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## An investigation on Italian researchers' collaborative habits

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### Abstract

The first step in the conventional methodology for breaking down research collaboration by field is to sort publications into subfields according to their intended audience. Using a more detailed classification of the authors' respective domains, we propose a novel technique in this work. The proposed technique provides a more accurate benchmark against which to evaluate an individual's propensity to collaborate. This research takes the new approach and applies it to all Italian university researchers in the hard sciences, assessing their propensity to work with different types of collaborators and in different contexts (intramural peers, domestic partners, and international partners). We show, using simulation, that the results greatly deviate from those obtained by conventional means.

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### Introduction

Over the last several decades, there has been a dramatic growth in collaboration aimed at advancing scientific knowledge. Co-authorship studies have revealed that the number of articles authored by a single person is declining (Abt, 2007; Udine et al., 2012). Intramural/extramural, domestic/international, interdisciplinary/interdisciplinary participation may be affected by contextual factors starting with the

research field (Gazni et al., 2012; Yossarian & Voyageur, 2004). Publications in the so-called "big scientific" domains often contain much more authors than those in other subjects due to factors such as the high cost of equipment, the need for large sample sizes, and the manner of assigning authorship (Cronin, 2001; Glance and Schubert, 2004). There may be a broad variety of activating collaboration techniques, even within a same field, due to the different areas of expertise required and the different

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cooperation tendencies of the individual scientists (Pipette & Ross, 1992). Newman (2001) and Moody (2004) both agree that. Having a good grip on the different ways in which cooperation expresses across domains and disciplines is crucial for investigating the processes at the very root of cooperation and defining the most applicable policies for its management. According to Wagner and Clydesdale (2005), this might increase the amount of research completed. This article's focus is on the collaborative efforts of academics from different disciplines. Research of this kind often begins with a classification of relevant publications. Instead, we have based our approach on the conventional hierarchies that have long existed within the scientific community. This option exists because of a quirk in the Italian schooling system. The Italian Ministry of Education and Research (MIUR) maintains an extraordinary database<sup>2</sup> of the country's academics, categorizing everyone into a single SDS (scientific discipline sector). There are 370 of these areas of study, which are divided among 14 academic divisions (Adas) in colleges and universities. Assigning authors to works allows us to see how often and in what ways scholars from different fields collaborate on studies. Using the standard method based on the classification of publications for the

same population, we can put a numerical value on the discrepancy between the two sets of results.

### **Scientific collaboration and its determinants**

In the early stages of a scientific collaboration, when it is required to enhance familiarity and establish a climate of trust among collaborators, the ability to communicate effectively, informally, and flexibly is one of the key components for growth of productive scientific collaboration (Tractor and Landry, 1997). Therefore, it is not surprising that many partnerships are launched via in-person interactions, whether they take place in the workplace, at conferences, or as part of a well orchestrated kickoff event (Lauder, 2001; Wagner and Clydesdale, 2005). In long-distance cooperation, when monitoring is more challenging, face-to-face interactions might assist to alleviate coordination issues during the implementation phase by preventing "free riding" and reducing partner dispute (Hinds and Bailey, 2003). Houseman et al. (2010), Abram et al. (2009), and Larivière et al. (2006) all find that the likelihood of collaboration decreases as the distance between the scientists' respective home organizations increases. This trend may be due to the significance of face-to-face contacts. This would also explain why scientists from different sized universities use different forms of co-authorship (Kate, 2000): those from large universities

tend to collaborate primarily with colleagues from the same university or from foreign organizations, while those from smaller universities, due to the scarcity of their own intramural colleagues and the lower "relational" value of these connections, tend to work with colleagues belonging to other domestic universities. The overall decrease in travel expenditures in recent years is most likely connected to the rise in scientific cooperation, particularly on an international scale (Houseman et al., 2010). However, the single most significant element in the noticeable growth in extramural scientific partnerships is the spread of low-cost new communications technologies that considerably minimize the qualitative distinction between remote and face-to-face contact (Cairn cross, 1997; Olsen and Olsen, 2000).

