

Assessing the scientific impact of research

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The constantly increasing size of scientific output



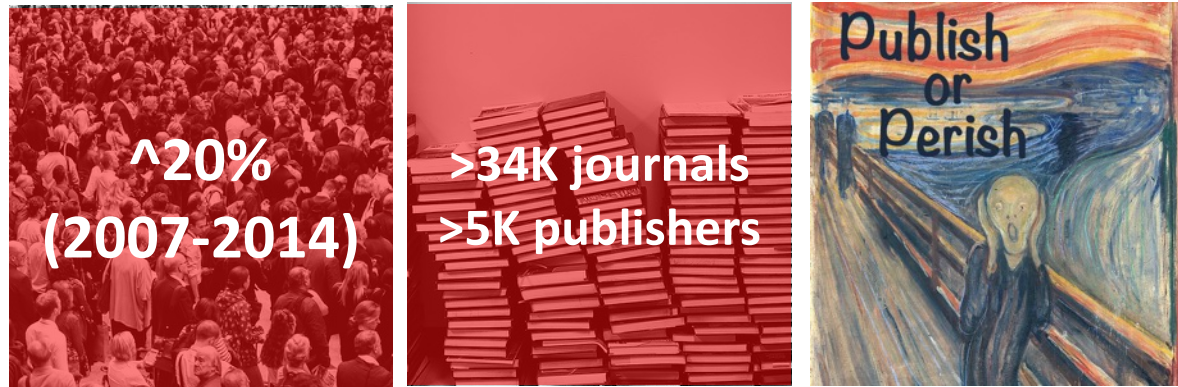
- **Exponential growth of scientific publications**
 - Increase in the number of researchers worldwide ^[1]
 - Large number of journals ^[2]
 - “Publish or perish” ^[3]

The constantly increasing size of scientific output



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In many cases, articles contain low quality research^[4,5]...



honest mistakes because
of time pressure (publish
or perish)



predatory publishers



data manipulation
e.g. p-hacking



peer-review scams
plagiarism

Assessing scientific impact is crucial for...



Graphics: Designed by macrovector / Freepik

A major problem...

Useful data for research assessment:

- Publication content
- References (citations)
- Social media data (e.g., #tweets)
- Usage data (e.g., #downloads)

These data have been **isolated in data silos** of publishers or/and research organizations



Open Science - Open Access - Science 2.0

- Ongoing change:
 - Many **open science & open access initiatives**
 - BOAI <https://www.budapestopenaccessinitiative.org/>
 - cOAlation S <https://www.scienceeurope.org/coalition-s/>
 - Scientific data should be **FAIR**: Findable, Accessible, Interoperable, Reusable
 - Most research is **publicly funded** → the results should be open



Scientific impact has many aspects

Many **diverge indicators/metrics** have been proposed in the literature.

Each captures a slightly or completely **different impact aspect**.

We will focus on **citation-based metrics**.

