

How to fit an animal model

An ecologist guide

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Preface

This book is a collection of tutorial from the excellent paper by (?). Instead of just copy pasting the tutorial in a bookdown format, the tutorials have been updated to work with the newest version of the softwares and extended to present other softwares. **However, this is still a work in progress.**



Do not take anything in this manual as gospel.

Contributors

List of people who contributed to update and extend tutorials:

- Eric Postma updated the previous tutorial from asreml-r 3 to asreml-r 4
- Julien Martin developed and maintains the site
- Mathieu Videlier added (and continues to add) more details in the document

Chapitre 1

Introduction

The book is provides a series of tutorials (and accompanying data files) to fit animal model in R using different packages (**ASReml-R**, **gremlin**, **MCMCglmm** and **brms/stan**) . You will need to carefully follow the instructions below to first download the data files and second install the R packages. Before beginning the tutorial, we assume the reader has successfully installed the chosen R package on their computer and has saved the required data files to an appropriate directory from which they will be read. Full instructions for how to do this are provided with software distributions.

To work though the different tutorial I would recommend to create a folder where you will save your different R scripts for the tutorials.

In addition, the tutorial is here to help researchers in their coding and understanding of models and outputs, but it is required that you read and understand the literature in quantitative genetics and animal model.

1.1 Data

1.1.1 Data files

You will need to download 3 data files for the tutorial in R:

- gryphon.csv: data on gryphon birth weight and morphology
- gryphonRM.csv: data on gryphon repeated measurement of lay date.
- gryphonped.csv: data on the associated pedigree of the data gryphon

In addition, some models presented in the tutorials can take a while to run (sometimes > 1 hour), thus we are also providing the model outputs to allow you continue the tutorial without waiting for the model to run. (But you are free to run models)

The files are available [here](#) I recommend to save the data and Rdata files in a subfolder **data** in the folder you will use as your working directory for R and where you will save your R scripts. It should be noted that the tutorial are using this structure to read or save data.

1.1.2 Notes on data and pedigree

It is always important to take time to think carefully about the strengths and potential limitations of your pedigree information before embarking on quantitative genetic analyses. Pedigree Viewer, written by Brian Kinghorn, is an extremely useful application for visualizing pedigrees, and can be downloaded from: <http://www-personal.une.edu.au/~bkinghor/pedigree.htm>. Pedantics an R package written by Michael Morrissey and distributed through CRAN (<http://cran.r-project.org/>) can also be used for this and offers some nice additional features for visualizing pedigree structures and generating associated statistics. Before you begin running through the tutorials, we advise taking a moment to look at the pedigree files provided with them using Pedigree Viewer or Pedantics.

1.2 R

You should check that you have the most current version of R and R packages. You can check the number of the current version on CRAN. If you need to update (or install) R packages, use `install.packages()` and follow the prompted instructions.

1.2.1 R packages

1.2.1.1 asreml-r

ASReml-R is commercial software published by VSN international (<http://www.vsni.co.uk/software/asreml/>). This package is not free and requires a key access. Additional information and guide can be find in the Asreml-R manual: (<https://asreml.kb.vsni.co.uk/wp-content/uploads/sites/3/2018/02/ASReml-R-Reference-Manual-4.pdf>)

1.2.1.2 gremlin

`gremlin` is a little monster appearing if you feed a mugwai after midnight. It is also a great and promising software written by Pr. Matthew E. Wolak to fit mixed models using a frequentist approach .

1.2.1.3 MCMCglmm

`MCMCglmm` is an R package for Bayesian mixed model analysis written by Pr. Jarrod Hadfield. It is a freeware distributed through CRAN (<http://cran.r-project.org/>). Information and guide about the package can be find in the user manual and vignettes (<http://cran.r-project.org/web/packages/MCMCglmm/index.html>). Reference: (? , ?).

This module provides some information that applies to MCMCglmm-based analyses in general, but that will not be included in other tutorials. Most importantly, this applies to some of the simplest ways of determining the performance of a run using MCMCglmm, i.e., verification of the validity of the posterior distribution. This tutorial is not a substitute for working through the MCMCglmm course notes, which is available from CRAN (the Comprehensive R ArchiveNetwork, <http://cran.r-project.org/>, or can be accessed in R using the command `vignette("CourseNotes", "MCMCglmm")`). These tutorials do not introduce one of the main advantages of using MCMCglmm for analyses of data from natural populations -the ability to

properly model non-normal responses. These capabilities are introduced in the documentation that is distributed with MCMCglmm, and available from CRAN. Another specific animal guide for MCMCglmm can be find (https://devillemerueil.legtux.org/wp-content/uploads/2021/09/tuto_en.pdf). Pr. Pierre de Villemereuil provide more information in Bayesian concept and focus more on non-gaussian variable.

1.2.1.4 brms

`brms` provides an interface to fit Bayesian generalized multivariate (non-)linear multilevel models using `Stan`, which is a C++ package for obtaining full Bayesian inference (see <https://mc-stan.org/>). The formula syntax is an extended version of the syntax applied in the ‘lme4’ package to provide a familiar and simple interface for performing regression analyses.

It should be noted that if `brms` is able to fit animal model the parametrization used is not the most efficient and can take quite longer than using a different parametrization directly in `stan`.

Chapitre 2

Univariate animal model

This tutorial will demonstrate how to run a univariate animal model to estimate genetic variance in birth weight in the mighty gryphons.

2.1 Scenario and data

2.1.1 Scenario

In a population of gryphons there is strong positive selection on birth weight with heavier born individuals having, on average higher fitness. To find out whether increased birth weight will evolve in response to the selection, and if so how quickly, we want to estimate the heritability of birth weight.

2.1.2 Data files

Open `gryphonped.csv` and `gryphon.csv` in your text editor. The structure and contents of these files is fairly self-explanatory. The pedigree file `gryphonped.csv` contains three columns containing unique IDs that correspond to each animal, its father, and its mother. Note that this is a multigenerational pedigree, with the earliest generation (for which parentage information is necessarily missing) at the beginning of the file. For later-born individuals maternal identities are all known but paternity information is incomplete (a common situation in real world applications).

The phenotype data, as well as additional factors and covariates that we may wish to include in our model are contained in `gryphon.csv`. Columns correspond to individual identity (`animal`), maternal identity (`mother`), year of birth (`byear`), sex (`sex`, where 1 is female and 2 is male), birth weight (`bwt`), and tarsus length (`tarsus`). Each row of the data file contains a record for a different offspring individual. Note that all individuals included in the data file must be included as offspring in the pedigree file.

We can read the data file, using `read.csv()` which consider by default that `NA` is the symbol for missing values and that the first line of the file contains the column headers.

It is a good idea to make sure that all variables are correctly assigned as numeric or factors:

```

gryphon$animal <- as.factor(gryphon$animal)
gryphon$mother <- as.factor(gryphon$mother)
gryphon$byear <- as.factor(gryphon$byear)
gryphon$sex <- as.factor(gryphon$sex)
gryphon$bwt <- as.numeric(gryphon$bwt)
gryphon$tarsus <- as.numeric(gryphon$tarsus)
str(gryphon)

## 'data.frame': 1084 obs. of 6 variables:
## $ animal: Factor w/ 1084 levels "1","2","3","5",...: 864 1076 549 989 1030 751 987 490 ...
## $ mother: Factor w/ 429 levels "1","2","3","8",...: 362 268 216 375 396 289 328 255 347 ...
## $ byear : Factor w/ 34 levels "968","970","971",...: 1 1 2 2 2 2 3 3 3 ...
## $ sex   : Factor w/ 2 levels "1","2": 1 1 2 1 2 1 2 1 1 ...
## $ bwt   : num 10.77 9.3 3.98 5.39 12.12 ...
## $ tarsus: num 24.8 22.5 12.9 20.5 NA ...

```

Similarly we can read in the pedigree file, using `read.csv()` which consider by default that `NA` is the symbol for missing values and that the first line of the file contains the column headers.

```

## 'data.frame': 1309 obs. of 3 variables:
## $ id    : int 1306 1304 1298 1293 1290 1288 1284 1283 1282 1278 ...
## $ father: int NA NA NA NA NA NA NA NA NA ...
## $ mother: int NA NA NA NA NA NA NA NA NA ...

gryphonped$id <- as.factor(gryphonped$id)
gryphonped$father <- as.factor(gryphonped$father)
gryphonped$mother <- as.factor(gryphonped$mother)
str(gryphonped)

## 'data.frame': 1309 obs. of 3 variables:
## $ id    : Factor w/ 1309 levels "1","2","3","4",...: 1306 1304 1298 1293 1290 1288 1284 ...
## $ father: Factor w/ 158 levels "4","13","18",...: NA NA NA NA NA NA NA NA NA ...
## $ mother: Factor w/ 429 levels "1","2","3","8",...: NA NA NA NA NA NA NA NA NA ...

```

Now that we have imported the data and the pedigree file, we are ready to fit an animal model.

2.2 Asreml-R

2.2.1 Running the model

First we need to load the `asreml` library:

```
library(asreml)
```

To be able to fit an animal model, Asreml-r needs (the inverse of) the relationship matrix using the `ainverse` function:

```
ainv <- ainverse(gryphonped)
```

We are now ready to specify our first model:

```
model1 <- asreml(
  fixed = bwt ~ 1, random = ~ vm(animal, ainv),
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:53:29 2022
##          LogLik      Sigma2      DF    wall      cpu
## 1     -4128.454      1.0  853 16:53:29      0.0
## 2     -3284.272      1.0  853 16:53:29      0.0
## 3     -2354.992      1.0  853 16:53:29      0.0
## 4     -1710.357      1.0  853 16:53:29      0.0
## 5     -1363.555      1.0  853 16:53:29      0.0
## 6     -1263.516      1.0  853 16:53:29      0.0
## 7     -1247.854      1.0  853 16:53:29      0.0
## 8     -1247.185      1.0  853 16:53:29      0.0
## 9     -1247.183      1.0  853 16:53:29      0.0
```

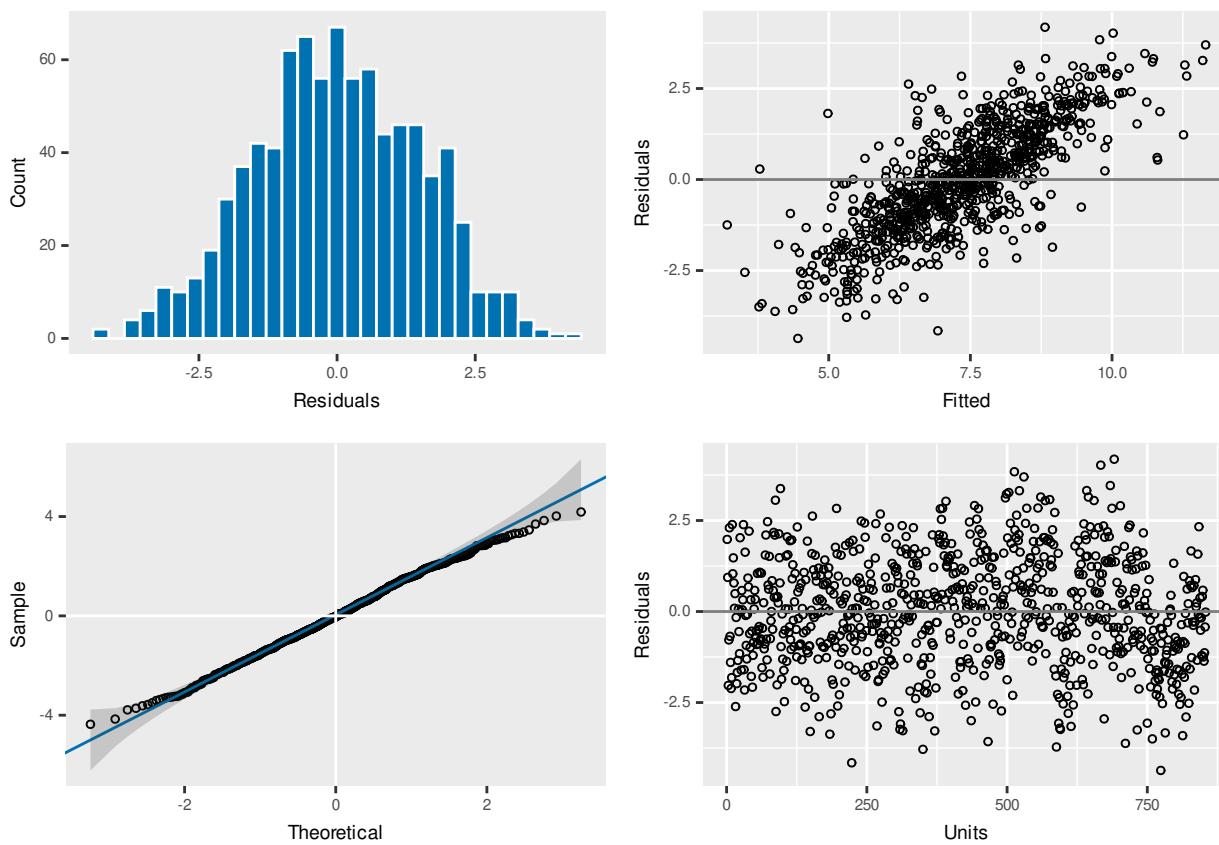
In this model, `bwt` is the response variable and the only fixed effect is the intercept, denoted as 1. The only random effect we have fitted is `animal`, which will provide an estimate of V_A . Our random `animal` effect is connected to the inverse related matrix `ainv` which integrate the relativeness or pedigree information.

`data=` specifies the name of the dataframe that contains our variables. Finally, we inform `asreml()` what to do when it encounters NAs in either the dependent or predictor variables (in this case we choose to remove the records). If you use the argument “include” instead of “omit”, model will keep the NA. With `x=“include”`, the model will exchange NA with 0. Be careful you need to standardize your trait so the mean will be equal to 0, if not estimates (including covariance in multivariate models) could be strongly biased due to the missing values considered as 0. `y=“include”` will exchange NA with a factor labeled `mv` which will be included in the sparse equation. For more details see Asreml-R manual.

A note of the specification of the structure of the residuals: This simple univariate model will run fine without `residual=~idv(units)`. However, if you are going to use `vpredict()` to calculate the heritability (see below), without specifying the residuals in this way will result in a standard error for the heritability that is incorrect.

Any model has assumption which need to be checked. The model can be plotted which help visualizing the distribution of the model residual and check the different assumptions.

```
plot(model1)
```



To see the estimates for the variance components, we run:

```
summary(model1)$varcomp
```

	component	std.error	z.ratio	bound	%ch	
## vm(animal, ainv)	3.395398	0.6349915	5.347154	P	0	
## units!units	3.828602	0.5185919	7.382687	P	0	
## units!R	1.000000		NA	NA	F	0

We fitted a single random effect so we partitioned the phenotypic variance into two components. The `vm(animal, ainv)` variance component is V_A and is estimated as 3.4. Given that the ratio of V_A to its standard error (`z.ratio`) is considerably larger than 2 (*i.e.* the parameter estimate is more than 2 SEs from zero), this looks likely to be significant. The `units!units` component refers to the residual variance V_R , and `units$R` should be ignored. If you don't include `residual=~idv(units)` in your model specification, `units$R` will provide you with the residual variance.

2.2.2 Estimating heritability

We can calculate the h^2 of birth weight from the components above since $h^2 = V_A/V_P = V_A/(V_A + V_R)$. Thus according to this model, $h^2 = 3.4 / (3.4 + 3.83) = 0.47$.

Alternatively we can use the `vpredict()` function to calculate h^2 and its standard error. `v predict()` function has two structures, first the model used (here `model1`) and then the estimate name with its associated equation. The equation used different V and their associated numbers depend of the order of the different random and residual effects included in the model.

```
v predict(model1, h2.bwt ~ V1 / (V1 + V2))
```

```
##           Estimate        SE
## h2.bwt  0.4700163 0.07650881
```

2.2.3 Adding fixed effects

To add fixed effects to a univariate model, we simply modify the model statement. For example, we might know (or suspect) that birth weight is a sexually dimorphic trait and therefore fit in the model.

```
model2 <- asreml(
  fixed = bwt ~ 1 + sex,
  random = ~ vm(animal, ainv),
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit"))
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:53:30 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1     -3364.126    1.0  852 16:53:30    0.0
## 2     -2702.117    1.0  852 16:53:30    0.0
## 3     -1978.916    1.0  852 16:53:30    0.0
## 4     -1487.834    1.0  852 16:53:30    0.0
## 5     -1236.350    1.0  852 16:53:30    0.0
## 6     -1172.771    1.0  852 16:53:30    0.0
## 7     -1165.270    1.0  852 16:53:30    0.0
## 8     -1165.093    1.0  852 16:53:30    0.0
## 9     -1165.093    1.0  852 16:53:30    0.0
```

Now we can look at the fixed effects parameters and assess their significance with a conditional Wald F-test:

```
summary(model2, coef = TRUE)$coef.fixed
wald.asreml(model2, ssType = "conditional", denDF = "numeric")
```

```
##           solution std.error z.ratio
## sex_1       0.000000      NA       NA
## sex_2      2.206996 0.1619974 13.62365
```

```

## (Intercept) 6.058669 0.1718244 35.26082
## Model fitted using the sigma parameterization.
## Warning in asreml(fixed = bwt ~ 1 + sex, random = ~vm(animal, ainv), residual =
## ~idv(units), : Algebraic derivatives for denominator df not available.
## ASReml 4.1.0 Tue Nov 29 16:53:30 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1    -1165.093       1.0    852 16:53:30      0.0
## 2    -1165.093       1.0    852 16:53:31      0.0
## Calculating denominator DF
## $Wald
##
##          Df denDF   F.inc   F.con Margin      Pr
## (Intercept) 1    251 3491.0 3491.0      0.00000e+00
## sex         1     831 185.6 185.6      A 2.70204e-38
##
## $stratumVariances
##          df Variance vm(animal, ainv) units!units
## vm(animal, ainv) 752.28476 5.957254           0.9864077 1
## units!units      99.71524 2.938413           0.0000000 1

```

The very small probability (Pr) in the Wald test above shows that `sex` is a highly significant fixed effect, and from the parameter estimates (`summary(model2,coef=T)$coef.fixed`) we can see that the average male (sex 2) is 2.2 kg (± 0.16 SE) heavier than the average female (sex 1). However, when we look at the variance components in the model including `sex` as a fixed effect, we see that they have changed slightly from the previous model:

```
summary(model2)$varcomp
```

```

##          component std.error z.ratio bound %ch
## vm(animal, ainv) 3.060441 0.5243571 5.836558      P  0
## units!units      2.938412 0.4161473 7.060991      P  0
## units!R          1.000000        NA        NA      F  0

```

In fact since `sex` effects were previously contributing to the residual variance of the model, our estimate of V_R (denoted `units!R` in the output) is now slightly lower than before. This has an important consequence for estimating heritability since if we calculate V_P as $V_A + V_R$ then as we include fixed effects we will soak up more residual variance driving V_P . Assuming that V_A is more or less unaffected by the fixed effects fitted then as V_P goes down we expect our estimate of h^2 will go up:

```
(h2.1 <- vpredict(model1, h2.bwt ~ V1 / (V1 + V2)))
```

```

##          Estimate      SE
## h2.bwt 0.4700163 0.07650881

```

```
(h2.2 <- vpredict(model2, h2.bwt ~ V1 / (V1 + V2)))
```

```
##          Estimate       SE
## h2.bwt  0.510171 0.07432388
```

Here h^2 has increased slightly from 0.47 to 0.51. Which is the better estimate? It depends on what your question is. The first is an estimate of the proportion of variance in birth weight explained by additive effects, the latter is an estimate of the proportion of variance in birth weight *after conditioning on sex* that is explained by additive effects.

An important piece of advice, each researcher should be consistent in how they name their estimates and always correctly describe which estimates they are using conditional or not (to avoid any confusion).

2.2.4 Adding random effects

This is done by simply modifying the model statement in the same way. For instance fitting:

```
model3 <- asreml(
  fixed = bwt ~ 1 + sex,
  random = ~ vm(animal, ainv) + byear,
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:53:31 2022
##           LogLik      Sigma2      DF    wall      cpu
## 1     -2742.658      1.0    852 16:53:31      0.0
## 2     -2237.268      1.0    852 16:53:31      0.0
## 3     -1690.453      1.0    852 16:53:31      0.0
## 4     -1328.910      1.0    852 16:53:31      0.0
## 5     -1154.597      1.0    852 16:53:31      0.0
## 6     -1116.992      1.0    852 16:53:31      0.0
## 7     -1113.809      1.0    852 16:53:31      0.0
## 8     -1113.772      1.0    852 16:53:31      0.0
## 9     -1113.772      1.0    852 16:53:31      0.0
```

```
summary(model3)$varcomp
```

	component	std.error	z.ratio	bound	%ch
## byear	0.8862604	0.2695918	3.287416	P	0
## vm(animal, ainv)	2.7068665	0.4422140	6.121169	P	0
## units!units	2.3092415	0.3451025	6.691466	P	0
## units!R	1.0000000	NA	NA	F	0

```
(h2.3 <- vpredict(model3, h2.bwt ~ V2 / (V1 + V2 + V3)))
```

```
##          Estimate       SE
## h2.bwt  0.4586068 0.06740364
```

Here the variance in `bwt` explained by `byear` is 0.89 and, based on the `z.ratio`, appears to be significant (>2). Thus we would conclude that year-to-year variation (e.g., in weather, resource abundance) contributes to V_P . Note that although V_A has changed somewhat, as most of what is now partitioned as a birth year effect was previously partitioned as V_R . Thus what we have really done here is to partition environmental effects into those arising from year-to-year differences versus everything else, and we do not really expect much change in h^2 (since now $h^2 = V_A/(V_A + V_{BY} + V_R)$).

However, we get a somewhat different result if we also add a random effect of `mother` to test for maternal effects:

```
model4 <- asreml(
  fixed = bwt ~ 1 + sex,
  random = ~ vm(animal, ainv) + byear + mother,
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:53:31 2022
##          LogLik      Sigma2      DF    wall      cpu
## 1     -2033.178      1.0    852 16:53:31      0.0
## 2     -1723.734      1.0    852 16:53:31      0.0
## 3     -1396.354      1.0    852 16:53:31      0.0
## 4     -1193.012      1.0    852 16:53:31      0.0
## 5     -1107.946      1.0    852 16:53:31      0.0
## 6     -1095.327      1.0    852 16:53:31      0.0
## 7     -1094.816      1.0    852 16:53:31      0.0
## 8     -1094.815      1.0    852 16:53:31      0.0
```

```
summary(model4)$varcomp
```

```
##          component std.error   z.ratio bound %ch
## byear        0.8820313 0.2632455 3.350604      P  0
## mother       1.1184698 0.2386239 4.687167      P  0
## vm(animal, ainv) 2.2985320 0.4962496 4.631806      P  0
## units!units   1.6290034 0.3714154 4.385934      P  0
## units!R       1.0000000        NA         NA      F  0
```

```
(h2.4 <- vpredict(model4, h2.bwt ~ V1 / (V1 + V2 + V3 + V4)))
```

```
##           Estimate       SE
## h2.bwt 0.1487898 0.03861552
```

Here partitioning of significant maternal variance has resulted in a further decrease in V_R but also a decrease in V_A . The latter is because maternal effects of the sort we simulated (fixed differences between mothers) will have the consequence of increasing similarity among maternal siblings. Consequently they can look very much like additive genetic effects and if present, but unmodelled, represent a type of “common environment effect” that can - and will - cause upward bias in V_A and so h^2 . The “common environment” can be conceived as the inextricable sum of the maternal additive genetic effect (such as maternal loci) and the maternal environment or permanent environment (such as litter or nest environment created or modified by the mother).

2.2.5 Testing significance of random effects

An important point to note in this tutorial is that while the `z.ratio` (`component/std.error`) reported is a good indicator of likely statistical significance ($>1.96?$), the standard errors are approximate and are not recommended for formal hypothesis testing. A better approach is to use likelihood-ratio tests (LRT).

For example, to test the significance of maternal effects we could compare models with and without the inclusion of maternal identity as a random effect and compare the final log-likelihoods of these models.

```
model4$loglik
```

```
## [1] -1094.815
```

shows that the model including maternal identity has a log-likelihood of -1094.815, and

```
model3$loglik
```

```
## [1] -1113.772
```

shows that the model excluding maternal identity has a log-likelihood of -1113.772.

A test statistic equal to twice the absolute difference in these log-likelihoods is assumed to be distributed as Chi square with one degree of freedom (one term of difference between the two models). In this case we would conclude that the maternal effects are highly significant since: $2 \times (-1094.8145793 - -1113.7719147)$ equals 37.9146708, and the p-value that comes with this is:

```
1 - pchisq(2 * (model4$loglik - model3$loglik), 1)
```

```
## [1] 7.390738e-10
```

As $P < 0.0001$ we would therefore conclude that the additional of maternal identity as a random effect significantly improves the fit of the model, given an increase in log-likelihood of approximately 19.

2.2.6 Further partitioning the variance

A population can be further fragmented into different groups or categories (such as females and males, juveniles and adults or treated and untreated). Some scientific questions require further and deeper analysis of the variance. To avoid multiple model (one for each group), we can directly partition the variance between groups in a unique model. In addition, by doing so, we can also test if the variance are different between groups.

As example, we decide to take the model4 and partition its additive genetic variance and residual variance by sex. It is possible to further partition the other random effects but it will complexity the animal model and requires sufficient sample size.

First, it required to order the dataset by group (here sex).

```
gryphon <- gryphon[order(gryphon$sex), ]
```

To partition variances between sex, two distinct functions are require `at()` for the random level, and `dsum()` for the residual level:

```
model_SEX <- asreml(
  fixed = bwt ~ 1 + sex,
  random = ~ at(sex):vm(animal, ainv) + byear + mother,
  residual = ~ dsum(~ units | sex),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:53:31 2022
##          LogLik      Sigma2      DF    wall      cpu
## 1     -1142.164      1.0    852 16:53:31      0.0
## 2     -1126.308      1.0    852 16:53:31      0.0
## 3     -1111.536      1.0    852 16:53:31      0.0
## 4     -1105.383      1.0    852 16:53:31      0.0
## 5     -1104.375      1.0    852 16:53:31      0.0
## 6     -1104.364      1.0    852 16:53:31      0.0
```

```
summary(model_SEX)$varcomp
```

	component	std.error	z.ratio	bound	%ch
## byear	0.9001595	0.2690012	3.346303	P	0.0
## mother	1.3396184	0.2663118	5.030263	P	0.0

```
## at(sex, 1):vm(animal, ainv) 1.4372390 0.6514306 2.206281      P 0.1
## at(sex, 2):vm(animal, ainv) 1.9861434 0.9974302 1.991261      P 0.3
## sex_1!R                      2.1706213 0.5542492 3.916327      P 0.0
## sex_2!R                      1.7112948 0.8246188 2.075256      P 0.3
```

By partitioning the additive genetic variance and the residual variance, the model estimates the V_A and V_R for each group (sex). Doing so, we can calculate the h^2 for each group of sex. Here, it's important to know in which order the variances are estimated to extract the correct variance in the heritability equation.

```
(h2.F <- vpredict(model_SEX, h2.bwt ~ V3 / (V1 + V2 + V3 + V5)))
```

```
##           Estimate       SE
## h2.bwt 0.2457811 0.1070794
```

```
(h2.M <- vpredict(model_SEX, h2.bwt ~ V4 / (V1 + V2 + V4 + V6)))
```

```
##           Estimate       SE
## h2.bwt 0.3345244 0.1619218
```

To test if the variances are different between sexes, we can compare the model partitioned `model_SEX` and the previous model without the partitioning `model4` in a likelihood ratio test (LRT) with 2 degrees of freedom since models have two components of variance of difference.

```
model_SEX$loglik
```

```
## [1] -1104.364
```

```
model4$loglik
```

```
## [1] -1094.815
```

```
1 - pchisq(2 * (model_SEX$loglik - model4$loglik), 2)
```

```
## [1] 1
```

Here, we can see the point estimates of h^2 seems to differ between sexes (0.25 and 0.33), but their SE overlaps. LRT give more information and showed that partitioning the variance and the residual between sexes did not improved the fit of the model and so their variance are not significantly different.

```
h2.sex <- rbind(h2.F, h2.M)
```

```
plot(c(0.95, 1.05) ~ h2.sex[, 1], xlim = c(0, 0.8), ylim = c(0.5, 1.5), , xlab = "", ylab
arrows(y0 = 0.95, x0 = h2.sex[1, 1] - h2.sex[1, 2], y1 = 0.95, x1 = h2.sex[1, 1] + h2.sex[1, 2])
```

```
arrows(y0 = 1.05, x0 = h2.sex[2, 1] - h2.sex[2, 2], y1 = 1.05, x1 = h2.sex[2, 1] + h2.sex[2, 2], line = 3)
mtext("Narrow-sense heritability ( $\pm$ se)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6)
axis(2, at = 1, labels = c("birth weight"), las = 3, cex.axis = 1.6)
```

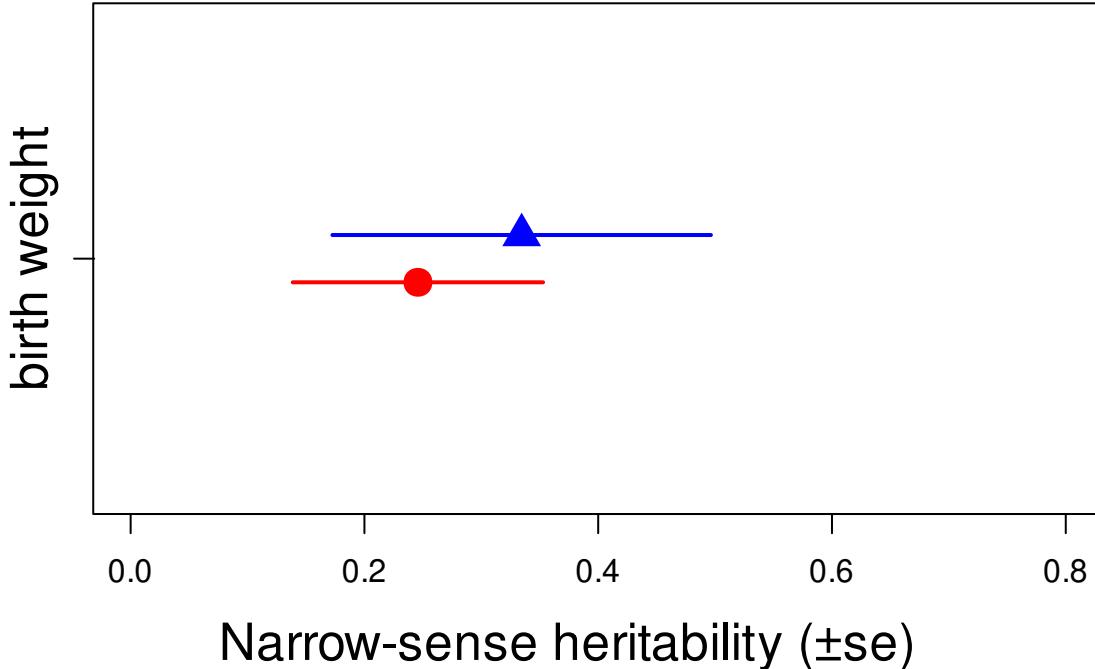


Figure 2.1: Female and male heritability of birth weight

2.2.7 Modification of the variance matrix parameters

Variance represents the deviation of the distribution and it is expected to be a positive value. Due to a lack of power, a structural problem in the dataset or a very low variance, Asreml-r often fixes the variance to a boundary B instead of a positive value P. When it happens, it is generally a good idea to examine it.

To examine the boundary effect, we can explore an alternative model where the model allowed an unstructured parameter for the variance of interest or the entire variance matrix. For this example: we allowed the model to estimate any values (so allowing possible negative values of estimates) for the random and residual matrix.

First, we create a temporary model `model.temp` with the exact structure to modify.

```
model.temp <- asreml(
  fixed = bwt ~ 1,
```

```

random = ~ vm(animal, ainv) + byear + mother,
residual = ~ idv(units),
data = gryphon,
na.action = na.method(x = "omit", y = "omit"),
start.values = T
)
G.temp <- model.temp$vparameters[(1:3), ]
G.temp$Constraint <- "U"
R.temp <- model.temp$vparameters[-(1:3), ]
R.temp$Constraint[2] <- "U"

```

The argument `start.values=T` allowed the `model.temp` to change its random parameters. We can create the two different matrices and specify which parameters will be modified. For this example we modified the G and the R matrix to fit all variance to be U unstructured. it is important to note for the R matrix the line `units!R` has to be fix to 1, so it will never change.

The object `G.temp` and `R.temp` can be implemented in the following model as new parameters using the argument `R.param` and `G.param`.

```

model5 <- asreml(
  fixed = bwt ~ 1 + sex,
  random = ~ vm(animal, ainv) + byear + mother,
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit"),
  R.param = R.temp, G.param = G.temp
)

```

```

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:53:32 2022
##          LogLik      Sigma2      DF    wall      cpu
## 1     -2033.178      1.0    852 16:53:32      0.0
## 2     -1723.734      1.0    852 16:53:32      0.0
## 3     -1396.354      1.0    852 16:53:32      0.0
## 4     -1193.012      1.0    852 16:53:32      0.0
## 5     -1107.946      1.0    852 16:53:32      0.0
## 6     -1095.327      1.0    852 16:53:32      0.0
## 7     -1094.816      1.0    852 16:53:32      0.0
## 8     -1094.815      1.0    852 16:53:32      0.0

```

```
summary(model5)$varcomp
```

	component	std.error	z.ratio	bound	%ch
## byear	0.8820313	0.2632455	3.350604	U	0
## mother	1.1184698	0.2386239	4.687167	U	0

```
## vm(animal, ainv) 2.2985320 0.4962496 4.631806      U    0
## units!units      1.6290034 0.3714154 4.385934      U    0
## units!R          1.0000000      NA      NA      F    0
```

Since model4 did not showed boundary, the model5 is very similar.

2.2.8 Covariance between two random effects

Some research questions require to estimate the covariance between two random effects within a univariate model. To do so, we can use the argument `str`. As an example, we fit a model which estimate the covariance between the additive genetic variance and the mother variance. Both variances require to operate on the same level, thus `animal` and `mother` require to be associated to the pedigree information.

The argument `str` has two components: first the equation term with the two random effects `~vm(animal, ainv)+vm(mother, ainv)` and second the structural term `~us(2):id(number)`. Here within the structural term, we fit a 2x2 unstructured matrix `us(2)` which estimated the variance and the covariance between the random effects in the equation term. To successfully work, the structural term also requires the number of level identified within `id()`. Here a small tip, if you don't know the number of level identified within `id()`, run the model with a random number. The model will not converge and a error message will appear like this one: `Size of direct product (4) does not conform with total size of included terms (2618)`. The error message can help you determine the required level within the `str` function, as here 2618 divide by 2. In addition, it is necessary the random effects

```
model.temp2 <- asreml(
  fixed = bwt ~ 1,
  random = ~ str(~ vm(animal, ainv) + vm(mother, ainv), ~ us(2):id(1309)) + byear,
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit"),
  start.values = T
)

G.temp2 <- model.temp2$vparameters[(1:4), ]
G.temp2$Constraint <- "U"
model6 <- asreml(
  fixed = bwt ~ 1 + sex,
  random = ~ str(~ vm(animal, ainv) + vm(mother, ainv), ~ us(2):id(1309)) + byear,
  residual = ~ idv(units),
  data = gryphon,
  na.action = na.method(x = "omit", y = "omit"),
  # equate.levels = c("animal", "mother"),
  , G.param = G.temp2
)
summary(model6)$varcomp
```

We have successfully produced a code to estimate the covariance between two random effects. However for this example, the dataset is not sufficient to properly estimate it and the model did not converge but you have the idea of how to use the function `str`.

Additional and final tip: It is happen that Asreml will estimate negative variance if you allow the variance matrix to be unstructured. A negative variance is counter-intuitive meaning statistically the mean within the random effect is less similar than expected by chance. However a possible biological reason can be hypothesized such as a sibling competition within the nest creating a negative among-individual covariance within the nest. Thus to test this hypotheses, it is required to estimate the covariance between two random effects.

2.3 gremlin

TODO (maybe just bother Matthew to do it)

Meanwhile



Figure 2.2: Keep it dry and do no feed after midnight.

2.4 MCMCglmm

2.4.1 Running the model

First load MCMCglmm:

```
library(MCMCglmm)
```

The first model we will fit is a simple animal model with no fixed effects, and only an ‘animal’ random effect relating individuals to their additive genetic values through the pedigree.

First we are going to define the priors. In a way we might want to avoid using priors, because we would like all of the information in our analysis to come from our data. By default MCMCglmm uses improper priors, but this can cause inferential and numerical problems. We will specify priors for the animal effect and the residual variance using the following code:

```
prior1.1 <- list(
  G = list(G1 = list(V = 1, nu = 0.002)),
  R = list(V = 1, nu = 0.002)
)
```

A prior allowed the model to fit different variance structures. With the unique random effect “animal”, we partitioned the phenotypic variance into two distinct variances matrices G (additive genetic) and R (residual). This prior specification is the simplistic one and often used because it was believed to be relatively uninformative, and is equivalent to an inverse-gamma prior with shape and scale equal to 0.001. In many cases it is relatively uninformative but when the posterior distribution for the variances has support close to zero it can behave poorly. Parameter expanded priors (See Chapter 8 of the MCMCglmm CourseNotes, available from CRAN) are gaining in popularity due to their better behaviour but for the purposes of this tutorial we will stick with the inverse-gamma prior.

We have told MCMCglmm to pay little heed to our prior expectation (V) by specifying a small degree of belief parameter (nu) of 0.002. Since this is a univariate analysis, the priors are matrix of order 1 and thus nu>0 is the smallest degree of belief that provides what is known as a ‘proper’ prior, avoiding numerical problems. In fact, there is a lot of information in the data regarding the marginal distributions of the parameters, and MCMCglmm will run most of the models that we suggest in these tutorials without priors. However, this is poor practice, but we will therefore use this simple priors throughout these tutorials. We can now fit an animal model. The model to decompose variation in birth weight into genetic and residual effects is as follows:

The lower case “animal” is a can be a **special** word for MCMCglmm. If a **pedigree** argument is provided then **MCMCglmm** will recognize the term **animal** as the term to use to estimate additive genetic variance. When the argument **pedigree** is not provided then the word **animal** is not different than any other variable. However, instead of providing a pedigree argument to the call to MCMCglmm function, it is much more flexible to use the **ginv** argument to specify the random effect that must be linked to the pedigree (with the inverse relatedness matrix). We thus first estimate the inverse relatedness matrix using **inverseA()** then fit the animal model.

```
Ainv <- inverseA(gryphonped)$Ainv
model1.1 <- MCMCglmm(bwt ~ 1,
  random = ~animal, ginv = list(animal = Ainv),
  data = gryphon, prior = prior1.1
```

```

)
##                                     MCMC iteration = 0
##
##                                     MCMC iteration = 1000
##
##                                     MCMC iteration = 2000
##
##                                     MCMC iteration = 3000
##
##                                     MCMC iteration = 4000
##
##                                     MCMC iteration = 5000
##
##                                     MCMC iteration = 6000
##
##                                     MCMC iteration = 7000
##
##                                     MCMC iteration = 8000
##
##                                     MCMC iteration = 9000
##
##                                     MCMC iteration = 10000
##
##                                     MCMC iteration = 11000
##
##                                     MCMC iteration = 12000
##
##                                     MCMC iteration = 13000

```

After typing this code, MCMCglmm will run, taking about 20 seconds on a modern desktop computer. The progress of the run will be printed to the screen. Also, note the warning message will be printed at the end of the run. This is natural too. In order for the MCMC algorithm to work, MCMCglmm must keep track of effects associated with unmeasured individuals appearing in the pedigree. This will not affect the answers, but when many unmeasured individuals exist, it can hinder the ability of the algorithm to explore the parameter space (more on this, and a solution, later). Lets have a look at the MCMCglmm outputs. First we will evaluate how confident we can be that MCMCglmm found good answers. By entering

```
plot(model1.1$Sol)
```

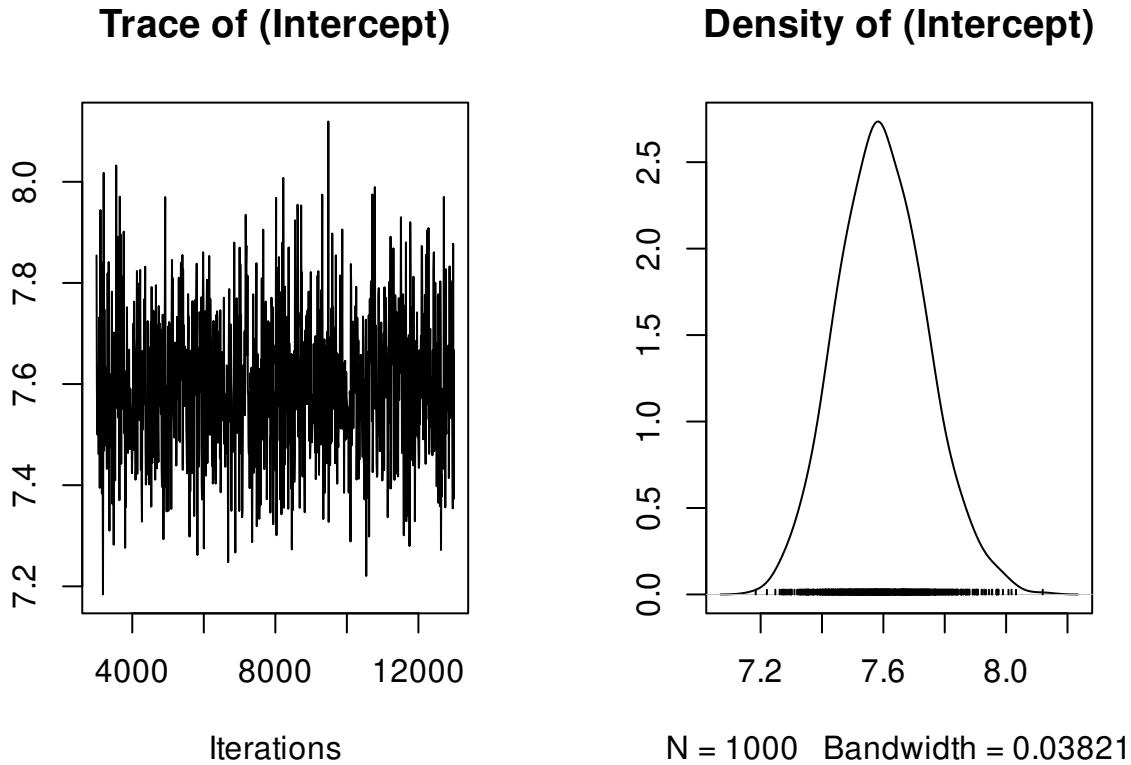


Figure 2.3: The posterior distribution of the fixed effect (the intercept, or mean) in model 1.1

in the console, we get Figure 2.2. The plot on the left shows a time series of the values of 1000 samples of the posterior distribution of the the model intercept (mean birth weight). The plot on the right shows the same data as a distribution. Complicated statistical methods for estimating population means are of course of little interest; rather, we are examining these outputs to check that MCMCglmm's algorithms worked well for our data and for this model. The important point here is that a consistent amount of variation around a largely unchanging mean value of the intercept was obtained (which give this fluctuating trace concentrated around the mean), and the posterior distribution of the intercept appears to be valid. More rigorous means of evaluation the independence of the samples in the posterior distribution (evaluating autocorrelation) are discussed in the MCMCglmm CourseNotes, available from CRAN. Note that your output for `model1.1` may not be identical to this due to Monte Carlo (random number) error. So every times, you run the model, you will get similar but slightly different results.

The posterior distributions of the the variance components are generally of more interest to animal model users. We can view plots of the posterior distribution for the variance components for model 1.1 by

```
plot(model1.1$VCV)
```

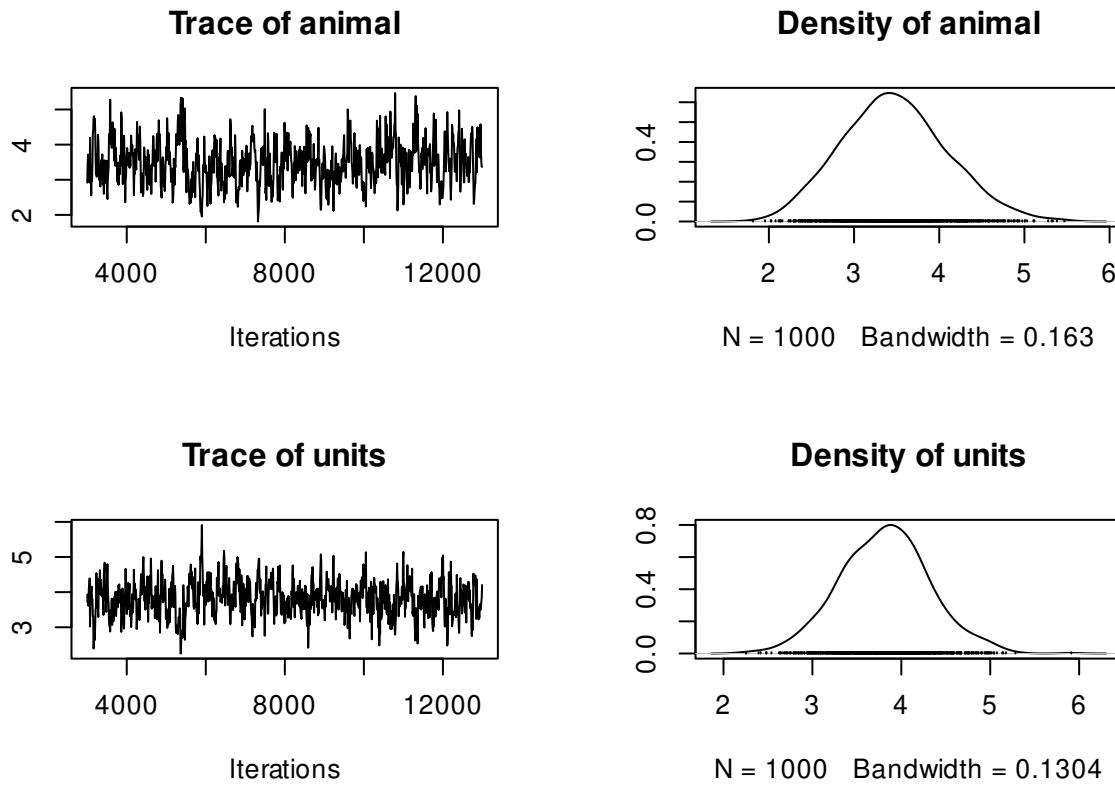


Figure 2.4: The posterior distributions of the variance components of model 1.1, based on an analysis with the default values for nitt, burnin, and thin in MCMCglmm

which generates Figure 2.3. Here we see distributions of the estimates of the additive genetic (animal) and residual (units) effects. These samples contain some autocorrelation, i.e., trends are apparent in the left-hand plot. We can deal with this easily.

2.4.2 Change in iteration and sampling

We will simply re-run the model for a longer number of iterations, and sample the chain less frequently. So far we have been running MCMCglmm with its default values. These defaults are a total run length of 13000 iterations, the first 3000 of which are discarded as a ‘burn-in’ period to make sure that the converges to the part of the parameter space where the maximum likelihood exists. The remaining 10000 iterations are sampled (estimates retained) every 10 iterations (the thinning interval). Because the values in the left-hand plots in figure 2.2 to appear to have different values at the beginning of the run, we might suspect that a longer burn-in period might be required. We can reduce the autocorrelation by lengthening the rest of the run and sampling the chain less frequently. The following code runs the same model 1.1, but is likely to produce better samples of the posterior distributions. This model should take about two minutes to analyze.

```
model1.1 <- MCMCglmm(bwt ~ 1,
  random = ~animal, ginv = list(animal = Ainv),
```

```

  data = gryphon, nitt = 65000, thin = 50, burnin = 15000,
  prior = prior1.1, verbose = FALSE
)

```

Notes that we have now included the argument `verbose=FALSE` in the `MCMCglmm` call. We will continue this throughout the tutorial so that more complete screen outputs can be included in this document without using too much space. Note that the autocorrelation is much reduced. A more compact way to evaluate the validity of the posterior distributions is to calculate autocorrelation among samples, as follows:

```
autocorr.diag(model1.1$VCV)
```

```

##           animal      units
## Lag 0     1.000000000 1.000000000
## Lag 50    0.209039004 0.173955831
## Lag 250   -0.017811690 -0.028870690
## Lag 500   -0.007328492 0.008719608
## Lag 2500   0.050325531 0.056367451

```

We will consider these levels of autocorrelation acceptable, at least for the purposes of this tutorial. Ideally, all samples of the posterior distribution should be independent, and the autocorrelation for all lag values greater than zero should be near zero. However, in practice this will not strictly be achievable for all analytic scenarios. Certainly the levels of autocorrelation observed here should not be tolerated in any formal analysis. Note that the validity of posterior distributions of any analysis should always be checked; however, for brevity we will not continue to be so consistently diligent throughout the rest of these tutorials. We can now proceed with confidence to recover some more information from these samples. We can obtain estimates of the additive genetic and residual variance by calculating the modes of the posterior distributions:

```
posterior.mode(model1.1$VCV)
```

```

##   animal      units
## 3.310074 3.728226

```

We can obtain the Bayesian equivalent of confidence intervals by calculating the the values of the estimates that bound 95% (or any other proportion) of the posterior distributions:

```
HPDinterval(model1.1$VCV)
```

```

##           lower      upper
## animal  2.185154 4.567025
## units   2.899000 4.922814
## attr(,"Probability")
## [1] 0.95

```

2.4.3 Change priors parameters

We specified weak priors in this analyses. Now we will check whether or not proper priors would have influenced the results that we obtained. The simplest way to do this is to re-run the model with different priors. In the previous model we specified a prior where the size of genetic and residual variance were similar. Here we construct priors with a larger degree of belief parameter (nu), and we will specify that a large proportion (95%) of the variation is under genetic control (V). Thus, the residual variance contains 05% of the phenotypic variance.

```
p.var <- var(gryphon$bwt, na.rm = TRUE)
prior1.1.2 <- list(
  G = list(G1 = list(V = matrix(p.var * 0.95), nu = 1)),
  R = list(V = matrix(p.var * 0.05), nu = 1)
)

model1.1.2 <- MCMCglmm(bwt ~ 1,
  random = ~animal, ginv = list(animal = Ainv),
  data = gryphon, prior = prior1.1.2, nitt = 65000, thin = 50,
  burnin = 15000, verbose = FALSE
)
posterior.mode(model1.1$VCV)
```

```
##   animal     units
## 3.310074 3.728226
```

```
posterior.mode(model1.1.2$VCV)
```

```
##   animal     units
## 3.299524 4.026746
```

and we can therefore conclude that the difference in the priors has little effect on the outcome of the analysis. This is typical for an analysis where lots of data are available relative to the complexity of the model, but is often not the case. In all cases, it is important to check the effect of priors on conclusions drawn from a model. In addition, you can also specify the prior with previous knowledge or expectation for the variance.

2.4.4 Estimating heritability

A useful property of Bayesian posterior distributions is that we can apply almost any transformation to these distributions and they will remain valid. This applies to the calculation of heritability. We can obtain an estimate of the heritability by applying the basic formula $h^2 = V_A/V_P$ to each sample of the posterior distribution:

```
posterior.hereditability1.1 <- model1.1$VCV[, "animal"] /
  (model1.1$VCV[, "animal"] + model1.1$VCV[, "units"])
```

```
posterior.mode(posterior.heritability1.1)
```

```
##      var1
## 0.4603614
```

```
HPDinterval(posterior.heritability1.1, 0.95)
```

```
##      lower      upper
## var1 0.3236453 0.6079976
## attr(,"Probability")
## [1] 0.95
```

Generate a plot of the posterior distribution of this heritability estimate:

```
plot(posterior.heritability1.1)
```

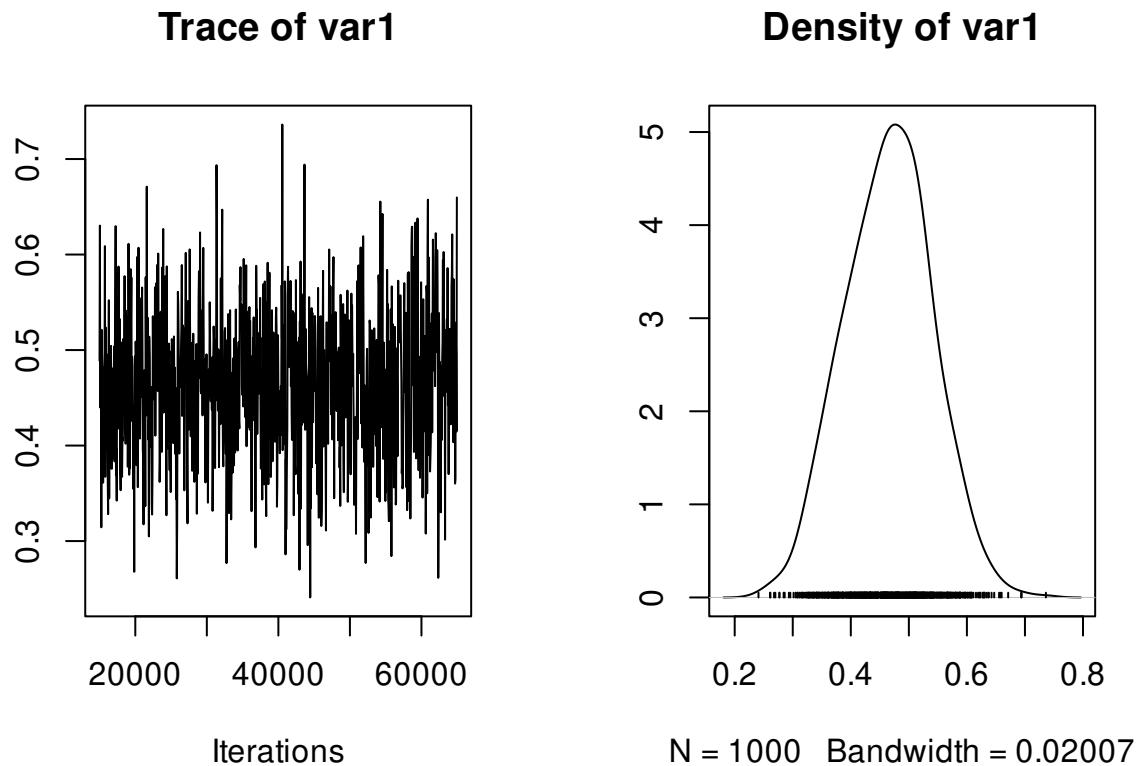


Figure 2.5: The posterior distributions the heritability from model 1.1

2.4.5 Adding fixed effects

To add effects to a univariate model, we simply modify the fixed effect part of the model specification:

```
model1.2 <- MCMCglmm(bwt ~ sex,
  random = ~animal, ginv = list(animal = Ainv),
  data = gryphon, prior = prior1.1,
  nitt = 65000, thin = 50, burnin = 15000, verbose = FALSE
)
summary(model1.2)
```

```
##
## Iterations = 15001:64951
## Thinning interval = 50
## Sample size = 1000
##
## DIC: 3719.3
##
## G-structure: ~animal
##
##          post.mean l-95% CI u-95% CI eff.samp
## animal      3.049    2.093    4.092    716.9
##
## R-structure: ~units
##
##          post.mean l-95% CI u-95% CI eff.samp
## units      2.974    2.168    3.74     775.5
##
## Location effects: bwt ~ sex
##
##          post.mean l-95% CI u-95% CI eff.samp pMCMC
## (Intercept)   6.053    5.726    6.359    1201 <0.001 ***
## sex2         2.214    1.895    2.536    1000 <0.001 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

We can assess the significance of `sex` as a fixed effect by examining its posterior distribution. Important notes here, it is important to know how the model names their fixed effect level to call them properly.

```
posterior.mode(model1.2$Sol[, "sex2"])
```

```
##     var1
## 2.16871
```

```
HPDinterval(model1.2$Sol[, "sex2"], 0.95)
```

```
##      lower     upper
## var1 1.89504 2.536397
## attr(,"Probability")
## [1] 0.95
```

The posterior distribution of the `sex2` term does not overlap zero. Thus, we can infer that sex has an effect on birth weight (presence of a sexual dimorphism) in this model and is a useful addition to the model, for most purposes. It is also worth noting that the variance components have changed slightly:

```
posterior.mode(model1.2$VCV)
```

```
##    animal     units
## 2.960190 3.117091
```

In fact since sex effects were previously contributing to the residual variance of the model our estimate of V_R (denoted 'units' in the output) is now slightly lower than before. This has an important consequence for estimating heritability since if we calculate V_P as $V_A + V_R$ then as we include fixed effects we will soak up more residual variance driving V_P . Assuming that V_A is more or less unaffected by the fixed effects fitted then as V_P goes down we expect our estimate of h^2 will go up.

```
posterior.heritability1.2 <- model1.2$VCV[, "animal"] /
  (model1.2$VCV[, "animal"] + model1.2$VCV[, "units"])
posterior.mode(posterior.heritability1.2)
```

```
##      var1
## 0.4915129
```

```
HPDinterval(posterior.heritability1.2, 0.95)
```

```
##      lower     upper
## var1 0.364643 0.6437182
## attr(,"Probability")
## [1] 0.95
```

Here h^2 has increased slightly from 0.4829 to 0.5079 (again, your values may differ slightly due to Monte Carlo error). Which is the better estimate? It depends on what your question is. The first is an estimate of the proportion of variance in birth weight explained by additive effects, the latter is an estimate of the proportion of variance in birth weight after conditioning on sex that is explained by additive effects. An important piece of advice, each researcher should be consistent in how they name their estimates and always correctly describe which estimates they are using conditional or not (to avoid any confusion).

2.4.6 Adding random effects

This is done by simply modifying the model statement in the same way, but requires addition of a prior for the new random effect. For instance, we can fit an effect of birth year:

```

prior1.3 <- list(
  G = list(G1 = list(V = 1, nu = 0.002), G2 = list(V = 1, nu = 0.002)),
  R = list(V = 1, nu = 0.002)
)

model1.3 <- MCMCglmm(bwt ~ sex,
  random = ~ animal + byear, ginv = list(animal = Ainv),
  data = gryphon,
  nitt = 65000, thin = 50, burnin = 15000,
  prior = prior1.3, verbose = FALSE
)

posterior.mode(model1.3$VCV)

##      animal      byear      units
## 2.7656801 0.8753865 2.3439002

```

Here the variance in birth weight explained by birth year is 0.88. Note that although V_A has changed somewhat, most of what is now partitioned as a `birth year` effect was previously partitioned as V_R . Thus what we have really done here is to partition environmental effects into those arising from year to year differences versus everything else, and we do not really expect much change in h^2 (since now $h^2 = V_A/(V_A + V_{BY} + V_R)$). However, we get a somewhat different result if we also add a random effect of `mother` to test for maternal effects:

```

prior1.4 <- list(
  G = list(
    G1 = list(V = 1, nu = 0.002),
    G2 = list(V = 1, nu = 0.002),
    G3 = list(V = 1, nu = 0.002)
  ),
  R = list(V = 1, nu = 0.002)
)

model1.4 <- MCMCglmm(bwt ~ sex,
  random = ~ animal + byear + mother,
  ginv = list(animal = Ainv), data = gryphon,
  nitt = 65000, thin = 50, burnin = 15000,
  prior = prior1.4, verbose = FALSE
)

posterior.mode(model1.4$VCV)

```

```
##     animal      byear     mother     units
## 2.5454307 0.7545662 1.0474161 1.8486924
```

Here partitioning of significant maternal variance has resulted in a further decrease in V_R but also a decrease in V_A . The latter is because maternal effects of the sort we simulated (fixed differences between mothers) will have the consequence of increasing similarity among maternal siblings. Consequently they can look very much like an additive genetic effects and if present, but unmodelled, represent a type of ‘common environment effect’ that can - and will- cause upward bias in V_A and so h^2 . Let’s compare the estimates of heritability from each of models 1.2, 1.3 and 1.4:

```
posterior.heritability1.3 <- model1.3$VCV[, "animal"] /
  (model1.3$VCV[, "animal"] + model1.3$VCV[, "byear"] + model1.3$VCV[, "units"])
posterior.heritability1.4 <- model1.4$VCV[, "animal"] /
  (model1.4$VCV[, "animal"] + model1.4$VCV[, "byear"] + model1.4$VCV[, "mother"] + model1.
posterior.mode(posterior.heritability1.2)
```

```
##     var1
## 0.4915129
```

```
posterior.mode(posterior.heritability1.3)
```

```
##     var1
## 0.4536882
```

```
posterior.mode(posterior.heritability1.4)
```

```
##     var1
## 0.3790414
```

2.4.7 Testing significance of variance components

While testing the significance of fixed effects by evaluating whether or not their posterior distributions overlap zero was simple and valid, this approach does not work for variance components. Variance components are bounded to be positive (given a proper prior), and thus even when a random effect is not meaningful, its posterior distribution will never overlap zero. Model comparisons can be performed using the deviance information criterion (DIC), although it should be noted that the properties of DIC are not well understood and that the DIC may be focused at the wrong level for most people’s intended level of inference - particularly with non-Gaussian responses. The implementation of DIC in MCMCglmm is further described in the reference manual. DIC values are calculated by MCMCglmm by default. Briefly, DIC like other information criteria balance model fit and model complexity simultaneously, and small values of DIC are preferred. We can compare models 1.4 and 1.3, i.e., models with and without the mother term:

```
model1.3$DIC
```

```
## [1] 3548.159
```

```
model1.4$DIC
```

```
## [1] 3327.724
```

model 1.4 has a much lower DIC value. Since the maternal effect term is the only difference between the models, we can consider the inclusion of this term statistically justifiable. We should note however that DIC has a large sampling variance and should probably only be calculated based on much longer MCMC runs.

2.4.8 Further partitioning variance

A population can be further fragmented into different groups or categories (such as females and males, juveniles and adults or treated and untreated). Some scientific questions require further and deeper analysis of the variance. To avoid multiple model (one for each group), we can directly partition the variance between groups in a unique model. In addition, by doing so, we can also test if the variance are different between groups.

As example, we can partition the additive genetic variance and residual variance by sex. It is impossible to further partition the other variances but complexity an animal model requires sufficient sample size.

```
prior1.4.SEX <- list(
  G = list(G1 = list(V = diag(2), nu = 1.002), G2 = list(V = 1, nu = 0.002), G3 = list(V =
  R = list(V = diag(2), nu = 1.002)
)
```

```
model1.4.SEX <- MCMCglmm(bwt ~ sex,
  random = ~ idh(sex):animal + byear + mother,
  rcov = ~ idh(sex):units,
  ginv = list(animal = Ainv), data = gryphon, nitt = 65000, thin = 50, burnin = 15000,
  prior = prior1.4.SEX, verbose = FALSE
)
```

```
posterior.mode(model1.4.SEX$VCV)
```

sex1.animal	sex2.animal	byear	mother	sex1.units	sex2.units
1.2979861	1.6649978	0.7149751	1.2746095	2.4459575	0.9091363

```
posterior.heredity1.4.FEM <- model1.4.SEX$VCV[, "sex1.animal"] /
  (model1.4.SEX$VCV[, "sex1.animal"] + model1.4.SEX$VCV[, "byear"] +
  model1.4.SEX$VCV[, "mother"] + model1.4.SEX$VCV[, "sex1.units"])
```

```
posterior.heritability1.4.MAL <- model1.4.SEX$VCV[, "sex2.animal"] /
  (model1.4.SEX$VCV[, "sex2.animal"] + model1.4.SEX$VCV[, "byear"] +
    model1.4.SEX$VCV[, "mother"] + model1.4.SEX$VCV[, "sex2.units"])
```

```
posterior.mode(posterior.heritability1.4.FEM)
```

```
##      var1
## 0.2507729
```

```
HPDinterval(posterior.heritability1.4.FEM, 0.95)
```

```
##           lower      upper
## var1 0.04358691 0.4369285
## attr(,"Probability")
## [1] 0.95
```

```
posterior.mode(posterior.heritability1.4.MAL)
```

```
##      var1
## 0.4497827
```

```
HPDinterval(posterior.heritability1.4.MAL, 0.95)
```

```
##           lower      upper
## var1 0.1041657 0.6197678
## attr(,"Probability")
## [1] 0.95
```

Here, we can estimate the heritability for each sex. Both doesn't overlap with zero, so we can conclude both sexes have significant heritability. However due to their overlaps CIs, we can not conclude the heritability is not significantly different between sexes. An important quote to remember is “A difference in significance is not a significant difference”

```
h2.sex <- rbind(
  cbind(posterior.mode(posterior.heritability1.4.FEM), HPDinterval(posterior.heritability1.4.FEM)),
  cbind(posterior.mode(posterior.heritability1.4.MAL), HPDinterval(posterior.heritability1.4.MAL))
)
```

```
plot(c(0.95, 1.05) ~ h2.sex[, 1], xlim = c(0, 0.8), ylim = c(0.5, 1.5), , xlab = "", ylab =
arrows(y0 = 0.95, x0 = h2.sex[1, 2], y1 = 0.95, x1 = h2.sex[1, 3], code = 3, angle = 90, lwd = 2)
arrows(y0 = 1.05, x0 = h2.sex[2, 2], y1 = 1.05, x1 = h2.sex[2, 3], code = 3, angle = 90, lwd = 2)
mtext("Narrow-sense heritability (±CI)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6)
axis(2, at = 1, labels = c("birth weight"), las = 3, cex.axis = 1.6)
```

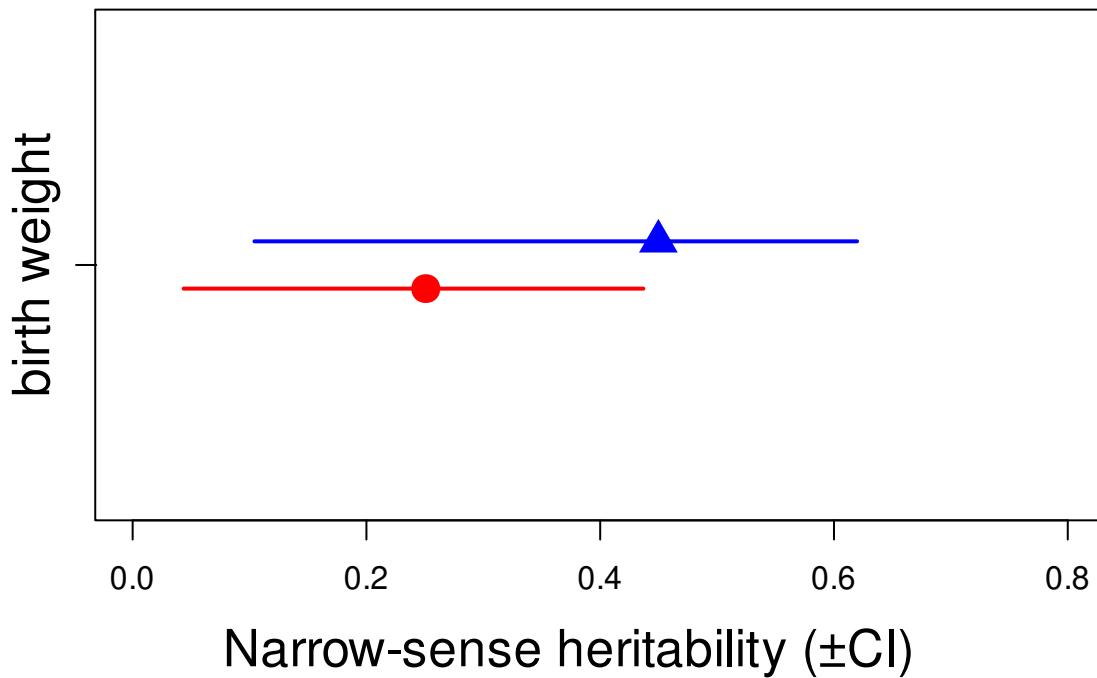


Figure 2.6: Female and male heritability of birth weight

2.4.9 Modification of model parameter

Unfortunately (to our knowledge), it is not possible to alter the variance matrices and refit them within the model.

2.4.10 Covariance between two random effects

Some research questions require to estimate the covariance between two random effects within a univariate model. To do so, we can use the argument `str`. A similar argument or linking.function `mm` can be used but it will forced the variance of `animal` and `mother` to be equal and the covariance to 1. As an example, we fit a model which estimate the covariance between the additive genetic variance and the mother variance. Both variances require to operate on the same level, thus `animal` and `mother` require to be associated to the pedigree information. The `ginverse` list name has to correspond to the first term in the argument or linking.function

2.5 brms

2.5.1 Running the model

First we need to load the `brms` library:

```
library(brms)
```

To be able to fit an animal model, `brms` needs the relativeness (relationship) matrix of the pedigree and not its inverse (as in other softwares). This can be estimated using the `nadiv` package created by Pr. Matthew Wolak (<https://cran.r-project.org/web/packages/nadiv/index.html>).

```
Amat <- as.matrix(nadiv::makeA(gryphonped))
```

We are now ready to specify our first model: The structure of a `bmrs` model is similar to `lme4`, thus the random effect is added to the model with the term `(1 | gr(animal, cov = Amat))` which associate the id animal to the matrix of relativeness. In addition to the synthase of `lme4`, we includes other features or parameters within the models such as `chain` which represent the number of Markov chains (defaults to 4), `core` which represents the number of cores to use when executing the chains in parallel and `iter` which represents the number of total iterations per chain. For more parameters such as `thin` or `warmup/burnin`, you can read the Cran R page of the package (<https://cran.r-project.org/web/packages/brms/brms.pdf>)

`bmrs` is a Bayesian Multilevel Models using Stan, doing so we can apply a prior to the model to better shape the distribution of the different variances estimated by the model. Given that `bmrs` fit the model using a Bayesian approach via the software `stan`, we need to specify priors for the model. Default priors in `brms` work relatively well, however we strongly suggest to carefully select an adequate prior for your analysis. In this tutorial we will use the default priors. To get the prior used by default, we can use the `get_prior()` function.

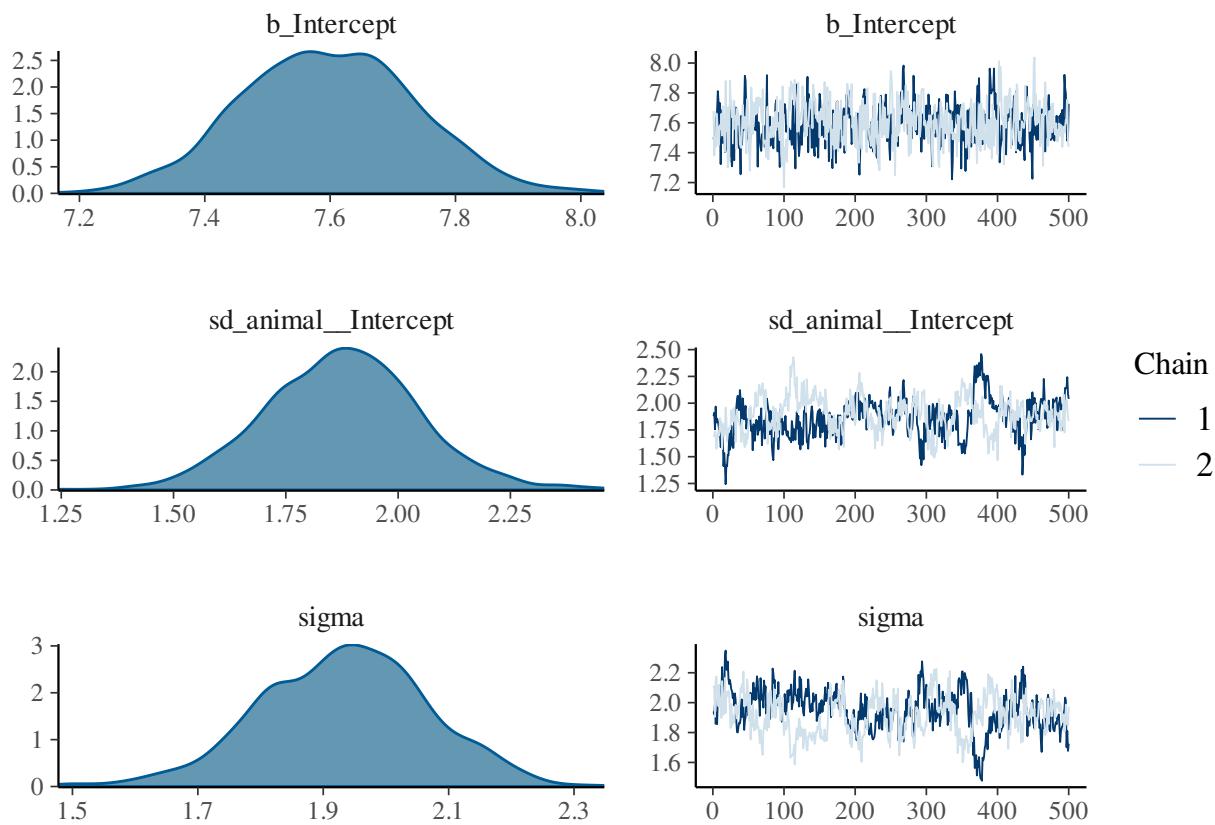
```
brms_m1.1 <- brm(
  bwt ~ 1 + (1 | gr(animal, cov = Amat)),
  data = gryphon,
  data2 = list(Amat = Amat),
  family = gaussian(),
  chains = 1, cores = 1, iter = 100
)

save(brms_m1.1, file = "data/brms_m1_1.rda")
```

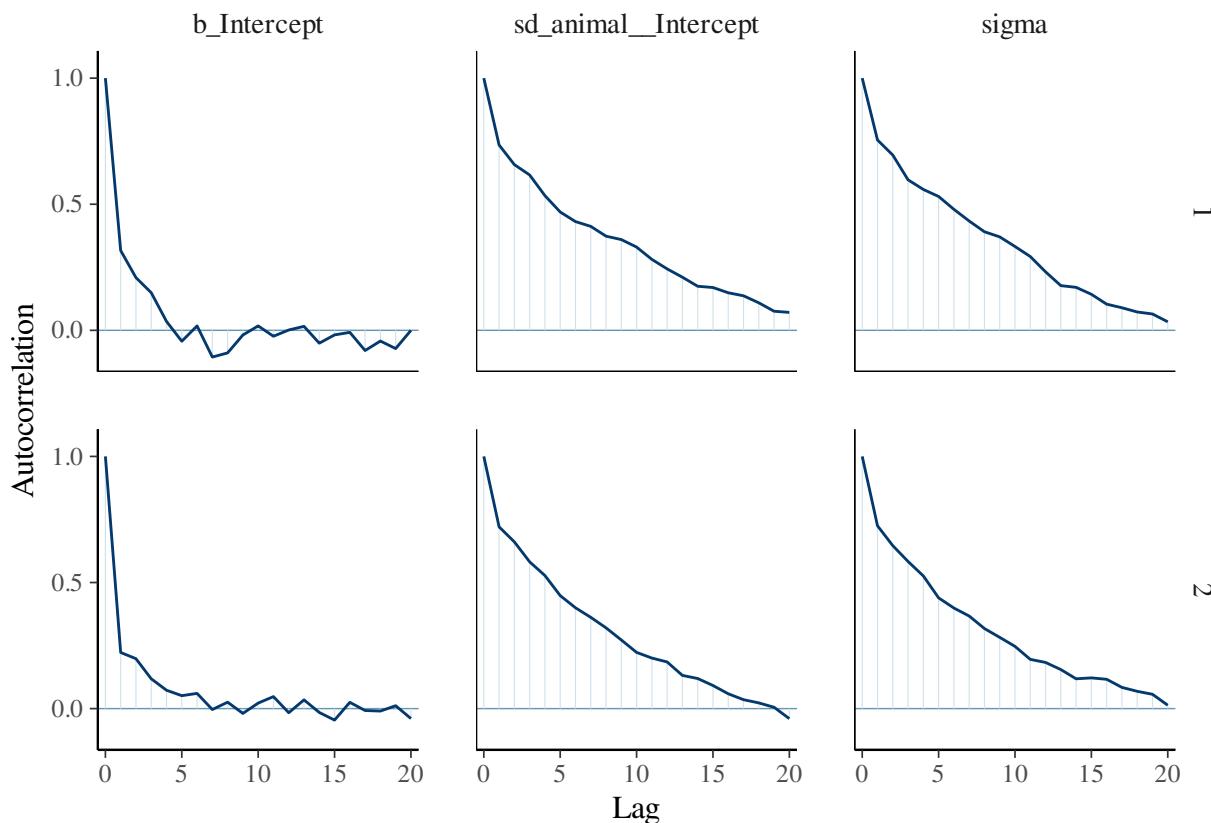
The result of the long model calculation is save in a spare file `brms_m1_1.rda`. To help readers, we can directly reloading it. Two distinct plot can be produce to produce some diagnostics graphs `mcmc_plot`. Note, that `sigma` represents the residual standard deviation.

Next, we examine (or directly using the model) the variance estimate and their distributions (via `summary` or `plot`).

