

1	Justify Your Alpha
2 3	In Press, Nature Human Behavior
4	Daniel Lakens*1, Federico G. Adolfi ² , Casper J. Albers ³ , Farid Anvari ⁴ , Matthew A. J. Apps ⁵ ,
5	Shlomo E. Argamon ⁶ , Thom Baguley ⁷ , Raymond B. Becker ⁸ , Stephen D. Benning ⁹ , Daniel E.
6	Bradford ¹⁰ , Erin M. Buchanan ¹¹ , Aaron R. Caldwell ¹² , Ben van Calster ¹³ , Rickard Carlsson ¹⁴ ,
7	Sau-Chin Chen ¹⁵ , Bryan Chung ¹⁶ , Lincoln J Colling ¹⁷ , Gary S. Collins ¹⁸ , Zander Crook ¹⁹ ,
8	Emily S. Cross ²⁰ , Sameera Daniels ²¹ , Henrik Danielsson ²² , Lisa DeBruine ²³ , Daniel J.
9	Dunleavy ²⁴ , Brian D. Earp ²⁵ , Michele I. Feist ^{26,} Jason D. Ferrell ²⁷ , James G. Field ²⁸ , Nicholas
10	W. Fox ²⁹ , Amanda Friesen ³⁰ , Caio Gomes ³¹ , Monica Gonzalez-Marquez ³² , James A.
11	Grange ³³ , Andrew P. Grieve ³⁴ , Robert Guggenberger ³⁵ , James Grist ³⁶ , Anne-Laura van
12	Harmelen ³⁷ , Fred Hasselman ³⁸ , Kevin D. Hochard ³⁹ , Mark R. Hoffarth ⁴⁰ , Nicholas P.
13	Holmes ⁴¹ , Michael Ingre ⁴² , Peder M. Isager ⁴³ , Hanna K. Isotalus ⁴⁴ , Christer Johansson ⁴⁵ ,
14	Konrad Juszczyk ⁴⁶ , David A. Kenny ⁴⁷ , Ahmed A. Khalil ⁴⁸ , Barbara Konat ⁴⁹ , Junpeng Lao ⁵⁰ ,
15	Erik Gahner Larsen ⁵¹ , Gerine M. A. Lodder ⁵² , Jiří Lukavský ⁵³ , Christopher R. Madan ⁵⁴ , David
16	Manheim ⁵⁵ , Stephen R. Martin ⁵⁶ , Andrea E. Martin ⁵⁷ , Deborah G. Mayo ⁵⁸ , Randy J.
17	McCarthy ⁵⁹ , Kevin McConway ⁶⁰ , Colin McFarland ⁶¹ , Amanda Q. X. Nio ⁶² , Gustav Nilsonne ⁶³ ,
18	Cilene Lino de Oliveira ⁶⁴ , Jean-Jacques Orban de Xivry ⁶⁵ , Sam Parsons ⁶⁶ , Gerit Pfuhl ⁶⁷ ,
19	Kimberly A. Quinn ⁶⁸ , John J. Sakon ⁶⁹ , S. Adil Saribay ⁷⁰ , Iris K. Schneider ⁷¹ , Manojkumar
20	Selvaraju ⁷² , Zsuzsika Sjoerds ⁷³ , Samuel G. Smith ⁷⁴ , Tim Smits ⁷⁵ , Jeffrey R. Spies ⁷⁶ , Vishnu
21	Sreekumar ⁷⁷ , Crystal N. Steltenpohl ⁷⁸ , Neil Stenhouse ⁷⁹ , Wojciech Świątkowski ⁸⁰ , Miguel A.
22	Vadillo ⁸¹ , Marcel A. L. M. Van Assen ⁸² , Matt N. Williams ⁸³ , Samantha E. Williams ⁸⁴ , Donald
23	R. Williams ⁸⁵ , Tal Yarkoni ⁸⁶ , Ignazio Ziano ⁸⁷ , Rolf A. Zwaan ⁸⁸
24	
25	Affiliations
26	
27	*1Human-Technology Interaction, Eindhoven University of Technology, Den Dolech,
~~	

