Two scholarly publishing cultures? Open access drives a divergence in European academic publishing practices

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Results showed significant differences in publishing practices, with a clear and robust separation of EU countries into two groups with distinct scholarly publishing cultures.

INTRODUCTION

One of the mainstays of evaluating the performance of universities is their performance in research, and a major plank of that evaluation is constituted by publication in academic journals. Likewise, the evaluation of individual academics follows similar processes. We have previously discussed the controversies involved in using publication metrics in academic journals to evaluate individual academics [1] and many of those issues apply to the use of publication in academic journals apply to the evaluation of universities.

Incentives shaping researchers publication strategies include peer pressure, institutional expectations, and national science systems. Universities and funding agencies often reward high-impact journal publications and frequent output, while competition for recognition drives researchers to conform. At the same time, open science tenets, promoting transparency and collaboration, challenge these pressures, pushing researchers to balance traditional goals with broader contributions to scientific progress. For example in one study [2], the authors examined how the introduction of bibliometric evaluations led researchers to change their behavior, focusing more on self-citations and strategic citations to improve their evaluation scores. This collective shift in behavior demonstrates how policy changes can significantly influence academic norms, leading to a new equilibrium where citation gaming becomes the dominant strategy to achieve career advancements.

The current system of scholarly publishing is often criticized for being slow, expensive, and not transparent. The rise of open access publishing, broader open science initiatives together with calls for research assessment reforms are the results of these criticisms. The emergence of new open access publishers presents a unique opportunity to empirically test how universities and countries respond to shifts in the academic publishing landscape.

These new actors challenge traditional publishing models, offering faster review times and broader accessibility, which could influence strategic publishing decisions. Researchers may choose open access venues to increase visibility and compliance with open science mandates, while universities and national science systems might adapt their evaluation criteria, balancing prestige with the growing importance of transparency and public access. This dynamic provides fertile ground for studying how institutions adjust their incentives and how these shifts affect researchers publishing strategies.

To show how two distinct cultures can emerge from a single culture within the realm of scholarly publishing, we use a game theory approach, a highly stylized approach to model very complex socio-dynamic phenomena by simplifying them to a set of rules and strategies.

METHODS AND RESULTS

OpenAlex [3] is an open-access platform designed to index scientific publications. It provides access to metadata about scientific papers, journals, authors, and ultimately institutions. The OpenAlex API was utilized for simplified and automatized access to data in the OpenAlex repository. The OpenAlex API is freely available and does not necessitate an API key for utilization, making it fairly easy to use. Due to the rate limits, which are implemented to prevent abuse and ensure fair usage, we had to send requests in 30 second intervals.

We analyzed the publishing data for all universities that were ranked in CTWS Open Ranking [4]. For each institution, we fetched its data based on ROR ID [5] and publication year. This dataset was then further processed by considering the publication type and calculating the total number of publications. By aggregating ROR ID, year, and journal type, we derived new features such as the number of publications published with MDPI, Taylor&Francis, Springer Nature, Wiley, Sage, and Elsevier,

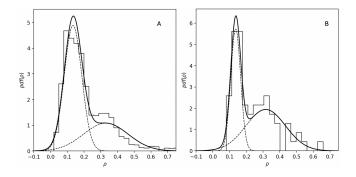


Figure 1. **Figure title.** (A) ρ distribution for 2022 at the university level, Full line shows two-gaussian mixture fit, individual gaussians are plotted with dashed lines. (B) ρ distributions for 2022 at the country level.

the number of retracted publications, the number of open access publications, and the number of gold open access publications. These features can be used to examine various indicators of publishing habits and serve as a starting point for comparing institutions publishing culture.

We selected MDPI as a representative open access publisher because it is successful, relatively new, widely used and offers journals across almost all subjects as a comparison with legacy journals published by more established publishers. All publishers now offer open access publishing, and established publishers, such as Wiley, Elsevier and Springer Nature publish a suite of hybrid journals offering both pay to view and pay to publish alongside a developing suite of open access journals. MDPI, on the other hand, offer only open access journals and compared with the Big Five (Springer Nature, Wiley, Elsevier, Taylor&Francis, Sage) publishers offer lower article processing charges (APCs) and higher rates of acceptance [6].

Here we define and focus on the ratio, ρ , between the number of publications published with MDPI and the Big Five in a given year defined as:

$$\rho = \frac{N_{\rm MDPI}}{N_{\rm Big\ Five} + N_{\rm MDPI}} \tag{1}$$

where $N_{\rm MDPI}$ is the number of publications published with MDPI and $N_{\rm Big\ Five}$ is the number of publications published with the Big Five publishers.

