

# Scientists connected with society are more active academically

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## Abstract

Today, most scientific institutions acknowledge the importance of opening the so-called “ivory tower” of academic research through popularization, industrial innovation or teaching. However, little is known about the actual openness of scientific institutions and how their proclaimed priorities translate into concrete measures. This paper helps getting an idea on the actual practices by studying three key points: the proportion of researchers who are active in dissemination, the academic productivity of these active scientists, and the institutional recognition of their activity in terms of careers. This paper answers these questions by analyzing extensive data about the academic production, career recognition and teaching or public/industrial outreach of several thousand CNRS scientists from many disciplines. We find that, contrary to what is often suggested, scientists active in dissemination are also more active academically. However, their dissemination activities have almost no impact (positive or negative) on their career.

## 1 Introduction

Researchers and academic institutions seem to have admitted the importance of establishing strong ties between science and society. In the United Kingdom, Martin Rees, president of the Royal Society, points out that “Researchers need to engage more fully with the public. The Royal Society recognizes this, and is keen to ensure that such engagement is helpful and effective”. A recent survey carried out by the Royal Society finds that “Most researchers have highlighted that social and ethical implications exist in their research, agree that the public needs to know about them, and believe that researchers themselves have a duty, as well as a primary responsibility, for communicating their research and its implications to the non-specialist public.” [Society (2006)]

In France, in the document supposed to steer his long-term policy, the “Multi-year action plan” [CNRS (2004)], the CNRS declares that one of the six top priorities is “to transfer research results to industries” and another “to strengthen the relations between science and society”. In February 2007, CNRS organized an official workshop on “Science and Society in transformation”, in presence of many CNRS officials [Allix (2007)]. This attitude seems to be shared by the majority of

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researchers: in her study on the attitudes of researchers towards popularization [Cheveigné (2007)], Suzanne de Cheveigné concluded: “All interviewed researchers unanimously declared: popularization is now a key and unavoidable component of research work.” Motivations provided by researchers are numerous: the yearning to inform the public, to make one’s field of research better known and encourage students to take up science, or the need to account to civil society for the use of funds provided to laboratories.

The reality on the field is generally aloof from these generous ideas. For example, in the CNRS report for candidating to the “Directeur de Recherche” (Senior Scientist) position, a mere 9 lines are provided to summarize twenty years of research dissemination. Likewise, the Royal Society survey concludes that, for most scientists, “research is the only game in town”, and popularization has to be done after one is through with “real” work.

The purpose of this paper is to obtain an empirical picture of dissemination practices in CNRS. We have presented in a previous study [Jensen and Croissant (2007)] a statistical view of scientists involved in popularization. Here, we also include data on teaching and industrial collaborations. Moreover, we correlate these data with scientists’ academic activity, as quantified by bibliometric records. Therefore, we are able, for the first time, to answer two important questions about scientists active in dissemination: are they “bad scientists” as some scientists suggest [Society (2006)]? Do they get any institutional recognition in terms of careers ? We answer these questions by analyzing extensive data about the academic production, career recognition and teaching or public/industrial outreach of several thousand CNRS scientists from many disciplines.

## 2 Methodology

Thanks to the CNRS Human Resources Direction, we have gathered data on the dissemination activities (public outreach, industrial collaborations and teaching) of the 11 000 CNRS scientists over a three-year period (2004-2006). It should be noted that these data are *declared* by scientists in their annual report (“Compte Rendus Annuels des Chercheurs” or CRACs). This annual report is not judged very important for the career, serious evaluations taking place only when scientists candidate to senior positions. However, filling out the report is mandatory and most researchers (over 90% each year) do fill it in due time. Many reasons could lead to some underestimation of the amount of activities declared, including fear of misperception of popularization activities by committees, laziness to report faithfully these minor activities, etc. Inversely, lack of control of these items could favor some overreporting of dissemination activities, although this is not likely since they have almost no perceived impact on career. Hence, we could anticipate some underestimation of dissemination activities in the figures below.

“Popularization” activities include public or school conferences, interviews in newspapers, collaboration with associations... Clearly, there is no entirely satisfactory definition of popularization. As Stephen Hilgartner convincingly shows [Hilgartner (1990)], there exists in fact a continuous gradation going from technical literature to popular science, with no clear cut indicating where popularization begins. Here, popularization actions are declared by scientists themselves, according to the following operational criterion : popularization means wide audience, actions aiming at non specialized public. For more details, we refer the reader to Appendix 1 and reference [Jensen and Croissant (2007)]. “Industrial

collaborations” mainly means contracts with industrial partners or funding from non academic sources (regional, specific programs ...). For natural sciences, contracts with industrial partners dominate, while funding from specific programs accounts for most of such actions in the social and human sciences. Both these types of actions account for 81% of the actions, while patents represent 16.5% and licenses 2.5% of the actions. “Teaching” is only characterized by the annual number of hours dedicated to this activity. CNRS researchers have no teaching duties.

We have described previously [Jensen, Rouquier, and Croissant (submitted, 2007)] how we manage to obtain a large but robust database of bibliometric indicators for the CNRS scientists. Briefly stated, our method uses the “Author search” of [Web of Science] on the subset of 8750 scientists having filled out the CNRS report the last three years. We exclude researchers in Social Sciences (their bibliographic record is not well documented in WoS) and in High Energy Physics (too few records in the CNRS database), leading to 6900 names. After filtering records suspected to be erroneous, we obtain a database of 3659 scientists with reliable bibliometric indicators, as checked by close inspection of several hundred records and good prediction of scientists promotions. A more detailed description of our method is given in Appendix 2 and reference [Jensen, Rouquier, and Croissant (submitted, 2007)].

We have used several bibliometric indicators as proxies for academic activity: number of papers published, number of papers published per year, number of citations or Hirsch index [Hirsch (2005)]. It could be argued that  $h$  is not a good measure when comparing the scientific activity of researchers with very different career lengths, because it automatically increases with time. A more relevant measure might be  $h$  divided by the career length in years  $h_y$  [Hirsch (2005)], although we have shown that it is not perfect either, since its average value for CNRS scientists decreases with scientist’s age [Jensen, Rouquier, and Croissant (submitted, 2007)]. However, since  $h_y$  is closer to a constant than  $h$  for scientists with different career lengths [Jensen, Rouquier, and Croissant (submitted, 2007)], we will use it, along with the average number of papers published per year (which is indeed constant over scientists’ ages) and other bibliometric indicators.

### 3 Proportion of active scientists

A summary of the subdisciplines encompassed by our database, together with some characteristic average values, is shown in tables 1 and 2. In Figures 1 and 2, we show the proportion of active scientists by age and grade, for each of the three dissemination activities. Overall, CNRS scientists carry out more than 7 000 popularization actions and more than 4 000 industrial collaborations per year, and these figures are increasing. Table 1 shows however that the activity is very unequally distributed: over a three-year period, about half of scientists remained inactive in popularization or industrial collaborations. For more details on popularization activities, see our previous analysis [Jensen (2005), Jensen and Croissant (2007)].

These large scientific domains are in fact heterogeneous. It is interesting to study more disaggregated data, at the discipline level (corresponding to CNRS scientific “sections”). Table 2 shows in detail the proportion of active scientists in each of the subdisciplines for popularization, teaching and industrial collaborations.