28 5600MB, Eindhoven, The Netherlands

1	² Laboratory of Experimental Psychology and Neuroscience (LPEN), Institute of Cognitive
2	and Translational Neuroscience (INCYT), INECO Foundation, Favaloro University,
3	Pacheco de Melo 1860, Buenos Aires, Argentina
4	² National Scientific and Technical Research Council (CONICET), Godoy Cruz 2290, Buenos
5	Aires, Argentina
6	³ Heymans Institute for Psychological Research, University of Groningen, Grote Kruisstraat
7	2/1, 9712TS Groningen, The Netherlands
8	⁴ College of Education, Psychology & Social Work, Flinders University, Adelaide, GPO Box
9	2100, Adelaide, SA, 5001, Australia
10	⁵ Department of Experimental Psychology, University of Oxford, New Radcliffe House,
11	Oxford, OX2 6GG, UK
12	⁶ Department of Computer Science, Illinois Institute of Technology, Chicago, IL, 10 W. 31st
13	Street, Chicago, IL 60645, USA
14	⁷ Department of Psychology, Nottingham Trent University, Nottingham, 50 Shakespeare
15	Street, Nottingham, NG1 4FQ, UK
16	⁸ Faculty of Linguistics and Literature, Bielefeld University, Bielefeld, Universitätsstraße 25,
17	33615 Bielefeld, Germany
18	⁹ Psychology, University of Nevada, Las Vegas, Las Vegas, 4505 S. Maryland Pkwy., Box
19	455030, Las Vegas, NV 89154-5030, USA
20	¹⁰ Psychology, University of Wisconsin-Madison, Madison, 1202 West Johnson St. Madison
21	WI. 53706, USA
22	¹¹ Psychology, Missouri State University, 901 S. National Ave, Springfield, MO, 65897, USA
23	¹² Health, Human Performance, and Recreation, University of Arkansas, Fayetteville, 155
24	Stadium Drive, HPER 321, Fayetteville, AR, 72701, USA
25	¹³ Department of Development and Regeneration, KU Leuven, Leuven, Herestraat 49 box
26	805, 3000 Leuven, Belgium, Belgium
27	¹³ Department of Medical Statistics and Bioinformatics, Leiden University Medical Center,
28	Postbus 9600, 2300 RC, Leiden, The Netherlands

1	¹⁴ Department of Psychology, Linnaeus University, Kalmar, Stagneliusgatan 14, 392 34,
2	Kalmar, Sweden
3	¹⁵ Department of Human Development and Psychology, Tzu-Chi University, No. 67, Jieren
4	St., Hualien City, Hualien County, 97074, Taiwan
5	¹⁶ Department of Surgery, University of British Columbia, Victoria, #301 - 1625 Oak Bay Ave,
6	Victoria BC Canada, V8R 1B1, Canada
7	¹⁷ Department of Psychology, University of Cambridge, Cambridge CB2 3EB, UK
8	¹⁸ Centre for Statistics in Medicine, University of Oxford, Windmill Road, Oxford, OX3 7LD,
9	UK
10	¹⁹ Department of Psychology, The University of Edinburgh, 7 George Square, Edinburgh, EH8
11	9JZ, UK
12	²⁰ School of Psychology, Bangor University, Bangor, Adeilad Brigantia, Bangor, Gwynedd,
13	LL57 2AS, UK
14	²¹ Ramsey Decision Theoretics, 4849 Connecticut Ave. NW #132, Washington, DC 20008,
15	USA
16	²² Department of Behavioural Sciences and Learning, Linköping University, SE-581 83,
17	Linköping, Sweden
18	²³ Institute of Neuroscience and Psychology, University of Glasgow, Glasgow, 58 Hillhead
19	Street, UK
20	²⁴ College of Social Work, Florida State University, 296 Champions Way, University Center C,
21	Tallahassee, FL, 32304, USA
22	²⁵ Departments of Psychology and Philosophy, Yale University, 2 Hillhouse Ave, New Haven
23	CT 06511, USA
24	²⁶ Department of English, University of Louisiana at Lafayette, P. O. Box 43719, Lafayette LA
25	70504, USA
26	²⁷ Department of Psychology, St. Edward's University, 3001 S. Congress, Austin, TX 78704,
27	USA

1	²⁷ Department of Psychology, University of Texas at Austin, 108 E. Dean Keeton Stop A8000,
2	Austin, TX 78712-1043, USA
3	²⁸ Department of Management, West Virginia University, 1602 University Avenue,
4	Morgantown, WV 26506, USA
5	²⁹ Department of Psychology, Rutgers University, New Brunswick, 53 Avenue E, Piscataway
6	NJ 08854, USA
7	³⁰ Department of Political Science, Indiana University Purdue University, Indianapolis,
8	Indianapolis, 425 University Blvd CA417, Indianapolis, IN 46202, USA
9	³¹ Booking.com, Herengracht 597, 1017 CE Amsterdam, The Nederlands
10	³² Department of English, American and Romance Studies, RWTH - Aachen University,
11	Aachen, Kármánstraße 17/19, 52062 Aachen, Germany
12	³³ School of Psychology, Keele University, Keele, Staffordshire, ST5 5BG, UK
13	³⁴ Centre of Excellence for Statistical Innovation, UCB Celltech, 208 Bath Road, Slough,
14	Berkshire SL1 3WE, UK
15	³⁵ Translational Neurosurgery, Eberhard Karls University Tübingen, Tübingen, Germany
16	³⁵ University Tübingen, International Centre for Ethics in Sciences and Humanities, Germany
17	³⁶ Department of Radiology, University of Cambridge, Box 218, Cambridge Biomedical
18	Campus, CB2 0QQ, UK
19	³⁷ Department of Psychiatry, University of Cambridge, Cambridge, 18b Trumpington Road,
20	CB2 8AH, UK
21	³⁸ Behavioural Science Institute, Radboud University Nijmegen, Montessorilaan 3, 6525 HR,
22	Nijmegen, The Netherlands
23	³⁹ Department of Psychology, University of Chester, Chester, Department of Psychology,
24	University of Chester, Chester, CH1 4BJ, UK
25	⁴⁰ Department of Psychology, New York University, 4 Washington Place, New York, NY
26	10003, USA
27	⁴¹ School of Psychology, University of Nottingham, Nottingham, University Park, NG7 2RD,
28	UK

