CORRESPONDENCE

Issues with data transformation in genome-wide association studies for phenotypic variability [version 1; peer review: 2 approved]

Xia Shen¹, Lars Rönnegård^{2,3}

¹Division of Computational Genetics, Department of Clinical Sciences, Swedish University of Agricultural Sciences, Uppsala, SE-750 07, Sweden

²Statistics, School of Technology and Business Studies, Dalarna University, Falun, SE-791 88, Sweden

³Division of Quantitative Genetics, Department of Animal Breeding and Genetics, Swedish University of Agricultural Sciences, Uppsala, SE-750 07, Sweden

V1 First published: 02 Oct 2013, 2:200
<https://doi.org/10.12688/f1000research.2-200.v1>
Latest published: 02 Oct 2013, 2:200
<https://doi.org/10.12688/f1000research.2-200.v1>

Abstract

The purpose of this correspondence is to discuss and clarify a few points about data transformation used in genome-wide association studies, especially for phenotypic variability. By commenting on the recent publication by Sun *et al.* in the *American Journal of Human Genetics*, we emphasize the importance of statistical power in detecting functional loci and the real meaning of the scale of the phenotype in practice.

Open Peer Review

Approval Status

	1	2
version 1
02 Oct 2013	view	view

1. **William G. Hill**, University of Edinburgh, Edinburgh, UK
Ian White, University of Edinburgh,
2. **Yurii Aulchenko**, Siberian Division of the Russian Academy of Sciences, Novosibirsk, Russian Federation
Sodbo Sharapov, Novosibirsk State University, Novosibirsk, Russian Federation
Yakov Tsepilov , Novosibirsk State University, Novosibirsk, Russian Federation

Any reports and responses or comments on the article can be found at the end of the article.

Corresponding author: Xia Shen (xia.shen@slu.se)

Competing interests: The authors declare no competing interest.

Grant information: XS is funded by a Future Research Leaders grant from Swedish Foundation for Strategic Research (SSF) to Örjan Carlborg. LR is funded by the Swedish Research Council for Environment, Agricultural Sciences and Spatial Planning (FORMAS). *The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.*

Copyright: © 2013 Shen X and Rönnegård L. This is an open access article distributed under the terms of the [Creative Commons Attribution License](#), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited. Data associated with the article are available under the terms of the [Creative Commons Zero "No rights reserved" data waiver](#) (CC0 1.0 Public domain dedication).

How to cite this article: Shen X and Rönnegård L. **Issues with data transformation in genome-wide association studies for phenotypic variability [version 1; peer review: 2 approved]** F1000Research 2013, 2:200 <https://doi.org/10.12688/f1000research.2-200.v1>

First published: 02 Oct 2013, 2:200 <https://doi.org/10.12688/f1000research.2-200.v1>

Correspondence

Recently, Sun *et al.*¹ raised an interesting suggestion concerning the use of variance-stabilization transformations in genome-wide association studies (GWAS) for phenotypic variability. Specifically, Sun *et al.* revisited Yang *et al.*'s² results on the variability-controlling locus *FTO* for human body mass index (BMI) and claimed that the underlying variability across genotypes might not be as large as Yang *et al.* had seen. Although it was an important point that Sun *et al.* discussed, especially when quantitatively studying phenotypic variability has become such a hot topic, it is our opinion that there are some issues with the transformation approach that Sun *et al.* proposed.

First of all, if we take Sun *et al.*'s transformation according to Yang *et al.*'s phenotypic mean and variance per *FTO* genotype class, *i.e.* a one-to-one map through an inverse hyperbolic sine function, the BMI scale will become rather different compared with the ordinary measurement that we normally use (Figure 1). On the transformed scale of BMI, the difference between two persons who have a BMI of 24 and 25 kg/m² is much larger than that between two BMIs of 20 and 21 kg/m², which is strange in reality since the original BMI scale is what we commonly use and also what we care about. Sun *et al.*'s main argument here is that nearly all the measurement units are manmade. However, considering one of the traits of most interest, *e.g.* height, why should we regard the difference between 160cm and 170cm different from 170cm and 180cm? Although the definitions of most units can be arbitrary, some measurement scales do have meaning in real life.

Secondly, a key problem with Sun *et al.*'s transformation in practice is that such a transformation is marker-specific. Namely, when performing a GWAS, one needs to transform the phenotypic records differently for different markers, according to the phenotypic distribution across the genotypes per marker. This does not make much sense in practical analyses, because if there is a "best" scale of the phenotype, it should be used for all the markers across the genome, before testing the association between the phenotype and the markers. Using the tested marker to determine the transformation of the phenotype is strange. If a marker-specific transformation can be estimated, one should estimate a genome-specific transformation for GWAS, instead of doing different transformations marker-by-marker.