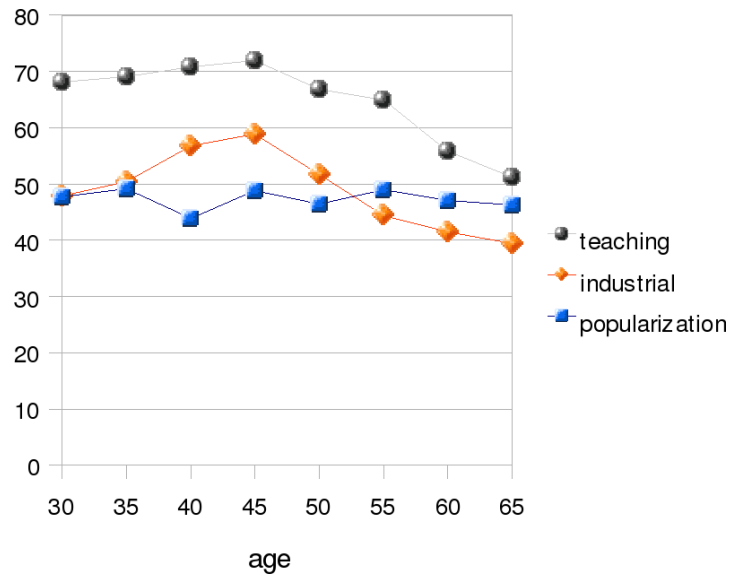


Figure 1: Evolution of the proportion of scientists active in dissemination as a function of their age. Data correspond to the whole database, i.e. before filtering with bibliometric indicators.

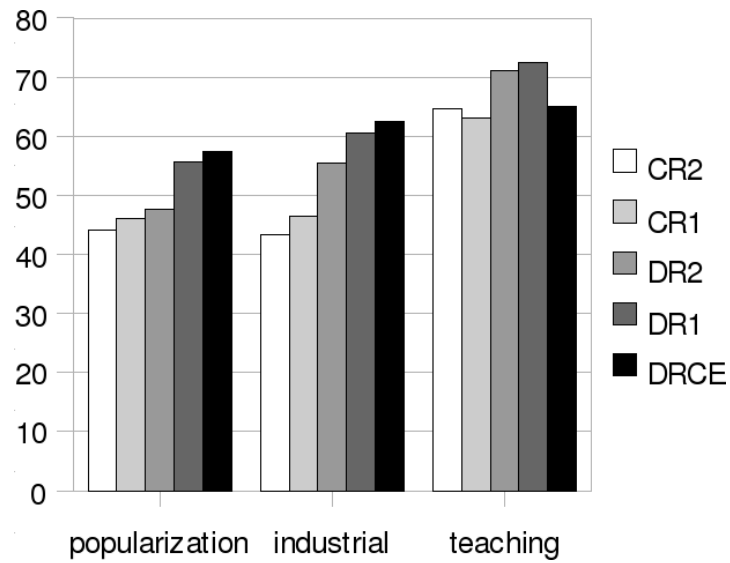


Figure 2: Evolution of the proportion of scientists active in dissemination as a function of their position. Data correspond to the whole database, i.e. before filtering with bibliometric indicators.

## 4 Academic achievement of open scientists

A large fraction of scientists view dissemination activities as a low status occupation, done by “those who are not good enough for an academic career” [Society (2006)]. This common perception is captured by the well-known “Sagan effect”: popularity and celebrity with the general public are thought to be inversely proportional to the quantity and quality of real science being done [Hartz and Chappell (1997)]. Sagan’s biographers [Shermer (2002)] have shown that Harvard’s refusal of Sagan’s bid for tenure, and the National Academy of Science’s rejection of the nomination of Sagan for membership, was a direct result of this perception. By analyzing his publication record, they have also shown that there is no such effect: “Throughout his career, which began in 1957 and ended in December 1996, upon his untimely death, Sagan averaged a scientific peer-reviewed paper per month. The ‘Sagan Effect’, at least when applied to Sagan himself, is a Chimera” [Shermer (2002)]. In the following, we will test on a larger scale whether such an effect exists for the CNRS scientists, i.e. whether scientists engaged in dissemination are academically less active than average. To anticipate our conclusion, let us say that we find exactly the opposite correlation: scientists connected with society are more active than average, for reasons we then investigate.

### 4.1 Comparing bibliometric indicators of active and inactive scientists

We begin by comparing the average academic activity of active (in dissemination) and inactive scientists. The precise question we investigate is: if we choose randomly a scientist and ask her whether she is active in dissemination, does the answer tell us something about her academic activity? According to the common view quoted above, the answer should be that an active scientist has, on average, a weaker academic activity, which should correspond to lower bibliometric indicators. Our data shows exactly the opposite effect.

#### 4.1.1 Comparing bibliometric indicators of active and inactive scientists

Figure 3 shows that activity in dissemination is correlated with higher academic indicators. Scientists inactive in both popularization and industrial collaborations (roughly 30%) have a lower academic activity ( $h_y = 0.65$ ), which still decreases for the ones also inactive in teaching (15%,  $h_y = 0.62$ ). If one uses the number of papers published per year, the conclusion is similar: the average value is 2.28, while dissemination active scientists have significantly higher average values (popularization: 2.38, p-value  $2.610^{-5}$ , industrial collaboration 2.45, p-value  $< 2.210^{-16}$ , teaching: 2.35, p-value  $810^{-6}$ ).

A potential danger of this kind of general comparison is the well-known variability of average  $h$  indexes among different scientific disciplines [Iglesias and Pecharromán (2007)]. Therefore, we calculated the differences in  $h_y$  between scientists of the same discipline but different dissemination activities. Our results (Table 5) confirm the fact that open scientists are always, on average, academically more active than the inactive ones, even if the smaller number of scientists investigated prevents the results from being statistically significant for some disciplines. Note also that  $h_y$  is not always the best indicator of academic activity: for Engineering, the average number of papers published accounts better for promotion [Jensen, Rouquier, and Croissant (submitted, 2007)]. Taking this indicator, differences between active and inactive scientists become significant in favour of active scientists (for example, popularization active scientists from Engineering

Table 1: Percentage of inactive (no action or no teaching respectively), active (less than 10 outreach actions or less than 4 industrial collaborations or less than 210 teaching hours respectively) and very active scientists for the CNRS scientific domains. This division in subpopulations is more instructive than the mean number of actions, as the activity is very unequally distributed among researchers: the 5% most active account for half of the actions [Jensen (2005)]. Figures correspond to the activity cumulated over 2004-2006. CNRS researchers have no teaching duties.

DOMAIN	Popularization			Industrial Collaboration			Teaching		
	inactive	active	very active	inactive	active	very active	inactive	active	very active
Physical sciences	59	39	2	63	35	2	46	50	4
High Energy Physics	45	54	1	95	5	0	71	27	2
Life sciences	64	34	2	43	53	4	33	66	1
Engineering	52	46	2	19	74	7	22	69	8
Chemistry	65	34	1	39	52	9	42	55	2
Earth Sciences, Astrophysics	36	57	7	59	41	1	31	67	1
Social Sciences	27	62	10	68	32	0	17	76	8
All CNRS	53	43	4	49	47	4	33	63	4

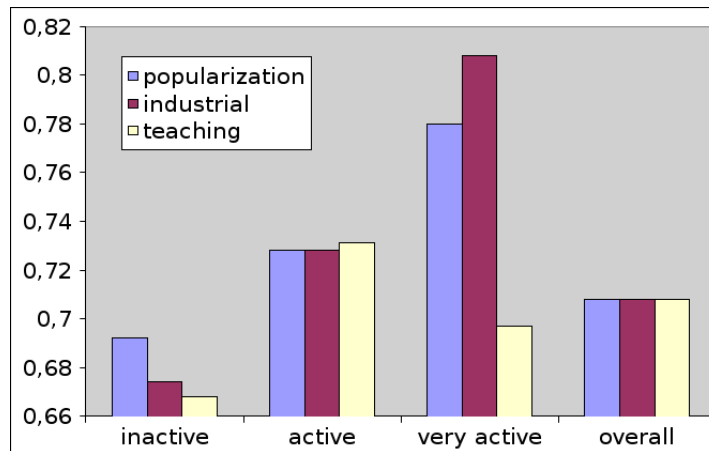


Figure 3: Average  $h_y$  for inactive, active, or very active scientists (see Table 1 for the definitions) in the different dissemination activities. Again, we exclude researchers in Social Sciences because their bibliographic record is not well documented in WoS. Variance tests on the indicators ensures that they are strongly significant (for popularization:  $F = 6.9$ , p-value = 0.01; for industrial collaborations:  $F = 18.6$ , p-value = 0.00004. For teaching, active scientists have a significantly higher  $h_y$  than the non active, p-value = 0.0003). However, contrary to dissemination, the very active ones have the same  $h_y$  than the mean (the small difference is not statistically significant). Our data point to an “optimal” value of roughly 20 - 30 teaching hours per year, additional hours lowering  $h_y$ .

