

# BGmisc: An R Package for Extended Behavior Genetics Analysis

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## Summary

Behavior genetics focuses on understanding genetic and environmental influences on individual differences. Traditionally, twin studies have been at the forefront of this discipline. However, research has moved beyond the classical twin design to embrace more complex family structures such as children of twins (CoT) (D'Onofrio et al., 2003), mother-daughter-aunt-niece (MDAN) (Rodgers et al., 2008), and other extended family designs. This expansion allows for a deeper, more nuanced exploration of genetic and environmental influences, but it also introduces challenges, particularly in data structuring and modeling.

A notable challenge arises when researchers venture into extended family data. The data structures inherent in such family designs are orders of magnitude larger than traditional designs. In the classical twin study, a family will consist of a single pair of twins (i.e., two people), whereas in the MDAN design a family consists of two mother-daughter pairs (i.e. four people). This problem becomes intractable, when applied to pedigrees, where a single family can be of any size. The BGmisc package addresses this gap by offering a comprehensive suite of functions for structuring and modeling such data.

## Statement of need

As behavior genetics delves into more complex data structures like pedigrees, the limitations of current tools become evident. The BGmisc R package addresses these challenges, going beyond what is available in tools like OpenMx and EasyMx that mainly focus on classical twin models.

Two widely used R packages in genetics modeling are OpenMx (Neale et al., 2016) and kinship2 (J. P. Sinnwell et al., 2014; J. Sinnwell & Therneau, 2022). The OpenMx (Neale et al., 2016) package is a workhorse in behavior genetic research. Not only is it a general-purpose software for structural equation modeling that is popular among behavior geneticists (Garrison, 2018), but also for its unique features – the `mxCheckIdentification()` function. This function checks whether a model is identified, determining if there is a unique solution to estimate the model's parameters based on the observed data. In addition, EasyMx (Hunter, 2023) is a more user-friendly package that streamlines the process of building and estimating structural equation models. It seamlessly integrates with OpenMx's infrastructure. Its functionalities range from foundational matrix builders like `emxCholeskyVariance` and `emxGeneticFactorVariance` to more specialized functions like `emxTwinModel` designed for classical twin models. Despite their strengths, EasyMx and OpenMx have limitations when handling extended family data. Notably, they lack functions for handling modern molecular designs (Kirkpatrick et al., 2021),

42 modeling genetic complex relationships, inferring relatedness, or simulating pedigrees.

43 Although not a staple in behavior genetics, the kinship2 (J. P. Sinnwell et al., 2014) package  
44 provides core features to the broader statistical genetics scientific community, such as plotting  
45 pedigrees and computing genetic relatedness matrices. It uses the Lange algorithm (Lange,  
46 2002) to compute relatedness coefficients. This recursive algorithm is discussed in great detail  
47 elsewhere, laying out several boundary conditions and recurrence rules. The BGmisc package  
48 extends the capabilities of kinship2 by introducing an alternative algorithm to calculate the  
49 relatedness coefficient based on network models. By applying classic path-tracing rules to the  
50 entire network, this new method is computationally more efficient by eliminating the need for  
51 a multi-step recursive approach.

## 52 Features

53 The BGmisc package offers various features tailored for extended behavior genetics analysis.  
54 These features are grouped under two main categories, mirroring the structure presented in  
55 our vignettes.

### 56 Modeling and Relatedness:

- 57 ■ Model Identification: BGmisc evaluates whether a variance components model is identified  
58 and fits the model's estimated variance components to observed covariance data. The  
59 technical aspects related to model identification have been described in Hunter et al.  
60 (2021).
- 61 ■ Relatedness Coefficient Calculation: Using path tracing rules first described in (Wright,  
62 1922) and formalized in (McArdle & McDonald, 1984), BGmisc calculates the (sparse)  
63 relatedness coefficients between all pairs of individuals in extended pedigrees based purely  
64 on mother and father identifiers.
- 65 ■ Relatedness Inference: BGmisc infers the relatedness between two groups based on their  
66 observed total correlation, given additive genetic and shared environmental parameters.

### 67 Pedigree Analysis and Simulation:

- 68 ■ Pedigree Conversion: BGmisc converts pedigrees into various relatedness matrices, in-  
69 cluding additive genetics, mitochondrial, common nuclear, and extended environmental  
70 relatedness matrices.
- 71 ■ Pedigree Simulation: BGmisc simulates pedigrees based on parameters including the  
72 number of children per mate, generations, sex ratio of newborns, and mating rate.

73 Collectively, these tools provide a valuable resource for behavior geneticists and others who  
74 work with extended family data. They were developed as part of a grant and have been used  
75 in several ongoing projects (Burt, 2023; Garrison et al., 2023; Hunter et al., 2023; Lyu et al.,  
76 2023), and theses (Lyu, 2023).

## 77 Availability

78 The BGmisc package is open-source and available on both GitHub at [https://github.com/](https://github.com/R-Computing-Lab/BGmisc)  
79 [R-Computing-Lab/BGmisc](https://github.com/R-Computing-Lab/BGmisc) and the Comprehensive R Archive Network (CRAN) at [https://](https://cran.r-project.org/package=BGmisc)  
80 [cran.r-project.org/package=BGmisc](https://cran.r-project.org/package=BGmisc). It is licensed under the GNU General Public License.

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