Publish (your data) or (let the data) perish! Why not publish your data too?

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ABSTRACT

The authors argue that upon publication of a paper, the data should be made available through online archives or repositories. Reasons for not sharing data are discussed and contrasted with advantages of sharing, which include abiding by the scientific principle of openness, keeping the data for posterity, increasing one’s impact, facilitation of secondary analyses and collaborations, prevention and correction of errors, and meeting funding agencies’ increasingly stringent stipulations concerning the dissemination of data. Practicing what they preach, the authors include data as an online appendix to this editorial. These data are from a cohort of psychology freshmen who completed Raven’s Advanced Progressive Matrices, tests of Numerical Ability, Number Series, Hidden Figures, Vocabulary, Verbal Analogies, and Logical Reasoning, two Big Five personality inventories, and scales for social desirability and impression management. Student’s sex and grade point average (GPA) are also included. Data could be used to study predictive validity of cognitive ability tests, Extraversion, Neuroticism, Conscientiousness, Openness to Experience, Agreeableness, and the general factor of personality, as well as sex differences, differential prediction, and relations between personality and intelligence.

1. Introduction

Congratulations! You have just received the good news from the editor, popped open the bubbly to celebrate your latest scientific contribution, and – after having sobered up – are now finalizing your manuscript for publication in this journal. You have written an interesting paper, and are eager to share your empirical findings with the world. So why not go all the way, and publish your raw data too?

2. The cons of data sharing

Of course, there may be insurmountable reasons not to share the data. You may not own the data to begin with. Or sharing may be hampered by ethical considerations relating to privacy. The data may be part of a longitudinal study that will be featured in other articles, and are thus part of an ongoing project.

Perhaps you are hesitant to share your data for other reasons. You probably invested considerable time and resources in collecting and analyzing the data, and may therefore consider yourself the rightful owner (even if, say, the taxpayer helped finance the data). You may be afraid that some nitpicking statistician may one day discover an error in one of the statistical analyses that you reported in your paper or that a researcher with opposing views will disagree with your analyses. You may feel that you will be able to publish another paper with the data, provided you had the time. You may consider it too much trouble to document the data properly, which clearly takes time and effort. But such impediments to data sharing may be petty when compared to the benefits of sharing.

3. The pros of data sharing

You may not realize that there are many advantages to publishing the data following publication. First, the required...
data archiving, i.e., making them suitable for dissemination, ensures that your data are not (ultimately) lost. One of us once requested data from a close colleague, who responded: “Sure I will send you those data, but it’s like seven computers ago, and so please allow me some time to hunt them down”. Apparently, archiving is not every psychologist’s cup of tea; in a survey among 192 Dutch research psychologists, two thirds indicated to not archive their data (Voorbrood, 2010). Those who indicated to do so often confused proper archiving of data with the simple act of saving the file on their current computer.

When Wicherts et al. requested data from 141 psychological researchers back in 2005 (Wicherts, Borsboom, Cats, & Molenaar, 2006), the relevant papers had been published less than 12 months earlier and only 27% of the researchers were willing and able to share (some of) their data. The low response rate was mostly due to the disappearance of the data and researchers’ failure to document the data file in an accessible format. Only a small number of researchers flatly refused. Although we do expect your data handling practices to be relatively good in light of the relative complexity of intelligence test data, it would still be good to outsource archiving so as to save the data for posterity.

Second, sharing the data is consistent with the universal scientific norms of openness and rigor. We recently found that psychologists who fail to share their data after publication commit considerably more errors in the reporting of statistical results and report findings that are more likely to be debatable (Wicherts, Bakker, & Molenaar, 2011). Openness with respect to data is good for you and scientific progress, while concealment may facilitate all sorts of trouble, including faulty analyses, overly positive reporting of results, and even misconduct (Wicherts, 2011).