```
load("data/brms_m1_1.rda")
plot(brms_m1.1)
```



```
mcmc_plot(brms_m1.1, type = "acf")
```



```
summary(brms_m1.1)
```

```
## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: bwt ~ 1 + (1 | gr(animal, cov = Amat))
## Data: gryphon (Number of observations: 854)
## Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##        total post-warmup draws = 1000
##
## Group-Level Effects:
## ~animal (Number of levels: 854)
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(Intercept)    1.88     0.17     1.54     2.23 1.03      74      99
##
## Population-Level Effects:
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## Intercept       7.60     0.14     7.33     7.86 1.01      428      727
##
## Family Specific Parameters:
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma         1.93     0.13     1.66     2.18 1.04      71      112
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
```

```
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
```

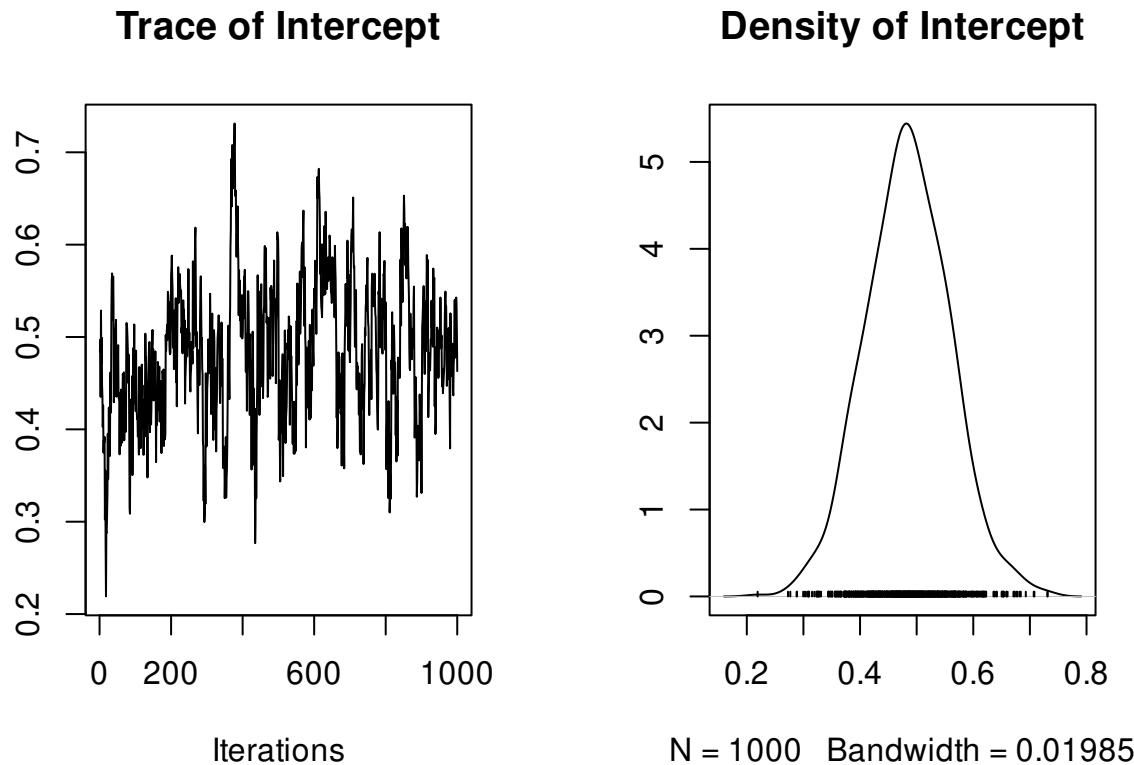
The `plot` of variance showed that the different variances have an normal distribution, the auto-correlation plot or ‘acf’ show that the autocorrelation is close to 0. The `summary` exposes the mean (Estimate) of each variance or fixed effect (here just the intercept) associated to their posterior distribution with standard deviation (Est.Error) and two-sided 95% Credible intervals. `Rhat` provides information on the estimate convergence. If it’s greater than 1, the chains have not yet converged and it will be require to run more iterations and/or set stronger priors. `ESS` represents the Effective sample values as the number of independent samples from the posterior distribution. However, for the purpose of this guide, the `Rhat` values are acceptable.

It is also possible to calculate the heritability using the function ‘`as.mcmc`’

```
v_animal <- (VarCorr(brms_m1.1, summary = FALSE)$animal$sd)^2
v_r <- (VarCorr(brms_m1.1, summary = FALSE)$residual$sd)^2
h.bwt.1 <- as.mcmc(v_animal / (v_animal + v_r))
summary(h.bwt.1)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD     Naive SE Time-series SE
## 0.484221  0.074533  0.002357  0.009275
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## 0.3433 0.4338 0.4841 0.5350 0.6369
```

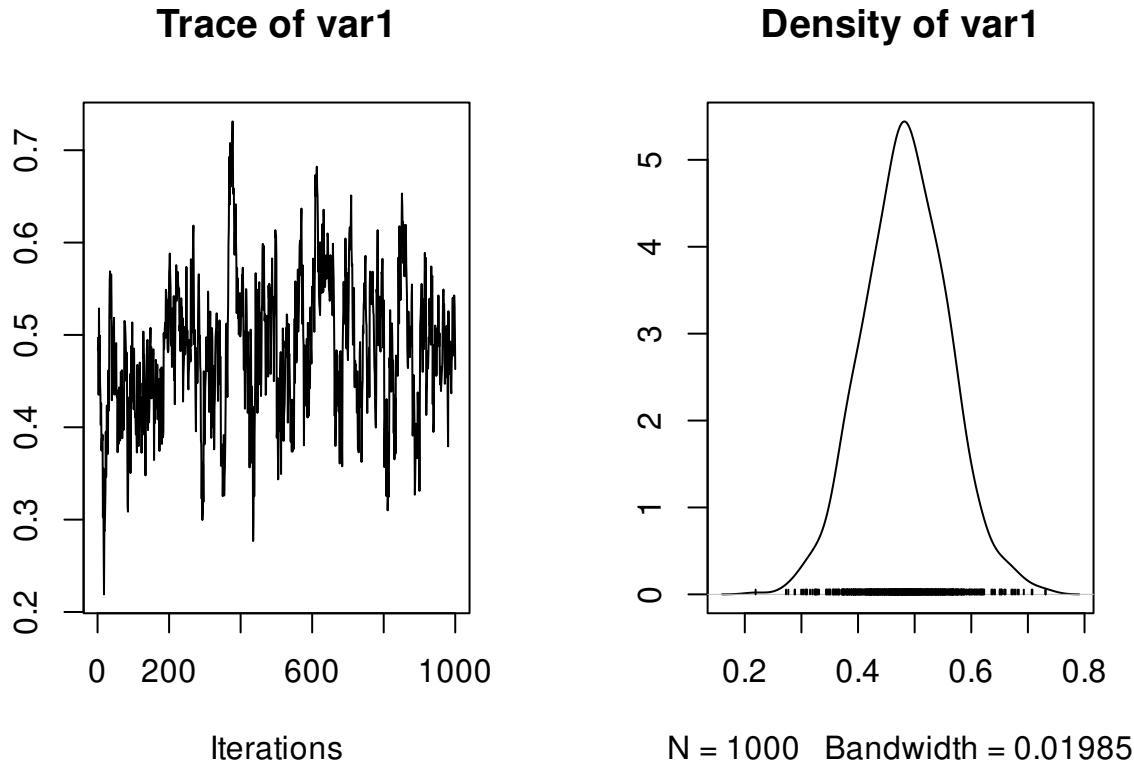
```
plot(h.bwt.1)
```



```
# or
Var.table <- as_draws_df(brms_m1.1)
Var.table$h.bwt.1 <- as.mcmc((Var.table$sd_animal_Intercept)^2 / ((Var.table$sd_animal_I
summary(Var.table$h.bwt.1)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD     Naive SE Time-series SE
## 0.484221 0.074533 0.002357 0.009275
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## 0.3433 0.4338 0.4841 0.5350 0.6369
```

```
plot(Var.table$h.bwt.1)
```



2.5.2 Adding fixed effects

To add effects to a univariate model, we simply modify the priors and the fixed effect portion of the model specification:

```
brms_m1.2 <- brm(
  bwt ~ 1 + sex + (1 | gr(animal, cov = Amat)),
  data = gryphon,
  data2 = list(Amat = Amat),
  family = gaussian(),
  chains = 2, cores = 2, iter = 1000
)

save(brms_m1.2, file = "data/brms_m1_2.rda")
```

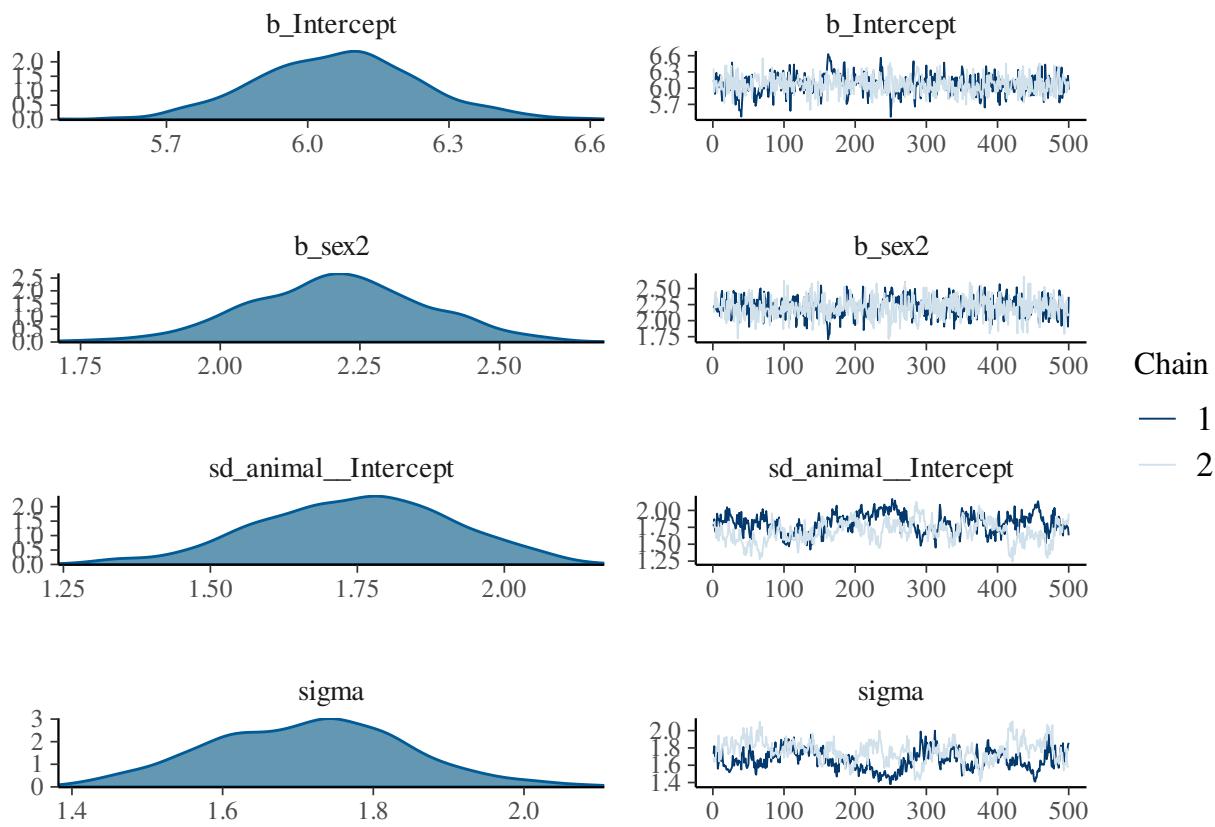
To save time, the results of the calculation is stored in the spare file `brms_m1_2.rda`. We can assess the significance of `sex` as a fixed effect by examining its posterior distribution.

```
load("data/brms_m1_2.rda")
summary(brms_m1.2)
```

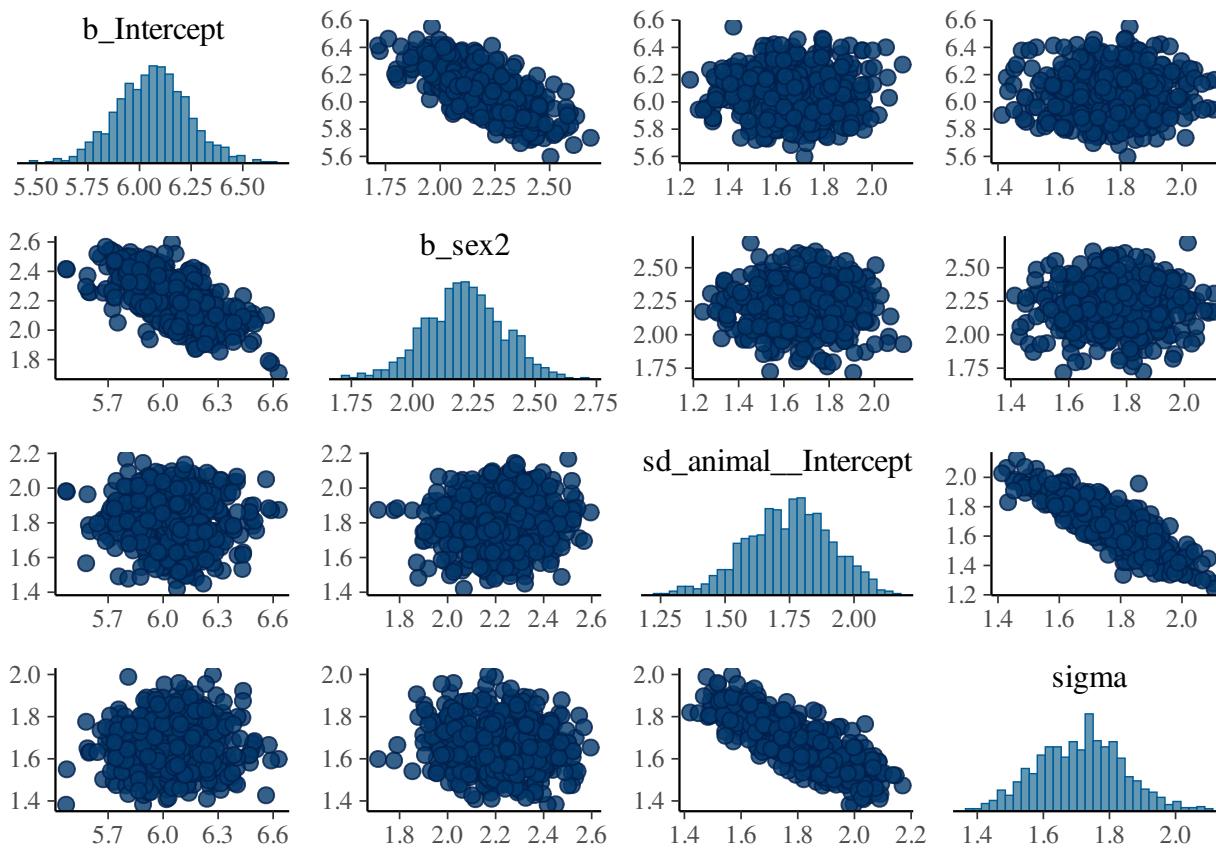
```
## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.
```

```
## Family: gaussian
##   Links: mu = identity; sigma = identity
## Formula: bwt ~ 1 + sex + (1 | gr(animal, cov = Amat))
##   Data: gryphon (Number of observations: 854)
##   Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##          total post-warmup draws = 1000
##
## Group-Level Effects:
##   ~animal (Number of levels: 854)
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(Intercept)     1.75      0.16     1.42    2.05 1.13        13      113
##
## Population-Level Effects:
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## Intercept      6.06      0.17     5.74    6.41 1.00       358      574
## sex2           2.21      0.16     1.90    2.52 1.00       723      657
##
## Family Specific Parameters:
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma         1.71      0.13     1.47    1.98 1.12        14      97
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
```

```
plot(brms_m1.2)
```



```
mcmc_plot(brms_m1.2, type = "pairs")
```



```
summary(brms_m1.2)$fixed
```

```
## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

##           Estimate Est.Error l-95% CI u-95% CI      Rhat Bulk_ESS Tail_ESS
## Intercept 6.064853 0.1726459 5.735170 6.410117 1.002990 357.5666 574.2620
## sex2       2.210675 0.1574542 1.898645 2.520026 1.002313 722.8016 657.2121
```

```
summary(brms_m1.2)$random
```

```
## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## $animal
##           Estimate Est.Error l-95% CI u-95% CI      Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 1.747083 0.1632548 1.419377 2.050884 1.126639 13.10221 113.2445
```

The posterior distribution of the `sex2` term does not overlap zero. Thus, we can infer that sex has an effect on birth weight (presence of a sexual dimorphism) in this model and is a useful addition to the model, for most purposes. It is also worth noting that the variance components have changed slightly:

```
summary(brms_m1.2)$random
```

```
## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.
```

```
## $animal
```

	Estimate	Est.Error	l-95% CI	u-95% CI	Rhat	Bulk_ESS	Tail_ESS
## sd(Intercept)	1.747083	0.1632548	1.419377	2.050884	1.126639	13.10221	113.2445

In fact since sex effects were previously contributing to the residual variance of the model our estimate of V_R (denoted 'units' in the output) is now slightly lower than before. This has an important consequence for estimating heritability since if we calculate V_P as $V_A + V_R$ then as we include fixed effects we will soak up more residual variance driving V_P . Assuming that V_A is more or less unaffected by the fixed effects fitted then as V_P goes down we expect our estimate of h^2 will go up.

```
v_animal <- (VarCorr(brms_m1.2, summary = FALSE)$animal$sd)^2
v_r <- (VarCorr(brms_m1.2, summary = FALSE)$residual$sd)^2
h.bwt.2 <- as.mcmc(v_animal / (v_animal + v_r))
```

```
summary(h.bwt.2)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD      Naive SE Time-series SE
## 0.508677  0.080357  0.002541   0.011998
##
## 2. Quantiles for each variable:
##
##    2.5%    25%    50%    75%   97.5%
## 0.3427  0.4549  0.5107  0.5675  0.6576
```

```
summary(h.bwt.1)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
```

```

## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean          SD   Naive SE Time-series SE
## 0.484221     0.074533     0.002357      0.009275
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%   97.5%
## 0.3433 0.4338 0.4841 0.5350 0.6369

```

Here h^2 has increased slightly from 0.5010 to 0.4192 (again, your values may differ slightly due to Monte Carlo error). Which is the better estimate? It depends on what your question is. The first is an estimate of the proportion of variance in birth weight explained by additive effects, the latter is an estimate of the proportion of variance in birth weight after conditioning on sex that is explained by additive effects. An important piece of advice, each researcher should be consistent in how they name their estimates and always correctly describe which estimates they are using conditional or not (to avoid any confusion).

2.5.3 Adding random effects

This is done by simply modifying the model statement in the same way, but requires addition of a prior for the new random effect. For instance, we can fit an effect of birth year:

```

brms_m1.3 <- brm(
  bwt ~ 1 + sex + (1 | gr(animal, cov = Amat)) + (1 | byear) + (1 | mother),
  data = gryphon,
  data2 = list(Amat = Amat),
  family = gaussian(),
  chains = 2, cores = 2, iter = 1000
)

save(brms_m1.3, file = "data/brms_m1_3.rda")

```

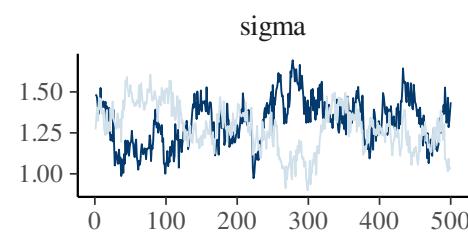
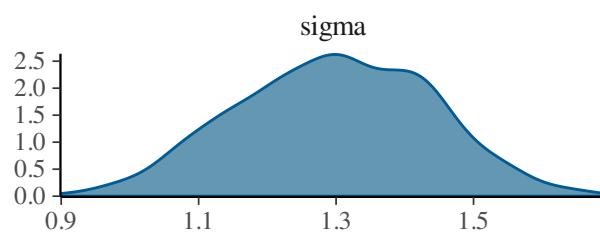
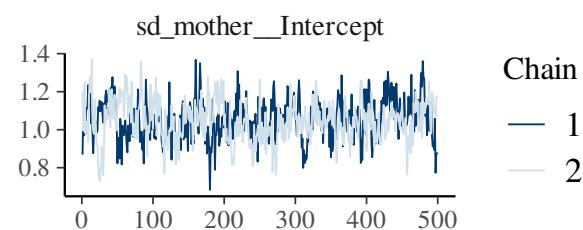
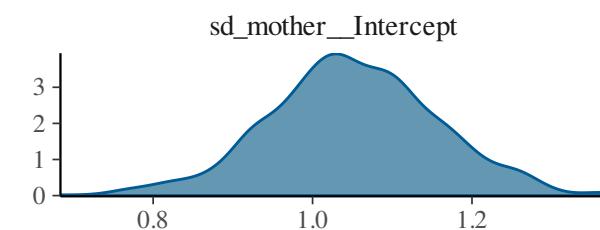
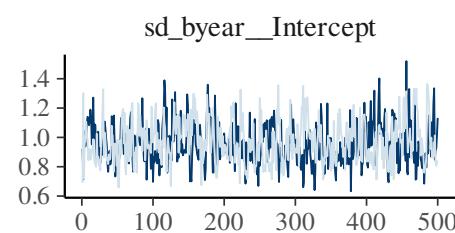
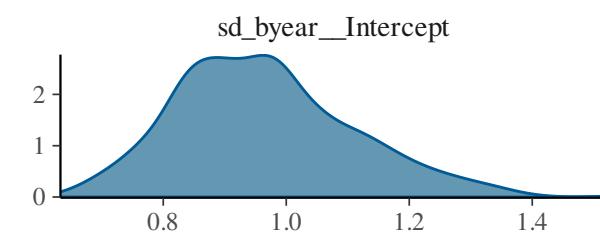
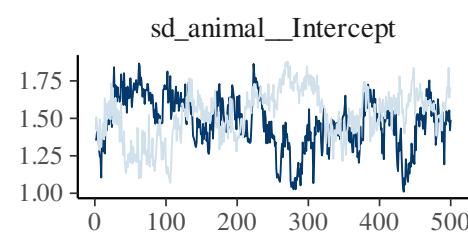
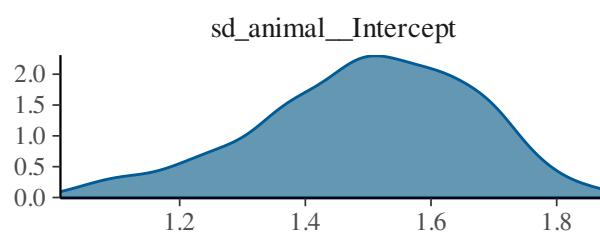
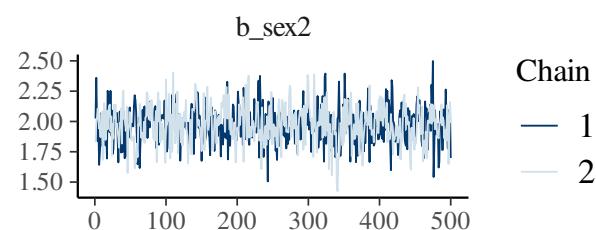
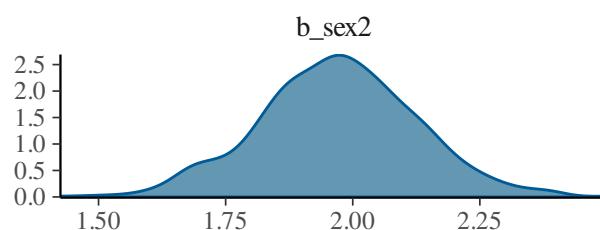
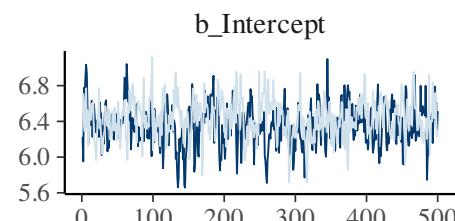
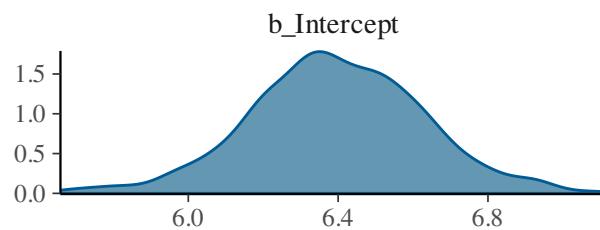
To save time, the results of the calculation is stored in the spare file `brms_m1_3.rda`. We can assess the significance of `sex` as a fixed effect by examining its posterior distribution.

```

load("data/brms_m1_3.rda")

plot(brms_m1.3, ask = FALSE, N = 3)

```



```
summary(brms_m1.3)$random

## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## $animal
##           Estimate Est.Error l-95% CI u-95% CI      Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 1.496464 0.1708587 1.105797 1.782062 1.163193 9.153202 50.78299
##
## $byear
##           Estimate Est.Error l-95% CI u-95% CI      Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 0.9642702 0.144174 0.7147103 1.280642 1.002215 399.5806 642.4663
##
## $mother
##           Estimate Est.Error l-95% CI u-95% CI      Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 1.050225 0.1057319 0.8292744 1.258439 1.011062 192.6606 379.2063
```

Here partitioning of significant birth year and maternal variance has resulted in a further decrease in V_R but also a decrease in V_A . The latter is because maternal effects of the sort we simulated (fixed differences between mothers) will have the consequence of increasing similarity among maternal siblings. Consequently they can look very much like an additive genetic effects and if present, but unmodelled, represent a type of ‘common environment effect’ that can - and will- cause upward bias in V_A and so h^2 . Let’s compare the estimates of heritability from each of models 1.2, 1.3 and 1.4:

```
v_animal <- (VarCorr(brms_m1.3, summary = FALSE)$animal$sd)^2
v_byear <- (VarCorr(brms_m1.3, summary = FALSE)$byear$sd)^2
v_mother <- (VarCorr(brms_m1.3, summary = FALSE)$mother$sd)^2
v_r <- (VarCorr(brms_m1.3, summary = FALSE)$residual$sd)^2
h.bwt.3 <- as.mcmc(v_animal / (v_animal + v_byear + v_mother + v_r))
summary(h.bwt.3)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean          SD      Naive SE Time-series SE
## 0.375509   0.078767   0.002491     0.012711
##
## 2. Quantiles for each variable:
```

```
## 
##   2.5%    25%    50%    75%  97.5%
## 0.2140 0.3240 0.3764 0.4322 0.5182
```

```
summary(h.bwt.2)
```

```
## 
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD     Naive SE Time-series SE
## 0.508677 0.080357 0.002541 0.011998
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## 0.3427 0.4549 0.5107 0.5675 0.6576
```

```
summary(h.bwt.1)
```

```
## 
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD     Naive SE Time-series SE
## 0.484221 0.074533 0.002357 0.009275
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## 0.3433 0.4338 0.4841 0.5350 0.6369
```

or

```
Var.table <- as_draws_df(brms_m1.3)
```

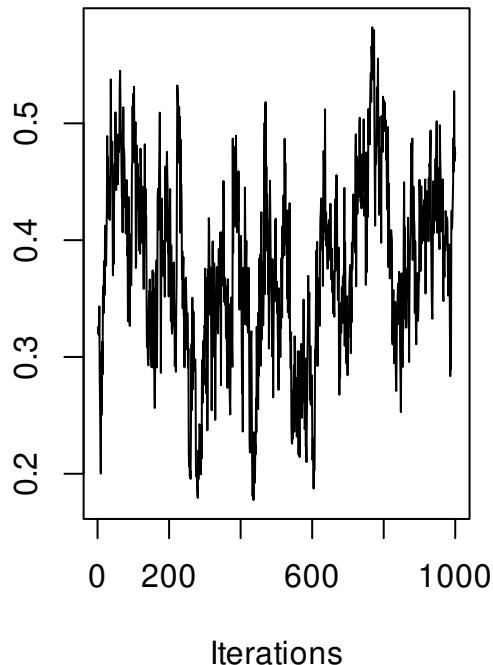
```
Var.table$h.bwt.3 <- as.mcmc((Var.table$sd_animal_Intercept)^2 / ((Var.table$sd_animal_I
```

```
summary(Var.table$h.bwt.3)
```

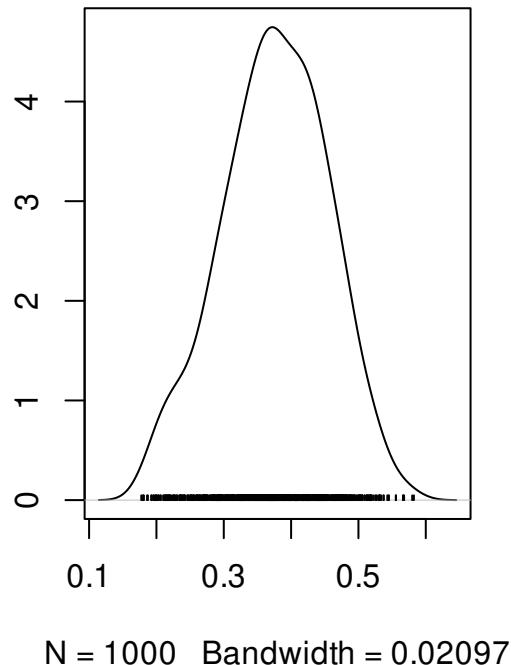
```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##     plus standard error of the mean:
##
##           Mean        SD    Naive SE Time-series SE
## 0.375509   0.078767  0.002491   0.012711
##
## 2. Quantiles for each variable:
##
##    2.5%    25%    50%    75%   97.5%
## 0.2140  0.3240  0.3764  0.4322  0.5182
```

```
plot(Var.table$h.bwt.3)
```

Trace of var1



Density of var1



2.5.4 Testing significance of variance components

While testing the significance of fixed effects by evaluating whether or not their posterior distributions overlap zero was simple and valid, this approach does not work for variance components. Variance components are bounded to be positive (given a proper prior), and thus even when a random effect is not meaningful, its posterior distribution will never overlap zero.

Model comparisons can be performed using the function `loo_compare` using `waic` or weighted AIC.

```
brms_m1.3 <- add_criterion(brms_m1.3, "waic")

## Warning:
## 311 (36.4%) p_waic estimates greater than 0.4. We recommend trying loo instead.

brms_m1.1 <- add_criterion(brms_m1.1, "waic")

## Warning:
## 236 (27.6%) p_waic estimates greater than 0.4. We recommend trying loo instead.

loo_compare(brms_m1.3, brms_m1.1, criterion = "waic")

##           elpd_diff se_diff
## brms_m1.3      0.0      0.0
## brms_m1.1 -282.7     14.0
```

2.5.5 Further partitioning of the variance

Depending of the research question and the presence of different group within the dataset, `brms` allowed to partition the variance at different groups. Two distinct approach can be done to partition the different random effect: using an extra argument `by=sex` or by adding `(0+sex|)` before the `|`. Notes, here we used `||` which not estimate a possible covariance between groups (female and male) for the random effect.

```
brms_m1.4 <- brm(
  # bwt ~ 1 + sex + (1 / gr(animal, cov = Amat, by = sex)) + (1 / gr(byear, by = sex)) +
  bwt ~ 1 + sex + (0 + sex || gr(animal, cov = Amat)) + (0 + sex || byear) + (0 + sex || m
  data = gryphon,
  data2 = list(Amat = Amat),
  family = gaussian(),
  chains = 2, cores = 2, iter = 1000
)

save(brms_m1.4, file = "data/brms_m1_4.rda")
```

To save time, the results of the calculation is stored in the spare file `brms_m1_4.rda`.

```

load("data/brms_m1_4.rda")
summary(brms_m1.4)

## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## Family: gaussian
##   Links: mu = identity; sigma = identity
## Formula: bwt ~ 1 + sex + (0 + sex || gr(animal, cov = Amat)) + (0 + sex || byear) + (0 +
##   Data: gryphon (Number of observations: 854)
##   Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##          total post-warmup draws = 1000
##
## Group-Level Effects:
##   ~animal (Number of levels: 854)
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
##   sd(sex1)     1.39     0.22     0.88     1.77 1.02      56     116
##   sd(sex2)     1.06     0.31     0.35     1.57 1.06      29      63
##
##   ~byear (Number of levels: 34)
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
##   sd(sex1)     0.91     0.18     0.61     1.29 1.02     381     542
##   sd(sex2)     1.09     0.20     0.77     1.56 1.01     310     626
##
##   ~mother (Number of levels: 394)
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
##   sd(sex1)     0.90     0.21     0.47     1.29 1.01     143     219
##   sd(sex2)     1.34     0.16     1.00     1.65 1.01     143     396
##
## Population-Level Effects:
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## Intercept    6.29     0.24     5.85     6.78 1.01     435     444
## sex2         2.03     0.35     1.31     2.69 1.00     568     679
##
## Family Specific Parameters:
##             Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma        1.45     0.16     1.09     1.72 1.05      27      35
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

```

We can see the model estimate variance for both sexes. However, the residual level or sigma is not splitted by sexes. A futher and more complex code need to be performed, thus we can estimate the sex-specific heritability.

```

bf_m1.5 <- bf(
  bwt ~ 1 + sex + (0 + sex || gr(animal, cov = Amat)) + (0 + sex || mother) + (0 + sex || 
  sigma ~ sex - 1
)

brms_m1.5 <- brm(bf_m1.5,
  data = gryphon,
  data2 = list(Amat = Amat),
  family = gaussian(),
  chains = 1, cores = 1, iter = 1000
)

save(brms_m1.5, file = "data/brms_m1_5.rda")

```

To save time, the results of the calculation is stored in the spare file `brms_m1_4.rda`.

```

load("data/brms_m1_5.rda")
summary(brms_m1.5)

## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## Family: gaussian
##   Links: mu = identity; sigma = log
## Formula: bwt ~ 1 + sex + (0 + sex || gr(animal, cov = Amat)) + (0 + sex || mother) + (0
##           sigma ~ sex - 1
## Data: gryphon (Number of observations: 854)
## Draws: 1 chains, each with iter = 1000; warmup = 500; thin = 1;
##        total post-warmup draws = 500
##
## Group-Level Effects:
##   ~animal (Number of levels: 854)
##     Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
##   sd(sex1)    1.56      0.29     1.02     2.09 1.17       4      30
##   sd(sex2)    1.61      0.41     0.52     2.08 1.36       2      21
##
##   ~byear (Number of levels: 34)
##     Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
##   sd(sex1)    0.91      0.18     0.59     1.36 1.01      153     229
##   sd(sex2)    1.06      0.20     0.75     1.49 1.00      170     143
##
##   ~mother (Number of levels: 394)
##     Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
##   sd(sex1)    0.88      0.21     0.41     1.25 1.01       73     134
##   sd(sex2)    1.27      0.18     0.88     1.59 1.01       31      64

```

```

##  

## Population-Level Effects:  

##           Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS  

## Intercept     6.29      0.23    5.88    6.75 1.00     209     313  

## sex2         2.02      0.31    1.49    2.66 1.00     127     296  

## sigma_sex1   0.22      0.21   -0.25    0.54 1.15      5      12  

## sigma_sex2  -0.20      0.40   -0.82    0.54 1.59      2      15  

##  

## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS  

## and Tail_ESS are effective sample size measures, and Rhat is the potential  

## scale reduction factor on split chains (at convergence, Rhat = 1).  

#  

Var.table <- as_draws_df(brms_m1.5)  

Var.table$h.bwt.f <- as.mcmc((Var.table$sd_animal__sex1)^2 / ((Var.table$sd_animal__sex1)^  

Var.table$h.bwt.m <- as.mcmc((Var.table$sd_animal__sex2)^2 / ((Var.table$sd_animal__sex2)^  

summary(Var.table$h.bwt.f)  

##  

## Iterations = 1:500  

## Thinning interval = 1  

## Number of chains = 1  

## Sample size per chain = 500  

##  

## 1. Empirical mean and standard deviation for each variable,  

##     plus standard error of the mean:  

##  

##           Mean          SD       Naive SE Time-series SE  

##     0.575443     0.126621     0.005663     0.031251  

##  

## 2. Quantiles for each variable:  

##  

##   2.5%    25%    50%    75%  97.5%  

## 0.3075  0.4863  0.5811  0.6741  0.7800  

summary(Var.table$h.bwt.m)  

##  

## Iterations = 1:500  

## Thinning interval = 1  

## Number of chains = 1  

## Sample size per chain = 500  

##  

## 1. Empirical mean and standard deviation for each variable,  

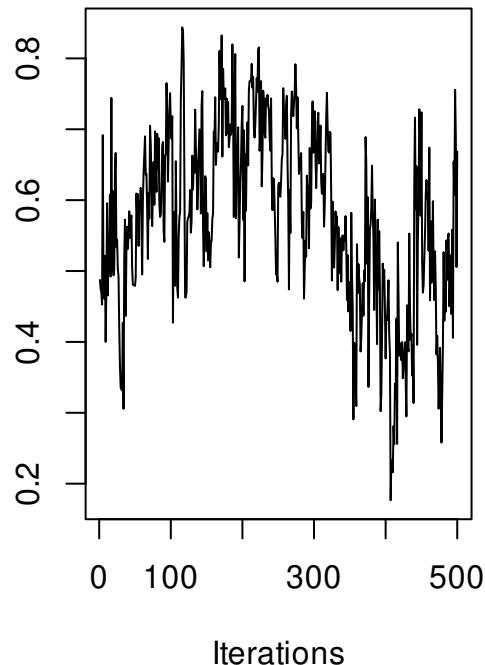
##     plus standard error of the mean:  

##
```

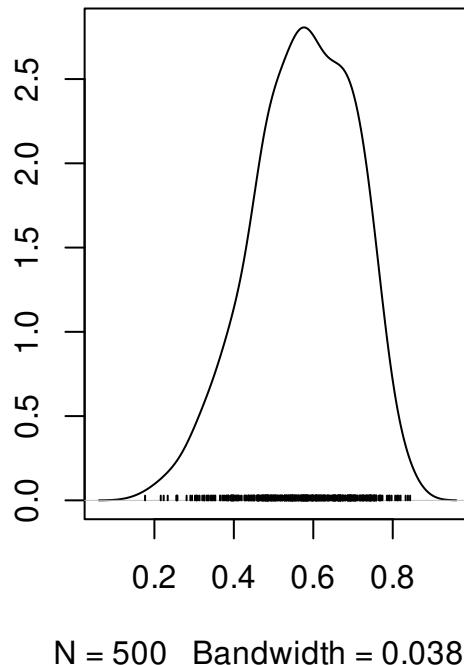
```
##           Mean          SD      Naive SE Time-series SE
##       0.463879   0.155395  0.006949    0.078323
## 
## 2. Quantiles for each variable:
## 
##    2.5%     25%     50%     75%   97.5%
## 0.06693 0.43668 0.50150 0.55729 0.66016
```

```
plot(Var.table$h.bwt.f)
```

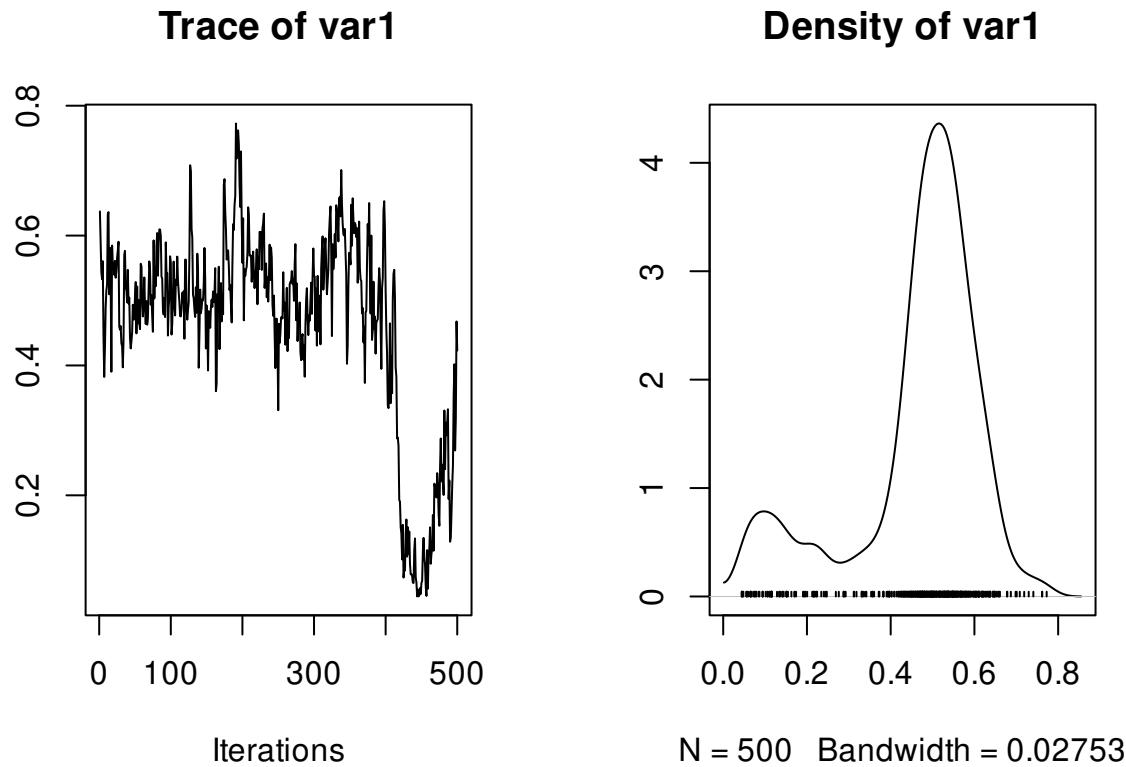
Trace of var1



Density of var1



```
plot(Var.table$h.bwt.m)
```



Here, we can plot the point estimates of the h^2 which seems to differ between sexes, but their CI overlaps.

```

h2.sex <- rbind(
  cbind(summary(Var.table$h.bwt.f)$statistics[1], summary(Var.table$h.bwt.f)$quantiles[1]),
  cbind(summary(Var.table$h.bwt.m)$statistics[1], summary(Var.table$h.bwt.m)$quantiles[1]),
)

plot(c(0.95, 1.05) ~ h2.sex[, 1], xlim = c(0, 0.8), ylim = c(0.5, 1.5), , xlab = "", ylab =
arrows(y0 = 0.95, x0 = h2.sex[1, 2], y1 = 0.95, x1 = h2.sex[1, 3], code = 3, angle = 90, l
arrows(y0 = 1.05, x0 = h2.sex[2, 2], y1 = 1.05, x1 = h2.sex[2, 3], code = 3, angle = 90, l
mtext("Narrow-sense heritability (\u00b1CI)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6
axis(2, at = 1, labels = c("birth weight"), las = 3, cex.axis = 1.6)

```

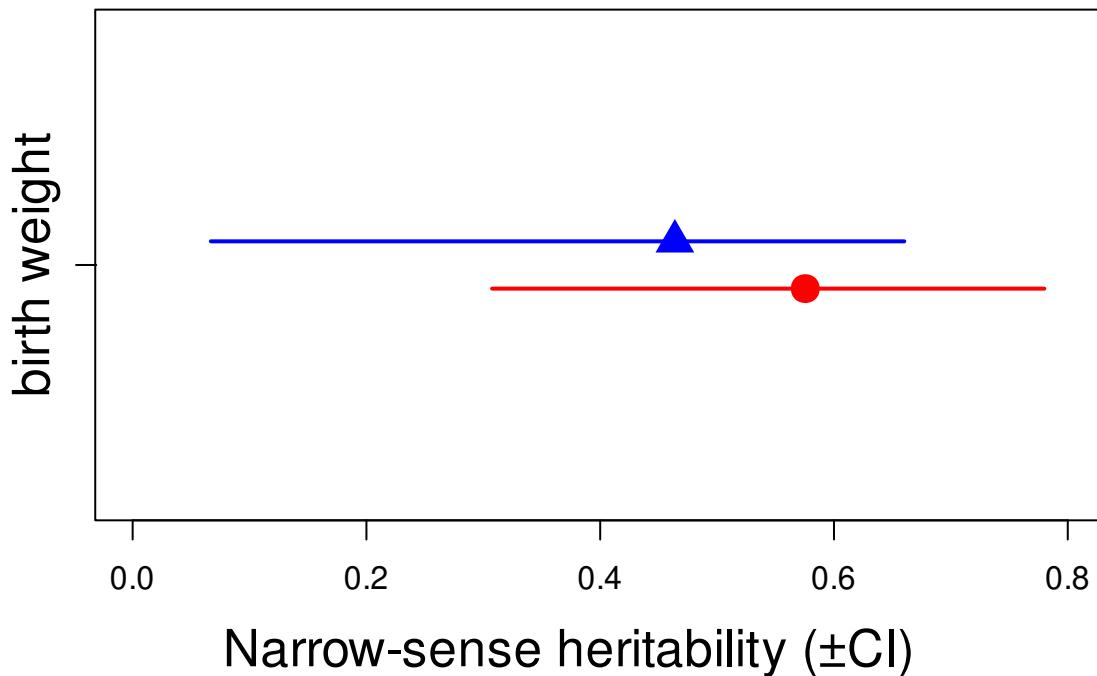


Figure 2.7: Female and male heritability of birth weight

2.5.6 Modification of model parameter

Unfortunately (to our knowledge), it is not possible to alter the variance matrices and refit them within the model.

2.5.7 Covariance between two random effects

Some research questions require to estimate the covariance between two random effects within a univariate model. Unfortunately (to our knowledge), it is not possible to create a covariance between distinct random effects (<https://github.com/paul-buerkner/brms/issues/502>). However, a multi-membership model can be fit using the linking.function `mm`, thus forcing the variance of two variables to be equal and the covariance to 1.

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Chapitre 3

Multivariate animal model

This tutorial will demonstrate how to run a multivariate animal model looking at birth weight and tarsus length of the phenomenal gryphons.

3.1 Scenario and data

3.1.1 Scenario

Since natural selection rarely acts on single traits, to understand how birth weight might evolve in our population of gryphons, we may also want to think about possible covariance with other traits. If tarsus length at fledging is also under positive selection, what implications does it have for birth weight and vice versa? If the two traits are positively genetically correlated then this will facilitate evolution of larger size (since response of one trait will induce a positively correlated response in the other). If there is negative genetic covariance then this could act as an evolutionary constraint.

Using multivariate models allows the estimation of parameters relating to each trait alone (*i.e.* V_A , h^2 , etc), but also yields estimates of covariance components between traits. These include the (additive) genetic covariance COV_A which is often rescaled to give the additive genetic correlation r_A . However, covariance can also arise through other random effects (*e.g.* maternal covariance) and these sources can also be explicitly modeled in a bivariate analysis.

3.1.2 gryphon files

gryphonpedigree and phenotypic data files are the same as those used in tutorial 1 (*i.e.* `gryphonped.csv` and `gryphon.csv` respectively).

Reading the data

```
gryphon <- read.csv("data/gryphon.csv")
gryphon$animal <- as.factor(gryphon$animal)
gryphon$mother <- as.factor(gryphon$mother)
gryphon$byear <- as.factor(gryphon$byear)
gryphon$sex <- as.factor(gryphon$sex)
```

```
gryphon$bwt <- as.numeric(gryphon$bwt)
gryphon$tarsus <- as.numeric(gryphon$tarsus)
```

Reading the pedigree

```
gryphonped <- read.csv("data/gryphonped.csv")
gryphonped$id <- as.factor(gryphonped$id)
gryphonped$father <- as.factor(gryphonped$father)
gryphonped$mother <- as.factor(gryphonped$mother)
```

3.2 Asreml-R

3.2.1 Running the model

First we need to load the `asreml` library:

```
library(asreml)
```

For running multivariate analyses in ASReml-R, the code is slightly more complex than for the univariate case. This is because ASReml-R allows us to make different assumptions about the way in which traits might be related. We need to explicitly specify a covariance structure with difference covariance functions `us()`, `idh()` or `corgh()` which for example would estimate an unconstrained (co)variance matrix, an identity matrix and a variance and correlation matrix respectively. We can also specify some starting values for the variance matrices. These can be very approximate *guesstimates* or not at all, but having reasonable starting values can help convergence. It is also possible to let the model running without specifying starting values. Finally, we have increased the default maximum number of iterations (`maxiter`) which can help to achieve convergence for more complicated models. Another way to increase the number of iteration will be to use the `update` function. Notes that if the `LogLik` is not stabilized after several iterations, it is good indication of the model require more iteration.

```
ainv <- ainverse(gryphonped)
modela <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait,
  random = ~ us(trait, init = c(1, 0.1, 1)):vm(animal, ainv),
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:16 2022
```

```

##          LogLik      Sigma2      DF     wall      cpu
## 1    -7108.741      1.0  1535 16:54:16      0.0
## 2    -5837.803      1.0  1535 16:54:16      0.0
## 3    -4437.495      1.0  1535 16:54:16      0.0
## 4    -3459.378      1.0  1535 16:54:16      0.0
## 5    -2914.034      1.0  1535 16:54:16      0.0
## 6    -2729.131      1.0  1535 16:54:16      0.0
## 7    -2684.659      1.0  1535 16:54:16      0.0
## 8    -2679.838      1.0  1535 16:54:16      0.0
## 9    -2679.742      1.0  1535 16:54:16      0.0
## 10   -2679.741      1.0  1535 16:54:16      0.0

```

```
modela <- update(modela)
```

```

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:16 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1    -2679.741      1.0  1535 16:54:16      0.0
## 2    -2679.741      1.0  1535 16:54:16      0.0

```

`modela` has fitted a bivariate model of `bwt` and `tarsus`, with the mean for each of the traits as a fixed effect (`trait`). The additive genetic variance-covariance matrix (**G**) is unstructured (`us`; i.e. all elements are free to vary) and the starting values for V_A for `bwt`, COV_A between `bwt` and `tarsus`, and V_A for `tarsus` are set to 1, 0.1 and 1, respectively. Similarly, the residual matrix is unstructured and uses the same starting values.

Note that the argument `na.action = na.method(x = "include", y = "include")` can be added to the model. In a bivariate model, it will help calculate the covariance between two traits with different missing information `NA` and so help imbalance phenotypage and save sample size. However, it is important to scale (`mean =0`, `var =1`) the two traits to correctly adjust the model(see Asreml-R manual for more information).

Let's have a look at the variance components, and notice that there are now seven (co)variance components reported in the table:

```
summary(modela)$varcomp
```

	component	std.error	z.ratio	bound
## trait:vm(animal, ainv)!trait_bwt:bwt	3.368397	0.6348307	5.305977	P
## trait:vm(animal, ainv)!trait_tarsus:bwt	2.459809	1.0732644	2.291895	P
## trait:vm(animal, ainv)!trait_tarsus:tarsus	12.345792	3.0744285	4.015638	P
## units:trait!R	1.000000	NA	NA	F
## units:trait!trait_bwt:bwt	3.849916	0.5200101	7.403541	P
## units:trait!trait_tarsus:bwt	3.313282	0.9129234	3.629310	P
## units:trait!trait_tarsus:tarsus	17.646432	2.6670380	6.616491	P
## trait:vm(animal, ainv)!trait_bwt:bwt	%ch			
	0			

```
## trait:vm(animal, ainv)!trait_tarsus:bwt      0
## trait:vm(animal, ainv)!trait_tarsus:tarsus   0
## units:trait!R                                0
## units:trait!trait_bwt:bwt                   0
## units:trait!trait_tarsus:bwt                 0
## units:trait!trait_tarsus:tarsus              0
```

The first three terms are related to the genetic matrix and, in order are $V_{A,bwt}$, COV_A , $V_{A,tarsus}$. Below is again a line where the `units:trait!R` component equals to 1, which again can be ignored. The final three terms relate to the residual matrix and correspond to $V_{R,bwt}$, COV_R , $V_{R,tarsus}$. Based on our quick and dirty check (`is z.ratio > 1.96?`) all components look to be statistically significant.

We can calculate the genetic correlation as $COV_A / \sqrt{V_{A,bwt} \cdot V_{A,tarsus}}$. Thus this model gives an estimate of $r_A = 0.38$. It is also possible to estimate the residual correlation $r_{res} = 0.4$.

Both correlations are distinct in nature. The genetic correlation reflects how much the traits are linked by genetic via polygenic effect or linkage disequilibrium, whereas the residual correlation reflects the environmental correlation or errors measurement correlation.

Although we can calculate this by hand, we can also use `vpredict()`, which also provides an (approximate) standard error:

```
vpredict(modela, r_A ~ V2 / sqrt(V1 * V3))
```

```
##           Estimate        SE
## r_A  0.3814436 0.1299759
```

```
vpredict(modela, r_res ~ V6 / sqrt(V5 * V7))
```

```
##           Estimate        SE
## r_res 0.4019799 0.08607104
```

Of course we can also calculate the heritability of `bwt` and `tarsus` from this model:

```
vpredict(modela, h2.bwt ~ V1 / (V1 + V5))
```

```
##           Estimate        SE
## h2.bwt 0.466646 0.07671533
```

```
vpredict(modela, h2.tarsus ~ V3 / (V3 + V7))
```

```
##           Estimate        SE
## h2.tarsus 0.4116331 0.09305863
```

3.2.2 Adding fixed and random effects

Fixed and random effects can be added just as for the univariate case. Given that our full model of bwt from tutorial 1 had sex as a fixed effect as well as birth year and mother as random effects, we could specify a bivariate formulation with the same complexity:

```
modelb <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:16 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1     -4672.301      1.0  1533 16:54:16      0.0
## 2     -4005.615      1.0  1533 16:54:16      0.0
## 3     -3271.483      1.0  1533 16:54:16      0.0 (1 restrained)
## 4     -2761.414      1.0  1533 16:54:16      0.0 (1 restrained)
## 5     -2481.357      1.0  1533 16:54:17      0.0
## 6     -2395.858      1.0  1533 16:54:17      0.0
## 7     -2381.050      1.0  1533 16:54:17      0.0
## 8     -2380.251      1.0  1533 16:54:17      0.0
## 9     -2380.246      1.0  1533 16:54:17      0.0
```

```
modelb <- update(modelb)

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:17 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1     -2380.246      1.0  1533 16:54:17      0.0
## 2     -2380.246      1.0  1533 16:54:17      0.0
```

Note that we have specified a covariance structure for each random effect and an estimate of the effect of sex on both birth weight and tarsus length.

There will now be thirteen (co)variance components reported after running the code:

```
summary(modelb)$varcomp
```

	component	std.error	z.ratio
## trait:byear!trait_bwt:bwt	0.9746385	0.2825727	3.4491602

```

## trait:byear!trait_tarsus:bwt          0.1624076 0.4185079 0.3880635
## trait:byear!trait_tarsus:tarsus      3.7383721 1.2065992 3.0982716
## trait:mother!trait_bwt:bwt          1.1445184 0.2302182 4.9714512
## trait:mother!trait_tarsus:bwt      -1.5567306 0.4051848 -3.8420260
## trait:mother!trait_tarsus:tarsus    4.8206132 1.3201300 3.6516202
## trait:vm(animal, ainv)!trait_bwt:bwt 1.9893546 0.4410246 4.5107569
## trait:vm(animal, ainv)!trait_tarsus:bwt 3.3170404 0.9032323 3.6724110
## trait:vm(animal, ainv)!trait_tarsus:tarsus 10.2294887 2.8077066 3.6433610
## units:trait!R                      1.0000000 NA NA
## units:trait!trait_bwt:bwt          1.8443110 0.3443178 5.3564203
## units:trait!trait_tarsus:bwt      4.0142841 0.7412540 5.4155308
## units:trait!trait_tarsus:tarsus   12.4845955 2.2893363 5.4533690
##
## bound %ch
## trait:byear!trait_bwt:bwt          P 0
## trait:byear!trait_tarsus:bwt      P 0
## trait:byear!trait_tarsus:tarsus    P 0
## trait:mother!trait_bwt:bwt        P 0
## trait:mother!trait_tarsus:bwt    P 0
## trait:mother!trait_tarsus:tarsus  P 0
## trait:vm(animal, ainv)!trait_bwt:bwt  P 0
## trait:vm(animal, ainv)!trait_tarsus:bwt  P 0
## trait:vm(animal, ainv)!trait_tarsus:tarsus  P 0
## units:trait!R                      F 0
## units:trait!trait_bwt:bwt          P 0
## units:trait!trait_tarsus:bwt      P 0
## units:trait!trait_tarsus:tarsus   P 0

```

we can estimate the different correlations using `vpredict`:

```
vpredict(modelb, r_byear ~ V2 / sqrt(V1 * V3))
```

```

##           Estimate       SE
## r_byear  0.08508312 0.2134209

```

```
vpredict(modelb, r_M ~ V5 / sqrt(V4 * V6))
```

```

##           Estimate       SE
## r_M     -0.6627518 0.2487963

```

```
vpredict(modelb, r_A ~ V8 / sqrt(V7 * V9))
```

```

##           Estimate       SE
## r_A     0.7353053 0.1094747

```

```
vpredict(modelb, r_res ~ V12 / sqrt(V11 * V13))
```

```
##           Estimate       SE
## r_res  0.8365729 0.07366762
```

Now we can look at the fixed effects parameters and assess their significance with a conditional Wald F-test:

```
summary(modelb, coef = TRUE)$coef.fi
wald.asreml(modelb, denDF = "default", ssType = "conditional")$Wald
```

```
##                               solution std.error z.ratio
## at(trait, tarsus):sex_1 0.0000000      NA      NA
## at(trait, tarsus):sex_2 -0.0684413 0.3823448 -0.1790041
## at(trait, bwt):sex_1    0.0000000      NA      NA
## at(trait, bwt):sex_2    1.9502053 0.1480467 13.1729086
## trait_bwt                 6.3844483 0.2328210 27.4221324
## trait_tarsus              20.5936436 0.5098944 40.3880569

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:17 2022
##          LogLik     Sigma2      DF    wall      cpu
## 1     -2380.246      1.0   1533 16:54:17      0.0
## 2     -2380.246      1.0   1533 16:54:17      0.0
## Calculating denominator DF

##
##                               Df denDF   F.inc   F.con Margin      Pr
## trait                      2  52.6 1396.00 1396.00      0.00000
## at(trait, bwt):sex         1 812.8  298.40  173.50      B 0.00000
## at(trait, tarsus):sex      1 747.9   0.03    0.03      B 0.85798
```

Note that it is possible to specify a fixed effect to a specific trait by adding the number of order within `cbind` inside the argument `at(trait,x)`. For example, here we apply the fixed effect `sex` only to the response variable `tarsus`.

```
modelb_2 <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait, 2):sex,
  random = ~ us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
```

```

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:17 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1    -4810.563      1.0  1534 16:54:18      0.1
## 2    -4129.799      1.0  1534 16:54:18      0.0
## 3    -3382.529      1.0  1534 16:54:18      0.0 (1 restrained)
## 4    -2864.076      1.0  1534 16:54:18      0.0
## 5    -2574.891      1.0  1534 16:54:18      0.0
## 6    -2478.879      1.0  1534 16:54:18      0.0
## 7    -2458.305      1.0  1534 16:54:18      0.0
## 8    -2456.425      1.0  1534 16:54:18      0.0
## 9    -2456.377      1.0  1534 16:54:18      0.0
## 10   -2456.376      1.0  1534 16:54:18      0.0

```

```

summary(modelb_2, coef = TRUE)$coef.fi
wald.asreml(modelb_2, denDF = "default", ssType = "conditional")$Wald

```

```

##                                solution std.error z.ratio
## at(trait, tarsus):sex_1  0.000000      NA      NA
## at(trait, tarsus):sex_2 -3.267042  0.2953279 -11.06242
## trait_bwt                 7.636226  0.2389515  31.95722
## trait_tarsus              22.703658 0.4827348  47.03133

```

Model fitted using the sigma parameterization.

ASReml 4.1.0 Tue Nov 29 16:54:18 2022

```

##          LogLik      Sigma2      DF     wall      cpu
## 1    -2456.376      1.0  1534 16:54:18      0.0
## 2    -2456.376      1.0  1534 16:54:18      0.0

```

Calculating denominator DF

```

##
##                               Df denDF  F.inc  F.con Margin          Pr
## trait                      2  50.7 1233.0 1233.0      0.00000e+00
## at(trait, tarsus):sex  1 522.9 122.4 122.4      B 1.02886e-25

```

3.2.3 Significance testing

Under the model above r_M is estimated as -0.66 and the `z.ratio` associated with the corresponding covariance (COV_M) is >2 (in absolute terms). We might therefore infer that there is evidence for a strong negative correlation between the traits with respect to the mother and that while maternal identity explains variance in both traits those mothers that tend to produce heavier offspring actually tend to produce offspring with shorter tarsus lengths.

To formally test if COV_M is significantly different from zero, we can compare the log-likelihood for this model:

```
modelb$loglik
```

```
## [1] -2380.246
```

to a model in which we specify that $COV_M=0$. Since this constraint reduces the number of parameters to be estimated by one, we can use a likelihood ratio test (LRT) with one degree of freedom. To run the constrained model, we modify the G structure defined for the `mother` random effect to diagonal (`diag`), which means we only estimate the variances (the diagonal of the matrix) but not the covariance (the covariance are fixed to 0):

```
modelc <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    diag(trait, init = c(1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
```

```
## Model fitted using the sigma parameterization.
```

```
## ASReml 4.1.0 Tue Nov 29 16:54:19 2022
```

	LogLik	Sigma2	DF	wall	cpu
## 1	-4677.820	1.0	1533	16:54:19	0.0
## 2	-4010.442	1.0	1533	16:54:19	0.0
## 3	-3275.409	1.0	1533	16:54:19	0.0
## 4	-2763.519	1.0	1533	16:54:19	0.0
## 5	-2483.732	1.0	1533	16:54:19	0.0
## 6	-2400.242	1.0	1533	16:54:19	0.0
## 7	-2386.663	1.0	1533	16:54:19	0.0
## 8	-2386.049	1.0	1533	16:54:19	0.0
## 9	-2386.045	1.0	1533	16:54:19	0.0

You can run `summary(modelc)$varcomp` to confirm this worked. We can now obtain the log-likelihood of this model and compare this to that of `modelb` using a likelihood ratio test:

```
modelc$loglik
```

```
## [1] -2386.045
```

We can see that the model log-likelihood is now -2386.05. And comparing the models using a likelihood ratio test:

```
2 * (modelb$loglik - modelc$loglik)
```

```
## [1] 11.59835
```

So our chi-square test statistic is $\chi^2_1 = 11.6$. The p-value that goes with this is obtained by:

```
1 - pchisq(2 * (modelb$loglik - modelc$loglik), 1)
```

```
## [1] 0.0006601037
```

We would therefore conclude that the maternal covariance is significantly different from zero.

We could apply the same procedure to show that the residual (environmental) covariance and the genetic covariance estimates are significantly greater than zero (*i.e.*, heavier individuals tend to have longer tarsus lengths). In contrast, we should find that the byear covariance between the two traits is non-significant.

```
modeld <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    diag(trait, init = c(1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
```

Model fitted using the sigma parameterization.

ASReml 4.1.0 Tue Nov 29 16:54:19 2022

	LogLik	Sigma2	DF	wall	cpu
## 1	-4672.708	1.0	1533	16:54:19	0.1
## 2	-4005.953	1.0	1533	16:54:19	0.0
## 3	-3271.737	1.0	1533	16:54:19	0.0 (1 restrained)
## 4	-2761.626	1.0	1533	16:54:19	0.0 (1 restrained)
## 5	-2481.649	1.0	1533	16:54:19	0.0
## 6	-2395.992	1.0	1533	16:54:19	0.0
## 7	-2381.136	1.0	1533	16:54:19	0.0
## 8	-2380.331	1.0	1533	16:54:19	0.0
## 9	-2380.326	1.0	1533	16:54:19	0.0

```
2 * (modelb$loglik - modeld$loglik)
```

```
## [1] 0.1600641
```

```
1 - pchisq(2 * (modelb$loglik - modeld$loglik), 1)
```

```
## [1] 0.6890975
```

3.2.4 Estimate directly the genetic correlation within the model

Within Asreml-r, different matrix structure can be specify such as `us`, `corg`, `diag`, etc (cf see the Asreml-r guide). Instead of the fitting an unstructured matrix with the argument `us` or a reduced model with no covariance with the argument `diag`, we can also directly estimate the genetic correlation between the `bwt` and `tarsus` with `corg`.

Here we decide to estimate directly the additive genetic correlation.

```
modele <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ corg(trait, init = c(0.1, 1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:20 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1     -4672.301      1.0  1533 16:54:20      0.1
## 2     -4003.183      1.0  1533 16:54:20      0.0
## 3     -3266.521      1.0  1533 16:54:20      0.0 (1 restrained)
## 4     -2757.188      1.0  1533 16:54:20      0.0 (1 restrained)
## 5     -2479.293      1.0  1533 16:54:20      0.0
## 6     -2395.477      1.0  1533 16:54:20      0.0
## 7     -2381.026      1.0  1533 16:54:20      0.0
## 8     -2380.251      1.0  1533 16:54:20      0.0
## 9     -2380.246      1.0  1533 16:54:20      0.0
```

```
modele <- update(modele)

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:20 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1     -2380.246      1.0  1533 16:54:20      0.0
## 2     -2380.246      1.0  1533 16:54:20      0.0
```

```
summary(modele)$varcomp

##                                         component std.error
## trait:byear!trait_bwt:bwt           0.9746386 0.2825728
## trait:byear!trait_tarsus:bwt        0.1624071 0.4185082
```

```

## trait:byear!trait_tarsus:tarsus           3.7383734 1.2066018
## trait:mother!trait_bwt:bwt              1.1445186 0.2302183
## trait:mother!trait_tarsus:bwt          -1.5567316 0.4051850
## trait:mother!trait_tarsus:tarsus        4.8206154 1.3201324
## trait:vm(animal, ainv)!trait!tarsus:!trait!bwt.cor 0.7353061 0.1094807
## trait:vm(animal, ainv)!trait_bwt       1.9893543 0.4410243
## trait:vm(animal, ainv)!trait_tarsus     10.2294850 2.8077055
## units:trait!R                          1.0000000 NA
## units:trait!trait_bwt:bwt             1.8443112 0.3443178
## units:trait!trait_tarsus:bwt          4.0142825 0.7412540
## units:trait!trait_tarsus:tarsus      12.4845977 2.2893355
##
##                                     z.ratio bound %ch
## trait:byear!trait_bwt:bwt            3.449159   P  0
## trait:byear!trait_tarsus:bwt         0.388062   P  0
## trait:byear!trait_tarsus:tarsus     3.098266   P  0
## trait:mother!trait_bwt:bwt          4.971450   P  0
## trait:mother!trait_tarsus:bwt       -3.842027  P  0
## trait:mother!trait_tarsus:tarsus    3.651615   P  0
## trait:vm(animal, ainv)!trait!tarsus:!trait!bwt.cor 6.716310   U  0
## trait:vm(animal, ainv)!trait_bwt     4.510758   P  0
## trait:vm(animal, ainv)!trait_tarsus  3.643361   P  0
## units:trait!R                      NA   F  0
## units:trait!trait_bwt:bwt          5.356422   P  0
## units:trait!trait_tarsus:bwt       5.415529   P  0
## units:trait!trait_tarsus:tarsus    5.453372   P  0

```

It is important to note that using `corgh` change the order of the estimate (co)variance/correlation. Thus, the initial values need to be reorder and all different calculation need to be adjust in consequence. It is also important to check the difference between the model with `us` and `corgh` to make sure any mistake are made.

```
summary(modelb)$loglik
```

```
## [1] -2380.246
```

```
summary(modele)$loglik
```

```
## [1] -2380.246
```

There two main advantages to use `corgh`: first, a direct estimation of correlation within the G matrix can avoid mistake in the `vpredict` calculation; second, it is possible to test if the correlation is significantly different than 0 (similar result as LRT with the covariance) but also to -1 and 1 which correspond of the correlation boundaries. The following code showed how to create a reduced model with the correlation close to 1 and compared to the initial model. Since we compared the correlation to its boundary, the degree of freedom is only half as a one tail LTR.

```

MODEL_MODIF <- update.asreml(modele, start.values = T)
G_MOD <- MODEL_MODIF$vparameters.table[(1:9), ]
G_MOD[1, 2] <- 0.99999
G_MOD[1, 3] <- "F"
modele.red <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ corgh(trait, init = c(0.1, 1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20,
  G.param = G_MOD
)

```

```

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:20 2022
##          LogLik      Sigma2       DF     wall      cpu
## 1    -2545.233      1.0  1533 16:54:20      0.0
## 2    -2483.883      1.0  1533 16:54:20      0.0
## 3    -2423.504      1.0  1533 16:54:21      0.0
## 4    -2392.509      1.0  1533 16:54:21      0.0
## 5    -2383.661      1.0  1533 16:54:21      0.0
## 6    -2383.084      1.0  1533 16:54:21      0.0
## 7    -2383.033      1.0  1533 16:54:21      0.0
## 8    -2383.022      1.0  1533 16:54:21      0.0
## 9    -2383.019      1.0  1533 16:54:21      0.1
## 10   -2383.019      1.0  1533 16:54:21      0.1

```

```

2 * (modele$loglik - modele.red$loglik)

## [1] 5.544679

1 - pchisq(2 * (modele$loglik - modele.red$loglik), df = 0.5)

## [1] 0.006598676

```

Here, the correlation is significantly different than 1 (~0.99999).

3.2.5 Visualisation of the correlation (aka BLUP extraction)

When estimating correlation between traits, having a visualization of it can help the interpretation. In addition, visualizing the correlation can spot outliers in the dataset. Thanks to mixed model, each breeding values is stored within the model and can be extract as BLUP (Best Linear Unbiased

Predictor).BLUP should be normally distributed, if not you need to check the assumption of your animal model.

To simplify the following code, we rename the variable T1 and T2.