1	⁴² None, Independent, Stockholm, Skåpvägen 5, 12245 ENSKEDE, Sweden
2	⁴³ Department of Clinical and Experimental Medicine, University of Linköping, 581 83
3	Linköping,, Sweden
4	⁴⁴ School of Clinical Sciences, University of Bristol, Bristol, Level 2 academic offices, L&R
5	Building, Southmead Hospital, BS10 5NB, UK
6	⁴⁵ Occupational Orthopaedics and Research, Sahlgrenska University Hospital, 413 45
7	Gothenburg, Sweden
8	⁴⁶ The Faculty of Modern Languages and Literatures, Institute of Linguistics, Psycholinguistics
9	Department, Adam Mickiewicz University, Al. Niepodległości 4, 61-874, Poznań, Poland
10	⁴⁷ Department of Psychological Sciences, University of Connecticut, Storrs, CT, Department
11	of Psychological Sciences, U-1020, Storrs, CT 06269-1020, USA
12	⁴⁸ Center for Stroke Research Berlin, Charité - Universitätsmedizin Berlin, Hindenburgdamm
13	30, 12200 Berlin, Germany
14	⁴⁸ Max Planck Institute for Human Cognitive and Brain Sciences, Stephanstraße 1a, 04103
15	Leipzig, Germany
16	⁴⁸ Berlin School of Mind and Brain, Humboldt-Universität zu Berlin, Luisenstraße 56, 10115
17	Berlin, Germany
18	⁴⁰ Social Sciences, Adam Mickiewicz University, Poznań, Szamarzewskiego 89, 60-568
19	Poznan, Poland
20	⁵⁰ Department of Psychology, University of Fribourg, Faucigny 2, 1700 Fribourg, Switzerland
21	⁵¹ School of Politics and International Relations, University of Kent, Canterbury CT2 7NX, UK
22	⁵² Department of Sociology / ICS, University of Groningen, Grote Rozenstraat 31, 9712 TG
23	Groningen, The Netherlands
24	⁵³ Institute of Psychology, Czech Academy of Sciences, Hybernská 8, 11000 Prague, Czech
25	Republic
26	⁵⁴ School of Psychology, University of Nottingham, Nottingham, NG7 2RD, UK
27	⁵⁵ Pardee RAND Graduate School, RAND Corporation, 1200 S Hayes St, Arlington, VA
28	22202, USA

1 ⁵⁶Psychology and Neuroscience, Baylor University, Waco, One Bear Place 97310, Waco TX,

2 USA

- 3 ⁵⁷Psychology of Language Department, Max Planck Institute for Psycholinguistics, Nijmegen,
- 4 Wundtlaan 1, 6525XD, The Netherlands
- 5 ⁵⁷Department of Psychology, School of Philosophy, Psychology, and Language Sciences,
- 6 University of Edinburgh, 7 George Square, EH8 9JZ Edinburgh, UK
- 7 ⁵⁸Dept of Philosophy, Major Williams Hall, Virginia Tech, Blacksburg, VA, US
- 8 ⁵⁹Center for the Study of Family Violence and Sexual Assault, Northern Illinois University,
- 9 DeKalb, IL, 125 President's BLVD., DeKalb, IL 60115, USA
- 10 ⁶⁰School of Mathematics and Statistics, The Open University, Milton Keynes, Walton Hall,
- 11 Milton Keynes MK7 6AA, UK
- 12 ⁶¹Skyscanner, 15 Laurison Place, Edinburgh, EH3 9EN, UK
- 13 ⁶²School of Biomedical Engineering and Imaging Sciences, King's College London, London,
- 14 UK
- 15 ⁶³Stress Research Institute, Stockholm University, Stockholm, Frescati Hagväg 16A, SE-
- 16 10691 Stockholm, Sweden
- ⁶³Department of Clinical Neuroscience, Karolinska Institutet, Nobels väg 9, SE-17177
- 18 Stockholm, Sweden
- ⁶³Department of Psychology, Stanford University, 450 Serra Mall, Stanford, CA 94305, USA
- 20 ⁶⁴Laboratory of Behavioral Neurobiology, Department of Physiological Sciences, Federal
- 21 University of Santa Catarina, Florianópolis, Campus Universitário Trindade, 88040900,
- 22 Brazil
- 23 ⁶⁵Department of Kinesiology, KU Leuven, Leuven, Tervuursevest 101 box 1501, B-3001
- 24 Leuven, Belgium
- 25 ⁶⁶Department of Experimental Psychology, University of Oxford, Oxford, UK
- ⁶⁷Department of Psychology, UiT The Arctic University of Norway, Tromsø, Norway
- ⁶⁸Department of Psychology, DePaul University, Chicago, 2219 N Kenmore Ave, Chicago, IL
- 28 60657, USA