Table 2: Details of Table 1 by subdiscipline. The precise names of the CNRS “sections” have been shortened for simplicity. The discipline “High Energy Physics” of Table 1 corresponds to “Interactions, particles & strings”. As in Table 1, we show the percentage of active (less than 10 popularization actions or less than 4 industrial collaborations or less than 210 teaching hours respectively) and very active scientists (i.e. more active than the previous figures) for the CNRS scientific subdisciplines. For simplicity, we have not shown the percentages of inactive (no action or no teaching respectively) scientists. These can be easily calculated from the difference to 100% of the sum of the two columns “active” and “very active”. Figures correspond to the activity cumulated over 2004-2006.

Subdiscipline		popularization		industrial		teaching	
		active	very active	active	very active	active	very active
Physical sciences	1 Mathematics	30	3	16	1	61	7
	2 Physics, theory & method	29	3	25	1	43	3
	3 Interactions, particles & strings	54	1	5	0	27	2
	4 Atoms & molecules	43	1	40	3	51	4
	5 Condensed matter: dynamics	44	3	45	2	51	3
	6 Condensed matter: structure	42	1	42	1	38	6
Engineering	7 Information science	47	1	71	3	76	6
	8 Micro & nano-technologies	44	2	72	15	59	11
	9 Materials & structure	48	2	80	3	76	14
	10 Fluids & reactants	45	4	80	5	68	6
Chemistry	11 Super/macromolecular systems	42	1	52	11	62	1
	12 Molecular architecture synthesis	29	0	47	8	54	4
	13 Physical chemistry	33	1	43	3	53	2
	14 Coordination chemistry	37	1	53	14	44	3
	15 Materials chemistry	35	1	59	12	57	1
	16 Biochemistry	28	1	52	7	66	0
Earth sciences, astrophysics	17 Solar systems & the universe	56	14	20	0	44	1
	18 Earth & earth plants	59	4	39	0	79	3
	19 Earth systems: superficial layers	56	5	52	1	64	1
	20 Continental surface	56	3	66	1	80	1
Life sciences	21 Molecular basis of life systems	27	1	58	6	61	3
	22 Genomic organization	26	1	43	2	62	2
	23 Cellular biology	25	0	50	4	63	2
	24 Cellular interaction	29	0	50	5	66	2
	25 Physiology	32	2	57	3	64	0
	26 Development, evolution	32	1	46	2	67	0
	27 Behavior, cognition & brain	59	6	63	1	78	1
	28 Integrative vegetal biology	34	0	52	4	60	0
	29 Biodiversity, evolution	54	8	60	1	81	1
	30 Therapy, pharmacology	38	1	58	12	63	1
Human & social sciences	31 Human evolution	63	16	24	0	82	5
	32 Ancient & medieval history	66	9	13	0	71	8
	33 Modern & contemporary history	65	10	19	0	75	5
	34 Languages, language & speech	54	4	32	1	75	8
	35 Philosophy	58	10	20	1	63	8
	36 Sociology	68	11	42	0	80	9
	37 Economics & management	43	6	61	1	80	7
	38 Society & cultures	69	9	26	0	78	2
	39 Environment & society	71	9	62	0	82	11
	40 Politics, power	64	19	49	1	79	15

Table 3: Percentage of active scientists in popularization, industrial collaborations or teaching as a function of their age (activity cumulated over 2004-2006, whole database including Social Sciences and High Energy Physics). CNRS researchers have no teaching duties.

age	pop	indus	teach
<30	47.7	47.9	68.1
35	49.1	50.4	69.1
40	43.9	56.7	70.9
45	48.8	58.8	71.9
50	46.4	51.7	66.8
55	48.9	44.4	64.9
60	47.1	41.5	55.8
>65	46.3	39.4	51.2

Table 4: Percentage of active scientists in popularization, industrial collaborations or teaching as a function of their position (activity cumulated over 2004-2006, whole database including Social Sciences and High Energy Physics). The different positions of CNRS scientists are, by increasing hierarchical importance: “Chargé de Recherche 2<sup>e</sup> classe” (CR2), “Chargé de Recherche 1<sup>re</sup> classe” (CR1), “Directeur de Recherche 2<sup>e</sup> classe” (DR2), “Directeur de Recherche 1<sup>re</sup> classe” (DR1) and “Directeur de Recherche de Classe Exceptionnelle” (DRCE). The proportion of CNRS scientists for each position are given in the last column. CNRS researchers have no teaching duties.

position	pop	indus	teach	%
CR2	44.1	43.3	64.7	5.9
CR1	46.1	46.4	63.2	52.9
DR2	47.5	55.5	71.0	31.5
DR1	55.7	60.6	72.4	8.8
DRCE	57.5	62.5	65.0	.9

sciences have an average publication rate of 2.12, instead of 1.93 for inactive ones, p-value = 0.07).

#### 4.1.2 Scientists active in all dissemination activities

It is also interesting to look at the academic records of the scientists active in all three dissemination actions. They represent roughly 20% of our 3659 database, which is much more than expected if the engagements in the three different dissemination activities (teaching, industrial collaboration and popularization) were uncorrelated (14 %). This points to an “open” attitude, which makes a scientist practicing popularization more prone to teach or establish industrial collaborations. This high percentage is contrary to what one could expect from a “time consumption” argument, where each of these activities lowers the activity in the others. From an academic point of view, scientists who cumulate more than one dissemination activity are more active academically than those who carry out only one of them. The precise values are: scientists active in all three dissemination activities have a  $h_y$  of 0.75 against 0.70 for the others (p-value = 0.0001), those active in industrial collaborations and teaching 0.74 against 0.69 (p-value =  $2.0 \cdot 10^{-5}$ ), those active in industrial collaborations and popularization 0.74 against 0.70 (p-value = 0.012), those active in popularization and teaching 0.74 against 0.70 (p-value = 0.0008).

#### 4.1.3 Dissemination activity of the “best” scientists

One can also look at the dissemination attitude of the (academically) most active scientists, taken as those whose  $h$  increases faster than their career time ( $h_y > 1$ , totalizing 1/6 of CNRS researchers). They are more active in both outreach (44%



of active instead of 37%, p-value = 0.0035), industrial dissemination (56% of active instead of 51%, p-value = 0.035) and teaching (69% of active instead of 60%, p-value =  $7.5 \cdot 10^{-5}$ ). The same correlations are found by discipline, even if, again, the differences are less significative, except for biology and physics.

#### 4.1.4 Dissemination activity of “those who are not good enough”

Finally, one can investigate whether “those who are not good enough for an academic career” [Society (2006)] are the most active in dissemination. Our previous result suggest that this is not so, which is confirmed by a statistical analysis. Taking as “not good enough” the 25% of CNRS scientists with the lowest  $h_y$  (lower than .5), we find that these scientists are less active in dissemination, the precise figures being 39.4% active for popularization instead of 41.4% for the rest of the scientists (p-value = .25 i.e. non significative), 52.1% active for industrial collaboration instead of 57.8% (p-value = .0025) and 60% active for teaching instead of 67% (p-value = .0002). Even stronger differences (all highly significative statistically) are found if the number of publications per career year is used as the bibliometric indicator, which could be more appropriate when comparing scientists of different ages [Jensen, Rouquier, and Croissant (submitted, 2007)] (for example, 35.7% of the “not good enough” are active for popularization instead of 42.6% in the rest, p-value = .00021).

