

# The Availability of Research Data Declines Rapidly with Article Age

Timothy H. Vines,<sup>1,2,\*</sup> Arianne Y.K. Albert,<sup>3</sup> Rose L. Andrew,<sup>1</sup> Florence Débarre,<sup>1,4</sup> Dan G. Bock,<sup>1</sup> Michelle T. Franklin,<sup>1,5</sup> Kimberly J. Gilbert,<sup>1</sup> Jean-Sébastien Moore,<sup>1,6</sup> Sébastien Renaut,<sup>1</sup> and Diana J. Rennison<sup>1</sup>

<sup>1</sup>Biodiversity Research Centre, University of British Columbia, 6270 University Boulevard, Vancouver, BC V6T 1Z4, Canada

<sup>2</sup>Molecular Ecology Editorial Office, 6270 University Boulevard, Vancouver, BC V6T 1Z4, Canada

<sup>3</sup>Women's Health Research Institute, 4500 Oak Street, Vancouver, BC V6H 3N1, Canada

<sup>4</sup>Centre for Ecology & Conservation Biosciences, University of Exeter, Cornwall Campus, Tremough, Penryn TR10 9EZ, UK

<sup>5</sup>Institute for Sustainable Horticulture, Kwantlen Polytechnic University, 12666 72<sup>nd</sup> Avenue, Surrey, BC V3W 2M8, Canada

<sup>6</sup>Department of Biology, Université Laval, 1030 Avenue de la Médecine, Laval, QC G1V 0A6, Canada

## Summary

Policies ensuring that research data are available on public archives are increasingly being implemented at the government [1], funding agency [2–4], and journal [5, 6] level. These policies are predicated on the idea that authors are poor stewards of their data, particularly over the long term [7], and indeed many studies have found that authors are often unable or unwilling to share their data [8–11]. However, there are no systematic estimates of how the availability of research data changes with time since publication. We therefore requested data sets from a relatively homogenous set of 516 articles published between 2 and 22 years ago, and found that availability of the data was strongly affected by article age. For papers where the authors gave the status of their data, the odds of a data set being extant fell by 17% per year. In addition, the odds that we could find a working e-mail address for the first, last, or corresponding author fell by 7% per year. Our results reinforce the notion that, in the long term, research data cannot be reliably preserved by individual researchers, and further demonstrate the urgent need for policies mandating data sharing via public archives.

## Results

We investigated how research data availability changes with article age. To avoid potential confounding effects of data type and different research community practices, we focused on recovering data from articles containing morphological data from plants or animals that made use of a discriminant function analysis (DFA). Our final data set consisted of 516 articles published between 1991 and 2011. We found at least one apparently working e-mail for 385 papers (74%), either in the article itself or by searching online. We received 101 data sets (19%) and were told that another 20 (4%) were still in use and could not be shared, such that a total of 121 data

sets (23%) were confirmed as extant. Table 1 provides a breakdown of the data by year.

We used logistic regression to formally investigate the relationships between the age of the paper and (1) the probability that at least one e-mail appeared to work (i.e., did not generate an error message), (2) the conditional probability of a response given that at least one e-mail appeared to work, (3) the conditional probability of getting a response that indicated the status of the data (data lost, data exist but unwilling to share, or data shared) given that a response was received, and, finally, (4) the conditional probability that the data were extant (either “shared” or “exists but unwilling to share”) given that an informative response was received.

There was a negative relationship between the age of the paper and the probability of finding at least one apparently working e-mail either in the paper or by searching online (odds ratio [OR] = 0.93 [0.90–0.96, 95% confidence interval (CI)],  $p < 0.00001$ ). The odds ratio suggests that for every year since publication, the odds of finding at least one apparently working e-mail decreased by 7% (Figure 1A). Since we searched for e-mails in both the paper and online, four factors contribute to the probability of finding a working e-mail: (1) the number of e-mails in the paper and (2) the chance that any of those worked and (3) the number of e-mails we could find by searching online and (4) the chance that any of those worked. The total number of e-mail addresses we found in the paper decreased with age (Poisson regression coefficient =  $-0.07$ , SE = 0.01,  $p < 0.0001$ ) from an average of 1.17 in 2011 to 0.42 in 1991 (Figure 2A), and there was a slight positive effect of article age on the number of e-mails we found online (Poisson regression coefficient = 0.015, SE = 0.007,  $p < 0.05$ ; Figure 2C). Moreover, the chance that an e-mail found in the paper or online appeared to work also showed a relationship with article age (OR = 0.96 [0.926–0.998, 95% CI],  $p < 0.05$ ; and OR = 0.97 [0.936–0.997, 95% CI],  $p < 0.05$ ; respectively), such that the odds that an e-mail appeared to work declined by 4% and 3% per year since publication, respectively (Figures 2B and 2D).