1	⁶⁹ Center for Neural Science, New York University, 4 Washington PI Room 809 New York, NY
2	10003, USA
3	⁷⁰ Department of Psychology, Boğaziçi University, Bebek, 34342, Istanbul, Turkey
4	⁷¹ Psychology, University of Cologne, Cologne, Herbert-Lewin-St. 2, 50931, Cologne,
5	Germany
6	⁷² Saudi Human Genome Program, King Abdulaziz City for Science and Technology
7	(KACST); Integrated Gulf Biosystems, Riyadh, Saudi Arabia
8	⁷³ Cognitive Psychology Unit, Institute of Psychology, Leiden University, Wassenaarseweg
9	52, 2333 AK Leiden, The Netherlands
10	⁷³ Leiden Institute for Brain and Cognition, Leiden University, Leiden, The Netherlands
11	⁷⁴ Leeds Institute of Health Sciences, University of Leeds, Leeds, LS2 9NL, UK
12	⁷⁵ Institute for Media Studies, KU Leuven, Leuven, Belgium
13	⁷⁶ Center for Open Science, 210 Ridge McIntire Rd Suite 500, Charlottesville, VA 22903, USA
14	⁷⁶ Department of Engineering and Society, University of Virginia, Thornton Hall, P.O. Box
15	400259, Charlottesville, VA 22904, USA
16	⁷⁷ Surgical Neurology Branch, National Institute of Neurological Disorders and Stroke,
17	National Institutes of Health, Bethesda, MD 20892, USA
18	⁷⁸ Department of Psychology, University of Southern Indiana, 8600 University Boulevard,
19	Evansville, Indiana, USA
20	⁷⁹ Life Sciences Communication, University of Wisconsin-Madison, Madison, Wisconsin, 1545
21	Observatory Drive, Madison, WI 53706, USA
22	⁸⁰ Department of Social Psychology, Institute of Psychology, University of Lausanne, Quartier
23	UNIL-Mouline, Bâtiment Géopolis, CH-1015 Lausanne, Switzerland
24	⁸¹ Departamento de Psicología Básica, Universidad Autónoma de Madrid, c/ Ivan Pavlov 6,
25	28049 Madrid, Spain
26	⁸² Department of Methodology and Statistics, Tilburg University, Warandelaan 2, 5000 LE
27	Tilburg, The Netherlands

1 ⁸²Department of Sociology, Utrecht University, Padualaan 14, 3584 CH, Utrecht, The 2 Netherlands 3 ⁸³School of Psychology, Massey University, Auckland, Private Bag 102904, North Shore, 4 Auckland, 0745, New Zealand 5 ⁸⁴Psychology, Saint Louis University, St. Louis, MO, 3700 Lindell Blvd, St. Louis, MO 63108, 6 USA 7 ⁸⁵Psychology, University of California, Davis, Davis, One Shields Ave, Davis, CA 95616, USA 8 ⁸⁶Department of Psychology, University of Texas at Austin, 108 E. Dean Keeton Stop A8000, 9 Austin, TX 78712-1043, USA 10 ⁸⁷Marketing Department, Ghent University, Tweekerkenstraat 2, 9000 Ghent, Belgium 11 ⁸⁸Department of Psychology, Education, and Child Studies, Erasmus University Rotterdam, 12 Rotterdam, Burgemeester Oudlaan 50, 3000 DR, Rotterdam, The Netherlands 13 14 Author Contributions. Daniel Lakens, Nicholas W. Fox, Monica Gonzalez-Marguez, James 15 A. Grange, Nicholas P. Holmes, Ahmed A. Khalil, Stephen R. Martin, Vishnu Sreekumar, 16 and Crystal N. Steltenpohl participated in brainstorming, drafting the commentary, and data-17 analysis. Casper J. Albers, Shlomo E. Argamon, Thom Baguley, Erin M. Buchanan, Ben van Calster, Zander Crook, Sameera Daniels, Daniel J. Dunleavy, Brian D. Earp, Jason D. 18 19 Ferrell, James G. Field, Anne-Laura van Harmelen, Michael Ingre, Peder M. Isager, Hanna 20 K. Isotalus, Junpeng Lao, Gerine M. A. Lodder, David Manheim, Andrea E. Martin, Kevin 21 McConway, Amanda Q. X. Nio, Gustav Nilsonne, Cilene Lino de Oliveira, Jean-Jacques 22 Orban de Xivry, Gerit Pfuhl, Kimberly A. Quinn, Iris K. Schneider, Zsuzsika Sjoerds, Samuel 23 G. Smith, Jeffrey R. Spies, Marcel A. L. M. Van Assen, Matt N. Williams, Donald R. Williams, 24 Tal Yarkoni, and Rolf A. Zwaan participated in brainstorming and drafting the commentary. 25 Federico G. Adolfi, Raymond B. Becker, Michele I. Feist, and Sam Parsons participated in 26 drafting the commentary, and data-analysis. Matthew A. J. Apps, Stephen D. Benning,