## 5 Which scientists are active in dissemination?

In the previous section, we have shown that scientists engaged in dissemination, be it popularization, teaching or industrial collaborations, are academically more active than inactive researchers. To be precise, we have shown that if you select randomly a scientist and ask him whether he is active in dissemination, a positive answer implies higher bibliometric indicators. We now investigate the separate effects of all the available scientists’ characteristics on the probability that they are active in dissemination. This will help us to interpret the correlations between academic and dissemination activities (next section).

We have conducted a statistical analysis intending to single out the individual effects of each one of these characteristics, all other things being equal. For example, for a (hypothetical) average researcher, we analyze the effect of her position, i.e. how much it separately increases (or diminishes) her probability of being active in popularization. Since the variable that we investigate is a logical variable (either active or inactive), we have used a standard logit regression model \*. In this model, the probability of being active is written as  $P(y_i = 1) = F(\beta'x_i)$  where  $\beta'$  is the vector of fitted coefficients and  $x_i$  the vector of characteristics of scientist  $i$  (age, position...). The marginal effect of a variation of the variable  $x_{ik}$  (where  $k$  refers to one of the characteristics) on the probability of being active can be written as  $\frac{\partial P(y_i=1)}{\partial x_{ik}} = \frac{\partial F}{\partial x_{ik}}(\beta'x_i) = \beta_k F'(\beta'x_i)$ . In a logit model,  $F(z) = \frac{e^z}{1+e^z}$ , which leads to  $F'(z) = \frac{e^z}{(1+e^z)^2} = F(z)(1 - F(z))$ . This function reaches its maximum for  $z = 0$ , which corresponds to an activity probability of 0.5, leading to a proportionality coefficient of 1/4. Therefore, a simple interpretation of the effect of a scientist characteristic on its probability of being active is the following : the maximum marginal effect of a characteristic equals the corresponding coefficient divided by a factor 4. For example, the isolated effect of an age increase of one year is a decrease of about  $0.027/4 * 100 \simeq .6\%$  of the probability of popularizing.

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\*The statistical analysis was carried out with the open software “R” (<http://www.r-project.org/>)

Being “CR1” decreases the probability of being active in industrial collaborations by 18% compared to a “DR2” sharing the same characteristics (age, sex, subdiscipline...).

Our results are summarized in Tables 6 and 7. The main influences of scientists’ characteristics, all other things being equal, are:

1. *position*: as scientists reach higher positions, they become significantly more active in all dissemination activities
2. *academic record*: there is no significant influence except for industrial collaborations. For this activity, scientists with higher Hirsch index are more active
3. *age*: dissemination activities decrease with age
4. *gender*: women are more active in popularization, men in teaching, and there is no significative difference in industrial collaborations

Results (1) and (3) represent a clear example of the usefulness of a regression study. Since age and position are strongly correlated, a simple study of the evolution of the proportion of active scientists with age is not concluding. Figure 4 confirms that popularization activity decreases with age for all positions, but that scientists in higher hierarchical positions are much more active.

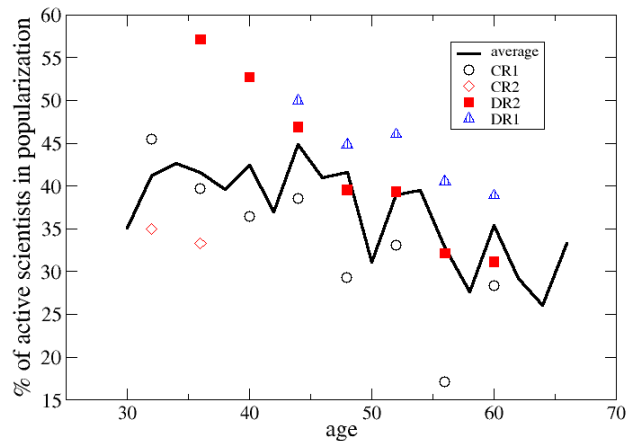


Figure 4: Evolution of the proportion of scientists active in popularization as a function of their age, on average and for different positions. Data correspond to the filtered database, i.e. without social sciences and particle physics, which explains the differences with Figures 1 and 2.

It should be noted that there are strong positive correlations between different types of dissemination. As argued previously, this shows that different activities do not compete for scientist time. Table 6 also show the specificities of the subdisciplines for the different dissemination activities. For example, mathematicians are much more active in teaching, less in industrial collaborations ...

Table 7 shows how scientists’ characteristics influence the probability of being active in *all* the dissemination activities. In general, the effects are similar to those seen for single activities, but the effects are stronger. The main difference is the

strong effect of academic activity, which is even stronger than for industrial collaborations alone. This further confirms that the different activities (academic and dissemination) are not competing but tend to reinforce mutually. Table 7 also shows the effects of age, gender, etc. on the probability of being inactive in all dissemination activities. These effects are consistent (i.e. opposite) with those seen for the probability of being active, except for the lack of significant effect of the academic activity.

To interpret the correlations between academic and dissemination activities, one has to take into account that our regression models quantify academic achievements in two different ways: the position and the bibliometric indicator  $h$ . Since we include both in our regressions, the effect of  $h$  is considered “all other things being equal”, i.e. only within each position. Therefore, the lack of influence of  $h$  for the probability of being active in popularization (third column of Table 6) is not in contradiction with our previous findings (Table 5 for example). Indeed, since positions are strongly correlated with  $h$  and higher positions are much more active in popularization (Table 6), our regression shows that that popularization activity is more influenced by the hierarchical position than by the bibliometric indicators within each category<sup>†</sup>. The opposite is observed for the probability of participating in “open days” events (second column of Table 6). Here, position is not relevant (none affects significantly the participation even if a definite trend exists) but Hirsch index is. A similar (but in the opposite direction) conclusion can be drawn for popularization in the press (first column of Table 6): position is only slightly relevant (except for the youngest scientists, which rarely participate), but  $h$  is very important, as is scientist’s age. Briefly, the main influences for the other types of actions are the following: contracts with industrial partners are strongly (positively) influenced by the position, Hirsch index and (decreasing) age. Instead, popularization through television and conferences or industrial collaborations through patents or licenses only depend on position and age and not bibliometric indicators. Finally, participation in popularization in schools decreases with age but is unaffected by the position or academic activity.

## 6 Interpreting our results

What can we learn from this statistical study of dissemination activities? We will examine different hypothesis and compare them to our findings. Our interpretations are centered on popularization practices. We will deal with industrial collaborations in future work.

- **H0: Dissemination is done by “those who are not good enough for an academic career”.** If we admit that bibliometric indicators are a good proxy for “being good enough for an academic career”, then our study clearly invalidates this hypothesis. First, randomly chosen active scientists have higher academic indicators than inactive ones. Second, all other things being equal, the probability of disseminating increases with academic position (Table 6). Furthermore, better academic records increase the probability of being active in industrial collaborations.
- **H1: Dissemination is done by people close to retirement.** Our study has also shown that this hypothesis is incorrect. Table 3 shows that scientists close to retirement are less active than average. Our statistical also shows that, as their age increases, scientists become less active in dissemination, all other things being equal.

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<sup>†</sup>If positions are omitted from the regressions,  $h$  becomes strongly significative.