We note that eight e-mail addresses generated an error message but did lead to a response from the authors. It also seems likely that some addresses failed but did not generate an error message, leading us to record a “no response” rather than “e-mail not working,” although unfortunately the frequency of these cannot be estimated from our data.

There was no relationship between age of the paper and the probability of a response given that there was an apparently working e-mail (50% response rate, OR = 1.00 [0.97–1.04, 95% CI]; Figure 1B). There was also no relationship between article age and the probability that the response indicated the status of the data, given a response was received (83% useful responses, OR = 1.00 [0.95–1.07, 95% CI]; Figure 1C).

Finally, there was a strong negative relationship between the age of the paper and the probability that the data set was still extant (either “shared” or “exists but unwilling to share”), given that a response indicating the status of the data was received (OR = 0.83 [0.79–0.90, 95% CI],  $p < 0.0001$ ; Figure 1D). The odds ratio suggests that for every yearly increase in article age, the odds of the data set being extant decreased by 17%.

\*Correspondence: [vines@zoology.ubc.ca](mailto:vines@zoology.ubc.ca)

Table 1. Breakdown of Data Availability by Year of Publication

Year	No Working E-Mail	No Response to E-Mail	Response Did Not Give Status of Data	Data Lost	Data Exist, Unwilling to Share	Data Received	Data Extant (Unwilling to Share + Received)	Number of Papers
1991	9 (35%)	9 (35%)	2 (8%)	4 (15%)	1 (4%)	1 (4%)	2 (8%)	26
1993	14 (39%)	11 (31%)	3 (8%)	7 (19%)	0 (0%)	1 (3%)	1 (3%)	36
1995	11 (31%)	9 (26%)	0 (0%)	7 (20%)	2 (6%)	6 (17%)	8 (23%)	35
1997	11 (37%)	9 (30%)	1 (3%)	2 (7%)	3 (10%)	4 (13%)	7 (23%)	30
1999	19 (48%)	13 (32%)	1 (2%)	1 (2%)	0 (0%)	6 (15%)	6 (15%)	40
2001	13 (30%)	15 (35%)	3 (7%)	4 (9%)	0 (0%)	8 (19%)	8 (19%)	43
2003	9 (20%)	20 (43%)	4 (9%)	2 (4%)	0 (0%)	11 (24%)	11 (24%)	46
2005	11 (24%)	14 (31%)	6 (13%)	1 (2%)	0 (0%)	13 (29%)	13 (29%)	45
2007	12 (18%)	31 (47%)	2 (3%)	4 (6%)	1 (2%)	16 (24%)	17 (26%)	66
2009	9 (13%)	34 (49%)	3 (4%)	5 (7%)	6 (9%)	12 (17%)	18 (26%)	69
2011	13 (16%)	29 (36%)	8 (10%)	0 (0%)	7 (9%)	23 (29%)	30 (38%)	80
Totals	131 (25%)	194 (38%)	33 (6%)	37 (7%)	20 (4%)	101 (19%)	121 (23%)	516

Data are displayed as n (%); the percentages are calculated by rows.

## Discussion

We found a strong effect of article age on the availability of data from these 516 studies. The decline in data availability could arise because the authors of older papers were less likely to respond, but this was not supported by the data. Instead, researchers were equally likely to respond (Figure 1B) and to indicate the status of their data (Figure 1C) across the entire range of article ages.