Thirdly, if the transformation of the phenotype is determined by one marker showing a significant effect on the phenotypic variability before testing the other markers, another significant effect on the phenotypic variability might be created due to such a transformation. In such a situation, it is problematic to decide which phenotypic scale we should choose.

Fourthly, several recent studies discussed that gene-gene or gene-environment interactions could cause significant variance heterogeneity across genotypes³⁻⁶, which makes testing variance-controlling loci a powerful tool to reveal potential interaction effects. Reducing the difference in variance across genotypes using a marker-specific variance-stabilization transformation would dramatically reduce such power. Regarding the biological sense of genetically regulated variance heterogeneity, empirical evidence has shown that a

Figure 1. Comparison of the original scale of body mass index (BMI) and the transformed scale using Sun *et al.*'s¹ transformation. The transformation was determined by the phenotypic distribution across *FTO* genotypes reported by Yang *et al.*².

single causal locus could show a much higher significant effect on variance compared with the mean⁶. In a particular population, such a locus may only be mappable through testing the variability rather than the magnitude of the phenotype.

The above issues cause us to question Sun *et al.*'s transformation in practice. The scale of the phenotype is certainly an important concern when interpreting an effect on phenotypic variability⁷. However, one needs to be careful for the points above before applying any transformation on the data. In particular, the statistical power in detecting functional loci and the real meaning of the scale used should be emphasized.

Author contributions

XS and LR initiated the study. XS performed the analysis. Both authors contributed to writing the report.

Competing interests

The authors declare no competing interest.

Grant information

XS is funded by a Future Research Leaders grant from Swedish Foundation for Strategic Research (SSF) to Örjan Carlberg. LR is funded by the Swedish Research Council for Environment, Agricultural Sciences and Spatial Planning (FORMAS).

The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

References

1. Sun X, Elston R, Morris N, *et al.*: **What Is the Significance of Difference in Phenotypic Variability across SNP Genotypes?** *Am J Hum Genet.* 2013; **93**(2): 390–397.
[PubMed Abstract](#) | [Publisher Full Text](#)
2. Yang J, Loos RJ, Powell JE, *et al.*: **FTO genotype is associated with phenotypic variability of body mass index.** *Nature.* 2012; **490**(7419): 267–272.
[PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)
3. Rönnegård L, Valdar W: **Detecting major genetic loci controlling phenotypic variability in experimental crosses.** *Genetics.* 2011; **188**(2): 435–447.
[PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)
4. Struchalin MV, Dehghan A, Witterman JC, *et al.*: **Variance heterogeneity analysis for detection of potentially interacting genetic loci: method and its limitations.** *BMC Genet.* 2010; **11**: 92.
[PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)
5. Paré G, Cook NR, Ridker PM, *et al.*: **On the use of variance per genotype as a tool to identify quantitative trait interaction effects: a report from the Women's Genome Health Study.** *PLoS Genet.* 2010; **6**(6): e1000981.
[PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)
6. Shen X, Pettersson M, Rönnegård L, *et al.*: **Inheritance Beyond Plain Heritability: Variance-Controlling Genes in Arabidopsis thaliana.** *PLoS Genet.* 2012; **8**(8): e1002839.
[PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)
7. Rönnegård L, Valdar W: **Recent developments in statistical methods for detecting genetic loci affecting phenotypic variability.** *BMC Genet.* 2012; **13**: 63.
[PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)

Open Peer Review

Current Peer Review Status:

Version 1

Reviewer Report 01 November 2013

<https://doi.org/10.5256/f1000research.2505.r1948>

© 2013 Aulchenko Y et al. This is an open access peer review report distributed under the terms of the [Creative Commons Attribution License](#), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Yurii Aulchenko

Institute of Cytology and Genetics, Siberian Division of the Russian Academy of Sciences,
Novosibirsk, Russian Federation

Yakov Tsepilov

Novosibirsk State University, Novosibirsk, Russian Federation

Sodbo Sharapov

Novosibirsk State University, Novosibirsk, Russian Federation

We agree with criticism raised by Shen and Ronnegard in their points 2 and 3 concerning the application of the transformation of Sun *et al.* in the context of whole-genome scans. Indeed, applying this transformation in SNP-specific manner is difficult to adopt conceptually. Sun *et al.* rightly suggest that “the scales on which we measure interval-scale quantitative traits are man-made and have little intrinsic biological relevance”, but the underlying intrinsic scale, and the function reflecting this scale into the observed, is likely to be unique and does not change with SNP. In that, the transformation applied to a trait should not change through the markers studied. Practically, this is not very difficult to implement, and as a simplest option one could think of the estimation of Sun’s transformation parameters from upper, middle and lower tertiles of the total phenotypic distribution. A more general approach (without restricting the data into three groups, but modelling the variance as a function of the mean) should be straightforward to implement.

