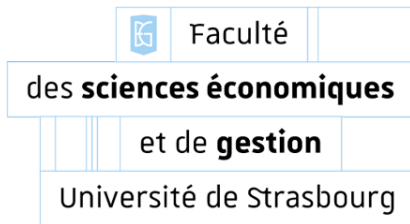


Novelty Components of Scientific Productions

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Data Visualization



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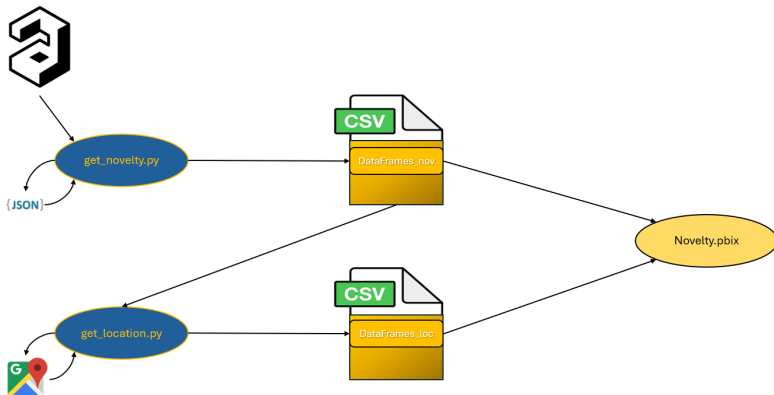
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- 1 Novelty Indicator
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Introduction

Our Program Diagram



⇒ Github

Co-occurrence Matrix

- The co-occurrence matrix represents how frequently two references (journals) appear together in a scientific article.
- Each cell (i, j) contains the number of times journal i and journal j are cited together.

| Reference | Journal A | Journal B | Journal C | Journal D |
|-----------|-----------|-----------|-----------|-----------|
| Journal A | 10 | 3 | 2 | 0 |
| Journal B | 3 | 15 | 1 | 1 |
| Journal C | 2 | 1 | 12 | 0 |
| Journal D | 0 | 1 | 0 | 8 |

Table: Example of a co-occurrence matrix

\Rightarrow `get_novelty.py`

Geocoding

\implies *get_location.py*

Limitations of the Novelty Indicator

- **Dependence on citation patterns:** The indicator relies heavily on the availability and accuracy of citation data.
- **Field-specific dynamics:** It may not account for differences in citation practices across disciplines.
- **Temporal lag:** It only reflects novelty after a reference has been cited, potentially overlooking emerging ideas not yet widely cited.
- **Computational limitations:** In very large datasets, constructing co-occurrence matrices and calculating frequencies can be resource-intensive.
- **Simplistic measure of novelty:** The indicator may not capture the full spectrum of creativity or innovation in the research being analyzed.