The major cause of the reduced data availability for older papers was the rapid increase in the proportion of data sets reported as either lost or on inaccessible storage media. For papers where authors reported the status of their data, the odds of the data being extant decreased by 17% per year (Figure 1D). There was a continuum of author responses between the data being reported lost and being stored on inaccessible media, and they seemed to vary with the amount of time and effort involved in retrieving the data. Responses included

authors being sure that the data were lost (e.g., on a stolen computer) or thinking that they might be stored in some distant location (e.g., their parent's attic) to authors having some degree of certainty that the data are on a Zip or floppy disk in their possession but no longer having the appropriate hardware to access it. In the latter two cases, the authors would have to devote hours or days to retrieving the data. Our reason for needing the data (a reproducibility study) was not especially compelling for authors, and we may have received more of these inaccessible data sets if we had offered authorship on the subsequent paper or said that the data were needed for an important medical or conservation project.

The odds that we were able to find an apparently working e-mail address (either in the paper or by searching online) for any of the contacted authors did decrease by about 7% per year. This decrease was partly driven by a dearth of e-mail addresses in articles published before 2000 (0.38 per paper on average for 1991–1999) compared with those

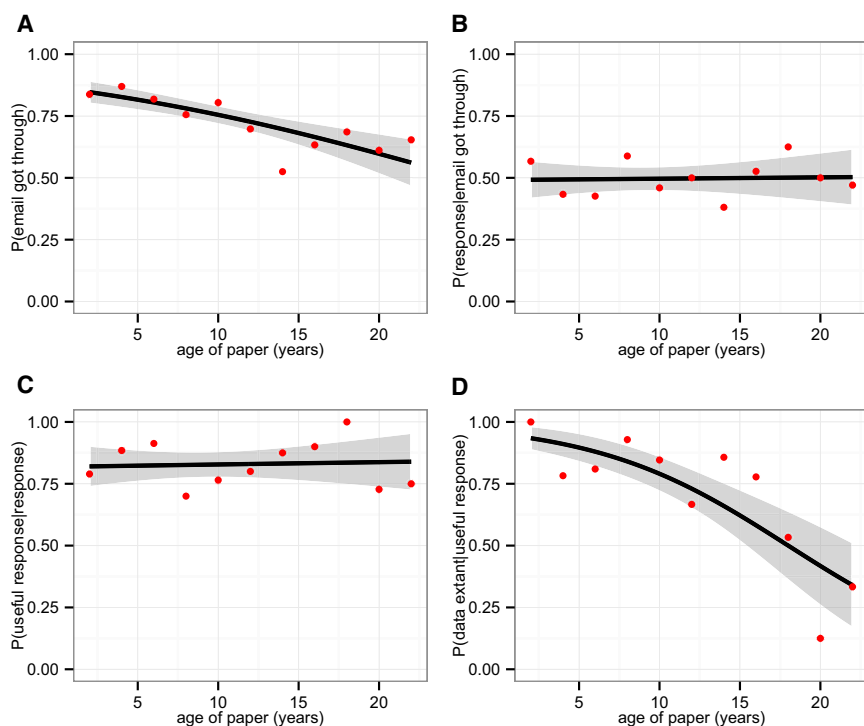


Figure 1. The Effect of Article Age on Four Obstacles to Receiving Data from the Authors

(A) Predicted probability that the paper had at least one apparently working e-mail.