- **H2: Popularization is driven by demand upon the scientific elite.** This hypothesis assumes that popularization is driven by an external demand (institutions or journalists). Then, the scientific elite, with higher bibliometric indices, is more visible from outside the scientific community, and is therefore more solicited. Our data supports this interpretation: scientists engaged in the type of popularization actions mostly driven by demand (radio, television, press, and to a lesser extent, conferences) have a higher  $h_y$  than average (Table 8). Instead, scientists performing the popularization activities that are mostly driven by the offer, and symbolically less important (open days, school conferences, web sites...) have a slightly lower average  $h_y$  than the other scientists (Tables 6 and 8).
- **H3: Active personality.** We have repeatedly pointed out that being active in one activity is positively correlated to being active in the others. This suggests that the correlations observed could arise from some internal characteristic of the scientists involved, call it “active personality” or “intellectual capacity”. For example, it could be argued that popularization is intellectually demanding, for it is difficult to explain complex scientific issues in simple terms. Therefore, good popularization demands a deep understanding of the subject, as anyone preparing lectures has experienced. This “intellectual capacity” would in turn generate higher academic records. Alternatively, one could interpret an “active personality” as one able to “sell” his work, both to journalists and to those colleagues in charge of refereeing papers and citing them. The “active personality” argument is supported by the the fact that scientists active in all dissemination activities are also very active academically speaking.
- **H4: Social and cognitive hierarchies.** The observed correlations between position and dissemination activities can also be understood by referring to sociological studies of scientific communities. Terry Shinn [Shinn (1988)] studied a French physics lab for several years, looking for correlations between hierarchical positions and cognitive work. He noticed a clear work division between junior and senior scientists. Junior staff devote most of their time to experiments or “local” questions. By “local”, Shinn means questions focused on particular points : a single experiment, a thorough investigation of a very precise subtopic . . . . In contrast, senior scientists devote most of their time to “general” questions, i.e. how the local results can be inserted into global theoretical or conceptual frameworks. They also spend much time establishing and maintaining social networks both inside and outside the scientific community. Both these activities are clearly more in line with dissemination activities, which demand putting scientific problems into perspective. Moreover, senior scientists generally have a team of junior researchers working with them, which is active even when seniors are outside of the lab disseminating... One could also argue that the scientific elite is able to transform its symbolic capital [Bourdieu (1984)], gained in the academic arena, into public arenas, thus popularizing not only on issues directly related to their own domain but on virtually any issue. This would lead to a correlation between higher popularization activity and academic records, as for H2. In an old study of popularization practices of CNRS scientists, Luc Boltanski [?] has observed that Senior scientists (“Directeur de Recherche”) have the legitimacy to speak to the public in the name of the institution. Instead, scientists in the lowest positions can only express their own point of view, and popularization is mostly seen as a waste of time or a personal occupation.
- **H5: Benefits of dissemination.** It is the reverse causality from H2 and H4. Dissemination activities compel scientists to open up their horizon, to discuss with people having other points of view on their research topics, giving

new insights, contacts, which could improve their academic research. Actually, H2, H4 and H5 could act together in a reinforcing way. It seems difficult to argue that this effect is dominating, but it could contribute to the observed correlations, mainly in the case of industrial collaborations which strongly correlate with higher academic indicators.

As a summary, it is likely that the strong correlations observed between dissemination and academic activity result from the cumulative effects of H2-H5.

## 7 Are dissemination activities good for the career?

It is commonly recognized that scientists engaged in dissemination do not get much reward, and that their involvement can even be bad for their career [Society (2006)]. In France, the CNRS director stated recently the importance of taking into consideration “scientific culture popularization actions” for the evaluation of researchers: “one must insist that they give equal importance to scientific work and to activities related to the popularization and dissemination of scientific culture: participations in “open doors” events, or the publication in magazines or other popularization works, in events organized for non-specialized audiences, newspaper articles or TV appearances, etc.” (letter sent to CNRS scientists in 2005, our translation). In the document that was supposed to steer his long-term policy, the “Multi-year action plan” [CNRS (2004)], the CNRS thus declares that: “If current [evaluation] practice is suitable for the purpose of evaluating academic research, the same cannot be said for interdisciplinary activities and for other facets of scientific work: transfer of scientific knowledge, teaching and popularization. Consequently, the work by CNRS researchers who choose to engage in these activities, which are very necessary for the CNRS, is not adequately acknowledged and researchers are therefore reluctant to proceed in this direction.”

Thanks to our large database, we are able to study statistically the influence of dissemination activities on the promotions of CNRS researchers to senior positions (“Directeur de Recherche”) over the 2004–2006 period. Table 9 shows the results of our regression analysis, for all CNRS disciplines and for each discipline separately. It turns out that dissemination activities are *not* bad for scientists careers. They are not very good either: the effects are generally weak, but positive, and rarely significant. The detailed study by discipline shows that the overall positive effect of popularization arises mainly from its recognition in Life sciences and the positive effect of teaching from Chemistry. However, it is interesting to note that all dissemination activities influence positively promotions for most of other disciplines, even if their effects are not significant.

Overall, two characteristics have strong effects: academic activity (h or the number of papers) and age (the “optimal” age for becoming DR2 is 46.6 years, for DR1 52.4). For DR1 promotion, there is a small (and positive) effect from industrial collaborations.

## 8 Discussion, Conclusions

Our statistical study on the correlations between dissemination activities and academic records of more than 3500 scientists from most disciplines has allowed us to establish several facts. First, we have clearly shown that scientists engaged

in dissemination are more active academically, formally refuting the common idea that “dissemination activities are carried out by those who are not good enough for an academic career” [Society (2006)]. We have even shown that some prestigious activities (presse, radio and television) are mostly carried out by the scientific “elite” in academic terms. One can certainly criticize the idea that bibliometric indicators do account properly for the academic quality of scientists [Jensen, Rouquier, and Croissant (submitted, 2007), Kostoff (1998), Leydesdorff (1998), Liu (1993), Brooks (1996)]. However, those who argue that dissemination activities are carried out by the “worst” scientists usually do accept this definition of scientific quality. Therefore, our paper should convince them that they are wrong.

This was the “easy” part of the discussion. The interpretation of our results is otherwise not easy, as there have been few qualitative studies on the perception by scientists of popularization or teaching practices. Concerning relations with industries, a group of the Catholic University of Leuven (K.U. Leuven, Belgium) examined the academic records of 32 scientists “inventors” in their university [B. Van Looy (2006)]. Their data suggest a “reinforcing or positive spillover effect on scientific performance from engaging in technology development efforts”, consistent with our findings in a much larger sample. A recent study has investigated the factors that predict scientists’ intentions to participate in public engagement [Poliakoff and Webb (2007)]. Thanks to a questionnaire distributed to academic staff and postgraduates, it was found that the main reasons why scientists decided not to participate in public engagement activities are the following : they had not participated in the past (a result consistent with our former finding in [Jensen and Croissant (2007)]), they have a negative attitude toward participation (it is seen as ‘pointless’ or ‘unenjoyable’), they feel they lack the skills and finally that they do not believe that their colleagues participate in such activities, which is interpreted as a signal of relative irrelevance of this activity. Notably, the lack of time or career recognition are not seen as important determinants of participation. Clearly, more qualitative studies on the relations of the scientific milieu and dissemination practices are needed. Here, we limit our discussion to a few issues:

- **What have we learnt about relations of scientists and popularization?** Our study suggests that popularization is mostly an activity of the (academically speaking) scientific “elite”. Our finding agrees with data from the Royal Society survey [Society (2006)], which shows that higher positions popularize more, the differences being even larger than in our database : senior staff is active at 86%, while junior staff is active at a mere 14%. The following comment follows : ”The seniority finding is borne out by the qualitative research which found that young researchers keen to climb the research career ladder were focused on research and publishing and/or felt that they needed more experience before they could engage with those outside their research community.” We may add (H4 above) that senior activities are more in line with dissemination than those carried out by junior staff. The question that our findings raise is then: why does a significant fraction of the scientific community feel that “only bad scientists” popularize? Is it a problem of jealousy for colleagues that manage to present their results to a wide audience? Is that because, cognitively speaking, creating knowledge is judged more important than disseminating it, as suggested by Shinn [Shinn (1988)]? This would imply that scientists are still prisoners of the “diffusion model” [Weigold (2001)], which ignores that to disseminate knowledge, one has to recreate it altogether, a creative and difficult endeavour.