27 Daniel E. Bradford, Sau-Chin Chen, Bryan Chung, Lincoln J Colling, Henrik Danielsson, Lisa

28 DeBruine, Mark R. Hoffarth, Erik Gahner Larsen, Randy J. McCarthy, John J. Sakon, S. Adil

1 Saribay, Tim Smits, Neil Stenhouse, Wojciech Światkowski, and Miguel A. Vadillo 2 participated in brainstorming. Farid Anvari, Aaron R. Caldwell, Rickard Carlsson, Emily S. 3 Cross, Amanda Friesen, Caio Gomes, Andrew P. Grieve, Robert Guggenberger, James 4 Grist, Kevin D. Hochard, Christer Johansson, Konrad Juszczyk, David A. Kenny, Barbara 5 Konat, Jiří Lukavský, Christopher R. Madan, Deborah G. Mayo, Colin McFarland, 6 Manojkumar Selvaraju, Samantha E. Williams, and Ignazio Ziano did not participate in 7 drafting the commentary because the points that they would have raised had already been 8 incorporated into the commentary, or endorse a sufficiently large part of the contents as if 9 participation had occurred. Except for the first author, authorship order is alphabetical.

10

11 Acknowledgements: We'd like to thank Dale Barr, Felix Cheung, David Colquhoun, Hans 12 IJzerman, Harvey Motulsky, and Richard Morey for helpful discussions while drafting this 13 commentary. Daniel Lakens was supported by NWO VIDI 452-17-013. Federico G. Adolfi 14 was supported by CONICET. Matthew Apps was funded by a Biotechnology and Biological 15 Sciences Research Council AFL Fellowship (BB/M013596/1). Gary Collins was supported by 16 the NIHR Biomedical Research Centre, Oxford. Zander Crook was supported by the 17 Economic and Social Research Council [grant number C106891X]. Emily S. Cross was 18 supported by the European Research Council (ERC-2015-StG-677270). Lisa DeBruine is 19 supported by the European Research Council (ERC-2014-CoG-647910 KINSHIP). Anne-20 Laura van Harmelen is funded by a Royal Society Dorothy Hodgkin Fellowship (DH150176). 21 Mark R. Hoffarth was supported by the National Science Foundation under grant SBE 22 SPRF-FR 1714446. Junpeng Lao was supported by the SNSF grant 100014 156490/1. 23 Cilene Lino de Oliveira was supported by AvH, Capes, CNPq. Andrea E. Martin was 24 supported by the Economic and Social Research Council of the United Kingdom [grant 25 number ES/K009095/1]. Jean-Jacques Orban de Xivry is supported by an internal grant from 26 the KU Leuven (STG/14/054) and by the Fonds voor Wetenschappelijk Onderzoek 27 (1519916N). Sam Parsons was supported by the European Research Council (FP7/2007-28 2013; ERC grant agreement no; 324176). Gerine Lodder was funded by NWO VICI 453-14-

1	016. Samuel Smith is supported by a Cancer Research UK Fellowship (C42785/A17965).
2	Vishnu Sreekumar was supported by the NINDS Intramural Research Program (IRP). Miguel
3	A. Vadillo was supported by Grant 2016-T1/SOC-1395 from Comunidad de Madrid. Tal
4	Yarkoni was supported by NIH award R01MH109682.
5	
6	Competing Interests: The authors declare no competing interests.
7	
8	Abstract : In response to recommendations to redefine statistical significance to $p \le .005$, we
9	propose that researchers should transparently report and justify all choices they make when
10	designing a study, including the alpha level.
11	

Justify Your Alpha Benjamin et al.¹ proposed changing the conventional "statistical significance" threshold (i.e., the alpha level) from $p \le .05$ to $p \le .005$ for all novel claims with relatively low prior odds.