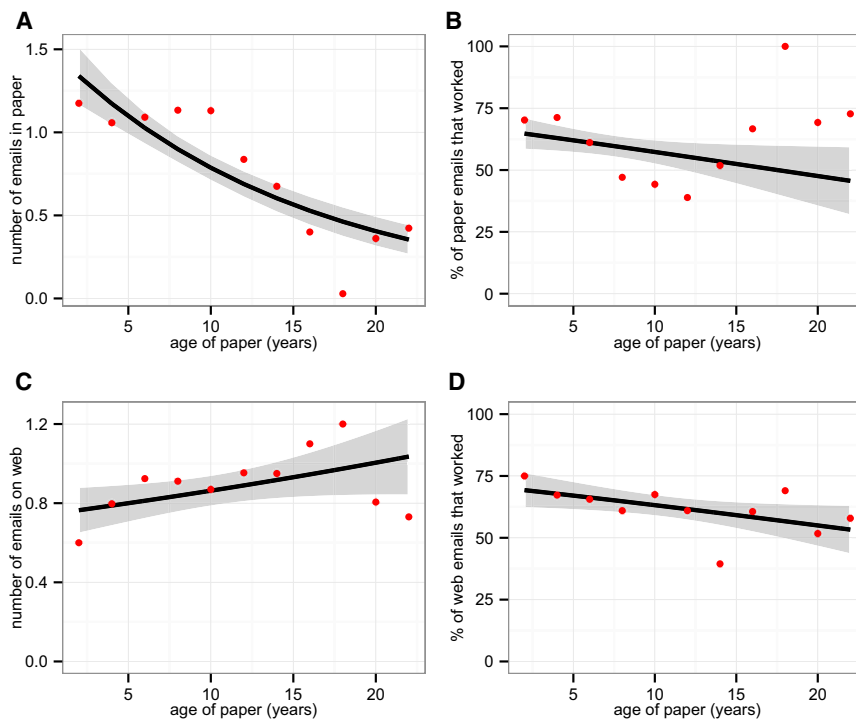
(B) Predicted probability of receiving a response, given that at least one e-mail was apparently working.

(C) Predicted probability of receiving a response giving the status of the data, given that we received a response.

(D) Predicted probability that the data were extant (either “shared” or “exist but unwilling to share”) given that we received a useful response. In all panels, the line indicates the predicted probability from the logistic regression, the gray area shows the 95% CI of this estimate, and the red dots indicate the actual proportions from the data.

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**Figure 2. The Effect of Article Age on the Number and Status of Author E-Mails**

(A) Number of e-mails found in the paper against article age.

(B) Predicted probability that an individual e-mail from the paper appeared to work against article age.

(C) Number of e-mails found by searching on the web against article age.

(D) Predicted probability that an individual e-mail found on the web appeared to work against article age.

The line indicates the predicted probability from a Poisson (A and C) or logistic (B and D) regression, the gray area shows the 95% CI of this estimate, and the red dots indicate the actual proportions from the data.

published after 2001 (1.08 per paper on average; [Figure 2A](#)). Wren et al. [12] found a similar increase in the number of e-mails in articles published after 2000. The larger number of e-mails in recent papers may mean that the issue of missing author e-mails is restricted to articles from before 2000: researchers in e.g., 2031 will be able to try a wider range of addresses in their attempts to contact authors of articles published in 2011.

The proportion of e-mails from the paper that appeared to work declined with article age between 2 and 14 years of age and then rose to around 80% for articles from 1991, 1993, and 1995 ([Figure 2B](#)). These latter three proportions are only based on a total of 13 e-mail addresses. Wren et al. [12] reported a steep decline with age in the proportion of functioning e-mails from papers published between 1995 and 2004, such that 84% of their 10-year-old e-mails returned an error message. Our proportions for 10-year-old e-mails are lower, with only 51% of e-mails from 2003 returning an error. It may be that e-mail addresses are becoming more stable through time, although this clearly requires additional study. The arrival of author identification initiatives like ORCID [13] and online research profiles such as ResearchGate or Google Scholar should make it easier to find working contact information for authors in the future.

Considering only the papers from 2011, our results show that asking authors for their data shortly after publication does yield a moderate proportion of data sets (~40%). A comparable study [11] received 59% of the requested data sets from papers that were less than a year old. It is hard to tell whether this difference is due to the slightly different research communities involved or the presence of an extra year between publication and the data request in this study. A related paper by Wicherts et al. in 2005 [9] received only 26% of requested psychology data sets.

Overall, we only received 19.5% of the requested data sets, and only 11% for articles published before 2000. We

found that several factors contribute to these low proportions: nonworking e-mails, a 50% response rate, and sometimes the lack of an informative response from the authors. However, when the authors did give the status of their data, the proportion of data sets that still existed dropped from 100% in 2011 to 33% in 1991 ([Figure 1D](#)). Unfortunately, many of these missing data

sets could be retrieved only with considerable effort by the authors, and others are completely lost to science.