```
gryphon$T1 <- gryphon$bwt
gryphon$T2 <- gryphon$tarsus
#####
modele <- asreml(
  fixed = cbind(T1, T2) ~ trait + at(trait):sex,
  random = ~ corgh(trait, init = c(0.1, 1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ id(units):us(trait, init = c(1, 0.1, 1)),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:21 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1     -4672.301      1.0  1533 16:54:21      0.1
## 2     -4003.183      1.0  1533 16:54:21      0.0
## 3     -3266.521      1.0  1533 16:54:21      0.0 (1 restrained)
## 4     -2757.188      1.0  1533 16:54:21      0.0 (1 restrained)
## 5     -2479.293      1.0  1533 16:54:21      0.0
## 6     -2395.477      1.0  1533 16:54:21      0.0
## 7     -2381.026      1.0  1533 16:54:21      0.0
## 8     -2380.251      1.0  1533 16:54:21      0.0
## 9     -2380.246      1.0  1533 16:54:22      0.0
```

```
modele <- update(modele)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:22 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1     -2380.246      1.0  1533 16:54:22      0.0
## 2     -2380.246      1.0  1533 16:54:22      0.0
```

```
summary(modele)$varcomp
```

	component	std.error	z.ratio
## trait:byear!trait_T1:T1	0.9746386	0.2825728	3.449159
## trait:byear!trait_T2:T1	0.1624071	0.4185082	0.388062
## trait:byear!trait_T2:T2	3.7383734	1.2066018	3.098266

```

## trait:mother!trait_T1:T1          1.1445186 0.2302183 4.971450
## trait:mother!trait_T2:T1          -1.5567316 0.4051850 -3.842027
## trait:mother!trait_T2:T2          4.8206154 1.3201324 3.651615
## trait:vm(animal, ainv)!trait!T2:!trait!T1.cor 0.7353061 0.1094807 6.716310
## trait:vm(animal, ainv)!trait_T1      1.9893543 0.4410243 4.510758
## trait:vm(animal, ainv)!trait_T2      10.2294850 2.8077055 3.643361
## units:trait!R                     1.0000000 NA NA
## units:trait!trait_T1:T1           1.8443112 0.3443178 5.356422
## units:trait!trait_T2:T1           4.0142825 0.7412540 5.415529
## units:trait!trait_T2:T2           12.4845977 2.2893355 5.453372
##                                         bound %ch
##                                         P 0
##                                         F 0
##                                         P 0
##                                         P 0
##                                         P 0

```

```

#####
DvsS <- data.frame(
  Trait = rownames(modele$coefficients$random),
  BLUP = modele$coefficients$random,
  SE = sqrt(modele$vcoeff$random * modele$sigma2)
)
DvsS$ID <- substr(DvsS$Trait, 27, 30)
DvsS$TRAIT <- substr(DvsS$Trait, 7, 8)
DvsS <- DvsS[927:3544, ] # keep only row associated to animal
summary(factor(DvsS$TRAIT)) # 1309 each

```

```

##   T1   T2
## 1309 1309

```

```
#
```

```

DvsS$Trait <- NULL
colnames(DvsS)[1] <- "BLUP"
BLUPS <- reshape(DvsS, v.names = c("BLUP", "SE"), idvar = "ID", timevar = "TRAIT", direction = "long")
nrow(BLUPS)

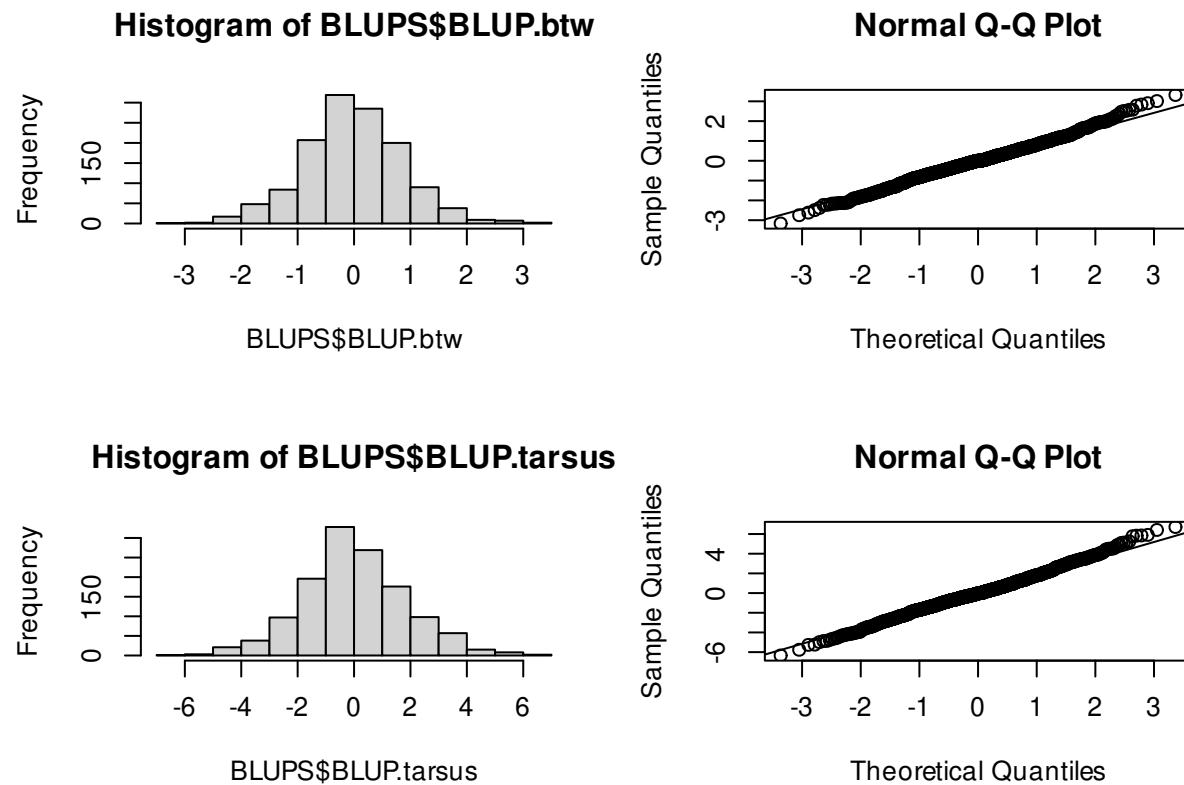
```

```
## [1] 1309
```

```
rownames(BLUPS) <- c()
colnames(BLUPS) <- c("ID", "BLUP.btw", "SE.btw", "BLUP.tarsus", "SE.tarsus")
summary(BLUPS)
```

```
##      ID          BLUP.btw        SE.btw       BLUP.tarsus
## Length:1309    Min.   :-3.165474   Min.   :0.7984   Min.   :-6.34104
## Class :character  1st Qu.:-0.559280  1st Qu.:0.9967  1st Qu.:-1.14428
## Mode  :character   Median :-0.001912   Median :1.0367  Median :-0.02641
##                   Mean   :-0.009008   Mean   :1.0933  Mean   : 0.02134
##                   3rd Qu.: 0.533972  3rd Qu.:1.2210  3rd Qu.: 1.18107
##                   Max.   : 3.319657  Max.   :1.4377  Max.   : 6.71502
##      SE.tarsus
##      Min.   :1.928
##      1st Qu.:2.371
##      Median :2.451
##      Mean   :2.576
##      3rd Qu.:2.811
##      Max.   :3.287
```

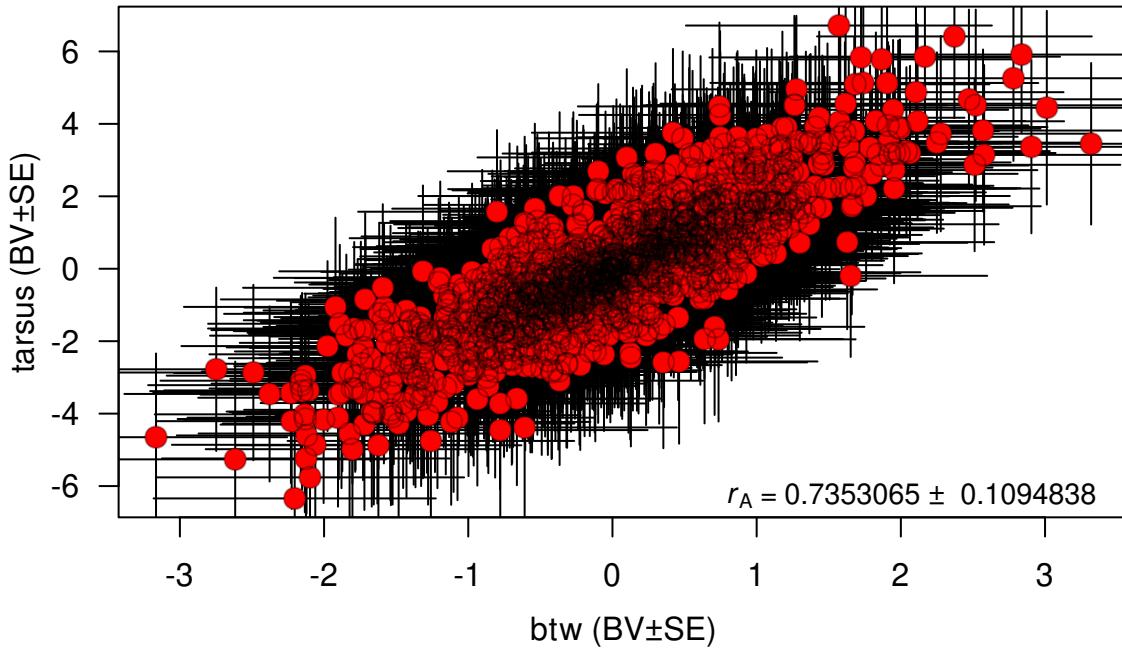
```
# write.csv(BLUPS,file="BLUPS_6x6.csv",row.names=F)
#####
par(mfrow = c(2, 2))
hist(BLUPS$BLUP.btw)
qqnorm(BLUPS$BLUP.btw)
qqline(BLUPS$BLUP.btw)
hist(BLUPS$BLUP.tarsus)
qqnorm(BLUPS$BLUP.tarsus)
qqline(BLUPS$BLUP.tarsus)
```



#

Here, some simple code to plot the genetic correlation.

```
plot(BLUP.tarsus ~ BLUP.btw, BLUPS, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = BLUPS$BLUP.btw, y0 = BLUPS$BLUP.tarsus - BLUPS$SE.tarsus, x1 = BLUPS$BLUP.btw,
arrows(x0 = BLUPS$BLUP.btw - BLUPS$SE.btw, y0 = BLUPS$BLUP.tarsus, x1 = BLUPS$BLUP.btw + BLUPS$SE.btw)
points(BLUP.tarsus ~ BLUP.btw, BLUPS, pch = 16, col = "red", cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, BLUPS, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("btw (BV±SE)", side = 1, line = 2.4)
mtext("tarsus (BV±SE)", side = 2, line = 2, las = 3)
mtext(expression(paste(italic(r)[A], " = 0.7353065 ± 0.1094838")), side = 1, line = -1, las = 1)
```



3.2.6 Partitionning (co)variance between groups

Similar to the univariate model, it is possible to partition the variance and also the covariance between different groups within the dataset. Here, we can estimate sex-specific genetic correlation. Note, to partition a correlation, it is require to have important sample size within each group. For this example, we simplify the model !

```

gryphon <- gryphon[order(gryphon$sex), ]
model_sex <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ at(sex):us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ dsum(~ id(units):us(trait) | sex),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)
## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:23 2022
##          LogLik      Sigma2       DF     wall      cpu
## 1     -2522.729        1.0   1807 16:54:23     0.1 (1 restrained)

```

```

## Warning in asreml(fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex, random =
## = ~at(sex):us(trait, : US updates modified 1 times in iteration 2 to remain
## positive definite.

## 2      -2459.512          1.0   1807 16:54:23    0.1 (3 restrained)
## 3      -2408.940          1.0   1807 16:54:23    0.1
## 4      -2392.691          1.0   1807 16:54:23    0.2
## 5      -2388.962          1.0   1807 16:54:23    0.2
## 6      -2388.743          1.0   1807 16:54:24    0.1
## 7      -2388.736          1.0   1807 16:54:24    0.1
## 8      -2388.736          1.0   1807 16:54:24    0.1

## Warning in asreml(fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex, random =
## ~at(sex):us(trait, : US variance structures were modified in 1 instances to make
## them positive definite

```

```
model_sex <- update(model_sex)
```

```

## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:24 2022
##           LogLik      Sigma2     DF    wall    cpu
## 1      -2388.736      1.0   1807 16:54:24    0.2
## 2      -2388.736      1.0   1807 16:54:24    0.1

```

```
summary(model_sex)$varcomp
```

	component	std.error
## trait:byear!trait_bwt:bwt	0.9858478	0.2863878
## trait:byear!trait_tarsus:bwt	0.1525063	0.4334263
## trait:byear!trait_tarsus:tarsus	3.9981983	1.2798747
## trait:mother!trait_bwt:bwt	1.3312734	0.2484444
## trait:mother!trait_tarsus:bwt	-1.6174228	0.4283851
## trait:mother!trait_tarsus:tarsus	4.7542338	1.3546517
## at(sex, 1):trait:vm(animal, ainv)!trait_bwt:bwt	1.3402853	0.5670773
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:bwt	2.3608392	1.1348473
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:tarsus	6.0625993	3.1304394
## at(sex, 2):trait:vm(animal, ainv)!trait_bwt:bwt	1.8645998	0.8888206
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:bwt	5.0954811	2.0684729
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:tarsus	14.9771870	6.4479787
## sex_1!R	1.0000000	NA
## sex_1!trait_bwt:bwt	2.3079850	0.5015651
## sex_1!trait_tarsus:bwt	4.4287898	1.0376370
## sex_1!trait_tarsus:tarsus	13.4857819	2.9284922
## sex_2!R	1.0000000	NA
## sex_2!trait_bwt:bwt	1.7956612	0.7549779
## sex_2!trait_tarsus:bwt	2.6340448	1.7685804
## sex_2!trait_tarsus:tarsus	9.6094528	5.4917853

```

##                                     z.ratio bound %ch
## trait:byear!trait_bwt:bwt          3.4423530   P  0
## trait:byear!trait_tarsus:bwt      0.3518622   P  0
## trait:byear!trait_tarsus:tarsus   3.1238982   P  0
## trait:mother!trait_bwt:bwt       5.3584371   P  0
## trait:mother!trait_tarsus:bwt     -3.7756279  P  0
## trait:mother!trait_tarsus:tarsus  3.5095618   P  0
## at(sex, 1):trait:vm(animal, ainv)!trait_bwt:bwt  2.3634965  P  0
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:bwt 2.0803144  P  0
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:tarsus 1.9366608  P  0
## at(sex, 2):trait:vm(animal, ainv)!trait_bwt:bwt      2.0978361  P  0
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:bwt     2.4634024  P  0
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:tarsus  2.3227724  P  0
## sex_1!R                               NA   F  0
## sex_1!trait_bwt:bwt                  4.6015657   P  0
## sex_1!trait_tarsus:bwt               4.2681493   P  0
## sex_1!trait_tarsus:tarsus           4.6050257   P  0
## sex_2!R                               NA   F  0
## sex_2!trait_bwt:bwt                  2.3784288   P  0
## sex_2!trait_tarsus:bwt               1.4893554   P  0
## sex_2!trait_tarsus:tarsus           1.7497867   P  0

```

we can estimate the different correlations using `vpredict`:

```
vpredict(model_sex, r_byear ~ V2 / sqrt(V1 * V3))
```

```

##           Estimate      SE
## r_byear  0.07681584 0.213141

```

```
vpredict(model_sex, r_M ~ V5 / sqrt(V4 * V6))
```

```

##           Estimate      SE
## r_M      -0.6429092 0.248944

```

```
vpredict(model_sex, r_A.1 ~ V8 / sqrt(V7 * V9))
```

```

##           Estimate      SE
## r_A.1    0.8282059 0.1723596

```

```
vpredict(model_sex, r_A.2 ~ V11 / sqrt(V10 * V12))
```

```

##           Estimate      SE
## r_A.2    0.9642225 0.1241668

```

```
vpredict(model_sex, r_res.1 ~ V15 / sqrt(V14 * V16))
```

```
##           Estimate      SE
## r_res.1 0.7938355 0.07892634
```

```
vpredict(model_sex, r_res.2 ~ V19 / sqrt(V18 * V20))
```

```
##           Estimate      SE
## r_res.2 0.6341057 0.1894837
```

and the heritability too:

```
vpredict(model_sex, h2.bwt.1 ~ V7 / (V1 + V4 + V7 + V14))
```

```
##           Estimate      SE
## h2.bwt.1 0.2246768 0.09176827
```

```
vpredict(model_sex, h2.bwt.2 ~ V10 / (V1 + V4 + V10 + V18))
```

```
##           Estimate      SE
## h2.bwt.2 0.3119425 0.1442547
```

```
vpredict(model_sex, h2.tarsus.1 ~ V9 / (V3 + V6 + V9 + V16))
```

```
##           Estimate      SE
## h2.tarsus.1 0.21422 0.1070464
```

```
vpredict(model_sex, h2.tarsus.2 ~ V12 / (V3 + V6 + V12 + V20))
```

```
##           Estimate      SE
## h2.tarsus.2 0.4492383 0.1833858
```

Now we can look at the fixed effects parameters and assess their significance with a conditional Wald F-test:

```
summary(model_sex, coef = TRUE)$coef.fi
wald.asreml(model_sex, denDF = "default", ssType = "conditional")$Wald
```

	solution	std error	z.ratio
## at(trait, tarsus):sex_1	0.0000000	NA	NA
## at(trait, tarsus):sex_2	-0.0554799	0.4758708	-0.1165861
## at(trait, bwt):sex_1	0.0000000	NA	NA
## at(trait, bwt):sex_2	1.9393688	0.1903239	10.1898321

```

## trait_bwt           6.3779149 0.2311766 27.5889321
## trait_tarsus        20.5838787 0.4942649 41.6454395

## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:25 2022
##          LogLik      Sigma2      DF     wall      cpu
## 1      -2388.736      1.0    1807 16:54:25      0.2
## 2      -2388.736      1.0    1807 16:54:25      0.1
## Calculating denominator DF

##
##          Df denDF   F.inc   F.con Margin      Pr
## trait            2  44.8 1522.00 1522.00      0.00000
## at(trait, bwt):sex 1 137.5 220.90 103.80      B 0.00000
## at(trait, tarsus):sex 1 138.6  0.01   0.01      B 0.90737

```

To assess the significant of the covariance, a LTR test can be done with a reduced model where a specific covariance can be fixed to 0 (for example the female covariance, following code).

```

model_modif <- update.asreml(model_sex, start.values = T)
G <- model_modif$vparameters[(1:12), ]
G$Constraint[(2)] <- "F"
G$Value[(2)] <- 0
#
reduc.model_sex <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ at(sex):us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ dsum(~ id(units):us(trait) | sex),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20,
  G.param = G
)

## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:26 2022

## Warning in asreml(fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex, random
## = ~at(sex):us(trait, : US updates modified 1 times in iteration 1 to remain
## positive definite.

##          LogLik      Sigma2      DF     wall      cpu
## 1      -2474.972      1.0    1807 16:54:26      0.1 (3 restrained)
## 2      -2406.283      1.0    1807 16:54:26      0.1
## 3      -2394.010      1.0    1807 16:54:27      0.1
## 4      -2391.718      1.0    1807 16:54:27      0.1
## 5      -2391.480      1.0    1807 16:54:27      0.1

```

```

## 6      -2391.477          1.0   1807 16:54:27    0.1

## Warning in asreml(fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex, random =
## ~at(sex):us(trait, : US variance structures were modified in 1 instances to make
## them positive definite

reduc.model_sex <- update(reduc.model_sex)

## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:27 2022
##           LogLik     Sigma2      DF     wall      cpu
## 1      -2391.476      1.0   1807 16:54:27    0.2
## 2      -2391.476      1.0   1807 16:54:27    0.1

summary(reduc.model_sex)$varcomp

##                                         component std.error
## trait:byear!trait_bwt:bwt            0.9794331 0.2848997
## trait:byear!trait_tarsus:bwt         0.1428995 0.4322719
## trait:byear!trait_tarsus:tarsus      4.0021595 1.2818624
## trait:mother!trait_bwt:bwt          1.4956509 0.2568074
## trait:mother!trait_tarsus:bwt        -1.2460057 0.4438357
## trait:mother!trait_tarsus:tarsus     5.3945609 1.4035705
## at(sex, 1):trait:vm(animal, ainv)!trait_bwt:bwt 0.5265716 0.3579555
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:bwt 0.0000000 NA
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:tarsus 1.4223969 1.9103795
## at(sex, 2):trait:vm(animal, ainv)!trait_bwt:bwt      1.5835813 0.8671365
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:bwt     4.4288714 2.0173971
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:tarsus 12.9349047 6.2946996
## sex_1!R                                1.0000000 NA
## sex_1!trait_bwt:bwt                     2.9539767 0.4196755
## sex_1!trait_tarsus:bwt                  6.3138301 0.6802598
## sex_1!trait_tarsus:tarsus                17.3577089 2.4730547
## sex_2!R                                1.0000000 NA
## sex_2!trait_bwt:bwt                     1.9341439 0.7416691
## sex_2!trait_tarsus:bwt                  2.9467290 1.7370018
## sex_2!trait_tarsus:tarsus                10.7245912 5.4025888
##                                         z.ratio bound %ch
## trait:byear!trait_bwt:bwt            3.4378175 P  0
## trait:byear!trait_tarsus:bwt         0.3305778 P  0
## trait:byear!trait_tarsus:tarsus      3.1221444 P  0
## trait:mother!trait_bwt:bwt          5.8240170 P  0
## trait:mother!trait_tarsus:bwt        -2.8073580 P  0
## trait:mother!trait_tarsus:tarsus     3.8434556 P  0
## at(sex, 1):trait:vm(animal, ainv)!trait_bwt:bwt 1.4710530 P  0
## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:bwt NA   F  NA

```

```

## at(sex, 1):trait:vm(animal, ainv)!trait_tarsus:tarsus  0.7445625    P  0
## at(sex, 2):trait:vm(animal, ainv)!trait_bwt:bwt      1.8262193    P  0
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:bwt      2.1953395    P  0
## at(sex, 2):trait:vm(animal, ainv)!trait_tarsus:tarsus  2.0548883    P  0
## sex_1!R                                         NA  F  0
## sex_1!trait_bwt:bwt                         7.0387165    P  0
## sex_1!trait_tarsus:bwt                      9.2814981    P  0
## sex_1!trait_tarsus:tarsus                  7.0187323    P  0
## sex_2!R                                         NA  F  0
## sex_2!trait_bwt:bwt                         2.6078261    P  0
## sex_2!trait_tarsus:bwt                      1.6964455    P  0
## sex_2!trait_tarsus:tarsus                  1.9850837    P  0

```

```
2 * (model_sex$loglik - reduc.model_sex$loglik)
```

```
## [1] 5.481033
```

```
1 - pchisq(2 * (model_sex$loglik - reduc.model_sex$loglik), df = 1)
```

```
## [1] 0.0192239
```

In addition, it is also possible to test the sexes if sexes has significant differences with another reduced model where both covariance are fixed to their average values.

```

# code provided as an example for the moment since the model cannot run on this data
model_modif <- update.asreml(model_sex, start.values = T)
G <- model_modif$vparameters[(1:12), ]
G$fac <- factor(
  c(
    1, 2, 3, 4, 2, 6, # Additive genetic matrix 2 =5
    7, 8, 9, # byear matrix
    10, 11, 12 # mother matrix
  )
)
Modif <- vcm.lm(~fac, data = G)
attr(Modif, "assign") <- NULL
attr(Modif, "contrasts") <- NULL
#
reduc.model_sex_2 <- asreml(
  fixed = cbind(bwt, tarsus) ~ trait + at(trait):sex,
  random = ~ at(sex):us(trait, init = c(1, 0.1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ dsum(~ id(units):us(trait) | sex),
  data = gryphon,
```

```

na.action = na.method(x = "include", y = "include"),
maxit = 20,
G.param = G, vcm = Modif
)
reduc.model_sex_2 <- update(reduc.model_sex_2)
summary(reduc.model_sex_2)$varcomp

2 * (model_sex$loglik - reduc.model_sex_2$loglik)
1 - pchisq(2 * (model_sex$loglik - reduc.model_sex_2$loglik), df = 2)

```

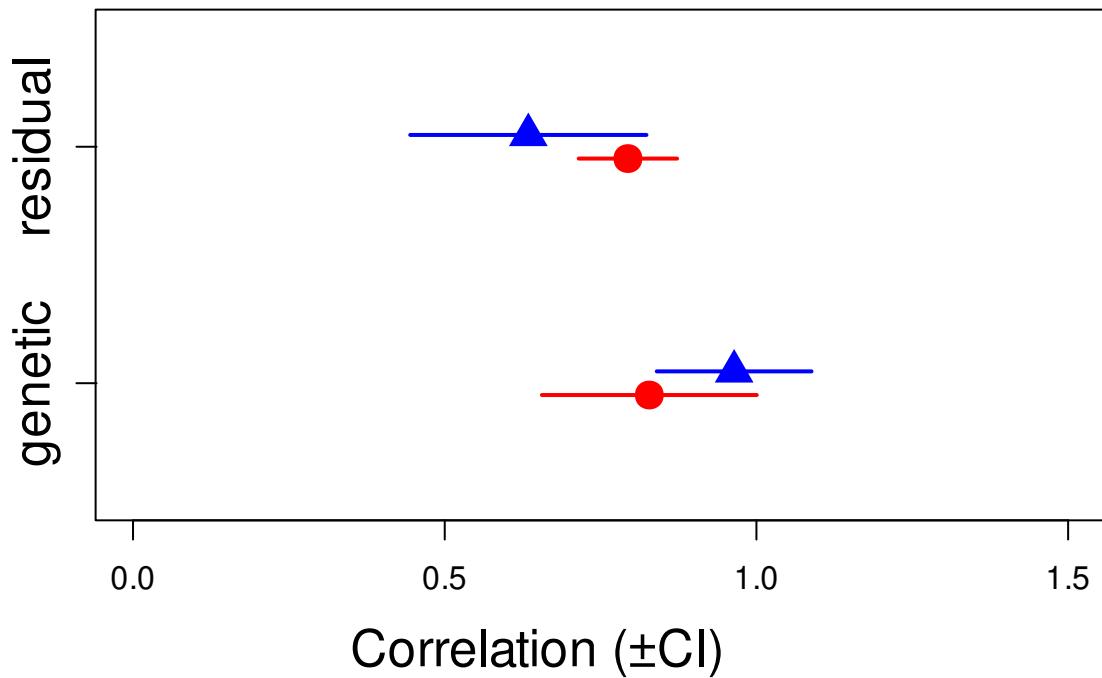
Here a plot to visualize the overlaps of covariances.

```

genetic.correlation.F <- vpredict(model_sex, r_A.1 ~ V8 / sqrt(V7 * V9))
genetic.correlation.M <- vpredict(model_sex, r_A.2 ~ V11 / sqrt(V10 * V12))
residual.correlation.F <- vpredict(model_sex, r_res.1 ~ V15 / sqrt(V14 * V16))
residual.correlation.M <- vpredict(model_sex, r_res.2 ~ V19 / sqrt(V18 * V20))
cor.est <- rbind(genetic.correlation.F, genetic.correlation.M, residual.correlation.F, res

plot(c(0.95, 1.05, 1.95, 2.05) ~ cor.est[, 1], xlim = c(0, 1.5), ylim = c(0.5, 2.5), xlab =
arrows(y0 = 0.95, x0 = cor.est[1, 1] - cor.est[1, 2], y1 = 0.95, x1 = cor.est[1, 1] + cor.
arrows(y0 = 1.05, x0 = cor.est[2, 1] - cor.est[2, 2], y1 = 1.05, x1 = cor.est[2, 1] + cor.
arrows(y0 = 1.95, x0 = cor.est[3, 1] - cor.est[3, 2], y1 = 1.95, x1 = cor.est[3, 1] + cor.
arrows(y0 = 2.05, x0 = cor.est[4, 1] - cor.est[4, 2], y1 = 2.05, x1 = cor.est[4, 1] + cor.
mtext("Correlation (\u00b1CI)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6)
axis(2, at = 1, labels = c("genetic"), las = 3, cex.axis = 1.6)
axis(2, at = 2, labels = c("residual"), las = 3, cex.axis = 1.6)

```



By using `corgf`, we can extract the BLUPs and plot the sex-specific correlation.

```

gryphon$T1 <- gryphon$bwt
gryphon$T2 <- gryphon$tarsus
#####
model_sex <- asreml(
  fixed = cbind(T1, T2) ~ trait + at(trait):sex,
  random = ~ at(sex):corgf(trait, init = c(0.1, 1, 1)):vm(animal, ainv) +
    us(trait, init = c(1, 0.1, 1)):byear +
    us(trait, init = c(1, 0.1, 1)):mother,
  residual = ~ dsum(~ id(units):us(trait) | sex),
  data = gryphon,
  na.action = na.method(x = "include", y = "include"),
  maxit = 20
)

```

```

## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:27 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1     -2522.729      1.0  1807 16:54:28      0.1 (2 restrained)
## 2     -2457.755      1.0  1807 16:54:28      0.1 (2 restrained)
## 3     -2407.462      1.0  1807 16:54:28      0.1 (2 restrained)
## 4     -2394.143      1.0  1807 16:54:28      0.1 (1 restrained)

```

```
## 5 -2389.368 1.0 1807 16:54:28 0.1
## 6 -2388.741 1.0 1807 16:54:28 0.1
## 7 -2388.736 1.0 1807 16:54:28 0.1
```

```
model_sex <- update(model_sex)
```

```
## Multi-section model using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:54:28 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1     -2388.736      1.0 1807 16:54:28 0.1
## 2     -2388.736      1.0 1807 16:54:29 0.1
```

```
DvsS <- data.frame(
  Trait = rownames(model_sex$coefficients$random),
  BLUP = model_sex$coefficients$random,
  SE = sqrt(model_sex$vcoeff$random * model_sex$sigma2)
)
DvsS$ID <- substr(DvsS$Trait, 38, 40)
DvsS$TRAIT <- substr(DvsS$Trait, 18, 19)
DvsS$SEX <- substr(DvsS$Trait, 9, 9)
DvsS <- DvsS[927:6162, ] # keep only row associated to animal
summary(factor(DvsS$TRAIT)) # 1309 each
```

```
##   T1   T2
## 2618 2618
```

```
#  
DvsS$Trait <- NULL  
colnames(DvsS)[1] <- "BLUP"  
BLUPS <- reshape(DvsS, v.names = c("BLUP", "SE"), idvar = c("ID", "SEX"), timevar = "TRAIT")
```

```
## Warning in reshapeWide(data, idvar = idvar, timevar = timevar, varying =
## varying, : multiple rows match for TRAIT=T1: first taken
## Warning in reshapeWide(data, idvar = idvar, timevar = timevar, varying =
## varying, : multiple rows match for TRAIT=T2: first taken
```

```
nrow(BLUPS)
```

```
## [1] 1998
```

```
rownames(BLUPS) <- c()  
colnames(BLUPS) <- c("ID", "SEX", "BLUP.btw", "SE.btw", "BLUP.tarsus", "SE.tarsus")  
summary(BLUPS)
```

```

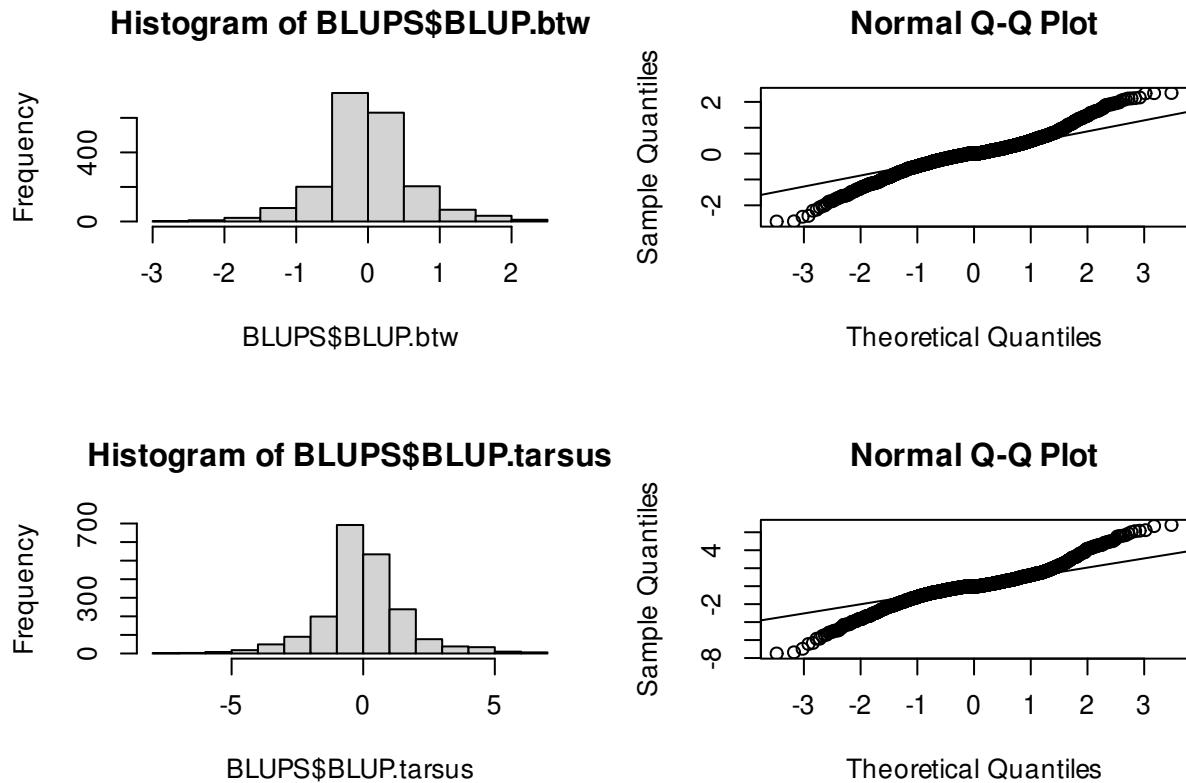
##      ID          SEX        BLUP.btw       SE.btw
## Length:1998  Length:1998   Min.  :-2.625088  Min.  :0.8383
## Class  :character  Class  :character  1st Qu.:-0.281979  1st Qu.:0.9412
## Mode   :character  Mode   :character  Median  :0.000000  Median  :1.1018
##                               Mean   :0.008687  Mean   :1.0931
##                               3rd Qu.: 0.293571  3rd Qu.:1.1777
##                               Max.   : 2.344440  Max.   :1.3650
##      BLUP.tarsus    SE.tarsus
## Min.  :-7.49444  Min.   :1.829
## 1st Qu.:-0.66206 1st Qu.:2.345
## Median : 0.00000  Median :2.462
## Mean   : 0.02557  Mean   :2.730
## 3rd Qu.: 0.71508  3rd Qu.:3.326
## Max.   : 6.80950  Max.   :3.868

```

```

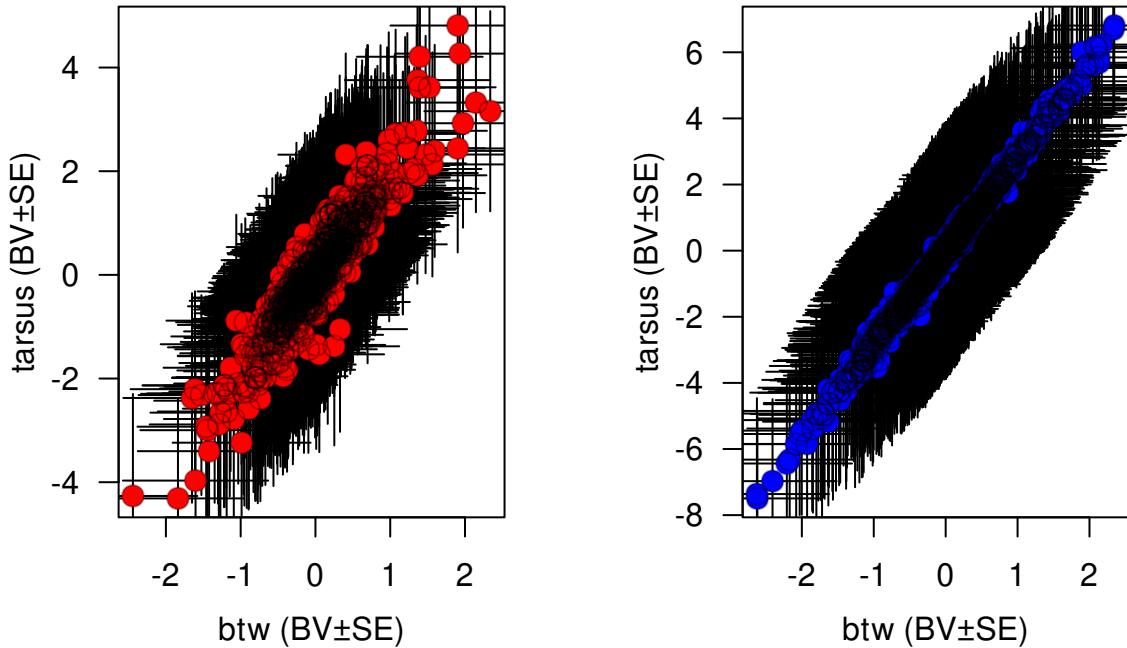
# write.csv(BLUPS,file="BLUPS_6x6_SEX.csv",row.names=F)
#####
par(mfrow = c(2, 2))
hist(BLUPS$BLUP.btw)
qqnorm(BLUPS$BLUP.btw)
qqline(BLUPS$BLUP.btw)
hist(BLUPS$BLUP.tarsus)
qqnorm(BLUPS$BLUP.tarsus)
qqline(BLUPS$BLUP.tarsus)

```



Here, some simple codes to plot the genetic correlation.

```
FEM <- subset(BLUPS, SEX == "1")
MAL <- subset(BLUPS, SEX == "2")
#
par(mfrow = c(1, 2))
#
plot(BLUP.tarsus ~ BLUP.btw, FEM, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = FEM$BLUP.btw, y0 = FEM$BLUP.tarsus - FEM$SE.tarsus, x1 = FEM$BLUP.btw, y1 = FEM$BLUP.tarsus + FEM$SE.tarsus)
arrows(x0 = FEM$BLUP.btw - FEM$SE.btw, y0 = FEM$BLUP.tarsus, x1 = FEM$BLUP.btw + FEM$SE.btw, y1 = FEM$BLUP.tarsus)
points(BLUP.tarsus ~ BLUP.btw, FEM, pch = 16, col = "red", cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, FEM, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("btw (BV±SE)", side = 1, line = 2.4)
mtext("tarsus (BV±SE)", side = 2, line = 2, las = 3)
#
plot(BLUP.tarsus ~ BLUP.btw, MAL, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = MAL$BLUP.btw, y0 = MAL$BLUP.tarsus - MAL$SE.tarsus, x1 = MAL$BLUP.btw, y1 = MAL$BLUP.tarsus + MAL$SE.tarsus)
arrows(x0 = MAL$BLUP.btw - MAL$SE.btw, y0 = MAL$BLUP.tarsus, x1 = MAL$BLUP.btw + MAL$SE.btw, y1 = MAL$BLUP.tarsus)
points(BLUP.tarsus ~ BLUP.btw, MAL, pch = 16, col = "blue", cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, MAL, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("btw (BV±SE)", side = 1, line = 2.4)
mtext("tarsus (BV±SE)", side = 2, line = 2, las = 3)
```



3.2.7 Between groups (co)variances and the B-matrix

Animal models are amazing model. With different group within a population, it is also possible to estimate how much the different groups shared the same genetic via the cross-group genetic covariance. This covariance is essential to understand ontogenetic or sexual conflict, which can constraint or enhanced response to evolution. As an example, we estimate the cross-sex genetic correlation r_{fm}

First, we need to dissociate the trait values for females and males into distinct variables. Then, we use a bivariate model (for one trait: `tarsus`) and a multivariate model (for various traits: `tarsus` and `bwt`). With a multivariate model, the cross-sex-cross trait covariance matrix is also named B matrix.

The coding is a bit complex but pretty straightforward. It is important to modify the covariance matrix at the residual level to avoid the calculation of a cross-sex residual covariance (no individual switched sex during the experiment).

```
gryphon$bwt.1 <- NA
gryphon$tarsus.1 <- NA
animal <- gryphon[gryphon$sex == "1", ]$animal
for (i in unique(animal)) {
  gryphon$bwt.1[which(gryphon$animal == i)] <- gryphon$bwt[which(gryphon$animal == i)]
  gryphon$tarsus.1[which(gryphon$animal == i)] <- gryphon$tarsus[which(gryphon$animal == i)]
```

```

}

#
gryphon$bwt.2 <- NA
gryphon$tarsus.2 <- NA
animal <- gryphon[gryphon$sex == "2", ]$animal
for (i in unique(animal)) {
  gryphon$bwt.2[which(gryphon$animal == i)] <- gryphon$bwt[which(gryphon$animal == i)]
  gryphon$tarsus.2[which(gryphon$animal == i)] <- gryphon$tarsus[which(gryphon$animal == i)]
}

#####
temp <- asreml(cbind(tarsus.1, tarsus.2) ~ trait,
  random = ~ us(trait):vm(animal, ainv) +
    diag(trait):byear + diag(trait):mother,
  residual = ~ units:us(trait),
  data = gryphon, na.action = na.method(y = "include", x = "include"), maxiter = 20,
  start.values = T
)
G <- temp$vparameters[(1:7), ]
R <- temp$vparameters[-(1:7), ]
#
G$Constraint <- "U"
R$Value[3] <- 0
R$Constraint[3] <- "F"
#
model.BiV_Sex <- asreml(cbind(tarsus.1, tarsus.2) ~ trait,
  random = ~ us(trait):vm(animal, ainv) +
    diag(trait):byear + diag(trait):mother,
  residual = ~ units:us(trait),
  data = gryphon, na.action = na.method(y = "include", x = "include"), maxiter = 20,
  G.param = G, R.param = R
)

```

Model fitted using the sigma parameterization.

ASReml 4.1.0 Tue Nov 29 16:54:30 2022

	LogLik	Sigma2	DF	wall	cpu
## 1	-1494.807	1.0	681	16:54:30	0.0 (1 restrained)
## 2	-1484.793	1.0	681	16:54:30	0.0 (1 restrained)
## 3	-1475.726	1.0	681	16:54:30	0.0 (1 restrained)
## 4	-1471.905	1.0	681	16:54:30	0.0 (1 restrained)
## 5	-1470.716	1.0	681	16:54:30	0.0
## 6	-1468.154	1.0	681	16:54:30	0.0
## 7	-1467.969	1.0	681	16:54:30	0.0
## 8	-1467.967	1.0	681	16:54:30	0.0

```
model.BiV_Sex <- update.asreml(model.BiV_Sex)
```

```
## Model fitted using the sigma parameterization.
```

```
## ASReml 4.1.0 Tue Nov 29 16:54:30 2022
```

	LogLik	Sigma2	DF	wall	cpu
## 1	-1467.967	1.0	681	16:54:30	0.0
## 2	-1467.967	1.0	681	16:54:30	0.0

```
#
```

```
summary(model.BiV_Sex)$varcomp
```

	component	std.error	z.ratio
## trait:byear!trait_tarsus.1	3.280319	1.532909	2.1399299
## trait:byear!trait_tarsus.2	4.743134	1.891252	2.5079332
## trait:mother!trait_tarsus.1	1.875132	2.424092	0.7735398
## trait:mother!trait_tarsus.2	4.314158	2.785254	1.5489283
## trait:vm(animal, ainv)!trait_tarsus.1:tarsus.1	6.582654	3.636467	1.8101781
## trait:vm(animal, ainv)!trait_tarsus.2:tarsus.1	8.396245	3.278591	2.5609306
## trait:vm(animal, ainv)!trait_tarsus.2:tarsus.2	12.898424	8.038362	1.6046084
## units:trait!R	1.000000	NA	NA
## units:trait!trait_tarsus.1:tarsus.1	14.872757	3.637545	4.0886803
## units:trait!trait_tarsus.2:tarsus.1	0.000000	NA	NA
## units:trait!trait_tarsus.2:tarsus.2	10.760849	6.294585	1.7095406
##	bound	%ch	
## trait:byear!trait_tarsus.1	U	0	
## trait:byear!trait_tarsus.2	U	0	
## trait:mother!trait_tarsus.1	U	0	
## trait:mother!trait_tarsus.2	U	0	
## trait:vm(animal, ainv)!trait_tarsus.1:tarsus.1	U	0	
## trait:vm(animal, ainv)!trait_tarsus.2:tarsus.1	U	0	
## trait:vm(animal, ainv)!trait_tarsus.2:tarsus.2	U	0	
## units:trait!R	F	0	
## units:trait!trait_tarsus.1:tarsus.1	P	0	
## units:trait!trait_tarsus.2:tarsus.1	F	NA	
## units:trait!trait_tarsus.2:tarsus.2	P	0	

The cross-sex genetic correlation can estimate form the output of the model. For tarsus length at fledging, sexes shared a lot of genetic variance which is commun for a trait with low sexual dimorphism. If the selection is antagonistic between males and females, sexes can not evolve freely form the other sexes and a sexual conflict appears.

```
vpredict(model.BiV_Sex, r_fm ~ V6 / sqrt(V5 * V7))
```

	Estimate	SE
## r_fm	0.9112054	0.4229764

We can estimate directly the correlation and plot the cross-sex genetic correlation

```

temp <- asreml(cbind(tarsus.1, tarsus.2) ~ trait,
  random = ~ corgh(trait):vm(animal, ainv) +
    diag(trait):byear + diag(trait):mother,
  residual = ~ units:corgh(trait),
  data = gryphon, na.action = na.method(y = "include", x = "include"), maxiter = 20,
  start.values = T
)
G <- temp$vparameters[(1:7), ]
R <- temp$vparameters[-(1:7), ]
#
G$Constraint <- "U"
R$Value[2] <- 0
R$Constraint[2] <- "F"
#
model.BiV_Sex <- asreml(cbind(tarsus.1, tarsus.2) ~ trait,
  random = ~ corgh(trait):vm(animal, ainv) +
    diag(trait):byear + diag(trait):mother,
  residual = ~ units:corgh(trait),
  data = gryphon, na.action = na.method(y = "include", x = "include"), maxiter = 20,
  G.param = G, R.param = R
)

```

Model fitted using the sigma parameterization.

ASReml 4.1.0 Tue Nov 29 16:54:30 2022

	LogLik	Sigma2	DF	wall	cpu
## 1	-1494.323	1.0	681	16:54:31	0.1 (1 restrained)
## 2	-1482.996	1.0	681	16:54:31	0.0 (1 restrained)
## 3	-1472.827	1.0	681	16:54:31	0.0 (1 restrained)
## 4	-1468.707	1.0	681	16:54:31	0.0
## 5	-1467.984	1.0	681	16:54:31	0.0
## 6	-1467.968	1.0	681	16:54:31	0.0
## 7	-1467.967	1.0	681	16:54:31	0.0

```
model.BiV_Sex <- update.asreml(model.BiV_Sex)
```

Model fitted using the sigma parameterization.

ASReml 4.1.0 Tue Nov 29 16:54:31 2022

	LogLik	Sigma2	DF	wall	cpu
## 1	-1467.967	1.0	681	16:54:31	0.1
## 2	-1467.967	1.0	681	16:54:31	0.0

```
#
summary(model.BiV_Sex)$varcomp
```

```

##                                         component std.error
## trait:byear!trait_tarsus.1          3.2803263 1.5329224
## trait:byear!trait_tarsus.2          4.7431679 1.8913244
## trait:mother!trait_tarsus.1         1.8751274 2.4240942
## trait:mother!trait_tarsus.2         4.3141262 2.7852550
## trait:vm(animal, ainv)!trait!tarsus.2:!trait!tarsus.1.cor 0.9111864 0.4230261
## trait:vm(animal, ainv)!trait_tarsus.1           6.5826478 3.6364929
## trait:vm(animal, ainv)!trait_tarsus.2           12.8988848 8.0388517
## units:trait!R                         1.0000000      NA
## units:trait!trait!tarsus.2:!trait!tarsus.1.cor 0.0000000      NA
## units:trait!trait_tarsus.1             14.8727602 3.6375549
## units:trait!trait_tarsus.2             10.7604420 6.2948051
##                                         z.ratio bound %ch
## trait:byear!trait_tarsus.1            2.1399167      U  0
## trait:byear!trait_tarsus.2            2.5078553      U  0
## trait:mother!trait_tarsus.1           0.7735373      U  0
## trait:mother!trait_tarsus.2           1.5489160      U  0
## trait:vm(animal, ainv)!trait!tarsus.2:!trait!tarsus.1.cor 2.1539720      U  0
## trait:vm(animal, ainv)!trait_tarsus.1           1.8101638      U  0
## trait:vm(animal, ainv)!trait_tarsus.2           1.6045681      U  0
## units:trait!R                         NA            F  0
## units:trait!trait!tarsus.2:!trait!tarsus.1.cor  NA            F  NA
## units:trait!trait_tarsus.1             4.0886696      P  0
## units:trait!trait_tarsus.2             1.7094162      P  0

```

```

#####
DvsS <- data.frame(
  Trait = rownames(model.BiV_Sex$coefficients$random),
  BLUP = model.BiV_Sex$coefficients$random,
  SE = sqrt(model.BiV_Sex$vcoeff$random * model.BiV_Sex$sigma2)
)
DvsS$ID <- substr(DvsS$Trait, 33, 35)
DvsS$TRAIT <- substr(DvsS$Trait, 7, 14)
DvsS <- DvsS[927:3544, ] # keep only row associated to animal
summary(factor(DvsS$TRAIT))
```

```

## tarsus.1 tarsus.2
##     1309     1309
```

```
#
```

```
DvsS$Trait <- NULL
colnames(DvsS)[1] <- "BLUP"
BLUPS <- reshape(DvsS, v.names = c("BLUP", "SE"), idvar = "ID", timevar = "TRAIT", directi
```

```
## Warning in reshapeWide(data, idvar = idvar, timevar = timevar, varying =
## varying, : multiple rows match for TRAIT=tarsus.1: first taken
```

```
## Warning in reshapeWide(data, idvar = idvar, timevar = timevar, varying =
## varying, : multiple rows match for TRAIT=tarsus.2: first taken
```

```
nrow(BLUPS)
```

```
## [1] 999
```

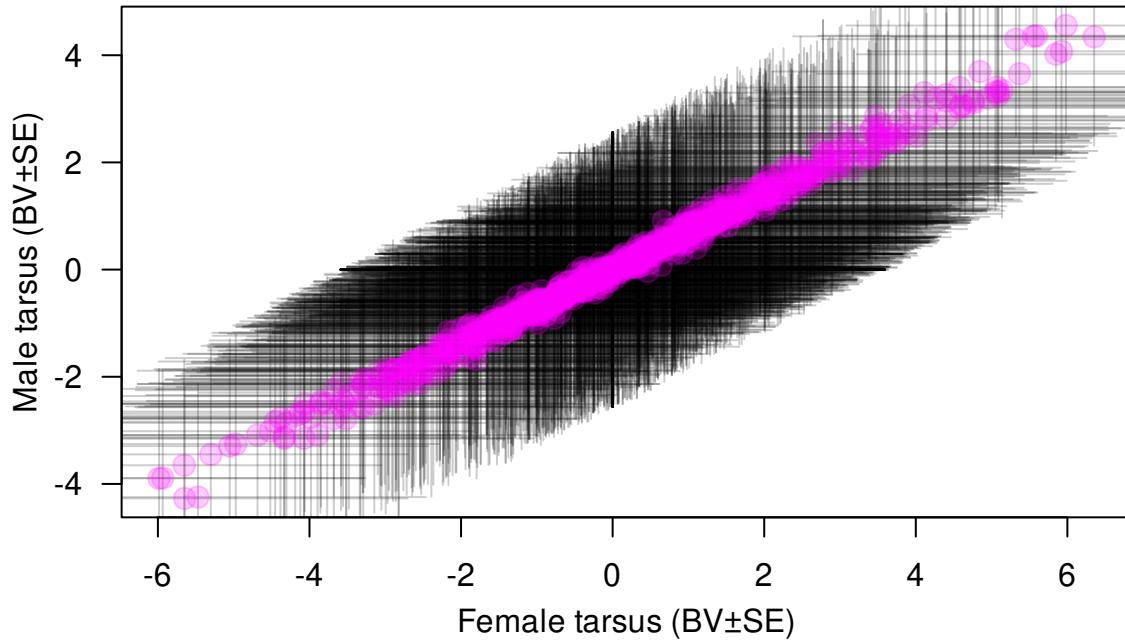
```
rownames(BLUPS) <- c()
colnames(BLUPS) <- c("ID", "BLUP.1", "SE.1", "BLUP.2", "SE.2")
summary(BLUPS)
```

	ID	BLUP.1	SE.1	BLUP.2
## Length:999		Min. :-4.27022	Min. :1.724	Min. :-5.98516
## Class :character		1st Qu.:-0.69816	1st Qu.:2.010	1st Qu.:-0.99808
## Mode :character		Median : 0.00000	Median :2.143	Median : 0.00000
		Mean : 0.05588	Mean :2.204	Mean : 0.06904
		3rd Qu.: 0.84250	3rd Qu.:2.426	3rd Qu.: 1.13426
		Max. : 4.55405	Max. :2.567	Max. : 6.35309
## SE.2				
## Min. :2.375				
## 1st Qu.:2.677				
## Median :3.055				
## Mean :3.047				
## 3rd Qu.:3.378				
## Max. :3.591				

```
#####
```

```
Y <- BLUPS$BLUP.1
X <- BLUPS$BLUP.2
se.Y <- BLUPS$SE.1
se.X <- BLUPS$SE.2
```

```
plot(X, Y, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = X, y0 = Y - se.Y, x1 = X, y1 = Y + se.Y, col = rgb(0, 0, 0, 0.2), code = 3, an
arrows(x0 = X - se.X, y0 = Y, x1 = X + se.X, y1 = Y, col = rgb(0, 0, 0, 0.2), code = 3, an
points(X, Y, pch = 1, col = rgb(1, 0, 1, 0.2), cex = 1.5)
points(X, Y, pch = 16, col = rgb(1, 0, 1, 0.2), cex = 1.5)
# abline(v=0,lty=3);abline(h=0,lty=3)
mtext("Male tarsus (BV±SE)", side = 2, line = 2, las = 3)
mtext("Female tarsus (BV±SE)", side = 1, line = 2.2)
```



The B matrix used the same code but in a multivariate animal model framework. Here some example code, however due to the nature of the dataset, the cross-sex genetic covariance for birth weight is hard to estimate making difficult to fit this multivariate animal model.

```

temp <- asreml(cbind(tarsus.1, bwt.1, tarsus.2, bwt.2) ~ trait,
  random = ~ us(trait):vm(animal, ainv) +
    diag(trait):byear + diag(trait):mother,
  residual = ~ units:us(trait),
  data = gryphon, na.action = na.method(y = "include", x = "include"), maxiter = 20,
  start.values = T
)
G <- temp$vparameters[(1:18), ]
R <- temp$vparameters[-(1:18), ]
#
G$Constraint <- "U"
R$Value[5:6] <- 0
R$Constraint[5:6] <- "F"
R$Value[8:9] <- 0
R$Constraint[8:9] <- "F"
#
# model.MultV_Sex<-asreml(cbind(tarsus.1,bwt.1,tarsus.2,bwt.2)~trait,
#   random=~us(trait):vm(animal,ainv)+
#     diag(trait):byear + diag(trait):mother,
```

```
#      residual = ~units:us(trait),
#      data=gryphon,na.action=na.method(y="include",x="include"),maxiter=20,
#      G.param=G,R.param=R)
# model.MultV_Sex<-update.asreml(model.MultV_Sex)
#
# summary(model.MultV_Sex)$varcomp
```

3.3 gremlin

No in the development plans

Meanwhile



Figure 3.1: Keep it dry and do no feed after midnight.

3.4 MCMCglmm

MCMCglmm has the advantage to keep automatically keep the lines with missing data and will try to fit the model use latent variables for missing data. We will remove the missing values from the data before fitting the model.

```
gryphon2 <- subset(gryphon, !is.na(bwt) & !is.na(tarsus))
```

First load MCMCglmm:

```
library(MCMCglmm)
Ainv <- inverseA(gryphonped)$Ainv
```

3.4.1 Fitting the model

Fitting a multivariate model in MCMCglmm involves several new consideration above those for fitting univariate models. First, we have to fit multivariate priors; second, we have to specify the ways in which effects on different traits may covary, including the nature of residual (co)variation; and third, we will have to be a little more specific when specifying to MCMCglmm what type of distributions from which we assume our data are drawn. Our most basic model can be specified as:

```
prior2.1 <- list(
  G = list(G1 = list(V = diag(2), nu = 1.002)),
  R = list(V = diag(2), nu = 1.002)
)

model2.1 <- MCMCglmm(cbind(bwt, tarsus) ~ trait - 1,
  random = ~ us(trait):animal,
  rcov = ~ us(trait):units,
  family = c("gaussian", "gaussian"),
  ginv = list(animal = Ainv),
  data = gryphon2, prior = prior2.1, verbose = FALSE
)
summary(model2.1)

##
## Iterations = 3001:12991
## Thinning interval = 10
## Sample size = 1000
##
## DIC: 7156.888
##
## G-structure: ~us(trait):animal
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traitbwt:traitbwt.animal      3.236  1.98774   4.656   117.84
## traittarsus:traitbwt.animal   2.045 -0.02795   4.612    97.11
## traitbwt:traittarsus.animal   2.045 -0.02795   4.612    97.11
## traittarsus:traittarsus.animal 11.357  5.21354  17.365    65.68
##
## R-structure: ~us(trait):units
##
```

```

## post.mean l-95% CI u-95% CI eff.samp
## traitbwt:traitbwt.units      3.951    2.790    5.074    154.3
## traittarsus:traitbwt.units   3.641    1.736    5.841    115.4
## traitbwt:traittarsus.units   3.641    1.736    5.841    115.4
## traittarsus:traittarsus.units 18.646   13.084   24.144   102.7
##
## Location effects: cbind(bwt, tarsus) ~ trait - 1
##
## post.mean l-95% CI u-95% CI eff.samp pMCMC
## traitbwt      7.483    7.166    7.749    609.9 <0.001 ***
## traittarsus   20.424   19.784   20.972   1000.0 <0.001 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
plot(model2.1$VCV[, "traittarsus:traittarsus.animal"])
```

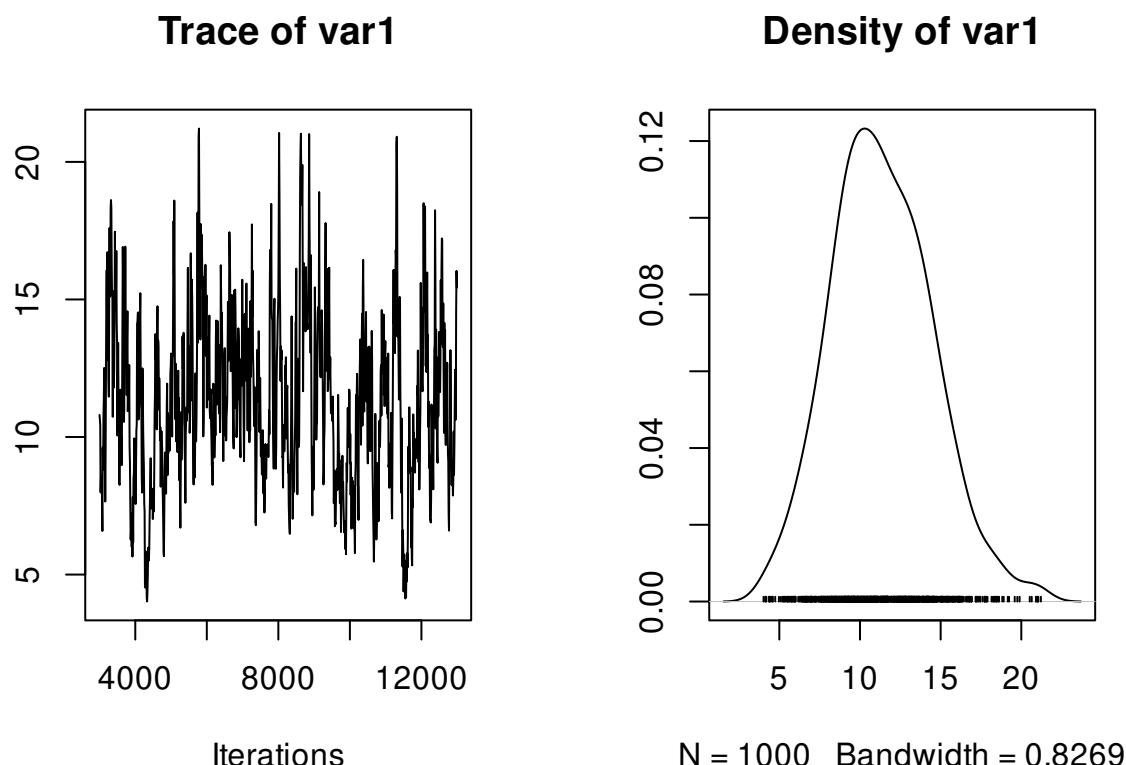


Figure 3.2: The posterior distribution of the additive genetic effect for tarsus length in a MCMCglmm run with default values

```
autocorr.diag(model2.1$VCV)[, "traittarsus:traittarsus.animal"][2]
```

```
##    Lag 10
## 0.8199896
```

We have constructed the prior similarly to those in the univariate models in tutorial 1, only we are specifying a 2x2 covariance matrix rather than a single variance. In order to provide proper priors, we have set the degree of belief parameter to greater than 1 (1.002). Those priors are not necessarily weak or uninformative in all circumstances. We will consider them adequate nonetheless for this tutorial. Please see the vignette of the MCMCglmm package (?) for more information on priors. In tutorial 1, we used full autocorrelation tables to evaluate the validity of the posterior distribution. Note that we have not done this here.

For a bivariate model this table can become very complex. Nonetheless, it is worth evaluating, rather it is simply too large to include here. It can be viewed in the console as before. Here we have displayed only the autocorrelation for estimates of additive genetic effects for tarsus length with a lag of one samples (10 iterations given this MCMCglmm run with default values). This lag of 0.8199896 is clearly unacceptable. The posterior distribution of the additive genetic effect on tarsus length is shown in Figure 4, note the autocorrelation evident in the left-hand plot.

We will opt to run the analysis for longer. This longer run could be run using the following code (including a line to save the output):

```
model2.1 <- MCMCglmm(cbind(bwt, tarsus) ~ trait - 1,
  random = ~ us(trait):animal,
  rcov = ~ us(trait):units,
  family = c("gaussian", "gaussian"),
  ginv = list(animal = Ainv),
  data = gryphon2,
  nitt = 130000, thin = 100, burnin = 30000,
  prior = prior2.1, verbose = FALSE
)
save(model2.1, file = "data/MCMCglmm_model2_1_LongRun.rda")
```

However, this run might take as long as an hour. For the purpose of this tutorial we have provided an output for such a run. It can be obtained and manipulated as follows, assuming that the file `MCMCglmm_model2_1_LongRun.rda` is available at the specified location:

```
load(file = "data/MCMCglmm_model2_1_LongRun.rda")
autocorr.diag(model2.1$VCV) [, "traittarsus:traittarsus.animal"] [2]
```

```
##    Lag 100
## 0.2805747
```

This level of autocorrelation is more acceptable, at least for the purpose of demonstration in this tutorial. We can recover variance components, heritabilities, and genetic correlations from the posterior distribution of this model:

```
posterior.mode(model2.1$VCV)
```

```
##          traitbwt:traitbwt.animal      traittarsus:traitbwt.animal
##                      3.147319                  2.390698
##          traitbwt:traittarsus.animal  traittarsus:traittarsus.animal
##                      2.390698                  10.863567
##          traitbwt:traitbwt.units     traittarsus:traitbwt.units
##                      3.823107                  4.044831
##          traitbwt:traittarsus.units  traittarsus:traittarsus.units
##                      4.044831                  17.734520
```

```
heritability.bwt2.1 <- model2.1$VCV[, "traitbwt:traitbwt.animal"] / (model2.1$VCV[, "traitbwt:traitbwt.animal"])
posterior.mode(heritability.bwt2.1)
```

```
##      var1
## 0.4999336
```

```
heritability.tarsus2.1 <- model2.1$VCV[, "traittarsus:traittarsus.animal"] / (model2.1$VCV[, "traittarsus:traittarsus.animal"])
posterior.mode(heritability.tarsus2.1)
```

```
##      var1
## 0.3698826
```

```
genetic.correlation2.1 <- model2.1$VCV[, "traitbwt:traittarsus.animal"] / sqrt(model2.1$VCV[, "traitbwt:traittarsus.animal"])
posterior.mode(genetic.correlation2.1)
```

```
##      var1
## 0.3815069
```

3.4.2 Adding fixed and random effects

Fixed and random effects can be added just as for the univariate case. Given that our full model of bwt from tutorial 1 had sex as a fixed effect as well as random effects of byear and mother, we could specify a bivariate formulation of this using the following code (including a line to save the output):

```
prior2.2 <- list(
  G = list(
    G1 = list(V = diag(2), nu = 1.002),
    G2 = list(V = diag(2), nu = 1.002),
    G3 = list(V = diag(2), nu = 1.002)
  ),
  R = list(V = diag(2), nu = 1.002)
```

```
)
model2.2 <- MCMCglmm(cbind(bwt, tarsus) ~ trait - 1 + trait:sex,
  random = ~ us(trait):animal + us(trait):byear + us(trait):mother,
  rcov = ~ us(trait):units,
  family = c("gaussian", "gaussian"),
  ginv = list(animal = Ainv), data = gryphon2,
  nitt = 130000, thin = 100, burnin = 30000,
  prior = prior2.2, verbose = FALSE
)
save(model2.2, file = "data/MCMCglmm_model2_2_LongRun.rda")
```

Again we have provided the data from one such run. It can be accessed using the code:

```
load(file = "data/MCMCglmm_model2_2_LongRun.rda")
summary(model2.2)
```

```
##
## Iterations = 30001:129901
## Thinning interval = 100
## Sample size = 1000
##
## DIC: 5832.952
##
## G-structure: ~us(trait):animal
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traitbwt:traitbwt.animal      1.558   0.5616   2.488   230.8
## traittarsus:traitbwt.animal   2.290   0.3241   4.264   274.8
## traitbwt:traittarsus.animal   2.290   0.3241   4.264   274.8
## traittarsus:traittarsus.animal 8.083   0.9063  13.599   228.1
##
##                               ~us(trait):byear
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traitbwt:traitbwt.byear      0.96775  0.4124   1.5053   1000
## traittarsus:traitbwt.byear    0.07332 -0.8100   0.9791   1000
## traitbwt:traittarsus.byear    0.07332 -0.8100   0.9791   1000
## traittarsus:traittarsus.byear 3.80720  1.6291   6.3986   1000
##
##                               ~us(trait):mother
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traitbwt:traitbwt.mother     1.335   0.8564   1.8090   871.2
## traittarsus:traitbwt.mother   -1.508  -2.1667  -0.8288   648.6
## traitbwt:traittarsus.mother   -1.508  -2.1667  -0.8288   648.6
```

```

## traittarsus:traittarsus.mother      4.292   2.2380   6.6336   796.0
##
## R-structure: ~us(trait):units
##
##          post.mean l-95% CI u-95% CI eff.samp
## traitbwt:traitbwt.units           2.13    1.304   2.939   469.2
## traittarsus:traitbwt.units       4.81    3.111   6.568   414.7
## traitbwt:traittarsus.units       4.81    3.111   6.568   414.7
## traittarsus:traittarsus.units    14.51   9.419  19.892   261.3
##
## Location effects: cbind(bwt, tarsus) ~ trait - 1 + trait:sex
##
##          post.mean l-95% CI u-95% CI eff.samp pMCMC
## traitbwt           6.2734   5.8152   6.7272   1205 <0.001 ***
## traittarsus        20.3985  19.4021  21.4106   1000 <0.001 ***
## traitbwt:sex2      2.0354   1.7347   2.3529   1000 <0.001 ***
## traittarsus:sex2    0.0705  -0.6949   0.7686   1000  0.868
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
autocorr(model2.2$VCV)[, , "traittarsus:traittarsus.animal"] [3, 4]
```

```
## [1] 0.1026744
```

We can evaluate the fixed effect, their Ci evaluate their significance.

```
posterior.mode(model2.2$Sol)
```

```

##      traitbwt      traittarsus      traitbwt:sex2 traittarsus:sex2
##      6.26902047     20.35816977      2.06048779     -0.06501522

```

```
HPDinterval(model2.2$Sol, 0.95)
```

```

##          lower      upper
## traitbwt      5.8151983  6.7272503
## traittarsus    19.4021008 21.4106029
## traitbwt:sex2    1.7347121  2.3528879
## traittarsus:sex2 -0.6948574  0.7686074
## attr(,"Probability")
## [1] 0.95

```

```
plot(model2.2$Sol)
```

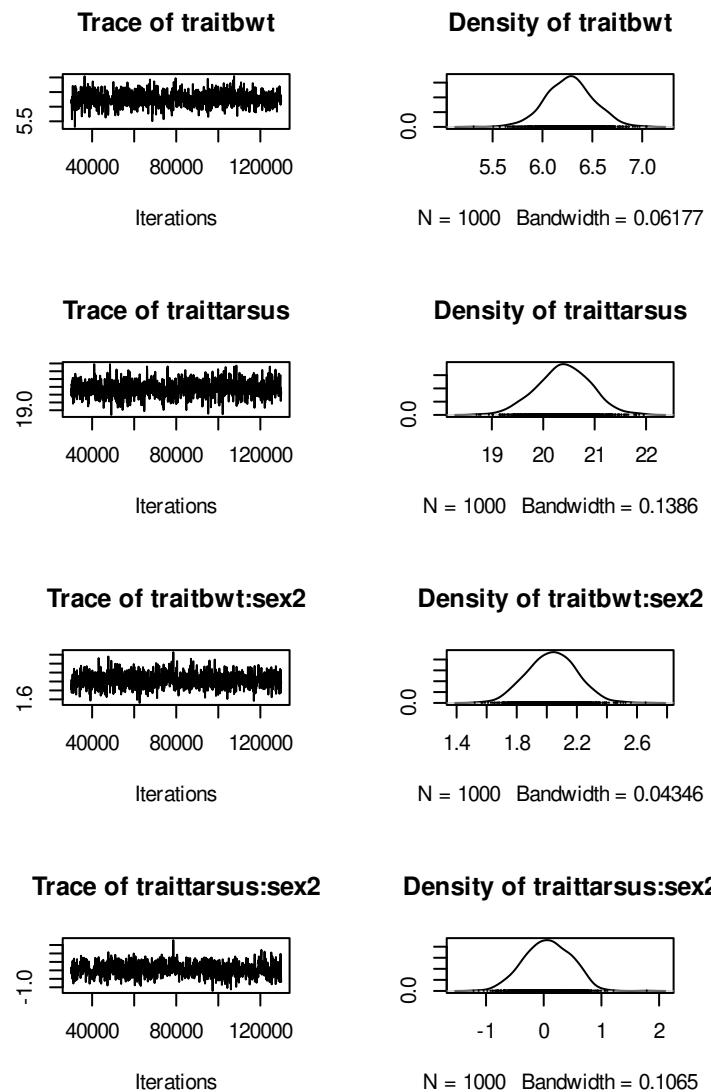


Figure 3.3: Posterior trace and distribution for the fixed effects in model 2.2

As before we can obtain the raw variance component estimates and genetic correlations for the random effects:

```
posterior.mode(model2.2$VCV)
```

```
##          traitbwt:traitbwt.animal      traittarsus:traitbwt.animal
##                1.3294950                  2.0622374
## traitbwt:traittarsus.animal traittarsus:traittarsus.animal
##                2.0622374                  8.3900676
##          traitbwt:traitbwt.byear      traittarsus:traitbwt.byear
##                0.8118565                  0.2327381
## traitbwt:traittarsus.byear traittarsus:traittarsus.byear
##                0.2327381                  3.7375906
```

```

##          traitbwt:traitbwt.mother      traittarsus:traitbwt.mother
##                           1.4089440                  -1.4963686
##          traitbwt:traittarsus.mother  traittarsus:traittarsus.mother
##                           -1.4963686                   3.9386669
##          traitbwt:traitbwt.units     traittarsus:traitbwt.units
##                           2.2353960                  4.3432849
##          traitbwt:traittarsus.units  traittarsus:traittarsus.units
##                           4.3432849                  15.0853981

genetic.correlation2.2 <- model2.2$VCV[, "traitbwt:traittarsus.animal"] / sqrt(model2.2$VCV[, "traitbwt:traittarsus.animal"])
maternal.correlation2.2 <- model2.2$VCV[, "traitbwt:traittarsus.mother"] / sqrt(model2.2$VCV[, "traitbwt:traittarsus.mother"])
posterior.mode(genetic.correlation2.2)

```

```

##      var1
## 0.6932486

```

```
posterior.mode(maternal.correlation2.2)
```

```

##      var1
## -0.7431221

```

Evaluation of the statistical support for these genetic and maternal correlations is straightforward. Because we imposed no constraint on their estimation, we can evaluate the extent to which the posterior distributions overlap zero:

```
HPDinterval(genetic.correlation2.2, 0.95)
```

```

##           lower      upper
## var1 0.3062932 0.9197543
## attr(,"Probability")
## [1] 0.95

```

```
HPDinterval(maternal.correlation2.2, 0.95)
```

```

##           lower      upper
## var1 -0.9432297 -0.3210149
## attr(,"Probability")
## [1] 0.95

```

Neither or these posterior distributions overlaps zero, so we can consider them both statistically supported.

3.4.3 Direct estimate of the correlation instead of the covariance.

For this example, we just estimate the correlation at the genetic level, the covariance for the other random effect (`mother` and `byear`) and the residual level was not estimate to help the model to

converge and compute faster. The prior will be the same but we change the pr argument to be TRUE to keep the posterior distribution of random effects. To simplify the following code and facilitate the BLUP extraction, we rename the variable T1 and T2 and estimate correlation only for the additive genetic and residual matrices.