Third, publishing one’s data has been shown to increase citation scores of the papers which first feature the data (Piwowar, Day, & Fridsma, 2007). So sharing is good for the publishing journal and it is good for the citation score on your resume. Notwithstanding our critique of Lynn and Vanhanen’s (2006) work on national IQ (e.g., Wicherts, Dolan, Carlson, & van der Maas, 2010), their work was followed up by many researchers simply because Lynn and Vanhanen published their data in an accessible form.

Fourth, sharing the data encourages more research because it enables secondary (novel) analyses, for which you may currently lack time, knowledge, expertise, or even interest. Advances in both statistical techniques and knowledge of our research topic will certainly open up novel possibilities with your data in the future. We are sure that many of your colleagues, who use your published data for novel analyses, will have the courtesy to ask you to join the novel research project (as happened in, e.g., Dolan et al., 2006). Besides, having first-hand knowledge of the data, plus possible relevant background information, your help may be essential in those future studies. Gifts are often returned. In addition, meta-analysts often find a considerable number of published studies to be useless because primary researchers failed to report the effect size or correlation of interest, i.e., crucial information that is lost once the data are unavailable (Ortego & Botella, 2010). Proper data archiving is essential for the quality of meta-analyses (Cooper & Patall, 2009).

Fifth, sharing the data facilitates subsequent reanalyses, which may have diverse beneficial results. On the one hand, they may result in the correction of potential errors in the analyses and in the reporting of results (which are quite common; Bakker & Wicherts, 2011; Wicherts et al., 2011) and may even help prevent such errors in the first place (Wicherts, 2011). On the other hand, reanalyses may emphasize the robustness of your original substantive results. Of course, laying oneself open to criticism and subsequent correction is good scientific practice, but psychologically not necessarily palatable. However, if trivial errors are found, these are very unlikely to be publishable and will receive little attention (although ultimately one is well served by knowledge of such errors). If serious errors are found, you should politely thank the person who found them and learn from the experience (which is, after all, the essence of scientific progress).

Sixth, many funding agencies including the National Institutes of Health (NIH), National Science Foundation (NSF), the Wellcome Trust, the Medical Research Council (MRC), the Deutsche Forschungsgemeinschaft (DFG), and the Netherlands Organization for Research (NWO) in our country have stipulated that grantees should either write a data sharing plan as part of proposals or make the data publically available upon completion of the project. In fact, you are already obliged keep your data in accessible form for at least five years and to share your data for verification purposes upon request (American Psychological Association, 2010). Obviously: you need not worry about such stipulations, if you routinely submit your data along with your paper.

4. Putting the money where our mouth is

Data can be archived in several online data repositories. For instance, data could be put on the website of the International Society for Intelligence Research (http://www.isironline.org/) by emailing Timothy Bates. Together with IT-specialists and librarians at the University of Amsterdam, we are currently developing a repository that will eventually include much of the data from over 40 years of our freshman-testing program. Moreover, Elsevier’s ScienceDirect offers the possibility to archive data as an online appendix (e.g., Hassal & Sherrat, 2011; Johnson & Bouchard, 2011). To illustrate the possibilities offered by Elsevier, we attach a data set to the current editorial that is potentially interesting for future use. The file includes data from our freshman-testing program called “Testweek” (Busato, Prins, Elshout, & Hamaker, 2000; Smits, Dolan, Vorst, Wicherts, & Timmerman, 2011; Wicherts & Vorst, 2010) in which 537 students (age: M = 21.0, SD = 4.3) took the Advanced Progressive Matrices (Raven, Court, & Raven, 1996), a test of Arithmetic, a Number Series test, a Hidden Figures Test, a test of Vocabulary, a test of Verbal Analogies, and a Logical Reasoning test (Elshout, 1976). Also included are data from a Dutch big five personality inventory (Elshout & Akkerman, 1975), the NEO-PI-R (Hoekstra, Ormel, & Fryt, 1996), scales of social desirability and impression management (based on work by Paulhus, 1984; Wicherts, 2002), sex of the participants, and grade point averages of the freshmen’s first trimester that may act as outcome variable. We had ambitious plans to use these data in studies of sex differences, differential prediction, advanced item response models, the general factor of personality (de Vries, 2011; Van der Linden, Te Nijenhuis, & Bakker.