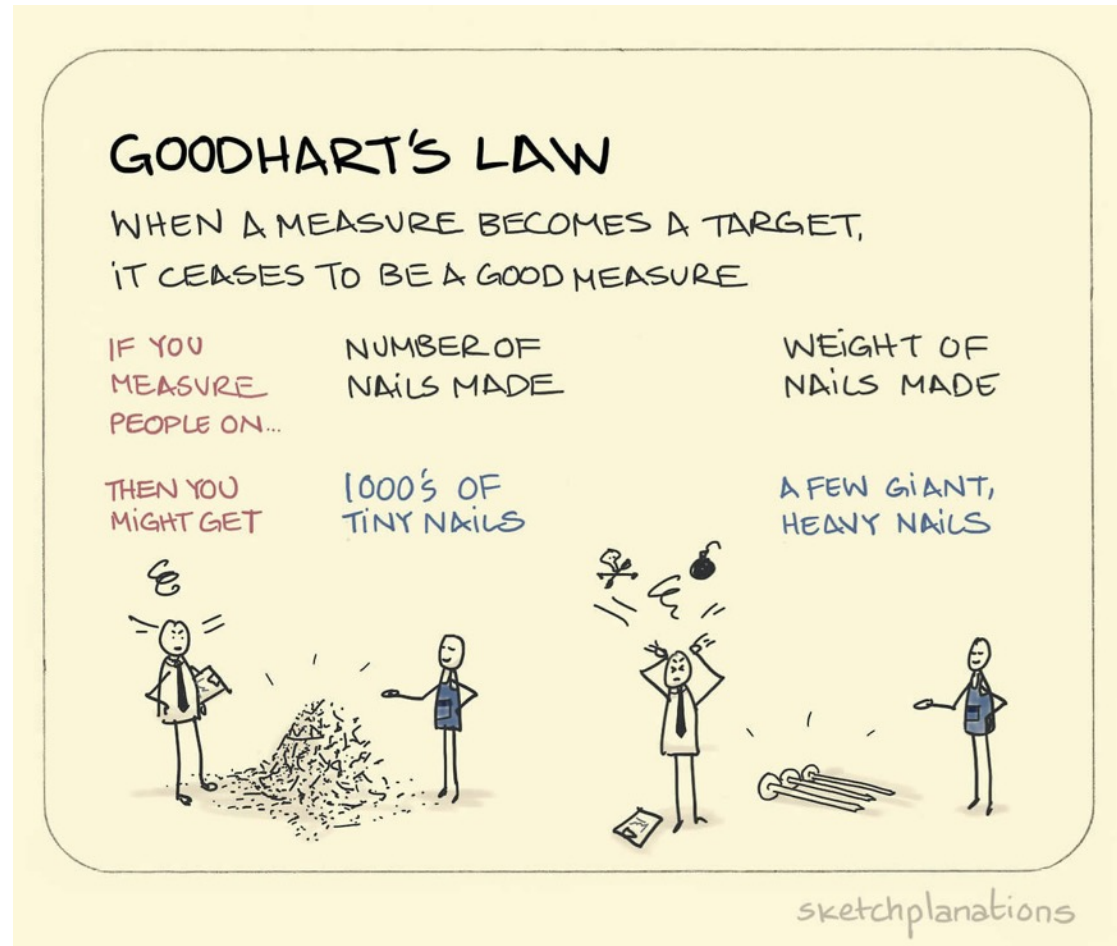


No silver bullets

- Naïve to believe that **one metric captures the “holistic impact”**.
- The impact of a paper has **many different aspects** ^[6]
 - **Popularity**: Short-term impact (having a “hype”)
 - **Influence**: Long-term impact (being “fundamental” for a discipline)
 - **Social impact**: Having hype in social media (e.g., altmetrics)
- Different aspects may be more important for different users.
 - Different algorithms **may capture better different impact aspects**.



Don't forget Goodhart's law!

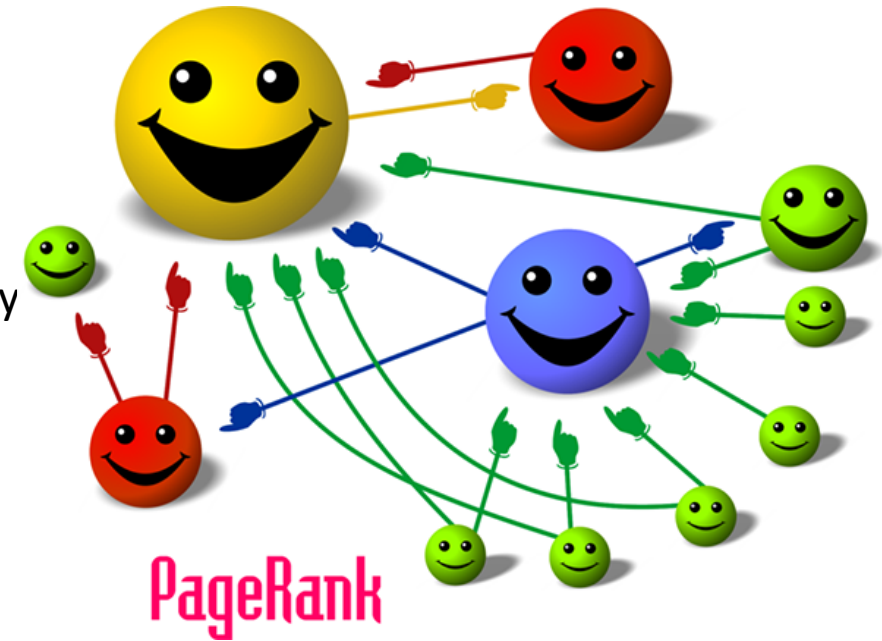


Measuring impact

■ How?

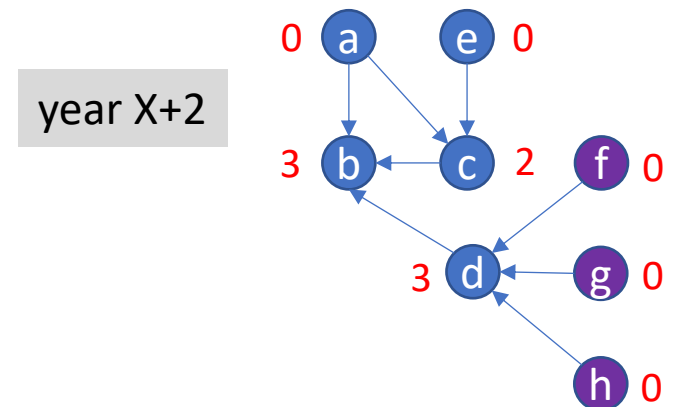
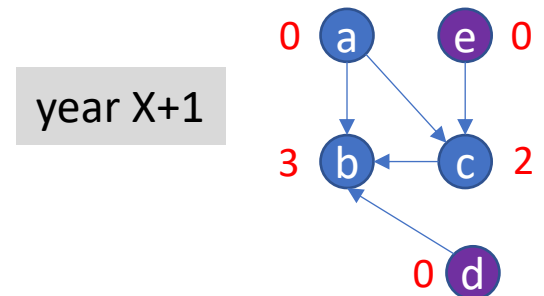
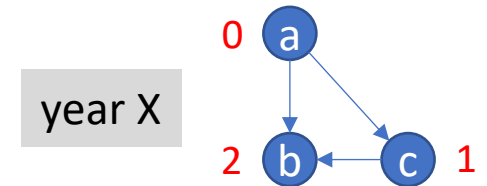
- Use **citation count** (CC) to measure impact
 - Not all citations equally important
- **Citation networks** ~ **Web** (page=paper, link=citation)
 - Use **PageRank** (PR) instead (link analysis – measures centrality)
 - Considers the impact/importance of those citing a paper

$$s_i = \alpha \sum_j \underbrace{P_{i,j}}_{1/\text{out_deg}(j)} s_j + \underbrace{(1 - \alpha)}_{\text{random jump probability}} \underbrace{v_i}_{\text{landing probability}}$$