```

gryphon2$T1 <- gryphon2$bwt
gryphon2$T2 <- gryphon2$tarsus
#
model2.3 <- MCMCglmm(cbind(T1, T2) ~ trait - 1 + trait:sex,
  random = ~ corg(trait):animal + corg(trait):byear + corg(trait):mother,
  rcov = ~ corg(trait):units,
  family = c("gaussian", "gaussian"),
  ginv = list(animal = Ainv), data = gryphon2,
  nitt = 130000, thin = 100, burnin = 30000,
  prior = prior2.2, verbose = FALSE, pr = TRUE,
)
save(model2.3, file = "data/MCMCglmm_model2_3_LongRun.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load(file = "data/MCMCglmm_model2_3_LongRun.rda")
summary(model2.3)

##
## Iterations = 30001:129901
## Thinning interval = 100
## Sample size = 1000
##
## DIC: 972.9044
##
## G-structure: ~corg(trait):animal
##
##          post.mean l-95% CI u-95% CI eff.samp
## traitT1:traitT1.animal      1       1       1       0
## traitT2:traitT1.animal     -1      -1      -1       0
## traitT1:traitT2.animal     -1      -1      -1       0
## traitT2:traitT2.animal      1       1       1       0
##
##          ~corg(trait):byear
##
##          post.mean l-95% CI u-95% CI eff.samp
## traitT1:traitT1.byear    1.0000  1.00000   1.000       0
## traitT2:traitT1.byear    0.2436 -0.09388   0.588    1000
## traitT1:traitT2.byear    0.2436 -0.09388   0.588    1000
## traitT2:traitT2.byear    1.0000  1.00000   1.000       0

```

```

## ~corg(trait):mother
##
##          post.mean l-95% CI u-95% CI eff.samp
## traitT1:traitT1.mother      1       1       1       0
## traitT2:traitT1.mother     -1      -1      -1       0
## traitT1:traitT2.mother     -1      -1      -1       0
## traitT2:traitT2.mother      1       1       1       0
##
## R-structure: ~corg(trait):units
##
##          post.mean l-95% CI u-95% CI eff.samp
## traitT1:traitT1.units       1       1       1       0
## traitT2:traitT1.units       1       1       1       0
## traitT1:traitT2.units       1       1       1       0
## traitT2:traitT2.units       1       1       1       0
##
## Location effects: cbind(T1, T2) ~ trait - 1 + trait:sex
##
##          post.mean l-95% CI u-95% CI eff.samp pMCMC
## traitT1        6.29252  5.79418  6.77846    1000 <0.001 ***
## traitT2        20.53149 20.04183 21.00833    1000 <0.001 ***
## traitT1:sex2    2.07254  1.83815  2.32478    1000 <0.001 ***
## traitT2:sex2    0.02657 -0.21610  0.26411    1000   0.864
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
autocorr(model2.3$VCV) [, , "traitT2:traitT1.animal"] [3, 4]
```

```
## [1] NaN
```

Here we can plot the genetic correlation by extraction the breeding values or BLUP. Just to remember it is an example, the correlation is close to 1 due to a weak prior and model parameters.

```

DvsS <- data.frame(
  Trait = colnames(model2.3$Sol),
  BLUP = posterior.mode(model2.3$Sol),
  CI = HPDinterval((model2.3$Sol))
)
DvsS <- DvsS[5:2622, ] # keep only rows associated with animal

DvsS$ID <- substr(DvsS$Trait, 16, 19)
DvsS$TRAIT <- substr(DvsS$Trait, 6, 7)
summary(factor(DvsS$TRAIT))

```

```
##   T1   T2
```

```
## 1309 1309
```

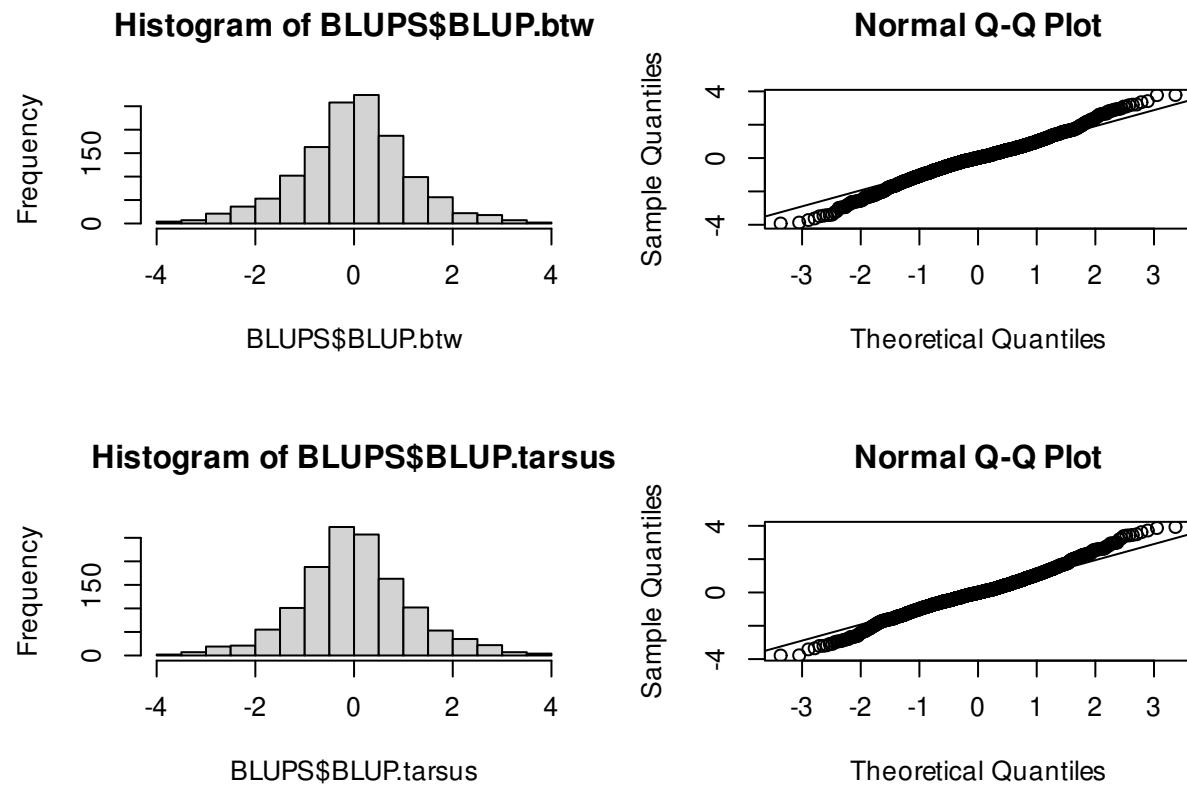
```
DvsS$Trait <- NULL
BLUPS <- reshape(DvsS, v.names = c("BLUP", "CI.lower", "CI.upper"), idvar = "ID", timevar
nrow(BLUPS)
```

```
## [1] 1309
```

```
rownames(BLUPS) <- c()
colnames(BLUPS) <- c("ID", "BLUP.btw", "CI.L.btw", "CI.U.btw", "BLUP.tarsus", "CI.L.tarsus")
summary(BLUPS)
```

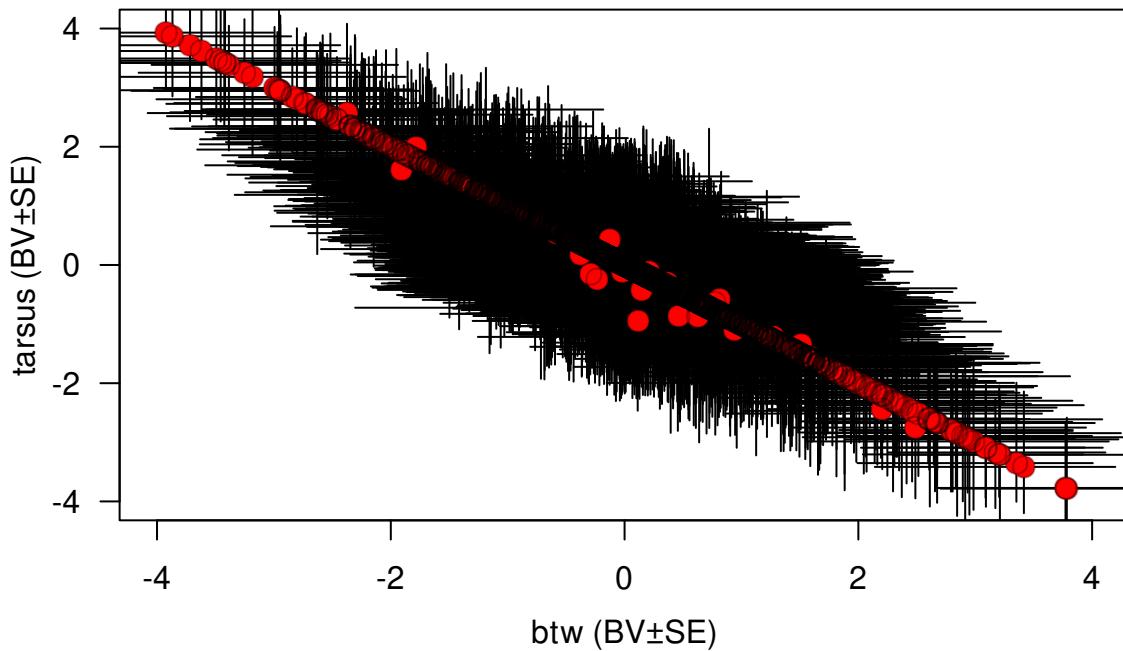
	ID	BLUP.btw	CI.L.btw	CI.U.btw
## Length:	1309	Min. : -3.92390	Min. : -5.0229	Min. : -3.0062
## Class :	character	1st Qu.: -0.65954	1st Qu.: -2.1218	1st Qu.: 0.7462
## Mode :	character	Median : 0.01183	Median : -1.5472	Median : 1.5745
		Mean : -0.02007	Mean : -1.4347	Mean : 1.3860
		3rd Qu.: 0.63842	3rd Qu.: -0.7509	3rd Qu.: 2.1033
		Max. : 3.78330	Max. : 2.7030	Max. : 4.7851
## BLUP.tarsus		CI.L.tarsus	CI.U.tarsus	
## Min. : -3.78353		Min. : -4.6656	Min. : -2.6778	
## 1st Qu.: -0.64453		1st Qu.: -2.1033	1st Qu.: 0.7505	
## Median : -0.02336		Median : -1.5741	Median : 1.5477	
## Mean : 0.01820		Mean : -1.3864	Mean : 1.4343	
## 3rd Qu.: 0.65947		3rd Qu.: -0.7485	3rd Qu.: 2.1219	
## Max. : 3.93053		Max. : 3.0061	Max. : 5.0231	

```
#  
par(mfrow = c(2, 2))  
hist(BLUPS$BLUP.btw)  
qqnorm(BLUPS$BLUP.btw)  
qqline(BLUPS$BLUP.btw)  
hist(BLUPS$BLUP.tarsus)  
qqnorm(BLUPS$BLUP.tarsus)  
qqline(BLUPS$BLUP.tarsus)
```



Here the code to plot the genetic correlation.

```
plot(BLUP.tarsus ~ BLUP.btw, BLUPS, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = BLUPS$BLUP.btw, y0 = BLUPS$CI.L.tarsus, x1 = BLUPS$BLUP.btw, y1 = BLUPS$CI.U.tarsus)
arrows(x0 = BLUPS$CI.L.btw, y0 = BLUPS$BLUP.tarsus, x1 = BLUPS$CI.U.btw, y1 = BLUPS$BLUP.tarsus)
points(BLUP.tarsus ~ BLUP.btw, BLUPS, pch = 16, col = "red", cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, BLUPS, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("btw (BV±SE)", side = 1, line = 2.4)
mtext("tarsus (BV±SE)", side = 2, line = 2, las = 3)
```



3.4.4 Partitioning (co)variances

As in the tutorial 1, it is possible to partition the variance-covariance matrix between groups (here sex). Note: the model is simplified without sex-specific covariance for the byear and mother random effect.

```

gryphon2 <- gryphon2[order(gryphon2$sex), ]

prior2.3 <- list(
  G = list(
    G1 = list(V = diag(2), nu = 1.002),
    G2 = list(V = diag(2), nu = 1.002),
    G3 = list(V = diag(2), nu = 1.002),
    G4 = list(V = diag(2), nu = 1.002)
  ),
  R = list(
    V1 = list(V = diag(2), nu = 1.002),
    V2 = list(V = diag(2), nu = 1.002)
  )
)

```

```

model2.4 <- MCMCglmm(cbind(bwt, tarsus) ~ trait - 1 + trait:sex,
  random = ~ us(at.level(sex, "1"):trait):animal + us(at.level(sex, "2"):trait):animal + i,
  rcov = ~ us(at.level(sex, "1"):trait):units + us(at.level(sex, "2"):trait):units,
  family = c("gaussian", "gaussian"),
  ginv = list(animal = Ainv), data = gryphon2,
  nitt = 130000, thin = 100, burnin = 30000,
  prior = prior2.3, verbose = FALSE, pr = TRUE,
)
save(model2.4, file = "data/MCMCglmm_model2_4_LongRun.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load(file = "data/MCMCglmm_model2_4_LongRun.rda")
summary(model2.4)

```

```

##
## Iterations = 30001:129901
## Thinning interval = 100
## Sample size = 1000
##
## DIC: 5576.328
##
## G-structure: ~us(at.level(sex, "1"):trait):animal
##
##                                         post.mean
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal      1.122
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal    1.127
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal    1.127
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal   3.379
##                                         1-95% CI
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal      0.1602
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal    -0.6531
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal    -0.6531
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal   0.1844
##                                         u-95% CI
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal      2.359
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal    3.496
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal    3.496
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal   8.918
##                                         eff.samp
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal      167.5
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal    119.3
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal    119.3
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal   102.6
##
```

```

## ~us(at.level(sex, "2"):trait):animal
##
## post.mean
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal      1.598
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal    3.099
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal    3.099
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal 10.218
##                                         1-95% CI
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal      0.1895
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal    -0.5506
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal    -0.5506
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal  0.2127
##                                         u-95% CI
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal      3.305
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal    7.864
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal    7.864
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal 24.230
##                                         eff.samp
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal      57.28
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal    42.01
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal    42.01
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal 37.21
##
## ~idh(trait):byear
##
## post.mean 1-95% CI u-95% CI eff.samp
## traitbwt.byear     0.9309   0.4614   1.463    1000
## traittarsus.byear  4.0310   1.9268   6.724    1000
##
## ~idh(trait):mother
##
## post.mean 1-95% CI u-95% CI eff.samp
## traitbwt.mother    1.924    1.406   2.398    667.6
## traittarsus.mother  7.093    4.626   9.681    698.5
##
## R-structure: ~us(at.level(sex, "1"):trait):units
##
## post.mean
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units      2.090
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units    4.533
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units    4.533
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units 14.113
##                                         1-95% CI
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units      0.9958
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units    2.4185
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units    2.4185
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units  8.1848

```

```

##                                         u-95% CI
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units      3.128
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.units    6.875
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units    6.875
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units 18.935
##                                         eff.samp
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units      207.2
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.units    168.4
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units    168.4
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units 185.3
##
##                                         ~us(at.level(sex, "2"):trait):units
##
##                                         post.mean
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units      1.782
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units    3.697
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units    3.697
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units 12.437
##                                         l-95% CI
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units      0.2776
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units   -0.3141
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units   -0.3141
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units  0.1776
##                                         u-95% CI
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units      3.115
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units    7.218
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units    7.218
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units 21.903
##                                         eff.samp
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units      52.55
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units    51.90
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units    51.90
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units 39.20
##
## Location effects: cbind(bwt, tarsus) ~ trait - 1 + trait:sex
##
##                                         post.mean l-95% CI u-95% CI eff.samp pMCMC
## traitbwt          6.30098  5.89218  6.78834  1000.0 <0.001 ***
## traiittarsus     20.45577 19.53577 21.34719  1129.8 <0.001 ***
## traitbwt:sex2    2.01306  1.63662  2.38011   887.4 <0.001 ***
## traiittarsus:sex2 0.05817 -0.86635  0.89119  1016.6  0.896
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
autocorr(model2.4$VCV)
```

```
## , , at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
##                                     at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                                         1.00000000
## Lag 100                                       0.64694479
## Lag 500                                       0.18648179
## Lag 1000                                      0.10392219
## Lag 5000                                      -0.04275072
##                                     at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                                         0.84704874
## Lag 100                                       0.60161240
## Lag 500                                       0.20180692
## Lag 1000                                      0.10068129
## Lag 5000                                      -0.03878312
##                                     at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                                         0.84704874
## Lag 100                                       0.60161240
## Lag 500                                       0.20180692
## Lag 1000                                      0.10068129
## Lag 5000                                      -0.03878312
##                                     at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                                         0.53041407
## Lag 100                                       0.39413485
## Lag 500                                       0.16964194
## Lag 1000                                      0.11264314
## Lag 5000                                      -0.01013697
##                                     at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                         -0.07132000
## Lag 100                                       -0.09608251
## Lag 500                                       -0.05360431
## Lag 1000                                      -0.02600414
## Lag 5000                                      -0.02326421
##                                     at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                                         -0.07404287
## Lag 100                                       -0.08742103
## Lag 500                                       -0.05376905
## Lag 1000                                      -0.03219125
## Lag 5000                                      0.02771727
##                                     at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                                         -0.07404287
## Lag 100                                       -0.08742103
## Lag 500                                       -0.05376905
## Lag 1000                                      -0.03219125
## Lag 5000                                      0.02771727
```

```

##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                               -0.06663301
## Lag 100                            -0.07398282
## Lag 500                            -0.03873715
## Lag 1000                           -0.03346102
## Lag 5000                           0.06535632
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      -0.002044905     0.06061428   -0.13681757    0.063034744
## Lag 100     -0.029101625     0.04741082   -0.09232454    0.062553003
## Lag 500     -0.025891155     0.04101237   -0.01510511   -0.026837884
## Lag 1000    0.029398462     0.02792539   -0.02514900    0.009578198
## Lag 5000   -0.016122661     0.03081539   0.04189460   -0.039791141
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                               -0.842319278
## Lag 100                            -0.569203867
## Lag 500                            -0.167844469
## Lag 1000                           -0.114647645
## Lag 5000                           -0.002132053
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                               -0.708901550
## Lag 100                            -0.517998161
## Lag 500                            -0.167589741
## Lag 1000                           -0.110500558
## Lag 5000                           0.002914291
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                               -0.708901550
## Lag 100                            -0.517998161
## Lag 500                            -0.167589741
## Lag 1000                           -0.110500558
## Lag 5000                           0.002914291
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0                               -0.438123204
## Lag 100                            -0.337083166
## Lag 500                            -0.129292647
## Lag 1000                           -0.103678560
## Lag 5000                           -0.001926232
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                               0.07807105
## Lag 100                            0.10707885
## Lag 500                            0.05568856
## Lag 1000                           0.02521629
## Lag 5000                           0.01402475
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                               0.06679340
## Lag 100                            0.08704308
## Lag 500                            0.05892190
## Lag 1000                           0.02676188

```

```

## Lag 5000                               -0.03056683
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0                                    0.06679340
## Lag 100                                 0.08704308
## Lag 500                                 0.05892190
## Lag 1000                                0.02676188
## Lag 5000                                -0.03056683
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0                                    0.04797898
## Lag 100                                 0.05730717
## Lag 500                                 0.04727555
## Lag 1000                                0.02677473
## Lag 5000                                -0.06608227
##
## , , at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                                    0.847048735
## Lag 100                                 0.596411029
## Lag 500                                 0.228550625
## Lag 1000                                0.137616124
## Lag 5000                                0.009429906
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                                    1.000000000
## Lag 100                                 0.71730236
## Lag 500                                 0.27616079
## Lag 1000                                0.13795063
## Lag 5000                                0.01144749
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                                    1.000000000
## Lag 100                                 0.71730236
## Lag 500                                 0.27616079
## Lag 1000                                0.13795063
## Lag 5000                                0.01144749
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                                    0.7989970
## Lag 100                                 0.6014134
## Lag 500                                 0.2515262
## Lag 1000                                0.1354306
## Lag 5000                                0.0136445
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                    -0.069644149
## Lag 100                                 -0.094348331
## Lag 500                                 -0.069174874
## Lag 1000                                -0.030980734
## Lag 5000                                -0.001770693
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal

```

```

## Lag 0                               -0.09266557
## Lag 100                             -0.10417316
## Lag 500                             -0.06908668
## Lag 1000                            -0.04934221
## Lag 5000                            0.03968797
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                               -0.09266557
## Lag 100                             -0.10417316
## Lag 500                             -0.06908668
## Lag 1000                            -0.04934221
## Lag 5000                            0.03968797
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                               -0.10219567
## Lag 100                             -0.10740690
## Lag 500                             -0.05829130
## Lag 1000                            -0.05667648
## Lag 5000                            0.08311412
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      -0.03731153      0.05572330     -0.12626725      0.06865980
## Lag 100    -0.04492620      0.05076637     -0.08142219      0.05404288
## Lag 500    -0.03460527      0.03246607     -0.03107773     -0.02899561
## Lag 1000   0.01459594      0.01717445     -0.05078674      0.01110690
## Lag 5000   -0.01688700     0.03883380     0.02698184     -0.03307579
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                               -0.73143141
## Lag 100                             -0.52689086
## Lag 500                             -0.22551523
## Lag 1000                            -0.12616708
## Lag 5000                            -0.04647821
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                               -0.82495927
## Lag 100                            -0.60990914
## Lag 500                            -0.24464022
## Lag 1000                           -0.12721295
## Lag 5000                           -0.03841367
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                               -0.82495927
## Lag 100                            -0.60990914
## Lag 500                            -0.24464022
## Lag 1000                           -0.12721295
## Lag 5000                           -0.03841367
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0                               -0.64394327
## Lag 100                            -0.48995337
## Lag 500                            -0.19725633
## Lag 1000                           -0.10852446
## Lag 5000                           -0.02105523

```

```

##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                               0.082481767
## Lag 100                            0.105417000
## Lag 500                            0.073280263
## Lag 1000                           0.028355398
## Lag 5000                           -0.006019758
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                               0.09308370
## Lag 100                            0.10680733
## Lag 500                            0.07810010
## Lag 1000                           0.04359553
## Lag 5000                           -0.04726853
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0                               0.09308370
## Lag 100                            0.10680733
## Lag 500                            0.07810010
## Lag 1000                           0.04359553
## Lag 5000                           -0.04726853
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0                               0.08843633
## Lag 100                            0.09343376
## Lag 500                            0.06886426
## Lag 1000                           0.05338682
## Lag 5000                           -0.09185034
##
## , , at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
##
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                               0.847048735
## Lag 100                            0.596411029
## Lag 500                            0.228550625
## Lag 1000                           0.137616124
## Lag 5000                           0.009429906
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                               1.000000000
## Lag 100                            0.71730236
## Lag 500                            0.27616079
## Lag 1000                           0.13795063
## Lag 5000                           0.01144749
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                               1.000000000
## Lag 100                            0.71730236
## Lag 500                            0.27616079
## Lag 1000                           0.13795063
## Lag 5000                           0.01144749
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                               0.7989970

```

```

## Lag 100                               0.6014134
## Lag 500                               0.2515262
## Lag 1000                              0.1354306
## Lag 5000                              0.0136445
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                -0.069644149
## Lag 100                             -0.094348331
## Lag 500                             -0.069174874
## Lag 1000                            -0.030980734
## Lag 5000                            -0.001770693
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                                -0.09266557
## Lag 100                             -0.10417316
## Lag 500                             -0.06908668
## Lag 1000                            -0.04934221
## Lag 5000                            0.03968797
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                                -0.09266557
## Lag 100                             -0.10417316
## Lag 500                             -0.06908668
## Lag 1000                            -0.04934221
## Lag 5000                            0.03968797
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                                -0.10219567
## Lag 100                             -0.10740690
## Lag 500                             -0.05829130
## Lag 1000                            -0.05667648
## Lag 5000                            0.08311412
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      -0.03731153      0.05572330     -0.12626725      0.06865980
## Lag 100     -0.04492620      0.05076637     -0.08142219      0.05404288
## Lag 500     -0.03460527      0.03246607     -0.03107773     -0.02899561
## Lag 1000    0.01459594      0.01717445     -0.05078674      0.01110690
## Lag 5000   -0.01688700      0.03883380     0.02698184     -0.03307579
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.73143141
## Lag 100                             -0.52689086
## Lag 500                             -0.22551523
## Lag 1000                            -0.12616708
## Lag 5000                            -0.04647821
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.82495927
## Lag 100                             -0.60990914
## Lag 500                             -0.24464022
## Lag 1000                            -0.12721295
## Lag 5000                            -0.03841367
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units

```

```

## Lag 0           -0.82495927
## Lag 100        -0.60990914
## Lag 500        -0.24464022
## Lag 1000       -0.12721295
## Lag 5000       -0.03841367
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0           -0.64394327
## Lag 100         -0.48995337
## Lag 500         -0.19725633
## Lag 1000        -0.10852446
## Lag 5000        -0.02105523
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0           0.082481767
## Lag 100         0.105417000
## Lag 500         0.073280263
## Lag 1000        0.028355398
## Lag 5000        -0.006019758
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0           0.09308370
## Lag 100         0.10680733
## Lag 500         0.07810010
## Lag 1000        0.04359553
## Lag 5000        -0.04726853
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0           0.09308370
## Lag 100         0.10680733
## Lag 500         0.07810010
## Lag 1000        0.04359553
## Lag 5000        -0.04726853
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0           0.08843633
## Lag 100         0.09343376
## Lag 500         0.06886426
## Lag 1000        0.05338682
## Lag 5000        -0.09185034
##
## , , at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0           0.5304141
## Lag 100         0.3737195
## Lag 500         0.1441203
## Lag 1000        0.1503417
## Lag 5000        0.1187940
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0           0.7989970
## Lag 100         0.5706521

```

```

## Lag 500          0.2430451
## Lag 1000         0.1680830
## Lag 5000         0.1259980
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0            0.7989970
## Lag 100          0.5706521
## Lag 500          0.2430451
## Lag 1000         0.1680830
## Lag 5000         0.1259980
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0            1.00000000
## Lag 100          0.73196692
## Lag 500          0.31335783
## Lag 1000         0.18501263
## Lag 5000         0.08438218
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0            -0.01785209
## Lag 100          -0.03508025
## Lag 500          -0.04733762
## Lag 1000         -0.01709422
## Lag 5000         -0.01586047
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0            -0.038020441
## Lag 100          -0.045171003
## Lag 500          -0.050004069
## Lag 1000         -0.054183547
## Lag 5000         -0.004955516
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0            -0.038020441
## Lag 100          -0.045171003
## Lag 500          -0.050004069
## Lag 1000         -0.054183547
## Lag 5000         -0.004955516
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0            -0.05447207
## Lag 100          -0.05998184
## Lag 500          -0.06158778
## Lag 1000         -0.08267333
## Lag 5000         0.02065741
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0            -0.060159939    0.06450755   -0.0973321863   -0.009350685
## Lag 100          -0.043720033    0.03483594   -0.0765923141   -0.006212912
## Lag 500          -0.052466217    0.02987272   -0.0662772868   -0.030465249
## Lag 1000         -0.001034192    0.03110963   -0.0728720391   0.009855596
## Lag 5000         -0.034160786    0.05472996   0.0008533055   0.003426058
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0            -0.4637838

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## Lag 100           -0.3337342
## Lag 500           -0.1618186
## Lag 1000          -0.1329387
## Lag 5000          -0.1239580
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0              -0.6583252
## Lag 100            -0.4881849
## Lag 500            -0.2287167
## Lag 1000           -0.1381960
## Lag 5000           -0.1179873
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0              -0.6583252
## Lag 100             0.4881849
## Lag 500             0.2287167
## Lag 1000            0.1381960
## Lag 5000            0.1179873
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0              -0.76001059
## Lag 100             0.57358014
## Lag 500             0.25179771
## Lag 1000            0.12123408
## Lag 5000            0.07986147
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0               0.02870904
## Lag 100              0.03399695
## Lag 500              0.05485675
## Lag 1000             0.01651664
## Lag 5000             0.01516504
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0               0.039633904
## Lag 100              0.041502118
## Lag 500              0.059018043
## Lag 1000             0.052958967
## Lag 5000             -0.002274568
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0               0.039633904
## Lag 100              0.041502118
## Lag 500              0.059018043
## Lag 1000             0.052958967
## Lag 5000             -0.002274568
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0               0.04584246
## Lag 100              0.04763329
## Lag 500              0.06852725
## Lag 1000             0.08362165
## Lag 5000             -0.03430204
##

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## , , at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                               -0.07132000
## Lag 100                             -0.06108550
## Lag 500                             -0.06344456
## Lag 1000                            -0.02628413
## Lag 5000                            0.10351490
##      at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                               -0.0696441487
## Lag 100                            -0.0685711479
## Lag 500                            -0.0543839240
## Lag 1000                           0.0004950661
## Lag 5000                           0.1221823016
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal
## Lag 0                               -0.0696441487
## Lag 100                            -0.0685711479
## Lag 500                            -0.0543839240
## Lag 1000                           0.0004950661
## Lag 5000                           0.1221823016
##      at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal
## Lag 0                               -0.0178520882
## Lag 100                            -0.0173974776
## Lag 500                            -0.0002494694
## Lag 1000                           0.0551913450
## Lag 5000                           0.1333840825
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                               1.0000000
## Lag 100                            0.8242352
## Lag 500                            0.5296802
## Lag 1000                           0.3040607
## Lag 5000                           -0.1276161
##      at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                               0.9099634
## Lag 100                            0.8047694
## Lag 500                            0.5857973
## Lag 1000                           0.3552775
## Lag 5000                           -0.1485103
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.animal
## Lag 0                               0.9099634
## Lag 100                            0.8047694
## Lag 500                            0.5857973
## Lag 1000                           0.3552775
## Lag 5000                           -0.1485103
##      at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.animal
## Lag 0                               0.7704756
## Lag 100                            0.7082472

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## Lag 500                               0.5569812
## Lag 1000                             0.3517296
## Lag 5000                            -0.1453072
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      -0.03784246     0.04916122    -0.025145260   -0.13999847
## Lag 100    -0.01888261     0.04551933    -0.009748633   -0.14616483
## Lag 500    -0.01864811     0.07395050     0.029035276   -0.12958636
## Lag 1000   -0.02117775     0.06164183     0.068666314   -0.09577992
## Lag 5000   0.01769136      0.04869291     0.037573009   0.01686724
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                                0.048426202
## Lag 100                              0.039795097
## Lag 500                              0.046311373
## Lag 1000                             0.005469282
## Lag 5000                            -0.104125437
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                                0.043207572
## Lag 100                             0.043210916
## Lag 500                             0.033044478
## Lag 1000                            -0.004411742
## Lag 5000                            -0.110707718
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                                0.043207572
## Lag 100                             0.043210916
## Lag 500                             0.033044478
## Lag 1000                            -0.004411742
## Lag 5000                            -0.110707718
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0                                0.01554818
## Lag 100                             0.02352457
## Lag 500                             0.01217491
## Lag 1000                            -0.02394172
## Lag 5000                            -0.13812594
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                                -0.9369474
## Lag 100                             -0.8092241
## Lag 500                             -0.5186132
## Lag 1000                            -0.2947735
## Lag 5000                            0.1226249
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                                -0.8725969
## Lag 100                             -0.7954903
## Lag 500                             -0.5688119
## Lag 1000                            -0.3390939
## Lag 5000                            0.1455303
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0                                -0.8725969

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## Lag 100          -0.7954903
## Lag 500          -0.5688119
## Lag 1000         -0.3390939
## Lag 5000          0.1455303
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0             -0.7525481
## Lag 100            0.7086874
## Lag 500            0.5421136
## Lag 1000           0.3389906
## Lag 5000           0.1485387
##
## , , at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.074042865
## Lag 100            0.072737049
## Lag 500            0.064855516
## Lag 1000           0.004245299
## Lag 5000           0.126495395
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.092665568
## Lag 100            0.096939661
## Lag 500            0.070837135
## Lag 1000           0.006501962
## Lag 5000           0.148898005
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.092665568
## Lag 100            0.096939661
## Lag 500            0.070837135
## Lag 1000           0.006501962
## Lag 5000           0.148898005
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.038020441
## Lag 100            0.039681669
## Lag 500            0.006820427
## Lag 1000           0.063529955
## Lag 5000           0.163665055
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0              0.9099634
## Lag 100            0.7863387
## Lag 500            0.5413307
## Lag 1000           0.3118422
## Lag 5000           -0.1191809
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0              1.0000000
## Lag 100            0.8933098
## Lag 500            0.6382613

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## Lag 1000          0.3875538
## Lag 5000         -0.1480316
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.animal
## Lag 0            1.0000000
## Lag 100          0.8933098
## Lag 500          0.6382613
## Lag 1000         0.3875538
## Lag 5000         -0.1480316
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.animal
## Lag 0            0.9445430
## Lag 100          0.8642926
## Lag 500          0.6377101
## Lag 1000         0.3971583
## Lag 5000         -0.1545401
## traitbwt.byear  traiittarsus.byear  traitbwt.mother  traiittarsus.mother
## Lag 0            -0.04691870    0.05505699    0.03372293   -0.18311492
## Lag 100          -0.03261563    0.04790144    0.03175029   -0.18180192
## Lag 500          -0.02904559    0.05050843    0.04377317   -0.16642684
## Lag 1000         -0.03811545    0.05361475    0.07019878   -0.12467546
## Lag 5000         0.04062218    0.04654678    0.03310770   0.01019974
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0            0.04443896
## Lag 100          0.04551916
## Lag 500          0.04357037
## Lag 1000         -0.01792034
## Lag 5000         -0.12752563
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0            0.04900783
## Lag 100          0.06050152
## Lag 500          0.04114675
## Lag 1000         -0.01839006
## Lag 5000         -0.13689966
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units
## Lag 0            0.04900783
## Lag 100          0.06050152
## Lag 500          0.04114675
## Lag 1000         -0.01839006
## Lag 5000         -0.13689966
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units
## Lag 0            0.02477085
## Lag 100          0.03939172
## Lag 500          0.02069295
## Lag 1000         -0.04020316
## Lag 5000         -0.16427556
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0            -0.8748666
## Lag 100          -0.7865075

```

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## Lag 500          -0.5347377
## Lag 1000         -0.3099292
## Lag 5000          0.1242901
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0             -0.9629330
## Lag 100            -0.8851422
## Lag 500            -0.6211970
## Lag 1000           -0.3754777
## Lag 5000            0.1511004
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0             -0.9629330
## Lag 100            -0.8851422
## Lag 500            -0.6211970
## Lag 1000           -0.3754777
## Lag 5000            0.1511004
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0             -0.9191068
## Lag 100            -0.8609263
## Lag 500            -0.6233180
## Lag 1000           -0.3879589
## Lag 5000            0.1606066
##
## , , at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.074042865
## Lag 100            -0.072737049
## Lag 500            -0.064855516
## Lag 1000           -0.004245299
## Lag 5000            0.126495395
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.092665568
## Lag 100            -0.096939661
## Lag 500            -0.070837135
## Lag 1000           0.006501962
## Lag 5000            0.148898005
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.092665568
## Lag 100            -0.096939661
## Lag 500            -0.070837135
## Lag 1000           0.006501962
## Lag 5000            0.148898005
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.038020441
## Lag 100            -0.039681669
## Lag 500            -0.006820427
## Lag 1000            0.063529955

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## Lag 5000                                0.163665055
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal      0.9099634
## Lag 0                                     0.7863387
## Lag 100                                    0.5413307
## Lag 500                                    0.3118422
## Lag 1000                                   -0.1191809
## Lag 5000

## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal    1.0000000
## Lag 0                                     0.8933098
## Lag 100                                    0.6382613
## Lag 500                                    0.3875538
## Lag 1000                                   -0.1480316
## Lag 5000

## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal     1.0000000
## Lag 0                                     0.8933098
## Lag 100                                    0.6382613
## Lag 500                                    0.3875538
## Lag 1000                                   -0.1480316
## Lag 5000

## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal   0.9445430
## Lag 0                                     0.8642926
## Lag 100                                    0.6377101
## Lag 500                                    0.3971583
## Lag 1000                                   -0.1545401
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0          -0.04691870      0.05505699      0.03372293      -0.18311492
## Lag 100         -0.03261563      0.04790144      0.03175029      -0.18180192
## Lag 500         -0.02904559      0.05050843      0.04377317      -0.16642684
## Lag 1000        -0.03811545      0.05361475      0.07019878      -0.12467546
## Lag 5000        0.04062218      0.04654678      0.03310770      0.01019974
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units       0.04443896
## Lag 0                                     0.04551916
## Lag 100                                    0.04357037
## Lag 500                                    -0.01792034
## Lag 1000                                   -0.12752563
## Lag 5000

## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units     0.04900783
## Lag 0                                     0.06050152
## Lag 100                                    0.04114675
## Lag 500                                    -0.01839006
## Lag 1000                                   -0.13689966
## Lag 5000

## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units      0.04900783
## Lag 0                                     0.06050152
## Lag 100                                    0.04114675

```

```

## Lag 1000          -0.01839006
## Lag 5000          -0.13689966
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0              0.02477085
## Lag 100             0.03939172
## Lag 500             0.02069295
## Lag 1000            -0.04020316
## Lag 5000            -0.16427556
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0              -0.8748666
## Lag 100             -0.7865075
## Lag 500             -0.5347377
## Lag 1000            -0.3099292
## Lag 5000             0.1242901
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0              -0.9629330
## Lag 100             -0.8851422
## Lag 500             -0.6211970
## Lag 1000            -0.3754777
## Lag 5000             0.1511004
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0              -0.9629330
## Lag 100             -0.8851422
## Lag 500             -0.6211970
## Lag 1000            -0.3754777
## Lag 5000             0.1511004
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0              -0.9191068
## Lag 100             -0.8609263
## Lag 500             -0.6233180
## Lag 1000            -0.3879589
## Lag 5000             0.1606066
##
## , , at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
##
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0              -0.066633008
## Lag 100             -0.069354252
## Lag 500             -0.053416684
## Lag 1000            0.001180564
## Lag 5000             0.142470162
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0              -0.102195672
## Lag 100             -0.107130141
## Lag 500             -0.073576929
## Lag 1000            -0.004189061
## Lag 5000             0.167339055

```

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##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0          -0.102195672
## Lag 100        -0.107130141
## Lag 500        -0.073576929
## Lag 1000       -0.004189061
## Lag 5000       0.167339055
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0          -0.05447207
## Lag 100        -0.05707224
## Lag 500        -0.01853426
## Lag 1000       0.04684883
## Lag 5000       0.17837197
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0          0.7704756
## Lag 100        0.6843951
## Lag 500        0.5050605
## Lag 1000       0.2977972
## Lag 5000       -0.1058223
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0          0.9445430
## Lag 100        0.8556548
## Lag 500        0.6303848
## Lag 1000       0.3904095
## Lag 5000       -0.1424423
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0          0.9445430
## Lag 100        0.8556548
## Lag 500        0.6303848
## Lag 1000       0.3904095
## Lag 5000       -0.1424423
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0          1.0000000
## Lag 100        0.9100529
## Lag 500        0.6616827
## Lag 1000       0.4121439
## Lag 5000       -0.1590568
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0          -0.03974818    0.04854354    0.04924110   -0.22289117
## Lag 100        -0.03767078    0.04176415    0.03930577   -0.20950998
## Lag 500        -0.03440434    0.03555315    0.05305906   -0.18185253
## Lag 1000       -0.02822560    0.04567963    0.06510782   -0.12092658
## Lag 5000       0.05272181    0.04246380    0.02211597   0.02036647
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0          0.04061110
## Lag 100        0.04206692
## Lag 500        0.03373343
## Lag 1000       -0.01881550

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## Lag 5000 -0.13859641
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.units 0.05879880
## Lag 0 0.06872711
## Lag 100 0.04634425
## Lag 500 -0.01579800
## Lag 1000 -0.14960797
## Lag 5000
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units 0.05879880
## Lag 0 0.06872711
## Lag 100 0.04634425
## Lag 500 -0.01579800
## Lag 1000 -0.14960797
## Lag 5000
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units 0.04254878
## Lag 0 0.05461840
## Lag 100 0.03646094
## Lag 500 -0.04023461
## Lag 1000 -0.17207003
## Lag 5000
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units -0.7532567
## Lag 0 -0.6923793
## Lag 100 -0.5032427
## Lag 500 -0.2991945
## Lag 1000 0.1154818
## Lag 5000
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units -0.9140715
## Lag 0 -0.8493221
## Lag 100 -0.6164356
## Lag 500 -0.3808768
## Lag 1000 0.1473936
## Lag 5000
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units -0.9140715
## Lag 0 -0.8493221
## Lag 100 -0.6164356
## Lag 500 -0.3808768
## Lag 1000 0.1473936
## Lag 5000
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units -0.9650519
## Lag 0 -0.9008410
## Lag 100 -0.6496396
## Lag 500 -0.4073019
## Lag 1000 0.1644354
##
## , , traitbwt.byear
##
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal

```

```

## Lag 0           -0.002044905
## Lag 100        0.018082206
## Lag 500        -0.019694583
## Lag 1000       -0.033624772
## Lag 5000       -0.025949000
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0           -0.03731153
## Lag 100         -0.01355344
## Lag 500         -0.01864081
## Lag 1000        -0.05745850
## Lag 5000        -0.01235998
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0           -0.03731153
## Lag 100         -0.01355344
## Lag 500         -0.01864081
## Lag 1000        -0.05745850
## Lag 5000        -0.01235998
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0           -0.06015994
## Lag 100         -0.03212487
## Lag 500         -0.02412236
## Lag 1000        -0.05846861
## Lag 5000        -0.02882580
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0           -0.03784246
## Lag 100         -0.04180932
## Lag 500         -0.04438042
## Lag 1000        -0.01257459
## Lag 5000        0.00630995
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0           -0.04691870
## Lag 100         -0.04133222
## Lag 500         -0.05230682
## Lag 1000        -0.02870414
## Lag 5000        -0.00469889
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0           -0.04691870
## Lag 100         -0.04133222
## Lag 500         -0.05230682
## Lag 1000        -0.02870414
## Lag 5000        -0.00469889
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0           -0.039748177
## Lag 100         -0.032934501
## Lag 500         -0.055295362
## Lag 1000        -0.027884156
## Lag 5000        -0.007044631

```

```

## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      1.00000000 -0.0251146296  0.03365469  0.03928862
## Lag 100    0.03109454  0.0004436899 -0.05764761 -0.01264335
## Lag 500    0.03937305  0.0006604187 -0.00457655  0.02746272
## Lag 1000   0.01680424 -0.0194711518  0.03737600 -0.04627035
## Lag 5000   0.03318792  0.0155533971 -0.02558374  0.05305580
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0          0.005983125
## Lag 100        -0.001889062
## Lag 500        -0.018793288
## Lag 1000       0.027363658
## Lag 5000       0.010334637
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0          0.052806759
## Lag 100         0.014657374
## Lag 500        -0.020921457
## Lag 1000       0.041519184
## Lag 5000       -0.000172048
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0          0.052806759
## Lag 100         0.014657374
## Lag 500        -0.020921457
## Lag 1000       0.041519184
## Lag 5000       -0.000172048
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0          0.072105699
## Lag 100         0.023891187
## Lag 500        -0.022626087
## Lag 1000       0.038071084
## Lag 5000       0.008022532
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0          0.02472261
## Lag 100         0.04922524
## Lag 500         0.05684465
## Lag 1000        0.02213746
## Lag 5000        -0.02587314
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0          0.035150019
## Lag 100         0.039823504
## Lag 500         0.058615425
## Lag 1000        0.039863168
## Lag 5000        -0.002768445
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0          0.035150019
## Lag 100         0.039823504
## Lag 500         0.058615425
## Lag 1000        0.039863168

```

```

## Lag 5000 -0.002768445
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units 0.032332494
## Lag 0 0.036826480
## Lag 100 0.056371336
## Lag 500 0.037506421
## Lag 1000 -0.002071877
## Lag 5000
##
## , , traittarsus.byear
##
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal 0.06061428
## Lag 0 0.06276970
## Lag 100 -0.02842127
## Lag 500 0.01799228
## Lag 1000 0.02740499
## Lag 5000
##
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal 0.05572330
## Lag 0 0.06655805
## Lag 100 -0.02673025
## Lag 500 0.04345968
## Lag 1000 0.05112113
## Lag 5000
##
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal 0.05572330
## Lag 0 0.06655805
## Lag 100 -0.02673025
## Lag 500 0.04345968
## Lag 1000 0.05112113
## Lag 5000
##
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal 0.064507548
## Lag 0 0.074840509
## Lag 100 -0.003777881
## Lag 500 0.058609933
## Lag 1000 0.063485567
## Lag 5000
##
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal 0.049161224
## Lag 0 0.009105861
## Lag 100 0.005065210
## Lag 500 0.016389664
## Lag 1000 -0.029590445
## Lag 5000
##
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal 0.055056994
## Lag 0 0.035711495
## Lag 100 0.012368434
## Lag 500 0.004770290
## Lag 1000 -0.009144398
## Lag 5000
##
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal

```

```

## Lag 0          0.055056994
## Lag 100       0.035711495
## Lag 500       0.012368434
## Lag 1000      0.004770290
## Lag 5000      -0.009144398
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0          0.048543542
## Lag 100        0.034488675
## Lag 500        0.009703880
## Lag 1000       -0.001685047
## Lag 5000       0.005011858
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0          -0.025114630   1.000000000  0.03708995  0.07084541
## Lag 100         -0.033801997  0.041927040  -0.06653239  0.04503853
## Lag 500         0.009533405  -0.020053055  0.01042960  -0.03755216
## Lag 1000        -0.003946143  0.011455578  -0.01588844  0.01986940
## Lag 5000        0.027020776  0.002689451  -0.02585871  0.02687208
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0          -0.10335662
## Lag 100        -0.05376108
## Lag 500        0.01626001
## Lag 1000       -0.01344940
## Lag 5000       -0.02833156
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0          -0.09746412
## Lag 100        -0.05607997
## Lag 500        0.01829969
## Lag 1000       -0.04558284
## Lag 5000       -0.04241014
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0          -0.09746412
## Lag 100        -0.05607997
## Lag 500        0.01829969
## Lag 1000       -0.04558284
## Lag 5000       -0.04241014
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0          -0.108393481
## Lag 100        -0.049448255
## Lag 500        0.002404817
## Lag 1000       -0.086421792
## Lag 5000       -0.055605953
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0          -0.025767223
## Lag 100        -0.006184423
## Lag 500        -0.002201914
## Lag 1000       -0.011174601
## Lag 5000       0.029950491

```

```

##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                               -0.049170830
## Lag 100                            -0.040276502
## Lag 500                            0.003528012
## Lag 1000                           -0.010662330
## Lag 5000                           0.002523808
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0                               -0.049170830
## Lag 100                            -0.040276502
## Lag 500                            0.003528012
## Lag 1000                           -0.010662330
## Lag 5000                           0.002523808
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0                               -0.053929470
## Lag 100                            -0.043281273
## Lag 500                            0.009317392
## Lag 1000                           -0.010594624
## Lag 5000                           -0.013754908
##
## , , traitbwt.mother
##
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                               -0.13681757
## Lag 100                            -0.09694549
## Lag 500                            -0.06857367
## Lag 1000                           -0.04540954
## Lag 5000                           -0.01652050
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                               -0.12626725
## Lag 100                            -0.12449687
## Lag 500                            -0.05181080
## Lag 1000                           -0.03932960
## Lag 5000                           -0.01141931
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                               -0.12626725
## Lag 100                            -0.12449687
## Lag 500                            -0.05181080
## Lag 1000                           -0.03932960
## Lag 5000                           -0.01141931
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                               -0.09733219
## Lag 100                            -0.10049386
## Lag 500                            -0.04634235
## Lag 1000                           -0.01749975
## Lag 5000                           -0.03285757
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                               -0.02514526

```

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## Lag 100          0.01494004
## Lag 500         -0.05400749
## Lag 1000        -0.02644804
## Lag 5000        -0.02759428
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0           0.0337229276
## Lag 100         0.0355284011
## Lag 500         -0.0008562576
## Lag 1000        -0.0192570169
## Lag 5000        -0.0314028551
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.animal
## Lag 0           0.0337229276
## Lag 100         0.0355284011
## Lag 500         -0.0008562576
## Lag 1000        -0.0192570169
## Lag 5000        -0.0314028551
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.animal
## Lag 0           0.04924110
## Lag 100         0.04492689
## Lag 500         0.02422838
## Lag 1000        -0.04034312
## Lag 5000        -0.03784851
## traitbwt.byear traiittarsus.byear traitbwt.mother traiittarsus.mother
## Lag 0           0.033654686   0.037089946   1.000000000   -0.267715213
## Lag 100          0.020365368   0.031597781   0.039464037   0.006711605
## Lag 500          0.007110008   0.046188516   0.095615498   -0.023010721
## Lag 1000         -0.019597442   0.001266059   0.065362608   0.041196297
## Lag 5000          0.019704700   -0.034265234   -0.005121853   0.041919494
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0           0.093898173
## Lag 100         0.109444195
## Lag 500         0.055322096
## Lag 1000        0.036648121
## Lag 5000        0.003492676
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0           0.11659923
## Lag 100         0.12608289
## Lag 500         0.03369955
## Lag 1000        0.04312267
## Lag 5000        -0.01418292
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units
## Lag 0           0.11659923
## Lag 100         0.12608289
## Lag 500         0.03369955
## Lag 1000        0.04312267
## Lag 5000        -0.01418292
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units

```

```

## Lag 0                               0.1007331132
## Lag 100                             0.0846271381
## Lag 500                             0.0314559531
## Lag 1000                            0.0313378649
## Lag 5000                            0.0006030047
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                               -0.02385685
## Lag 100                             -0.01929710
## Lag 500                             0.05442789
## Lag 1000                            0.02999688
## Lag 5000                            0.02354946
##      at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                               -0.037784916
## Lag 100                             -0.040619404
## Lag 500                             0.009246757
## Lag 1000                            0.021585046
## Lag 5000                            0.029657103
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units
## Lag 0                               -0.037784916
## Lag 100                             -0.040619404
## Lag 500                             0.009246757
## Lag 1000                            0.021585046
## Lag 5000                            0.029657103
##      at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units
## Lag 0                               -0.04048151
## Lag 100                             -0.04664306
## Lag 500                             -0.02420607
## Lag 1000                            0.03414246
## Lag 5000                            0.04075949
##
## , , traiittarsus.mother
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                               0.06303474
## Lag 100                            0.09137304
## Lag 500                            0.05137407
## Lag 1000                           0.02569160
## Lag 5000                           -0.08057411
##      at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                               0.06865980
## Lag 100                            0.10336350
## Lag 500                            0.01711371
## Lag 1000                           0.03032742
## Lag 5000                           -0.08530728
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal
## Lag 0                               0.06865980
## Lag 100                            0.10336350

```

```

## Lag 500                                0.01711371
## Lag 1000                               0.03032742
## Lag 5000                              -0.08530728
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                                 -0.009350685
## Lag 100                               0.024288467
## Lag 500                               -0.016438414
## Lag 1000                             -0.005767054
## Lag 5000                            -0.063956859
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                -0.13999847
## Lag 100                             -0.14330129
## Lag 500                             -0.12944310
## Lag 1000                           -0.08446537
## Lag 5000                           0.04104776
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                                -0.18311492
## Lag 100                             -0.16614099
## Lag 500                             -0.14634974
## Lag 1000                           -0.09164415
## Lag 5000                           0.05689178
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                                -0.18311492
## Lag 100                             -0.16614099
## Lag 500                             -0.14634974
## Lag 1000                           -0.09164415
## Lag 5000                           0.05689178
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                                -0.22289117
## Lag 100                             -0.19084285
## Lag 500                             -0.14966648
## Lag 1000                           -0.07718413
## Lag 5000                           0.07429663
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0       0.039288617    0.07084541   -0.267715213   1.000000000
## Lag 100     0.047364166   -0.01932534   -0.036245609   0.088363955
## Lag 500     0.005475011   -0.04374386   -0.028017777   0.043041568
## Lag 1000    -0.022031785    0.01752292    0.005084865   -0.003286219
## Lag 5000    -0.004605383   -0.01801176   -0.022372822   -0.055614496
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.04283736
## Lag 100                             -0.05247358
## Lag 500                             -0.03520027
## Lag 1000                           -0.02685262
## Lag 5000                           0.08204082
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.080407800

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## Lag 100           -0.079283374
## Lag 500           0.007896803
## Lag 1000          -0.024892006
## Lag 5000          0.090297411
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units
## Lag 0             -0.080407800
## Lag 100           -0.079283374
## Lag 500           0.007896803
## Lag 1000          -0.024892006
## Lag 5000          0.090297411
##      at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units
## Lag 0             -0.09075088
## Lag 100           -0.04605175
## Lag 500           0.02816248
## Lag 1000          0.01663048
## Lag 5000          0.08246697
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0             0.14299876
## Lag 100           0.14159603
## Lag 500           0.13583757
## Lag 1000          0.06878998
## Lag 5000          -0.03559959
##      at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0             0.16291072
## Lag 100           0.16523003
## Lag 500           0.14972084
## Lag 1000          0.08594345
## Lag 5000          -0.05664141
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units
## Lag 0             0.16291072
## Lag 100           0.16523003
## Lag 500           0.14972084
## Lag 1000          0.08594345
## Lag 5000          -0.05664141
##      at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units
## Lag 0             0.17073426
## Lag 100           0.18983154
## Lag 500           0.15067362
## Lag 1000          0.07940280
## Lag 5000          -0.07670042
##
## , , at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.84231928
## Lag 100           -0.57945611
## Lag 500           -0.19062716

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## Lag 1000          -0.08668794
## Lag 5000          0.01749515
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.73143141
## Lag 100            0.54471632
## Lag 500           -0.20876950
## Lag 1000           0.08873424
## Lag 5000           0.01590906
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.73143141
## Lag 100            0.54471632
## Lag 500           -0.20876950
## Lag 1000           0.08873424
## Lag 5000           0.01590906
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.463783799
## Lag 100            0.353636210
## Lag 500           -0.169902631
## Lag 1000           0.092155921
## Lag 5000           -0.009813848
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0              0.048426202
## Lag 100             0.082944312
## Lag 500             0.033604930
## Lag 1000            0.005353498
## Lag 5000             0.014843101
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0              0.04443896
## Lag 100             0.05903744
## Lag 500             0.02909782
## Lag 1000            0.01112292
## Lag 5000           -0.02822596
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0              0.04443896
## Lag 100             0.05903744
## Lag 500             0.02909782
## Lag 1000            0.01112292
## Lag 5000           -0.02822596
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0              0.04061110
## Lag 100             0.04147746
## Lag 500             0.01338370
## Lag 1000            0.01709317
## Lag 5000           -0.05401662
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      0.005983125     -0.103356615     0.0938981727    -0.0428373611
## Lag 100     0.042918382     -0.047853012     0.0762514968    -0.0699317580

```

```

## Lag 500    -0.002564868    -0.017832691    0.0157985450    0.0235550018
## Lag 1000   0.014867426    -0.027311021    0.0003829822   -0.0006460503
## Lag 5000   0.009101393    0.004430949    -0.0233572527    0.0174497074
##          at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0           1.000000000
## Lag 100        0.503164974
## Lag 500        0.176766919
## Lag 1000       0.100125592
## Lag 5000       0.003858174
##          at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0           0.862131977
## Lag 100         0.467009268
## Lag 500         0.173071923
## Lag 1000        0.091405415
## Lag 5000        0.001214249
##          at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0           0.862131977
## Lag 100         0.467009268
## Lag 500         0.173071923
## Lag 1000        0.091405415
## Lag 5000        0.001214249
##          at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0           0.556754885
## Lag 100         0.300025056
## Lag 500         0.130914950
## Lag 1000        0.084742808
## Lag 5000        0.008877657
##          at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0           -0.066893803
## Lag 100         -0.088991169
## Lag 500         -0.041740101
## Lag 1000        -0.013627370
## Lag 5000        -0.004864431
##          at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0           -0.041695516
## Lag 100         -0.058375215
## Lag 500         -0.038590340
## Lag 1000        -0.008465406
## Lag 5000        0.033653455
##          at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0           -0.041695516
## Lag 100         -0.058375215
## Lag 500         -0.038590340
## Lag 1000        -0.008465406
## Lag 5000        0.033653455
##          at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0           -0.020866307

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## Lag 100                               -0.024784750
## Lag 500                               -0.024026474
## Lag 1000                              -0.008875445
## Lag 5000                             0.060413047
##
## , , at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
##
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                                -0.70890155
## Lag 100                              -0.51875562
## Lag 500                              -0.20600428
## Lag 1000                             -0.10994315
## Lag 5000                            -0.01039121
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                                -0.82495927
## Lag 100                             -0.61316330
## Lag 500                             -0.24079957
## Lag 1000                            -0.11403500
## Lag 5000                            -0.01904085
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                                -0.82495927
## Lag 100                             -0.61316330
## Lag 500                             -0.24079957
## Lag 1000                            -0.11403500
## Lag 5000                            -0.01904085
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                                -0.65832516
## Lag 100                             -0.49644020
## Lag 500                             -0.22156354
## Lag 1000                            -0.11606788
## Lag 5000                            -0.03431864
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                0.0432075717
## Lag 100                             0.0731729081
## Lag 500                             0.0351321242
## Lag 1000                            0.0008078044
## Lag 5000                            0.0060255376
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                                0.04900783
## Lag 100                             0.06034422
## Lag 500                             0.03147880
## Lag 1000                            0.01663910
## Lag 5000                            -0.02708842
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                                0.04900783
## Lag 100                             0.06034422
## Lag 500                             0.03147880

```

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## Lag 1000          0.01663910
## Lag 5000         -0.02708842
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0             0.05879880
## Lag 100            0.06066830
## Lag 500            0.02143188
## Lag 1000           0.02664768
## Lag 5000          -0.05794619
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      0.05280676   -0.097464115   0.11659923   -0.080407800
## Lag 100     0.06749027   -0.048884392   0.05889480   -0.042771302
## Lag 500     0.01581190   -0.013402510   0.01353140   0.023637489
## Lag 1000    0.02427770   -0.012851631   0.01997733   0.008272035
## Lag 5000    -0.00118757    0.004909443   -0.02042686   0.001970307
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0             0.8621320
## Lag 100            0.4520274
## Lag 500            0.1932988
## Lag 1000           0.1091092
## Lag 5000           0.0303842
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0             1.00000000
## Lag 100            0.52253994
## Lag 500            0.19894517
## Lag 1000           0.10703411
## Lag 5000           0.03011013
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0             1.00000000
## Lag 100            0.52253994
## Lag 500            0.19894517
## Lag 1000           0.10703411
## Lag 5000           0.03011013
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0             0.83204286
## Lag 100            0.40931694
## Lag 500            0.16394862
## Lag 1000           0.09392225
## Lag 5000           0.02432038
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0             -0.066443086
## Lag 100            -0.077647262
## Lag 500            -0.036441782
## Lag 1000           -0.007304102
## Lag 5000           -0.002379788
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0             -0.05557720
## Lag 100            -0.05970375

```

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## Lag 500          -0.03617882
## Lag 1000         -0.01614682
## Lag 5000          0.03234714
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0             -0.05557720
## Lag 100            0.05970375
## Lag 500            0.03617882
## Lag 1000           0.01614682
## Lag 5000          0.03234714
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0             -0.04717349
## Lag 100            0.04356349
## Lag 500            0.02677549
## Lag 1000           0.02460366
## Lag 5000          0.06745537
##
## , , at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.70890155
## Lag 100            0.51875562
## Lag 500            0.20600428
## Lag 1000           0.10994315
## Lag 5000          0.01039121
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0             -0.82495927
## Lag 100            0.61316330
## Lag 500            0.24079957
## Lag 1000           0.11403500
## Lag 5000          0.01904085
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.82495927
## Lag 100            0.61316330
## Lag 500            0.24079957
## Lag 1000           0.11403500
## Lag 5000          0.01904085
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0             -0.65832516
## Lag 100            0.49644020
## Lag 500            0.22156354
## Lag 1000           0.11606788
## Lag 5000          0.03431864
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0              0.0432075717
## Lag 100             0.0731729081
## Lag 500             0.0351321242
## Lag 1000            0.0008078044

```

```

## Lag 5000                               0.0060255376
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal   0.04900783
## Lag 0                                     0.06034422
## Lag 100                                    0.03147880
## Lag 500                                    0.01663910
## Lag 1000                                   -0.02708842
## Lag 5000
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal   0.04900783
## Lag 0                                     0.06034422
## Lag 100                                    0.03147880
## Lag 500                                    0.01663910
## Lag 1000                                   -0.02708842
## Lag 5000
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal   0.05879880
## Lag 0                                     0.06066830
## Lag 100                                    0.02143188
## Lag 500                                    0.02664768
## Lag 1000                                   -0.05794619
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0          0.05280676     -0.097464115    0.11659923    -0.080407800
## Lag 100         0.06749027     -0.048884392    0.05889480    -0.042771302
## Lag 500         0.01581190     -0.013402510    0.01353140    0.023637489
## Lag 1000        0.02427770     -0.012851631    0.01997733    0.008272035
## Lag 5000        -0.00118757      0.004909443    -0.02042686    0.001970307
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                           0.8621320
## Lag 100                          0.4520274
## Lag 500                          0.1932988
## Lag 1000                         0.1091092
## Lag 5000                         0.0303842
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                           1.00000000
## Lag 100                          0.52253994
## Lag 500                          0.19894517
## Lag 1000                         0.10703411
## Lag 5000                         0.03011013
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                           1.00000000
## Lag 100                          0.52253994
## Lag 500                          0.19894517
## Lag 1000                         0.10703411
## Lag 5000                         0.03011013
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0                           0.83204286
## Lag 100                          0.40931694
## Lag 500                          0.16394862

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```

## Lag 1000                      0.09392225
## Lag 5000                      0.02432038
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                          -0.066443086
## Lag 100                         -0.077647262
## Lag 500                         -0.036441782
## Lag 1000                        -0.007304102
## Lag 5000                        -0.002379788
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                          -0.05557720
## Lag 100                         -0.05970375
## Lag 500                         -0.03617882
## Lag 1000                        -0.01614682
## Lag 5000                        0.03234714
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0                          -0.05557720
## Lag 100                         -0.05970375
## Lag 500                         -0.03617882
## Lag 1000                        -0.01614682
## Lag 5000                        0.03234714
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0                          -0.04717349
## Lag 100                         -0.04356349
## Lag 500                         -0.02677549
## Lag 1000                        -0.02460366
## Lag 5000                        0.06745537
##
## , , at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
##
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                          -0.43812320
## Lag 100                         -0.32924526
## Lag 500                         -0.13020212
## Lag 1000                        -0.13254862
## Lag 5000                        -0.07309476
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                          -0.64394327
## Lag 100                         -0.46780888
## Lag 500                         -0.20079012
## Lag 1000                        -0.14480996
## Lag 5000                        -0.09575862
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                          -0.64394327
## Lag 100                         -0.46780888
## Lag 500                         -0.20079012
## Lag 1000                        -0.14480996
## Lag 5000                        -0.09575862

```

```

##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                               -0.76001059
## Lag 100                            -0.54663983
## Lag 500                            -0.24713058
## Lag 1000                           -0.16089016
## Lag 5000                           -0.08907569
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                               0.01554818
## Lag 100                            0.03347226
## Lag 500                            0.03054674
## Lag 1000                           0.01335762
## Lag 5000                           0.01644186
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                               0.0247708510
## Lag 100                            0.0298811053
## Lag 500                            0.0393319911
## Lag 1000                           0.0410794684
## Lag 5000                           0.0003106287
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                               0.0247708510
## Lag 100                            0.0298811053
## Lag 500                            0.0393319911
## Lag 1000                           0.0410794684
## Lag 5000                           0.0003106287
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                               0.04254878
## Lag 100                            0.04748420
## Lag 500                            0.04789053
## Lag 1000                           0.06796255
## Lag 5000                           -0.01851941
##           traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      0.072105699   -0.1083934806   0.1007331132   -0.090750879
## Lag 100    0.069867202   -0.0486077910   0.0545104394   -0.013859645
## Lag 500    0.026575267   -0.0133212485   0.0286661585   0.008613454
## Lag 1000   0.015663649   -0.0195256514   0.0413886079   0.020780006
## Lag 5000   -0.002018726  -0.0001505633   0.0007733641   -0.036366883
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                               0.55675488
## Lag 100                            0.28216537
## Lag 500                            0.12989226
## Lag 1000                           0.12353001
## Lag 5000                           0.07787577
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                               0.83204286
## Lag 100                            0.39760050
## Lag 500                            0.17265831
## Lag 1000                           0.12489795

```

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## Lag 5000                               0.08926567
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.units
## Lag 0                                    0.83204286
## Lag 100                                 0.39760050
## Lag 500                                 0.17265831
## Lag 1000                                0.12489795
## Lag 5000                                0.08926567
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units
## Lag 0                                    1.00000000
## Lag 100                                 0.43985525
## Lag 500                                 0.18705457
## Lag 1000                                0.11273003
## Lag 5000                                0.07297707
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                                   -0.03083664
## Lag 100                                 -0.03304267
## Lag 500                                 -0.03123886
## Lag 1000                                -0.02337670
## Lag 5000                                -0.02222725
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                                   -0.0304234955
## Lag 100                                 -0.0241051816
## Lag 500                                 -0.0412351901
## Lag 1000                                -0.0471997761
## Lag 5000                                0.0005985541
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units
## Lag 0                                   -0.0304234955
## Lag 100                                 -0.0241051816
## Lag 500                                 -0.0412351901
## Lag 1000                                -0.0471997761
## Lag 5000                                0.0005985541
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units
## Lag 0                                   -0.03748483
## Lag 100                                 -0.02874080
## Lag 500                                 -0.04666191
## Lag 1000                                -0.07162962
## Lag 5000                                0.02693394
##
## , , at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
##
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                                    0.07807105
## Lag 100                                 0.04809322
## Lag 500                                 0.07342298
## Lag 1000                                0.02747559
## Lag 5000                                -0.08865706
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal

```

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## Lag 0                               0.082481767
## Lag 100                             0.069884466
## Lag 500                             0.053669242
## Lag 1000                            -0.003247808
## Lag 5000                            -0.105675290
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                               0.082481767
## Lag 100                            0.069884466
## Lag 500                            0.053669242
## Lag 1000                           -0.003247808
## Lag 5000                           -0.105675290
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                               0.0287090438
## Lag 100                            0.0207457037
## Lag 500                            0.0006081912
## Lag 1000                           -0.0572950737
## Lag 5000                           -0.1156100914
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                               -0.9369474
## Lag 100                            -0.8074872
## Lag 500                            -0.5046449
## Lag 1000                           -0.2917207
## Lag 5000                           0.1083680
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                               -0.8748666
## Lag 100                            -0.8019088
## Lag 500                            -0.5689707
## Lag 1000                           -0.3490005
## Lag 5000                           0.1337402
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                               -0.8748666
## Lag 100                            -0.8019088
## Lag 500                            -0.5689707
## Lag 1000                           -0.3490005
## Lag 5000                           0.1337402
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                               -0.7532567
## Lag 100                            -0.7106727
## Lag 500                            -0.5497772
## Lag 1000                           -0.3502307
## Lag 5000                           0.1338020
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0       0.024722609     -0.02576722    -0.02385685     0.14299876
## Lag 100    0.017186912     -0.04974888    -0.00291607     0.12584167
## Lag 500    0.027788923     -0.07216401    -0.03437633     0.12502432
## Lag 1000   0.030599400     -0.05970350    -0.07019603     0.09866987
## Lag 5000   0.003688852     -0.03623879    -0.01956357    -0.02929408

```

```

##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                               -0.066893803
## Lag 100                            -0.023868973
## Lag 500                            -0.055699109
## Lag 1000                           -0.006728342
## Lag 5000                           0.089063097
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                               -0.06644309
## Lag 100                            -0.03252225
## Lag 500                            -0.03013542
## Lag 1000                           0.00413424
## Lag 5000                           0.09523719
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                               -0.06644309
## Lag 100                            -0.03252225
## Lag 500                            -0.03013542
## Lag 1000                           0.00413424
## Lag 5000                           0.09523719
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0                               -0.030836644
## Lag 100                            -0.012261003
## Lag 500                            -0.009828189
## Lag 1000                           0.015985541
## Lag 5000                           0.120961240
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                               1.0000000
## Lag 100                            0.7895424
## Lag 500                            0.4966907
## Lag 1000                           0.2824837
## Lag 5000                           -0.1086586
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                               0.9162648
## Lag 100                            0.7897342
## Lag 500                            0.5546655
## Lag 1000                           0.3340592
## Lag 5000                           -0.1357395
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0                               0.9162648
## Lag 100                            0.7897342
## Lag 500                            0.5546655
## Lag 1000                           0.3340592
## Lag 5000                           -0.1357395
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0                               0.7818908
## Lag 100                            0.7088815
## Lag 500                            0.5356249
## Lag 1000                           0.3410568

```

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## Lag 5000                                -0.1407239
##
## , , at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
##
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                                     0.066793402
## Lag 100                                    0.050378446
## Lag 500                                    0.077515159
## Lag 1000                                   0.008812794
## Lag 5000                                   -0.121465553
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                                     0.093083697
## Lag 100                                    0.084384646
## Lag 500                                    0.074812950
## Lag 1000                                   -0.007598499
## Lag 5000                                   -0.145198558
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0                                     0.093083697
## Lag 100                                    0.084384646
## Lag 500                                    0.074812950
## Lag 1000                                   -0.007598499
## Lag 5000                                   -0.145198558
##           at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0                                     0.039633904
## Lag 100                                    0.030631751
## Lag 500                                    0.008216661
## Lag 1000                                   -0.067574178
## Lag 5000                                   -0.158681853
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                     -0.8725969
## Lag 100                                    -0.7772224
## Lag 500                                    -0.5293889
## Lag 1000                                   -0.3047292
## Lag 5000                                   0.1032909
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                                     -0.9629330
## Lag 100                                    -0.8857980
## Lag 500                                    -0.6314814
## Lag 1000                                   -0.3830396
## Lag 5000                                   0.1372333
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                                     -0.9629330
## Lag 100                                    -0.8857980
## Lag 500                                    -0.6314814
## Lag 1000                                   -0.3830396
## Lag 5000                                   0.1372333
##           at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal

```

```

## Lag 0                                -0.9140715
## Lag 100                             -0.8566927
## Lag 500                             -0.6364708
## Lag 1000                            -0.3975119
## Lag 5000                            0.1452190
## traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0      0.03515002    -0.04917083   -0.03778492   0.16291072
## Lag 100     0.03091121    -0.05304287   -0.03754739   0.16477625
## Lag 500     0.02594763    -0.04411058   -0.04091176   0.15960207
## Lag 1000    0.03943697    -0.05730659   -0.07505811   0.12634662
## Lag 5000    -0.02494957   -0.03580137   -0.02083442  -0.01955372
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.04169552
## Lag 100                             -0.02400995
## Lag 500                             -0.05483846
## Lag 1000                            0.01413904
## Lag 5000                            0.12137475
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.05557720
## Lag 100                             -0.04292481
## Lag 500                             -0.04373833
## Lag 1000                            0.01820822
## Lag 5000                            0.13061028
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                                -0.05557720
## Lag 100                             -0.04292481
## Lag 500                             -0.04373833
## Lag 1000                            0.01820822
## Lag 5000                            0.13061028
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0                                -0.03042350
## Lag 100                             -0.02364770
## Lag 500                             -0.02090983
## Lag 1000                            0.03823823
## Lag 5000                            0.15598999
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0                                0.9162648
## Lag 100                             0.7742948
## Lag 500                             0.5235576
## Lag 1000                            0.3032634
## Lag 5000                            -0.1115658
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0                                1.0000000
## Lag 100                             0.8752177
## Lag 500                             0.6156893
## Lag 1000                            0.3715418
## Lag 5000                            -0.1435973

```