Many data sets produced in scientific research are unique to their time and location, and once lost they cannot be replaced [14]. Since it is impossible to know what uses would have been found for these data or when they would become important, leaving their preservation to authors denies future researchers any chance of reusing them. Fortunately, one effective solution is to require that authors share it on a public archive at publication: the data will be preserved in perpetuity and can no longer be withheld or lost by authors. Some journals have already enacted policies to this effect (e.g., [5, 6]), and we hope that the worrying magnitude of the issues reported here will encourage others to draft similar policies in due course.

**Experimental Procedures**

It is likely that expectations on data sharing will differ between academic communities and that some data types are easier to preserve than others. Moreover, the types of data being collected change through time. We attempted to control for these effects by focusing on a single type of data that has been collected in the same way for many decades: data on morphological dimensions from plants or animals, as is typically collected by biologists and taxonomists. We are also conducting a parallel study on how the reproducibility of statistical analyses changes through time, and this study is working on reproducing DFAs, which are commonly applied to morphometric data [15]. We therefore also set the condition that the data must have been used in a DFA. Since this study is focused on the data from publicly available articles and not individual researchers, we determined that it did not require review by a research ethics board (Article 2.1, second edition of the Tri-Council Policy Statement: Ethical Conduct for Research Involving Humans).

We searched Web of Science for articles matching “morpholog\* and discriminant\*” in the topic field for the years 1980 to 2011. Only 24 papers were identified before 1991, and these were excluded. To reduce the total number of articles, we chose to focus on odd years from 1991 to 2011, leaving 1,009 papers. These papers were randomly assigned to the working group for data collection. Papers were excluded if the article text was

not available to us either online or via the University of British Columbia library, if the analysis did not include morphological data from a biological organism, or if the paper did not report the results of a DFA. Papers were also excluded if the data were already available as a supplementary file or appendix or on another website, as curation of these data sets is no longer the responsibility of the author. Due to the effort involved in checking all 1,009 papers for details on analysis and author contact information, we stopped data collection after a random subset of 526 papers had been assessed. Of these, ten did not meet the inclusion criteria (e.g., were not DFAs on morphology or had data already available in a supplementary file or appendix) and were dropped. This left 516 papers, with a minimum of 26 papers for any given year, and over 40 for most years (Table 1). Interestingly, we found that only 2.4% (13 of 529) of otherwise eligible papers had made their data available at publication: one paper each from 1999, 2001, 2003, and 2007, three papers in 2005, two in 2009, and four in 2011.

Data collected from the papers included information on the DFA used, the results (for the reproducibility analyses), and author contact information. In every case, we attempted to find e-mail addresses for the first, corresponding, and last authors of every paper. Often these were not mutually exclusive (e.g., a single author), and there were many different combinations. We attempted to extract the e-mails from the article text, but quickly determined that older papers would be more likely to have nonworking e-mail addresses [12] or no e-mails at all. We therefore also searched online for a maximum of 5 min per author for a recent or current e-mail address.

We used R [16] to generate data request e-mails, with all available e-mail addresses in the “to:” field, and used an R script to automatically send them out on April 15, 2013. Reminder e-mails were sent out to unresponsive authors 3 weeks later (May 8, 2013). When authors replied asking for more information, we provided additional details as required. The text for these two e-mails is included in the [Supplemental Experimental Procedures](#). The recording period for author responses ended on June 5, 2013, and the papers were sorted into different outcomes: (1) all e-mail addresses generated an error message, (2) no response was received, (3) a response was received but gave no information about the status of the data, (4) data were lost or stored on obsolete hardware, and (5) the authors had the data but were unwilling to share or (6) the data were received. Since outcomes (5) and (6) both implied that the data set still existed, we combined these into a single outcome, “data extant.”