5 They provided two arguments for why lowering the significance threshold would

- 6 "immediately improve the reproducibility of scientific research." First, a *p*-value near .05
- 7 provides weak evidence for the alternative hypothesis. Second, under certain assumptions,

8 an alpha of .05 leads to high false positive report probabilities (FPRP²; the probability that a

9 significant finding is a false positive).

10

1

2

3

4

11 We share their concerns regarding the apparent non-replicability of many scientific studies, 12 and agree that a universal alpha of .05 is undesirable. However, redefining "statistical 13 significance" to a lower, but equally arbitrary threshold, is inadvisable for three reasons: (1) 14 there is insufficient evidence that the current standard is a "leading cause of non-15 reproducibility"¹; (2) the arguments in favor of a blanket default of $p \le .005$ do not warrant the 16 immediate and widespread implementation of such a policy; and (3) a lower significance 17 threshold will likely have negative consequences not discussed by Benjamin and colleagues. 18 We conclude that the term "statistically significant" should no longer be used and suggest 19 that researchers employing null hypothesis significance testing justify their choice for an 20 alpha level before collecting the data, instead of adopting a new uniform standard.

21

22 Lack of evidence that $p \le .005$ improves replicability

23

Benjamin et al.¹ claimed that the expected proportion of replicable studies should be considerably higher for studies observing $p \le .005$ than for studies observing .005 ,due to a lower FPRP.*Theoretically*, replicability is related to the FPRP, and lower alphalevels will reduce false positive results in the literature. However,*in practice*, the impact oflowering alpha levels depends on several unknowns, such as the prior odds that the

examined hypotheses are true, the statistical power of studies, and the (change in) behavior
of researchers in response to any modified standards.

3

4 An analysis of the results of the Reproducibility Project: Psychology³ showed that 49% 5 (23/47) of the original findings with *p*-values below .005 yielded $p \le .05$ in the replication 6 study, whereas only 24% (11/45) of the original studies with $.005 yielded <math>p \le .05$ 7 $(\chi^2(1) = 5.92, p = .015, BF_{10} = 6.84)$. Benjamin and colleagues presented this as evidence of 8 "potential gains in reproducibility that would accrue from the new threshold." According to 9 their own proposal, however, this evidence is only "suggestive" of such a conclusion, and 10 there is considerable variation in replication rates across *p*-values (see Figure 1). 11 Importantly, lower replication rates for p-values just below .05 are likely confounded by p-12 hacking (the practice of flexibly analyzing data until the p-value passes the "significance" 13 threshold). Thus, the differences in replication rates between studies with .00514 compared to those with $p \le .005$ may not be entirely due to the level of evidence. Further 15 analyses are needed to explain the low (49%) replication rate of studies with $p \le .005$, before 16 this alpha level is recommended as a new significance threshold for novel discoveries 17 across scientific disciplines.

18

19 Weak justifications for the α = .005 threshold

20

21 We agree with Benjamin et al. that single *p*-values close to .05 never provide strong 22 "evidence" against the null hypothesis. Nonetheless, the argument that p-values provide 23 weak evidence based on Bayes factors has been questioned⁴. Given that the marginal 24 likelihood is sensitive to different choices for the models being compared, redefining alpha 25 levels as a function of the Bayes factor is undesirable. For instance, Benjamin and 26 colleagues stated that *p*-values of .005 imply Bayes factors between 14 and 26. However, 27 these upper bounds only hold for a Bayes factor based on a point null model and when the 28 p-value is calculated for a two-sided test, whereas one-sided tests or Bayes factors for non-

point null models would imply different alpha thresholds. When a test yields BF = 25 the data are interpreted as strong relative evidence for a specific alternative (e.g., $\mu = 2.81$), while a *p* $\leq .005$ only warrants the more modest rejection of a null effect without allowing one to reject even small positive effects with a reasonable error rate⁵. Benjamin et al. provided no rationale for why the new *p*-value threshold *should* align with equally arbitrary Bayes factor thresholds. We question the idea that the alpha level at which an error rate is controlled should be based on the amount of relative evidence indicated by Bayes factors.