```

##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units
## Lag 0                               1.0000000
## Lag 100                            0.8752177
## Lag 500                            0.6156893
## Lag 1000                           0.3715418
## Lag 5000                           -0.1435973
##           at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units
## Lag 0                               0.9475981
## Lag 100                            0.8522663
## Lag 500                            0.6217954
## Lag 1000                           0.3896947
## Lag 5000                           -0.1540929
##
## , , at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units
##
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0                               0.066793402
## Lag 100                            0.050378446
## Lag 500                            0.077515159
## Lag 1000                           0.008812794
## Lag 5000                           -0.121465553
##           at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0                               0.093083697
## Lag 100                            0.084384646
## Lag 500                            0.074812950
## Lag 1000                           -0.007598499
## Lag 5000                           -0.145198558
##           at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal
## Lag 0                               0.093083697
## Lag 100                            0.084384646
## Lag 500                            0.074812950
## Lag 1000                           -0.007598499
## Lag 5000                           -0.145198558
##           at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal
## Lag 0                               0.039633904
## Lag 100                            0.030631751
## Lag 500                            0.008216661
## Lag 1000                           -0.067574178
## Lag 5000                           -0.158681853
##           at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                               -0.8725969
## Lag 100                            -0.7772224
## Lag 500                            -0.5293889
## Lag 1000                           -0.3047292
## Lag 5000                           0.1032909
##           at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                               -0.9629330

```

```

## Lag 100          -0.8857980
## Lag 500          -0.6314814
## Lag 1000         -0.3830396
## Lag 5000          0.1372333
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0             -0.9629330
## Lag 100           -0.8857980
## Lag 500           -0.6314814
## Lag 1000          -0.3830396
## Lag 5000          0.1372333
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0             -0.9140715
## Lag 100            -0.8566927
## Lag 500            -0.6364708
## Lag 1000           -0.3975119
## Lag 5000           0.1452190
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0              0.03515002     -0.04917083    -0.03778492     0.16291072
## Lag 100             0.03091121     -0.05304287    -0.03754739     0.16477625
## Lag 500             0.02594763     -0.04411058    -0.04091176     0.15960207
## Lag 1000            0.03943697     -0.05730659    -0.07505811     0.12634662
## Lag 5000            -0.02494957    -0.03580137    -0.02083442    -0.01955372
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0               -0.04169552
## Lag 100              -0.02400995
## Lag 500              -0.05483846
## Lag 1000             0.01413904
## Lag 5000             0.12137475
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0               -0.05557720
## Lag 100              -0.04292481
## Lag 500              -0.04373833
## Lag 1000             0.01820822
## Lag 5000             0.13061028
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0               -0.05557720
## Lag 100              -0.04292481
## Lag 500              -0.04373833
## Lag 1000             0.01820822
## Lag 5000             0.13061028
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
## Lag 0               -0.03042350
## Lag 100              -0.02364770
## Lag 500              -0.02090983
## Lag 1000             0.03823823
## Lag 5000             0.15598999
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units

```

```

## Lag 0          0.9162648
## Lag 100       0.7742948
## Lag 500       0.5235576
## Lag 1000      0.3032634
## Lag 5000      -0.1115658
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0          1.0000000
## Lag 100        0.8752177
## Lag 500        0.6156893
## Lag 1000       0.3715418
## Lag 5000      -0.1435973
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
## Lag 0          1.0000000
## Lag 100        0.8752177
## Lag 500        0.6156893
## Lag 1000       0.3715418
## Lag 5000      -0.1435973
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
## Lag 0          0.9475981
## Lag 100        0.8522663
## Lag 500        0.6217954
## Lag 1000       0.3896947
## Lag 5000      -0.1540929
##
## , , at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
##
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
## Lag 0          0.047978981
## Lag 100        0.045079451
## Lag 500        0.067970538
## Lag 1000       -0.006909171
## Lag 5000      -0.132371879
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.animal
## Lag 0          0.08843633
## Lag 100        0.08879180
## Lag 500        0.08319912
## Lag 1000       -0.00586568
## Lag 5000      -0.15675135
## at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.animal
## Lag 0          0.08843633
## Lag 100        0.08879180
## Lag 500        0.08319912
## Lag 1000       -0.00586568
## Lag 5000      -0.15675135
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.animal
## Lag 0          0.04584246
## Lag 100        0.04189200

```

```

## Lag 500                               0.02195919
## Lag 1000                             -0.05747491
## Lag 5000                             -0.16717974
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
## Lag 0                                -0.75254810
## Lag 100                              -0.68626575
## Lag 500                              -0.50089775
## Lag 1000                             -0.29031256
## Lag 5000                             0.09258748
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
## Lag 0                                -0.9191068
## Lag 100                             -0.8549977
## Lag 500                             -0.6281177
## Lag 1000                            -0.3837276
## Lag 5000                            0.1299735
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
## Lag 0                                -0.9191068
## Lag 100                             -0.8549977
## Lag 500                             -0.6281177
## Lag 1000                            -0.3837276
## Lag 5000                            0.1299735
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
## Lag 0                                -0.9650519
## Lag 100                             -0.9020388
## Lag 500                             -0.6603730
## Lag 1000                            -0.4085407
## Lag 5000                            0.1451290
##      traitbwt.byear traittarsus.byear traitbwt.mother traittarsus.mother
## Lag 0       0.03233249    -0.05392947   -0.04048151   0.17073426
## Lag 100     0.03971497    -0.04863212   -0.04603016   0.19100707
## Lag 500     0.02833531    -0.02831174   -0.04356222   0.17180604
## Lag 1000    0.03018475    -0.05344380   -0.07230573   0.12082866
## Lag 5000    -0.04000580   -0.02623878   -0.01322038  -0.01980137
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.02086631
## Lag 100                             -0.02091564
## Lag 500                             -0.04738890
## Lag 1000                            0.02308918
## Lag 5000                            0.12589958
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
## Lag 0                                -0.04717349
## Lag 100                             -0.04819145
## Lag 500                             -0.05369332
## Lag 1000                            0.02184183
## Lag 5000                            0.13760532
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
## Lag 0                                -0.04717349

```

```

## Lag 100           -0.04819145
## Lag 500           -0.05369332
## Lag 1000          0.02184183
## Lag 5000          0.13760532
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.units
## Lag 0             -0.03748483
## Lag 100           -0.03257654
## Lag 500           -0.03463265
## Lag 1000          0.04295442
## Lag 5000          0.15951258
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
## Lag 0              0.7818908
## Lag 100            0.6906357
## Lag 500            0.4994770
## Lag 1000           0.2932880
## Lag 5000           -0.1017144
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traitbwt.units
## Lag 0              0.9475981
## Lag 100             0.8462154
## Lag 500             0.6150082
## Lag 1000            0.3748486
## Lag 5000            -0.1355925
## at.level(sex, "2"):traitbwt:at.level(sex, "2"):traiittarsus.units
## Lag 0              0.9475981
## Lag 100             0.8462154
## Lag 500             0.6150082
## Lag 1000            0.3748486
## Lag 5000            -0.1355925
## at.level(sex, "2"):traiittarsus:at.level(sex, "2"):traiittarsus.units
## Lag 0              1.0000000
## Lag 100             0.8919817
## Lag 500             0.6480975
## Lag 1000            0.4037684
## Lag 5000            -0.1516651

```

As before we can obtain the raw variance component estimates and genetic correlations for the random effects:

```
posterior.mode(model2.4$VCV)
```

```

##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.animal
##                                         0.9669729
##      at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traitbwt.animal
##                                         0.1859324
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traiittarsus.animal
##                                         0.1859324
## at.level(sex, "1"):traiittarsus:at.level(sex, "1"):traiittarsus.animal

```

```

##                                1.9958915
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.animal
##                                         0.8390295
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.animal
##                                         -0.1015747
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.animal
##                                         -0.1015747
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.animal
##                                         0.6444978
##                                         traitbwt.byear
##                                         0.8339224
##                                         traittarsus.byear
##                                         2.9032045
##                                         traitbwt.mother
##                                         1.8371233
##                                         traittarsus.mother
##                                         6.4835733
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traitbwt.units
##                                         2.1384066
##      at.level(sex, "1"):traittarsus:at.level(sex, "1"):traitbwt.units
##                                         4.8421570
##      at.level(sex, "1"):traitbwt:at.level(sex, "1"):traittarsus.units
##                                         4.8421570
## at.level(sex, "1"):traittarsus:at.level(sex, "1"):traittarsus.units
##                                         13.0119702
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traitbwt.units
##                                         2.1428481
##      at.level(sex, "2"):traittarsus:at.level(sex, "2"):traitbwt.units
##                                         6.0868703
##      at.level(sex, "2"):traitbwt:at.level(sex, "2"):traittarsus.units
##                                         6.0868703
## at.level(sex, "2"):traittarsus:at.level(sex, "2"):traittarsus.units
##                                         18.3836767

#
genetic.correlation2.4.F <- model2.4$VCV[, "at.level(sex, \"1\"):traittarsus:at.level(sex,
residual.correlation2.4.F <- model2.4$VCV[, "at.level(sex, \"1\"):traittarsus:at.level(sex
genetic.correlation2.4.M <- model2.4$VCV[, "at.level(sex, \"2\"):traittarsus:at.level(sex,
residual.correlation2.4.M <- model2.4$VCV[, "at.level(sex, \"2\"):traittarsus:at.level(sex
#
posterior.mode(genetic.correlation2.4.F)

##      var1
## 0.7852385

```

```
posterior.mode(residual.correlation2.4.F)
```

```
##      var1
## 0.8738073
```

```
posterior.mode(genetic.correlation2.4.M)
```

```
##      var1
## 0.9166822
```

```
posterior.mode(residual.correlation2.4.M)
```

```
##      var1
## 0.8658358
```

Evaluation of the statistical support for these sex-specific correlations is straightforward. Because we imposed no constraint on their estimation, we can evaluate the extent to which the posterior distributions overlap zero or overlap each other:

```
HPDinterval(genetic.correlation2.4.F, 0.95)
```

```
##           lower      upper
## var1 -0.3742433 0.963575
## attr(,"Probability")
## [1] 0.95
```

```
HPDinterval(genetic.correlation2.4.M, 0.95)
```

```
##           lower      upper
## var1 -0.3275444 0.9801895
## attr(,"Probability")
## [1] 0.95
```

```
HPDinterval(residual.correlation2.4.F, 0.95)
```

```
##           lower      upper
## var1 0.698292 0.962472
## attr(,"Probability")
## [1] 0.95
```

```
HPDinterval(residual.correlation2.4.M, 0.95)
```

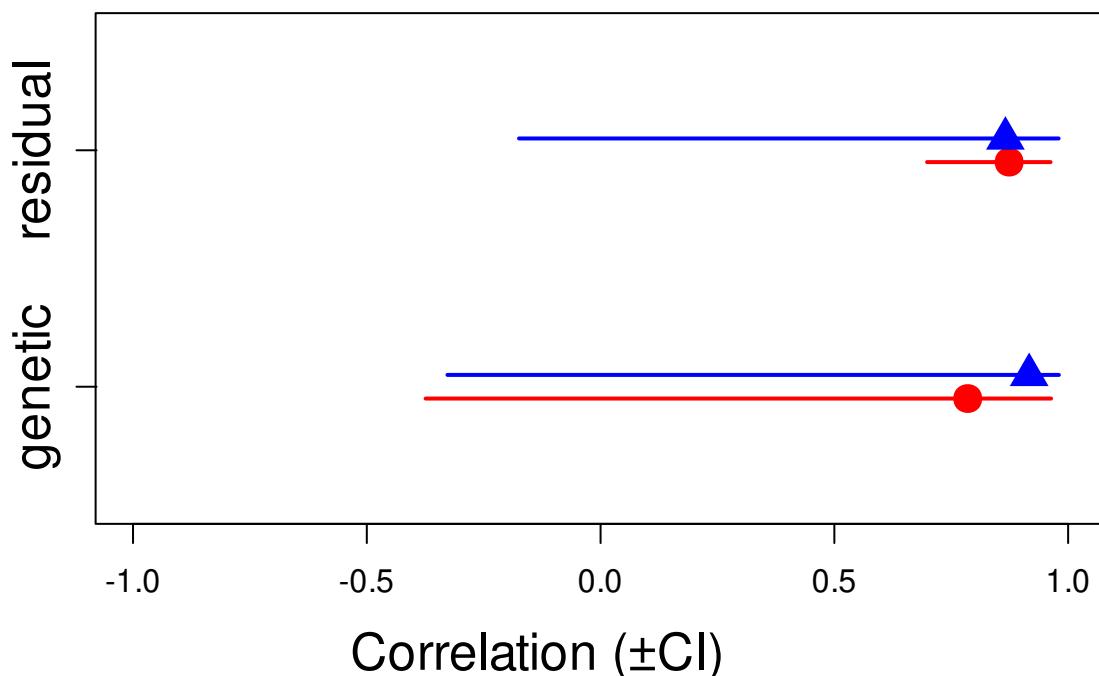
```
##           lower      upper
```

```
## var1 -0.1744922 0.9796736
## attr(,"Probability")
## [1] 0.95
```

Here a plot to visualize the overlaps of covariances.

```
cor.est <- rbind(
  cbind(posterior.mode(genetic.correlation2.4.F, ), HPDinterval(genetic.correlation2.4.F,
  cbind(posterior.mode(genetic.correlation2.4.M), HPDinterval(genetic.correlation2.4.M, 0,
  cbind(posterior.mode(residual.correlation2.4.F, ), HPDinterval(residual.correlation2.4.F,
  cbind(posterior.mode(residual.correlation2.4.M), HPDinterval(residual.correlation2.4.M,
))

plot(c(0.95, 1.05, 1.95, 2.05) ~ cor.est[, 1], xlim = c(-1, 1), ylim = c(0.5, 2.5), xlab =
arrows(y0 = 0.95, x0 = cor.est[1, 2], y1 = 0.95, x1 = cor.est[1, 3], code = 3, angle = 90,
arrows(y0 = 1.05, x0 = cor.est[2, 2], y1 = 1.05, x1 = cor.est[2, 3], code = 3, angle = 90,
arrows(y0 = 1.95, x0 = cor.est[3, 2], y1 = 1.95, x1 = cor.est[3, 3], code = 3, angle = 90,
arrows(y0 = 2.05, x0 = cor.est[4, 2], y1 = 2.05, x1 = cor.est[4, 3], code = 3, angle = 90,
mtext("Correlation (\u00b1CI)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6)
axis(2, at = 1, labels = c("genetic"), las = 3, cex.axis = 1.6)
axis(2, at = 2, labels = c("residual"), las = 3, cex.axis = 1.6)
```



These posterior distributions overlap between each other, which suggested the correlation were not

significantly different between sexes.

By using corgh instead of us, we can extract the BLUPs and plot the sex-specific correlation.

```
gryphon2$T1 <- gryphon2$bwt
gryphon2$T2 <- gryphon2$tarsus
#
model2.5 <- MCMCglmm(cbind(T1, T2) ~ trait - 1 + trait:sex,
  random = ~ corgh(at.level(sex, "1"):trait):animal + corgh(at.level(sex, "2"):trait):animal,
  rcov = ~ us(at.level(sex, "1"):trait):units + us(at.level(sex, "2"):trait):units,
  family = c("gaussian", "gaussian"),
  ginv = list(animal = Ainv), data = gryphon2,
  nitt = 130000, thin = 100, burnin = 30000,
  prior = prior2.3, verbose = FALSE, pr = TRUE,
)
save(model2.5, file = "data/MCMCglmm_model2_5_LongRun.rda")
```

Again we have provided the data from one such run. It can be accessed using the code:

```
load(file = "data/MCMCglmm_model2_5_LongRun.rda")
summary(model2.5)
```

```

## Iterations = 30001:129901
## Thinning interval = 100
## Sample size = 1000
##
## DIC: 6271.158
##
## G-structure: ~corgh(at.level(sex, "1"):trait):animal
##
## post.mean l-95% CI
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal      1      1
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal      1      1
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal      1      1
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal      1      1
##
## u-95% CI eff.samp
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal      1      0.0
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal      1     139.2
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal      1     139.2
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal      1      0.0
##
## ~corgh(at.level(sex, "2"):trait):animal
##
## post.mean l-95% CI

```

```

## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal      1      1
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal      1      1
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal      1      1
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal      1      1
##
##                                     u-95% CI eff.samp
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal      1      0.0
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal      1    117.2
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal      1    117.2
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal      1      0.0
##
##           ~idh(trait):byear
##
##           post.mean l-95% CI u-95% CI eff.samp
## traitT1.byear   0.9381   0.4298   1.542    1309
## traitT2.byear   3.9257   1.7247   6.792    1000
##
##           ~idh(trait):mother
##
##           post.mean l-95% CI u-95% CI eff.samp
## traitT1.mother  1.906    1.405    2.353    1000
## traitT2.mother  7.756    5.064    10.104   1000
##
## R-structure: ~us(at.level(sex, "1"):trait):units
##
##                                     post.mean l-95% CI
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units   2.138   1.538
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units   4.599   3.264
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units   4.599   3.264
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units  15.897  12.695
##
##                                     u-95% CI eff.samp
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units   2.724   1010
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units   5.762   1000
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units   5.762   1000
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units  19.557   1000
##
##           ~us(at.level(sex, "2"):trait):units
##
##                                     post.mean l-95% CI
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units   2.304   1.705
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units   5.578   4.305
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units   5.578   4.305
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units  20.399  16.622
##
##                                     u-95% CI eff.samp
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units   2.860   1000
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units   6.909  1099
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units   6.909  1099
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units  24.475   1000

```

```
##
## Location effects: cbind(T1, T2) ~ trait - 1 + trait:sex
##
##          post.mean l-95% CI u-95% CI eff.samp pMCMC
## traitT1      6.31127  5.82475  6.73755     1138 <0.001 ***
## traitT2     20.51915 19.62675 21.52179     1000 <0.001 ***
## traitT1:sex2  2.01404  1.61651  2.40538     1000 <0.001 ***
## traitT2:sex2  0.07302 -0.73542  0.84480      825  0.862
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
autocorr(model2.5$VCV)
```

```
## , , at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
##
##          at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0                               1.000000000
## Lag 100                            -0.001391944
## Lag 500                            -0.001397506
## Lag 1000                           -0.001404458
## Lag 5000                           -0.001983576
##
##          at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0                               -0.004648432
## Lag 100                            -0.004565467
## Lag 500                            -0.005210530
## Lag 1000                           -0.011578289
## Lag 5000                           -0.005168617
##
##          at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0                               -0.004648432
## Lag 100                            -0.004565467
## Lag 500                            -0.005210530
## Lag 1000                           -0.011578289
## Lag 5000                           -0.005168617
##
##          at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                               1.000000000
## Lag 100                            -0.001391944
## Lag 500                            -0.001397506
## Lag 1000                           -0.001404458
## Lag 5000                           -0.001983576
##
##          at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                               -0.0006350552
## Lag 100                            -0.0006356902
## Lag 500                            -0.0006382304
## Lag 1000                           -0.0010042903
## Lag 5000                           -0.0010296922
##
##          at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
```

```

## Lag 0                                -0.001456198
## Lag 100                             -0.001077247
## Lag 500                             -0.002490439
## Lag 1000                            -0.005282054
## Lag 5000                            -0.006612499
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0                                -0.001456198
## Lag 100                             -0.001077247
## Lag 500                             -0.002490439
## Lag 1000                            -0.005282054
## Lag 5000                            -0.006612499
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                                -0.0006350554
## Lag 100                             -0.0006356904
## Lag 500                             -0.0006382306
## Lag 1000                            -0.0010042906
## Lag 5000                            -0.0010296924
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0        0.01795130   0.009525697   0.01165057  -0.022459989
## Lag 100     -0.02230357   0.018537495   0.01908830  -0.004395967
## Lag 500      0.01133334   0.034563646   0.02966327  -0.024803073
## Lag 1000     0.02568159  -0.072909756   0.01174105   0.008341327
## Lag 5000    -0.00165152  -0.040694308   -0.01713262  -0.022155338
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                               0.022762565
## Lag 100                            0.013550715
## Lag 500                            0.023394975
## Lag 1000                           0.003721016
## Lag 5000                           -0.019698113
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                               0.022365674
## Lag 100                            0.041651846
## Lag 500                            0.012710980
## Lag 1000                           -0.011480840
## Lag 5000                           -0.001466941
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0                               0.022365674
## Lag 100                            0.041651846
## Lag 500                            0.012710980
## Lag 1000                           -0.011480840
## Lag 5000                           -0.001466941
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0                               -0.010314628
## Lag 100                            0.055525653
## Lag 500                            -0.003105893
## Lag 1000                           -0.025454553
## Lag 5000                           0.008550330

```

```

##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0                               0.041893462
## Lag 100                            -0.005652676
## Lag 500                            0.062306138
## Lag 1000                           0.021660306
## Lag 5000                           -0.009682721
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0                               0.0674542988
## Lag 100                            -0.0276604979
## Lag 500                            0.0278254174
## Lag 1000                           0.0335689393
## Lag 5000                           0.0008595584
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0                               0.0674542988
## Lag 100                            -0.0276604979
## Lag 500                            0.0278254174
## Lag 1000                           0.0335689393
## Lag 5000                           0.0008595584
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0                               0.081823186
## Lag 100                            -0.039860699
## Lag 500                            0.008092171
## Lag 1000                           0.008412668
## Lag 5000                           0.021467565
##
## , , at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
##
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0                               -0.004648432
## Lag 100                            -0.004625049
## Lag 500                            -0.003531126
## Lag 1000                           -0.004439099
## Lag 5000                           -0.006278871
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0                               1.000000000
## Lag 100                            0.723181023
## Lag 500                            0.292398811
## Lag 1000                           0.092837234
## Lag 5000                           -0.009749071
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0                               1.000000000
## Lag 100                            0.723181023
## Lag 500                            0.292398811
## Lag 1000                           0.092837234
## Lag 5000                           -0.009749071
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                               -0.004648437

```

```

## Lag 100                                -0.004625085
## Lag 500                                -0.003531141
## Lag 1000                               -0.004439104
## Lag 5000                               -0.006278871
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                                 -0.002312162
## Lag 100                               -0.002217551
## Lag 500                               -0.001837789
## Lag 1000                              -0.002977420
## Lag 5000                              -0.004726772
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0                                 -0.010685598
## Lag 100                               -0.008503067
## Lag 500                               -0.007387620
## Lag 1000                              -0.012150586
## Lag 5000                              -0.019200093
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0                                 -0.010685598
## Lag 100                               -0.008503067
## Lag 500                               -0.007387620
## Lag 1000                              -0.012150586
## Lag 5000                              -0.019200093
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                                 -0.002312162
## Lag 100                               -0.002217551
## Lag 500                               -0.001837789
## Lag 1000                              -0.002977420
## Lag 5000                              -0.004726772
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0       0.03206603   0.001388092   -0.01597305   -0.012121778
## Lag 100     0.02062954  -0.034683825    0.01430611   -0.050389774
## Lag 500     0.01419036   0.020814008    0.03068981   -0.018472864
## Lag 1000    0.02272278  -0.003124614   -0.03666507   0.008391569
## Lag 5000    -0.01889532  -0.003299003   -0.03161363   0.042393770
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                                -0.022308219
## Lag 100                               -0.004067301
## Lag 500                               -0.004473470
## Lag 1000                              0.048582961
## Lag 5000                              0.001202075
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                                -0.015759766
## Lag 100                               0.017032779
## Lag 500                               -0.003108186
## Lag 1000                              0.042126244
## Lag 5000                              -0.016955230
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units

```

```

## Lag 0           -0.015759766
## Lag 100        0.017032779
## Lag 500        -0.003108186
## Lag 1000       0.042126244
## Lag 5000       -0.016955230
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0           -0.008530358
## Lag 100         0.019755361
## Lag 500        -0.013140805
## Lag 1000       0.013898534
## Lag 5000       -0.033469898
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0           0.005815507
## Lag 100         0.001279903
## Lag 500        -0.004227436
## Lag 1000       -0.037409946
## Lag 5000       0.027223985
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0           -0.028111911
## Lag 100         -0.015823870
## Lag 500         -0.002146573
## Lag 1000        -0.016376141
## Lag 5000        0.038876537
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0           -0.028111911
## Lag 100         -0.015823870
## Lag 500         -0.002146573
## Lag 1000        -0.016376141
## Lag 5000        0.038876537
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0           -0.042456056
## Lag 100         -0.010621969
## Lag 500         0.009797485
## Lag 1000        -0.019305202
## Lag 5000        0.033404201
##
## , , at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0           -0.004648432
## Lag 100        -0.004625049
## Lag 500        -0.003531126
## Lag 1000       -0.004439099
## Lag 5000       -0.006278871
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0           1.000000000
## Lag 100        0.723181023

```

```

## Lag 500          0.292398811
## Lag 1000         0.092837234
## Lag 5000        -0.009749071
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0            1.000000000
## Lag 100          0.723181023
## Lag 500          0.292398811
## Lag 1000         0.092837234
## Lag 5000        -0.009749071
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0           -0.004648437
## Lag 100          -0.004625085
## Lag 500          -0.003531141
## Lag 1000         -0.004439104
## Lag 5000         -0.006278871
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0           -0.002312162
## Lag 100          -0.002217551
## Lag 500          -0.001837789
## Lag 1000         -0.002977420
## Lag 5000         -0.004726772
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0           -0.010685598
## Lag 100          -0.008503067
## Lag 500          -0.007387620
## Lag 1000         -0.012150586
## Lag 5000         -0.019200093
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0           -0.010685598
## Lag 100          -0.008503067
## Lag 500          -0.007387620
## Lag 1000         -0.012150586
## Lag 5000         -0.019200093
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0           -0.002312162
## Lag 100          -0.002217551
## Lag 500          -0.001837789
## Lag 1000         -0.002977420
## Lag 5000         -0.004726772
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0           0.03206603  0.001388092  -0.01597305  -0.012121778
## Lag 100          0.02062954 -0.034683825   0.01430611  -0.050389774
## Lag 500          0.01419036  0.020814008   0.03068981  -0.018472864
## Lag 1000         0.02272278 -0.003124614   -0.03666507  0.008391569
## Lag 5000         -0.01889532 -0.003299003   -0.03161363  0.042393770
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0           -0.022308219

```

```

## Lag 100           -0.004067301
## Lag 500           -0.004473470
## Lag 1000          0.048582961
## Lag 5000          0.001202075
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0             -0.015759766
## Lag 100           0.017032779
## Lag 500           -0.003108186
## Lag 1000          0.042126244
## Lag 5000          -0.016955230
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0             -0.015759766
## Lag 100           0.017032779
## Lag 500           -0.003108186
## Lag 1000          0.042126244
## Lag 5000          -0.016955230
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0             -0.008530358
## Lag 100           0.019755361
## Lag 500           -0.013140805
## Lag 1000          0.013898534
## Lag 5000          -0.033469898
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0             0.005815507
## Lag 100           0.001279903
## Lag 500           -0.004227436
## Lag 1000          -0.037409946
## Lag 5000          0.027223985
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0             -0.028111911
## Lag 100           -0.015823870
## Lag 500           -0.002146573
## Lag 1000          -0.016376141
## Lag 5000          0.038876537
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0             -0.028111911
## Lag 100           -0.015823870
## Lag 500           -0.002146573
## Lag 1000          -0.016376141
## Lag 5000          0.038876537
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0             -0.042456056
## Lag 100           -0.010621969
## Lag 500           0.009797485
## Lag 1000          -0.019305202
## Lag 5000          0.033404201
##

```

```
## , , at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0                               1.000000000
## Lag 100                            -0.001391944
## Lag 500                            -0.001397506
## Lag 1000                           -0.001404458
## Lag 5000                           -0.001983576
##
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0                               -0.004648437
## Lag 100                            -0.004565469
## Lag 500                            -0.005210530
## Lag 1000                           -0.011578289
## Lag 5000                           -0.005168617
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0                               -0.004648437
## Lag 100                            -0.004565469
## Lag 500                            -0.005210530
## Lag 1000                           -0.011578289
## Lag 5000                           -0.005168617
##
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                               1.000000000
## Lag 100                            -0.001391944
## Lag 500                            -0.001397506
## Lag 1000                           -0.001404458
## Lag 5000                           -0.001983576
##
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                               -0.0006350551
## Lag 100                            -0.0006356902
## Lag 500                            -0.0006382304
## Lag 1000                           -0.0010042903
## Lag 5000                           -0.0010296921
##
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0                               -0.001456198
## Lag 100                            -0.001077247
## Lag 500                            -0.002490439
## Lag 1000                           -0.005282054
## Lag 5000                           -0.006612499
##
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0                               -0.001456198
## Lag 100                            -0.001077247
## Lag 500                            -0.002490439
## Lag 1000                           -0.005282054
## Lag 5000                           -0.006612499
##
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                               -0.0006350554
## Lag 100                            -0.0006356904
```

```

## Lag 500                               -0.0006382306
## Lag 1000                             -0.0010042906
## Lag 5000                            -0.0010296924
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0       0.017951297   0.009525703   0.01165057  -0.022459984
## Lag 100    -0.022303567   0.018537496   0.01908830  -0.004395967
## Lag 500     0.011333343   0.034563646   0.02966327  -0.024803074
## Lag 1000    0.025681592  -0.072909754   0.01174105   0.008341327
## Lag 5000   -0.001651524  -0.040694308  -0.01713262  -0.022155336
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                                0.022762562
## Lag 100                              0.013550715
## Lag 500                              0.023394973
## Lag 1000                             0.003721019
## Lag 5000                            -0.019698116
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                                0.022365669
## Lag 100                             0.041651846
## Lag 500                             0.012710977
## Lag 1000                            -0.011480840
## Lag 5000                            -0.001466943
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0                                0.022365669
## Lag 100                             0.041651846
## Lag 500                             0.012710977
## Lag 1000                            -0.011480840
## Lag 5000                            -0.001466943
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0                                -0.010314633
## Lag 100                              0.055525653
## Lag 500                              -0.003105895
## Lag 1000                            -0.025454554
## Lag 5000                            0.008550329
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0                                0.041893463
## Lag 100                             -0.005652675
## Lag 500                              0.062306141
## Lag 1000                            0.021660305
## Lag 5000                            -0.009682725
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0                                0.0674542971
## Lag 100                             -0.0276604962
## Lag 500                              0.0278254213
## Lag 1000                            0.0335689362
## Lag 5000                            0.0008595534
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0                                0.0674542971

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## Lag 100          -0.0276604962
## Lag 500          0.0278254213
## Lag 1000         0.0335689362
## Lag 5000         0.0008595534
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0            0.081823182
## Lag 100          -0.039860695
## Lag 500          0.008092176
## Lag 1000         0.008412663
## Lag 5000         0.021467560
##
## , , at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.0006350552
## Lag 100          -0.0002728055
## Lag 500          -0.0002753457
## Lag 1000         -0.0002785209
## Lag 5000         -0.0005430012
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.0023121624
## Lag 100          -0.0018131603
## Lag 500          -0.0021171712
## Lag 1000         -0.0002636656
## Lag 5000         -0.0007281379
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0            -0.0023121624
## Lag 100          -0.0018131603
## Lag 500          -0.0021171712
## Lag 1000         -0.0002636656
## Lag 5000         -0.0007281379
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0            -0.0006350551
## Lag 100          -0.0002728055
## Lag 500          -0.0002753457
## Lag 1000         -0.0002785209
## Lag 5000         -0.0005430012
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0            1.0000000000
## Lag 100          -0.0001245872
## Lag 500          0.0472252652
## Lag 1000         -0.0002929252
## Lag 5000         -0.0003045261
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0            -0.0030464876
## Lag 100          -0.0012272569
## Lag 500          -0.0012428417

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## Lag 1000                               -0.0009101059
## Lag 5000                               -0.0098179074
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0                                    -0.0030464876
## Lag 100                                 -0.0012272569
## Lag 500                                 -0.0012428417
## Lag 1000                                -0.0009101059
## Lag 5000                                -0.0098179074
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                                    1.0000000000
## Lag 100                                 -0.0001245873
## Lag 500                                 0.0472252675
## Lag 1000                                -0.0002929253
## Lag 5000                                -0.0003045262
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0        0.007859897  -0.048647022   0.014169960  -0.01277587
## Lag 100     -0.031686328   0.044252052   0.001418991  -0.00470343
## Lag 500     -0.006041444  -0.003672647   0.026635103  -0.05216309
## Lag 1000    0.040389650   0.001576320  -0.008599945  -0.03098371
## Lag 5000    -0.014128580  -0.030023647   0.016865943  -0.08562409
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0          0.019317584
## Lag 100        -0.006883423
## Lag 500        0.005456272
## Lag 1000       -0.040428229
## Lag 5000       -0.019377516
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0          0.002652222
## Lag 100        -0.016499071
## Lag 500        -0.004463759
## Lag 1000       -0.044909299
## Lag 5000       -0.009389202
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0          0.002652222
## Lag 100        -0.016499071
## Lag 500        -0.004463759
## Lag 1000       -0.044909299
## Lag 5000       -0.009389202
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0          -0.002054456
## Lag 100         0.002237526
## Lag 500         0.013746887
## Lag 1000        -0.048289160
## Lag 5000         0.014216318
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0          -0.01258769
## Lag 100         0.01872011

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## Lag 500          0.01882522
## Lag 1000         0.03756773
## Lag 5000        -0.02411450
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0            -0.005906983
## Lag 100          0.013146223
## Lag 500          0.043916861
## Lag 1000         0.033507898
## Lag 5000        -0.008078930
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0            -0.005906983
## Lag 100          0.013146223
## Lag 500          0.043916861
## Lag 1000         0.033507898
## Lag 5000        -0.008078930
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0            0.0006335032
## Lag 100          0.0037825580
## Lag 500          0.0543707513
## Lag 1000         0.0360053763
## Lag 5000        -0.0061125594
##
## , , at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.001456198
## Lag 100          0.001715826
## Lag 500          -0.003740235
## Lag 1000         -0.005014592
## Lag 5000        -0.010487126
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.01068560
## Lag 100          -0.01235479
## Lag 500          -0.01399016
## Lag 1000         -0.01676244
## Lag 5000        -0.01431417
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0            -0.01068560
## Lag 100          -0.01235479
## Lag 500          -0.01399016
## Lag 1000         -0.01676244
## Lag 5000        -0.01431417
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0            -0.001456198
## Lag 100          0.001715826
## Lag 500          -0.003740235
## Lag 1000         -0.005014592

```

```

## Lag 5000 -0.010487126
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal -0.003046488
## Lag 0 -0.003153963
## Lag 100 -0.002352225
## Lag 500 -0.003782156
## Lag 1000 -0.002857518
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal 1.000000000
## Lag 0 0.83058654
## Lag 100 0.23387736
## Lag 500 0.12487172
## Lag 1000 -0.02765118
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal 1.000000000
## Lag 0 0.83058654
## Lag 100 0.23387736
## Lag 500 0.12487172
## Lag 1000 -0.02765118
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal -0.003046486
## Lag 0 -0.003153962
## Lag 100 -0.002352224
## Lag 500 -0.003782157
## Lag 1000 -0.002857519
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0 -0.013526377 -0.01774985 -0.013841217 0.022701877
## Lag 100 0.004093624 -0.01386591 0.001219680 0.015022006
## Lag 500 0.021634587 0.01429551 -0.019234622 0.022081044
## Lag 1000 -0.020668120 0.05928732 -0.059492408 0.033606722
## Lag 5000 -0.009719164 0.04539140 -0.005427111 -0.008612086
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units -0.024077943
## Lag 0 -0.002203887
## Lag 100 0.010909995
## Lag 500 0.018505117
## Lag 1000 -0.065135988
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units -0.0072931648
## Lag 0 0.0003911575
## Lag 100 0.0164572902
## Lag 500 0.0161183049
## Lag 1000 -0.0518609896
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units -0.0072931648
## Lag 0 0.0003911575
## Lag 100 0.0164572902

```

```

## Lag 1000          0.0161183049
## Lag 5000         -0.0518609896
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0             0.002102035
## Lag 100           0.002516452
## Lag 500           0.016614009
## Lag 1000          0.009102492
## Lag 5000          -0.057713146
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0             0.003830731
## Lag 100           -0.005606418
## Lag 500           0.005332862
## Lag 1000          0.009183940
## Lag 5000          -0.021145579
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0             0.001171218
## Lag 100           -0.012536100
## Lag 500           0.002347666
## Lag 1000          0.012765703
## Lag 5000          -0.035720789
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0             0.001171218
## Lag 100           -0.012536100
## Lag 500           0.002347666
## Lag 1000          0.012765703
## Lag 5000          -0.035720789
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0             0.011573448
## Lag 100           0.010597731
## Lag 500           0.009234753
## Lag 1000          0.017510534
## Lag 5000          -0.019205136
##
## , , at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
##
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0             -0.001456198
## Lag 100           0.001715826
## Lag 500           -0.003740235
## Lag 1000          -0.005014592
## Lag 5000          -0.010487126
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0             -0.01068560
## Lag 100           -0.01235479
## Lag 500           -0.01399016
## Lag 1000          -0.01676244
## Lag 5000          -0.01431417

```

```

##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0                           -0.01068560
## Lag 100                          -0.01235479
## Lag 500                          -0.01399016
## Lag 1000                         -0.01676244
## Lag 5000                         -0.01431417
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                           -0.001456198
## Lag 100                          0.001715826
## Lag 500                          -0.003740235
## Lag 1000                         -0.005014592
## Lag 5000                         -0.010487126
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                           -0.003046488
## Lag 100                          -0.003153963
## Lag 500                          -0.002352225
## Lag 1000                         -0.003782156
## Lag 5000                         -0.002857518
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0                           1.000000000
## Lag 100                          0.83058654
## Lag 500                          0.23387736
## Lag 1000                         0.12487172
## Lag 5000                         -0.02765118
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0                           1.000000000
## Lag 100                          0.83058654
## Lag 500                          0.23387736
## Lag 1000                         0.12487172
## Lag 5000                         -0.02765118
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                           -0.003046486
## Lag 100                          -0.003153962
## Lag 500                          -0.002352224
## Lag 1000                         -0.003782157
## Lag 5000                         -0.002857519
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0     -0.013526377   -0.01774985   -0.013841217    0.022701877
## Lag 100    0.004093624   -0.01386591    0.001219680    0.015022006
## Lag 500    0.021634587    0.01429551   -0.019234622    0.022081044
## Lag 1000   -0.020668120    0.05928732   -0.059492408    0.033606722
## Lag 5000   -0.009719164    0.04539140   -0.005427111   -0.008612086
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                           -0.024077943
## Lag 100                          -0.002203887
## Lag 500                          0.010909995
## Lag 1000                         0.018505117

```

```

## Lag 5000                               -0.065135988
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units   -0.0072931648
## Lag 0                                    0.0003911575
## Lag 100                                 0.0164572902
## Lag 500                                 0.0161183049
## Lag 1000                                -0.0518609896
## Lag 5000

## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units   -0.0072931648
## Lag 0                                    0.0003911575
## Lag 100                                 0.0164572902
## Lag 500                                 0.0161183049
## Lag 1000                                -0.0518609896
## Lag 5000

## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units   0.002102035
## Lag 0                                    0.002516452
## Lag 100                                 0.016614009
## Lag 500                                 0.009102492
## Lag 1000                                -0.057713146
## Lag 5000

## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units   0.003830731
## Lag 0                                    -0.005606418
## Lag 100                                 0.005332862
## Lag 500                                 0.009183940
## Lag 1000                                -0.021145579
## Lag 5000

## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units   0.001171218
## Lag 0                                    -0.012536100
## Lag 100                                 0.002347666
## Lag 500                                 0.012765703
## Lag 1000                                -0.035720789
## Lag 5000

## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units   0.001171218
## Lag 0                                    -0.012536100
## Lag 100                                 0.002347666
## Lag 500                                 0.012765703
## Lag 1000                                -0.035720789
## Lag 5000

## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units   0.011573448
## Lag 0                                    0.010597731
## Lag 100                                 0.009234753
## Lag 500                                 0.017510534
## Lag 1000                                -0.019205136
## Lag 5000

## , , at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal

```

```

## Lag 0           -0.0006350554
## Lag 100        -0.0002727894
## Lag 500        -0.0002753459
## Lag 1000       -0.0002785211
## Lag 5000       -0.0005430015
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0           -0.0023121624
## Lag 100         -0.0018131609
## Lag 500         -0.0021171719
## Lag 1000        -0.0002636663
## Lag 5000        -0.0007281383
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0           -0.0023121624
## Lag 100         -0.0018131609
## Lag 500         -0.0021171719
## Lag 1000        -0.0002636663
## Lag 5000        -0.0007281383
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0           -0.0006350554
## Lag 100         -0.0002727894
## Lag 500         -0.0002753459
## Lag 1000        -0.0002785211
## Lag 5000        -0.0005430015
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0           1.0000000000
## Lag 100         -0.0001245873
## Lag 500         0.0472252591
## Lag 1000        -0.0002929253
## Lag 5000        -0.0003045262
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0           -0.0030464862
## Lag 100         -0.0012272567
## Lag 500         -0.0012428427
## Lag 1000        -0.0009101069
## Lag 5000        -0.0098178961
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0           -0.0030464862
## Lag 100         -0.0012272567
## Lag 500         -0.0012428427
## Lag 1000        -0.0009101069
## Lag 5000        -0.0098178961
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0           1.0000000000
## Lag 100         -0.0001245874
## Lag 500         0.0472252614
## Lag 1000        -0.0002929254
## Lag 5000        -0.0003045263

```

```

##          traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0      0.007859899 -0.048647026  0.014169961 -0.012775870
## Lag 100    -0.031686329  0.044252052  0.001418992 -0.004703428
## Lag 500    -0.006041443 -0.003672647  0.026635104 -0.052163087
## Lag 1000   0.040389655  0.001576326 -0.008599948 -0.030983703
## Lag 5000   -0.014128581 -0.030023647  0.016865937 -0.085624096
##          at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                               0.019317585
## Lag 100                            -0.006883421
## Lag 500                            0.005456271
## Lag 1000                           -0.040428232
## Lag 5000                           -0.019377514
##          at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                               0.002652225
## Lag 100                            -0.016499068
## Lag 500                            -0.004463759
## Lag 1000                           -0.044909300
## Lag 5000                           -0.009389200
##          at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0                               0.002652225
## Lag 100                            -0.016499068
## Lag 500                            -0.004463759
## Lag 1000                           -0.044909300
## Lag 5000                           -0.009389200
##          at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0                               -0.002054453
## Lag 100                            0.002237526
## Lag 500                            0.013746888
## Lag 1000                           -0.048289160
## Lag 5000                           0.014216318
##          at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0                               -0.01258769
## Lag 100                            0.01872011
## Lag 500                            0.01882522
## Lag 1000                           0.03756772
## Lag 5000                           -0.02411450
##          at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0                               -0.005906985
## Lag 100                            0.013146221
## Lag 500                            0.043916862
## Lag 1000                           0.033507897
## Lag 5000                           -0.008078924
##          at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0                               -0.005906985
## Lag 100                            0.013146221
## Lag 500                            0.043916862
## Lag 1000                           0.033507897

```

```

## Lag 5000                               -0.008078924
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units      0.0006335012
## Lag 0                                    0.0037825564
## Lag 100                                 0.0543707522
## Lag 500                                 0.0360053770
## Lag 1000                                -0.0061125543
## Lag 5000

##
## , , traitT1.byear

##
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal      0.017951297
## Lag 0                                    -0.015650603
## Lag 100                                 -0.048310004
## Lag 500                                 0.006748945
## Lag 1000                                -0.024695487
## Lag 5000

## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal      0.03206603
## Lag 0                                    0.03644816
## Lag 100                                 0.02462867
## Lag 500                                 0.03134031
## Lag 1000                                -0.02813186
## Lag 5000

## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal      0.03206603
## Lag 0                                    0.03644816
## Lag 100                                 0.02462867
## Lag 500                                 0.03134031
## Lag 1000                                -0.02813186
## Lag 5000

## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal      0.017951297
## Lag 0                                    -0.015650603
## Lag 100                                 -0.048310005
## Lag 500                                 0.006748946
## Lag 1000                                -0.024695489
## Lag 5000

## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal      0.007859897
## Lag 0                                    0.008951424
## Lag 100                                 -0.075193837
## Lag 500                                 -0.056006662
## Lag 1000                                0.056756820
## Lag 5000

## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal      -0.013526377
## Lag 0                                    -0.008509234
## Lag 100                                 0.016651269
## Lag 500                                 -0.028261999
## Lag 1000                                0.036871813
## Lag 5000

## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal

```

```

## Lag 0           -0.013526377
## Lag 100        -0.008509234
## Lag 500         0.016651269
## Lag 1000        -0.028261999
## Lag 5000         0.036871813
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0           0.007859899
## Lag 100          0.008951426
## Lag 500         -0.075193836
## Lag 1000        -0.056006660
## Lag 5000         0.056756821
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0           1.0000000 -0.023340160 -0.06849446  0.027498789
## Lag 100        -0.05704935 -0.031640453  0.02105239  0.005021463
## Lag 500        -0.09276121  0.013578861 -0.02647627  0.017938010
## Lag 1000       -0.02013395 -0.042064858  0.03145873 -0.021329142
## Lag 5000       -0.04698339 -0.007786385 -0.04540802 -0.011391335
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0           0.03137990
## Lag 100          0.03791656
## Lag 500         -0.02619206
## Lag 1000        -0.05409955
## Lag 5000         -0.04538239
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0           0.0040622738
## Lag 100          0.0677383954
## Lag 500         -0.0005605853
## Lag 1000        -0.0411199654
## Lag 5000         -0.0290154960
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0           0.0040622738
## Lag 100          0.0677383954
## Lag 500         -0.0005605853
## Lag 1000        -0.0411199654
## Lag 5000         -0.0290154960
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0           -0.024177752
## Lag 100          0.080268776
## Lag 500          0.017538142
## Lag 1000        -0.040908697
## Lag 5000         -0.005206272
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0           -0.024662838
## Lag 100         -0.008402504
## Lag 500          0.095917314
## Lag 1000        -0.042564597
## Lag 5000         -0.002506247

```

```

##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0                               -0.02752719
## Lag 100                            -0.01695931
## Lag 500                             0.05741234
## Lag 1000                            -0.04606330
## Lag 5000                            0.04134386
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0                               -0.02752719
## Lag 100                            -0.01695931
## Lag 500                             0.05741234
## Lag 1000                            -0.04606330
## Lag 5000                            0.04134386
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0                               -0.008651050
## Lag 100                            -0.028966290
## Lag 500                             -0.004977943
## Lag 1000                            -0.040848683
## Lag 5000                            0.085512002
##
## , , traitT2.byear
##
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0                               0.009525697
## Lag 100                            -0.064974801
## Lag 500                             -0.028789100
## Lag 1000                            -0.043603295
## Lag 5000                            0.020618670
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0                               0.001388092
## Lag 100                            0.040474658
## Lag 500                             -0.014654139
## Lag 1000                            0.021983132
## Lag 5000                            0.038125394
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0                               0.001388092
## Lag 100                            0.040474658
## Lag 500                             -0.014654139
## Lag 1000                            0.021983132
## Lag 5000                            0.038125394
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                               0.009525703
## Lag 100                            -0.064974806
## Lag 500                             -0.028789102
## Lag 1000                            -0.043603295
## Lag 5000                            0.020618669
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                               -4.864702e-02

```

```

## Lag 100          5.788328e-02
## Lag 500          -3.017681e-05
## Lag 1000         -5.850992e-03
## Lag 5000         2.420463e-04
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0            -0.0177498454
## Lag 100          0.0063808442
## Lag 500          -0.0049424163
## Lag 1000         -0.0003719052
## Lag 5000         0.0521589142
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0            -0.0177498454
## Lag 100          0.0063808442
## Lag 500          -0.0049424163
## Lag 1000         -0.0003719052
## Lag 5000         0.0521589142
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0            -4.864703e-02
## Lag 100          5.788328e-02
## Lag 500          -3.016908e-05
## Lag 1000         -5.850992e-03
## Lag 5000         2.420437e-04
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0            -0.023340160  1.000000000  -0.040047907  0.07501106
## Lag 100           -0.025201131  0.041181362   0.026543851  -0.02763272
## Lag 500           -0.013876500  0.025109718   -0.007886112  -0.01358018
## Lag 1000          -0.001175324  0.050652913   -0.058106033  0.02511307
## Lag 5000          -0.027192208  0.002883567   -0.007135063  0.05428665
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0            -0.04513833
## Lag 100          0.04234089
## Lag 500          -0.03057247
## Lag 1000         -0.07029298
## Lag 5000         0.00168274
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0            -0.087126652
## Lag 100          0.019735995
## Lag 500          -0.036577541
## Lag 1000         -0.082722967
## Lag 5000         0.004687441
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0            -0.087126652
## Lag 100          0.019735995
## Lag 500          -0.036577541
## Lag 1000         -0.082722967
## Lag 5000         0.004687441
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units

```

```

## Lag 0           -0.123422570
## Lag 100        -0.036523126
## Lag 500        -0.009320346
## Lag 1000       -0.058354272
## Lag 5000       0.005124232
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0           0.02756337
## Lag 100         0.03558955
## Lag 500         0.00835486
## Lag 1000        -0.01369138
## Lag 5000        -0.01097255
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0           -0.02016065
## Lag 100         0.04165810
## Lag 500         0.01434291
## Lag 1000        -0.01828566
## Lag 5000        -0.02257444
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0           -0.02016065
## Lag 100         0.04165810
## Lag 500         0.01434291
## Lag 1000        -0.01828566
## Lag 5000        -0.02257444
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0           -0.05004816
## Lag 100         0.05046643
## Lag 500         -0.01284557
## Lag 1000        -0.02219803
## Lag 5000        -0.01385276
##
## , , traitT1.mother
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0           0.01165057
## Lag 100         0.08905786
## Lag 500         0.02244386
## Lag 1000        -0.03313281
## Lag 5000        0.02806631
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0           -0.01597305
## Lag 100         -0.00334208
## Lag 500         0.02847650
## Lag 1000        0.04457421
## Lag 5000        0.02068766
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0           -0.01597305
## Lag 100         -0.00334208

```

```

## Lag 500          0.02847650
## Lag 1000         0.04457421
## Lag 5000         0.02068766
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0            0.01165057
## Lag 100           0.08905786
## Lag 500           0.02244385
## Lag 1000          -0.03313282
## Lag 5000          0.02806631
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0            0.014169960
## Lag 100           0.079761436
## Lag 500           0.048204670
## Lag 1000          0.009553326
## Lag 5000          -0.004370446
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0            -0.013841217
## Lag 100           0.001743246
## Lag 500           -0.004551783
## Lag 1000          -0.020295969
## Lag 5000          0.017570665
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0            -0.013841217
## Lag 100           0.001743246
## Lag 500           -0.004551783
## Lag 1000          -0.020295969
## Lag 5000          0.017570665
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0            0.014169961
## Lag 100           0.079761439
## Lag 500           0.048204674
## Lag 1000          0.009553326
## Lag 5000          -0.004370444
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0            -0.06849446 -0.0400479067  1.000000000 -0.297139239
## Lag 100           0.01901561  0.0230734698 -0.009027192 -0.005184777
## Lag 500           -0.03709457 -0.0475231945 -0.006020676 -0.021229786
## Lag 1000          0.03016960 -0.0002406725  0.017633995  0.014194686
## Lag 5000          -0.01390477  0.0350936719 -0.001330014 -0.010498917
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0            -0.074914843
## Lag 100           -0.011337055
## Lag 500           0.019960354
## Lag 1000          0.019122566
## Lag 5000          0.006225052
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0            -0.008518268

```

```

## Lag 100           -0.016248199
## Lag 500           0.044751016
## Lag 1000          0.024827801
## Lag 5000          -0.001860750
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0             -0.008518268
## Lag 100           -0.016248199
## Lag 500           0.044751016
## Lag 1000          0.024827801
## Lag 5000          -0.001860750
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0              0.02902645
## Lag 100            -0.01253698
## Lag 500            0.05556811
## Lag 1000           0.02173176
## Lag 5000           -0.01956112
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0             -0.11230235
## Lag 100            -0.02682760
## Lag 500            0.00620533
## Lag 1000           -0.02055078
## Lag 5000           0.03081905
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0              0.011701233
## Lag 100            -0.010011619
## Lag 500            0.008428736
## Lag 1000           -0.022933341
## Lag 5000           0.023963672
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0              0.011701233
## Lag 100            -0.010011619
## Lag 500            0.008428736
## Lag 1000           -0.022933341
## Lag 5000           0.023963672
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0              0.06312437
## Lag 100            -0.01937043
## Lag 500            0.01541354
## Lag 1000           -0.02002003
## Lag 5000           0.02002943
##
## , , traitT2.mother
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0             -0.02245999
## Lag 100           0.01287933
## Lag 500           -0.06937135

```

```

## Lag 1000          -0.03408051
## Lag 5000          -0.03437527
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0              -0.01212178
## Lag 100             0.02432570
## Lag 500             0.03510369
## Lag 1000            0.02757162
## Lag 5000            -0.01614218
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0              -0.01212178
## Lag 100             0.02432570
## Lag 500             0.03510369
## Lag 1000            0.02757162
## Lag 5000            -0.01614218
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0              -0.02245998
## Lag 100             0.01287933
## Lag 500             -0.06937135
## Lag 1000            -0.03408051
## Lag 5000            -0.03437528
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0              -0.012775867
## Lag 100             0.043772797
## Lag 500             -0.011619254
## Lag 1000            -0.047193555
## Lag 5000             0.009181997
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0               0.022701877
## Lag 100             0.020020010
## Lag 500             -0.007184152
## Lag 1000            0.027822588
## Lag 5000            -0.010936045
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0               0.022701877
## Lag 100             0.020020010
## Lag 500             -0.007184152
## Lag 1000            0.027822588
## Lag 5000            -0.010936045
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0              -0.012775870
## Lag 100             0.043772795
## Lag 500             -0.011619258
## Lag 1000            -0.047193553
## Lag 5000             0.009181997
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0       0.027498789   0.075011059   -0.29713924    1.000000000
## Lag 100     -0.004330013   0.019387759    0.01135535    0.01111072

```

```

## Lag 500 -0.022534204 0.013976131 -0.03024673 0.03135127
## Lag 1000 0.018105482 -0.004480605 -0.01622036 0.01843071
## Lag 5000 -0.021285124 -0.015352704 0.01171495 -0.03266031
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0 0.010656057
## Lag 100 -0.014350075
## Lag 500 -0.025329139
## Lag 1000 -0.020213976
## Lag 5000 0.006746823
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0 -0.052640698
## Lag 100 -0.011683409
## Lag 500 -0.061476681
## Lag 1000 -0.016566750
## Lag 5000 -0.002144703
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0 -0.052640698
## Lag 100 -0.011683409
## Lag 500 -0.061476681
## Lag 1000 -0.016566750
## Lag 5000 -0.002144703
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0 -0.150664503
## Lag 100 -0.030124097
## Lag 500 -0.035301755
## Lag 1000 0.006326746
## Lag 5000 -0.034639670
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0 0.03843077
## Lag 100 0.10864998
## Lag 500 -0.01765222
## Lag 1000 -0.04802417
## Lag 5000 -0.02312820
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0 -0.07682037
## Lag 100 0.08857155
## Lag 500 -0.03270686
## Lag 1000 -0.04309358
## Lag 5000 -0.03198783
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0 -0.07682037
## Lag 100 0.08857155
## Lag 500 -0.03270686
## Lag 1000 -0.04309358
## Lag 5000 -0.03198783
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0 -0.186833159

```

```

## Lag 100          0.068802531
## Lag 500          0.010939808
## Lag 1000         -0.046597710
## Lag 5000         -0.009495369
##
## , , at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
##
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0            0.022762565
## Lag 100          0.058195840
## Lag 500          -0.006642960
## Lag 1000         -0.003238572
## Lag 5000         -0.006463485
##
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.022308219
## Lag 100          -0.042901544
## Lag 500          -0.007774208
## Lag 1000         -0.024987119
## Lag 5000         0.022428057
##
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0            -0.022308219
## Lag 100          -0.042901544
## Lag 500          -0.007774208
## Lag 1000         -0.024987119
## Lag 5000         0.022428057
##
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0            0.022762562
## Lag 100          0.058195842
## Lag 500          -0.006642962
## Lag 1000         -0.003238573
## Lag 5000         -0.006463481
##
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0            0.019317584
## Lag 100          0.006631109
## Lag 500          -0.010021686
## Lag 1000         -0.028126873
## Lag 5000         0.027531394
##
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0            -0.024077943
## Lag 100          -0.008773569
## Lag 500          -0.036834295
## Lag 1000         0.002484525
## Lag 5000         -0.006980967
##
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0            -0.024077943
## Lag 100          -0.008773569
## Lag 500          -0.036834295

```

```

## Lag 1000                               0.002484525
## Lag 5000                               -0.006980967
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                                    0.019317585
## Lag 100                                 0.006631109
## Lag 500                                 -0.010021687
## Lag 1000                                -0.028126868
## Lag 5000                                0.027531392
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0        0.03137990  -0.045138331  -0.074914843  0.010656057
## Lag 100      0.01846540  -0.057249918   0.023899199  0.064677145
## Lag 500      0.03717666   0.038894481  -0.058146358  -0.003011312
## Lag 1000     -0.02033719  -0.027871935   0.007427151  -0.040316314
## Lag 5000     -0.01443541  -0.006005922  -0.022039264  -0.041817387
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                               1.000000000
## Lag 100                             0.01369121
## Lag 500                             0.04194817
## Lag 1000                            -0.07246259
## Lag 5000                            0.02792893
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                               0.823958396
## Lag 100                            -0.008329624
## Lag 500                             0.062460340
## Lag 1000                            -0.039303741
## Lag 5000                            0.054712853
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0                               0.823958396
## Lag 100                            -0.008329624
## Lag 500                             0.062460340
## Lag 1000                            -0.039303741
## Lag 5000                            0.054712853
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0                               0.50406802
## Lag 100                            -0.04934121
## Lag 500                             0.04076741
## Lag 1000                            0.01261949
## Lag 5000                            0.07086807
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0                               -0.033043522
## Lag 100                            0.041554101
## Lag 500                            -0.009477397
## Lag 1000                            -0.009847229
## Lag 5000                            -0.007099129
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0                               -0.0215659725
## Lag 100                            0.0520737617

```

```

## Lag 500          -0.0336840982
## Lag 1000         -0.0001323435
## Lag 5000          0.0032024860
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0            -0.0215659725
## Lag 100           0.0520737617
## Lag 500          -0.0336840982
## Lag 1000         -0.0001323435
## Lag 5000          0.0032024860
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0            -0.02355430
## Lag 100           0.03658290
## Lag 500          -0.05089403
## Lag 1000          0.01436107
## Lag 5000          0.02999856
##
## , , at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0             0.02236567
## Lag 100           0.06353895
## Lag 500           0.01792884
## Lag 1000          -0.01728677
## Lag 5000          0.01065908
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.01575977
## Lag 100           0.04742525
## Lag 500           0.03550099
## Lag 1000          -0.02954787
## Lag 5000          -0.00496397
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0            -0.01575977
## Lag 100           0.04742525
## Lag 500           0.03550099
## Lag 1000          -0.02954787
## Lag 5000          -0.00496397
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0             0.02236567
## Lag 100           0.06353895
## Lag 500           0.01792884
## Lag 1000          -0.01728677
## Lag 5000          0.01065908
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0             0.002652222
## Lag 100           -0.010429352
## Lag 500           -0.021288103
## Lag 1000          -0.007999975

```

```

## Lag 5000                               0.019762979
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal   -0.007293165
## Lag 0                                    0.005227658
## Lag 100                                 -0.033764314
## Lag 500                                 0.017945899
## Lag 1000                                -0.001672632
## Lag 5000
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal   -0.007293165
## Lag 0                                    0.005227658
## Lag 100                                 -0.033764314
## Lag 500                                 0.017945899
## Lag 1000                                -0.001672632
## Lag 5000
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal   0.002652225
## Lag 0                                    -0.010429349
## Lag 100                                 -0.021288102
## Lag 500                                 -0.007999970
## Lag 1000                                0.019762979
## Lag 5000
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0        0.004062274   -0.08712665   -0.008518268   -0.05264070
## Lag 100      0.010963120   -0.01901086   -0.010654363    0.07192496
## Lag 500      0.020424796    0.02733555   -0.030878714   -0.01768252
## Lag 1000     -0.027055067   -0.02492168   -0.001611622   -0.04952808
## Lag 5000     -0.001142555   -0.01083790   -0.022179343   -0.03555068
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                                    0.823958396
## Lag 100                                 0.004401269
## Lag 500                                 0.017203529
## Lag 1000                                -0.030453790
## Lag 5000                                0.015804862
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                                    1.000000000
## Lag 100                                -0.006182391
## Lag 500                                 0.061955983
## Lag 1000                                -0.010693302
## Lag 5000                                0.036099351
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0                                    1.000000000
## Lag 100                                -0.006182391
## Lag 500                                 0.061955983
## Lag 1000                                -0.010693302
## Lag 5000                                0.036099351
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0                                    0.82256826
## Lag 100                                 -0.03667399
## Lag 5000                                0.04993746

```

```

## Lag 1000          0.01636640
## Lag 5000          0.04643465
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0             -0.024405654
## Lag 100            0.050538857
## Lag 500            0.004825519
## Lag 1000           0.006447716
## Lag 5000           -0.034705839
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0             -0.02293030
## Lag 100            0.04705166
## Lag 500            -0.01213092
## Lag 1000           0.01884484
## Lag 5000           -0.01173929
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0             -0.02293030
## Lag 100            0.04705166
## Lag 500            -0.01213092
## Lag 1000           0.01884484
## Lag 5000           -0.01173929
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0             -0.01731642
## Lag 100            0.03837677
## Lag 500            -0.02371802
## Lag 1000           0.04334219
## Lag 5000           0.02973740
##
## , , at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0              0.02236567
## Lag 100            0.06353895
## Lag 500            0.01792884
## Lag 1000           -0.01728677
## Lag 5000           0.01065908
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0             -0.01575977
## Lag 100            -0.04742525
## Lag 500            -0.03550099
## Lag 1000           -0.02954787
## Lag 5000           -0.00496397
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0             -0.01575977
## Lag 100            -0.04742525
## Lag 500            -0.03550099
## Lag 1000           -0.02954787
## Lag 5000           -0.00496397

```

```

##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                               0.02236567
## Lag 100                            0.06353895
## Lag 500                            0.01792884
## Lag 1000                           -0.01728677
## Lag 5000                           0.01065908
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                               0.002652222
## Lag 100                           -0.010429352
## Lag 500                           -0.021288103
## Lag 1000                          -0.007999975
## Lag 5000                           0.019762979
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0                               -0.007293165
## Lag 100                            0.005227658
## Lag 500                            -0.033764314
## Lag 1000                           0.017945899
## Lag 5000                           -0.001672632
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0                               -0.007293165
## Lag 100                            0.005227658
## Lag 500                            -0.033764314
## Lag 1000                           0.017945899
## Lag 5000                           -0.001672632
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0                               0.002652225
## Lag 100                           -0.010429349
## Lag 500                           -0.021288102
## Lag 1000                          -0.007999970
## Lag 5000                           0.019762979
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0      0.004062274   -0.08712665   -0.008518268   -0.05264070
## Lag 100    0.010963120   -0.01901086   -0.010654363    0.07192496
## Lag 500    0.020424796    0.02733555   -0.030878714   -0.01768252
## Lag 1000   -0.027055067   -0.02492168   -0.001611622   -0.04952808
## Lag 5000   -0.001142555   -0.01083790   -0.022179343   -0.03555068
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                               0.823958396
## Lag 100                            0.004401269
## Lag 500                            0.017203529
## Lag 1000                           -0.030453790
## Lag 5000                           0.015804862
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                               1.000000000
## Lag 100                           -0.006182391
## Lag 500                            0.061955983
## Lag 1000                           -0.010693302

```

```

## Lag 5000                               0.036099351
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units   1.000000000
## Lag 0                                    -0.006182391
## Lag 100                                 0.061955983
## Lag 500                                 -0.010693302
## Lag 1000                                0.036099351
## Lag 5000
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units   0.82256826
## Lag 0                                    -0.03667399
## Lag 100                                 0.04993746
## Lag 500                                 0.01636640
## Lag 1000                                0.04643465
## Lag 5000
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units   -0.024405654
## Lag 0                                    0.050538857
## Lag 100                                 0.004825519
## Lag 500                                 0.006447716
## Lag 1000                                -0.034705839
## Lag 5000
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units   -0.02293030
## Lag 0                                    0.04705166
## Lag 100                                 -0.01213092
## Lag 500                                 0.01884484
## Lag 1000                                -0.01173929
## Lag 5000
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units   -0.02293030
## Lag 0                                    0.04705166
## Lag 100                                 -0.01213092
## Lag 500                                 0.01884484
## Lag 1000                                -0.01173929
## Lag 5000
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units   -0.01731642
## Lag 0                                    0.03837677
## Lag 100                                 -0.02371802
## Lag 500                                 0.04334219
## Lag 1000                                0.02973740
##
## , , at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal   -0.01031463
## Lag 0                                    0.04023582
## Lag 100                                 0.04917285
## Lag 500                                 -0.03055049
## Lag 1000                                0.02723070
## Lag 5000
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal

```

```

## Lag 0           -0.008530358
## Lag 100        -0.035788854
## Lag 500        -0.049379232
## Lag 1000       -0.021325041
## Lag 5000       -0.010162718
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0           -0.008530358
## Lag 100         -0.035788854
## Lag 500         -0.049379232
## Lag 1000        -0.021325041
## Lag 5000        -0.010162718
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0           -0.01031463
## Lag 100          0.04023582
## Lag 500          0.04917285
## Lag 1000         -0.03055049
## Lag 5000         0.02723070
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0           -0.0020544560
## Lag 100          -0.0421619393
## Lag 500          -0.0255939041
## Lag 1000         0.0255731937
## Lag 5000         0.0001228549
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0           0.002102035
## Lag 100         -0.001379098
## Lag 500         -0.043434436
## Lag 1000        0.029664521
## Lag 5000        0.008098285
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0           0.002102035
## Lag 100         -0.001379098
## Lag 500         -0.043434436
## Lag 1000        0.029664521
## Lag 5000        0.008098285
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0           -0.0020544532
## Lag 100          -0.0421619357
## Lag 500          -0.0255939038
## Lag 1000         0.0255731963
## Lag 5000         0.0001228553
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0           -0.024177752   -0.12342257    0.029026453   -0.150664503
## Lag 100          0.002722644   0.01828947   -0.026421477   0.057871224
## Lag 500          0.001177915   0.02306467   -0.008302615   -0.001595395
## Lag 1000         -0.024450338  -0.01252743    0.007380387   -0.059682493
## Lag 5000          0.025952792  -0.01290027   -0.016603877   -0.008151565

```

```

## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0 0.50406802
## Lag 100 -0.01684331
## Lag 500 -0.02957066
## Lag 1000 -0.01048833
## Lag 5000 0.03126125
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0 0.822568265
## Lag 100 -0.018771091
## Lag 500 0.005255697
## Lag 1000 -0.022320801
## Lag 5000 0.035047964
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0 0.822568265
## Lag 100 -0.018771091
## Lag 500 0.005255697
## Lag 1000 -0.022320801
## Lag 5000 0.035047964
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0 1.000000000
## Lag 100 -0.023522308
## Lag 500 0.009564241
## Lag 1000 -0.003345884
## Lag 5000 0.033520801
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0 -0.016051070
## Lag 100 0.025579389
## Lag 500 0.032949698
## Lag 1000 0.002284379
## Lag 5000 -0.052985152
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0 -0.015097854
## Lag 100 0.015648340
## Lag 500 0.018098141
## Lag 1000 0.008134587
## Lag 5000 -0.022911727
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0 -0.015097854
## Lag 100 0.015648340
## Lag 500 0.018098141
## Lag 1000 0.008134587
## Lag 5000 -0.022911727
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0 -0.01042859
## Lag 100 0.01241322
## Lag 500 -0.01095842
## Lag 1000 0.04284066

```

```

## Lag 5000          0.01861245
##
## , , at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
##
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0             0.041893462
## Lag 100           0.013012827
## Lag 500           0.018612812
## Lag 1000          0.007230337
## Lag 5000          -0.035189323
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0             0.005815507
## Lag 100           -0.013543955
## Lag 500           0.005673448
## Lag 1000          0.036662349
## Lag 5000          -0.015403733
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0             0.005815507
## Lag 100           -0.013543955
## Lag 500           0.005673448
## Lag 1000          0.036662349
## Lag 5000          -0.015403733
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0             0.041893463
## Lag 100           0.013012825
## Lag 500           0.018612808
## Lag 1000          0.007230335
## Lag 5000          -0.035189323
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0             -0.012587692
## Lag 100           0.013086359
## Lag 500           -0.045715917
## Lag 1000          -0.011715657
## Lag 5000          0.005419636
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0             0.003830731
## Lag 100           0.018194535
## Lag 500           0.007394958
## Lag 1000          0.037075016
## Lag 5000          0.032991654
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0             0.003830731
## Lag 100           0.018194535
## Lag 500           0.007394958
## Lag 1000          0.037075016
## Lag 5000          0.032991654
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal

```

```

## Lag 0                               -0.012587692
## Lag 100                             0.013086358
## Lag 500                             -0.045715919
## Lag 1000                            -0.011715659
## Lag 5000                            0.005419638
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0      -0.0246628378   0.027563373  -0.1123023496   0.038430775
## Lag 100     -0.0006534773   0.023430140   0.0483097879  -0.071115069
## Lag 500     -0.0089061496   -0.023938075   0.0538343828  -0.035653713
## Lag 1000    0.0388264676   0.047908029  -0.0034643972  -0.069348371
## Lag 5000    0.0094923504   -0.007668143   0.0005980899   0.001697667
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0                               -0.03304352
## Lag 100                             0.02067113
## Lag 500                             -0.01234139
## Lag 1000                            0.02649335
## Lag 5000                            -0.02026354
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0                               -0.024405654
## Lag 100                             0.028902660
## Lag 500                             0.002513982
## Lag 1000                            0.029707035
## Lag 5000                            0.003612644
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0                               -0.024405654
## Lag 100                             0.028902660
## Lag 500                             0.002513982
## Lag 1000                            0.029707035
## Lag 5000                            0.003612644
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0                               -0.016051070
## Lag 100                             0.023913919
## Lag 500                             0.007815152
## Lag 1000                            0.007869038
## Lag 5000                            0.042763915
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0                               1.000000000
## Lag 100                            -0.010070063
## Lag 500                             0.008873360
## Lag 1000                            -0.034546071
## Lag 5000                            -0.003176807
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0                               0.834736019
## Lag 100                            -0.056847293
## Lag 500                             0.016590019
## Lag 1000                            -0.043298737
## Lag 5000                            0.004888737

```

```

##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0                               0.834736019
## Lag 100                            -0.056847293
## Lag 500                            0.016590019
## Lag 1000                           -0.043298737
## Lag 5000                           0.004888737
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0                               0.53614652
## Lag 100                            -0.05605043
## Lag 500                            0.01141045
## Lag 1000                           -0.03702388
## Lag 5000                           0.01411382
##
## , , at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
##
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0                               0.067454299
## Lag 100                            0.032113695
## Lag 500                            0.004944910
## Lag 1000                           0.004241478
## Lag 5000                           -0.024462625
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0                               -0.02811191
## Lag 100                            -0.03088823
## Lag 500                            -0.02192075
## Lag 1000                           0.02858712
## Lag 5000                           -0.02228574
##           at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0                               -0.02811191
## Lag 100                            -0.03088823
## Lag 500                            -0.02192075
## Lag 1000                           0.02858712
## Lag 5000                           -0.02228574
##           at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0                               0.067454297
## Lag 100                            0.032113696
## Lag 500                            0.004944906
## Lag 1000                           0.004241477
## Lag 5000                           -0.024462625
##           at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0                               -0.005906983
## Lag 100                            0.025695166
## Lag 500                            -0.026539717
## Lag 1000                           -0.002673988
## Lag 5000                           0.007084844
##           at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0                               0.001171218

```

```

## Lag 100          0.022089501
## Lag 500          0.011075437
## Lag 1000         0.024246787
## Lag 5000         0.009425114
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0             0.001171218
## Lag 100           0.022089501
## Lag 500           0.011075437
## Lag 1000          0.024246787
## Lag 5000          0.009425114
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0             -0.005906985
## Lag 100           0.025695166
## Lag 500           -0.026539718
## Lag 1000          -0.002673987
## Lag 5000          0.007084847
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0   -0.0275271884  -0.020160651   0.011701233   -0.07682037
## Lag 100  0.0089350042  -0.001230592   0.036811929   -0.08498764
## Lag 500  -0.0171963815  -0.030526948   0.032191211   -0.03111354
## Lag 1000 0.0202456175  -0.006661377   0.011866900   -0.05501113
## Lag 5000 -0.0002988564  0.002999329   -0.006049455   0.01140095
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0             -0.021565973
## Lag 100           0.022888235
## Lag 500           0.008505413
## Lag 1000          0.031209139
## Lag 5000          -0.020380719
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0             -0.0229302985
## Lag 100           0.0380673613
## Lag 500           0.0297742244
## Lag 1000          0.0451962823
## Lag 5000          0.0008947703
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0             -0.0229302985
## Lag 100           0.0380673613
## Lag 500           0.0297742244
## Lag 1000          0.0451962823
## Lag 5000          0.0008947703
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0             -0.01509785
## Lag 100           0.03378488
## Lag 500           0.01426613
## Lag 1000          0.02943689
## Lag 5000          0.01788225
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units

```