We also understand the reasoning behind the Shen and Ronnegard’s points 1 and 4, but here we are less certain that the problem raised could be easily addressed. Specifically, one could argue with point 1 (“why should we regard the difference between 160cm and 170cm different from 170cm and 180cm?”): it is not that hard to imagine a biologically relevant model in which same changes of an “intrinsic scale” lead to different changes on the observed scale as the mean advances (an example would be Michaelis–Menten kinetics). Also both points 1 and 4 (losing power after transformation) relate not only to Sun *et al.*’s transformation, but to almost any transformation in wide use (e.g. Log, Box-Cox, Gaussienization/inverse-normal). While it is true that analysis of transformed trait may lead to reduced power (and specifically in case of Sun’s transformation applied in marker-specific manner to the analysis of variance heterogeneity it *should*), we have a feeling that one still would like to check whether the variance heterogeneity

found can be modeled as a function of the mean (in which case any SNP affecting the mean is likely to show “control” of the variance as well).

Finally, we fully agree with comment of William Hill and Ian White who criticize Sun *et al.*'s statement that “In the absence of genotypic mean differences, we can hardly infer that differences in variances are *per se* of biological interest”. We think that the differences in variance *per se* are biologically and genetically plausible and interesting.

Competing Interests: No competing interests were disclosed.

We confirm that we have read this submission and believe that we have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Reviewer Report 07 October 2013

<https://doi.org/10.5256/f1000research.2505.r1949>

© 2013 Hill et al. This is an open access peer review report distributed under the terms of the [Creative Commons Attribution License](#), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

William G. Hill

Institute of Evolutionary Biology, University of Edinburgh, Edinburgh, UK

Ian White

University of Edinburgh

Shen and Rönnegård (SR) comment critically and succinctly on the paper by Sun *et al.* published in AJHG which advocates that, before any claim of differences in variance among genotypes in a GWAS or similar study, a check should first be made whether these can be removed by a monotonic transformation. Each of SR's four criticisms seems well justified.

As 105 or more SNPs may be fitted in a GWAS study, what biological interpretation could be given to that number of different transformations or even on a limited subset of loci showing possible variance differences? If some loci give signals of mean but not variance difference, should these then be transformed to eliminate the scale effect on mean and perhaps reveal variance differences? Any concept of an original scale of measurement is lost, as SR point out. It is not obvious why the mere existence of a transformation designed to minimise differences in variance should prevent discussion of variance heterogeneity on the chosen scale. Equivalently, if we considered means of the three genotypes at the locus rather than just average effects, would our ability to transform the data at each locus such that heterozygotes were intermediate imply there was no dominance, or only that it was on a particular scale?

On a further point. Sun *et al.* (p395) comment: 'In the absence of genotypic mean differences, we can hardly infer that differences in variances are *per se* of biological interest.' That is to take too narrow a view: the mean and phenotypic variance (or CV) of a quantitative trait in any species take typical values, e.g. the CV for adult human height is ca. 4% and for BMI ca. 16%. There is direct

evidence of genetic differences within species in environmental variance, with GWAS and other single gene studies, that cannot be removed by scale, so the level of the environmental variance is subject to evolutionary forces (e.g. Hill & Mulder 2010 *Genet. Res.* 92:381). To view variance as a biological phenomenon which is just some adjunct to the mean seems simplistic, as SR argue. Indeed one has to ask whether scale transformations have value unless there is a biological basis, such as a log transformation to account for multiplicative genetic effects; but that must then apply across all loci.

Competing Interests: No competing interests were disclosed.

We confirm that we have read this submission and believe that we have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

The benefits of publishing with F1000Research:

- Your article is published within days, with no editorial bias
- You can publish traditional articles, null/negative results, case reports, data notes and more
- The peer review process is transparent and collaborative
- Your article is indexed in PubMed after passing peer review
- Dedicated customer support at every stage

For pre-submission enquiries, contact research@f1000.com

F1000Research