Second, qualitative interviews indicate that many reasons push scientists to engage in popularization. In private

discussions, popularizers acknowledge that one of the main reasons is the pleasure of interacting with the public, of going out of the lab [Pérez, Kreimer, and Jensen (2008)]. For the Royal Society study [Society (2006)], i.e. in a more official environment, the strongest reason given to justify popularization is “informing the public”. We can wonder whether scientists are still prisoners of the so-called deficit model [Weigold (2001)]. This is an old model for scholars of the Science Studies field, dating back to 1960. It insists on the teaching of elementary scientific facts and methods to the public. Listening to the public seems important to only a few percent of the scientists interviewed in the UK [Society (2006)]. However, this idea should be one of the strongest with a more “generous” vision of the public in mind [Leblond (1992), Wagner (2007)]. Scientists also seem to ignore the numerous criticisms to the deficit model: the relation between the knowledge of scientific facts and its appreciation is empirically unsolved, the knowledge of the “facts” of science taken out of their context is more alienating than it is informative... It is also important to establish links in the other way, where scientists learn from society [Irwin and Wynne (1996), Jurdant (1993), Bensaude-Vincent (2001)]. The culture of the scientific milieu seems far from these ideas at the moment.

- **What do we learn about relations between science and society?** Our study shows that, even in the institution hosting the most fundamental sciences, roughly half of the scientists are in close contact with society, i.e. popularize or look for funding outside the academic sphere. This could worry some “fundamentalists” which think that science should be isolated from society needs, because society can only perturb science. For example, the recent French movement “Sauvons la recherche” wrote in its final document that “science can only work by developing its own questions, protected from the emergency and the deformation congenital to social and economics worlds”. However, our result will not surprise scholars in Science Studies, as they know that the “ivory tower” of science never existed. Scientists have always been connected with society, from which they depend for funding (see, among many others, [Biagioli (1993), Pestre (2003), Latour (1988)]). Even the most fundamental physical theories such as relativity [Galison (2004)] and quantum mechanics [Hoddeson et al. (1992)] have been inspired by applications.
- **Dissemination and career** Another contradiction between our study and common views among scientists is the idea that “dissemination activities are negative for the career”. For example, 20% of scientists of the Royal Society survey [Society (2006)] answered that scientists who engage in popularization are viewed less well by their peers. Here, we have shown that promotion is mainly determined by academic indicators, dissemination activities being marginal, counting only for specific disciplines (chemistry for teaching, life sciences for popularization). There is however no negative effect. How can we explain the common opposite idea that pervades the scientific community?

We have started our paper by the proclamation of many prestigious institutions that dissemination activities are priorities. We have shown that these activities are carried out by academically active scientists, that receive no reward for their engagement. We feel that institutions’ duty is now to invent ways of evaluating and rewarding the active scientists.

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## 9 Appendix 1 : A summary of popularization activities of CNRS scientists

To draw up their annual report, researchers must specify the type of popularization activities that they have performed. The following table displays a distribution of types of activities according to the different scientific departments of the CNRS for the 2006 data. The categories are chosen by the scientists themselves. Most categories names in the Table are self-explanatory. 'Associations' refer to popularization actions taken to help associations understand scientific aspects of their activity (think of patients or astronomical associations). 'Schools' refers to actions taking place in schools. 'Web' to popularization sites on the Web.

It is interesting to analyze the misrepresentation of certain disciplines for each type of activity. For instance the over-representation of Social Sciences researchers in Radio/Television and, to a lesser extent, in activities involving associations, the press and conferences. Not surprisingly, these researchers are by far under-represented in "open door" events. On the other hand, their weak presence in schools is food for thought for the community. The Nuclear physics, Chemistry and Engineering departments are over-represented in "open door" activities, which are relatively scarce in Life Sciences. These departments are rather absent from actions involving the press, radio or publishing.

	Chemistry	Nuclear	Earth	Life	Social	Engineering	Physics	Info tech
Other	0.14	0.11	0.06	0.10	0.05	0.10	0.10	0.10
Conference	0.18	0.24	0.28	0.19	0.30	0.20	0.24	0.23
Exhibition	0.09	0.08	0.06	0.05	0.06	0.11	0.11	0.09
Associations	0.02	0.03	0.05	0.05	0.05	0.04	0.04	0.02
Schools	0.14	0.14	0.09	0.14	0.02	0.13	0.11	0.10
Books/CD Rom	0.03	0.02	0.04	0.03	0.04	0.03	0.04	0.05
Open doors	0.16	0.09	0.08	0.09	0.01	0.13	0.12	0.13
Press	0.13	0.14	0.15	0.18	0.19	0.13	0.13	0.14
Radio/Television	0.06	0.06	0.14	0.14	0.22	0.08	0.07	0.06
Web	0.06	0.08	0.05	0.03	0.05	0.05	0.05	0.08

## 10 Appendix 2 : Obtaining reliable bibliometric indicators for several thousand scientists

In the following, we detail our procedure to obtain a large but reliable sample ( $\simeq 3\,500$  records) of bibliometric indicators (number of publications, citations and  $h$  index). The difficulty lies in the proper identification of the publications of each scientist. Two opposite dangers arise. The first one consists in including extra publications because the request is not precise enough. For example, if only surname and name initials are indicated to WoS, the obtained list may contain papers from homonyms. The second one consists in missing some papers. This can happen if scientists change initials from time to time, or if the surname corresponds to a woman who changed name after marriage. But this can also happen when one tries to be more precise to correct for the first danger, by adding other characteristics such as scientific discipline or French institutions for CNRS scientists. The problem is that both the records and the ISI classifications are far from ideal: the scientific field can be confusing for interdisciplinary research, the limitation to French institutions incorrect for people starting their career in foreign labs, etc.

Basically, our strategy consists in guessing if there are homonyms (see below how we manage to get a good idea on this). If we think there are no homonyms, then we count all papers, for any supplementary information (and the resulting selection) can lead to miss some records. If we guess that there are homonyms, then we carefully select papers by scientific domain and belonging to French institutions. After all the bibliometric records have been obtained in this way, we filter our results to eliminate “suspect” records by two criteria: average number of publications per year and scientist’s age at the first publication.

### 10.1 Evaluate the possibility that there exist homonyms

For this, compute the ratio of the number of papers found for the exact spelling (for example JENSEN P.) and all the variants proposed by WoS (JENSEN P.\*, meaning P.A. P.B., etc.). If this ratio is large (in our study, larger than .8), then the studied surname is probably not very common and the author might be the single scientist publishing. To get a more robust guess, we use the scientist’s age. We look at the total number of papers and compare it to a “maximum” normal rate of publishing, taken to be 6 papers a year. If the publishing rate is smaller than our threshold, this is a further indication that there is a single scientist behind all the records. Actually, our strategy can be misleading only when there are only homonyms with the same initial *and* all the homonyms have published very few papers.

### 10.2 Obtain the bibliometric records

**No homonyms** If we guess that there is a single scientist behind the publications obtained for the surname and initials (which happens for about 75% of the names), we record the citation analysis corresponding to all associated papers.

**Homonyms** If we estimate that there are homonyms, we try to eliminate them by using supplementary data we have. We refine the search by scientific field (“Subject category” in WoS terms, but one can select only one) and by selecting only

French institutions<sup>‡</sup>.

### 10.3 Eliminate suspect records

Finally, once all the data has been gathered according to the preceding steps, we eliminate “suspect” results by two criteria related to the scientist’s age. For a record to be accepted, the age of the first publication has to be between 21 and 30 years, and the average number of publications per year between 0.4 and 6. After this filtering process, we end up with 3659 records out of the 6900 initial scientists, i.e. an acceptance rate of 53%.