```

## Lag 0          0.834736019
## Lag 100       -0.009799499
## Lag 500        0.013020026
## Lag 1000      -0.045786597
## Lag 5000      -0.003017397
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0          1.000000000
## Lag 100       -0.047701283
## Lag 500        0.019555344
## Lag 1000      -0.056983801
## Lag 5000      -0.003963176
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0          1.000000000
## Lag 100       -0.047701283
## Lag 500        0.019555344
## Lag 1000      -0.056983801
## Lag 5000      -0.003963176
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0          0.84509620
## Lag 100       -0.04080851
## Lag 500        0.01669182
## Lag 1000      -0.04879806
## Lag 5000      -0.01025480
##
## , , at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0          0.067454299
## Lag 100       0.032113695
## Lag 500        0.0049444910
## Lag 1000      0.004241478
## Lag 5000      -0.024462625
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0          -0.02811191
## Lag 100       -0.03088823
## Lag 500        -0.02192075
## Lag 1000      0.02858712
## Lag 5000      -0.02228574
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal
## Lag 0          -0.02811191
## Lag 100       -0.03088823
## Lag 500        -0.02192075
## Lag 1000      0.02858712
## Lag 5000      -0.02228574
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal
## Lag 0          0.067454297
## Lag 100       0.032113696

```

```

## Lag 500          0.004944906
## Lag 1000         0.004241477
## Lag 5000        -0.024462625
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal
## Lag 0           -0.005906983
## Lag 100          0.025695166
## Lag 500          -0.026539717
## Lag 1000         -0.002673988
## Lag 5000         0.007084844
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal
## Lag 0            0.001171218
## Lag 100          0.022089501
## Lag 500          0.011075437
## Lag 1000         0.024246787
## Lag 5000         0.009425114
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal
## Lag 0            0.001171218
## Lag 100          0.022089501
## Lag 500          0.011075437
## Lag 1000         0.024246787
## Lag 5000         0.009425114
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.animal
## Lag 0           -0.005906985
## Lag 100          0.025695166
## Lag 500          -0.026539718
## Lag 1000         -0.002673987
## Lag 5000         0.007084847
##      traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0   -0.0275271884  -0.020160651   0.011701233  -0.07682037
## Lag 100  0.0089350042  -0.001230592   0.036811929  -0.08498764
## Lag 500  -0.0171963815  -0.030526948   0.032191211  -0.03111354
## Lag 1000 0.0202456175  -0.006661377   0.011866900  -0.05501113
## Lag 5000 -0.0002988564  0.002999329  -0.006049455   0.01140095
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units
## Lag 0           -0.021565973
## Lag 100          0.022888235
## Lag 500          0.008505413
## Lag 1000         0.031209139
## Lag 5000         -0.020380719
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0           -0.0229302985
## Lag 100          0.0380673613
## Lag 500          0.0297742244
## Lag 1000         0.0451962823
## Lag 5000         0.0008947703
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0           -0.0229302985

```

```

## Lag 100          0.0380673613
## Lag 500          0.0297742244
## Lag 1000         0.0451962823
## Lag 5000         0.0008947703
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0            -0.01509785
## Lag 100          0.03378488
## Lag 500          0.01426613
## Lag 1000         0.02943689
## Lag 5000         0.01788225
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0            0.834736019
## Lag 100          -0.009799499
## Lag 500          0.013020026
## Lag 1000         -0.045786597
## Lag 5000         -0.003017397
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0            1.000000000
## Lag 100          -0.047701283
## Lag 500          0.019555344
## Lag 1000         -0.056983801
## Lag 5000         -0.003963176
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0            1.000000000
## Lag 100          -0.047701283
## Lag 500          0.019555344
## Lag 1000         -0.056983801
## Lag 5000         -0.003963176
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0            0.84509620
## Lag 100          -0.04080851
## Lag 500          0.01669182
## Lag 1000         -0.04879806
## Lag 5000         -0.01025480
##
## , , at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
##
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.animal
## Lag 0            0.081823186
## Lag 100          0.025862129
## Lag 500          -0.001019422
## Lag 1000         0.022038934
## Lag 5000         0.008353223
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.animal
## Lag 0            -0.042456056
## Lag 100          -0.036956287
## Lag 500          -0.028788001

```

```

## Lag 1000          0.017511964
## Lag 5000          0.003358484
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.animal -0.042456056
## Lag 0             -0.036956287
## Lag 100            0.017511964
## Lag 500            0.003358484
## Lag 1000           0.017511964
## Lag 5000           0.003358484
## at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.animal 0.081823182
## Lag 0              0.025862131
## Lag 100             0.001019426
## Lag 500             0.022038934
## Lag 1000            0.008353222
## Lag 5000            0.0267444524
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.animal 0.0006335032
## Lag 0               0.0114320128
## Lag 100              -0.0042014027
## Lag 500              0.0067096438
## Lag 1000             0.0267444553
## Lag 5000             0.011573448
## at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.animal 0.031257650
## Lag 0               0.029970355
## Lag 100              0.007774208
## Lag 500              -0.008009508
## Lag 1000             0.011573448
## Lag 5000             0.031257650
## at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.animal 0.029970355
## Lag 0               0.007774208
## Lag 100              -0.008009508
## Lag 5000             0.0006335012
## Lag 0               0.0114320134
## Lag 100              -0.0042014032
## Lag 500              0.0067096460
## Lag 1000             0.0267444553
## traitT1.byear traitT2.byear traitT1.mother traitT2.mother
## Lag 0      -0.008651050   -0.05004816    0.063124373   -0.186833159
## Lag 100     0.010418974   -0.04495129    0.038216075   -0.079695676
## Lag 500     -0.023819545   -0.01200187   -0.001081368   -0.016737029
## Lag 1000    0.001822076   -0.05243956    0.006894149   -0.049173928
## Lag 5000    -0.003613643    0.01656821   -0.032479008    0.005560052
## at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT1.units -0.023554303
## Lag 0                  0.001740856
## Lag 100

```

```

## Lag 500          0.012843461
## Lag 1000         0.008680482
## Lag 5000        -0.009920137
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT1.units
## Lag 0            -0.017316421
## Lag 100          0.029916369
## Lag 500          0.033143611
## Lag 1000         0.017296002
## Lag 5000         0.006303038
##      at.level(sex, "1"):traitT1:at.level(sex, "1"):traitT2.units
## Lag 0            -0.017316421
## Lag 100          0.029916369
## Lag 500          0.033143611
## Lag 1000         0.017296002
## Lag 5000         0.006303038
##      at.level(sex, "1"):traitT2:at.level(sex, "1"):traitT2.units
## Lag 0            -0.010428588
## Lag 100          0.029648297
## Lag 500          0.009827887
## Lag 1000         0.014861996
## Lag 5000         0.005418763
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT1.units
## Lag 0            0.536146519
## Lag 100         -0.005826103
## Lag 500          0.037319533
## Lag 1000         -0.030233133
## Lag 5000         -0.008069233
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT1.units
## Lag 0            0.845096201
## Lag 100         -0.021641972
## Lag 500          0.044233606
## Lag 1000         -0.035207990
## Lag 5000         -0.008416608
##      at.level(sex, "2"):traitT1:at.level(sex, "2"):traitT2.units
## Lag 0            0.845096201
## Lag 100         -0.021641972
## Lag 500          0.044233606
## Lag 1000         -0.035207990
## Lag 5000         -0.008416608
##      at.level(sex, "2"):traitT2:at.level(sex, "2"):traitT2.units
## Lag 0            1.000000000
## Lag 100         -0.002381589
## Lag 500          0.036399260
## Lag 1000         -0.027816041
## Lag 5000         -0.019613782

```

Here the simple plot to plot the genetic correlation

```

DvsS <- data.frame(
  Trait = colnames(model2.5$Sol),
  BLUP = posterior.mode(model2.5$Sol),
  CI = HPDinterval((model2.5$Sol))
)
DvsS <- DvsS[5:5240, ] # keep only rows associated with animal
DvsS$ID <- substr(DvsS$Trait, 35, 38)
DvsS$TRAIT <- substr(DvsS$Trait, 25, 26)
DvsS$SEX <- substr(DvsS$Trait, 16, 16)
summary(factor(DvsS$TRAIT))

##    T1      T2
## 2618 2618

DvsS$Trait <- NULL
BLUPS <- reshape(DvsS, v.names = c("BLUP", "CI.lower", "CI.upper"), idvar = c("ID", "SEX"),
nrow(BLUPS)

## [1] 2618

rownames(BLUPS) <- c()
colnames(BLUPS) <- c("ID", "SEX", "BLUP.btw", "CI.L.btw", "CI.U.btw", "BLUP.tarsus", "CI.L.tarsus")
summary(BLUPS)

##           ID              SEX          BLUP.btw          CI.L.btw
## Length:2618    Length:2618    Min.   :-1.80272   Min.   :-3.1968
## Class :character Class :character  1st Qu.:-0.26298  1st Qu.:-2.0362
## Mode  :character Mode  :character Median : 0.02120  Median :-1.8561
##                                         Mean   : 0.01884  Mean   :-1.8044
##                                         3rd Qu.: 0.29973  3rd Qu.:-1.6116
##                                         Max.   : 1.98486  Max.   : 0.6303
##           CI.U.btw          BLUP.tarsus          CI.L.tarsus          CI.U.tarsus
## Min.   :-0.1541    Min.   :-1.80246    Min.   :-3.197   Min.   :-0.1557
## 1st Qu.: 1.6313    1st Qu.:-0.26067   1st Qu.:-2.038   1st Qu.: 1.6302
## Median : 1.8658    Median : 0.02262   Median :-1.854   Median : 1.8671
## Mean   : 1.8247    Mean   : 0.01982   Mean   :-1.804   Mean   : 1.8249
## 3rd Qu.: 2.0468    3rd Qu.: 0.30215   3rd Qu.:-1.612   3rd Qu.: 2.0483
## Max.   : 3.8419    Max.   : 1.99604   Max.   : 0.632   Max.   : 3.8431

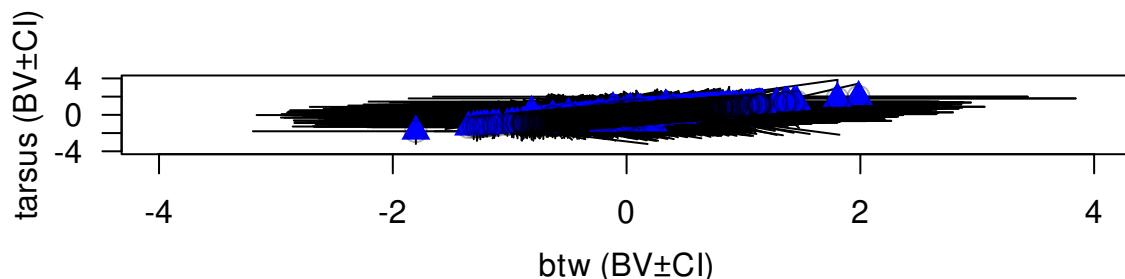
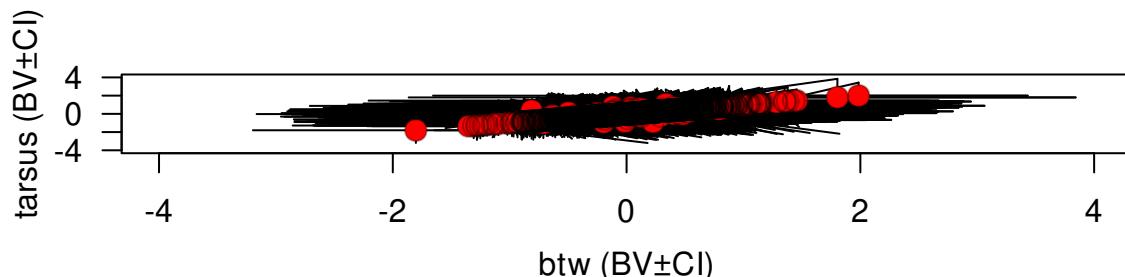
#
#
FEM <- subset(BLUPS, SEX == "1")
MAL <- subset(BLUPS, SEX == "0")
par(mfrow = c(2, 1))

```

```

plot(BLUP.tarsus ~ BLUP.btw, FEM, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = BLUPS$BLUP.btw, y0 = FEM$CI.L.tarsus, x1 = FEM$BLUP.btw, y1 = FEM$CI.U.tarsus,
arrows(x0 = BLUPS$CI.L.btw, y0 = FEM$BLUP.tarsus, x1 = FEM$CI.U.btw, y1 = FEM$BLUP.tarsus,
points(BLUP.tarsus ~ BLUP.btw, FEM, pch = 16, col = "red", cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, FEM, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("btw (BV±CI)", side = 1, line = 2.4)
mtext("tarsus (BV±CI)", side = 2, line = 2, las = 3)
#
plot(BLUP.tarsus ~ BLUP.btw, MAL, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = BLUPS$BLUP.btw, y0 = MAL$CI.L.tarsus, x1 = MAL$BLUP.btw, y1 = MAL$CI.U.tarsus,
arrows(x0 = BLUPS$CI.L.btw, y0 = MAL$BLUP.tarsus, x1 = MAL$CI.U.btw, y1 = MAL$BLUP.tarsus,
points(BLUP.tarsus ~ BLUP.btw, MAL, pch = 17, col = "blue", cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, MAL, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("btw (BV±CI)", side = 1, line = 2.4)
mtext("tarsus (BV±CI)", side = 2, line = 2, las = 3)

```



#

3.4.5 Between groups (co)variances and the B-matrix

Animal models are amazing model. With different group within a population, it is also possible to estimate how much the different groups shared the same genetic via the cross-group genetic covariance. This covariance is essential to understand ontogenetic or sexual conflict, which can constraint or enhanced response to evolution. As an example, we estimate the cross-sex genetic correlation `r_{fm}`

First, we need to dissociate the trait values for females and males into distinct variables. Then, we use a bivariate model (for one trait: `tarsus`) and a multivariate model (for various traits: `tarsus` and `bwt`). With a multivariate model, the cross-sex-cross trait covariance matrix is also named **B matrix**.

The coding is a bit complain but pretty straightforward. It is important to modify the covariance matrix at the residual level to avoid the calculation of a cross-sex residual covariance (no individual switched sex during the experiment).

```

gryphon2$bwt.1 <- NA
gryphon2$tarsus.1 <- NA
animal <- gryphon2[gryphon2$sex == "1", ]$animal
for (i in unique(animal)) {
  gryphon2$bwt.1[which(gryphon2$animal == i)] <- gryphon2$bwt[which(gryphon2$animal == i)]
  gryphon2$tarsus.1[which(gryphon2$animal == i)] <- gryphon2$tarsus[which(gryphon2$animal == i)]
}
#
gryphon2$bwt.2 <- NA
gryphon2$tarsus.2 <- NA
animal <- gryphon2[gryphon2$sex == "2", ]$animal
for (i in unique(animal)) {
  gryphon2$bwt.2[which(gryphon2$animal == i)] <- gryphon2$bwt[which(gryphon2$animal == i)]
  gryphon2$tarsus.2[which(gryphon2$animal == i)] <- gryphon2$tarsus[which(gryphon2$animal == i)]
}

#
prior2.4 <- list(
  G = list(
    G1 = list(V = diag(2), nu = 1.002),
    G2 = list(V = diag(2), nu = 1.002),
    G3 = list(V = diag(2), nu = 1.002)
  ),
  R = list(
    V1 = list(V = diag(2), nu = 1.002)
  )
)
#
model.BivSex <- MCMCglmm(cbind(tarsus.1, tarsus.2) ~ trait - 1,
  random = ~ us(trait):animal + idh(trait):byear + idh(trait):mother,
```

```

rcov = ~ us(trait):units,
family = c("gaussian", "gaussian"),
ginv = list(animal = Ainv), data = gryphon2,
nitt = 130000, thin = 100, burnin = 30000,
prior = prior2.4, verbose = FALSE, pr = TRUE
)

save(model.BivSex, file = "data/MCMCglmm_model_BivSex_LongRun.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load(file = "data/MCMCglmm_model_BivSex_LongRun.rda")
summary(model.BivSex)

```

```

##
## Iterations = 30001:129901
## Thinning interval = 100
## Sample size = 1000
##
## DIC: 1670.599
##
## G-structure: ~us(trait):animal
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traittarsus.1:traittarsus.1.animal      6.632    2.136   12.69    85.74
## traittarsus.2:traittarsus.1.animal      8.043    2.389   13.54   117.04
## traittarsus.1:traittarsus.2.animal      8.043    2.389   13.54   117.04
## traittarsus.2:traittarsus.2.animal     16.145    3.128   28.93   21.81
##
## ~idh(trait):byear
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traittarsus.1.byear      3.184    0.505   6.515   357.4
## traittarsus.2.byear      4.576    1.346   8.476   442.5
##
## ~idh(trait):mother
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traittarsus.1.mother     1.777  0.07858   4.714   299.68
## traittarsus.2.mother     2.980  0.12204   7.328   70.26
##
## R-structure: ~us(trait):units
##
##                               post.mean l-95% CI u-95% CI eff.samp
## traittarsus.1:traittarsus.1.units     15.455   8.998   21.84  104.923

```

```

## traittarsus.2:traittarsus.1.units      -1.497  -15.500   15.53   8.767
## traittarsus.1:traittarsus.2.units      -1.497  -15.500   15.53   8.767
## traittarsus.2:traittarsus.2.units      9.356    0.239   19.10  21.548
##
## Location effects: cbind(tarsus.1, tarsus.2) ~ trait - 1
##
##           post.mean l-95% CI u-95% CI eff.samp pMCMC
## traittarsus.1      20.48    19.62    21.48    703.9 <0.001 ***
## traittarsus.2      20.46    19.42    21.40    846.8 <0.001 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
autocorr(model.BivSex$VCV)
```

```

## , , traittarsus.1:traittarsus.1.animal
##
##           traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0                  1.00000000          0.48422763
## Lag 100                 0.73797990          0.27701934
## Lag 500                 0.43151596          0.08460564
## Lag 1000                0.27709357          0.03138071
## Lag 5000                0.09623473          0.02508175
##           traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0                  0.48422763         -0.20835012
## Lag 100                 0.27701934         -0.26518087
## Lag 500                 0.08460564         -0.27929151
## Lag 1000                0.03138071         -0.23408954
## Lag 5000                0.02508175         -0.02431482
##           traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0                  0.05650214        -0.059474225       -0.038396947
## Lag 100                 0.03805347        -0.063250151       -0.012779432
## Lag 500                 0.03353186        -0.031583736       0.059794328
## Lag 1000                0.02971111        0.032191172       -0.004310584
## Lag 5000                0.04021796        -0.001198619       -0.065221991
##           traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0                  0.1285643          -0.7394524
## Lag 100                 0.1416089          -0.5874113
## Lag 500                 0.1227812          -0.3892610
## Lag 1000                0.1054744          -0.2272794
## Lag 5000                0.1046093          -0.0490914
##           traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0                  -0.030661923       -0.030661923
## Lag 100                 -0.009066847       -0.009066847
## Lag 500                 -0.014021640       -0.014021640
## Lag 1000                -0.038727365       -0.038727365
## Lag 5000                -0.053406989       -0.053406989

```

```

##      traittarsus.2:traittarsus.2.units
## Lag 0                      0.24444681
## Lag 100                     0.28630940
## Lag 500                     0.28676382
## Lag 1000                    0.25016902
## Lag 5000                   -0.01889626
##
## , , traittarsus.2:traittarsus.1.animal
##
##      traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0                      0.48422763          1.00000000
## Lag 100                     0.25316393          0.68722653
## Lag 500                     -0.01525426          0.30994286
## Lag 1000                    -0.13117432          0.12750201
## Lag 5000                   -0.03481149          0.01967858
##      traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0                      1.00000000          0.4372811
## Lag 100                     0.68722653          0.3171927
## Lag 500                     0.30994286          0.2173200
## Lag 1000                    0.12750201          0.2034517
## Lag 5000                   0.01967858          0.1155543
##      traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0                      -0.028123268         -0.10117282          0.068035501
## Lag 100                     -0.036441213         -0.06326083          0.058376180
## Lag 500                     -0.017206400         -0.06159010          -0.002869166
## Lag 1000                    -0.004346341         0.06173108          -0.017320527
## Lag 5000                   -0.044798021         0.07734401          -0.083350187
##      traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0                      -0.261520775          -0.32691256
## Lag 100                     -0.219989670          -0.19037150
## Lag 500                     -0.171008994          0.03568009
## Lag 1000                    -0.073253782          0.15052971
## Lag 5000                   -0.002628408          0.06147675
##      traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0                      -0.15561461          -0.15561461
## Lag 100                     -0.11940169          -0.11940169
## Lag 500                     -0.10722180          -0.10722180
## Lag 1000                    -0.11888700          -0.11888700
## Lag 5000                   -0.07749379          -0.07749379
##      traittarsus.2:traittarsus.2.units
## Lag 0                      -0.3304177
## Lag 100                     -0.2624116
## Lag 500                     -0.1893048
## Lag 1000                    -0.2045441
## Lag 5000                   -0.1580773
##
## , , traittarsus.1:traittarsus.2.animal

```

```

##                                     traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0                               0.48422763                         1.000000000
## Lag 100                             0.25316393                         0.68722653
## Lag 500                             -0.01525426                        0.30994286
## Lag 1000                            -0.13117432                        0.12750201
## Lag 5000                            -0.03481149                        0.01967858
##                                     traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0                               1.00000000                         0.4372811
## Lag 100                            0.68722653                         0.3171927
## Lag 500                            0.30994286                         0.2173200
## Lag 1000                           0.12750201                         0.2034517
## Lag 5000                           0.01967858                         0.1155543
##                                     traittarsus.1:byear traittarsus.2:byear traittarsus.1:mother
## Lag 0                               -0.028123268                      -0.10117282                         0.068035501
## Lag 100                            -0.036441213                      -0.06326083                         0.058376180
## Lag 500                            -0.017206400                      -0.06159010                        -0.002869166
## Lag 1000                           -0.004346341                      0.06173108                         -0.017320527
## Lag 5000                           -0.044798021                      0.07734401                         -0.083350187
##                                     traittarsus.2:mother traittarsus.1:traittarsus.1:units
## Lag 0                               -0.261520775                      -0.32691256
## Lag 100                            -0.219989670                      -0.19037150
## Lag 500                            -0.171008994                      0.03568009
## Lag 1000                           -0.073253782                      0.15052971
## Lag 5000                           -0.002628408                      0.06147675
##                                     traittarsus.2:traittarsus.1:units traittarsus.1:traittarsus.2:units
## Lag 0                               -0.15561461                         -0.15561461
## Lag 100                            -0.11940169                         -0.11940169
## Lag 500                            -0.10722180                         -0.10722180
## Lag 1000                           -0.11888700                         -0.11888700
## Lag 5000                           -0.07749379                         -0.07749379
##                                     traittarsus.2:traittarsus.2:units
## Lag 0                               -0.3304177
## Lag 100                            -0.2624116
## Lag 500                            -0.1893048
## Lag 1000                           -0.2045441
## Lag 5000                           -0.1580773
##
## , , traittarsus.2:traittarsus.2:animal
##
##                                     traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0                               -0.2083501                         0.43728107
## Lag 100                            -0.2828015                         0.29272735
## Lag 500                            -0.3313152                         0.14658444
## Lag 1000                           -0.3483078                         0.06012732
## Lag 5000                           -0.1542996                         -0.06471744
##                                     traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal

```

```

## Lag 0           0.43728107      1.00000000
## Lag 100        0.29272735      0.87775419
## Lag 500        0.14658444      0.72227084
## Lag 1000       0.06012732      0.61062886
## Lag 5000       -0.06471744     0.08074184
##          traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0           -0.020084226    -0.013645105   0.007263224
## Lag 100         -0.015303798    0.015235251   0.013922907
## Lag 500         -0.009604004    -0.004276257  -0.022794079
## Lag 1000        -0.036443873    0.045216146   0.038406389
## Lag 5000        -0.046131641    0.012581835  -0.012187074
##          traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0           -0.50677816     0.23024103
## Lag 100         -0.46549011     0.26517948
## Lag 500         -0.35999881     0.31894466
## Lag 1000        -0.26148886     0.30787546
## Lag 5000        -0.09601646     0.09998448
##          traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0           0.03215520     0.03215520
## Lag 100         0.04234930     0.04234930
## Lag 500         0.01485912     0.01485912
## Lag 1000        -0.02777646    -0.02777646
## Lag 5000        -0.10000281    -0.10000281
##          traittarsus.2:traittarsus.2.units
## Lag 0           -0.90778147
## Lag 100         -0.84400702
## Lag 500         -0.70957533
## Lag 1000        -0.62450980
## Lag 5000        -0.06262314
##
## , , traittarsus.1.byear
##
##          traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0           0.05650214     -0.02812327
## Lag 100         0.06068975     -0.02687949
## Lag 500         0.02342116     -0.03285780
## Lag 1000        0.03532201     -0.03419221
## Lag 5000        0.02786557     -0.01481811
##          traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0           -0.02812327    -0.02008423
## Lag 100         -0.02687949    -0.02175035
## Lag 500         -0.03285780    -0.01196604
## Lag 1000        -0.03419221    -0.03085934
## Lag 5000        -0.01481811    0.01496886
##          traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0           1.00000000    -0.008840935   -0.034026465
## Lag 100         0.23201455    -0.040849216   0.006952432

```

```

## Lag 500      0.10173419      0.018677216     0.021915445
## Lag 1000    0.05656314      0.045347703     -0.069023041
## Lag 5000   -0.02674433      0.003879773     -0.031727649
##          traiittarsus.2.mother traiittarsus.1:traiittarsus.1.units
## Lag 0       0.041524616      -0.131561587
## Lag 100    0.022654125      -0.091920671
## Lag 500   -0.003137523      -0.023174815
## Lag 1000   0.093233105      0.017723781
## Lag 5000  -0.001576843      0.003971118
##          traiittarsus.2:traiittarsus.1.units traiittarsus.1:traiittarsus.2.units
## Lag 0       0.06264243      0.06264243
## Lag 100    0.06240060      0.06240060
## Lag 500    0.03789653      0.03789653
## Lag 1000   0.03352306      0.03352306
## Lag 5000   0.05768588      0.05768588
##          traiittarsus.2:traiittarsus.2.units
## Lag 0       0.013826338
## Lag 100    0.015690413
## Lag 500    0.020546446
## Lag 1000   -0.006052117
## Lag 5000   -0.020241900
##
## , , traiittarsus.2.byear
##
##          traiittarsus.1:traiittarsus.1.animal traiittarsus.2:traiittarsus.1.animal
## Lag 0       -0.059474225     -0.10117282
## Lag 100    -0.029542673     -0.06571088
## Lag 500    0.019127206      0.02553144
## Lag 1000   -0.006675716      0.03933732
## Lag 5000   -0.020363086     -0.01801919
##          traiittarsus.1:traiittarsus.2.animal traiittarsus.2:traiittarsus.2.animal
## Lag 0       -0.10117282     -0.013645105
## Lag 100    -0.06571088     -0.001743299
## Lag 500    0.02553144      0.024080673
## Lag 1000   0.03933732      0.020998801
## Lag 5000   -0.01801919     -0.036422613
##          traiittarsus.1.byear traiittarsus.2.byear traiittarsus.1.mother
## Lag 0       -0.008840935     1.00000000     0.019871390
## Lag 100    0.003895756      0.15813110     0.000200299
## Lag 500    -0.028206511     0.02861588     0.028238837
## Lag 1000   -0.022327614     0.02577769     0.069995240
## Lag 5000   0.054519302     -0.01863939     0.021348106
##          traiittarsus.2.mother traiittarsus.1:traiittarsus.1.units
## Lag 0       0.02790731      0.04582166
## Lag 100   -0.02533496      0.01584625
## Lag 500   -0.04228406     -0.01965332
## Lag 1000  -0.03776489     -0.05592538

```

```

## Lag 5000          0.04026220          0.02293994
##      traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0              -0.01692229         -0.01692229
## Lag 100             -0.03014201        -0.03014201
## Lag 500             -0.05685394        -0.05685394
## Lag 1000            -0.05775898        -0.05775898
## Lag 5000            -0.04474406        -0.04474406
##      traittarsus.2:traittarsus.2.units
## Lag 0              -0.05647151
## Lag 100             -0.01120227
## Lag 500             -0.02826105
## Lag 1000            -0.01902149
## Lag 5000             0.04653422
##
## , , traittarsus.1.mother
##
##      traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0              -0.03839695         0.068035501
## Lag 100             -0.01125448         0.090872487
## Lag 500             -0.01822303         0.074904451
## Lag 1000            -0.06684368        -0.009049799
## Lag 5000             0.01882807        -0.038704215
##      traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0              0.068035501         0.007263224
## Lag 100             0.090872487         0.009427782
## Lag 500             0.074904451        -0.008966636
## Lag 1000            -0.009049799        -0.062208885
## Lag 5000            -0.038704215        -0.090802212
##      traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0              -0.034026465         0.019871390        1.000000000
## Lag 100             -0.044009625         -0.022642909        0.53848292
## Lag 500              0.033149894         -0.058758981        0.05060514
## Lag 1000             -0.034811076         -0.041691079       -0.01263351
## Lag 5000              0.003520887         -0.007417182       -0.01561154
##      traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0              -0.053312892         -0.274551627
## Lag 100             -0.021399368         -0.183188795
## Lag 500              0.001127949         -0.004741630
## Lag 1000             0.017008055         0.049168722
## Lag 5000             -0.013865361         -0.003898312
##      traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0              0.021389916         0.021389916
## Lag 100             0.010504830         0.010504830
## Lag 500              0.005067944         0.005067944
## Lag 1000             0.024425517         0.024425517
## Lag 5000             -0.004626176        -0.004626176
##      traittarsus.2:traittarsus.2.units

```

```

## Lag 0           0.02429791
## Lag 100        0.01285473
## Lag 500        0.02017728
## Lag 1000       0.06093906
## Lag 5000       0.10720949
##
## , , traittarsus.2.mother
##
##          traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0           0.12856434          -0.2615208
## Lag 100         0.14693139          -0.2325836
## Lag 500         0.21342005          -0.1621874
## Lag 1000        0.23353394          -0.1119149
## Lag 5000        0.02771044          0.0396288
##          traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0           -0.2615208         -0.50677816
## Lag 100         -0.2325836         -0.48725853
## Lag 500         -0.1621874         -0.40888208
## Lag 1000        -0.1119149         -0.34699264
## Lag 5000        0.0396288          0.02549716
##          traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0           0.04152462         0.027907314        -0.05331289
## Lag 100         0.02194445         0.054336046        -0.04284508
## Lag 500         -0.03327513        -0.020857444        -0.02968159
## Lag 1000        0.02573536         -0.040238713        -0.03336616
## Lag 5000        0.01374507         0.005808512         0.07687084
##          traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0           1.00000000        -0.13910701
## Lag 100         0.70222369        -0.13691118
## Lag 500         0.35625516        -0.18735521
## Lag 1000        0.19638031        -0.21082439
## Lag 5000        -0.04114072        -0.04387536
##          traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0           0.04614450         0.04614450
## Lag 100         0.04486226         0.04486226
## Lag 500         0.04022585         0.04022585
## Lag 1000        0.05277963         0.05277963
## Lag 5000        0.04252673         0.04252673
##          traittarsus.2:traittarsus.2.units
## Lag 0           0.268960307
## Lag 100         0.289158808
## Lag 500         0.345252707
## Lag 1000        0.323960441
## Lag 5000        -0.002082367
##
## , , traittarsus.1:traittarsus.1.units
##

```

```

##          traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0                  -0.7394524                  -0.32691256
## Lag 100                 -0.5906993                 -0.20009588
## Lag 500                 -0.3828210                 -0.08127245
## Lag 1000                -0.2031251                  0.02298348
## Lag 5000                -0.0979703                 -0.01920779
##          traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0                  -0.32691256                  0.23024103
## Lag 100                 -0.20009588                  0.28084172
## Lag 500                 -0.08127245                  0.28272296
## Lag 1000                0.02298348                  0.28283381
## Lag 5000                -0.01920779                 0.05177093
##          traittarsus.1:byear traittarsus.2:byear traittarsus.1.mother
## Lag 0                  -0.13156159                  0.04582166                 -0.27455163
## Lag 100                 -0.05719958                  0.04337991                 -0.16825519
## Lag 500                 -0.06997182                  0.07140946                 -0.07087194
## Lag 1000                -0.02088140                  0.01426241                  0.03955072
## Lag 5000                -0.01887279                  0.04431921                  0.06767122
##          traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0                  -0.1391070                   1.00000000
## Lag 100                 -0.1379410                   0.56579808
## Lag 500                 -0.1301093                   0.35250381
## Lag 1000                -0.1253506                   0.15585076
## Lag 5000                -0.0735286                   0.05909602
##          traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0                      0.04896537                  0.04896537
## Lag 100                     0.02746645                  0.02746645
## Lag 500                     0.03218528                  0.03218528
## Lag 1000                    0.05596211                  0.05596211
## Lag 5000                    0.05413531                  0.05413531
##          traittarsus.2:traittarsus.2.units
## Lag 0                      -0.2610614
## Lag 100                     -0.3079509
## Lag 500                     -0.2969894
## Lag 1000                    -0.2847495
## Lag 5000                     -0.0427216
##
## , , traittarsus.2:traittarsus.1.units
##
##          traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0                  -0.030661923                 -0.15561461
## Lag 100                 0.003884354                 -0.11920858
## Lag 500                 0.042099074                 -0.08915914
## Lag 1000                0.042025296                 -0.07981143
## Lag 5000                -0.018911064                 -0.09774452
##          traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0                  -0.15561461                  0.03215520

```

```

## Lag 100           -0.11920858      0.04834806
## Lag 500           -0.08915914      0.05802195
## Lag 1000          -0.07981143      0.02656239
## Lag 5000          -0.09774452      -0.04245498
## traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0             0.06264243      -0.016922288     0.02138992
## Lag 100           0.06751797      -0.004991909     0.02085557
## Lag 500           0.06731777      -0.030284706     0.05500122
## Lag 1000          0.06629827      -0.044495833     0.10333723
## Lag 5000          0.08927471      -0.038637660     0.06016204
## traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0             0.04614450      0.048965370
## Lag 100           0.04880037      0.017340477
## Lag 500           0.04329157      -0.013844496
## Lag 1000          0.06724925      -0.039591736
## Lag 5000          0.04795409      -0.003568665
## traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0             1.0000000      1.0000000
## Lag 100           0.9675526      0.9675526
## Lag 500           0.9029928      0.9029928
## Lag 1000          0.8401123      0.8401123
## Lag 5000          0.4663345      0.4663345
## traittarsus.2:traittarsus.2.units
## Lag 0             -0.04275237
## Lag 100           -0.06227874
## Lag 500           -0.06672641
## Lag 1000          -0.04382724
## Lag 5000          0.03699475
##
## , , traittarsus.1:traittarsus.2.units
##
## traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0             -0.030661923     -0.15561461
## Lag 100           0.003884354     -0.11920858
## Lag 500           0.042099074     -0.08915914
## Lag 1000          0.042025296     -0.07981143
## Lag 5000          -0.018911064     -0.09774452
## traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0             -0.15561461      0.03215520
## Lag 100           -0.11920858      0.04834806
## Lag 500           -0.08915914      0.05802195
## Lag 1000          -0.07981143      0.02656239
## Lag 5000          -0.09774452      -0.04245498
## traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0             0.06264243      -0.016922288     0.02138992
## Lag 100           0.06751797      -0.004991909     0.02085557
## Lag 500           0.06731777      -0.030284706     0.05500122

```

```

## Lag 1000      0.06629827      -0.044495833      0.10333723
## Lag 5000      0.08927471      -0.038637660      0.06016204
##          traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0          0.04614450      0.048965370
## Lag 100        0.04880037      0.017340477
## Lag 500        0.04329157      -0.013844496
## Lag 1000       0.06724925      -0.039591736
## Lag 5000       0.04795409      -0.003568665
##          traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0          1.0000000      1.0000000
## Lag 100        0.9675526      0.9675526
## Lag 500        0.9029928      0.9029928
## Lag 1000       0.8401123      0.8401123
## Lag 5000       0.4663345      0.4663345
##          traittarsus.2:traittarsus.2.units
## Lag 0          -0.04275237
## Lag 100        -0.06227874
## Lag 500        -0.06672641
## Lag 1000       -0.04382724
## Lag 5000        0.03699475
##
## , , traittarsus.2:traittarsus.2.units
##
##          traittarsus.1:traittarsus.1.animal traittarsus.2:traittarsus.1.animal
## Lag 0          0.2444468      -0.33041768
## Lag 100        0.2953169      -0.24524259
## Lag 500        0.3257997      -0.12021844
## Lag 1000       0.3204576      -0.05040170
## Lag 5000       0.1966910      0.07386444
##          traittarsus.1:traittarsus.2.animal traittarsus.2:traittarsus.2.animal
## Lag 0          -0.33041768     -0.9077815
## Lag 100        -0.24524259     -0.8410651
## Lag 500        -0.12021844     -0.6969304
## Lag 1000       -0.05040170     -0.5893775
## Lag 5000        0.07386444     -0.1004775
##          traittarsus.1.byear traittarsus.2.byear traittarsus.1.mother
## Lag 0          0.01382634     -0.056471508     0.024297909
## Lag 100        0.01730830     -0.030701856     0.017739447
## Lag 500        0.03968933     0.004065694     0.010636299
## Lag 1000       0.03341332     -0.048898719     -0.028816725
## Lag 5000       0.05219060     -0.013503372     -0.005167407
##          traittarsus.2.mother traittarsus.1:traittarsus.1.units
## Lag 0          0.2689603      -0.2610614
## Lag 100        0.3013532      -0.3031510
## Lag 500        0.3094825      -0.3179085
## Lag 1000       0.2400190      -0.2838200
## Lag 5000       0.1233946      -0.1252443

```

```

##          traittarsus.2:traittarsus.1.units traittarsus.1:traittarsus.2.units
## Lag 0                  -0.04275237           -0.04275237
## Lag 100                 -0.05037125          -0.05037125
## Lag 500                 -0.01879502          -0.01879502
## Lag 1000                0.02100080          0.02100080
## Lag 5000                0.10654662          0.10654662
##          traittarsus.2:traittarsus.2.units
## Lag 0                  1.00000000
## Lag 100                0.87986649
## Lag 500                0.70214749
## Lag 1000               0.61035106
## Lag 5000               0.07522425

```

The cross-sex genetic correlation can estimate form the output of the model. For tarsus length at fledging, sexes shared a lot of genetic variance which is commun for a trait with low sexual dimorphism. If the selection is antagonistic between males and females, sexes can not evolve freely from the other sexes and a sexual conflict appears.

```

rfm <- model.BivSex$VCV[, "traittarsus.1:traittarsus.2.animal"] / sqrt(model.BivSex$VCV[, posterior.mode(rfm)]

```

```

##      var1
## 0.9664439

```

```
HPDinterval(rfm, 0.95)
```

```

##          lower     upper
## var1 0.4630817 0.992376
## attr(),"Probability")
## [1] 0.95

```

We can estimate directly the correlation and plot the cross-sex genetic correlation

```

DvsS <- data.frame(
  Trait = colnames(model.BivSex$Sol),
  BLUP = posterior.mode(model.BivSex$Sol),
  CI = HPDinterval((model.BivSex$Sol))
)
DvsS <- DvsS[3:2619, ] # keep only rows associated with animal
DvsS$ID <- substr(DvsS$Trait, 22, 26)
DvsS$TRAIT <- substr(DvsS$Trait, 6, 13)
summary(factor(DvsS$TRAIT))

```

```

## tarsus.1 tarsus.2
##      1309      1308

```

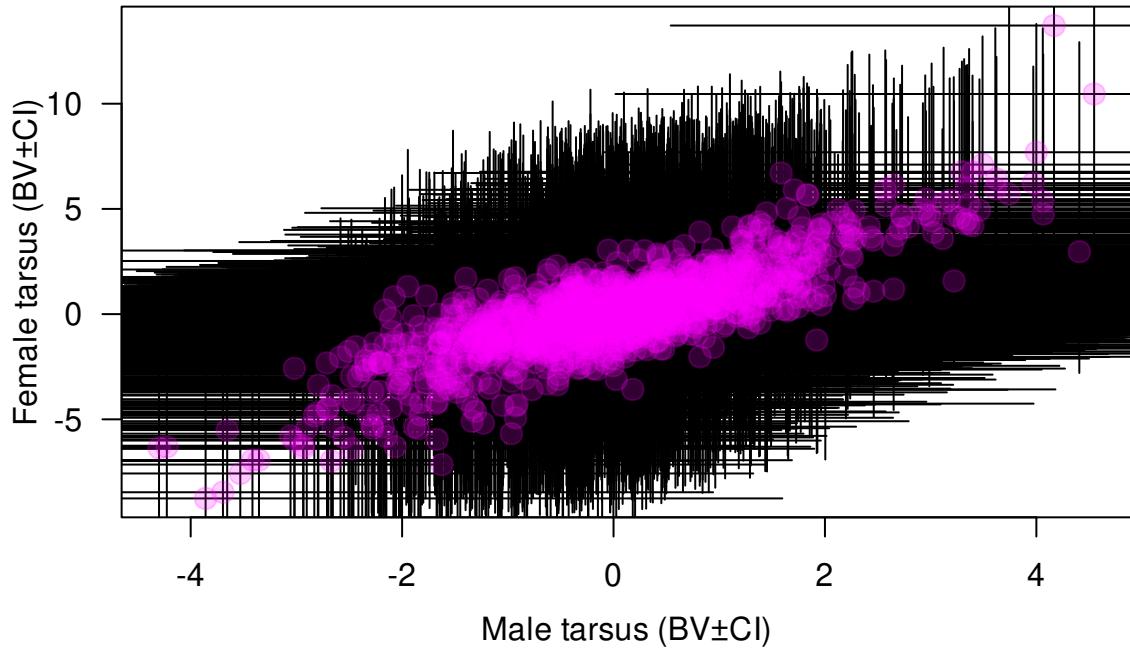
```
DvsS$Trait <- NULL
BLUPS <- reshape(DvsS, v.names = c("BLUP", "CI.lower", "CI.upper"), idvar = "ID", timevar = nrow(BLUPS))
```

```
## [1] 1309
```

```
rownames(BLUPS) <- c()
colnames(BLUPS) <- c("ID", "BLUP.btw", "CI.L.btw", "CI.U.btw", "BLUP.tarsus", "CI.L.tarsus")
summary(BLUPS)
```

	ID	BLUP.btw	CI.L.btw	CI.U.btw
## Length:	1309	Min. : -4.299559	Min. : -9.4393	Min. : 0.5871
## Class :	character	1st Qu.: -0.743429	1st Qu.: -5.2018	1st Qu.: 3.5625
## Mode :	character	Median : -0.000879	Median : -4.3976	Median : 4.5287
		Mean : 0.024573	Mean : -4.3251	Mean : 4.4753
		3rd Qu.: 0.762532	3rd Qu.: -3.4818	3rd Qu.: 5.3499
		Max. : 4.546380	Max. : 0.5408	Max. : 10.9441
##				
##	BLUP.tarsus	CI.L.tarsus	CI.U.tarsus	
##	Min. : -8.75836	Min. : -14.320	Min. : -0.3279	
##	1st Qu.: -1.05894	1st Qu.: -8.157	1st Qu.: 4.9941	
##	Median : 0.06830	Median : -6.720	Median : 6.8745	
##	Mean : 0.07182	Mean : -6.467	Mean : 6.6117	
##	3rd Qu.: 1.16904	3rd Qu.: -4.869	3rd Qu.: 8.3023	
##	Max. : 13.71503	Max. : 1.264	Max. : 16.7611	
##	NA's : 1	NA's : 1	NA's : 1	

```
plot(BLUP.tarsus ~ BLUP.btw, BLUPS, xlab = "", ylab = "", las = 1.2, bty = "o", col = "white")
arrows(x0 = BLUPS$BLUP.btw, y0 = BLUPS$CI.L.tarsus, x1 = BLUPS$BLUP.btw, y1 = BLUPS$CI.U.tarsus)
arrows(x0 = BLUPS$CI.L.btw, y0 = BLUPS$BLUP.tarsus, x1 = BLUPS$CI.U.btw, y1 = BLUPS$BLUP.tarsus)
points(BLUP.tarsus ~ BLUP.btw, BLUPS, pch = 16, col = rgb(1, 0, 1, 0.2), cex = 1.5)
points(BLUP.tarsus ~ BLUP.btw, BLUPS, pch = 1, col = rgb(1, 0, 1, 0.2), cex = c(1.5))
mtext("Male tarsus (BV±CI)", side = 1, line = 2.4)
mtext("Female tarsus (BV±CI)", side = 2, line = 2, las = 3)
```



#

The B matrix used the same code but in a multivariate animal model framework. Here some example code, however due to the nature of the dataset, the cross-sex genetic covariance for birth weight is hard to estimate making difficult to fit this multivariate animal model.

```

prior2.5 <- list(
  G = list(
    G1 = list(V = diag(4), nu = 1.002),
    G2 = list(V = diag(4), nu = 1.002),
    G3 = list(V = diag(4), nu = 1.002)
  ),
  R = list(
    V1 = list(V = diag(4), nu = 1.002)
  )
)
#
model.MultivSex <- MCMCglmm(cbind(tarsus.1, bwt.1, tarsus.2, bwt.2) ~ trait - 1,
  random = ~ us(trait):animal + idh(trait):byear + idh(trait):mother,
  rcov = ~ us(trait):units,
  family = c("gaussian", "gaussian", "gaussian", "gaussian"),
  ginv = list(animal = Ainv), data = gryphon2,

```

```

  nitt = 130000, thin = 100, burnin = 30000,
  prior = prior2.5, verbose = FALSE, pr = TRUE
)
save(model.MultivSex, file = "data/MCMCglmm_model_MultivSex_LongRun.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load(file = "data/MCMCglmm_model_MultivSex_LongRun.rda")
summary(model.MultivSex)

```

```

##
## Iterations = 30001:129901
## Thinning interval = 100
## Sample size = 1000
##
## DIC: 2590.513
##
## G-structure: ~us(trait):animal
##
##                                post.mean   l-95% CI u-95% CI eff.samp
## traittarsus.1:traittarsus.1.animal    5.2542  0.6145315  11.414   28.46
## traitbwt.1:traittarsus.1.animal     1.2682 -0.7026690   3.298   63.97
## traittarsus.2:traittarsus.1.animal    5.9945  0.3482679  13.405   21.59
## traitbwt.2:traittarsus.1.animal     1.4467 -0.3853824   4.110   35.84
## traittarsus.1:traitbwt.1.animal     1.2682 -0.7026690   3.298   63.97
## traitbwt.1:traitbwt.1.animal      1.7891  0.5620761   3.005   97.87
## traittarsus.2:traitbwt.1.animal     0.7719 -1.8038433   4.190   22.64
## traitbwt.2:traitbwt.1.animal      0.9939  0.0009052   2.069   50.24
## traittarsus.1:traittarsus.2.animal    5.9945  0.3482679  13.405   21.59
## traitbwt.1:traittarsus.2.animal     0.7719 -1.8038433   4.190   22.64
## traittarsus.2:traittarsus.2.animal    12.7341  1.4093961  23.690   28.81
## traitbwt.2:traittarsus.2.animal      2.7675 -0.5938976   6.927   20.64
## traittarsus.1:traitbwt.2.animal     1.4467 -0.3853824   4.110   35.84
## traitbwt.1:traitbwt.2.animal      0.9939  0.0009052   2.069   50.24
## traittarsus.2:traitbwt.2.animal      2.7675 -0.5938976   6.927   20.64
## traitbwt.2:traitbwt.2.animal      1.5560  0.2002403   3.046   27.21
##
##                                ~idh(trait):byear
##
##                                post.mean   l-95% CI u-95% CI eff.samp
## traittarsus.1.byear     3.3123  0.9318   6.450   391.3
## traitbwt.1.byear      0.6822  0.2403   1.253   522.1
## traittarsus.2.byear     4.2198  1.3966   7.713   245.2
## traitbwt.2.byear      1.1743  0.5405   1.992   669.6
##
```

```

## ~idh(trait):mother
##
##          post.mean l-95% CI u-95% CI eff.samp
## traittarsus.1.mother      4.858   0.5149   8.841   122.5
## traitbwt.1.mother        1.307   0.5752   2.041   369.0
## traittarsus.2.mother      5.389   0.7457   9.557   140.4
## traitbwt.2.mother        2.003   1.2844   2.770   409.1
##
## R-structure: ~us(trait):units
##
##          post.mean l-95% CI u-95% CI eff.samp
## traittarsus.1:traittarsus.1.units 14.0783  8.6213 20.698  76.808
## traitbwt.1:traittarsus.1.units    4.0764  2.2358  6.357  47.762
## traittarsus.2:traittarsus.1.units -3.6471 -16.9514 14.938  4.746
## traitbwt.2:traittarsus.1.units    -1.3655 -6.7308  4.970  7.185
## traittarsus.1:traitbwt.1.units    4.0764  2.2358  6.357  47.762
## traitbwt.1:traitbwt.1.units       1.7295  0.7344  2.785  57.968
## traittarsus.2:traitbwt.1.units    -1.1455 -5.8008  4.672  6.837
## traitbwt.2:traitbwt.1.units       -0.4245 -2.3300  1.630  7.646
## traittarsus.1:traittarsus.2.units -3.6471 -16.9514 14.938  4.746
## traitbwt.1:traittarsus.2.units    -1.1455 -5.8008  4.672  6.837
## traittarsus.2:traittarsus.2.units 10.8365  0.5947 19.576  26.795
## traitbwt.2:traittarsus.2.units    3.7358 -0.1168  6.848  25.426
## traittarsus.1:traitbwt.2.units    -1.3655 -6.7308  4.970  7.185
## traitbwt.1:traitbwt.2.units       -0.4245 -2.3300  1.630  7.646
## traittarsus.2:traitbwt.2.units    3.7358 -0.1168  6.848  25.426
## traitbwt.2:traitbwt.2.units       1.7825  0.2691  2.916  28.817
##
## Location effects: cbind(tarsus.1, bwt.1, tarsus.2, bwt.2) ~ trait - 1
##
##          post.mean l-95% CI u-95% CI eff.samp pMCMC
## traittarsus.1      20.424   19.488   21.324   484.6 <0.001 ***
## traitbwt.1         6.143    5.686    6.550    596.7 <0.001 ***
## traittarsus.2      20.487   19.421   21.461   587.3 <0.001 ***
## traitbwt.2         8.247    7.744    8.741    876.7 <0.001 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
autocorr(model.MultivSex$VCV)

## , , traittarsus.1:traittarsus.1.animal
##
##           traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                  1.0000000          0.6872795
## Lag 100                 0.8646238          0.6023267
## Lag 500                 0.6217623          0.4701016
```

```

## Lag 1000          0.4759845      0.3306117
## Lag 5000          0.1189988      0.1665003
##     traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0             0.6821642      0.5157628
## Lag 100            0.5837197      0.4557449
## Lag 500            0.4393984      0.3819538
## Lag 1000           0.3626317      0.3277211
## Lag 5000           0.1659171      0.2490413
##     traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0             0.6872795      0.2316436
## Lag 100            0.6023267      0.1917562
## Lag 500            0.4701016      0.1506507
## Lag 1000           0.3306117      0.1026122
## Lag 5000           0.1665003      0.1840457
##     traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0             0.4371184      0.2363148
## Lag 100            0.3790517      0.1969987
## Lag 500            0.3316844      0.1943316
## Lag 1000           0.2706112      0.1653149
## Lag 5000           0.2226417      0.2722600
##     traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0             0.6821642      0.4371184
## Lag 100            0.5837197      0.3790517
## Lag 500            0.4393984      0.3316844
## Lag 1000           0.3626317      0.2706112
## Lag 5000           0.1659171      0.2226417
##     traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0             0.10780644     0.12397199
## Lag 100            0.07023035     0.10315317
## Lag 500            0.03867618     0.09153994
## Lag 1000           0.03903023     0.09577409
## Lag 5000           0.21464371     0.32154133
##     traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0             0.5157628      0.2363148
## Lag 100            0.4557449      0.1969987
## Lag 500            0.3819538      0.1943316
## Lag 1000           0.3277211      0.1653149
## Lag 5000           0.2490413      0.2722600
##     traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             0.12397199     0.07790198
## Lag 100            0.10315317     0.07045248
## Lag 500            0.09153994     0.07781235
## Lag 1000           0.09577409     0.07872349
## Lag 5000           0.32154133     0.36408510
##     traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0             -0.0001991343    0.02533600    0.027261984
## Lag 100            0.0100119397    0.03699313    0.023145145

```

```

## Lag 500      0.0417069693    0.01039048    0.001545709
## Lag 1000     0.0504304342    0.06238542    -0.035646379
## Lag 5000     0.0318527577    0.07041675    0.069731639
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0        -0.01231264    -0.150788349   0.10750858
## Lag 100      -0.02806645    -0.118379074   0.11004526
## Lag 500      0.01040024    -0.006641147   0.07412673
## Lag 1000     -0.02300762    0.007943789   0.04372918
## Lag 5000     -0.01416454    0.056559933   -0.06110098
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0        0.01502226    0.050820670
## Lag 100      0.01411095    0.039228813
## Lag 500      0.05290822    0.008673539
## Lag 1000     0.06760176    -0.017960145
## Lag 5000     0.02399980    -0.029552126
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0          -0.6964793   -0.5806661
## Lag 100        -0.6332557   -0.5256332
## Lag 500        -0.5006482   -0.4211229
## Lag 1000       -0.3965468   -0.3341455
## Lag 5000       -0.1551518   -0.1703888
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0          0.1834123   0.1843936
## Lag 100        0.1989318   0.1972224
## Lag 500        0.2183849   0.2244201
## Lag 1000       0.2393546   0.2533883
## Lag 5000       0.1874294   0.2352808
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0          -0.5806661   -0.2972330
## Lag 100        -0.5256332   -0.2627074
## Lag 500        -0.4211229   -0.1975278
## Lag 1000       -0.3341455   -0.1864674
## Lag 5000       -0.1703888   -0.1779074
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0          0.1148525   0.1045202
## Lag 100        0.1322561   0.1205853
## Lag 500        0.1661494   0.1609281
## Lag 1000       0.2078063   0.2097516
## Lag 5000       0.2120391   0.2411848
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0          0.1834123   0.1148525
## Lag 100        0.1989318   0.1322561
## Lag 500        0.2183849   0.1661494
## Lag 1000       0.2393546   0.2078063
## Lag 5000       0.1874294   0.2120391
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0          -0.09279221  -0.09066377

```

```

## Lag 100           -0.06552157      -0.08137057
## Lag 500           -0.03120806      -0.06842888
## Lag 1000          -0.04688856      -0.07652447
## Lag 5000          -0.22692538      -0.31476665
##   traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.1843936       0.1045202
## Lag 100            0.1972224       0.1205853
## Lag 500            0.2244201       0.1609281
## Lag 1000           0.2533883       0.2097516
## Lag 5000           0.2352808       0.2411848
##   traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.09066377     -0.04305989
## Lag 100             -0.08137057    -0.04967401
## Lag 500             -0.06842888    -0.06533316
## Lag 1000            -0.07652447    -0.05918369
## Lag 5000            -0.31476665    -0.34290345
##
## , , traitbwt.1:traittarsus.1.animal
##
##   traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                0.6872795      1.0000000
## Lag 100              0.5870266      0.8045048
## Lag 500              0.4080911      0.4908098
## Lag 1000             0.3227808      0.3613161
## Lag 5000             0.1778345      0.1860991
##   traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                0.3863431      0.5234222
## Lag 100              0.3311212      0.4545868
## Lag 500              0.2390588      0.3427345
## Lag 1000             0.2351271      0.3004179
## Lag 5000             0.1671396      0.1589172
##   traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                1.0000000      0.7001005
## Lag 100              0.8045048      0.5457846
## Lag 500              0.4908098      0.3331394
## Lag 1000             0.3613161      0.2645656
## Lag 5000             0.1860991      0.1481314
##   traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                0.4680554      0.4293517
## Lag 100              0.3667985      0.3303034
## Lag 500              0.2878690      0.2612182
## Lag 1000             0.2588194      0.2248089
## Lag 5000             0.2468277      0.2268437
##   traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                0.3863431      0.4680554
## Lag 100              0.3311212      0.3667985
## Lag 500              0.2390588      0.2878690

```

```

## Lag 1000          0.2351271          0.2588194
## Lag 5000          0.1671396          0.2468277
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0             0.096601163         0.14997594
## Lag 100            0.063514030         0.11517198
## Lag 500            0.007343656         0.08445112
## Lag 1000           0.062005867         0.13299248
## Lag 5000           0.138711493         0.22091811
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0             0.5234222          0.4293517
## Lag 100            0.4545868          0.3303034
## Lag 500            0.3427345          0.2612182
## Lag 1000           0.3004179          0.2248089
## Lag 5000           0.1589172          0.2268437
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             0.14997594        0.14180302
## Lag 100            0.11517198        0.11177118
## Lag 500            0.08445112        0.09651159
## Lag 1000           0.13299248        0.13501835
## Lag 5000           0.22091811        0.25836873
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0             0.06781216        -0.04857154       -0.001115860
## Lag 100            0.05744185        -0.04365919       -0.003028122
## Lag 500            0.04182580        -0.01683610       -0.013871766
## Lag 1000           0.08409670        0.04075709       -0.088936973
## Lag 5000           0.07211397        0.04755605       0.061879136
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0             -0.026101857       0.015326642      -0.015001711
## Lag 100            -0.026229006      0.004791955      -0.002553299
## Lag 500            -0.004795139       0.018197593      -0.001734987
## Lag 1000           -0.025567829       -0.014419690     -0.054320480
## Lag 5000           -0.011401634       -0.003782365     -0.012255190
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0             0.05171415        -0.01155255
## Lag 100            0.04549989        -0.01648838
## Lag 500            0.09888411        -0.02579001
## Lag 1000           0.07526847        -0.03519968
## Lag 5000           0.07495551        -0.03929467
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0              -0.5374156        -0.7822168
## Lag 100             -0.4631107        -0.6561709
## Lag 500             -0.3115622        -0.4297448
## Lag 1000            -0.2502951        -0.3378405
## Lag 5000             -0.1757000        -0.1812141
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.1247101          0.1399193
## Lag 100             0.1426212          0.1604774

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## Lag 500          0.1604094          0.2063094
## Lag 1000         0.1703454          0.2249688
## Lag 5000         0.1859265          0.1911673
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            -0.7822168          -0.6444049
## Lag 100          -0.6561709          -0.5391921
## Lag 500          -0.4297448          -0.3429426
## Lag 1000         -0.3378405          -0.2744658
## Lag 5000         -0.1812141          -0.1591715
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            0.1003793          0.1067549
## Lag 100          0.1314503          0.1404067
## Lag 500          0.1658186          0.1981668
## Lag 1000         0.1899850          0.2354607
## Lag 5000         0.2090708          0.1964048
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            0.1247101          0.1003793
## Lag 100          0.1426212          0.1314503
## Lag 500          0.1604094          0.1658186
## Lag 1000         0.1703454          0.1899850
## Lag 5000         0.1859265          0.2090708
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0            -0.09677793         -0.12847847
## Lag 100          -0.06212434         -0.10051402
## Lag 500          -0.01671120         -0.07130994
## Lag 1000         -0.07460059         -0.11790351
## Lag 5000         -0.16171357         -0.21125583
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            0.1399193          0.1067549
## Lag 100          0.1604774          0.1404067
## Lag 500          0.2063094          0.1981668
## Lag 1000         0.2249688          0.2354607
## Lag 5000         0.1911673          0.1964048
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0            -0.12847847         -0.10947233
## Lag 100          -0.10051402         -0.09309607
## Lag 500          -0.07130994         -0.08545450
## Lag 1000         -0.11790351         -0.11149356
## Lag 5000         -0.21125583         -0.23203853
##
## , , traittarsus.2:traittarsus.1.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0            0.682164176          0.38634307
## Lag 100          0.591700877          0.34935277
## Lag 500          0.451559860          0.32828629
## Lag 1000         0.378373002          0.27249926

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## Lag 5000          0.004289439      0.08293975
## traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0            1.000000000      0.78206164
## Lag 100          0.87865696      0.70110541
## Lag 500          0.69161215      0.56910648
## Lag 1000         0.56265051      0.44347532
## Lag 5000         -0.04003226     0.01712343
## traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0            0.38634307      0.09679049
## Lag 100          0.34935277      0.10017661
## Lag 500          0.32828629      0.12775760
## Lag 1000         0.27249926      0.11433423
## Lag 5000         0.08293975      0.09000388
## traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            0.71265304      0.46511587
## Lag 100          0.64354820      0.41849123
## Lag 500          0.54967619      0.35510793
## Lag 1000         0.45077032      0.27846473
## Lag 5000         0.02763651      0.08826768
## traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            1.000000000      0.71265304
## Lag 100          0.87865696      0.64354820
## Lag 500          0.69161215      0.54967619
## Lag 1000         0.56265051      0.45077032
## Lag 5000         -0.04003226     0.02763651
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            0.5942621       0.5613010
## Lag 100          0.5307170       0.5141865
## Lag 500          0.4678297       0.4619296
## Lag 1000         0.4001045       0.3816713
## Lag 5000         0.1478270       0.1489578
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            0.78206164      0.46511587
## Lag 100          0.70110541      0.41849123
## Lag 500          0.56910648      0.35510793
## Lag 1000         0.44347532      0.27846473
## Lag 5000         0.01712343      0.08826768
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            0.5613010       0.4107260
## Lag 100          0.5141865       0.3793894
## Lag 500          0.4619296       0.3449678
## Lag 1000         0.3816713       0.2731469
## Lag 5000         0.1489578       0.1431489
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            0.01620495      0.04295234      0.052223393
## Lag 100          0.02029711      0.05648775      0.034069858
## Lag 500          0.04765917      0.03285861      0.007554307

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## Lag 1000      0.03767470      0.07587887      0.003191847
## Lag 5000     -0.02249363      0.05599353      0.040862718
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0       -0.04161269     -0.102952660      0.04055573
## Lag 100     -0.05474586     -0.077407539      0.03752540
## Lag 500     -0.01536913     -0.009985327      0.02041393
## Lag 1000    -0.02242552     -0.016880479      0.02095844
## Lag 5000     0.05382429      0.093039192     -0.03853343
## traittarsus.2.mother traitbwt.2.mother
## Lag 0       -0.15988076      0.11172969
## Lag 100     -0.14920373      0.08235912
## Lag 500     -0.10669450      0.07860308
## Lag 1000    -0.09563279      0.04882603
## Lag 5000     -0.03477033      0.03882608
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.48510213     -0.35513667
## Lag 100          -0.43884380     -0.32128424
## Lag 500          -0.37005382     -0.29419691
## Lag 1000         -0.29875897     -0.25541065
## Lag 5000         -0.06651357     -0.08518695
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            0.1906441      0.2159001
## Lag 100           0.2109187      0.2292953
## Lag 500           0.2433902      0.2525070
## Lag 1000          0.2670207      0.2801991
## Lag 5000          0.2055428      0.2202544
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.35513667     -0.16353731
## Lag 100          -0.32128424     -0.14737332
## Lag 500          -0.29419691     -0.14406290
## Lag 1000         -0.25541065     -0.15449858
## Lag 5000         -0.08518695     -0.09726346
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            0.1588923      0.1751169
## Lag 100           0.1763493      0.1863279
## Lag 500           0.2035612      0.2019259
## Lag 1000          0.2408050      0.2434815
## Lag 5000          0.2188271      0.2190020
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            0.1906441      0.1588923
## Lag 100           0.2109187      0.1763493
## Lag 500           0.2433902      0.2035612
## Lag 1000          0.2670207      0.2408050
## Lag 5000          0.2055428      0.2188271
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.5600992     -0.5208193
## Lag 100          -0.5263626     -0.5010452

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## Lag 500           -0.4620756          -0.4434018
## Lag 1000          -0.4037773          -0.3729971
## Lag 5000          -0.1493219          -0.1529356
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.2159001          0.1751169
## Lag 100             0.2292953          0.1863279
## Lag 500             0.2525070          0.2019259
## Lag 1000            0.2801991          0.2434815
## Lag 5000            0.2202544          0.2190020
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.5208193         -0.3939711
## Lag 100             -0.5010452         -0.3851102
## Lag 500             -0.4434018         -0.3487254
## Lag 1000            -0.3729971         -0.2846600
## Lag 5000            -0.1529356         -0.1449719
##
## , , traitbwt.2:traittarsus.1.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.51576277          0.52342220
## Lag 100            0.44159056          0.43580048
## Lag 500            0.30785687          0.33037079
## Lag 1000            0.24390874          0.26926704
## Lag 5000            -0.05005481          0.04687804
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.782061640          1.00000000
## Lag 100             0.696299215          0.87453824
## Lag 500             0.538889846          0.65219311
## Lag 1000            0.445389577          0.48957859
## Lag 5000            -0.006519724          0.01842864
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.52342220          0.32999449
## Lag 100            0.43580048          0.28148773
## Lag 500            0.33037079          0.26589289
## Lag 1000            0.26926704          0.24323871
## Lag 5000            0.04687804          0.09890983
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.59763628          0.6478170
## Lag 100             0.52643466          0.5573055
## Lag 500             0.44551868          0.4418750
## Lag 1000            0.37736602          0.3442054
## Lag 5000            0.08285167          0.1125165
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.782061640          0.59763628
## Lag 100             0.696299215          0.52643466
## Lag 500             0.538889846          0.44551868
## Lag 1000            0.445389577          0.37736602

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## Lag 5000           -0.006519724          0.08285167
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.5635407          0.6729793
## Lag 100             0.5125673          0.6052812
## Lag 500             0.4465910          0.5277439
## Lag 1000            0.4070664          0.4538221
## Lag 5000            0.1850370          0.1792542
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              1.0000000          0.6478170
## Lag 100             0.87453824         0.5573055
## Lag 500             0.65219311         0.4418750
## Lag 1000            0.48957859         0.3442054
## Lag 5000            0.01842864         0.1125165
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              0.6729793          0.6191252
## Lag 100             0.6052812          0.5529718
## Lag 500             0.5277439          0.4729419
## Lag 1000            0.4538221          0.3884032
## Lag 5000            0.1792542          0.1542326
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              0.063463728         0.02730871        0.051964042
## Lag 100             0.066558472         0.03961464        0.040967359
## Lag 500             0.076728186         0.05374642       -0.012930939
## Lag 1000            0.100842393         0.09752083       -0.062954038
## Lag 5000            -0.006751012        0.04984406        0.002434151
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.09337461         0.02006907       -0.06942502
## Lag 100             -0.09001277         0.02763705       -0.07262374
## Lag 500             -0.04869538         0.02000750       -0.09505843
## Lag 1000            -0.05831890         0.02560894       -0.08617648
## Lag 5000             0.04246982         0.06392166       -0.03722137
## traittarsus.2.mother traitbwt.2.mother
## Lag 0              -0.026436938        0.04826105
## Lag 100             -0.041329867        0.04078039
## Lag 500             -0.028290001        0.03056152
## Lag 1000            -0.036326123        0.04300492
## Lag 5000            -0.008439906        0.00653221
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0              -0.3969629895      -0.44235441
## Lag 100             -0.3457725771      -0.37788435
## Lag 500             -0.2409128696      -0.27612777
## Lag 1000            -0.2064627528      -0.23939733
## Lag 5000            -0.0008284304      -0.05932951
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.1332860          0.1894193
## Lag 100             0.1542105          0.2129929
## Lag 500              0.1772766          0.2426247

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## Lag 1000          0.1956287          0.2577964
## Lag 5000          0.1561845          0.1623293
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0             -0.44235441         -0.3300209
## Lag 100            -0.37788435         -0.2868342
## Lag 500            -0.27612777         -0.2200427
## Lag 1000           -0.23939733         -0.2196364
## Lag 5000           -0.05932951         -0.1105188
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.1392793          0.1845588
## Lag 100             0.1618837          0.2068320
## Lag 500             0.1824833          0.2322375
## Lag 1000            0.2062851          0.2545405
## Lag 5000            0.1708879          0.1679111
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.1332860          0.1392793
## Lag 100             0.1542105          0.1618837
## Lag 500             0.1772766          0.1824833
## Lag 1000            0.1956287          0.2062851
## Lag 5000            0.1561845          0.1708879
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              -0.5541670         -0.6200400
## Lag 100             -0.5163721         -0.5823652
## Lag 500             -0.4591717         -0.5082023
## Lag 1000            -0.4081394         -0.4277328
## Lag 5000            -0.1850899         -0.1717978
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.1894193          0.1845588
## Lag 100             0.2129929          0.2068320
## Lag 500             0.2426247          0.2322375
## Lag 1000            0.2577964          0.2545405
## Lag 5000            0.1623293          0.1679111
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.6200400         -0.5649761
## Lag 100             -0.5823652         -0.5353605
## Lag 500             -0.5082023         -0.4627844
## Lag 1000            -0.4277328         -0.3777429
## Lag 5000            -0.1717978         -0.1507443
##
## , , traittarsus.1:traitbwt.1.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.6872795          1.0000000
## Lag 100            0.5870266          0.8045048
## Lag 500            0.4080911          0.4908098
## Lag 1000            0.3227808          0.3613161
## Lag 5000            0.1778345          0.1860991

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##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                  0.3863431           0.5234222
## Lag 100                 0.3311212           0.4545868
## Lag 500                 0.2390588           0.3427345
## Lag 1000                0.2351271           0.3004179
## Lag 5000                0.1671396           0.1589172
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  1.0000000           0.7001005
## Lag 100                 0.8045048           0.5457846
## Lag 500                 0.4908098           0.3331394
## Lag 1000                0.3613161           0.2645656
## Lag 5000                0.1860991           0.1481314
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                  0.4680554           0.4293517
## Lag 100                 0.3667985           0.3303034
## Lag 500                 0.2878690           0.2612182
## Lag 1000                0.2588194           0.2248089
## Lag 5000                0.2468277           0.2268437
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                  0.3863431           0.4680554
## Lag 100                 0.3311212           0.3667985
## Lag 500                 0.2390588           0.2878690
## Lag 1000                0.2351271           0.2588194
## Lag 5000                0.1671396           0.2468277
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                  0.096601163          0.14997594
## Lag 100                 0.063514030          0.11517198
## Lag 500                 0.007343656          0.08445112
## Lag 1000                0.062005867          0.13299248
## Lag 5000                0.138711493          0.22091811
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                  0.5234222           0.4293517
## Lag 100                 0.4545868           0.3303034
## Lag 500                 0.3427345           0.2612182
## Lag 1000                0.3004179           0.2248089
## Lag 5000                0.1589172           0.2268437
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                  0.14997594          0.14180302
## Lag 100                 0.11517198          0.11177118
## Lag 500                 0.08445112          0.09651159
## Lag 1000                0.13299248          0.13501835
## Lag 5000                0.22091811          0.25836873
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0                  0.06781216          -0.04857154          -0.001115860
## Lag 100                 0.05744185          -0.04365919          -0.003028122
## Lag 500                 0.04182580          -0.01683610          -0.013871766
## Lag 1000                0.08409670          0.04075709          -0.088936973

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## Lag 5000          0.07211397      0.04755605      0.061879136
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.026101857     0.015326642     -0.015001711
## Lag 100           -0.026229006     0.004791955     -0.002553299
## Lag 500           -0.004795139     0.018197593     -0.001734987
## Lag 1000          -0.025567829     -0.014419690     -0.054320480
## Lag 5000          -0.011401634     -0.003782365     -0.012255190
## traittarsus.2.mother traitbwt.2.mother
## Lag 0             0.05171415      -0.01155255
## Lag 100            0.04549989      -0.01648838
## Lag 500            0.09888411      -0.02579001
## Lag 1000           0.07526847      -0.03519968
## Lag 5000           0.07495551      -0.03929467
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0              -0.5374156      -0.7822168
## Lag 100             -0.4631107      -0.6561709
## Lag 500             -0.3115622      -0.4297448
## Lag 1000            -0.2502951      -0.3378405
## Lag 5000            -0.1757000      -0.1812141
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0               0.1247101      0.1399193
## Lag 100              0.1426212      0.1604774
## Lag 500              0.1604094      0.2063094
## Lag 1000             0.1703454      0.2249688
## Lag 5000             0.1859265      0.1911673
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0              -0.7822168      -0.6444049
## Lag 100             -0.6561709      -0.5391921
## Lag 500             -0.4297448      -0.3429426
## Lag 1000            -0.3378405      -0.2744658
## Lag 5000            -0.1812141      -0.1591715
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0               0.1003793      0.1067549
## Lag 100              0.1314503      0.1404067
## Lag 500              0.1658186      0.1981668
## Lag 1000             0.1899850      0.2354607
## Lag 5000             0.2090708      0.1964048
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0               0.1247101      0.1003793
## Lag 100              0.1426212      0.1314503
## Lag 500              0.1604094      0.1658186
## Lag 1000             0.1703454      0.1899850
## Lag 5000             0.1859265      0.2090708
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              -0.09677793     -0.12847847
## Lag 100             -0.06212434     -0.10051402
## Lag 500             -0.01671120     -0.07130994