In figure 1 we show the ρ distribution with universities ranked in CWTS Leiden Open Ranking (panel A) for 2022. The distribution is well described by two-gaussian mixture $P(x) = \omega \mathcal{N}(x|\mu_1, \sigma_1) + (1 - \omega)\mathcal{N}(x|\mu_2, \sigma_2)$ peaked around mean values μ_1, μ_2 for ρ s with widths σ_1, σ_2 . ω and $1 - \omega$ are the weights of each gaussian peak in the mixture distribution. In panel B of Figure 1 we show the ρ distribution across countries with universities ranked in CWTS Leiden Open Ranking. The difference $\mu_2 - mu_1$ in gaussian means is increasing with time between 2019 and 2023 as: 0.116, 0.186, 0.219, 0.196, 0.199,

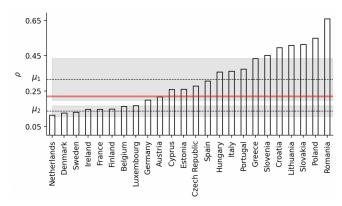


Figure 2. **Figure title.** ρ for 2022 for countries with universities ranked in CWTS Leiden Open Ranking. The thick horizontal line corresponds to the minimum of the two-gaussian mixture fit. The dotted lines are the means μ_1 and μ_2 of the two gaussians and the shaded bands are the widths σ_1 and σ_2 .

at the university level, and as: 0.086, 0.139, 0.174, 0.179, 0.232 at the country level. In both cases the separation between the two groups is increasing over time.

To see which countries comprise each of the two groups, we plotted the ρ for 2022 vs countries with universities ranked in CWTS Leiden Open Ranking in Figure 2. The thick horizontal line corresponds to minimum of the two-gaussian mixture fit and denotes the approximate separation of countries into two groups. The dotted lines are the means μ_1 and μ_2 of the mixed distribution model and the shaded bands are the corresponding standard deviations σ_1 and σ_2 of the two peaks.

The data in Figure 2 indicates a clear separation of countries into two groups: ρ_{high} , countries with a higher ρ moslty overlapping with the set of central and southeastern EU countries (Croatia, Cyprus, Czech Republic, Estonia, Greece, Hungary, Italy, Lithuania, Poland, Portugal, Romania, Slovakia, Slovenia, Spain) and ρ_{low} , countries with a lower ρ moslty overlapping with the set of central and north-western EU countries (Austria, Belgium, Denmark, Finland, France, Germany, Ireland, Luxembourg, Netherlands, Sweden).

To see if the division of countries into ρ_{high} and ρ_{low} groups is statistically significant, we performed a Mann-Whitney U test. The test showed that the mean ρ for ρ_{high} countries is significantly higher than for ρ_{low} countries with p < 0.01. Figure 3 displays the mean ρ separately for ρ_{high} and ρ_{low} countries. The gap and the increasing trend of the gap over the years in mean ρ between these two country groups is clearly visible.

We also tested the robustness of the separation of countries into ρ_{high} and ρ_{low} groups by randomly switching countries between the two groups. Figure 4 shows the mean p-values for the Mann-Whitney U test for the differences in mean ρ between ρ_{high} and ρ_{low} countries in which 0, 1 or 2 countries are randomly switched between

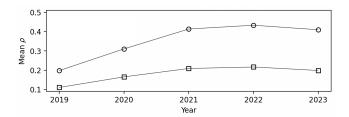


Figure 3. **Figure title.** Mean ρ vs years for ρ_{high} (open circles) and ρ_{low} (open squares) countries. Mann-Wittney U test shows statistically significant differences in mean ρ between ρ_{high} and ρ_{low} countries with p < 0.01.

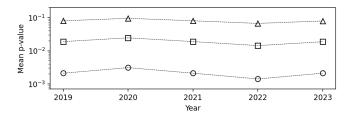


Figure 4. **Figure title.** Mean p-values for Mann-Whitney U test for differences in mean ρ between ρ_{high} and ρ_{low} countries. The p-values are calculated for 10000 random permutations of countries between the two groups in which 0 (open circles), 1 (open squares) or 2 (open triangles) countries are randomly switched between the two groups.

the two groups. The p-values are calculated for 10000 such random permutations of countries between the two groups. The mean p-values vary very little between years for each permutation regime. When only one country is switched between the two groups the division between ρ_{high} and ρ_{low} countries is still statistically significant, however when switching two countries the division is no longer statistically significant showing the robustness of the separation of countries into ρ_{high} and ρ_{low} groups.

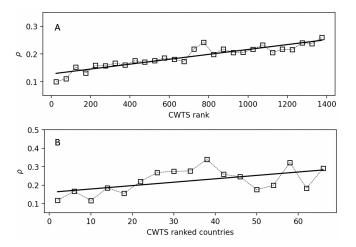


Figure 5. Figure title.