## Methodology

Studying research partnerships typically involves defining the type of partnership (intramural vs. extramural, infra-disciplinary vs. interdisciplinary, public-private vs. international, etc.), the setting (a discipline or a group of universities), and the tool (the co-authorships of the publications). Then, all the articles that may be linked to the given context are sorted according to the studied methods of cooperation. According to Gazing et al. (2012), for instance, the existence or absence of an author affiliated with a foreign organization is used to categorize the articles referable to a field as "international." The percentage of total publications in the field that are categorized as "international" provides a measure of the prevalence of international

cooperation within the field. Starting with Panamanian's (1983) "Degree of Collaboration," continuing on to Lawani's (1986) "Collaborative Index," Aquifer et al (1988) 's "Collaborative Coefficient," and finally Egg he's (1991) "Revised Collaborative Coefficient," this methodology underpins all of the principal indicators of co-authorship developed in the literature. Using a common measurement for all of your publishing data is another method for studying who wrote what. In order to assess the likelihood of scientists collaborating in the form under consideration, the single scientist is used as the primary analytical unit. For the phenomena studied by Gazing et al. (2012) once again, using individual scientists as the base analytical unit would allow for assessment of the tendency to international cooperation for the scientists that are part of a field. At least two other groups have taken a similar tack, and they are Martin-Simpered et al. (2002) and Abram et al (2011). The latter quantified Italian scholars' inclination for international cooperation by field, tallying the proportion of each scholar's total publications that were written in conjunction with foreign organization colleagues. Although limited to 93 Spanish university-based geologists, Martin-Simpered et al. determined each researcher's "degree of collaboration," which they defined as the ratio of coauthored publications to the researcher's total number of publications, and their "degree of national collaboration," which they defined as the ratio of coauthored publications with colleagues from at least one national organization to the researcher's total number of publications.

Methods of obtaining data and scope of study Our investigation relied on data from the aforementioned Ministry of Universities and Research database, which included profiles of Italian academics. The authors then take the data set of these authors' papers and pull it from the Italian Observatory of Public Research (ORP), a database they created and manage using data licensed from the Woos. By starting with the raw data of Italian publications in Woos between 2006 and 2010, and then applying a complex algorithm for disambiguate the true identity of the authors and their institutional affiliations (for details see Tangelo et al., 2011), we are able to attribute each publication<sup>4</sup> to the university scientist or scientists (full, associate, and assistant professors) that produced it, with a harmonic average of precision and recall (F-measure) equal to 96 (error of 4%). The biometric data set includes the following information for each publication: a full list of authors; a full list of authors' addresses; a sub-list of solely the academic authors, with their SDS/UDA and university affiliations.

## Indicators and methods

We will begin with a single scientist from a well-established field and compare their average propensity to collaborate in four different forms: in general, within their own institution, with researchers from other institutions in their home country, and with scientists from institutions in other countries. The first kind of cooperation is a super set of the others; it is the predisposition to cooperate in general. We build a "author-publication" matrix with m rows and n columns according to the number of active

academics and n publications. This matrix has a size of 36,211 by 197,460. Next, we link each scholar with his or her output (p) within that time frame. For each professor, we can determine the total number of collaborative publications (cp), the number of intramural (within the same institution) publications (clip), and the number of extramural (within other domestic institutions) publications (cede) because we know the total number of authors and the total number of Italian and foreign organizations involved in each publication (extramural international - cap). Using these numbers, we may calculate indications of people's varying propensities to work together, from which we can also get average inclinations by industry and specialty:

- Propensity to collaborate  $C = \frac{cp}{p}$
- Propensity to collaborate intramurally  $CI = \frac{clip}{p}$
- Propensity to collaborate extramurally at the domestic level  $CED = \frac{cede}{p}$
- Propensity to collaborate extramurally at the international level  $CEF = \frac{cefp}{p}$

## Results and discussion

The many types of co-authorship may be analyzed, and distinct UDAs and their individual SDSs can be described, using the calculated C, CI, CED, and CEF values based on the registered propensity values for respective member academics. Our results from sections 4.1 and 4.2 detail these analyses. Later, in Section 4.3, we look at how these four metrics are related to one another.

### Collaborative Tendency 4.1 Propensity to Work Together in Different Fields

Academics from the different UDAs studied had varying propensities for cooperation in all forms, including intramural, extramural domestic, and international partnerships. We give a table for each kind of partnership, illustrating, on a per UDA basis: I) the proportion of UDA faculty members with zero collaboration propensity; II) the percentage of UDA faculty members with maximum (100%) collaboration propensity; III) the average value of the UDA faculty's collaboration propensity. The Kruskal-Wallis test (Kruskal and Wallis, 1952) is applied to all of the UDAs, and the Mann-Whitney U test (Mann and Whitney, 1947) is used to confirm the differences in the inclinations recorded for the academics who belong to each field. These non-parametric tests allow us to see whether there is a greater or lesser tendency for academics to work together in one UDA compared to another. Using the `kruskal.test` and `wilcox.test` functions, we do this analysis, and the findings (available in Supplemental Material - S1) demonstrate that almost all of the comparisons we made had a high level of significance. Conclusions allow for grouping UDAs according to their varying degrees of cooperation. The values of cooperation propensity are shown in Table 2. These seem extraordinarily high, which is consistent with other previous studies using alternative methods demonstrating that the proportion of co-authored articles within the "biometric" fields is currently over 90%. (Abe, 2007; Gazing et al., 2012). Table 2 shows that many UDAs do not vary much from one another in their inclination to cooperate, despite the fact that these differences are typically statistically significant based on the findings of the