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## Lag 1000           -0.07460059          -0.11790351
## Lag 5000           -0.16171357          -0.21125583
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.1399193           0.1067549
## Lag 100             0.1604774           0.1404067
## Lag 500             0.2063094           0.1981668
## Lag 1000            0.2249688           0.2354607
## Lag 5000            0.1911673           0.1964048
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.12847847          -0.10947233
## Lag 100             -0.10051402          -0.09309607
## Lag 500             -0.07130994          -0.08545450
## Lag 1000            -0.11790351          -0.11149356
## Lag 5000            -0.21125583          -0.23203853
##
## , , traitbwt.1:traitbwt.1.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.23164362          0.70010050
## Lag 100            0.18054608          0.54456316
## Lag 500             0.08781934          0.30301892
## Lag 1000            0.03769032          0.17531595
## Lag 5000            0.09013727          0.08812775
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.09679049          0.329994492
## Lag 100            0.07358032          0.280145064
## Lag 500             0.03161374          0.197610680
## Lag 1000            0.07044623          0.170452855
## Lag 5000            0.09611101          0.004847697
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.70010050          1.0000000
## Lag 100            0.54456316          0.6856952
## Lag 500             0.30301892          0.3598203
## Lag 1000            0.17531595          0.2634115
## Lag 5000            0.08812775          0.0581774
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.2789729           0.50673141
## Lag 100            0.2383906           0.37724913
## Lag 500             0.1850306           0.26913927
## Lag 1000            0.1760392           0.21403804
## Lag 5000            0.1959228           0.08822723
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.09679049          0.2789729
## Lag 100            0.07358032          0.2383906
## Lag 500             0.03161374          0.1850306
## Lag 1000            0.07044623          0.1760392
## Lag 5000            0.09611101          0.1959228

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##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                      0.10548333      0.17131088
## Lag 100                     0.08666667      0.15141502
## Lag 500                     0.03855034      0.12592562
## Lag 1000                    0.09753802      0.15461525
## Lag 5000                    0.05868589      0.07881607
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                      0.329994492     0.50673141
## Lag 100                     0.280145064     0.37724913
## Lag 500                     0.197610680     0.26913927
## Lag 1000                    0.170452855     0.21403804
## Lag 5000                    0.004847697     0.08822723
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                      0.17131088      0.21046169
## Lag 100                     0.15141502      0.17525516
## Lag 500                     0.12592562      0.17738787
## Lag 1000                    0.15461525      0.17597147
## Lag 5000                    0.07881607      0.08328384
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0                      0.065188114    -0.054239888   -0.0315843380
## Lag 100                     0.038096586    -0.064724456   -0.0570202816
## Lag 500                     -0.002027739   0.020824115   0.0009269064
## Lag 1000                    0.057546907    0.008429484   -0.0456402525
## Lag 5000                    0.105389029    0.003598701   0.0158576632
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0                      -0.02489988     0.063535708   -0.301057181
## Lag 100                     -0.02027214     0.044670327   -0.205303237
## Lag 500                     -0.01893478     0.000885945   -0.083381850
## Lag 1000                    -0.01593842     -0.053672083   -0.098869582
## Lag 5000                    -0.06322669     0.022199101   0.007755258
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0                      0.02830176     -0.06829043
## Lag 100                     0.01508424     -0.03965583
## Lag 500                     0.07970255     -0.07134929
## Lag 1000                    0.03715760     -0.03341572
## Lag 5000                    0.05894613     -0.03420041
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                      -0.224526532    -0.57857216
## Lag 100                     -0.156949752    -0.45706691
## Lag 500                     -0.038181523    -0.25766540
## Lag 1000                    -0.006974297    -0.17163469
## Lag 5000                    -0.082791880    -0.06775113
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                      0.08095300      0.11864383
## Lag 100                     0.08975567      0.13271891
## Lag 500                     0.10744809      0.16596526
## Lag 1000                    0.10586708      0.15987797

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## Lag 5000          0.11020681          0.08507017
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0             -0.57857216         -0.72140390
## Lag 100            -0.45706691         -0.56413893
## Lag 500            -0.25766540         -0.34043397
## Lag 1000           -0.17163469         -0.22135712
## Lag 5000           -0.06775113         -0.03117589
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.1206260          0.15466547
## Lag 100             0.1358748          0.17988396
## Lag 500             0.1545988          0.21095632
## Lag 1000            0.1778949          0.23064318
## Lag 5000            0.1189483          0.08708938
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.08095300         0.1206260
## Lag 100             0.08975567         0.1358748
## Lag 500             0.10744809         0.1545988
## Lag 1000            0.10586708         0.1778949
## Lag 5000            0.11020681         0.1189483
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              -0.11346055        -0.1672807
## Lag 100             -0.08013985        -0.1375357
## Lag 500             -0.05850422        -0.1306059
## Lag 1000            -0.11497643        -0.1592872
## Lag 5000            -0.07297426        -0.0658664
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.11864383          0.15466547
## Lag 100             0.13271891          0.17988396
## Lag 500             0.16596526          0.21095632
## Lag 1000            0.15987797          0.23064318
## Lag 5000            0.08507017          0.08708938
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.1672807          -0.1878114
## Lag 100             -0.1375357          -0.1655553
## Lag 500             -0.1306059          -0.1703498
## Lag 1000            -0.1592872          -0.1664878
## Lag 5000            -0.0658664          -0.0614103
##
## , , traittarsus.2:traitbwt.1.animal
##
## traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.43711841          0.4680554
## Lag 100             0.38614287          0.3822935
## Lag 500             0.32973015          0.2828059
## Lag 1000            0.30898433          0.3053279
## Lag 5000            0.06068506          0.1038801
## traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal

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## Lag 0           0.71265304      0.59763628
## Lag 100        0.64495665      0.53517485
## Lag 500        0.56187680      0.46435370
## Lag 1000       0.52770068      0.43493174
## Lag 5000       -0.04596197     -0.09919957
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0           0.4680554       0.27897287
## Lag 100         0.3822935       0.24804717
## Lag 500         0.2828059       0.18131034
## Lag 1000        0.3053279       0.20383689
## Lag 5000        0.1038801       0.07466278
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           1.00000000     0.75320258
## Lag 100         0.84412396     0.63847339
## Lag 500         0.63408467     0.47179404
## Lag 1000        0.55041345     0.40200369
## Lag 5000        0.07989653     0.02863099
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           0.71265304     1.00000000
## Lag 100         0.64495665     0.84412396
## Lag 500         0.56187680     0.63408467
## Lag 1000        0.52770068     0.55041345
## Lag 5000       -0.04596197     0.07989653
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           0.39609257     0.50085427
## Lag 100         0.37249703     0.45415166
## Lag 500         0.36147616     0.41592880
## Lag 1000        0.38702161     0.40262387
## Lag 5000        0.05902071     0.05027704
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           0.59763628     0.75320258
## Lag 100         0.53517485     0.63847339
## Lag 500         0.46435370     0.47179404
## Lag 1000        0.43493174     0.40200369
## Lag 5000       -0.09919957     0.02863099
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0           0.50085427     0.47574281
## Lag 100         0.45415166     0.42373764
## Lag 500         0.41592880     0.36577921
## Lag 1000        0.40262387     0.31479935
## Lag 5000        0.05027704     0.04228229
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           0.029421346    0.01370326    0.08861601
## Lag 100         0.024425366    0.01320865    0.06508810
## Lag 500         0.050837464    0.03060238    0.05880699
## Lag 1000        0.065685010    0.04013922    0.02624809
## Lag 5000       -0.006279369    0.01390888    0.06953734

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## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0      -0.06566102      0.010266822     -0.002921727
## Lag 100    -0.05957647      0.001133316     -0.025849735
## Lag 500    -0.03310752      -0.035302602     -0.011049599
## Lag 1000   -0.01325750      -0.005132729     -0.030094999
## Lag 5000   0.07737072      0.051936326      0.028222378
## traittarsus.2.mother traitbwt.2.mother
## Lag 0      -0.01173970      0.004891297
## Lag 100    -0.01984404      0.001495609
## Lag 500    -0.01047992      0.019305497
## Lag 1000   -0.06166849      0.035361493
## Lag 5000   -0.03572076      0.037357859
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0      -0.3729388      -0.3850991
## Lag 100    -0.3393916      -0.3381615
## Lag 500    -0.2662669      -0.2603097
## Lag 1000   -0.2581177      -0.2862286
## Lag 5000   -0.1138844      -0.1255711
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0      0.1304165      0.1778597
## Lag 100    0.1523428      0.1937314
## Lag 500    0.1726064      0.2144724
## Lag 1000   0.1844905      0.2275146
## Lag 5000   0.1876933      0.1730132
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0      -0.3850991      -0.2793636
## Lag 100    -0.3381615      -0.2451230
## Lag 500    -0.2603097      -0.1889420
## Lag 1000   -0.2862286      -0.2310740
## Lag 5000   -0.1255711      -0.1098246
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0      0.1142625      0.1618783
## Lag 100    0.1516538      0.1886145
## Lag 500    0.1942499      0.2259939
## Lag 1000   0.1947749      0.2314168
## Lag 5000   0.1938069      0.1610221
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0      0.1304165      0.1142625
## Lag 100    0.1523428      0.1516538
## Lag 500    0.1726064      0.1942499
## Lag 1000   0.1844905      0.1947749
## Lag 5000   0.1876933      0.1938069
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0      -0.41858410     -0.47291409
## Lag 100    -0.40815749     -0.45309828
## Lag 500    -0.38612479     -0.40783101
## Lag 1000   -0.40349965     -0.40278702

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## Lag 5000           -0.06617356          -0.06076514
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.1778597           0.1618783
## Lag 100             0.1937314           0.1886145
## Lag 500             0.2144724           0.2259939
## Lag 1000            0.2275146           0.2314168
## Lag 5000            0.1730132           0.1610221
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.47291409          -0.4396042
## Lag 100             -0.45309828          -0.4158969
## Lag 500             -0.40783101          -0.3577962
## Lag 1000            -0.40278702          -0.3301300
## Lag 5000            -0.06076514          -0.0490055
##
## , , traitbwt.2:traitbwt.1.animal
##
## traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.23631479          0.42935174
## Lag 100             0.19631829          0.33111043
## Lag 500             0.13140023          0.20166811
## Lag 1000            0.13084790          0.23051272
## Lag 5000            -0.05046358          0.04857134
## traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.4651159           0.6478170
## Lag 100             0.4250027           0.5751741
## Lag 500             0.3644232           0.4605468
## Lag 1000            0.3695356           0.4130944
## Lag 5000            -0.0748654          -0.1705451
## traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.42935174          0.50673141
## Lag 100             0.33111043          0.36623376
## Lag 500             0.20166811          0.24599792
## Lag 1000            0.23051272          0.26598499
## Lag 5000            0.04857134          0.06879203
## traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.75320258          1.00000000
## Lag 100             0.64643730          0.80429897
## Lag 500             0.48359147          0.55454840
## Lag 1000            0.43668798          0.44468678
## Lag 5000            0.08606661          -0.01807979
## traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.4651159           0.75320258
## Lag 100             0.4250027           0.64643730
## Lag 500             0.3644232           0.48359147
## Lag 1000            0.3695356           0.43668798
## Lag 5000            -0.0748654          0.08606661
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal

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## Lag 0           0.35637772          0.518702070
## Lag 100        0.34579510          0.481108274
## Lag 500        0.32649959          0.434843612
## Lag 1000       0.37931051          0.436821292
## Lag 5000       0.03852246          0.006399124
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           0.6478170          1.000000000
## Lag 100         0.5751741          0.80429897
## Lag 500         0.4605468          0.55454840
## Lag 1000        0.4130944          0.44468678
## Lag 5000        -0.1705451         -0.01807979
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0           0.518702070          0.64907876
## Lag 100         0.481108274          0.56161598
## Lag 500         0.434843612          0.47984513
## Lag 1000        0.436821292          0.40938531
## Lag 5000        0.006399124         -0.01977366
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           0.034800266          0.007431509          0.10446126
## Lag 100         0.051936557          -0.001493245         0.08007660
## Lag 500         0.071129943          0.067797207         0.05478615
## Lag 1000        0.103624443          0.063225351         0.01489821
## Lag 5000        -0.000397122          0.008471637         -0.01350094
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0           -0.08333568          0.100888337         -0.21005355
## Lag 100         -0.07762909          0.079323546         -0.17806133
## Lag 500         -0.03825641          -0.007478373        -0.11051022
## Lag 1000        -0.02784742          0.017077993         -0.10296388
## Lag 5000        0.04558001          0.022598106          0.02483781
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0           0.018470325          -0.095372303
## Lag 100         -0.002048636          -0.029630368
## Lag 500         0.014260425          -0.034300090
## Lag 1000        -0.048567932          0.067004499
## Lag 5000        -0.025983993          0.005631837
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.25474138          -0.36301593
## Lag 100         -0.21067390          -0.29399524
## Lag 500         -0.11042093          -0.17823668
## Lag 1000        -0.13264168          -0.21582496
## Lag 5000        0.02716341          -0.03930422
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0           0.06808542          0.1473359
## Lag 100         0.08333997          0.1644163
## Lag 500         0.11004808          0.1967466
## Lag 1000        0.13239890          0.2087490
## Lag 5000        0.14121855          0.1213850

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##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  -0.36301593      -0.39024104
## Lag 100                 -0.29399524      -0.30803165
## Lag 500                 -0.17823668      -0.22428734
## Lag 1000                -0.21582496      -0.24800908
## Lag 5000                -0.03930422      -0.07900553
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0                   0.09800038       0.1786089
## Lag 100                  0.12840120       0.2150929
## Lag 500                  0.17479984       0.2654006
## Lag 1000                 0.19253125       0.2644751
## Lag 5000                 0.13671344       0.1087463
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0                   0.06808542       0.09800038
## Lag 100                  0.08333997       0.12840120
## Lag 500                  0.11004808       0.17479984
## Lag 1000                 0.13239890       0.19253125
## Lag 5000                 0.14121855       0.13671344
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0                   -0.38430674      -0.49584021
## Lag 100                  -0.36892511      -0.47162690
## Lag 500                  -0.35875583      -0.43506761
## Lag 1000                 -0.38190122      -0.42199356
## Lag 5000                 -0.04674851      -0.01287201
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0                   0.1473359        0.1786089
## Lag 100                  0.1644163        0.2150929
## Lag 500                  0.1967466        0.2654006
## Lag 1000                 0.2087490        0.2644751
## Lag 5000                 0.1213850        0.1087463
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0                   -0.49584021      -0.56988609
## Lag 100                  -0.47162690      -0.53124887
## Lag 500                  -0.43506761      -0.45896948
## Lag 1000                 -0.42199356      -0.40139643
## Lag 5000                 -0.01287201      0.01366036
##
## , , traittarsus.1:traittarsus.2.animal
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                   0.682164176       0.38634307
## Lag 100                  0.591700877       0.34935277
## Lag 500                  0.451559860       0.32828629
## Lag 1000                 0.378373002       0.27249926
## Lag 5000                 0.004289439       0.08293975
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                   1.000000000       0.78206164

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## Lag 100          0.87865696          0.70110541
## Lag 500          0.69161215          0.56910648
## Lag 1000         0.56265051          0.44347532
## Lag 5000         -0.04003226         0.01712343
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0            0.38634307          0.09679049
## Lag 100          0.34935277          0.10017661
## Lag 500          0.32828629          0.12775760
## Lag 1000         0.27249926          0.11433423
## Lag 5000         0.08293975          0.09000388
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            0.71265304          0.46511587
## Lag 100          0.64354820          0.41849123
## Lag 500          0.54967619          0.35510793
## Lag 1000         0.45077032          0.27846473
## Lag 5000         0.02763651          0.08826768
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            1.00000000          0.71265304
## Lag 100          0.87865696          0.64354820
## Lag 500          0.69161215          0.54967619
## Lag 1000         0.56265051          0.45077032
## Lag 5000         -0.04003226         0.02763651
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            0.5942621           0.5613010
## Lag 100          0.5307170           0.5141865
## Lag 500          0.4678297           0.4619296
## Lag 1000         0.4001045           0.3816713
## Lag 5000         0.1478270           0.1489578
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            0.78206164          0.46511587
## Lag 100          0.70110541          0.41849123
## Lag 500          0.56910648          0.35510793
## Lag 1000         0.44347532          0.27846473
## Lag 5000         0.01712343          0.08826768
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            0.5613010           0.4107260
## Lag 100          0.5141865           0.3793894
## Lag 500          0.4619296           0.3449678
## Lag 1000         0.3816713           0.2731469
## Lag 5000         0.1489578           0.1431489
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            0.01620495          0.04295234          0.052223393
## Lag 100          0.02029711          0.05648775          0.034069858
## Lag 500          0.04765917          0.03285861          0.007554307
## Lag 1000         0.03767470          0.07587887          0.003191847
## Lag 5000         -0.02249363          0.05599353          0.040862718
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother

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## Lag 0      -0.04161269    -0.102952660   0.04055573
## Lag 100     -0.05474586   -0.077407539   0.03752540
## Lag 500     -0.01536913   -0.009985327   0.02041393
## Lag 1000    -0.02242552   -0.016880479   0.02095844
## Lag 5000    0.05382429    0.093039192  -0.03853343
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0      -0.15988076    0.11172969
## Lag 100     -0.14920373   0.08235912
## Lag 500     -0.10669450   0.07860308
## Lag 1000    -0.09563279   0.04882603
## Lag 5000    -0.03477033   0.03882608
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0      -0.48510213   -0.35513667
## Lag 100     -0.43884380   -0.32128424
## Lag 500     -0.37005382   -0.29419691
## Lag 1000    -0.29875897   -0.25541065
## Lag 5000    -0.06651357   -0.08518695
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0      0.1906441    0.2159001
## Lag 100    0.2109187    0.2292953
## Lag 500    0.2433902    0.2525070
## Lag 1000   0.2670207    0.2801991
## Lag 5000   0.2055428    0.2202544
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0      -0.35513667  -0.16353731
## Lag 100     -0.32128424  -0.14737332
## Lag 500     -0.29419691  -0.14406290
## Lag 1000    -0.25541065  -0.15449858
## Lag 5000    -0.08518695  -0.09726346
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0      0.1588923    0.1751169
## Lag 100    0.1763493    0.1863279
## Lag 500    0.2035612    0.2019259
## Lag 1000   0.2408050    0.2434815
## Lag 5000   0.2188271    0.2190020
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0      0.1906441    0.1588923
## Lag 100    0.2109187    0.1763493
## Lag 500    0.2433902    0.2035612
## Lag 1000   0.2670207    0.2408050
## Lag 5000   0.2055428    0.2188271
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0      -0.5600992   -0.5208193
## Lag 100     -0.5263626   -0.5010452
## Lag 500     -0.4620756   -0.4434018
## Lag 1000    -0.4037773   -0.3729971
## Lag 5000    -0.1493219   -0.1529356

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##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0                  0.2159001      0.1751169
## Lag 100                 0.2292953      0.1863279
## Lag 500                 0.2525070      0.2019259
## Lag 1000                0.2801991      0.2434815
## Lag 5000                0.2202544      0.2190020
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0                  -0.5208193     -0.3939711
## Lag 100                 -0.5010452     -0.3851102
## Lag 500                 -0.4434018     -0.3487254
## Lag 1000                -0.3729971     -0.2846600
## Lag 5000                -0.1529356     -0.1449719
##
## , , traitbwt.1:traittarsus.2.animal
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                  0.43711841     0.4680554
## Lag 100                 0.38614287     0.3822935
## Lag 500                 0.32973015     0.2828059
## Lag 1000                0.30898433     0.3053279
## Lag 5000                0.06068506     0.1038801
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                  0.71265304     0.59763628
## Lag 100                 0.64495665     0.53517485
## Lag 500                 0.56187680     0.46435370
## Lag 1000                0.52770068     0.43493174
## Lag 5000                -0.04596197    -0.09919957
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  0.4680554      0.27897287
## Lag 100                 0.3822935      0.24804717
## Lag 500                 0.2828059      0.18131034
## Lag 1000                0.3053279      0.20383689
## Lag 5000                0.1038801      0.07466278
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                  1.00000000     0.75320258
## Lag 100                 0.84412396     0.63847339
## Lag 500                 0.63408467     0.47179404
## Lag 1000                0.55041345     0.40200369
## Lag 5000                0.07989653     0.02863099
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                  0.71265304     1.00000000
## Lag 100                 0.64495665     0.84412396
## Lag 500                 0.56187680     0.63408467
## Lag 1000                0.52770068     0.55041345
## Lag 5000                -0.04596197    0.07989653
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                  0.39609257     0.50085427

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## Lag 100           0.37249703      0.45415166
## Lag 500           0.36147616      0.41592880
## Lag 1000          0.38702161      0.40262387
## Lag 5000          0.05902071      0.05027704
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0             0.59763628      0.75320258
## Lag 100           0.53517485      0.63847339
## Lag 500           0.46435370      0.47179404
## Lag 1000          0.43493174      0.40200369
## Lag 5000          -0.09919957     0.02863099
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             0.50085427      0.47574281
## Lag 100           0.45415166      0.42373764
## Lag 500           0.41592880      0.36577921
## Lag 1000          0.40262387      0.31479935
## Lag 5000          0.05027704      0.04228229
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0             0.029421346     0.01370326     0.08861601
## Lag 100           0.024425366     0.01320865     0.06508810
## Lag 500           0.050837464     0.03060238     0.05880699
## Lag 1000          0.065685010     0.04013922     0.02624809
## Lag 5000          -0.006279369    0.01390888     0.06953734
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0             -0.06566102     0.010266822    -0.002921727
## Lag 100           -0.05957647     0.001133316    -0.025849735
## Lag 500           -0.03310752     -0.035302602   -0.011049599
## Lag 1000          -0.01325750     -0.005132729   -0.030094999
## Lag 5000          0.07737072     0.051936326    0.028222378
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0             -0.01173970     0.004891297
## Lag 100           -0.01984404     0.001495609
## Lag 500           -0.01047992     0.019305497
## Lag 1000          -0.06166849     0.035361493
## Lag 5000          -0.03572076     0.037357859
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0             -0.3729388     -0.3850991
## Lag 100           -0.3393916     -0.3381615
## Lag 500           -0.2662669     -0.2603097
## Lag 1000          -0.2581177     -0.2862286
## Lag 5000          -0.1138844     -0.1255711
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0             0.1304165      0.1778597
## Lag 100           0.1523428      0.1937314
## Lag 500           0.1726064      0.2144724
## Lag 1000          0.1844905      0.2275146
## Lag 5000          0.1876933      0.1730132
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units

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## Lag 0           -0.3850991      -0.2793636
## Lag 100        -0.3381615      -0.2451230
## Lag 500        -0.2603097      -0.1889420
## Lag 1000       -0.2862286      -0.2310740
## Lag 5000       -0.1255711      -0.1098246
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0           0.1142625       0.1618783
## Lag 100         0.1516538       0.1886145
## Lag 500         0.1942499       0.2259939
## Lag 1000        0.1947749       0.2314168
## Lag 5000        0.1938069       0.1610221
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0           0.1304165       0.1142625
## Lag 100         0.1523428       0.1516538
## Lag 500         0.1726064       0.1942499
## Lag 1000        0.1844905       0.1947749
## Lag 5000        0.1876933       0.1938069
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.41858410     -0.47291409
## Lag 100         -0.40815749     -0.45309828
## Lag 500         -0.38612479     -0.40783101
## Lag 1000        -0.40349965     -0.40278702
## Lag 5000        -0.06617356     -0.06076514
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0           0.1778597       0.1618783
## Lag 100         0.1937314       0.1886145
## Lag 500         0.2144724       0.2259939
## Lag 1000        0.2275146       0.2314168
## Lag 5000        0.1730132       0.1610221
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           -0.47291409     -0.4396042
## Lag 100         -0.45309828     -0.4158969
## Lag 500         -0.40783101     -0.3577962
## Lag 1000        -0.40278702     -0.3301300
## Lag 5000        -0.06076514     -0.0490055
##
## , , traittarsus.2:traittarsus.2.animal
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0           0.107806440      0.09660116
## Lag 100         0.063335645      0.07671143
## Lag 500         0.015199121      0.08067887
## Lag 1000        -0.004286261     0.05833877
## Lag 5000        -0.070723679     -0.04217947
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0           0.5942621       0.5635407
## Lag 100         0.5173844       0.5077019

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## Lag 500           0.4100415          0.4261910
## Lag 1000          0.3053499          0.3088610
## Lag 5000          -0.2062339         -0.1851005
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.09660116        0.10548333
## Lag 100             0.07671143        0.09974202
## Lag 500             0.08067887        0.11354731
## Lag 1000            0.05833877        0.12159525
## Lag 5000            -0.04217947       -0.06338610
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.3960926          0.3563777
## Lag 100             0.3697695          0.3333969
## Lag 500             0.3480333          0.3176505
## Lag 1000            0.2823012          0.2658600
## Lag 5000            -0.2062348         -0.1460308
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.5942621          0.3960926
## Lag 100             0.5173844          0.3697695
## Lag 500             0.4100415          0.3480333
## Lag 1000            0.3053499          0.2823012
## Lag 5000            -0.2062339         -0.2062348
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              1.0000000          0.89913093
## Lag 100             0.90510783         0.83322028
## Lag 500             0.73781155         0.70412544
## Lag 1000            0.57269345         0.54745411
## Lag 5000            -0.03830606        -0.09250669
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.5635407          0.3563777
## Lag 100             0.5077019          0.3333969
## Lag 500             0.4261910          0.3176505
## Lag 1000            0.3088610          0.2658600
## Lag 5000            -0.1851005         -0.1460308
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              0.89913093         0.6844205
## Lag 100             0.83322028         0.6434324
## Lag 500             0.70412544         0.5640206
## Lag 1000            0.54745411         0.4452587
## Lag 5000            -0.09250669        -0.0996112
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              0.035961574        -0.017469135        0.007171364
## Lag 100             0.026817555        -0.002104493        0.011633070
## Lag 500             0.008446146        -0.011130916        0.013837207
## Lag 1000            0.029054326        0.040567400        0.023793648
## Lag 5000            -0.032575030        0.020439004        -0.049218434
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.08333086         0.04651404        -0.06375729

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## Lag 100      -0.08053153      0.06577446      -0.05659105
## Lag 500      -0.03847864      0.11200938      -0.07300512
## Lag 1000     -0.04249559      0.08134977      -0.07101148
## Lag 5000     0.07016400      0.02226530      0.03188352
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0        -0.34477064      0.15871038
## Lag 100     -0.31220276      0.12040861
## Lag 500     -0.20777383      0.08824805
## Lag 1000    -0.12481403      0.05105578
## Lag 5000    0.01048624      0.01454861
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0        -0.10376953      -0.10398125
## Lag 100     -0.07294547      -0.07699566
## Lag 500     -0.06977230      -0.08979287
## Lag 1000    -0.03903802      -0.06150263
## Lag 5000    0.09531506      0.08613972
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0        0.18812698      0.21480017
## Lag 100     0.19528250      0.21914991
## Lag 500     0.19556551      0.21256430
## Lag 1000    0.18892755      0.20764610
## Lag 5000    0.06665558      0.07468684
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0        -0.10398125      -0.10182068
## Lag 100     -0.07699566      -0.08777589
## Lag 500     -0.08979287      -0.09511315
## Lag 1000    -0.06150263      -0.09765668
## Lag 5000    0.08613972      0.05634093
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0        0.22270138      0.23393547
## Lag 100     0.22287169      0.23382912
## Lag 500     0.20351827      0.21079497
## Lag 1000    0.20537971      0.21383439
## Lag 5000    0.03798515      0.03641653
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0        0.18812698      0.22270138
## Lag 100     0.19528250      0.22287169
## Lag 500     0.19556551      0.20351827
## Lag 1000    0.18892755      0.20537971
## Lag 5000    0.06665558      0.03798515
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0        -0.9107369      -0.84107771
## Lag 100     -0.8633544      -0.80717702
## Lag 500     -0.7252004      -0.68217338
## Lag 1000    -0.5917870      -0.54946518
## Lag 5000    0.0476902      0.09830838
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units

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## Lag 0           0.21480017   0.23393547
## Lag 100        0.21914991   0.23382912
## Lag 500        0.21256430   0.21079497
## Lag 1000       0.20764610   0.21383439
## Lag 5000       0.07468684   0.03641653
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           -0.84107771  -0.6791085
## Lag 100         -0.80717702  -0.6539764
## Lag 500         -0.68217338  -0.5583969
## Lag 1000        -0.54946518  -0.4464232
## Lag 5000        0.09830838   0.1028304
##
## , , traitbwt.2:traittarsus.2.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0           0.12397199   0.14997594
## Lag 100         0.08643519   0.11314881
## Lag 500         0.03605225   0.09323271
## Lag 1000        0.03084094   0.10586940
## Lag 5000        -0.09632022  -0.03936224
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0           0.5613010    0.6729793
## Lag 100         0.5033508    0.6010903
## Lag 500         0.4274531    0.5133373
## Lag 1000        0.3517565    0.4086869
## Lag 5000        -0.1677216   -0.1581302
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0           0.14997594   0.17131088
## Lag 100         0.11314881   0.15721379
## Lag 500         0.09323271   0.15933093
## Lag 1000        0.10586940   0.18501406
## Lag 5000        -0.03936224  -0.01791672
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           0.5008543    0.51870207
## Lag 100         0.4453244    0.46669360
## Lag 500         0.3907627    0.41453930
## Lag 1000        0.3366895    0.35501264
## Lag 5000        -0.1140445   -0.07692479
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           0.5613010    0.5008543
## Lag 100         0.5033508    0.4453244
## Lag 500         0.4274531    0.3907627
## Lag 1000        0.3517565    0.3366895
## Lag 5000        -0.1677216   -0.1140445
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           0.8991309    1.00000000
## Lag 100         0.8359176    0.91646465

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## Lag 500           0.7187715          0.77900191
## Lag 1000          0.5982337          0.63306113
## Lag 5000          0.0176654          -0.03136451
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.6729793          0.51870207
## Lag 100             0.6010903          0.46669360
## Lag 500             0.5133373          0.41453930
## Lag 1000            0.4086869          0.35501264
## Lag 5000            -0.1581302         -0.07692479
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              1.00000000          0.89818763
## Lag 100            0.91646465          0.82112793
## Lag 500            0.77900191          0.69592684
## Lag 1000            0.63306113          0.55065244
## Lag 5000            -0.03136451         -0.04949222
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              0.05124741          0.001646162        0.04068399
## Lag 100             0.04921037          0.008930910        0.03759685
## Lag 500             0.03899327          0.015514352        0.02580930
## Lag 1000            0.06406637          0.051682789        0.02147033
## Lag 5000            -0.01137788          0.015052216        -0.04729885
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.12402916          0.08639137         -0.09704390
## Lag 100             -0.10713277          0.09204331         -0.10427670
## Lag 500             -0.07124572          0.09145340         -0.11605857
## Lag 1000            -0.06554488          0.09437629         -0.10182520
## Lag 5000             0.06888625          0.02727995          0.01776003
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0              -0.15610253          0.09891371
## Lag 100             -0.16408367          0.08304541
## Lag 500             -0.12517718          0.04261766
## Lag 1000            -0.09948987          0.04824881
## Lag 5000            -0.01179365          -0.01278893
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                -0.14515627          -0.1467921
## Lag 100              -0.11636785          -0.1156256
## Lag 500              -0.08159624          -0.1042741
## Lag 1000             -0.07861755          -0.1076860
## Lag 5000              0.10186075          0.0712461
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                  0.18051752          0.23807593
## Lag 100                0.18953377          0.24534120
## Lag 500                0.17920764          0.22900261
## Lag 1000                0.16794260          0.21104881
## Lag 5000                0.04389169          0.05181362
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  -0.1467921          -0.14376430

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## Lag 100           -0.1156256          -0.12366939
## Lag 500           -0.1042741          -0.12078005
## Lag 1000          -0.1076860          -0.15700540
## Lag 5000          0.0712461           0.01799705
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.21778240         0.25816273
## Lag 100            0.22684710         0.26502282
## Lag 500            0.20506975         0.23751874
## Lag 1000           0.19012136         0.21960899
## Lag 5000           0.02816484         0.03038383
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.18051752         0.21778240
## Lag 100             0.18953377         0.22684710
## Lag 500             0.17920764         0.20506975
## Lag 1000            0.16794260         0.19012136
## Lag 5000            0.04389169         0.02816484
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              -0.8750125593        -0.93853010
## Lag 100             -0.8368578282        -0.89227332
## Lag 500             -0.7287281819        -0.75764556
## Lag 1000            -0.6175608959        -0.62307537
## Lag 5000            -0.0008665459         0.04348967
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.23807593          0.25816273
## Lag 100             0.24534120          0.26502282
## Lag 500             0.22900261          0.23751874
## Lag 1000            0.21104881          0.21960899
## Lag 5000            0.05181362          0.03038383
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.93853010        -0.85888343
## Lag 100             -0.89227332        -0.81357341
## Lag 500             -0.75764556        -0.67868373
## Lag 1000            -0.62307537        -0.54366192
## Lag 5000            0.04348967          0.05718606
##
## , , traittarsus.1:traitbwt.2.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.51576277          0.52342220
## Lag 100            0.44159056          0.43580048
## Lag 500            0.30785687          0.33037079
## Lag 1000           0.24390874          0.26926704
## Lag 5000            -0.05005481         0.04687804
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.782061640         1.00000000
## Lag 100             0.696299215         0.87453824
## Lag 500             0.538889846         0.65219311

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## Lag 1000           0.445389577          0.48957859
## Lag 5000          -0.006519724          0.01842864
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.52342220          0.32999449
## Lag 100             0.43580048          0.28148773
## Lag 500             0.33037079          0.26589289
## Lag 1000            0.26926704          0.24323871
## Lag 5000            0.04687804          0.09890983
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.59763628          0.6478170
## Lag 100             0.52643466          0.5573055
## Lag 500             0.44551868          0.4418750
## Lag 1000            0.37736602          0.3442054
## Lag 5000            0.08285167          0.1125165
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.782061640         0.59763628
## Lag 100             0.696299215         0.52643466
## Lag 500             0.538889846         0.44551868
## Lag 1000            0.445389577         0.37736602
## Lag 5000            -0.006519724        0.08285167
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.5635407          0.6729793
## Lag 100             0.5125673          0.6052812
## Lag 500             0.4465910          0.5277439
## Lag 1000            0.4070664          0.4538221
## Lag 5000            0.1850370          0.1792542
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              1.00000000         0.6478170
## Lag 100             0.87453824          0.5573055
## Lag 500             0.65219311          0.4418750
## Lag 1000            0.48957859          0.3442054
## Lag 5000            0.01842864          0.1125165
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              0.6729793          0.6191252
## Lag 100             0.6052812          0.5529718
## Lag 500             0.5277439          0.4729419
## Lag 1000            0.4538221          0.3884032
## Lag 5000            0.1792542          0.1542326
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              0.063463728         0.02730871         0.051964042
## Lag 100             0.066558472         0.03961464         0.040967359
## Lag 500             0.076728186         0.05374642        -0.012930939
## Lag 1000            0.100842393         0.09752083        -0.062954038
## Lag 5000            -0.006751012        0.04984406         0.002434151
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.09337461         0.02006907        -0.06942502
## Lag 100             -0.09001277         0.02763705        -0.07262374

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## Lag 500      -0.04869538      0.02000750      -0.09505843
## Lag 1000     -0.05831890      0.02560894      -0.08617648
## Lag 5000      0.04246982      0.06392166      -0.03722137
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0        -0.026436938     0.04826105
## Lag 100      -0.041329867     0.04078039
## Lag 500      -0.028290001     0.03056152
## Lag 1000     -0.036326123     0.04300492
## Lag 5000     -0.008439906     0.00653221
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.3969629895    -0.44235441
## Lag 100         -0.3457725771    -0.37788435
## Lag 500         -0.2409128696    -0.27612777
## Lag 1000        -0.2064627528    -0.23939733
## Lag 5000        -0.0008284304    -0.05932951
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            0.1332860      0.1894193
## Lag 100          0.1542105      0.2129929
## Lag 500          0.1772766      0.2426247
## Lag 1000         0.1956287      0.2577964
## Lag 5000         0.1561845      0.1623293
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.44235441     -0.3300209
## Lag 100         -0.37788435     -0.2868342
## Lag 500         -0.27612777     -0.2200427
## Lag 1000        -0.23939733     -0.2196364
## Lag 5000        -0.05932951     -0.1105188
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            0.1392793      0.1845588
## Lag 100          0.1618837      0.2068320
## Lag 500          0.1824833      0.2322375
## Lag 1000         0.2062851      0.2545405
## Lag 5000         0.1708879      0.1679111
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            0.1332860      0.1392793
## Lag 100          0.1542105      0.1618837
## Lag 500          0.1772766      0.1824833
## Lag 1000         0.1956287      0.2062851
## Lag 5000         0.1561845      0.1708879
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.5541670      -0.6200400
## Lag 100          -0.5163721      -0.5823652
## Lag 500          -0.4591717      -0.5082023
## Lag 1000         -0.4081394      -0.4277328
## Lag 5000         -0.1850899      -0.1717978
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            0.1894193      0.1845588

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## Lag 100          0.2129929          0.2068320
## Lag 500          0.2426247          0.2322375
## Lag 1000         0.2577964          0.2545405
## Lag 5000         0.1623293          0.1679111
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0            -0.6200400         -0.5649761
## Lag 100          -0.5823652         -0.5353605
## Lag 500          -0.5082023         -0.4627844
## Lag 1000         -0.4277328         -0.3777429
## Lag 5000         -0.1717978         -0.1507443
##
## , , traitbwt.1:traitbwt.2.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0            0.23631479        0.42935174
## Lag 100          0.19631829        0.33111043
## Lag 500          0.13140023        0.20166811
## Lag 1000         0.13084790        0.23051272
## Lag 5000         -0.05046358       0.04857134
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0            0.4651159         0.6478170
## Lag 100          0.4250027         0.5751741
## Lag 500          0.3644232         0.4605468
## Lag 1000         0.3695356         0.4130944
## Lag 5000         -0.0748654        -0.1705451
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0            0.42935174        0.50673141
## Lag 100          0.33111043        0.36623376
## Lag 500          0.20166811        0.24599792
## Lag 1000         0.23051272        0.26598499
## Lag 5000         0.04857134        0.06879203
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            0.75320258        1.00000000
## Lag 100          0.64643730        0.80429897
## Lag 500          0.48359147        0.55454840
## Lag 1000         0.43668798        0.44468678
## Lag 5000         0.08606661        -0.01807979
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            0.4651159         0.75320258
## Lag 100          0.4250027         0.64643730
## Lag 500          0.3644232         0.48359147
## Lag 1000         0.3695356         0.43668798
## Lag 5000         -0.0748654        0.08606661
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            0.35637772        0.518702070
## Lag 100          0.34579510        0.481108274
## Lag 500          0.32649959        0.434843612

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## Lag 1000           0.37931051           0.436821292
## Lag 5000           0.03852246           0.006399124
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.6478170            1.000000000
## Lag 100             0.5751741            0.80429897
## Lag 500             0.4605468            0.55454840
## Lag 1000            0.4130944            0.44468678
## Lag 5000            -0.1705451           -0.01807979
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              0.518702070          0.64907876
## Lag 100             0.481108274          0.56161598
## Lag 500             0.434843612          0.47984513
## Lag 1000            0.436821292          0.40938531
## Lag 5000            0.006399124          -0.01977366
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              0.034800266          0.007431509          0.10446126
## Lag 100             0.051936557          -0.001493245          0.08007660
## Lag 500             0.071129943          0.067797207          0.05478615
## Lag 1000            0.103624443          0.063225351          0.01489821
## Lag 5000            -0.000397122          0.008471637          -0.01350094
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.08333568          0.100888337          -0.21005355
## Lag 100             -0.07762909          0.079323546          -0.17806133
## Lag 500             -0.03825641          -0.007478373          -0.11051022
## Lag 1000            -0.02784742          0.017077993          -0.10296388
## Lag 5000             0.04558001          0.022598106          0.02483781
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0              0.018470325          -0.095372303
## Lag 100             -0.002048636          -0.029630368
## Lag 500             0.014260425          -0.034300090
## Lag 1000            -0.048567932          0.067004499
## Lag 5000             -0.025983993          0.005631837
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  -0.25474138          -0.36301593
## Lag 100                 -0.21067390          -0.29399524
## Lag 500                 -0.11042093          -0.17823668
## Lag 1000                -0.13264168          -0.21582496
## Lag 5000                  0.02716341          -0.03930422
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                  0.06808542           0.1473359
## Lag 100                 0.08333997           0.1644163
## Lag 500                 0.11004808           0.1967466
## Lag 1000                0.13239890           0.2087490
## Lag 5000                  0.14121855           0.1213850
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  -0.36301593          -0.39024104
## Lag 100                 -0.29399524          -0.30803165

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## Lag 500           -0.17823668          -0.22428734
## Lag 1000          -0.21582496          -0.24800908
## Lag 5000          -0.03930422          -0.07900553
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.09800038          0.1786089
## Lag 100             0.12840120          0.2150929
## Lag 500             0.17479984          0.2654006
## Lag 1000            0.19253125          0.2644751
## Lag 5000            0.13671344          0.1087463
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.06808542          0.09800038
## Lag 100             0.08333997          0.12840120
## Lag 500             0.11004808          0.17479984
## Lag 1000            0.13239890          0.19253125
## Lag 5000            0.14121855          0.13671344
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              -0.38430674         -0.49584021
## Lag 100             -0.36892511         -0.47162690
## Lag 500             -0.35875583         -0.43506761
## Lag 1000            -0.38190122         -0.42199356
## Lag 5000            -0.04674851         -0.01287201
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.1473359           0.1786089
## Lag 100             0.1644163           0.2150929
## Lag 500             0.1967466           0.2654006
## Lag 1000            0.2087490           0.2644751
## Lag 5000            0.1213850           0.1087463
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.49584021         -0.56988609
## Lag 100             -0.47162690         -0.53124887
## Lag 500             -0.43506761         -0.45896948
## Lag 1000            -0.42199356         -0.40139643
## Lag 5000            -0.01287201         0.01366036
##
## , , traittarsus.2:traitbwt.2.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.12397199          0.14997594
## Lag 100             0.08643519          0.11314881
## Lag 500             0.03605225          0.09323271
## Lag 1000            0.03084094          0.10586940
## Lag 5000            -0.09632022         -0.03936224
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.5613010           0.6729793
## Lag 100             0.5033508           0.6010903
## Lag 500             0.4274531           0.5133373
## Lag 1000            0.3517565           0.4086869

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## Lag 5000           -0.1677216          -0.1581302
## traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.14997594         0.17131088
## Lag 100             0.11314881         0.15721379
## Lag 500             0.09323271         0.15933093
## Lag 1000            0.10586940         0.18501406
## Lag 5000            -0.03936224        -0.01791672
## traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.5008543          0.51870207
## Lag 100             0.4453244          0.46669360
## Lag 500             0.3907627          0.41453930
## Lag 1000            0.3366895          0.35501264
## Lag 5000            -0.1140445         -0.07692479
## traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.5613010          0.5008543
## Lag 100             0.5033508          0.4453244
## Lag 500             0.4274531          0.3907627
## Lag 1000            0.3517565          0.3366895
## Lag 5000            -0.1677216         -0.1140445
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.8991309          1.00000000
## Lag 100             0.8359176          0.91646465
## Lag 500             0.7187715          0.77900191
## Lag 1000            0.5982337          0.63306113
## Lag 5000            0.0176654          -0.03136451
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.6729793          0.51870207
## Lag 100             0.6010903          0.46669360
## Lag 500             0.5133373          0.41453930
## Lag 1000            0.4086869          0.35501264
## Lag 5000            -0.1581302         -0.07692479
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              1.00000000         0.89818763
## Lag 100             0.91646465         0.82112793
## Lag 500             0.77900191         0.69592684
## Lag 1000            0.63306113         0.55065244
## Lag 5000            -0.03136451        -0.04949222
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              0.05124741         0.001646162        0.04068399
## Lag 100             0.04921037         0.008930910        0.03759685
## Lag 500             0.03899327         0.015514352        0.02580930
## Lag 1000            0.06406637         0.051682789        0.02147033
## Lag 5000            -0.01137788        0.015052216        -0.04729885
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.12402916        0.08639137       -0.09704390
## Lag 100             -0.10713277        0.09204331       -0.10427670
## Lag 500             -0.07124572        0.09145340       -0.11605857

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## Lag 1000      -0.06554488      0.09437629      -0.10182520
## Lag 5000      0.06888625      0.02727995      0.01776003
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0         -0.15610253      0.09891371
## Lag 100       -0.16408367      0.08304541
## Lag 500       -0.12517718      0.04261766
## Lag 1000      -0.09948987      0.04824881
## Lag 5000      -0.01179365      -0.01278893
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.14515627      -0.1467921
## Lag 100        -0.11636785      -0.1156256
## Lag 500        -0.08159624      -0.1042741
## Lag 1000       -0.07861755      -0.1076860
## Lag 5000       0.10186075      0.0712461
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0           0.18051752      0.23807593
## Lag 100         0.18953377      0.24534120
## Lag 500         0.17920764      0.22900261
## Lag 1000        0.16794260      0.21104881
## Lag 5000        0.04389169      0.05181362
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.1467921      -0.14376430
## Lag 100         -0.1156256      -0.12366939
## Lag 500         -0.1042741      -0.12078005
## Lag 1000        -0.1076860      -0.15700540
## Lag 5000         0.0712461      0.01799705
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0           0.21778240      0.25816273
## Lag 100         0.22684710      0.26502282
## Lag 500         0.20506975      0.23751874
## Lag 1000        0.19012136      0.21960899
## Lag 5000         0.02816484      0.03038383
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0           0.18051752      0.21778240
## Lag 100         0.18953377      0.22684710
## Lag 500         0.17920764      0.20506975
## Lag 1000        0.16794260      0.19012136
## Lag 5000         0.04389169      0.02816484
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.8750125593     -0.93853010
## Lag 100         -0.8368578282     -0.89227332
## Lag 500         -0.7287281819     -0.75764556
## Lag 1000        -0.6175608959     -0.62307537
## Lag 5000         -0.0008665459      0.04348967
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0           0.23807593      0.25816273
## Lag 100         0.24534120      0.26502282

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## Lag 500           0.22900261          0.23751874
## Lag 1000          0.21104881          0.21960899
## Lag 5000          0.05181362          0.03038383
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0             -0.93853010         -0.85888343
## Lag 100            -0.89227332        -0.81357341
## Lag 500            -0.75764556        -0.67868373
## Lag 1000           -0.62307537        -0.54366192
## Lag 5000            0.04348967         0.05718606
##
## , , traitbwt.2:traitbwt.2.animal
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.07790198          0.14180302
## Lag 100             0.04618642          0.09666633
## Lag 500             0.01084756          0.06603768
## Lag 1000            0.02267992          0.11225922
## Lag 5000            -0.11897702         -0.02765292
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.4107260           0.6191252
## Lag 100             0.3762287           0.5533278
## Lag 500             0.3541228           0.4906565
## Lag 1000            0.3161080           0.4110850
## Lag 5000            -0.1301409          -0.1457415
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.14180302          0.21046169
## Lag 100             0.09666633          0.16894953
## Lag 500             0.06603768          0.16336058
## Lag 1000            0.11225922          0.19977861
## Lag 5000            -0.02765292          0.02292947
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.47574281          0.64907876
## Lag 100             0.42046513          0.55150494
## Lag 500             0.36650261          0.46465819
## Lag 1000            0.33756450          0.39404623
## Lag 5000            -0.04573143         -0.05021868
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.4107260           0.47574281
## Lag 100             0.3762287           0.42046513
## Lag 500             0.3541228           0.36650261
## Lag 1000            0.3161080           0.33756450
## Lag 5000            -0.1301409          -0.04573143
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.6844205           0.89818763
## Lag 100             0.6509791           0.82577566
## Lag 500             0.5935139           0.71300098
## Lag 1000            0.5428655           0.62246323

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## Lag 5000          0.0305572      -0.02000725
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            0.6191252      0.64907876
## Lag 100          0.5533278      0.55150494
## Lag 500          0.4906565      0.46465819
## Lag 1000         0.4110850      0.39404623
## Lag 5000         -0.1457415     -0.05021868
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            0.89818763     1.00000000
## Lag 100          0.82577566     0.87344161
## Lag 500          0.71300098     0.71296418
## Lag 1000         0.62246323     0.58027397
## Lag 5000         -0.02000725    -0.05065402
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            0.05353374     0.007606345   0.07295919
## Lag 100          0.05915538     0.016873424   0.06649764
## Lag 500          0.05495199     0.034153584   0.05087275
## Lag 1000         0.08867289     0.063336260   0.03210710
## Lag 5000         0.01157646     0.010648297   -0.07267030
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.11303991    0.128729447   -0.162810059
## Lag 100          -0.10682241    0.132664466   -0.165762472
## Lag 500          -0.06622430    0.082749840   -0.142795999
## Lag 1000         -0.06796661    0.092840268   -0.120838816
## Lag 5000         0.07422776    0.007625067   0.007301082
## traittarsus.2.mother traitbwt.2.mother
## Lag 0            -0.04298505    -0.033866092
## Lag 100          -0.06870971    0.016347544
## Lag 500          -0.06335390    0.003806205
## Lag 1000         -0.06335680    0.042702082
## Lag 5000         -0.03319016    -0.022982672
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0            -0.13106867    -0.13241719
## Lag 100          -0.10278270    -0.09227117
## Lag 500          -0.06959318    -0.07769390
## Lag 1000         -0.08442761    -0.10813364
## Lag 5000         0.12643968    0.06443195
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            0.13071885    0.21162923
## Lag 100          0.13765665    0.22039206
## Lag 500          0.12262833    0.19792348
## Lag 1000         0.11563474    0.17509971
## Lag 5000         0.01775142    0.02534375
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            -0.13241719    -0.1400636
## Lag 100          -0.09227117    -0.1032594
## Lag 500          -0.07769390    -0.1120628

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## Lag 1000           -0.10813364          -0.1695827
## Lag 5000            0.06443195          -0.0137381
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0               0.172783869         0.24161056
## Lag 100              0.184368154         0.25618384
## Lag 500              0.169565117         0.23188706
## Lag 1000             0.152615170         0.20023901
## Lag 5000             0.008979599         0.01431812
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0                 0.13071885         0.172783869
## Lag 100                0.13765665         0.184368154
## Lag 500                0.12262833         0.169565117
## Lag 1000               0.11563474         0.152615170
## Lag 5000               0.01775142         0.008979599
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0                 -0.70436348        -0.85484247
## Lag 100                -0.67503813        -0.80874929
## Lag 500                -0.61496101        -0.69869786
## Lag 1000               -0.55915090        -0.60389172
## Lag 5000               -0.01252908        0.03149869
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0                 0.21162923         0.24161056
## Lag 100                0.22039206         0.25618384
## Lag 500                0.19792348         0.23188706
## Lag 1000               0.17509971         0.20023901
## Lag 5000               0.02534375         0.01431812
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0                 -0.85484247        -0.90703776
## Lag 100                -0.80874929        -0.84260101
## Lag 500                -0.69869786        -0.68593227
## Lag 1000               -0.60389172        -0.56191283
## Lag 5000               0.03149869         0.05571691
##
## , , traittarsus.1:byear
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                 -0.0001991343         0.06781216
## Lag 100                0.0149599532         0.07212249
## Lag 500                -0.0052702707         0.06733748
## Lag 1000               0.0022177857         0.01064963
## Lag 5000                -0.0477855881        -0.02770227
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                 0.016204947          0.06346373
## Lag 100                0.029428446          0.06520358
## Lag 500                0.003988555          0.04831408
## Lag 1000               -0.009687121         -0.01510108
## Lag 5000                -0.057748697         -0.07984786

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##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  0.06781216      0.06518811
## Lag 100                 0.07212249      0.07480376
## Lag 500                 0.06733748      0.11082992
## Lag 1000                0.01064963      0.05008650
## Lag 5000                -0.02770227     -0.01341916
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                  0.02942135      0.034800266
## Lag 100                 0.02240012      0.032507546
## Lag 500                 0.02748906      0.041726424
## Lag 1000                0.01673477      0.007404418
## Lag 5000                -0.04116453     -0.064540096
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                  0.016204947     0.02942135
## Lag 100                 0.029428446     0.02240012
## Lag 500                 0.003988555     0.02748906
## Lag 1000                -0.009687121    0.01673477
## Lag 5000                -0.057748697    -0.04116453
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                  0.035961574     0.05124741
## Lag 100                 0.039140531     0.04519151
## Lag 500                 -0.008894598    0.03130914
## Lag 1000                -0.036681278    -0.01447981
## Lag 5000                -0.088569223    -0.10213928
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                  0.06346373      0.034800266
## Lag 100                 0.06520358      0.032507546
## Lag 500                 0.04831408      0.041726424
## Lag 1000                -0.01510108     0.007404418
## Lag 5000                -0.07984786     -0.064540096
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                  0.05124741      0.05353374
## Lag 100                 0.04519151      0.03671240
## Lag 500                 0.03130914      0.04212493
## Lag 1000                -0.01447981     -0.01321894
## Lag 5000                -0.10213928     -0.09784272
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0                  1.000000000    -0.011500754    -0.014207130
## Lag 100                 0.29498964      0.032775932      0.007424037
## Lag 500                 0.01534486      -0.007063004    -0.008141387
## Lag 1000                -0.00531817      0.074940752    -0.088872220
## Lag 5000                -0.04617778      -0.015458692   -0.020304988
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0                  -0.0081906518    0.107143639    -0.06029833
## Lag 100                 -0.0458427594    0.096733446    -0.08150148
## Lag 500                 0.0004877642    0.068025342    -0.06046456
## Lag 1000                0.0105349484    -0.002189929   -0.01019378

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## Lag 5000 -0.0644916207 -0.020934152 0.04368155
## traittarsus.2.mother traitbwt.2.mother
## Lag 0 -0.03378213 0.028992058
## Lag 100 -0.04096317 0.005259162
## Lag 500 0.04192974 -0.033349969
## Lag 1000 -0.02480675 -0.011058052
## Lag 5000 0.04747283 -0.001107231
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0 -0.145066040 -0.104716754
## Lag 100 -0.089626633 -0.078525674
## Lag 500 -0.055797909 -0.064960700
## Lag 1000 -0.002415045 -0.025253406
## Lag 5000 0.031464141 0.006356867
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0 -0.038470154 -0.03236997
## Lag 100 -0.037857054 -0.03429257
## Lag 500 -0.023813333 -0.02985769
## Lag 1000 -0.004466145 -0.01855463
## Lag 5000 0.031916697 0.01037899
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0 -0.104716754 -0.071954690
## Lag 100 -0.078525674 -0.071715488
## Lag 500 -0.064960700 -0.078535410
## Lag 1000 -0.025253406 -0.053257563
## Lag 5000 0.006356867 -0.008584678
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0 -0.02749973 -0.02370119
## Lag 100 -0.02036580 -0.01896999
## Lag 500 -0.02537270 -0.03447923
## Lag 1000 0.01650965 0.00885959
## Lag 5000 0.03965003 0.01362772
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0 -0.038470154 -0.02749973
## Lag 100 -0.037857054 -0.02036580
## Lag 500 -0.023813333 -0.02537270
## Lag 1000 -0.004466145 0.01650965
## Lag 5000 0.031916697 0.03965003
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0 -0.029532571 -0.04533921
## Lag 100 -0.006520508 -0.01935854
## Lag 500 0.015074249 -0.01914032
## Lag 1000 0.040326869 0.01182073
## Lag 5000 0.077261030 0.09163015
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0 -0.03236997 -0.02370119
## Lag 100 -0.03429257 -0.01896999
## Lag 500 -0.02985769 -0.03447923

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## Lag 1000           -0.01855463      0.00885959
## Lag 5000            0.01037899      0.01362772
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.04533921      -0.044815839
## Lag 100             -0.01935854      -0.018188105
## Lag 500             -0.01914032      -0.044430332
## Lag 1000            0.01182073      0.006995154
## Lag 5000             0.09163015      0.087372099
##
## , , traitbwt.1.byear
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                0.02533600      -0.048571545
## Lag 100               0.04296781      -0.011259967
## Lag 500               0.04385496      0.001433897
## Lag 1000              0.00311012      -0.028185947
## Lag 5000              -0.03791093     -0.035032491
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                0.042952342      0.027308707
## Lag 100               0.053766787      0.050479200
## Lag 500               0.061397390      0.045799389
## Lag 1000              0.026376244      0.045575916
## Lag 5000              -0.007467705      0.004959177
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                -0.048571545     -0.054239888
## Lag 100               -0.011259967     -0.013618719
## Lag 500               0.001433897     -0.023666680
## Lag 1000              -0.028185947     -0.035975542
## Lag 5000              -0.035032491     -0.004849703
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                0.01370326      0.007431509
## Lag 100               0.02573311      0.031124056
## Lag 500               0.01003935      -0.020817482
## Lag 1000              -0.01201110     -0.011831635
## Lag 5000              -0.03020983      0.005318089
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                0.042952342      0.01370326
## Lag 100               0.053766787      0.02573311
## Lag 500               0.061397390      0.01003935
## Lag 1000              0.026376244      -0.01201110
## Lag 5000              -0.007467705     -0.03020983
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                -0.01746914      0.001646162
## Lag 100               -0.00918084      0.014660207
## Lag 500                0.03727433      0.029738658
## Lag 1000               0.05233753      0.058558627
## Lag 5000               0.05442837      0.036237222

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##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                      0.027308707 0.007431509
## Lag 100                     0.050479200 0.031124056
## Lag 500                     0.045799389 -0.020817482
## Lag 1000                    0.045575916 -0.011831635
## Lag 5000                    0.004959177 0.005318089
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                      0.001646162 0.007606345
## Lag 100                     0.014660207 0.025027029
## Lag 500                     0.029738658 0.007127126
## Lag 1000                    0.058558627 0.051997571
## Lag 5000                    0.036237222 0.030982100
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0                      -0.011500754 1.000000000 -0.032565228
## Lag 100                     -0.008778968 0.12997956 -0.003228198
## Lag 500                     -0.004505343 0.03863671 -0.017827698
## Lag 1000                    0.025463698 -0.01655625 0.013769326
## Lag 5000                    0.001254535 -0.04239004 0.001901775
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0                      -0.039257246 -0.02879748 0.065880688
## Lag 100                     0.015007558 -0.03208555 0.058933430
## Lag 500                     -0.005393351 0.03210587 0.004743831
## Lag 1000                    0.043913747 0.05551756 0.012244934
## Lag 5000                    -0.026820593 0.05843803 0.010088100
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0                      0.008939011 0.059268463
## Lag 100                     -0.001571382 0.086064468
## Lag 500                     -0.023366713 0.008732966
## Lag 1000                    0.002165536 -0.002113612
## Lag 5000                    -0.038345540 -0.007683096
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                      -0.0435586501 0.010219509
## Lag 100                     -0.0505048908 -0.031324102
## Lag 500                     -0.0630828149 -0.004357406
## Lag 1000                    0.0199630207 0.069957559
## Lag 5000                    0.0005522977 -0.005454216
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                      0.05293519 0.05768879
## Lag 100                     0.04858011 0.05892110
## Lag 500                     0.05854491 0.06372575
## Lag 1000                    0.04969484 0.03639502
## Lag 5000                    0.02832954 0.02784365
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                      0.010219509 -0.028183637
## Lag 100                     -0.031324102 -0.040106071
## Lag 500                     -0.004357406 0.006591277
## Lag 1000                    0.069957559 0.063937711

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## Lag 5000           -0.005454216          -0.027544320
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.04996250           0.04700881
## Lag 100             0.04397513           0.04673808
## Lag 500             0.04695403           0.04531904
## Lag 1000            0.03436998           0.02563582
## Lag 5000            0.02256523           0.02398399
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.05293519           0.04996250
## Lag 100             0.04858011           0.04397513
## Lag 500             0.05854491           0.04695403
## Lag 1000            0.04969484           0.03436998
## Lag 5000            0.02832954           0.02256523
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              0.01013932          -0.001638125
## Lag 100             0.01806400           0.001129150
## Lag 500             -0.01420490          -0.017077159
## Lag 1000            -0.04995245          -0.056833617
## Lag 5000            -0.04322171          -0.023913807
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              0.05768879           0.04700881
## Lag 100             0.05892110           0.04673808
## Lag 500             0.06372575           0.04531904
## Lag 1000            0.03639502           0.02563582
## Lag 5000            0.02784365           0.02398399
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.001638125         -0.007601342
## Lag 100             0.001129150         -0.013307489
## Lag 500             -0.017077159         -0.008568690
## Lag 1000            -0.056833617         -0.049677065
## Lag 5000            -0.023913807         -0.009694337
##
## , , traittarsus.2:byear
##
## traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.02726198          -0.00111586
## Lag 100            0.05357730           0.01828074
## Lag 500            0.04315378           0.02964535
## Lag 1000            0.08991282           0.06959065
## Lag 5000            0.05518086           0.09966305
## traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.05222339           0.05196404
## Lag 100             0.07930026           0.07496121
## Lag 500             0.04558008           0.04893690
## Lag 1000            0.07097163           0.08902753
## Lag 5000            0.06038135           0.02474492
## traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal

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## Lag 0           -0.00111586      -0.03158434
## Lag 100        0.01828074      -0.03831807
## Lag 500        0.02964535       0.03186606
## Lag 1000       0.06959065       0.06084561
## Lag 5000       0.09966305       0.04703834
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           0.08861601       0.10446126
## Lag 100         0.08291699       0.08928353
## Lag 500         0.07447628       0.09304440
## Lag 1000        0.05734586       0.08937241
## Lag 5000        0.08247995       0.05423744
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           0.05222339       0.08861601
## Lag 100         0.07930026       0.08291699
## Lag 500         0.04558008       0.07447628
## Lag 1000        0.07097163       0.05734586
## Lag 5000        0.06038135       0.08247995
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           0.007171364      0.04068399
## Lag 100         0.043732946      0.05987656
## Lag 500         0.015444417      0.03370999
## Lag 1000        0.019544081      0.04502976
## Lag 5000        0.006685864     -0.01642810
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           0.05196404       0.10446126
## Lag 100         0.07496121       0.08928353
## Lag 500         0.04893690       0.09304440
## Lag 1000        0.08902753       0.08937241
## Lag 5000        0.02474492       0.05423744
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0           0.04068399       0.07295919
## Lag 100         0.05987656       0.07152626
## Lag 500         0.03370999       0.07468887
## Lag 1000        0.04502976       0.05593692
## Lag 5000        -0.01642810     -0.02210630
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           -0.014207130      -0.03256523      1.00000000
## Lag 100         0.004989327      0.01606065      0.26418064
## Lag 500         0.044052804      0.03428138      0.09537772
## Lag 1000        -0.001440518      0.01039861      0.10874833
## Lag 5000        0.022958892      -0.01931981     -0.01473189
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0           0.0396353651      0.00119313      0.009996629
## Lag 100         0.0105436230      0.04327018      0.062928368
## Lag 500         -0.0175690949      0.04625644      0.018026408
## Lag 1000        -0.0005183552      0.06289415     -0.081508956
## Lag 5000        0.0441472029      -0.04105115     0.035686320

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##          traittarsus.2.mother traitbwt.2.mother
## Lag 0           0.04535223    0.002495676
## Lag 100         0.02030029    0.045447997
## Lag 500         0.02446377    0.009649222
## Lag 1000        0.03899831    0.024762534
## Lag 5000        -0.06415993   0.094134035
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.04924174   0.0004983836
## Lag 100          -0.07260472   -0.0274666354
## Lag 500          -0.08762084   -0.0545545565
## Lag 1000         -0.11600745   -0.0941075726
## Lag 5000         -0.01429942   -0.0306094192
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            0.06126409   0.06054337
## Lag 100           0.06861858   0.06794736
## Lag 500           0.06487160   0.06219436
## Lag 1000          0.05154526   0.06299063
## Lag 5000          -0.03961272  -0.02077759
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            0.0004983836  0.045243199
## Lag 100          -0.0274666354  0.002618325
## Lag 500          -0.0545545565  -0.045086068
## Lag 1000         -0.0941075726  -0.078265909
## Lag 5000         -0.0306094192  -0.039652118
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            0.04569537   0.05361674
## Lag 100           0.05790943   0.06228633
## Lag 500           0.06629817   0.07181673
## Lag 1000          0.05677668   0.07820270
## Lag 5000          -0.04900870  -0.01725220
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            0.06126409   0.04569537
## Lag 100           0.06861858   0.05790943
## Lag 500           0.06487160   0.06629817
## Lag 1000          0.05154526   0.05677668
## Lag 5000          -0.03961272  -0.04900870
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0            -0.05907849  -0.07410803
## Lag 100          -0.06985906  -0.08123809
## Lag 500          -0.02440940  -0.02862324
## Lag 1000         -0.03153841  -0.04394287
## Lag 5000          0.01609465   0.02558327
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            0.06054337   0.05361674
## Lag 100          0.06794736   0.06228633
## Lag 500          0.06219436   0.07181673
## Lag 1000         0.06299063   0.07820270

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## Lag 5000           -0.02077759          -0.01725220
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.07410803          -0.09539294
## Lag 100             -0.08123809          -0.10192651
## Lag 500             -0.02862324          -0.04787149
## Lag 1000            -0.04394287          -0.05696449
## Lag 5000            0.02558327          0.02318439
##
## , , traitbwt.2.byear
##
## traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              -0.012312644         -0.026101857
## Lag 100             0.001679568          -0.022719854
## Lag 500             -0.022095064          0.036781391
## Lag 1000            -0.044649058          0.008440477
## Lag 5000            0.012471911          0.017412057
## traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              -0.041612693         -0.093374609
## Lag 100             -0.044670676         -0.085938138
## Lag 500             -0.073851310         -0.057467051
## Lag 1000            -0.112427556         -0.073729661
## Lag 5000            -0.002302422         -0.007211525
## traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              -0.026101857         -0.024899884
## Lag 100             -0.022719854         -0.008317247
## Lag 500             0.036781391          0.041716499
## Lag 1000            0.008440477          -0.015226273
## Lag 5000            0.017412057          -0.007915616
## traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              -0.06566102          -0.08333568
## Lag 100             -0.07750349          -0.08378159
## Lag 500             -0.08329811          -0.05814502
## Lag 1000            -0.08817705          -0.08501262
## Lag 5000            0.01626488          -0.01063817
## traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              -0.041612693         -0.06566102
## Lag 100             -0.044670676         -0.07750349
## Lag 500             -0.073851310         -0.08329811
## Lag 1000            -0.112427556         -0.08817705
## Lag 5000            -0.002302422          0.01626488
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              -0.08333086          -0.12402916
## Lag 100             -0.07538028          -0.10747454
## Lag 500             -0.05024835          -0.08333129
## Lag 1000            -0.08252216          -0.09539109
## Lag 5000            -0.05270947          -0.04030638
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal

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## Lag 0           -0.093374609      -0.08333568
## Lag 100        -0.085938138      -0.08378159
## Lag 500        -0.057467051      -0.05814502
## Lag 1000       -0.073729661      -0.08501262
## Lag 5000       -0.007211525      -0.01063817
##     traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0           -0.12402916       -0.11303991
## Lag 100         -0.10747454       -0.09513886
## Lag 500         -0.08333129       -0.08666156
## Lag 1000        -0.09539109       -0.10675958
## Lag 5000        -0.04030638       -0.02380443
##     traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           -0.008190652      -0.039257246     0.03963537
## Lag 100          -0.031080960      -0.037426765     0.00903900
## Lag 500          0.006051004      -0.020419576     -0.06809153
## Lag 1000         0.018314138      -0.002583670     -0.02323337
## Lag 5000         -0.026798307      -0.007024901     -0.01177870
##     traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0           1.000000000      -0.03484939      0.019595061
## Lag 100          0.148773685      -0.05262924      0.013805361
## Lag 500          0.023071337      0.01145834      0.001042167
## Lag 1000         0.001235911      0.02017891      0.034032045
## Lag 5000         -0.035993191      -0.01462895     -0.011384188
##     traittarsus.2.mother traitbwt.2.mother
## Lag 0           0.015981233      -0.007859673
## Lag 100          -0.022156260      -0.004707385
## Lag 500          -0.006728665      -0.029061201
## Lag 1000         -0.021805472      0.003816387
## Lag 5000         -0.009007212      0.012478432
##     traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           0.01051817       0.02175859
## Lag 100          0.02304329       0.01564866
## Lag 500          0.01520723       -0.00804956
## Lag 1000         0.04140673       0.03546143
## Lag 5000         -0.01488195      -0.01947941
##     traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0           -0.04179855      -0.03235648
## Lag 100          -0.04317226      -0.03935256
## Lag 500          -0.03459143      -0.04109499
## Lag 1000         -0.03581518      -0.03377959
## Lag 5000         -0.04487869      -0.06427439
##     traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           0.02175859       0.008779544
## Lag 100          0.01564866       0.004367987
## Lag 500          -0.00804956      -0.020584034
## Lag 1000         0.03546143       0.024592354
## Lag 5000         -0.01947941      -0.008436479

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##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0                  -0.05587517      -0.03458419
## Lag 100                 -0.04924453      -0.03546666
## Lag 500                 -0.03327777      -0.03542884
## Lag 1000                -0.02872355      -0.01384796
## Lag 5000                -0.05472119      -0.06947879
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0                  -0.04179855      -0.05587517
## Lag 100                 -0.04317226      -0.04924453
## Lag 500                 -0.03459143      -0.03327777
## Lag 1000                -0.03581518      -0.02872355
## Lag 5000                -0.04487869      -0.05472119
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0                  0.07499456       0.09224924
## Lag 100                 0.07208086       0.09762890
## Lag 500                 0.07251351       0.09224810
## Lag 1000                0.08656386       0.08963227
## Lag 5000                0.01938132       0.01875251
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0                  -0.03235648      -0.03458419
## Lag 100                 -0.03935256      -0.03546666
## Lag 500                 -0.04109499      -0.03542884
## Lag 1000                -0.03377959      -0.01384796
## Lag 5000                -0.06427439      -0.06947879
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0                  0.09224924       0.07926090
## Lag 100                 0.09762890       0.09616451
## Lag 500                 0.09224810       0.09549623
## Lag 1000                0.08963227       0.09029398
## Lag 5000                0.01875251       0.01366050
##
## , , traittarsus.1.mother
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                  -0.15078835       0.015326642
## Lag 100                 -0.14463422       0.003365171
## Lag 500                 -0.15804338      -0.076509373
## Lag 1000                -0.09886124      -0.083743318
## Lag 5000                -0.10523556      -0.044881123
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                  -0.10295266       0.020069065
## Lag 100                 -0.08557369       0.024944654
## Lag 500                 -0.09200768      -0.007032787
## Lag 1000                -0.07176616      -0.035638619
## Lag 5000                -0.04022520      -0.053311401
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  0.015326642      0.063535708

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## Lag 100          0.003365171      0.050510724
## Lag 500          -0.076509373     0.005284771
## Lag 1000         -0.083743318     -0.009316143
## Lag 5000         -0.044881123     0.017585405
##   traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            0.01026682      0.10088834
## Lag 100          0.02777302      0.10524313
## Lag 500          -0.02208151      0.06189318
## Lag 1000         -0.03691548      0.02840216
## Lag 5000         -0.05057754      -0.05628831
##   traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            -0.10295266     0.01026682
## Lag 100          -0.08557369     0.02777302
## Lag 500          -0.09200768     -0.02208151
## Lag 1000         -0.07176616     -0.03691548
## Lag 5000         -0.04022520     -0.05057754
##   traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            0.046514036    0.086391373
## Lag 100          0.041591585    0.092682345
## Lag 500          0.014159903    0.054749349
## Lag 1000         -0.006046443    0.037148101
## Lag 5000         0.023770495    -0.002799893
##   traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            0.020069065    0.10088834
## Lag 100          0.024944654    0.10524313
## Lag 500          -0.007032787    0.06189318
## Lag 1000         -0.035638619    0.02840216
## Lag 5000         -0.053311401    -0.05628831
##   traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            0.086391373    0.128729447
## Lag 100          0.092682345    0.137837082
## Lag 500          0.054749349    0.091445074
## Lag 1000         0.037148101    0.065510680
## Lag 5000         -0.002799893   -0.009892994
##   traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            0.10714364     -0.028797482   0.001193130
## Lag 100          0.12475726     -0.002859105   0.017423692
## Lag 500          0.05635178     0.024729770   0.044602685
## Lag 1000         -0.01346005    0.015026664   -0.004151145
## Lag 5000         -0.03482857    -0.062611057   -0.134069816
##   traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.03484939    1.000000000  -0.271213796
## Lag 100          -0.06696141    0.63404978   -0.170737492
## Lag 500          -0.04137534    0.29008219   -0.136893424
## Lag 1000         0.02864811    0.07345948   -0.100567641
## Lag 5000         0.02750431    -0.03892620   0.007641615
##   traittarsus.2.mother traitbwt.2.mother