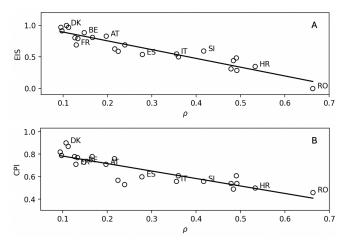


Figure 6. **Figure title.** Innovation potential, corruption perception and scholarly publishing culture

Research output measured through scholarly publications is one of key indicators in almost all university rankings so it should not be surprising that the scholarly publishing culture of universities might be closely related to their ranking. We checked this by comparing the ρ of universities with their rank in CWTS Leiden Open Ranking. In figure 5, panel A, we display the relationship between the ρ and the rank of universities included in CWTS Leiden Open Ranking table where the individual data points in the figure are the averages of ρ s of universities in bins each 50 ranks wide. This relationship can be well described by a simple linear relationship between the rank of the university and its scholarly publishing culture. In panel B of figure 5 we show the relationship between the ρ and the rank of countries of universities included in CWTS Leiden Open Ranking table.

Scholarly publishing culture of countries can also be related to their socio-economic context. Bringing together the data on innovation potential (European Innovation Scoreboard, EIS) [7] and corruption perception (Corruption Perception Index, CPI) [8] with the ρ of countries we can see how these factors are related to the scholarly publishing culture of countries. In figure 6 we show the relationship between the (normalized) EIS (panel A), CPI (panel B) and ρ ρ at a country level. For clarity only some of countries are indicated with country codes. There are clear and strong correlations between ρ and EIS and between ρ and CPI, with correlation coefficients $a = \text{Corr}(\rho, EIS) = -0.93$ and $b = \text{Corr}(\rho, CPI) = -0.86$ respectively, computed from the datasets. Since for EIS higher values indicate better innovation potential and for CPI higher values indicate lower corruption perception, the strong correlations between ρ and EIS and between ρ and CPI indicate that countries with higher innovation potential and lower corruption perception are associated with lower ρ .

The strong correlations between ρ and EIS and be-

tween ρ and CPI serve also as a validation of choosing the ρ as a measure of scholarly publishing culture in the sense that due to correlation transitivity in such case one should also observe a strong positive correlation between EIS and CPI. Indeed, the empirical correlation between EIS and CPI is Corr(EIS, CPI) = 0.87, while the theoretical lower bound for the correlation coefficient between EIS and CPI is $ab - \sqrt{1-a^2}\sqrt{1-b^2} = 0.61$.

EVOLUTION OF TWO CULTURES IN SCHOLARLY PUBLISHING

We showed that the ρ distribution in case of universities and countries is well described by two-gaussian mixed distribution which is a weighted sum of two gaussian distributions. Suppose that means and variances evolve with time much slower than the values of the weight $\omega \in [0,1]$. Therefore, the shape of P(x) will be mostly determined by the time evolution of ω . The emergence of two cultures in scholarly publishing when $\omega > 0$ or when the second peak with $\mu_2 > \mu_1$ occurs.

Changes in strategic choices are mostly driven by current publish or perish culture pervailing in academia where honest reporting of research results is often overshadowed by the need to publish in high-impact journals leading to publishing anxiety, questionable research practices, and even research misconduct.

The game involves researchers choosing between two strategies: publishing in established, legacy journals with typically slow reviewing processes (strategy B) or in new, fast-reviewing, open-access journals (strategy A). Initially, most researchers follow the strategy B, reflecting the traditional approach. However, an increasing push for open science and the appeal of rapid dissemination create incentives for adopting the strategy A.

$$u_i = P(e_i, e_j)R - e_i \tag{2}$$

where u_i is the payoff of researcher i, $P(e_i, e_j)$ is the probability, R is the reward for coordination, and e_i is the effort of researcher i. $P(e_i, e_j)$ is determined by factors such as the coordination between researchers, citation practices, visibility, academic recognition, etc.

Researchers payoffs depend on coordination with peers; citation practices, visibility, academic recognition-prestige, cost, speed

DISCUSSION

Our findings reveal significant insights into the current state and evolving trends of scholarly publishing practices among universities and EU countries. Using open

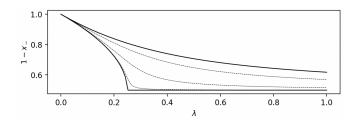


Figure 7. **Figure title.** Game theory model of scholarly publishing culture evolution. Researchers choose between two strategies: publishing in established, legacy journals with typically slow reviewing processes (strategy B) or in new, fast-reviewing, open-access journals (strategy A).

research information of scholarly publications and university ranking we showed that there is a clear bifurcation in publishing cultures in Europe.