Mann Whitney U test. The average willingness to work together approaches 100% in the fields of medicine, agriculture and veterinary medicine, biology, and chemistry. All in all, these findings corroborate those of Haiti and Hong (1997) and Gazing et al.

*Table 2: Propensity to collaborate, per UDA (percentage values)*

UDA	Mean C	% C = 0%	% C = 100%
Medicine (MED)	99.4	0.1	94.8
Chemistry (CHE)	99.2	0.1	94.8
Agricultural and veterinary sciences (AVS)	99.1	0.2	95.7
Biology (BIO)	99.1	0.1	94.4
Earth sciences (EAR)	97.6	0.7	90.6
Industrial and information engineering (IIE)	97.1	0.5	85.5
Pedagogy and psychology (PPS)	96.7	1.4	89.8
Physics (PHY)	96.6	1.0	81.5
Civil engineering (CEN)	94.3	1.7	81.5
Mathematics and computer sciences (MAT)	89.1	3.3	68.6
Economics and statistics (ECS)	84.0	8.3	70.1
Total	97.2	0.9	89.0

When comparing intramural cooperation, the disparities between the different UDAs seem to be considerably more evident. Table 3 demonstrates a disparity of about 40% between the UDA with the highest value (Chemistry) and the one with the lowest value (Physics) (Economics and statistics). Similar to what was shown using the Mann-Whitney U test, the probability of collaborating with colleagues from the same institution is rather high in the four UDAs of Medicine, Agricultural and Veterinary Sciences, Biology, and Chemistry. This finding makes sense when you take into account the fact that professors in these fields often work in laboratories owned by their own institution, which are sometimes shared for budgetary reasons between other colleagues, encouraging the growth of cooperation. When it comes to industries, Industrial and information engineering have the second greatest

tendency for this kind of cooperation. This finding makes sense when you think about the fact that, like many other academic fields, this one relies heavily on shared resources like labs, equipment, and software between faculty members at the same institution, making it easier to foster teamwork. In addition, many engineering studies are the end result of research projects commissioned by businesses and carried out by academics, who typically collaborate with their peers at the same institution rather than those at other institutions to save money on travel, time, and other overhead costs while increasing their reach throughout the region.

Table 3: Propensity to intramural collaboration, per UDA (percentage values)

UDA	Mean CI	% CI = 0%	% CI = 100%
Chemistry (CHE)	83.5	2.4	46.1
Industrial and information engineering (IIE)	82.2	3.9	46.9
Agricultural and veterinary sciences (AVS)	81.2	4.3	51.8
Medicine (MED)	81.1	3.6	45.9
Biology (BIO)	78.8	4.2	45.8
Civil engineering (CEN)	73.4	8.8	46.3
Physics (PHY)	66.7	8.7	29.2
Earth sciences (EAR)	62.0	11.4	31.1
Pedagogy and psychology (PPS)	59.6	18.2	35.8
Mathematics and computer sciences (MAT)	54.1	20.5	25.4
Economics and statistics (ECS)	44.0	36.0	26.7
Total	75.4	7.2	42.3

One way to categorize extramural partnerships is by whether or not the extramural organization is located inside or outside of the same country as the collaborating institution. Average propensities to work with scientists from different domestic organizations outside of the institution are shown in Table 4. Once again, disparities across fields are quite large, with almost half a century separating the UDAs with the highest (Physics) and lowest (Social Studies) average propensities, respectively (Industrial and information engineering).