Can we understand why half of the records are lost? First of all, let us detail how the different filters eliminate records. Deleting scientists who published their first paper after 30 years old eliminates 1347 “suspect” names, which are probably related to errors or missing papers in the WoS database, to married women for whom we miss the first papers published under their own surname and to people who started their career in non French institutions and had homonyms. Deleting scientists who published their first paper before 21 years old eliminates 1235 additional “suspect” names, which are probably related to errors in the WoS database, to scientists with older homonyms which we could not discriminate. Deleting scientists whose record contained less than .4 papers per year in average leads to the elimination of 121 names. These wrong records can be explained by the method missing some publications, as in the case “first publication after 30 years old”. Deleting scientists whose record contained more than 6 papers per year in average leads to the elimination of further 178 names. These wrong records can be explained by the presence of homonyms we could not discriminate. Finally, to make our database more robust, we decided to eliminate records suspect of containing homonyms even after selection of discipline and institution. This is done by eliminating the 359 scientists for which the number of papers kept after selection is smaller than 20% of the total number of papers for the same surname and initials. In those cases, we do not trust enough our selection criteria to keep such a fragile record.

### 10.4 A robust bibliometric database

In summary, our method leads to a reliable database of around 3500 scientists from all “hard” scientific fields. It only discriminates married women having changed surname. It also suffers from the unavoidable wrong WoS records<sup>§</sup>. We stress that the main drawback of the elimination of half the records is the resulting difficulty in obtaining good statistics. But at least we are pretty sure of the robustness of the filtered database.

Our filtering criteria are based on homonym detection, age of first publication and publication rate. The first criteria correlates only with scientist’s surnames, therefore we can expect that it introduces no bias except for married women. Actually, there is a lower woman proportion after filtering: 24.9% women in the 3659 selection, against 29.6% in the 6900 database. This is consistent with the preferential elimination of married women who changed surnames and have an incomplete bibliographical record.

The two other criteria could discriminate scientific disciplines with lower publication rates or underrepresented in WoS.

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<sup>‡</sup>Unfortunately, WoS allows the selection to be made only on the institutions of all coauthors as a whole. So we might retain articles of homonyms that have coauthored a paper with a French scientist.

<sup>§</sup>For a noticeable fraction of scientists, WoS records start only in the 1990s, even if there exist much older publications, which can be found for example by Google Scholar.

For example, we see in the following table that more scientists from the Engineering Department have been eliminated in the filtering. The mean age is somewhat lower in the filtered database (46.4 years) to be compared to 46.8 in the whole dataset, probably because the records from older scientists have a higher probability of containing errors.

However, overall, the filtered database is very similar to the initial one. For example, the percentage of candidates to senior positions is 16.0% in the 3659 selection, against 16.4% in the 6900 database, and the respective promoted percentages are 4.9% and 5.0%. The proportions of scientists from each position is also similar : none of the small differences between the filtered and unfiltered values is statistically significant.

As noted previously, the robustness of our filtered database is validated by the significantly better indicators found for scientists in higher positions. An even stronger test (because the effect is subtler) resides in testing the correlations of the scientist's age at his(her) first publication with several variables : age, position, subdiscipline and gender. We find [?] a progressive decrease of the age of first publication when a scientist has a higher position (all things being equal, for example scientist's age), an effect that is intuitively appealing but certainly small. The fact that we can recover such a subtle effect is a good indication of the robustness of our procedure. We also recover the intuitive effect of scientist's age (older scientists have begun their career later). The gender effect (men publish their first paper 2 months later than women, all other things being equal) is more difficult to interpret, since it mixes many effects : our discrimination (in the filtering procedure) of married women, the unknown effects of marriage and children on scientists' careers, etc.

Table 5: Differences in scientific activity — as measured by the normalized Hirsch index — for different subpopulations, characterized by the strength of their dissemination activities. To simplify the presentation, we only keep two categories: inactive and active, the latter grouping the “active” and “very active” categories of Table 1. We also show the p-values obtained by a standard “Welch Two Sample t-test” and the number of scientists in each domain.

Popularization

	inactive $h_y$	active $h_y$	p-value	Number of scientists
Physical sciences	0.68	0.73	0.036 *	669
Life sciences	0.75	0.81	0.0018**	1275
Engineering	0.50	0.52	0.38	504
Chemistry	0.73	0.74	0.54	848
Earth Sciences, Astrophysics	0.69	0.77	0.037 *	363

Industrial Collaboration

	inactive $h_y$	active $h_y$	p-value	Number of scientists
Physical sciences	0.65	0.78	$< 1.E-6^{***}$	669
Life sciences	0.69	0.83	$< 1.E-6^{***}$	1275
Engineering	0.47	0.52	0.17	504
Chemistry	0.69	0.75	0.0066**	848
Earth Sciences, Astrophysics	0.74	0.74	0.96	363

Teaching

	inactive $h_y$	active $h_y$	p-value	Number of scientists
Physical sciences	0.69	0.70	0.77	669
Life sciences	0.67	0.81	$< 1.E-6^{***}$	1275
Engineering	0.47	0.52	0.12	504
Chemistry	0.69	0.76	0.0004***	848
Earth Sciences, Astrophysics	0.74	0.76	0.58	363

Table 6: Binomial regressions to explain dissemination activities for 3659 CNRS scientists (filtered database). “Active” means at least one action in the three-year period encompassed by our study (2004-2006). For popularization, this represents 1495 scientists (40.9%), for industrial collaborations 2058 scientists (56.2%) and for teaching 2373 scientists (64.9%). These percentages are different from those of Table 1 because Social sciences are excluded from the filtered database. The explanatory variables are: sex, age, position, subdiscipline and Hirsch index ( $h$ ) as bibliometric quantifier. The reference levels are: “Condensed matter: structure” for the subdiscipline and DR2 for the position. The columns give the coefficients of the fit, together with their significance (standard significance codes for the p-values have been used: “\*\*\*” for  $< 0.001$ , “\*\*” for  $< 0.01$ , “\*” for  $< 0.05$ , “.” for  $< 0.1$ .) The position “DRCE” is almost never significative because it includes less than 100 scientists (Table 3).