The separation, evident at both the university and country levels, indicates a growing divergence in publication strategies, influenced by factors such as policy changes and institutional rankings. The increasing gap between ρ_{high} and ρ_{low} countries underscores the impact of regional dynamics on academic publishing. Additionally, the correlation between publishing practices and factors like innovation potential and corruption perception suggests that broader socio-economic contexts play a role in shaping research outputs.

The tendency of researchers from ρ_{high} countries and lower-ranked institutions to publish more in open-access MDPI journals, which provide faster turnaround and are often perceived as less stringent, may be explained in several ways. As publishing behavior is generally the result of a reasoned process, our findings can be interpreted through the lens of the Theory of Planned Behavior [9], which recognizes the critical role of individuals ability (i.e., perceived behavioral control) and motivation (i.e., intentions, which are heavily influenced by subjective norms and attitudes). While this theory has not yet been studied in the context of explaining researchers' decision to publish in MDPI over more traditional journals, Moksness and colleagues [10] have used it to explain intentions to publish in open-access journals.

First, in relation to perceived behavioral control, researchers from ρ_{high} and lower-ranked countries may perceive important internal and external barriers to publishing in traditional, legacy journals, such as resource constraints (including financial aspects), fewer international collaboration opportunities, language barriers, and less prestigious institutional affiliations. In fact, recent research suggests that some of these barriers objectively exist; for example, a recent study by Sverdlichenko and colleagues [11] revealed that journal editors may be influenced by author institutional affiliations when deciding whether a manuscript should be sent out for peer review. Second, regarding subjective norms, academics working

in ρ_{high} and lower-ranked countries may feel a stronger social pressure to publish frequently, potentially driven by institutional policies that still reward the quantity (instead of quality) of publications. This may lead to a snowball effect, with early adopters of such practices imposing pressure on others to do the same to remain competitive in their academic environment. Over time, publishing in MDPI over legacy journals can become completely normalized or even desirable in certain academic environments. The important role of social norms, as opposed to solely top-down regulations, in determining researchers publication choices has been previously discussed by other authors (e.g., Migheli and Ramello [12]). Lastly, several aspects may contribute to more favorable attitudes towards MDPI in ρ_{high} and lower-ranked countries. For example, they may believe that publishing in MDPI will help them achieve their academic goals, such as career advancement, quicker, or increase their visibility (due to the open-access nature of these journals).

We also found that researchers from countries with higher corruption perception are more likely to publish in MDPI over the Big Five journals. While this relationship has not yet been investigated in other countries, some parallels can be drawn from the broader literature on the characteristics of individuals who live in countries perceived as highly corrupt. For example, it is well-known that perceptions of corruption are associated with lower institutional trust [13], lower meritocratic ideology [14], and adaptive behaviors, whereby societal levels of corruption result in norm and rule violations on the individual level [15]. In the specific context of scholarly publishing, researchers may distrust the fairness of the peer-review processes and traditional publishing practices and merits in general. Moreover, environments with high corruption may sometimes be characterized by uncertain career advancement procedures and less transparent and equitable funding, motivating researchers to publish quickly to bolster their CVs and improve their chances in these ambiguous circumstances.

The observed relationship between EIS, CPI, and ρ highlights how academic publishing choices can mirror broader socio-economic and governance factors, with ρ acting as a significant indicator of these underlying dynamics. Countries with higher innovation (EIS) and lower corruption (CPI) may be more entrenched in established academic practices, favoring legacy journals, while countries with lower scores may rely more heavily on newer, open access journals, possibly due to less stringent academic systems, fewer resources, or a push for faster academic output. The choice to publish in newer, fast-review open access journals versus legacy journals might also be tied to institutional strength and governance quality.

Our findings raise concerns about the potential inequalities in scholarly publishing across Europe. The preference for MDPI publications in ρ_{high} countries and

lower-ranked universities may be influenced by factors such as publication speed and perceived ease of acceptance but could also indicate various challenges faced by researchers in these environments, such as fewer resources available for research and systemic issues that push researchers towards certain publication outlets. In line with this, governments and institutions in ρ_{high} countries should continue enhancing their support for research activities (e.g., transparent funding procedures, budgets for article processing charges) and empowering academics with better infrastructure, mentoring, and collaboration options. Moreover, extensive efforts should be dedicated to promoting transparent and fair evaluation metrics that go beyond the mere quantity of publications.

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Code and data availability. The data and the code used in this work is available at https://github.com/deankorosak/two-cultures/.

Author contributions. All authors contributed substantially to all aspects of the study.

Conflict of interest. The authors declare no conflict of interest, financial or otherwise.

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