8

9 The second argument for α = .005 is that the FPRP can be high with α = .05. Calculating the 10 FPRP requires a definition of the alpha level, the power of the tests examining true effects, 11 and the ratio of true to false hypotheses tested (the prior odds). Figure 2 in Benjamin et al. 12 displays FPRPs for scenarios where most hypotheses are false, with prior odds of 1:5, 1:10, 13 and 1:40. The recommended $p \le .005$ threshold reduces the *minimum* FPRP to less than 14 5%, assuming 1:10 prior odds (the true FPRP might still be substantially higher in studies 15 with very low power). This prior odds estimate is based on data from the Reproducibility 16 Project: Psychology³ using an analysis modelling publication bias for 73 studies⁶. Without 17 stating the reference class for the "base-rate of true nulls" (e.g., does this refer to all 18 hypotheses in science, in a discipline, or by a single researcher?), the concept of "prior odds 19 that H1 is true" has little meaning. Furthermore, there is insufficient representative data to 20 accurately estimate the prior odds that researchers examine a true hypothesis, and thus, 21 there is currently no strong argument based on FPRP to redefine statistical significance.

22

23 How a threshold of $p \le .005$ might harm scientific practice

24

Benjamin et al. acknowledged that their proposal has strengths as well as weaknesses, but
believe that its "efficacy gains would far outweigh losses." We are not convinced and see at
least three likely negative consequences of adopting a lowered threshold.

28

1 Risk of fewer replication studies. All else being equal, lowering the alpha level requires larger 2 sample sizes and creates an even greater strain on already limited resources. Achieving 3 80% power with α = .005, compared to α = .05, requires a 70% larger sample size for 4 between-subjects designs with two-sided tests (88% for one-sided tests). While Benjamin et 5 al. propose α = .005 exclusively for "new effects" (and not replications), designing larger 6 original studies would leave fewer resources (i.e., time, money, participants) for replication 7 studies, assuming fixed resources overall. At a time when replications are already relatively 8 rare and unrewarded, lowering alpha to .005 might therefore reduce resources spent on 9 replicating the work of others. More generally, recommendations for evidence thresholds 10 need to carefully balance statistical and non-statistical considerations (e.g., the value of 11 evidence for a novel claim vs. the value of independent replications).

12

13 Risk of reduced generalisability and breadth. Requiring larger sample sizes across scientific 14 disciplines may exacerbate over-reliance on convenience samples (e.g., undergraduate 15 students, online samples). Specifically, without (1) increased funding, (2) a reward system 16 that values large-scale collaboration, and (3) clear recommendations for how to evaluate 17 research with sample size constraints, lowering the significance threshold could adversely 18 affect the breadth of research questions examined. Compared to studies that use 19 convenience samples, studies with unique populations (e.g., people with rare genetic 20 variants, patients with post-traumatic stress disorder) or with time- or resource-intensive data 21 collection (e.g., longitudinal studies) require considerably more research funds and effort to 22 increase the sample size. Thus, researchers may become less motivated to study unique 23 populations or collect difficult-to-obtain data, reducing the generalisability and breadth of 24 findings.

25

Risk of exaggerating the focus on single p-values. Benjamin et al.'s proposal risks (1)
reinforcing the idea that relying on *p*-values is a sufficient, if imperfect, way to evaluate
findings, and (2) discouraging opportunities for more fruitful changes in scientific practice

and education. Even though Benjamin et al. do not propose *p* ≤ .005 as a publication
threshold, some bias in favor of significant results will remain, in which case redefining *p* ≤
.005 as "statistically significant" would result in greater upward bias in effect size estimates.
Furthermore, it diverts attention from the cumulative evaluation of findings, such as
converging results of multiple (replication) studies.

6

7 No one alpha to rule them all

8

9 We have two key recommendations. First, we recommend that the label "statistically 10 significant" should no longer be used. Instead, researchers should provide more meaningful 11 interpretations of the theoretical or practical relevance of their results. Second, authors 12 should transparently specify-and justify-their design choices. Depending on their choice of 13 statistical approach, these may include the alpha level, the null and alternative models, 14 assumed prior odds, statistical power for a specified effect size of interest, the sample size, 15 and/or the desired accuracy of estimation. We do not endorse a single value for any design 16 parameter, but instead propose that authors justify their choices before data are collected. 17 Fellow researchers can then evaluate these decisions, ideally also prior to data collection, 18 for example, by reviewing a Registered Report submission⁷. Providing researchers (and 19 reviewers) with accessible information about ways to justify (and evaluate) design choices, 20 tailored to specific research areas, will improve current research practices.

21

Benjamin et al. noted that some fields, such as genomics and physics, have lowered the "default" alpha level. However, in genomics the overall false positive rate is still controlled at 5%; the lower alpha level is only used to correct for multiple comparisons. In physics, researchers have argued against a blanket rule, and for an alpha level based on factors such as the surprisingness of the predicted result and its practical or theoretical impact⁸. In non-human animal research, minimizing the number of animals used needs to be directly balanced against the probability and cost of false positives. Depending on these and other

considerations, the optimal alpha level for a given research question could be higher or
 lower than the current convention of .05^{9,10,11}.

