

The file drawer effect – a long-lasting issue in the sciences

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ABSTRACT

Most of the scientific outputs produced by researchers are inaccessible since they are not published in scientific journals: they remain in the researchers' drawers, forming what we call the Dark Science. This is a long-standing issue in research, creating a misleading view of the scientific facts. Contrary to the current literature overfed with positive findings, the Dark Science is nurtured with null findings, replications, flawed experimental designs and other research outputs. Publishers, researchers, institutions and funders all play an important role in the accumulation of those unpublished works, but it is only once we understand the reasons and the benefits of publishing all the scientific findings that we can collectively act to solve the Dark Science problem. In this article, we discuss the causes and consequences of the Dark Science expansion, arguing that science and scientists would benefit from getting all their findings to the light of publication.

Keywords: file drawer effect, dark science, publication bias, null findings, replications, flawed designs

The Dark Science, a.k.a The File Drawer Effect

Science and scientists work best when they have full access to literature. Unfortunately, most of the scientific outputs produced are inaccessible. Half of the published experiments are still blocked behind paywalls¹ and an even bigger number is tucked away in file drawers, composing what we call the '*Dark Science*'.

While the growth of preprint platforms and open access scientific journals – where the published content is freely accessible – has begun to solve the former issue (+25 % of open access articles since 2000¹), the Dark Science problem remains relatively unhandled.

The Dark Science problem, commonly called the file drawer problem, refers to a culture of seeking and publishing only certain kinds of results, mostly those that reach the magic $p=0.05$. In 1979², Rosenthal described that scientists' file drawers were filled with 95% of non-significant results while the published literature was mainly composed of positive findings - with a probabilistic 5% of false positives (Type I error)-. In a recent report, Allen and Melher³ confirmed Rosenthal's claim. The current literature contains from 5 to 20% of null findings, while the literature obtained from pre-registered reports – when the methods and the hypotheses of an experiment are published before acquiring the data – has 4 to 10 times more. This clearly suggests that the traditional literature is a misleading representation of the scientific facts.

While it is still unclear how many experiments fill the Dark Science every year, we know that it is widespread in the sciences^{4,5} and that it has most probably increased with time, since positive findings grew by over 15% from 1990 to 2007⁶.

In the following sections of this article, we identify (1) the scientific outputs that are more likely to be stored in file drawers rather than being published in top scientific journals, (2) the causes and consequences of the Dark Science expansion and (3) the benefits of bringing the Dark Science back to the sight of science and scientists.

Dark Science, content, causes and consequences

Nowadays, articles stuck in the Dark Science are mostly null (statistically non-significant) findings, replications and flawed designed experiments.

Null findings are not all similar⁷. In fact, we can categorize them into two main types: inconclusive and conclusive findings⁸. Inconclusive findings often emanate from studies with small sample sizes (such as sometimes in field studies) that lack sufficient statistical power and are hardly interpretable^{9,10}. Conclusive findings, instead, can be interpreted with more certainty.

Replications are also less likely to be published in top scientific journals. This is particularly true for replications yielding no effect or confirming established results. However, replications demonstrating the opposite effect of established results seem to escape more easily from the Dark Science¹¹.

79 Flawed designs are poorly documented too¹¹. Often, scientists fail at
80 designing proper experimental designs to answer their hypothesis, but
81 also fail at disseminating failures.

82 Null findings, replications and flawed designs are common in
83 research. In fact, they probably constitute the majority of the research
84 outputs and are often discussed informally in laboratories during lab
85 meetings given their importance into the making of scientific discoveries.
86 However, the rest of the scientific community does not benefit from them
87 as they remain unpublished.

88 In 2014, null findings were 40 % less likely to see the light of
89 publication and in 60% of the case, they were not even written up⁴. It
90 appears difficult to come up with one main reason explaining this
91 approach, still, we can point out how publishers, researchers and funders
92 (either passively or actively), all participate in this phenomenon.

94 Publishers

95 Most of the time, scientific journals seek breakthroughs. This leads
96 to publication biases, which occur, according to Dickersin¹², when
97 “publication of study results is based on the direction or significance of
98 the findings”. Publication biases are globally observed in all the scientific
99 disciplines and countries⁶. As stated in the previous section³, the scientific
100 literature is only composed of 5 to 20% of null findings, while it should
101 approximate 60%. One reason is that null findings (as well as replications
102 and flawed designs) do not fill the knowledge gap. At least, not at first.

Therefore, their impact is difficult to evaluate at the moment of publication. According to Johnson and Dickersin⁷, a less common but more worrying explanation to publication bias is that editors direct their article selection towards eye-catching content for readers; in the worst case, wittingly for financial purposes (in 2017 scientific publishing annual market represented \$10 billion¹³).

Besides, the publishing costs indirectly affect publication bias. Researchers with limited funds prioritize their work that will be most likely to be published: novel, positive and therefore more attractive to most journals.

However, more and more journals and publishing platforms are seeking reliability rather than novelty. They willingly encourage scientists to submit their null findings, replications and 'failures' to disseminate them to the broadest audience (see In&Sight, PLOS one, PeerJ, JOTE, etc.). Publishers and journals appear to not be the only ones at fault in the Dark Science expansion, researchers play a major role too.

