

Can Open Access help fight fake news?





Scientific articles - a huge source of knowledge but many locks







Our purpose: to ease the reusability of scientific results in society as a whole!

What we do

Open Access Publishing





CC-BY Licence



APC Free



Open Source (OJS)

Opscidia - SciTA **Scientific Text Analysis**



SaaS platform



Concept detection



Smart navigation



Standard dashboards



Links to research articles

Projects



Data fusion



Weak signal detection



Personalized dashboards



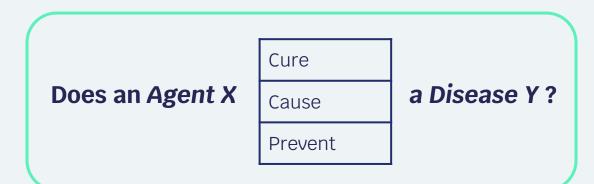
Automated flux







Our "Science-checking" prototype



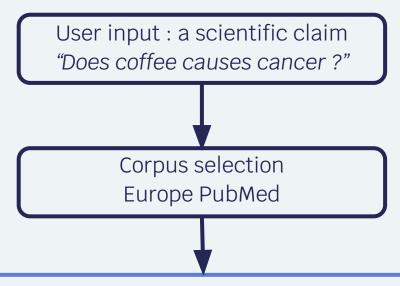
Based on FPMC data



Objectives

- Demonstrate the use of Open Access to fight fake news
- Educate people There is no "debate killer study"
- Help users identify the scientific consensus (or lack thereof)

Our appraoch



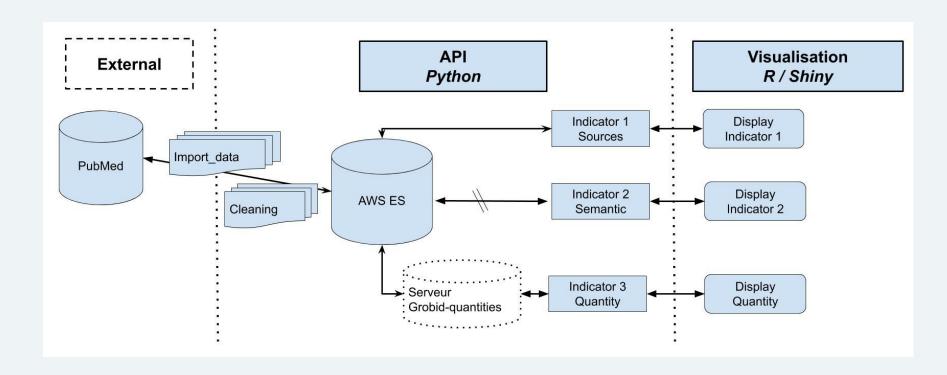
Calculation of indicators with NLP algorithms

Timeline of the sources

Semantic analysis

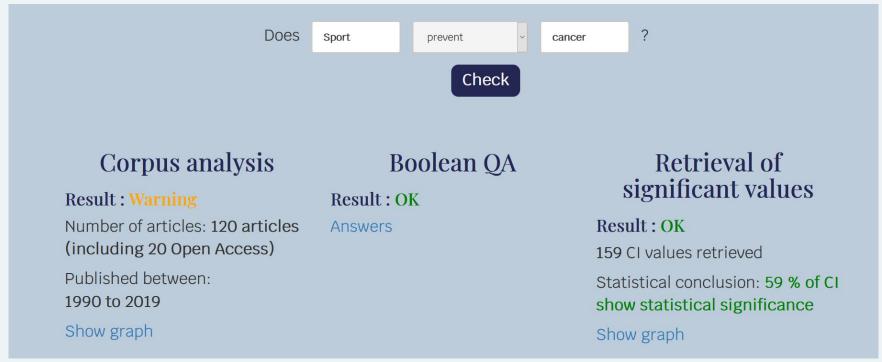
Numerical conclusions

Integrated pipeline - General architecture



Science Checker

What does the scientific consensus say about it?



Indicator 1: analysis of the sources

A standard method to fight fake news

In standard "fake news" detection:

- <u>difficult</u> to know whether a statement is true
- Easier to find where it comes from
- Sometimes enough to tell the whole story!