We used logistic regression to investigate the relationship between the age of a paper and the probability that the data were still extant. Further subanalyses were conducted on subsets of the data to investigate the relationships between the age of the paper and (1) the probability that at least one e-mail appeared to work, (2) the conditional probability of a response given that at least one e-mail appeared to work, (3) the conditional probability of getting a response that indicated the status of the data (data lost, data exist but unwilling to share, or data shared) given that a response was received, and, finally, (4) the conditional probability that the data was extant given that an informative response was received. We also used Poisson regressions to investigate the relationship between article age and the number of e-mails found in the paper or online. Lastly, logistic regressions were used to examine how article age affected the chance that an e-mail address appeared to work. All analyses were carried out in R 3.0.1 [16].

#### Accession Numbers

The analysis code and data are available on Dryad under DOI number 10.5061/dryad.q3g37.

#### Supplemental Information

Supplemental Information includes Supplemental Experimental Procedures and can be found with this article online at <http://dx.doi.org/10.1016/j.cub.2013.11.014>.

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#### References

1. Holdren, J.P. (2013). Increasing Access to the Results of Federally Funded Scientific Research. [http://www.whitehouse.gov/sites/default/files/microsites/ostp/ostp\\_public\\_access\\_memo\\_2013.pdf](http://www.whitehouse.gov/sites/default/files/microsites/ostp/ostp_public_access_memo_2013.pdf).
2. Biotechnology and Biological Sciences Research Council (2007). BBSRC Data Sharing Policy: Version 1.1 (June 2010 Update). <http://www.bbsrc.ac.uk/datasharing>.
3. National Institutes of Health (2003). NIH Data Sharing Policy and Implementation Guidance. [http://grants.nih.gov/grants/policy/data\\_sharing/data\\_sharing\\_guidance.htm](http://grants.nih.gov/grants/policy/data_sharing/data_sharing_guidance.htm).
4. Thorley, M. (2010). NERC DataPolicy – GuidanceNotes. <http://www.nerc.ac.uk/research/sites/data/documents/datapolicy-guidance.pdf>.
5. Whitlock, M.C., McPeck, M.A., Rausher, M.D., Rieseberg, L., and Moore, A.J. (2010). Data archiving. *Am. Nat.* 175, 145–146.
6. Groves, T. (2010). BMJ policy on data sharing. *BMJ* 340, c564.
7. Michener, W.K., Brunt, J.W., Helly, J.J., Kirchner, T.B., and Stafford, S.G. (1997). Nongeospatial metadata for the ecological sciences. *Ecol. Appl.* 7, 330–342.
8. Leberg, P.L., and Neigel, J.E. (1999). Enhancing the retrievability of population genetic survey data? An assessment of animal mitochondrial DNA studies. *Evolution* 53, 1961–1965.
9. Wicherts, J.M., Borsboom, D., Kats, J., and Molenaar, D. (2006). The poor availability of psychological research data for reanalysis. *Am. Psychol.* 61, 726–728.
10. Savage, C.J., and Vickers, A.J. (2009). Empirical study of data sharing by authors publishing in PLoS journals. *PLoS ONE* 4, e7078.
11. Vines, T.H., Andrew, R.L., Bock, D.G., Franklin, M.T., Gilbert, K.J., Kane, N.C., Moore, J.-S., Moyers, B.T., Renaut, S., Rennison, D.J., et al. (2013). Mandated data archiving greatly improves access to research data. *FASEB J.* 27, 1304–1308.
12. Wren, J.D., Grissom, J.E., and Conway, T. (2006). E-mail decay rates among corresponding authors in MEDLINE. The ability to communicate with and request materials from authors is being eroded by the expiration of e-mail addresses. *EMBO Rep.* 7, 122–127.
13. Haak, L.L., Fenner, M., Paglione, L., Pentz, E., and Ratner, H. (2012). ORCID: a system to uniquely identify researchers. *Learn. Publ.* 25, 259–264.
14. Wolkovich, E.M., Regetz, J., and O'Connor, M.I. (2012). Advances in global change research require open science by individual researchers. *Glob. Change Biol.* 18, 2102–2110.
15. Rayment, R.A., Blackith, R.E., and Campbell, N.A. (1984). *Multivariate Morphometrics*, Second Edition (London: Academic Press).
16. R Core Team (2013). R: A language and environment for statistical computing (Vienna, Austria: R Foundation for Statistical Computing), URL <http://www.R-project.org/>.