	Subdiscipline	press	open days	active in pop.	active in industrial	active in teaching
	(Intercept)	-1.8 *	-1.7 **	0.71 .	1.9 ***	3.1 ***
	$h$	0.037**	-0.033*	0.0020	0.019*	0.0067
	age	-0.057***	-0.0010	-0.027***	-0.052***	-0.078***
	CR2	-0.97 **	0.21	-0.58 **	-1.5 ***	-1.7 ***
	CR1	-0.28	0.089	-0.25 *	-0.72 ***	-0.94 ***
	DR2			reference		
	DR1	0.21	-0.37	0.24 .	0.70 ***	0.31 *
	DRCE	0.37	-14	0.26	0.96 *	-0.15
	active in popularization	X	X	X	0.40 ***	0.39 ***
	active in industrial collab.	0.37 *	0.25 .	0.47 ***	X	0.61 ***
	active in teaching	0.44 **	0.42 **	0.60 ***	0.40 ***	X
	male	0.15	-0.21	-0.19 *	-0.08	0.45 ***
Physical sciences	1 Mathematics	0.31	-1.6 **	-0.41	-1.2 ***	0.96 ***
	2 Physics, theory and method	1.1 .	-0.92 .	-0.37	-0.64 *	0.06
	4 Atoms and molecules	1.5 **	0.34	0.097	-0.18	0.19
	5 Condensed matter: dynamics	0.5	-0.017	0.15	0.21	0.17
	6 Condensed matter: structure			reference		
Engineering	7 Information science	1.3 *	-0.72	-0.48	1.5 ***	2.0 ***
	8 Micro and nano-technologies	1.2 *	-0.75 .	-0.11 .	2.5 ***	0.69 *
	9 Materials and structure	1.4 *	-0.53	-0.32	1.9 ***	2.4 ***
	10 Fluids and reactants	1.4 *	-0.41	0.032	2.1 ***	1.0 ***
Chemistry	11 Super/macromolecular systems	-0.18	-0.04	-0.35	0.92 ***	0.80 **
	12 Molecular architecture	-1.4	-1.1 *	-0.91 **	0.86 **	0.73 **
	13 Physical chemistry	-0.077	-0.43	-0.59 *	0.011	0.46 .
	14 Coordination chemistry	-0.50	-0.16	-0.35	1.0 ***	0.027
	15 Materials chemistry	0.83	-0.8 .	-0.52 *	1.3 ***	0.65 *
	16 Biochemistry	0.28	-0.51	-0.81 **	0.97 ***	0.82 **
Earth sciences, astrophysics	17 Solar systems, universe	2.3 ***	0.56	1.2 ***	-1.1 ***	-0.11
	18 Earth and earth plants	1.6 **	-0.19	0.63 *	-0.37	1.9 ***
	19 Earth systems: superficial layers	2.3 ***	-0.58	0.76 *	0.17	0.48
	20 Continental surface	1.8 **	-0.82	0.72 *	1.0 **	1.4 ***
Life sciences	21 Molecular basis of life	0.064	-1.0 *	-1.1 ***	1.1 ***	1.1 ***
	22 Genomic organization	-0.065	-1.5 **	-1.0 ***	0.12	1.1 ***
	23 Cellular biology	-1.6	-0.97 *	-1.0 ***	0.51 .	1.2 ***
	24 Cellular interaction	0.31	-1.4 **	-0.73 **	0.40	1.4 ***
	25 Physiology	-0.17	-1.9 **	-0.69 **	0.61 *	1.1 ***
	26 Development, evolution	0.19	-0.92 .	-0.70 **	0.37	1.1 ***
	27 Behavior, cognition and brain	2.7 ***	-1.6 **	0.56 *	0.78 **	1.8 ***
	28 Integrative vegetal biology	0.53	0.063	-0.35	0.57 *	0.79 **
	29 Biodiversity, evolution	2.1 ***	-0.13	0.4	0.76 *	1.6 ***
	30 Therapy, pharmacology	1.0	-1.3 *	-0.58 *	1.1 ***	0.85 **

Table 7: Binomial regressions to explain dissemination activities for 3659 CNRS scientists (filtered database). “Active in all dissemination activities” means at least one action in each of the activities over the three year period and “Inactive” no action in any of the three dissemination activities. The explanatory variables are: sex, age, position, subdiscipline and Hirsch index ( $h$ ) as bibliometric quantifier. The reference levels are: “Condensed matter: structure” for the subdiscipline and DR2 for the position. The columns give the coefficients of the fit, together with their significance (standard significance codes for the p-values have been used: “\*\*\*” 0.001 “\*\*” 0.01 “\*” 0.05 “.” 0.1. )

Subdiscipline		inactive for all	active for all
	(Intercept)	-6.4 ***	1.5 **
	$h$	-0.01	0.027 **
	age	0.10 ***	-0.071 ***
	CR2	2.3 ***	-1.7 ***
	CR1	1.1 ***	-0.86 ***
	DR2		reference
	DR1	-0.71 **	0.60 ***
	DRCE	-0.23	0.20
	male	-0.32 **	-0.15
Physical sciences	1 Mathematics	0.19	-0.63
	2 Physics, theory and method	0.77 *	-0.32
	4 Atoms and molecules, lasers and optics	-0.02	-0.25
	5 Condensed matter: organization and dynamics	0.19	0.74 *
	6 Condensed matter: structure		reference
Engineering	7 Information science and technology	-1.9 **	1.2 ***
	8 Micro and nano-technologies, electronics and photonics	-1.9 ***	1.4 ***
	9 Materials and structure engineering	-1.7 **	1.7 ***
	10 Fluids and reactants: transport and transfer	-1.5 ***	1.7 ***
Chemistry	11 Super and macromolecular systems, properties and functions	-0.79 *	0.55
	12 Molecular architecture synthesis	-0.25	0.31
	13 Physical chemistry: molecules and environment	0.31	-0.18
	14 Coordination chemistry: interfaces and procedures	-0.26	0.31
	15 Materials chemistry: nanomaterials and procedures	-0.85 *	0.72 *
	16 Biochemistry	-0.66 .	0.45
Earth sciences, astrophysics	17 Solar systems and the universe	-0.06	-0.40
	18 Earth and earth plants	-2.4 ***	0.62 .
	19 Earth systems: superficial layers	-1.9 **	0.34
	20 Continental surface and interfaces	-1.2 *	1.9 ***
Life sciences	21 Molecular basis and structure of life systems	-0.47	0.41
	22 Genomic organization, expression and evolution	0.03	0.087
	23 Cellular biology: organization and function	-0.39	0.27
	24 Cellular interaction	-0.48	0.26
	25 Molecular and integrative physiology	-0.74 *	0.34
	26 Development, evolution, reproduction and aging	-0.43	0.14
	27 Behavior, cognition and brain	-1.7 ***	1.7 ***
	28 Integrative vegetal biology	-0.32	0.61 .
	29 Biodiversity, evolution and biological adaptation	-1.4 **	1.6 ***
	30 Therapy, pharmacology and bioengineering	-0.46	0.80 *



Table 8: Differences in scientific activity — as measured by the normalized Hirsch index — for different subpopulations, characterized by the type of their popularization activities. The first group refers to prestigious activities regulated mainly by the outside demand, whereas the second group gathers less prestigious activities mainly driven by scientists’ offer. The p-values give the statistical significativity of the differences. They are obtained by a standard ‘Welch Two Sample t-test’. Note that the differences for the first group are highly significant even if the number of scientists active in those activities is quite low: between 220 (radio/television) and 430 (public conferences).

type of action	$h_y$		p-value
	active	inactive	
Press	0.82	0.70	$1.8 \cdot 10^{-6}$ ***
Radio, Television	0.81	0.70	$5.5 \cdot 10^{-5}$ ***
Public Conference	0.75	0.70	0.019 *
School conference	0.73	0.71	0.18
Open days	0.68	0.71	0.11
Web sites	0.73	0.71	0.55

Table 9: Binomial regressions to explain promotions to senior positions (from CR1 to DR2) on the 586 candidates and 179 promotions from all scientific disciplines of our filtered database. The last column shows the regressions for the DR2 to DR1 promotion, on 376 candidates and 67 promotions from all scientific disciplines of our filtered database. The explanatory variables are: sex, age, subdiscipline (not shown to simplify since none is significative) and  $h$  as bibliometric quantifier (except for Engineering, where the number of articles accounts much better for the promotions, see [Jensen, Rouquier, and Croissant (submitted, 2007)]). The columns give the coefficients of the fit for each scientific domain, together with their significance. Standard significance codes for the p-values have been used: 0 “\*\*\*\*” 0.001 “\*\*\*” 0.01 “\*\*” 0.05 “.” 0.1.

	all	Physics	ENG	Earth	Chemistry	Life	DR1
(Intercept)	-41.2	-78.2	-71.8 *	-35.41 .	-42.94	3.53	-49.6 **
act pop	0.4 .	0.01	0.068	-0.82	0.38	0.91 *	-.29
act indus	0.12	0.88	-1.22	-0.085	0.06	0.14	.65 .
act teach	0.79 **	0.76	1.66	1.08	1.06 *	0.41	.33
male	-0.23	-0.99	0.72	0.5	-0.68	0.23	-.36
$h$ (or art for ENG)	0.13 ***	0.26 ***	0.038 *	0.12 .	0.11 *	0.15 **	.106 ***
age	1.12 ***	2.48 *	3.2 *	1.26	0.96 .	0.46	1.78 **
age <sup>2</sup>	-0.012 ***	-0.026 *	-0.036 *	-0.012	-0.0096 .	-0.0057	-0.017 **