and perform analysis on the relationship between popularity in main stream media and popularity in the scientific community.

3. DATA

We collected data from the Proceedings of the National Academy of Sciences (PNAS journal) for 13 years (2004–2016). This gave us 44,109 papers, which cover a wide range of disciplines, including mathematics, physics, chemistry, biology, etc. Each paper has an Altmetric score[1] (a score indicating the media attention a paper gets), its coverage in the popular news media, social media (Twitter, Facebook, Reddit, etc) and scientific popularity stats like number of downloads, number of readers on Mendeley, etc. For each paper, we also obtained the number of citations from the Web of Science database. Figure 1 shows an example for one paper.[2] This paper has an Altmetric score of 138, and mentioned by 11 news media outlets. Since social media attention is a relatively new phenomenon (starting around 2008), we only consider mentions in main stream news as a sign of media attention to a paper. Our datasets are available for download on the project webpage.[3]

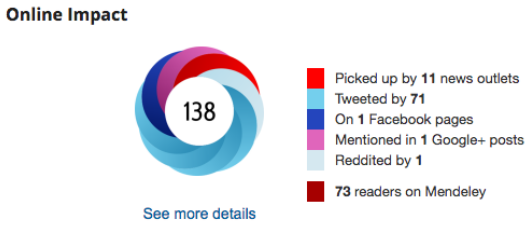

Figure 1: Example showing media metrics for a PNAS paper.

4. FINDINGS

Media attention over the years PNAS publishes a new version of the journal every week. Each paper is associated with a week in which it was published. For each paper published in a week, we extract the number of mentions in news media and compute the average number of news media mentions over all papers in each week. We plot this data over a period of 654 weeks (2004–2016). Figure 2 shows the average number of media mentions for all papers (blue) and for a specific field (Neuroscience, the most frequent field in our data, consisting of 3,500 papers). Both have a slight positive slope for a linear regression line fit (Slope 0.0071 for all papers and 0.0073 for Neuroscience papers). We tested the value of the non-zero slope for statistical significance using a t-test ($p < 0.0001$). Similar trends were found for the other fields in our data, indicating a slow but steady increase in media attention to science over the last decade.

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http://dx.doi.org/10.1145/3041021.3054215.

[1] http://bit.ly/2iqm7k3
[2] e.g. see http://bit.ly/2iHxXEc
[3] https://users.ics.aalto.fi/kiran/mediaAttentionScience/

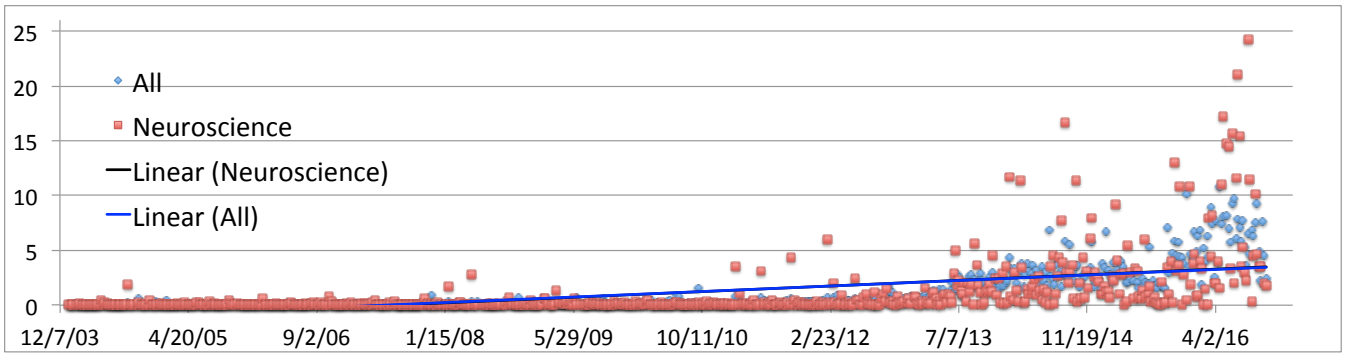

Figure 2: Average number of news media mentions over time. The regression fit lines overlap.

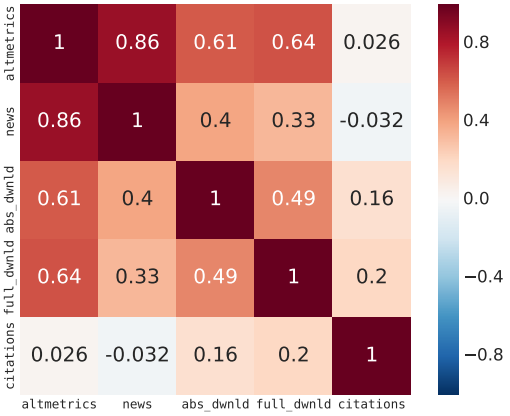

Figure 3: Heatmap of correlations. *altmetrics*: **Altmetric score,** *news*: **Number of mentions in news media,** $\{abs, full\}$*_dwnld*: **Number of downloads of the abstract and full version of the paper.**

Media attention vs. scientific attention Next, we compared the attention that a paper gets in popular media with scientific attention metrics such as: (i) Number of downloads of the abstract, and the full paper (both these are a proxy for the amount of interest in the paper and visits to the web version of the paper), and (ii) Number of citations (a proxy for the amount of scientific interest). We computed correlations between the various values. Figure 3 shows a heatmap of the Pearson correlations between the various metrics. From the figure, we can observe that: (i) Number of citations is not correlated with media popularity, (ii) Number of downloads of the paper is correlated with the media attention, indicating that media attention results in increased visits to the paper by inquisitive readers, but this does not necessarily translate into scientific popularity.

Can we predict media attention? We built regression models to predict the amount of media attention a paper gets (target variable: number of news mentions) based on features extracted from: (i) words in the paper title, (ii) words in the paper abstract, (iii) author's name(s), (iv) author's affiliation(s), and (v) research field of the paper. We experimented various standard regression models including Lasso, Linear Regression, Ridge, and Decision Trees, but find that the performance of these models is almost random. i.e., it is *not* possible to predict the media attention a paper gets. We suspect this bad performance to be because the target variable is power-law distributed. For example, 84% of papers have no mentions in the media. We also experimented with the regression only on papers which have at least one news mention, but the results are similar.

Table 1: Top 10 correlated features with media attention.

Feature	Coefficient
psychological_and_cognitive_sciences (field)	0.154
health (abstract)	0.112
people (abstract)	0.095
united (abstract)	0.08
earth__atmospheric__and_planetary_sciences (field)	0.08
sustainability_science (field)	0.078
decades (abstract)	0.077
climate (abstract)	0.074
mortality (abstract)	0.074
century (abstract)	0.074

Feature analysis Next, we analyzed the features that are most positvely correlated with media attention. For each feature, we computed its Pearson correlation coefficient with the media attention a paper containing this feature gets. From Table 1, we can get a sense of what factors could potentially lead to news media attention. For example, papers with keywords 'health', 'mortality', 'climate', etc in the abstract get more attention and fields like Psychology and Sustainability, studying mental health, and climate change, get more media attention.

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5. REFERENCES

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