Table 4 shows the national average UDA's propensity for extramural cooperation (percentage values)

UDA	Mean CED	% CED = 0%	% CED = 100%
Physics (PHY)	72.5	5.6	24.9
Medicine (MED)	62.4	8.2	20.6
Earth Sciences (EAR)	58.6	13.1	23.3
Biology (BIO)	57.4	9.8	17.5
Chemistry (CHE)	49.8	8.3	9.3
Pedagogy and Psychology (PPS)	48.5	26.3	22.2
Agricultural and veterinary sciences (AVS)	47.1	17.3	13.9
Economics and Statistics (ECS)	38.0	38.9	19.1
Mathematics and Computer Sciences (MAT)	33.6	33.3	10.7
Civil engineering (CEN)	26.0	44.3	8.2
Industrial and information engineering (IIE)	24.8	33.1	5.3
Total	50.3	17.0	15.7

According to the findings about the inclinations to interact in various ways (Section 4.1), in certain UDAs, academics have a tendency to work with scientists from both their own institution and from other organizations. This trend is most pronounced in areas of study where a significant amount of institutional support is needed or when collaboration with other institutions is essential. Sometimes, scientists may prefer one kind of cooperation over another due to the varying degrees of coordination necessary under each. The trend toward publications with a small number of authors may also influence the selection of a single mode of cooperation in particular fields. The Spearman non-parametric correlation between each professor's results on the four indicators C, CI, CED, and CEF is calculated using the R record function (R Development Core Team, 2012) to assess the connections between the various collaboration propensities. Table 8 displays the findings, allowing one to assess the degree to which the four types of cooperation are associated both globally and for each UDA.

Table 8 shows the association between markers of collaboration propensity using the Spearman test, as reported by UDA.

UDA	C-CI	C-CED	C-CEF	CI-CED	CI-CEF	CED-CEF
AVS	0.20***	0.10***	0.05**	-0.35***	-0.25***	0.00
BIO	0.23***	0.12***	0.02	-0.21***	-0.25***	-0.03
CEN	0.43***	0.14***	0.10***	-0.44***	-0.25***	0.04
CHE	0.25***	0.12***	0.04*	-0.28***	-0.21***	-0.01
EAR	0.21***	0.21***	0.11***	-0.28***	-0.25***	0.08**
ECS	0.41***	0.35***	0.27***	-0.20***	-0.23***	-0.03
IIE	0.38***	0.12***	0.06***	-0.44***	-0.32***	0.15***
MAT	0.43***	0.25***	0.20***	-0.32***	-0.30***	0.03
MED	0.17***	0.10***	0.02	-0.29***	-0.26***	0.07***
PHY	0.33***	0.34***	0.20***	-0.02	-0.04	0.37***
PPS	0.24***	0.21***	0.14***	-0.30***	-0.30***	-0.03
Total	0.36***	0.21***	0.08***	-0.21***	-0.27***	0.12***

Significance level: \*\*\*  $p < 0.001$ ; \*\*  $p < 0.01$ ; \*  $p < 0.05$

As the statistics reveal, there is a positive relationship between C and every other metric of interest. This finding holds true across all UDAs, albeit to varying degrees, and hints at how scientists who embrace various types of cooperation show a stronger propensity to cooperate in general. With such wide variation across UDAs, it might be difficult to make sense of the correlation between C and CEF, or the tendency to interact at the international level. In Economics and statistics, in particular, there is a favorable and statistically significant association between academic collaboration and the activation of relations with foreign organizations (Table 4, which shows that only 60% of academics in this field communicate).

## Conclusions

Several scholars have taken an interest in studying the various types of research cooperation in an effort to both observe and speculate about disparities in collaboration patterns across fields of study and

professional fields. Until date, counting articles has been employed as an indicator of prevalence. The authors here instead argue for a fresh methodological approach that uses the individual researcher as its basic building block. This approach has numerous advantages, one of which is that it makes it easier to investigate collaborative efforts across disciplines. Since productivity, apart from collaboration intensity, is not distributed in a homogeneous fashion (the real-world situation), the proposed method allows for a more accurate depiction of researchers' propensity to collaborate in various forms, whether with direct colleagues or with other organizations. Implementing trustworthy collaboration measurement systems is crucial for accurately defining the ex ante and controlling the ex post circumstances for different types of cooperation within any reference framework. It's no wonder that many nations have policies in place to promote international scientific cooperation, given the positive impact it may have on a nation's capacity to produce and distribute new knowledge. The policy's influence may be proven on the actors who are ultimately the policy's purpose since the tendency of the individual scientist to participate can be quantified. Further, by collecting information on individual scientists, it is possible to generate a quantitative measure of the tendency to cooperate for the research group and the organizational unit at higher levels, which may then be the focus of targeted policy. When it comes to supporting the execution of policy that aims to influence scientific partnerships, our approach gives more suitable measurements than those previously given in the literature. Our approach, applied to the work of



Italian academics, allowed us to measure the degree to which they are willing to collaborate across disciplines on their research. The results may be used to assess the effectiveness of current efforts at particular institutions and throughout the research economy, and to guide future policy choices about how to best stimulate cooperation between academics working in different fields.

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