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2010), the link between IQ and personality, etc., but we simply lacked the time. We therefore gladly make these data available for the consideration of the scientific community.

The data are anonymized and may be used for any purpose. Our only request is a proper citation to this editorial to make our point. The data are described in Table 1 and are given in an Excel file. The PDF in the supplement provides additional information on univariate distributions, some data-handling, number of items, time limits, and scale reliabilities. Fig. 1 gives the path diagram with standardized loadings of a reasonably well-fitting factor model of the seven cognitive tests: $\chi^2_{SB} = 16.95, \text{DF} = 10, p = .076$, $\text{RMSEA} = .040$, $\text{SRMR} = .028$, $\text{CFI} = .992$. This model was fitted in LISREL 8.80 to the data of 429 participants who completed all seven subtests and who were native speakers of Dutch. The Satorra Bentler-scaling was inspired by the non-normality in some of the variables (caveat lector!).

Perhaps the most interesting feature of sharing data is that anyone may disagree with the primary authors’ analytic choices and run alternative models, use other selections of the data, possibly leading to different conclusions. To be sure: there is simply nothing more scientific than a debate about the raw data. Subscale-level data (for the NEO-PI-R), item-level data, and test materials (in Dutch) are available upon request.

5. Conclusion

By not archiving our data, much valuable data get lost every year. For instance, you may still have a 5¼ inch floppy disk with valuable data, but nowhere to insert it. We should stop this waste and start archiving in an accessible and robust way. This particularly applies to the types of wealthy multivariate data sets that we as intelligence researchers often collect. Perhaps the best example of secondary analyses in our field is Carroll’s (1993) treatise on the structure of individual differences in cognitive abilities in which he factor analyzed data from over 460 studies by using their correlation matrices. This shows that it would be wise to always publish the full covariance matrix to accompany any factor analytic application. But the raw data could be essential. For instance, Molenaar and colleagues (Molenaar, Dolan, & Maas, 2011; Molenaar, Dolan, Wicherts, & van der Maas, 2010) recently developed novel methods to study Spearman’s Law of Diminishing Returns and these methods can only be applied with the raw data. Our understanding of this phenomenon as well as many others rests greatly on the possibilities to replicate these results on the basis of “old” data and in fact