Researchers

For Johnson and Dickersin⁷, the main reason why null findings are not published is that researchers fail to submit them. This is somehow confirmed by Franco and collaborators⁴ who demonstrated that 60% of null findings are not even written up although most scientists agree that they are worth being published^{11,14}. There are multiple reasons that lead researchers to keep part of their research unpublished. As described

above, the cost to publish is one of them. Researchers have limited funds, and it costs on average \$1500 to publish an article in an open access journal. Therefore, they prioritize some work at the expense of others. Time is limited as well, and it appears to be the main reason why researchers do not publish all of their findings¹⁴. Considering that publishing is a long process – ranging from a couple of months to one year to publish an article¹⁵ – and that publishers are mostly interested in novel findings, it is not surprising that much (if not most) research ends up in the Dark Science. For replications and flawed designs experiments, the same reasons probably apply¹⁶. Specifically related to flawed designs, or what we could call failures, researchers could feel embarrassed with their work and, therefore, more likely to put it aside. Going back to null findings, another reason that pop-up from Echevarria's study¹⁴ is that null findings are less cited than positive results (although this is only true in specific fields¹⁷ and therefore, are not submitted for publication to journals. This is a very interesting point as it brings the '*incentives to publish*' component on the table: how researchers are evaluated by their institutions and funding agencies is of fundamental relevance; these last are therefore responsible too in the Dark Science expansion.

Institutions/Funders

The growing competition for research funding and academic positions, combined with the use of specific bibliometrics (h-index, impact factor) led researchers into a 'publish or perish' culture: publish as many articles

as possible in high-impact factor journals^{18,19}. In Life Sciences, it is still widely believed that you need to author a paper in high-impact factor journals such as Cell, Nature or Science (CNS) to secure a faculty position and receive job offers^{20,21}. As those journals are highly competitive and seeking breakthroughs, it strongly influences what researchers focus on and, very likely, prevent them from spending time on publishing other findings, which instead end up in the Dark Science²². Until institutions and funders won't use new/different metrics to select the next scientists' generation, researchers will continue to look for high-impact factor journals, with publishers promoting theirs to attract researchers – a never-ending loop -.

The pressures to publish are indeed intense and inevitably influence researchers but it is also good to remember that publication metrics do not necessarily correlate with the percentage of job offers received²¹ and that next-generation metrics in line with open science practices are arising^{23,24}. By changing the way institutions and funders evaluate researchers, the incentives to publish can change toward better practices; practices that could be beneficial to science and the community.

As stated at the beginning of this section, it is difficult to come up with one reason to explain the Dark Science expansion which appears inevitable in today's research culture. Publishers, researchers and funders/institutions all play a role and the entire ecosystem requires changes.

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177 **Publishing the Dark Science, what are the benefits?**

178 We argue that many findings of the Dark Science such as null findings,
179 replications and flawed designs, would be beneficial to science if
180 published and addressed adequately.

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182 Null but conclusive findings have the power to dissuade researchers
183 from unfruitful avenues. Contrarily, null but inconclusive findings are
184 hardly interpretable. However, they are useful in laying down scientific
185 questions that remain to be tackled by the scientific community.
186 Surprisingly or not, once published and grouped in systematic reviews,
187 those underpowered studies can be lifesaving. This is well illustrated by
188 the Stroke Unit Trialists' collaborative systematic review who gathered a
189 series of null and underpowered studies²⁵ which once together
190 demonstrated that having a stroke unit care reduces the odds of death by
191 17% (95% confidence interval 4-29%). Similar examples have been
192 described in cancer, cardiovascular and thyroid-related research,
193 demonstrating the importance of publishing such studies²⁶⁻²⁸.
194 Replications validate or invalidate published findings and make science
195 reliable. They are therefore crucial to science. However, at the moment
196 less than 25% of the replicated studies confirm the original results^{29,30} and
197 many papers fail to disseminate enough methodological details to allow
198 replications³¹. Although they may appear embarrassing, flawed designs
199 are important too. Those studies should be used to learn from each other's

mistakes. In today's publishing process, flawed designs are mostly identified by peer reviewers prior to publication. Unfortunately, in most cases, it leads to the rejection of the article and not to the discussion of the detected problem. Yes, science is often flawed, but it is time we embrace it since this is certainly the only activity which allow us to be actually wrong (at least 5% of the times).

Overall, publishing the Dark Science can only lead to a more reliable, collaborative and complete science. Beyond doubts, publishing the Dark Science would save time, effort and funds. The amount of resources saved is difficult to evaluate, but Glazsiou and Chalmers³² did get an approximation of the funds lost every year in medical research only. They estimated that 85 % of the research was wasted, which corresponds to \$170 billion per year; a number that should chill up your spine. As for saving funds, the amount of time we could save is incalculable. However, we can count the number of hours scientists spend reviewing articles that are then rejected. It represents 15 million hours every year³³. Necessary, but wasted as the articles are buried in the Dark Science, potentially with precious information to convey to the scientific community.

To make reliable science and discoveries, scientists need access to all the information. However, this brings up an important limitation that should be tackled. As the number of scientific articles increases every year, it is more and more difficult for scientists to stay up to date and find relevant literature for their research. Publishing the Dark Science would

make the search for relevant content even harder, as it would double the number of articles published every year. However, the internet has offered tools previously inconceivable. Those tools, and the use we make of them, have the possibility to change the way we select, verify, publish and search science. After decades of keeping our research in the drawers, it is time to unleash our Dark Science with open eyes.

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