Bias against recent papers

- Centrality-based impact (CC,PR) inserts **bias against recently published papers** ^[7,8,9]
 - New papers have almost **zero in-degree**.
 - Citations to these papers will appear after months or even years (**citation lag**). ^[10,11,12,13]
 - However, these “late” citations indicate **contemporary impact**, not future impact (papers being read now)
- In this case, **network evolution is much slower than in the case of the Web** making the “bias-against-recent-nodes” problem more crucial.
 - Ranking algorithms trying to alleviate this issue have been proposed in the last years
 - Large room for improvements (as you will see)



Existing approaches

23+1 methods

7 approaches

Method	Basic PR variants	Time Aware		Metadata		Multiple Networks	Ensemble	Other
		Network Matrix	Landing Probability	Venue	Author			
Non-Linear PageRank (NPR) [50]	✓							
SPR [53]	✓							
SCEAS [41]	✓							
Focused PageRank [31]	✓							
Weighted Citation (WC) [49]		✓		✓				
Retained Adjacency Matrix (RAM) [19]		✓						
Timed PageRank [51,52]		✓		✓	✓			
Effective Contagion Matrix (ECM) [19]		✓						
NewRank (NR) [13]		✓	✓					
NTUWeightedPR [11]		✓	✓	✓	✓			
EWPR [34]		✓		✓	✓			✓
CiteRank (CR) [45]			✓					
FutureRank (FR) [40]			✓		✓	✓		
YetRank (YR) [24]			✓	✓				
Wang et al. [47]			✓	✓	✓	✓		
PopRank [36]						✓		
MutualRank [27]						✓		
NTUTriPartite (WSDM) [17]				✓	✓	✓	✓	
NTUEnsemble [9]		✓	✓	✓	✓	✓	✓	
bletchleypark [22]		✓		✓	✓		✓	
ALEF [48]					✓		✓	
S-RCR [38]								✓
Citation Wake [29]								✓

[clear all](#)

Influence

☐ Exceptional (Top 0.01%)
☐ Substantial (Top 1%)
☒ Average (All)

Popularity

☐ Exceptional (Top 0.01%)
☐ Substantial (Top 1%)
☒ Average (All)

Start Year

Starting Publication Year

End Year

Ending Publication Year

Venue ?

Select Venues

FILTERS



BIP! Finder

neural networks

Order by: ☒ Popularity ☐ Influence ☐ Year

Combine with keyword relevance: ☒ Yes ☐ No

148781 results (7440 pages)

« 1 2 3 4 5 »

💡 Click on entries for comparison

		Venue	Year	Impact
Gradient-Based Learning Applied to Document Recognition <i>i</i>	context <i>👁</i>	Proceedings Of The IEEE	1998	<i>🔥</i> <i>🏛</i> <i>🔖</i>
Reducing the Dimensionality of Data with Neural Networks <i>i</i>	context <i>👁</i>	Science	2006	<i>🔥</i> <i>🏛</i> <i>🔖</i>
Deep learning in neural networks: an overview. <i>i</i>	context <i>👁</i>	Neural Networks	2015	<i>🔥</i> <i>🏛</i> <i>🔖</i>
Mastering the game of Go with deep neural networks and tree search. <i>i</i>	context <i>👁</i>	Nature	2016	<i>🔥</i> <i>🏛</i> <i>🔖</i>
Extreme learning machine: Theory and applications. <i>i</i>	context <i>👁</i>	Neurocomputing	2006	<i>🔥</i> <i>🏛</i> <i>🔖</i>
Human-level control through deep reinforcement learning. <i>i</i>	context <i>👁</i>	N/A	2015	<i>🔥</i> <i>🏛</i> <i>🔖</i>
Dermatologist-level classification of skin cancer with deep neural networks <i>i</i>	context <i>👁</i>	Nature	2017	<i>🔥</i> <i>🏛</i> <i>🔖</i>

Website: <https://bip.imsi.athenarc.gr/>

Twitter: [@BipFinder](https://twitter.com/BipFinder)



BIP! comparison

Title	Venue	Year	Impact		
● Fast Pattern Matching in Strings ⓘ	SIAM J Comput	1973	🔥 🏛️	🔖	✕
● BRIEF: binary robust independent elementary features ⓘ	European Conference On Co (...)	2010	🔥 🏛️	🔖	✕

Impact aspects and other metrics

Citations per year ⓘ





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That's all Folks!

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