3

4 Benjamin et al. stated that a "critical mass of researchers" endorse the standard of a $p \le p$ 5 .005 threshold for "statistical significance." However, the presence of a critical mass can only 6 be identified after a norm has been widely adopted, not before. Even if a $p \le .005$ threshold 7 were widely accepted, this would only reinforce the misconception that a single alpha level is 8 universally applicable. Ideally, the alpha level is determined by comparing costs and benefits against a utility function using decision theory¹². This cost-benefit analysis (and thus the 9 alpha level)¹³ differs when analyzing large existing datasets compared to collecting data from 10 11 hard-to-obtain samples.

12

13 Conclusion

14

15 Science is diverse, and it is up to scientists to justify the alpha level they decide to use. As 16 Fisher noted¹⁴: "...no scientific worker has a fixed level of significance at which, from year to 17 year, and in all circumstances, he rejects hypotheses; he rather gives his mind to each particular case in the light of his evidence and his ideas." Research should be guided by 18 19 principles of *rigorous science*¹⁵, not by heuristics and arbitrary blanket thresholds. These 20 principles include not only sound statistical analyses, but also experimental redundancy 21 (e.g., replication, validation, and generalisation), avoidance of logical traps, intellectual 22 honesty, research workflow transparency, and accounting for potential sources of error. 23 Single studies, regardless of their *p*-value, are never enough to conclude that there is strong 24 evidence for a substantive claim. We need to train researchers to assess cumulative 25 evidence and work towards an unbiased scientific literature. We call for a broader mandate 26 beyond p-value thresholds whereby all justifications of key choices in research design and 27 statistical practice are transparently evaluated, fully accessible, and pre-registered whenever 28 feasible.

1	References
2	
3	1. Benjamin, D. J., et al. Nature Human Behaviour 2, 6-10 https://doi.org/10.1038/s41562-
4	017-0189-z (2017).
5	2. Wacholder, S., Chanock, S., Garcia-Closas, M., El Ghormli, L., & Rothman, N. Journal of
6	the National Cancer Institute 96, 434-442 https://doi.org/10.1093/jnci/djh075 (2004).
7	3. Open Science Collaboration. (2015). Science 349 (6251), 1-8
8	https://doi.org/10.1126/science.aac4716 (2015).
9	4. Senn, S. Statistical issues in drug development (2nd ed). (John Wiley & Sons, 2007).
10	5. Mayo, D. Statistical inference as severe testing: How to get beyond the statistics wars.
11	(Cambridge University Press, 2018).
12	6. Johnson, V. E., Payne, R. D., Wang, T., Asher, A., & Mandal, S. Journal of the American
13	Statistical Association 112(517), 1–10
14	https://doi.org/10.1080/01621459.2016.1240079 (2017).
15	7. Chambers, C.D., Dienes, Z., McIntosh, R.D., Rotshtein, P., & Willmes, K. Cortex 66, A1-2
16	https://doi.org/10.1016/j.cortex.2015.03.022 (2015).
17	8. Lyons, L. Discovering the Significance of 5 sigma. Preprint at
18	http://arxiv.org/abs/1310.1284 (2013).
19	9. Field, S. A., Tyre, A. J., Jonzen, N., Rhodes, J. R., & Possingham, H. P. Ecology Letters
20	7(8), 669-675 https://doi.org/10.1111/j.1461-0248.2004.00625.x (2004).
21	10. Grieve, A. P. Pharmaceutical Statistics 14(2), 139–150 https://doi.org/10.1002/pst.1667
22	(2015).
23	11. Mudge, J. F., Baker, L. F., Edge, C. B., & Houlahan, J. E. PLOS ONE 7(2), e32734
24	https://doi.org/10.1371/journal.pone.0032734 (2012).
25	12. Skipper, J. K., Guenther, A. L., & Nass, G. The American Sociologist 2(1), 16–18 (1967).
26	13. Neyman, J., & Pearson, E. S. Philosophical Transactions of the Royal Society of London
27	A: Mathematical, Physical and Engineering Sciences 231 694–706
28	https://doi.org/10.1098/rsta.1933.0009 (1933).

- 1 14. Fisher R. A. Statistical methods and scientific inferences. (Hafner, 1956).
- 2 15. Casadevall, A., & Fang, F. C. mBio 7(6), e01902-16. https://doi.org/10.1128/mbio.01902-
- 3 16 (2016).

1 Figure Caption

- 2
- 3 *Figure 1*. The proportion of studies³ replicated at α = .05 (with a bin width of .005). Window
- 4 start and end positions are plotted on the horizontal axis. The error bars denote 95%
- 5 Jeffreys confidence intervals. R code to reproduce Figure 1 is available from
- 6 https://osf.io/by2kc/.