Examples: Flying saucer, a riot looks like a riot, project Reveal

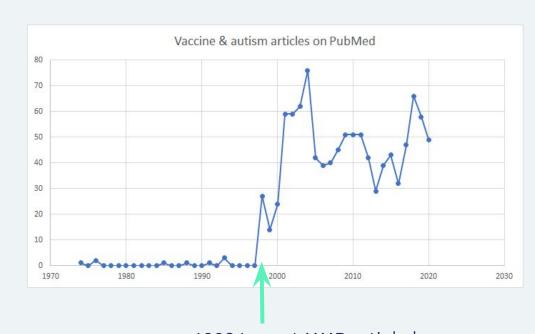
In the scientific field:

Example: Vaccines & autism

- We know where it comes from...
- Realize the massive amount of research resources to counter a successful (but retracted) article

Conclusion from this indicator:

- Is it still discussed
- When did it start?
- Is it widely covered



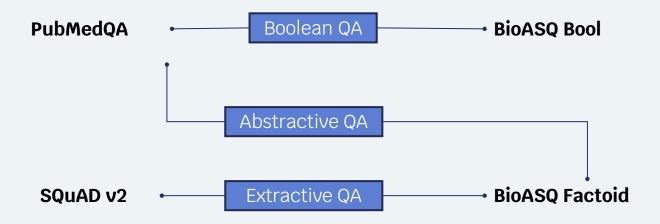
1998 Lancet MMR article by Andrew Wakefield - **retracted**

Indicator 2 : Semantic analysis

Semantic classification

Task : classify research articles wrt the input statement

- Supporting? Neutral? Contradicting?



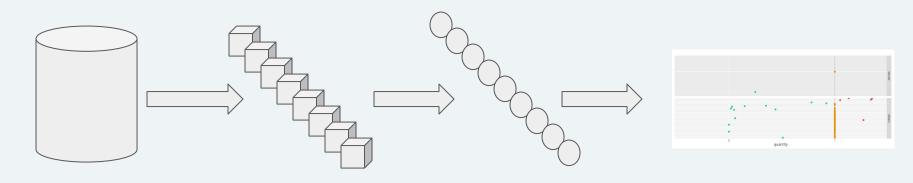
Results

Modèle	Transformer	Données	F1 - score	Accuracy
Boolean	RoBERTa _{BASE}	PubMedQA	77,98	97,01
	RoBERTa _{LARGE}		82,12	98,36
Extractive	BERT _{LARGE}	SQuAD	77,26	
	RoBERTa _{BASE}		74,02	92,64
Abstractive	T5 _{SMALL}	PubMedQA		98,33
	T5 _{BASE}			98,80

Results on test BioASQ Bool sample

Indicator 3: Numerical data retrieval

Numerical data retrieval



Copus of relevant articles

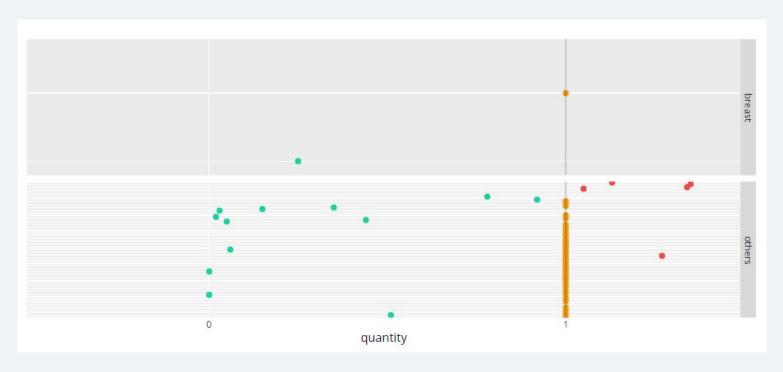
Identification of the right sentences

Extraction of the numbers

Visualization of the results

Numerical data retrieval

Does sport prevent cancer?



Conclusion and next steps

- We have built a pipeline based on 3-indicators
 - To detect scientific consensus
 - To help the public understand what is a research article
- Open Access can be useful to fight fake news
- Next step : online integration (prototype level)
- Looking for collaboration & funding to go further

Aknowledgements



For the good work!
Charles Letaillieur, Loic Rakotoson, Timothée Babinet



For funding! The Vietsch Foundation

Thank you!



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