Table 1

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
<th>N</th>
<th>Min</th>
<th>Max</th>
<th>Mean</th>
<th>SD</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sex</td>
<td>Sex of participant</td>
<td>534</td>
<td>1(=M)</td>
<td>2(=F)</td>
<td>1.72</td>
<td>.451</td>
</tr>
<tr>
<td>Language</td>
<td>Native Dutch speaker?</td>
<td>523</td>
<td>1(=Y)</td>
<td>2(=N)</td>
<td>1.10</td>
<td>.305</td>
</tr>
<tr>
<td>ZCPA</td>
<td>Z of 1st tri. Grade Point Average</td>
<td>314</td>
<td>-3.06</td>
<td>22.56</td>
<td>0.00</td>
<td>1.00</td>
</tr>
<tr>
<td>RAV_until</td>
<td>APM completed until item no.</td>
<td>521</td>
<td>14</td>
<td>36</td>
<td>27.24</td>
<td>4.13</td>
</tr>
<tr>
<td>RAV_score</td>
<td>APM no. correct</td>
<td>521</td>
<td>3.00</td>
<td>31.00</td>
<td>21.17</td>
<td>4.33</td>
</tr>
<tr>
<td>E5pft</td>
<td>Extraversion – 5PF</td>
<td>517</td>
<td>30</td>
<td>90</td>
<td>64.00</td>
<td>10.81</td>
</tr>
<tr>
<td>A5pft</td>
<td>Agreeableness – 5PF</td>
<td>517</td>
<td>25</td>
<td>96</td>
<td>72.17</td>
<td>9.27</td>
</tr>
<tr>
<td>C5pft</td>
<td>Conscientiousness – 5PF</td>
<td>517</td>
<td>33</td>
<td>90</td>
<td>61.34</td>
<td>9.74</td>
</tr>
<tr>
<td>N5pft</td>
<td>Neuroticism – 5PF</td>
<td>517</td>
<td>19</td>
<td>83</td>
<td>45.89</td>
<td>12.19</td>
</tr>
<tr>
<td>O5pft</td>
<td>Openness – 5PF</td>
<td>517</td>
<td>39</td>
<td>96</td>
<td>66.40</td>
<td>9.52</td>
</tr>
<tr>
<td>Lretot</td>
<td>Logical reasoning total score</td>
<td>513</td>
<td>14</td>
<td>40</td>
<td>12.62</td>
<td>9.01</td>
</tr>
<tr>
<td>Nsetot</td>
<td>Number series total score</td>
<td>527</td>
<td>-23</td>
<td>23</td>
<td>6.68</td>
<td>7.46</td>
</tr>
<tr>
<td>Voctot</td>
<td>Vocabulary total score</td>
<td>523</td>
<td>-28</td>
<td>21</td>
<td>-8.82</td>
<td>8.57</td>
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<tr>
<td>Hfigtot</td>
<td>Hidden figures total score</td>
<td>512</td>
<td>-26</td>
<td>29</td>
<td>7.96</td>
<td>10.24</td>
</tr>
<tr>
<td>Vantot</td>
<td>Verbal analogies total score</td>
<td>519</td>
<td>-30</td>
<td>30</td>
<td>5.52</td>
<td>11.74</td>
</tr>
<tr>
<td>Artot</td>
<td>Arithmetic total score</td>
<td>521</td>
<td>-52</td>
<td>48</td>
<td>12.74</td>
<td>7.86</td>
</tr>
<tr>
<td>lm</td>
<td>Impression Management</td>
<td>515</td>
<td>10</td>
<td>39</td>
<td>23.86</td>
<td>4.56</td>
</tr>
<tr>
<td>sd</td>
<td>Self Deception</td>
<td>515</td>
<td>8</td>
<td>39</td>
<td>19.03</td>
<td>6.19</td>
</tr>
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<td>nNEO</td>
<td>Neuroticism – NEO-PI-R</td>
<td>500</td>
<td>79</td>
<td>210</td>
<td>135.67</td>
<td>21.73</td>
</tr>
<tr>
<td>eNeo</td>
<td>Extraversion – NEO-PI-R</td>
<td>500</td>
<td>112</td>
<td>212</td>
<td>167.37</td>
<td>17.04</td>
</tr>
<tr>
<td>oNEO</td>
<td>Openness – NEO-PI-R</td>
<td>500</td>
<td>127</td>
<td>220</td>
<td>172.47</td>
<td>16.33</td>
</tr>
<tr>
<td>aNEO</td>
<td>Agreeableness – NEO-PI-R</td>
<td>500</td>
<td>93</td>
<td>212</td>
<td>164.88</td>
<td>16.83</td>
</tr>
<tr>
<td>cNEO</td>
<td>Conscientiousness – NEO-PI-R</td>
<td>500</td>
<td>87</td>
<td>212</td>
<td>153.97</td>
<td>18.60</td>
</tr>
</tbody>
</table>

Note: Valid N (listwise) = 280.
Roberto Colom kindly shared the Spanish WAIS-III standardization data precisely for that purpose (see also Dolan et al., 2006).

The call for more openness with respect to research data is rapidly increasing in many scientific fields (Nature, 2009), and, as intelligence researchers we should not lag behind. We owe much of our knowledge of intelligence from statistical analyses that have evolved over the years and are subject to further future refinement. Our ideas may change, which may lead us to revisit our data. Not so long ago both editors and publishers, in view of journal space, desired journal articles to be written as densely as possible. But times have changed. With the onset of online publishing, we can publish sizeable files as online appendices to journal articles. Published. With the onset of online publishing, we can publish it to our research participants and to science.

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Appendix A. Supplementary material

Supplementary data to this article can be found online at doi:10.1016/j.intell.2012.01.004.

References


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