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## Lag 0           0.03387619   -0.05622667
## Lag 100        0.07299528   -0.04780327
## Lag 500        0.04150672   -0.03149897
## Lag 1000       0.02244428    0.03981911
## Lag 5000       -0.02167543   0.03773247
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.28511353   -0.039522955
## Lag 100         -0.16904760   -0.001067247
## Lag 500         -0.02585760    0.071889419
## Lag 1000        0.03550783    0.091662892
## Lag 5000        0.11461631    0.071952509
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0           0.1697424     0.17038749
## Lag 100         0.1698287     0.16772504
## Lag 500         0.1774711     0.18621633
## Lag 1000        0.1539551     0.18005891
## Lag 5000        0.1039239     0.09992436
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.039522955   0.02826569
## Lag 100         -0.001067247   0.01968611
## Lag 500         0.071889419    0.06722240
## Lag 1000        0.091662892    0.06515118
## Lag 5000        0.071952509   -0.00146006
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0           0.1570314     0.15014199
## Lag 100         0.1589658     0.15363210
## Lag 500         0.1753944     0.19214621
## Lag 1000        0.1525229     0.17920894
## Lag 5000        0.0981355     0.09929337
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0           0.1697424     0.1570314
## Lag 100         0.1698287     0.1589658
## Lag 500         0.1774711     0.1753944
## Lag 1000        0.1539551     0.1525229
## Lag 5000        0.1039239     0.0981355
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.05365944   -0.08507034
## Lag 100         -0.05114312   -0.08650307
## Lag 500         -0.03294118   -0.06365200
## Lag 1000        -0.01342963   -0.04929539
## Lag 5000        -0.03255967   -0.00364547
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0           0.17038749   0.15014199
## Lag 100         0.16772504   0.15363210
## Lag 500         0.18621633   0.19214621
## Lag 1000        0.18005891   0.17920894
## Lag 5000        0.09992436   0.09929337

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##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           -0.08507034           -0.11755479
## Lag 100         -0.08650307           -0.11806828
## Lag 500         -0.06365200           -0.08269424
## Lag 1000        -0.04929539           -0.07064952
## Lag 5000        -0.00364547           0.01527502
##
## , , traitbwt.1.mother
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0           0.10750858           -0.015001711
## Lag 100          0.11345621           -0.018373787
## Lag 500          0.12529867           0.019269248
## Lag 1000         0.09676891           0.063965485
## Lag 5000         0.03295446           0.002551181
##
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0           0.04055573           -0.06942502
## Lag 100          0.04246078           -0.06694784
## Lag 500          0.05073917           -0.04628001
## Lag 1000         0.02033431           -0.02832344
## Lag 5000         0.02405793           0.06116511
##
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0           -0.015001711          -0.301057181
## Lag 100          -0.018373787          -0.217367082
## Lag 500          0.019269248          -0.084030109
## Lag 1000         0.063965485          0.016600029
## Lag 5000         0.002551181          0.006633359
##
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           -0.002921727          -0.21005355
## Lag 100          -0.017967358          -0.17453411
## Lag 500          -0.048665239          -0.13167639
## Lag 1000         -0.013679253          -0.06743430
## Lag 5000         -0.015338823          0.06036364
##
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           0.04055573           -0.002921727
## Lag 100          0.04246078           -0.017967358
## Lag 500          0.05073917           -0.048665239
## Lag 1000         0.02033431           -0.013679253
## Lag 5000         0.02405793           -0.015338823
##
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           -0.063757286          -0.09704390
## Lag 100          -0.078555235          -0.11278771
## Lag 500          -0.075850107          -0.13102396
## Lag 1000         -0.051714924          -0.09839159
## Lag 5000         0.004416861           0.02809940
##
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           -0.06942502           -0.21005355

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## Lag 100           -0.06694784      -0.17453411
## Lag 500           -0.04628001      -0.13167639
## Lag 1000          -0.02832344      -0.06743430
## Lag 5000          0.06116511       0.06036364
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             -0.09704390      -0.16281006
## Lag 100           -0.11278771      -0.15402583
## Lag 500           -0.13102396      -0.18417663
## Lag 1000          -0.09839159      -0.13265931
## Lag 5000          0.02809940       0.06774942
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0             -0.06029833      0.065880688    0.009996629
## Lag 100            -0.06060175     0.002155759    0.013541244
## Lag 500            0.02852613       -0.001260087   0.019839054
## Lag 1000           -0.02910626     0.001290157    0.048245645
## Lag 5000           -0.01024266     0.050399045    0.056737041
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              0.0195950614   -0.2712137958  1.000000000
## Lag 100             -0.0008933777  -0.1720110752  0.379551682
## Lag 500             0.0366638614   -0.0393624856  0.041329213
## Lag 1000            -0.0151426446  0.0476618317  0.002159042
## Lag 5000            0.0123140414   0.0001252644  -0.018222262
## traittarsus.2.mother traitbwt.2.mother
## Lag 0              0.031492934    0.14519001
## Lag 100            0.002534347    0.07587057
## Lag 500            -0.013284762   0.04165749
## Lag 1000           -0.031081556   -0.02212041
## Lag 5000           0.013850724   -0.02074787
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0              0.01973629      0.03665672
## Lag 100            -0.02293101      0.01498518
## Lag 500            -0.11066356      -0.05474960
## Lag 1000           -0.09135360      -0.03226117
## Lag 5000           -0.07860493      -0.07560389
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.04294093      0.012376673
## Lag 100            0.03955764      0.010683501
## Lag 500            0.01752126      -0.006221666
## Lag 1000           0.02219142      -0.001297180
## Lag 5000           0.01239642      0.030951874
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0              0.03665672      -0.007047622
## Lag 100            0.01498518      0.037977225
## Lag 500            -0.05474960      0.019970710
## Lag 1000           -0.03226117      -0.007739455
## Lag 5000           -0.07560389      -0.062919612
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units

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## Lag 0           -0.002752353   -0.03097991
## Lag 100        -0.003093096   -0.03192449
## Lag 500        -0.009027187   -0.04102552
## Lag 1000       -0.020527419   -0.03483856
## Lag 5000       0.006886321    0.02535835
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0           0.04294093    -0.002752353
## Lag 100         0.03955764    -0.003093096
## Lag 500         0.01752126    -0.009027187
## Lag 1000        0.02219142    -0.020527419
## Lag 5000        0.01239642    0.006886321
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           0.08205572    0.11704839
## Lag 100         0.07607819    0.11403531
## Lag 500         0.10265484    0.14762922
## Lag 1000        0.06476226    0.10125717
## Lag 5000        -0.03096688   -0.06021103
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0           0.012376673   -0.03097991
## Lag 100         0.010683501   -0.03192449
## Lag 500         -0.006221666   -0.04102552
## Lag 1000        -0.001297180   -0.03483856
## Lag 5000        0.030951874   0.02535835
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           0.11704839    0.15336397
## Lag 100         0.11403531    0.15628950
## Lag 500         0.14762922    0.17953508
## Lag 1000        0.10125717    0.12282290
## Lag 5000        -0.06021103   -0.09027737
##
## , , traittarsus.2.mother
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0           0.01502226    0.051714153
## Lag 100         0.02971209    0.049905332
## Lag 500         0.07537838    0.055173503
## Lag 1000        0.12660306    0.083736831
## Lag 5000        0.02592685    0.003920255
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0           -0.15988076   -0.026436938
## Lag 100         -0.12481132   -0.008713726
## Lag 500         -0.06365003   -0.003368880
## Lag 1000        -0.01077423   0.048477788
## Lag 5000        0.17321257   0.100587628
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0           0.051714153   0.02830176
## Lag 100         0.049905332   0.03057360

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## Lag 500           0.055173503      0.03073470
## Lag 1000          0.083736831      0.04723145
## Lag 5000          0.003920255      0.01299831
##     traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0             -0.01173970      0.0184703252
## Lag 100            -0.01697136      0.0251626879
## Lag 500            -0.05188574      -0.0005911452
## Lag 1000           -0.04833371      0.0163990890
## Lag 5000           0.24669771      0.1785739883
##     traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0             -0.15988076      -0.01173970
## Lag 100            -0.12481132      -0.01697136
## Lag 500            -0.06365003      -0.05188574
## Lag 1000           -0.01077423      -0.04833371
## Lag 5000           0.17321257      0.24669771
##     traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0             -0.3447706       -0.1561025
## Lag 100            -0.2981644       -0.1438613
## Lag 500            -0.2223198       -0.1486635
## Lag 1000           -0.1932667       -0.1478947
## Lag 5000           0.1277913        0.1562519
##     traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0             -0.026436938      0.0184703252
## Lag 100            -0.008713726      0.0251626879
## Lag 500            -0.003368880      -0.0005911452
## Lag 1000           0.048477788      0.0163990890
## Lag 5000           0.100587628      0.1785739883
##     traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             -0.1561025       -0.04298505
## Lag 100            -0.1438613       -0.03256173
## Lag 500            -0.1486635       -0.08441125
## Lag 1000           -0.1478947       -0.11538109
## Lag 5000           0.1562519        0.16485805
##     traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0             -0.033782129      0.0089390110      0.04535223
## Lag 100            0.009970602      -0.0185988821      0.06005459
## Lag 500            -0.031227653      0.0075697533      -0.02997353
## Lag 1000           0.011336122      -0.0006296073      -0.05122879
## Lag 5000           0.049341702      -0.0529101788      0.11732582
##     traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              0.015981233      0.033876186      0.031492934
## Lag 100             -0.001350435      0.027458021      -0.034002906
## Lag 500             -0.067109166      -0.078872225      -0.017665845
## Lag 1000            -0.038517473      -0.030003877      -0.040451967
## Lag 5000            -0.086683580      -0.004367187      0.002148712
##     traittarsus.2.mother traitbwt.2.mother
## Lag 0               1.000000000      -0.32918284

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## Lag 100      0.57125197   -0.19561543
## Lag 500      0.26951589   -0.09671746
## Lag 1000     0.09211216   -0.01393679
## Lag 5000    -0.02238888   -0.04415446
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0         -0.003303414  -0.033595784
## Lag 100       -0.026601660  -0.053415941
## Lag 500        0.023989474  -0.005220881
## Lag 1000      -0.059547268  -0.037570106
## Lag 5000      -0.043225022  -0.018285255
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0         -0.032394558  -0.038463991
## Lag 100       -0.025527001  -0.031296257
## Lag 500        0.026572307  -0.025766415
## Lag 1000      -0.039244262  -0.042965449
## Lag 5000      -0.005888125  -0.008787816
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0         -0.033595784  -0.047856148
## Lag 100       -0.053415941  -0.036979690
## Lag 500        0.005220881  0.004374636
## Lag 1000      -0.037570106  -0.014564806
## Lag 5000      -0.018285255  -0.019587618
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0         -0.044767110  -0.050917805
## Lag 100       -0.038505551  -0.046892201
## Lag 500        0.026544619  -0.030053092
## Lag 1000      -0.047964926  -0.058484565
## Lag 5000        0.005513663  0.008103974
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0         -0.032394558  -0.044767110
## Lag 100       -0.025527001  -0.038505551
## Lag 500        0.026572307  -0.026544619
## Lag 1000      -0.039244262  -0.047964926
## Lag 5000      -0.005888125  0.005513663
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0          0.1360740   0.1159929
## Lag 100        0.1741515   0.1418886
## Lag 500        0.1700049   0.1445076
## Lag 1000       0.1777767   0.1521923
## Lag 5000       -0.1463666  -0.1686025
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0         -0.038463991  -0.050917805
## Lag 100       -0.031296257  -0.046892201
## Lag 500        0.025766415  -0.030053092
## Lag 1000      -0.042965449  -0.058484565
## Lag 5000      -0.008787816  0.008103974
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units

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## Lag 0           0.1159929   0.0979877
## Lag 100        0.1418886   0.1021546
## Lag 500        0.1445076   0.1134636
## Lag 1000       0.1521923   0.1278208
## Lag 5000       -0.1686025  -0.1687524
##
## , , traitbwt.2.mother
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0           0.05082067  -0.011552553
## Lag 100         0.04450359  -0.023686867
## Lag 500         0.04179199  0.004658431
## Lag 1000        0.02883236  -0.006496015
## Lag 5000        0.04112469  0.028334393
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0           0.11172969  0.04826105
## Lag 100         0.10951168  0.03784269
## Lag 500         0.08924566  0.03087286
## Lag 1000        0.05576158  0.03858511
## Lag 5000        -0.02612490 0.02719676
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0           -0.011552553 -0.068290432
## Lag 100         -0.023686867 -0.061106862
## Lag 500         0.004658431 -0.008368994
## Lag 1000        -0.006496015 0.002748701
## Lag 5000        0.028334393  0.008620918
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           0.004891297 -0.095372303
## Lag 100         0.018900269 -0.050527725
## Lag 500         0.034858608 -0.018316704
## Lag 1000        0.021136189 -0.004218270
## Lag 5000        -0.083516144 0.008525728
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           0.11172969  0.004891297
## Lag 100         0.10951168  0.018900269
## Lag 500         0.08924566  0.034858608
## Lag 1000        0.05576158  0.021136189
## Lag 5000        -0.02612490 -0.083516144
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           0.15871038  0.09891371
## Lag 100         0.13316302  0.07837814
## Lag 500         0.10976919  0.06061434
## Lag 1000        0.08100730  0.06929365
## Lag 5000        -0.02040836 -0.01741179
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           0.04826105  -0.095372303
## Lag 100         0.03784269  -0.050527725

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## Lag 500          0.03087286      -0.018316704
## Lag 1000         0.03858511      -0.004218270
## Lag 5000          0.02719676      0.008525728
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             0.09891371      -0.033866092
## Lag 100            0.07837814      -0.011360083
## Lag 500            0.06061434      0.012969192
## Lag 1000           0.06929365      0.045752648
## Lag 5000           -0.01741179     -0.001193243
## traittarsus.1:byear traitbwt.1:byear traittarsus.2:byear
## Lag 0              0.02899206      0.05926846      0.0024956756
## Lag 100             -0.01583481     0.05520016      0.0239528518
## Lag 500              0.03917478     0.01148328      0.0002419691
## Lag 1000             -0.03694049     0.02504159      0.0667961149
## Lag 5000             -0.06196001     0.02957462      0.0049061407
## traitbwt.2:byear traittarsus.1:mother traitbwt.1:mother
## Lag 0              -0.007859673    -0.05622667      0.14519001
## Lag 100             0.010189964    -0.05585546      0.13743423
## Lag 500              0.007357905    0.02124383      0.03427438
## Lag 1000             0.029340288    0.06934074     -0.01797194
## Lag 5000              0.011164278    0.01230641      0.01184408
## traittarsus.2:mother traitbwt.2:mother
## Lag 0              -0.329182837    1.000000000
## Lag 100             -0.235728358    0.284175599
## Lag 500             -0.148665968    0.029162936
## Lag 1000             -0.007592145    -0.008559910
## Lag 5000              0.006615767    0.006917489
## traittarsus.1:traittarsus.1:units traitbwt.1:traittarsus.1:units
## Lag 0                -0.04849521     -0.01247903
## Lag 100               -0.04229095     -0.01179876
## Lag 500               -0.06190225     -0.03479262
## Lag 1000              -0.06244621     -0.01333269
## Lag 5000               -0.04472803     -0.03405548
## traittarsus.2:traittarsus.1:units traitbwt.2:traittarsus.1:units
## Lag 0                  0.06299708      0.07575889
## Lag 100                 0.05582189      0.06465068
## Lag 500                 0.05304268      0.04802379
## Lag 1000                0.07638864      0.07282218
## Lag 5000                 0.04062409      0.05791562
## traittarsus.1:traitbwt.1:units traitbwt.1:traitbwt.1:units
## Lag 0                 -0.01247903     -0.009223763
## Lag 100                -0.01179876     -0.004801511
## Lag 500                -0.03479262     -0.021159851
## Lag 1000               -0.01333269     -0.005423005
## Lag 5000                -0.03405548     -0.040128813
## traittarsus.2:traitbwt.1:units traitbwt.2:traitbwt.1:units
## Lag 0                  0.07595500      0.08248472

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## Lag 100           0.06741840           0.07062568
## Lag 500           0.04493967           0.04348999
## Lag 1000          0.05963657           0.06140170
## Lag 5000          0.05397711           0.06956257
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0             0.06299708           0.07595500
## Lag 100           0.05582189           0.06741840
## Lag 500           0.05304268           0.04493967
## Lag 1000          0.07638864           0.05963657
## Lag 5000          0.04062409           0.05397711
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0             -0.063098741          -0.063844474
## Lag 100           -0.069885333          -0.060995678
## Lag 500           -0.075115678          -0.052589081
## Lag 1000          -0.091682781          -0.072696501
## Lag 5000          0.004523339          0.009847437
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0             0.07575889           0.08248472
## Lag 100           0.06465068           0.07062568
## Lag 500           0.04802379           0.04348999
## Lag 1000          0.07282218           0.06140170
## Lag 5000          0.05791562           0.06956257
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0             -0.063844474          -0.079128573
## Lag 100           -0.060995678          -0.038220443
## Lag 500           -0.052589081          -0.022953575
## Lag 1000          -0.072696501          -0.037549862
## Lag 5000          0.009847437          -0.001314642
##
## , , traittarsus.1:traittarsus.1.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0             -0.69647928           -0.53741563
## Lag 100           -0.62623971           -0.47900505
## Lag 500           -0.45854096           -0.35990127
## Lag 1000          -0.34465891           -0.24167598
## Lag 5000          -0.04540565           -0.08362331
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0             -0.4851021            -0.3969630
## Lag 100           -0.4391087            -0.3672331
## Lag 500           -0.3514192            -0.3303341
## Lag 1000          -0.2817756            -0.2606715
## Lag 5000          -0.1379797            -0.1850268
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0             -0.53741563           -0.2245265
## Lag 100           -0.47900505           -0.1935129
## Lag 500           -0.35990127           -0.1387130

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## Lag 1000           -0.24167598      -0.1103208
## Lag 5000           -0.08362331      -0.1059888
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              -0.3729388       -0.2547414
## Lag 100             -0.3377067       -0.2223712
## Lag 500             -0.2941086       -0.2173660
## Lag 1000            -0.2338232       -0.1759256
## Lag 5000            -0.1733008       -0.1754188
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              -0.4851021       -0.3729388
## Lag 100             -0.4391087       -0.3377067
## Lag 500             -0.3514192       -0.2941086
## Lag 1000            -0.2817756       -0.2338232
## Lag 5000            -0.1379797       -0.1733008
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              -0.10376953      -0.1451563
## Lag 100             -0.07852837      -0.1303753
## Lag 500             -0.06730801      -0.1313359
## Lag 1000            -0.06713014      -0.1425029
## Lag 5000            -0.15697371      -0.2346428
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              -0.3969630       -0.2547414
## Lag 100             -0.3672331       -0.2223712
## Lag 500             -0.3303341       -0.2173660
## Lag 1000            -0.2606715       -0.1759256
## Lag 5000            -0.1850268       -0.1754188
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              -0.1451563       -0.1310687
## Lag 100             -0.1303753       -0.1210976
## Lag 500             -0.1313359       -0.1454273
## Lag 1000            -0.1425029       -0.1484290
## Lag 5000            -0.2346428       -0.2608162
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              -0.14506604      -0.043558650      -0.04924174
## Lag 100             -0.12498034      -0.053994536      -0.03871946
## Lag 500             -0.05225308      -0.005205432      -0.05590755
## Lag 1000            -0.05643316      -0.037921140      0.00532337
## Lag 5000            -0.01904738      -0.070857345      0.03761236
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              0.01051817      -0.28511353      0.01973629
## Lag 100             0.03224747      -0.21849925      0.02374680
## Lag 500             0.03546917      -0.12897815      0.01858679
## Lag 1000            0.01230618      -0.03335119      0.03274960
## Lag 5000            0.02825979      -0.05125552      0.04540862
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0              -0.003303414     -0.048495208
## Lag 100             -0.027103529     -0.040346222

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## Lag 500      -0.025758939      0.002369226
## Lag 1000     -0.025407178      -0.024008377
## Lag 5000     -0.010658268      0.012747460
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0          1.000000000      0.71266479
## Lag 100        0.69601632      0.48602800
## Lag 500        0.44391095      0.33331074
## Lag 1000       0.32879898      0.24626220
## Lag 5000       0.09190517      0.09278524
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0         -0.2592850      -0.2645685
## Lag 100        -0.2655767      -0.2702771
## Lag 500        -0.2886811      -0.2950022
## Lag 1000       -0.2858651      -0.3159051
## Lag 5000       -0.2005196      -0.2351461
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0          0.71266479      0.3606694
## Lag 100        0.48602800      0.2474550
## Lag 500        0.33331074      0.1595171
## Lag 1000       0.24626220      0.1323001
## Lag 5000       0.09278524      0.1142750
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0         -0.2009925      -0.1971901
## Lag 100        -0.2133808      -0.2127295
## Lag 500        -0.2513257      -0.2534202
## Lag 1000       -0.2644654      -0.2860502
## Lag 5000       -0.2208762      -0.2453635
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0         -0.2592850      -0.2009925
## Lag 100        -0.2655767      -0.2133808
## Lag 500        -0.2886811      -0.2513257
## Lag 1000       -0.2858651      -0.2644654
## Lag 5000       -0.2005196      -0.2208762
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0          0.09543017      0.11602826
## Lag 100        0.06131639      0.09650511
## Lag 500        0.06137108      0.11179013
## Lag 1000       0.08079826      0.13634360
## Lag 5000       0.18979188      0.24893895
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0         -0.2645685      -0.1971901
## Lag 100        -0.2702771      -0.2127295
## Lag 500        -0.2950022      -0.2534202
## Lag 1000       -0.3159051      -0.2860502
## Lag 5000       -0.2351461      -0.2453635
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0          0.11602826      0.10062986

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## Lag 100          0.09650511      0.09093566
## Lag 500          0.11179013      0.12421446
## Lag 1000         0.13634360      0.14583460
## Lag 5000         0.24893895      0.25949670
##
## , , traitbwt.1:traittarsus.1.units
##
##           traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0            -0.5806661      -0.7822168
## Lag 100          -0.5195230      -0.6845208
## Lag 500          -0.3880444      -0.4391722
## Lag 1000         -0.2850697      -0.3219049
## Lag 5000         -0.1858774      -0.1615545
##           traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0            -0.3551367      -0.4423544
## Lag 100          -0.3116917      -0.4053917
## Lag 500          -0.2504438      -0.3222535
## Lag 1000         -0.2164808      -0.2714011
## Lag 5000         -0.1964028      -0.1739972
##           traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0            -0.7822168      -0.5785722
## Lag 100          -0.6845208      -0.5036377
## Lag 500          -0.4391722      -0.3105590
## Lag 1000         -0.3219049      -0.2500610
## Lag 5000         -0.1615545      -0.1051180
##           traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            -0.3850991      -0.3630159
## Lag 100          -0.3306453      -0.3097920
## Lag 500          -0.2840530      -0.2549302
## Lag 1000         -0.2403468      -0.2124423
## Lag 5000         -0.2587649      -0.2036370
##           traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            -0.3551367      -0.3850991
## Lag 100          -0.3116917      -0.3306453
## Lag 500          -0.2504438      -0.2840530
## Lag 1000         -0.2164808      -0.2403468
## Lag 5000         -0.1964028      -0.2587649
##           traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            -0.10398125     -0.1467921
## Lag 100          -0.07656994     -0.1228344
## Lag 500          -0.05666776     -0.1265197
## Lag 1000         -0.08628503     -0.1605079
## Lag 5000         -0.11080574     -0.1945688
##           traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            -0.4423544      -0.3630159
## Lag 100          -0.4053917      -0.3097920
## Lag 500          -0.3222535      -0.2549302

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## Lag 1000           -0.2714011          -0.2124423
## Lag 5000           -0.1739972          -0.2036370
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              -0.1467921          -0.1324172
## Lag 100             -0.1228344          -0.1123196
## Lag 500             -0.1265197          -0.1443413
## Lag 1000            -0.1605079          -0.1717954
## Lag 5000            -0.1945688          -0.2243561
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              -0.10471675         0.01021951        0.0004983836
## Lag 100             -0.08194667         0.01380504       -0.0044687157
## Lag 500             -0.01278894         0.03290932       -0.0005650915
## Lag 1000            -0.06407190         -0.02170329        0.0530406940
## Lag 5000            -0.07516029         -0.06130238       -0.0224119290
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              0.02175859          -0.039522955       0.03665672
## Lag 100            0.03477326          -0.019056605       0.05315654
## Lag 500            -0.01973469          -0.002626844       0.03028858
## Lag 1000            0.04003712          0.003169525       0.06449203
## Lag 5000            0.02550062          -0.010727363       0.04503907
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0              -0.03359578         -0.012479028
## Lag 100             -0.04488434         -0.002934766
## Lag 500             -0.06697376          0.021613573
## Lag 1000            -0.08457909          0.019481160
## Lag 5000            -0.08600099          0.051485249
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  0.7126648          1.0000000
## Lag 100                0.4891102          0.6680602
## Lag 500                0.3020606          0.3997173
## Lag 1000               0.2345067          0.2999511
## Lag 5000                0.1759098          0.1561292
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                  -0.1706739         -0.1801399
## Lag 100                -0.1802375         -0.1994176
## Lag 500                -0.2103921         -0.2440747
## Lag 1000               -0.2087782         -0.2510893
## Lag 5000                -0.1812853         -0.1905911
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  1.0000000          0.8197350
## Lag 100                0.6680602          0.5536026
## Lag 500                0.3997173          0.3326177
## Lag 1000               0.2999511          0.2334663
## Lag 5000                0.1561292          0.1202757
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0                  -0.1666030          -0.1700866
## Lag 100                 -0.1812316          -0.1939150

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## Lag 500           -0.2194596      -0.2430469
## Lag 1000          -0.2384353      -0.2743602
## Lag 5000          -0.2008085      -0.2014312
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              -0.1706739      -0.1666030
## Lag 100             0.1802375      -0.1812316
## Lag 500             0.2103921      -0.2194596
## Lag 1000            0.2087782      -0.2384353
## Lag 5000            0.1812853      -0.2008085
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0               0.10875399     0.13276682
## Lag 100              0.05904373     0.09500984
## Lag 500              0.06612909     0.11639092
## Lag 1000             0.10572947     0.15285062
## Lag 5000             0.15545551     0.20383203
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0               -0.1801399     -0.1700866
## Lag 100              -0.1994176     -0.1939150
## Lag 500              -0.2440747     -0.2430469
## Lag 1000             -0.2510893     -0.2743602
## Lag 5000             -0.1905911     -0.2014312
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0               0.13276682     0.11344391
## Lag 100              0.09500984     0.08651744
## Lag 500              0.11639092     0.13209239
## Lag 1000             0.15285062     0.15588819
## Lag 5000             0.20383203     0.21723655
##
## , , traittarsus.2:traittarsus.1.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0               0.1834123      0.124710076
## Lag 100              0.1968309      0.130500544
## Lag 500              0.2068864      0.139328738
## Lag 1000             0.1939570      0.142277480
## Lag 5000             0.1325680      0.009474997
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0               0.1906441      0.13328596
## Lag 100              0.2063827      0.13989222
## Lag 500              0.2249905      0.15953298
## Lag 1000             0.2243627      0.17422740
## Lag 5000             0.1612708      0.09845918
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0               0.124710076     0.08095300
## Lag 100              0.130500544     0.07950890
## Lag 500              0.139328738     0.07894049
## Lag 1000             0.142277480     0.08409986

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## Lag 5000          0.009474997      -0.03476503
## traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            0.13041646       0.06808542
## Lag 100           0.13619199       0.07439607
## Lag 500           0.15236746       0.10133254
## Lag 1000          0.16354531       0.12513909
## Lag 5000          0.09727094       0.09702098
## traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            0.1906441        0.13041646
## Lag 100           0.2063827        0.13619199
## Lag 500           0.2249905        0.15236746
## Lag 1000          0.2243627        0.16354531
## Lag 5000          0.1612708        0.09727094
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            0.1881270        0.1805175
## Lag 100           0.2019192        0.1895957
## Lag 500           0.2143083        0.2038416
## Lag 1000          0.2049762        0.2031078
## Lag 5000          0.2015150        0.2236163
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            0.13328596       0.06808542
## Lag 100           0.13989222       0.07439607
## Lag 500           0.15953298       0.10133254
## Lag 1000          0.17422740       0.12513909
## Lag 5000          0.09845918       0.09702098
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            0.1805175        0.1307188
## Lag 100           0.1895957        0.1387545
## Lag 500           0.2038416        0.1596754
## Lag 1000          0.2031078        0.1718471
## Lag 5000          0.2236163        0.2388384
## traittarsus.1:byear traitbwt.1:byear traittarsus.2:byear
## Lag 0            -0.03847015       0.05293519       0.06126409
## Lag 100           -0.03144044       0.04956734       0.06250612
## Lag 500           -0.01780154       0.05591029       0.07921148
## Lag 1000          -0.02643841       0.04337011       0.09113494
## Lag 5000          -0.01140358       0.03031863      -0.02594409
## traitbwt.2:byear traittarsus.1:mother traitbwt.1:mother
## Lag 0            -0.041798551      0.1697424        0.04294093
## Lag 100           -0.034073821      0.1688986        0.04145441
## Lag 500           -0.048890409      0.1720568        0.02794912
## Lag 1000          -0.040999443      0.1790535        0.02144402
## Lag 5000          -0.002784642      0.1836021      -0.02796491
## traittarsus.2:mother traitbwt.2:mother
## Lag 0            -0.03239456       0.06299708
## Lag 100           -0.03271179       0.06442310
## Lag 500           -0.02932124       0.06841853

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## Lag 1000      -0.01370787      0.05547883
## Lag 5000      -0.04156138      0.02258640
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.2592850      -0.17067395
## Lag 100         -0.2666810      -0.16983312
## Lag 500         -0.2761829      -0.17499382
## Lag 1000        -0.2748628      -0.18004636
## Lag 5000        -0.2185399      -0.04226407
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            1.0000000      0.9603400
## Lag 100          0.9783825      0.9444274
## Lag 500          0.9365031      0.9054416
## Lag 1000         0.8964880      0.8617258
## Lag 5000         0.6039003      0.6022291
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.17067395     -0.12741602
## Lag 100          -0.16983312     -0.12362277
## Lag 500          -0.17499382     -0.11972769
## Lag 1000         -0.18004636     -0.12502460
## Lag 5000         -0.04226407     0.01790192
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            0.9609935      0.9034487
## Lag 100          0.9411829      0.8893146
## Lag 500          0.8990207      0.8539291
## Lag 1000         0.8622161      0.8157349
## Lag 5000         0.5753236      0.5773694
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            1.0000000      0.9609935
## Lag 100          0.9783825      0.9411829
## Lag 500          0.9365031      0.8990207
## Lag 1000         0.8964880      0.8622161
## Lag 5000         0.6039003      0.5753236
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.1856903      -0.1801544
## Lag 100          -0.1921763      -0.1830867
## Lag 500          -0.2110415      -0.2009371
## Lag 1000         -0.2033807      -0.1989511
## Lag 5000         -0.1866476      -0.2075534
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            0.9603400      0.9034487
## Lag 100          0.9444274      0.8893146
## Lag 500          0.9054416      0.8539291
## Lag 1000         0.8617258      0.8157349
## Lag 5000         0.6022291      0.5773694
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           -0.1801544      -0.1345421
## Lag 100          -0.1830867      -0.1370653

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## Lag 500           -0.2009371          -0.1566901
## Lag 1000          -0.1989511          -0.1643648
## Lag 5000          -0.2075534          -0.2123302
##
## , , traitbwt.2:traittarsus.1.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.1843936          0.13991929
## Lag 100             0.1916063          0.14051238
## Lag 500             0.1997384          0.15350375
## Lag 1000            0.1779242          0.14333349
## Lag 5000            0.1059309          0.02408511
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.2159001          0.18941930
## Lag 100             0.2292097          0.19756460
## Lag 500             0.2470830          0.21294438
## Lag 1000            0.2430281          0.21181509
## Lag 5000            0.1564420          0.09549646
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.13991929         0.11864383
## Lag 100             0.14051238         0.11475981
## Lag 500             0.15350375         0.12191537
## Lag 1000            0.14333349         0.12719171
## Lag 5000            0.02408511        -0.00131242
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.1778597          0.1473359
## Lag 100             0.1832967          0.1558541
## Lag 500             0.2035349          0.1829320
## Lag 1000            0.2085231          0.1947599
## Lag 5000            0.1096782          0.1076230
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.2159001          0.1778597
## Lag 100             0.2292097          0.1832967
## Lag 500             0.2470830          0.2035349
## Lag 1000            0.2430281          0.2085231
## Lag 5000            0.1564420          0.1096782
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.2148002          0.2380759
## Lag 100             0.2260582          0.2490150
## Lag 500             0.2434250          0.2659527
## Lag 1000            0.2396319          0.2621966
## Lag 5000            0.2311987          0.2321063
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.18941930         0.1473359
## Lag 100             0.19756460         0.1558541
## Lag 500             0.21294438         0.1829320
## Lag 1000            0.21181509         0.1947599

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## Lag 5000          0.09549646          0.1076230
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0             0.2380759           0.2116292
## Lag 100            0.2490150           0.2231338
## Lag 500            0.2659527           0.2429222
## Lag 1000           0.2621966           0.2441661
## Lag 5000           0.2321063           0.2236298
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            -0.032369970          0.05768879          0.06054337
## Lag 100           -0.031641858          0.05864059          0.06580478
## Lag 500           -0.001530924          0.04503326          0.07399288
## Lag 1000          -0.016005519          0.03161375          0.08585903
## Lag 5000          0.010619935          0.01730446         -0.02417570
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.032356483          0.1703875           0.012376673
## Lag 100           -0.034654911          0.1647608           0.013581085
## Lag 500           -0.052597308          0.1600416           0.005553001
## Lag 1000          -0.039591608          0.1872703         -0.009042770
## Lag 5000          0.004455343          0.1972374         -0.027185781
## traittarsus.2.mother traitbwt.2.mother
## Lag 0            -0.03846399          0.07575889
## Lag 100           -0.03390447          0.06853459
## Lag 500           -0.02862102          0.06274714
## Lag 1000          -0.01799649          0.06065063
## Lag 5000          -0.05555704          0.02202828
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0             -0.2645685          -0.18013988
## Lag 100            -0.2616234          -0.17607364
## Lag 500            -0.2652457          -0.18034511
## Lag 1000           -0.2671219          -0.17702256
## Lag 5000           -0.2093233          -0.06028968
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.9603400          1.0000000
## Lag 100             0.9432077          0.9762869
## Lag 500             0.9124477          0.9252632
## Lag 1000            0.8838890          0.8792611
## Lag 5000            0.6117319          0.6007465
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            -0.18013988          -0.15026999
## Lag 100           -0.17607364          -0.14749648
## Lag 500           -0.18034511          -0.15145709
## Lag 1000          -0.17702256          -0.15538707
## Lag 5000          -0.06028968          -0.02621414
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.9303986          0.9527160
## Lag 100             0.9147001          0.9321247
## Lag 500              0.8850543          0.8884276

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## Lag 1000          0.8576398          0.8431799
## Lag 5000          0.5876791          0.5807055
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0             0.9603400          0.9303986
## Lag 100            0.9432077          0.9147001
## Lag 500            0.9124477          0.8850543
## Lag 1000           0.8838890          0.8576398
## Lag 5000           0.6117319          0.5876791
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0             -0.2006563         -0.2276707
## Lag 100            -0.2124295         -0.2354203
## Lag 500            -0.2379446         -0.2556157
## Lag 1000           -0.2389714         -0.2561306
## Lag 5000           -0.2130152         -0.2137306
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              1.0000000          0.9527160
## Lag 100             0.9762869          0.9321247
## Lag 500             0.9252632          0.8884276
## Lag 1000            0.8792611          0.8431799
## Lag 5000            0.6007465          0.5807055
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0             -0.2276707         -0.2128094
## Lag 100            -0.2354203         -0.2168142
## Lag 500            -0.2556157         -0.2324758
## Lag 1000           -0.2561306         -0.2373888
## Lag 5000           -0.2137306         -0.1974646
##
## , , traittarsus.1:traitbwt.1.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0             -0.5806661         -0.7822168
## Lag 100            -0.5195230         -0.6845208
## Lag 500            -0.3880444         -0.4391722
## Lag 1000           -0.2850697         -0.3219049
## Lag 5000            -0.1858774         -0.1615545
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0             -0.3551367         -0.4423544
## Lag 100            -0.3116917         -0.4053917
## Lag 500            -0.2504438         -0.3222535
## Lag 1000           -0.2164808         -0.2714011
## Lag 5000            -0.1964028         -0.1739972
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0             -0.7822168         -0.5785722
## Lag 100            -0.6845208         -0.5036377
## Lag 500            -0.4391722         -0.3105590
## Lag 1000           -0.3219049         -0.2500610
## Lag 5000            -0.1615545         -0.1051180

```

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##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           -0.3850991              -0.3630159
## Lag 100          -0.3306453              -0.3097920
## Lag 500          -0.2840530              -0.2549302
## Lag 1000         -0.2403468              -0.2124423
## Lag 5000         -0.2587649              -0.2036370
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           -0.3551367              -0.3850991
## Lag 100          -0.3116917              -0.3306453
## Lag 500          -0.2504438              -0.2840530
## Lag 1000         -0.2164808              -0.2403468
## Lag 5000         -0.1964028              -0.2587649
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           -0.10398125             -0.1467921
## Lag 100          -0.07656994             -0.1228344
## Lag 500          -0.05666776             -0.1265197
## Lag 1000         -0.08628503             -0.1605079
## Lag 5000         -0.11080574             -0.1945688
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           -0.4423544              -0.3630159
## Lag 100          -0.4053917              -0.3097920
## Lag 500          -0.3222535              -0.2549302
## Lag 1000         -0.2714011              -0.2124423
## Lag 5000         -0.1739972              -0.2036370
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0           -0.1467921              -0.1324172
## Lag 100          -0.1228344              -0.1123196
## Lag 500          -0.1265197              -0.1443413
## Lag 1000         -0.1605079              -0.1717954
## Lag 5000         -0.1945688              -0.2243561
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           -0.10471675             0.01021951            0.0004983836
## Lag 100          -0.08194667             0.01380504            -0.0044687157
## Lag 500          -0.01278894             0.03290932            -0.0005650915
## Lag 1000         -0.06407190             -0.02170329            0.0530406940
## Lag 5000         -0.07516029             -0.06130238            -0.0224119290
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            0.02175859             -0.039522955            0.03665672
## Lag 100          0.03477326             -0.019056605            0.05315654
## Lag 500          -0.01973469             -0.002626844            0.03028858
## Lag 1000         0.04003712              0.003169525            0.06449203
## Lag 5000         0.02550062             -0.010727363            0.04503907
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0           -0.03359578             -0.012479028
## Lag 100          -0.04488434             -0.002934766
## Lag 500          -0.06697376              0.021613573
## Lag 1000         -0.08457909              0.019481160

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## Lag 5000      -0.08600099    0.051485249
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0          0.7126648     1.0000000
## Lag 100        0.4891102     0.6680602
## Lag 500        0.3020606     0.3997173
## Lag 1000       0.2345067     0.2999511
## Lag 5000       0.1759098     0.1561292
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0          -0.1706739    -0.1801399
## Lag 100        -0.1802375    -0.1994176
## Lag 500        -0.2103921    -0.2440747
## Lag 1000       -0.2087782    -0.2510893
## Lag 5000       -0.1812853    -0.1905911
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0          1.0000000     0.8197350
## Lag 100        0.6680602     0.5536026
## Lag 500        0.3997173     0.3326177
## Lag 1000       0.2999511     0.2334663
## Lag 5000       0.1561292     0.1202757
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0          -0.1666030    -0.1700866
## Lag 100        -0.1812316    -0.1939150
## Lag 500        -0.2194596    -0.2430469
## Lag 1000       -0.2384353    -0.2743602
## Lag 5000       -0.2008085    -0.2014312
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0          -0.1706739    -0.1666030
## Lag 100        -0.1802375    -0.1812316
## Lag 500        -0.2103921    -0.2194596
## Lag 1000       -0.2087782    -0.2384353
## Lag 5000       -0.1812853    -0.2008085
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0          0.10875399   0.13276682
## Lag 100        0.05904373   0.09500984
## Lag 500        0.06612909   0.11639092
## Lag 1000       0.10572947   0.15285062
## Lag 5000       0.15545551   0.20383203
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0          -0.1801399    -0.1700866
## Lag 100        -0.1994176    -0.1939150
## Lag 500        -0.2440747    -0.2430469
## Lag 1000       -0.2510893    -0.2743602
## Lag 5000       -0.1905911    -0.2014312
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0          0.13276682   0.11344391
## Lag 100        0.09500984   0.08651744
## Lag 500        0.11639092   0.13209239

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## Lag 1000          0.15285062          0.15588819
## Lag 5000          0.20383203          0.21723655
##
## , , traitbwt.1:traitbwt.1.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              -0.29723299         -0.6444049
## Lag 100             -0.26194069        -0.5583089
## Lag 500             -0.17460653        -0.3429615
## Lag 1000            -0.09216281        -0.2402326
## Lag 5000            -0.14907830        -0.1125298
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              -0.16353731         -0.33002093
## Lag 100             -0.13444098        -0.30242196
## Lag 500             -0.07860312        -0.20182430
## Lag 1000            -0.07437860        -0.16489835
## Lag 5000            -0.15240021        -0.07855392
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              -0.6444049          -0.72140390
## Lag 100             -0.5583089          -0.60150571
## Lag 500             -0.3429615          -0.35781515
## Lag 1000            -0.2402326          -0.29870892
## Lag 5000            -0.1125298          -0.04717998
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              -0.2793636          -0.3902410
## Lag 100             -0.2441693          -0.3411243
## Lag 500             -0.1881361          -0.2431591
## Lag 1000            -0.1803300          -0.2025644
## Lag 5000            -0.2243905          -0.1275516
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              -0.16353731         -0.2793636
## Lag 100             -0.13444098        -0.2441693
## Lag 500             -0.07860312        -0.1881361
## Lag 1000            -0.07437860        -0.1803300
## Lag 5000            -0.15240021        -0.2243905
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              -0.10182068         -0.14376430
## Lag 100             -0.07461870         -0.12129614
## Lag 500             -0.03475445         -0.09690301
## Lag 1000            -0.07582375         -0.13287333
## Lag 5000            -0.06172035         -0.10552070
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              -0.33002093         -0.3902410
## Lag 100             -0.30242196         -0.3411243
## Lag 500             -0.20182430         -0.2431591
## Lag 1000            -0.16489835         -0.2025644
## Lag 5000            -0.07855392         -0.1275516

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##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                  -0.14376430      -0.1400636
## Lag 100                 -0.12129614      -0.1267043
## Lag 500                 -0.09690301      -0.1357567
## Lag 1000                -0.13287333      -0.1570152
## Lag 5000                -0.10552070      -0.1203712
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0                  -0.071954690     -0.028183637     0.04524320
## Lag 100                 -0.044770596     0.002882978     0.03767827
## Lag 500                 -0.005911696     0.003838461     0.01619977
## Lag 1000                -0.056263708     0.008443638     0.05346543
## Lag 5000                -0.087263536     -0.023015448    -0.01948411
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0                  0.008779544      0.028265685    -0.007047622
## Lag 100                 0.013764785      0.026135300     0.054993361
## Lag 500                 -0.033989612      0.002688467     0.096589744
## Lag 1000                0.025220721      0.005441149     0.099042795
## Lag 5000                0.047628522      -0.007718921     0.047775710
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0                  -0.04785615      -0.009223763
## Lag 100                 -0.05001172      0.003907093
## Lag 500                 -0.07099090      0.044036802
## Lag 1000                -0.05777658      0.037746309
## Lag 5000                -0.09694236      0.069793617
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  0.36066940      0.8197350
## Lag 100                 0.25257032      0.5640460
## Lag 500                 0.11373127      0.3085415
## Lag 1000                0.07190253      0.2121382
## Lag 5000                0.13152446      0.1159014
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                  -0.1274160      -0.1502700
## Lag 100                 -0.1314926      -0.1663342
## Lag 500                 -0.1561690      -0.2058338
## Lag 1000                -0.1493199      -0.1963753
## Lag 5000                -0.1353839      -0.1194476
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  0.8197350      1.00000000
## Lag 100                 0.5640460      0.66059880
## Lag 500                 0.3085415      0.35457689
## Lag 1000                0.2121382      0.23677816
## Lag 5000                0.1159014      0.06278176
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0                  -0.1670809      -0.1902055
## Lag 100                 -0.1716919      -0.2107271
## Lag 500                 -0.1984597      -0.2462995
## Lag 1000                -0.2124167      -0.2635936

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## Lag 5000           -0.1444361           -0.1252641
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              -0.1274160           -0.1670809
## Lag 100             -0.1314926           -0.1716919
## Lag 500             -0.1561690           -0.1984597
## Lag 1000            -0.1493199           -0.2124167
## Lag 5000            -0.1353839           -0.1444361
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0               0.11081955          0.13741353
## Lag 100              0.06249231          0.09952814
## Lag 500              0.04779758          0.10118050
## Lag 1000             0.09619023          0.13533567
## Lag 5000             0.09963826          0.10989738
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0               -0.1502700          -0.1902055
## Lag 100              -0.1663342          -0.2107271
## Lag 500              -0.2058338          -0.2462995
## Lag 1000             -0.1963753          -0.2635936
## Lag 5000             -0.1194476          -0.1252641
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0               0.13741353          0.1314508
## Lag 100              0.09952814          0.1030693
## Lag 500              0.10118050          0.1373268
## Lag 1000             0.13533567          0.1500819
## Lag 5000             0.10989738          0.1138216
##
## , , traittarsus.2:traitbwt.1.units
##
## traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0               0.1148525          0.10037928
## Lag 100              0.1303566          0.10703233
## Lag 500              0.1453432          0.12751184
## Lag 1000             0.1449075          0.12663225
## Lag 5000             0.1652756          0.03681818
## traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0               0.1588923          0.1392793
## Lag 100              0.1783256          0.1463284
## Lag 500              0.2040736          0.1612874
## Lag 1000             0.2230315          0.1896293
## Lag 5000             0.1750420          0.0827739
## traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0               0.10037928         0.12062597
## Lag 100              0.10703233         0.11559443
## Lag 500              0.12751184         0.11837717
## Lag 1000             0.12663225         0.10630852
## Lag 5000             0.03681818         -0.04309778
## traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal

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## Lag 0           0.1142625      0.09800038
## Lag 100        0.1363961      0.11259128
## Lag 500        0.1781857      0.14358029
## Lag 1000       0.1973172      0.16595148
## Lag 5000       0.1261037      0.08832747
##     traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0           0.1588923      0.1142625
## Lag 100         0.1783256      0.1363961
## Lag 500         0.2040736      0.1781857
## Lag 1000        0.2230315      0.1973172
## Lag 5000        0.1750420      0.1261037
##     traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0           0.2227014      0.2177824
## Lag 100         0.2383512      0.2331011
## Lag 500         0.2644369      0.2625082
## Lag 1000        0.2577927      0.2670893
## Lag 5000        0.1772736      0.1885761
##     traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0           0.1392793      0.09800038
## Lag 100         0.1463284      0.11259128
## Lag 500         0.1612874      0.14358029
## Lag 1000        0.1896293      0.16595148
## Lag 5000        0.0827739      0.08832747
##     traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0           0.2177824      0.1727839
## Lag 100         0.2331011      0.1869982
## Lag 500         0.2625082      0.2195418
## Lag 1000        0.2670893      0.2381230
## Lag 5000        0.1885761      0.1980800
##     traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           -0.027499725    0.04996250    0.04569537
## Lag 100          -0.023937020   0.05747431    0.04863214
## Lag 500          -0.009997799   0.05757848    0.05649481
## Lag 1000         -0.028336813   0.03932991    0.08794045
## Lag 5000         -0.028421868   0.02850733    -0.02604903
##     traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0           -0.0558751650   0.1570314     -0.002752353
## Lag 100          -0.0390036691   0.1577619     -0.005577721
## Lag 500          -0.0604361094   0.1762753     -0.008689304
## Lag 1000         -0.0544777158   0.1711484     -0.003879605
## Lag 5000         0.0009703848   0.1611550     0.001378773
##     traittarsus.2.mother traitbwt.2.mother
## Lag 0           -0.04476711    0.07595500
## Lag 100          -0.04782246   0.07837204
## Lag 500          -0.04024464   0.07773574
## Lag 1000         -0.01332984   0.04580976
## Lag 5000         -0.03731338   0.02497852

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##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  -0.2009925      -0.16660301
## Lag 100                 -0.2135650     -0.16581994
## Lag 500                 -0.2387289     -0.17888627
## Lag 1000                -0.2382720     -0.17175572
## Lag 5000                -0.2290178     -0.06304922
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                  0.9609935      0.9303986
## Lag 100                 0.9414150      0.9167945
## Lag 500                 0.9016587      0.8805530
## Lag 1000                0.8646228      0.8343108
## Lag 5000                0.5874768      0.5773601
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  -0.16660301    -0.167080898
## Lag 100                 -0.16581994    -0.156138622
## Lag 500                 -0.17888627    -0.151199166
## Lag 1000                -0.17175572    -0.139083793
## Lag 5000                -0.06304922    0.007994506
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0                  1.0000000      0.9457989
## Lag 100                 0.9647777      0.9182192
## Lag 500                 0.8967026      0.8587466
## Lag 1000                0.8499110      0.8058180
## Lag 5000                0.5555747      0.5516644
##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0                  0.9609935      1.0000000
## Lag 100                 0.9414150      0.9647777
## Lag 500                 0.9016587      0.8967026
## Lag 1000                0.8646228      0.8499110
## Lag 5000                0.5874768      0.5555747
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0                  -0.2146171     -0.2167857
## Lag 100                 -0.2184104     -0.2221790
## Lag 500                 -0.2572470     -0.2584745
## Lag 1000                -0.2569503     -0.2596379
## Lag 5000                -0.1645148     -0.1699811
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0                  0.9303986      0.9457989
## Lag 100                 0.9167945      0.9182192
## Lag 500                 0.8805530      0.8587466
## Lag 1000                0.8343108      0.8058180
## Lag 5000                0.5773601      0.5516644
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0                  -0.2167857     -0.1800172
## Lag 100                 -0.2221790     -0.1849007
## Lag 500                 -0.2584745     -0.2183863
## Lag 1000                -0.2596379     -0.2235166

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## Lag 5000           -0.1699811          -0.1684397
##
## , , traitbwt.2:traitbwt.1.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.1045202          0.10675486
## Lag 100             0.1144301          0.10909010
## Lag 500             0.1244192          0.13415361
## Lag 1000            0.1101292          0.11544191
## Lag 5000            0.1331335          0.04367713
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.1751169          0.18455878
## Lag 100             0.1936998          0.19497627
## Lag 500             0.2199790          0.20989089
## Lag 1000            0.2292872          0.21238609
## Lag 5000            0.1693953          0.06254965
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.10675486         0.15466547
## Lag 100             0.10909010         0.15753844
## Lag 500             0.13415361         0.17187977
## Lag 1000            0.11544191         0.14889411
## Lag 5000            0.04367713        -0.01393939
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.1618783          0.17860886
## Lag 100             0.1807130          0.20406604
## Lag 500             0.2315550          0.24587070
## Lag 1000            0.2433453          0.24000705
## Lag 5000            0.1313066          0.07959284
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.1751169          0.1618783
## Lag 100             0.1936998          0.1807130
## Lag 500             0.2199790          0.2315550
## Lag 1000            0.2292872          0.2433453
## Lag 5000            0.1693953          0.1313066
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.2339355          0.2581627
## Lag 100             0.2476485          0.2736295
## Lag 500             0.2790724          0.3078386
## Lag 1000            0.2842308          0.3166841
## Lag 5000            0.2101542          0.1923682
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.18455878         0.17860886
## Lag 100             0.19497627         0.20406604
## Lag 500             0.20989089         0.24587070
## Lag 1000            0.21238609         0.24000705
## Lag 5000            0.06254965         0.07959284
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal

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## Lag 0           0.2581627           0.2416106
## Lag 100        0.2736295           0.2608676
## Lag 500        0.3078386           0.2963880
## Lag 1000       0.3166841           0.3068141
## Lag 5000       0.1923682           0.1712560
##          traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0           -0.023701193      0.04700881    0.05361674
## Lag 100         -0.021944414      0.05837761    0.06114321
## Lag 500         -0.004019018      0.04124133    0.05472308
## Lag 1000        -0.014592442      0.02897141    0.08299624
## Lag 5000        -0.001455363      0.01713092   -0.03330245
##          traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0           -0.03458419       0.1501420     -0.030979909
## Lag 100         -0.02882752       0.1495980     -0.034573917
## Lag 500         -0.05044049       0.1542703     -0.037881298
## Lag 1000        -0.04326520       0.1753867     -0.049179995
## Lag 5000        0.01976572       0.1700994     -0.002621514
##          traittarsus.2.mother traitbwt.2.mother
## Lag 0           -0.05091781       0.08248472
## Lag 100         -0.04612708       0.07643625
## Lag 500         -0.04592155       0.07211789
## Lag 1000        -0.02428388       0.04810929
## Lag 5000        -0.05454058       0.02230047
##          traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0           -0.1971901      -0.17008659
## Lag 100         -0.2006748      -0.16466175
## Lag 500         -0.2146434      -0.17764629
## Lag 1000        -0.2152829      -0.15594317
## Lag 5000        -0.2163189      -0.07951097
##          traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0           0.9034487       0.9527160
## Lag 100         0.8867034       0.9318204
## Lag 500         0.8558412       0.8820862
## Lag 1000        0.8281770       0.8332104
## Lag 5000        0.5777198       0.5582022
##          traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.17008659     -0.1902055
## Lag 100         -0.16466175     -0.1881759
## Lag 500         -0.17764629     -0.1918480
## Lag 1000        -0.15594317     -0.1592987
## Lag 5000        -0.07951097     -0.0307547
##          traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0           0.9457989       1.0000000
## Lag 100         0.9171312       0.9591397
## Lag 500         0.8650556       0.8846934
## Lag 1000        0.8286470       0.8252821
## Lag 5000        0.5520607       0.5384118

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##          traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0                      0.9034487      0.9457989
## Lag 100                     0.8867034      0.9171312
## Lag 500                     0.8558412      0.8650556
## Lag 1000                    0.8281770      0.8286470
## Lag 5000                    0.5777198      0.5520607
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0                      -0.2187484     -0.2507033
## Lag 100                     -0.2283894     -0.2593514
## Lag 500                     -0.2686154     -0.2983604
## Lag 1000                    -0.2818961     -0.3082824
## Lag 5000                    -0.1951997     -0.1732268
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0                      0.9527160      1.0000000
## Lag 100                     0.9318204      0.9591397
## Lag 500                     0.8820862      0.8846934
## Lag 1000                    0.8332104      0.8252821
## Lag 5000                    0.5582022      0.5384118
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0                      -0.2507033     -0.2443613
## Lag 100                     -0.2593514     -0.2551879
## Lag 500                     -0.2983604     -0.2884325
## Lag 1000                    -0.3082824     -0.2942164
## Lag 5000                    -0.1732268     -0.1455134
##
## , , traittarsus.1:traittarsus.2.units
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                      0.1834123      0.124710076
## Lag 100                     0.1968309      0.130500544
## Lag 500                     0.2068864      0.139328738
## Lag 1000                    0.1939570      0.142277480
## Lag 5000                    0.1325680      0.009474997
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                      0.1906441      0.13328596
## Lag 100                     0.2063827      0.13989222
## Lag 500                     0.2249905      0.15953298
## Lag 1000                    0.2243627      0.17422740
## Lag 5000                    0.1612708      0.09845918
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                      0.124710076     0.08095300
## Lag 100                     0.130500544     0.07950890
## Lag 500                     0.139328738     0.07894049
## Lag 1000                    0.142277480     0.08409986
## Lag 5000                    0.009474997     -0.03476503
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                      0.13041646     0.06808542

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## Lag 100          0.13619199      0.07439607
## Lag 500          0.15236746      0.10133254
## Lag 1000         0.16354531      0.12513909
## Lag 5000         0.09727094      0.09702098
##   traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            0.1906441       0.13041646
## Lag 100          0.2063827       0.13619199
## Lag 500          0.2249905       0.15236746
## Lag 1000         0.2243627       0.16354531
## Lag 5000         0.1612708       0.09727094
##   traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            0.1881270       0.1805175
## Lag 100          0.2019192       0.1895957
## Lag 500          0.2143083       0.2038416
## Lag 1000         0.2049762       0.2031078
## Lag 5000         0.2015150       0.2236163
##   traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            0.13328596     0.06808542
## Lag 100          0.13989222     0.07439607
## Lag 500          0.15953298     0.10133254
## Lag 1000         0.17422740     0.12513909
## Lag 5000         0.09845918     0.09702098
##   traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            0.1805175       0.1307188
## Lag 100          0.1895957       0.1387545
## Lag 500          0.2038416       0.1596754
## Lag 1000         0.2031078       0.1718471
## Lag 5000         0.2236163       0.2388384
##   traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            -0.03847015    0.05293519      0.06126409
## Lag 100          -0.03144044    0.04956734      0.06250612
## Lag 500          -0.01780154    0.05591029      0.07921148
## Lag 1000         -0.02643841    0.04337011      0.09113494
## Lag 5000         -0.01140358    0.03031863      -0.02594409
##   traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.041798551   0.1697424       0.04294093
## Lag 100          -0.034073821   0.1688986       0.04145441
## Lag 500          -0.048890409   0.1720568       0.02794912
## Lag 1000         -0.040999443   0.1790535       0.02144402
## Lag 5000         -0.002784642   0.1836021      -0.02796491
##   traittarsus.2.mother traitbwt.2.mother
## Lag 0            -0.03239456   0.06299708
## Lag 100          -0.03271179   0.06442310
## Lag 500          -0.02932124   0.06841853
## Lag 1000         -0.01370787   0.05547883
## Lag 5000         -0.04156138   0.02258640
##   traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units

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## Lag 0           -0.2592850      -0.17067395
## Lag 100        -0.2666810      -0.16983312
## Lag 500        -0.2761829      -0.17499382
## Lag 1000       -0.2748628      -0.18004636
## Lag 5000       -0.2185399      -0.04226407
##   traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0           1.0000000      0.9603400
## Lag 100         0.9783825      0.9444274
## Lag 500         0.9365031      0.9054416
## Lag 1000        0.8964880      0.8617258
## Lag 5000        0.6039003      0.6022291
##   traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0           -0.17067395     -0.12741602
## Lag 100         -0.16983312     -0.12362277
## Lag 500         -0.17499382     -0.11972769
## Lag 1000        -0.18004636     -0.12502460
## Lag 5000        -0.04226407     0.01790192
##   traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0           0.9609935      0.9034487
## Lag 100         0.9411829      0.8893146
## Lag 500         0.8990207      0.8539291
## Lag 1000        0.8622161      0.8157349
## Lag 5000        0.5753236      0.5773694
##   traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0           1.0000000      0.9609935
## Lag 100         0.9783825      0.9411829
## Lag 500         0.9365031      0.8990207
## Lag 1000        0.8964880      0.8622161
## Lag 5000        0.6039003      0.5753236
##   traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.1856903     -0.1801544
## Lag 100         -0.1921763     -0.1830867
## Lag 500         -0.2110415     -0.2009371
## Lag 1000        -0.2033807     -0.1989511
## Lag 5000        -0.1866476     -0.2075534
##   traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0           0.9603400      0.9034487
## Lag 100         0.9444274      0.8893146
## Lag 500         0.9054416      0.8539291
## Lag 1000        0.8617258      0.8157349
## Lag 5000        0.6022291      0.5773694
##   traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           -0.1801544     -0.1345421
## Lag 100         -0.1830867     -0.1370653
## Lag 500         -0.2009371     -0.1566901
## Lag 1000        -0.1989511     -0.1643648
## Lag 5000        -0.2075534     -0.2123302

```

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## 
## , , traitbwt.1:traittarsus.2.units
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                  0.1148525      0.10037928
## Lag 100                 0.1303566      0.10703233
## Lag 500                 0.1453432      0.12751184
## Lag 1000                0.1449075      0.12663225
## Lag 5000                0.1652756      0.03681818
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                  0.1588923      0.1392793
## Lag 100                 0.1783256      0.1463284
## Lag 500                 0.2040736      0.1612874
## Lag 1000                0.2230315      0.1896293
## Lag 5000                0.1750420      0.0827739
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  0.10037928     0.12062597
## Lag 100                 0.10703233     0.11559443
## Lag 500                 0.12751184     0.11837717
## Lag 1000                0.12663225     0.10630852
## Lag 5000                0.03681818    -0.04309778
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                  0.1142625      0.09800038
## Lag 100                 0.1363961      0.11259128
## Lag 500                 0.1781857      0.14358029
## Lag 1000                0.1973172      0.16595148
## Lag 5000                0.1261037      0.08832747
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                  0.1588923      0.1142625
## Lag 100                 0.1783256      0.1363961
## Lag 500                 0.2040736      0.1781857
## Lag 1000                0.2230315      0.1973172
## Lag 5000                0.1750420      0.1261037
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                  0.2227014      0.2177824
## Lag 100                 0.2383512      0.2331011
## Lag 500                 0.2644369      0.2625082
## Lag 1000                0.2577927      0.2670893
## Lag 5000                0.1772736      0.1885761
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                  0.1392793      0.09800038
## Lag 100                 0.1463284      0.11259128
## Lag 500                 0.1612874      0.14358029
## Lag 1000                0.1896293      0.16595148
## Lag 5000                0.0827739      0.08832747
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                  0.2177824      0.1727839

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## Lag 100           0.2331011          0.1869982
## Lag 500           0.2625082          0.2195418
## Lag 1000          0.2670893          0.2381230
## Lag 5000          0.1885761          0.1980800
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            -0.027499725        0.04996250        0.04569537
## Lag 100          -0.023937020        0.05747431        0.04863214
## Lag 500          -0.009997799        0.05757848        0.05649481
## Lag 1000         -0.028336813        0.03932991        0.08794045
## Lag 5000         -0.028421868        0.02850733       -0.02604903
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.0558751650       0.1570314        -0.002752353
## Lag 100          -0.0390036691       0.1577619       -0.005577721
## Lag 500          -0.0604361094       0.1762753       -0.008689304
## Lag 1000         -0.0544777158       0.1711484       -0.003879605
## Lag 5000          0.0009703848       0.1611550        0.001378773
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0            -0.04476711        0.07595500
## Lag 100          -0.04782246        0.07837204
## Lag 500          -0.04024464        0.07773574
## Lag 1000         -0.01332984        0.04580976
## Lag 5000          0.03731338        0.02497852
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0             -0.2009925       -0.16660301
## Lag 100           -0.2135650       -0.16581994
## Lag 500           -0.2387289       -0.17888627
## Lag 1000          -0.2382720       -0.17175572
## Lag 5000          -0.2290178       -0.06304922
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.9609935        0.9303986
## Lag 100            0.9414150        0.9167945
## Lag 500            0.9016587        0.8805530
## Lag 1000           0.8646228        0.8343108
## Lag 5000            0.5874768        0.5773601
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0             -0.16660301       -0.167080898
## Lag 100           -0.16581994       -0.156138622
## Lag 500           -0.17888627       -0.151199166
## Lag 1000          -0.17175572       -0.139083793
## Lag 5000          -0.06304922        0.007994506
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              1.0000000        0.9457989
## Lag 100            0.9647777        0.9182192
## Lag 500            0.8967026        0.8587466
## Lag 1000           0.8499110        0.8058180
## Lag 5000            0.5555747        0.5516644
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units

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```

## Lag 0           0.9609935   1.0000000
## Lag 100        0.9414150   0.9647777
## Lag 500        0.9016587   0.8967026
## Lag 1000       0.8646228   0.8499110
## Lag 5000       0.5874768   0.5555747
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0           -0.2146171  -0.2167857
## Lag 100         -0.2184104  -0.2221790
## Lag 500         -0.2572470  -0.2584745
## Lag 1000        -0.2569503  -0.2596379
## Lag 5000        -0.1645148  -0.1699811
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0           0.9303986   0.9457989
## Lag 100         0.9167945   0.9182192
## Lag 500         0.8805530   0.8587466
## Lag 1000        0.8343108   0.8058180
## Lag 5000        0.5773601   0.5516644
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0           -0.2167857  -0.1800172
## Lag 100         -0.2221790  -0.1849007
## Lag 500         -0.2584745  -0.2183863
## Lag 1000        -0.2596379  -0.2235166
## Lag 5000        -0.1699811  -0.1684397
##
## , , traittarsus.2:traittarsus.2.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0           -0.09279221 -0.09677793
## Lag 100         -0.06813005 -0.09295582
## Lag 500         -0.02512501 -0.08099239
## Lag 1000        -0.02287160 -0.08741368
## Lag 5000        0.07141747  0.05570893
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0           -0.5600992  -0.5541670
## Lag 100         -0.5121175  -0.5169149
## Lag 500         -0.4106125  -0.4295480
## Lag 1000        -0.3066655  -0.3332778
## Lag 5000        0.2026604   0.1966380
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0           -0.09677793 -0.11346055
## Lag 100         -0.09295582 -0.11622763
## Lag 500         -0.08099239 -0.09958123
## Lag 1000        -0.08741368 -0.12997232
## Lag 5000        0.05570893  0.07322980
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0           -0.4185841  -0.3843067
## Lag 100         -0.3914286  -0.3603585

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## Lag 500           -0.3364209          -0.3074805
## Lag 1000          -0.2650097          -0.2595029
## Lag 5000           0.1809913          0.1272979
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              -0.5600992          -0.4185841
## Lag 100             -0.5121175          -0.3914286
## Lag 500             -0.4106125          -0.3364209
## Lag 1000            -0.3066655          -0.2650097
## Lag 5000             0.2026604          0.1809913
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              -0.91073694         -0.87501256
## Lag 100             -0.86504036         -0.83310883
## Lag 500             -0.70973046         -0.68729285
## Lag 1000            -0.55757696         -0.54103224
## Lag 5000             0.02356453         0.08002102
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              -0.5541670          -0.3843067
## Lag 100             -0.5169149          -0.3603585
## Lag 500             -0.4295480          -0.3074805
## Lag 1000            -0.3332778          -0.2595029
## Lag 5000             0.1966380          0.1272979
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              -0.87501256         -0.70436348
## Lag 100             -0.83310883         -0.67114014
## Lag 500             -0.68729285         -0.55786587
## Lag 1000            -0.54103224         -0.43429408
## Lag 5000             0.08002102         0.08588843
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              -0.0295325711        0.01013932         -0.05907849
## Lag 100             -0.0258002546        0.01023903         -0.03799824
## Lag 500             -0.0008409219        0.01222834         -0.02632830
## Lag 1000            -0.0645850913        -0.04476975         -0.04078215
## Lag 5000             0.0499883850        -0.01870262         0.03463537
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0               0.07499456          -0.05365944         0.08205572
## Lag 100              0.07846166          -0.07198121         0.07982740
## Lag 500              0.06122895          -0.08239796         0.07480035
## Lag 1000             0.04893466          -0.08171638         0.06996265
## Lag 5000             -0.05752959          -0.01133586        -0.02071995
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0                0.13607402          -0.063098741
## Lag 100              0.16280409          -0.056620767
## Lag 500              0.14111652          -0.077496866
## Lag 1000              0.12664216          -0.055395330
## Lag 5000              0.01649277          -0.003196996
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  0.09543017          0.10875399

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## Lag 100          0.06598285      0.08293245
## Lag 500          0.05295587      0.08321338
## Lag 1000         0.06481928      0.08697241
## Lag 5000         -0.09911194     -0.09382365
## traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            -0.18569026     -0.20065634
## Lag 100          -0.19134621     -0.20598290
## Lag 500          -0.19158726     -0.20439959
## Lag 1000         -0.17346895     -0.18874412
## Lag 5000         -0.04965337     -0.05447094
## traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            0.10875399      0.11081955
## Lag 100          0.08293245      0.09377309
## Lag 500          0.08321338      0.07955802
## Lag 1000         0.08697241      0.10737966
## Lag 5000         -0.09382365     -0.07165472
## traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            -0.21461708     -0.2187484
## Lag 100          -0.21502402     -0.2194216
## Lag 500          -0.20000700     -0.2023328
## Lag 1000         -0.19364240     -0.1958537
## Lag 5000         -0.02836013     -0.0258784
## traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            -0.18569026     -0.21461708
## Lag 100          -0.19134621     -0.21502402
## Lag 500          -0.19158726     -0.20000700
## Lag 1000         -0.17346895     -0.19364240
## Lag 5000         -0.04965337     -0.02836013
## traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0            1.0000000      0.92685734
## Lag 100          0.8892164      0.83343965
## Lag 500          0.7197831      0.67640907
## Lag 1000         0.5723349      0.53670401
## Lag 5000         -0.0393405     -0.08858763
## traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            -0.20065634     -0.2187484
## Lag 100          -0.20598290     -0.2194216
## Lag 500          -0.20439959     -0.2023328
## Lag 1000         -0.18874412     -0.1958537
## Lag 5000         -0.05447094     -0.0258784
## traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0            0.92685734      0.75421487
## Lag 100          0.83343965      0.68043059
## Lag 500          0.67640907      0.55259827
## Lag 1000         0.53670401      0.43176584
## Lag 5000         -0.08858763     -0.08998996
##

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## , , traitbwt.2:traittarsus.2.units
##
##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              -0.09066377      -0.12847847
## Lag 100             -0.06791265     -0.11660390
## Lag 500             -0.02813480     -0.08793510
## Lag 1000            -0.02464042     -0.10829057
## Lag 5000            0.09059977      0.04232726
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              -0.5208193      -0.6200400
## Lag 100             -0.4814739      -0.5760115
## Lag 500             -0.4160646      -0.4939845
## Lag 1000            -0.3309504      -0.3961937
## Lag 5000            0.1719378       0.1704323
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              -0.12847847     -0.16728073
## Lag 100             -0.11660390     -0.16431801
## Lag 500             -0.08793510     -0.14580635
## Lag 1000            -0.10829057     -0.17627597
## Lag 5000            0.04232726      0.02621324
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              -0.4729141      -0.49584021
## Lag 100             -0.4400289      -0.46178772
## Lag 500             -0.3667218      -0.38855190
## Lag 1000            -0.3075053      -0.32632756
## Lag 5000            0.1176403       0.08762877
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              -0.5208193      -0.4729141
## Lag 100             -0.4814739      -0.4400289
## Lag 500             -0.4160646      -0.3667218
## Lag 1000            -0.3309504      -0.3075053
## Lag 5000            0.1719378       0.1176403
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              -0.84107771     -0.93853010
## Lag 100             -0.80537663     -0.88900373
## Lag 500             -0.70011206     -0.74957513
## Lag 1000            -0.58965340     -0.62094832
## Lag 5000            -0.01326034      0.04348338
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              -0.6200400      -0.49584021
## Lag 100             -0.5760115      -0.46178772
## Lag 500             -0.4939845      -0.38855190
## Lag 1000            -0.3961937      -0.32632756
## Lag 5000            0.1704323       0.08762877
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              -0.93853010     -0.85484247
## Lag 100             -0.88900373     -0.80629890

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## Lag 500           -0.74957513          -0.66699739
## Lag 1000          -0.62094832          -0.53588327
## Lag 5000           0.04348338          0.06798066
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              -0.04533921         -0.001638125        -0.07410803
## Lag 100             -0.04526568         0.002752216        -0.05435921
## Lag 500             -0.02361959         -0.010266769        -0.04066778
## Lag 1000            -0.08661480         -0.055915794        -0.03404716
## Lag 5000             0.02028860         -0.015718402        0.05000555
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0               0.09224924         -0.08507034         0.117048388
## Lag 100              0.09712475         -0.09090158         0.119562799
## Lag 500              0.06704383         -0.06950074         0.093258667
## Lag 1000             0.06946025         -0.09719184         0.091569829
## Lag 5000             -0.05957284         -0.01506617        -0.004965446
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0               0.11599294         -0.06384447
## Lag 100              0.12644815         -0.06061567
## Lag 500              0.12233108         -0.05933391
## Lag 1000             0.11302925         -0.05409993
## Lag 5000              0.02623229         0.01443371
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  0.11602826          0.13276682
## Lag 100                 0.08412251          0.10350782
## Lag 500                 0.06419072          0.10039070
## Lag 1000                0.08359941          0.11689682
## Lag 5000                -0.10317497        -0.07583069
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0                  -0.1801544          -0.2276707
## Lag 100                 -0.1866591          -0.2347923
## Lag 500                 -0.1769783          -0.2196866
## Lag 1000                -0.1617965          -0.1997778
## Lag 5000                -0.0380065          -0.0414611
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0                  0.13276682          0.13741353
## Lag 100                 0.10350782          0.11641231
## Lag 500                 0.10039070          0.12001620
## Lag 1000                0.11689682          0.15480317
## Lag 5000                -0.07583069        -0.03093518
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0                  -0.21678569          -0.2507033
## Lag 100                 -0.22198433          -0.2562972
## Lag 500                 -0.20689452          -0.2323026
## Lag 1000                -0.19261030          -0.2149386
## Lag 5000                -0.02955808          -0.0282441
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0                  -0.1801544          -0.21678569

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## Lag 100           -0.1866591          -0.22198433
## Lag 500           -0.1769783          -0.20689452
## Lag 1000          -0.1617965          -0.19261030
## Lag 5000          -0.0380065          -0.02955808
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              0.926857338         1.00000000
## Lag 100            0.838930228         0.89252017
## Lag 500            0.714319171         0.73419722
## Lag 1000           0.601447081         0.60685285
## Lag 5000           -0.005767447        -0.05722557
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              -0.2276707         -0.2507033
## Lag 100             -0.2347923        -0.2562972
## Lag 500             -0.2196866        -0.2323026
## Lag 1000            -0.1997778        -0.2149386
## Lag 5000            -0.0414611        -0.0282441
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              1.000000000         0.92347712
## Lag 100            0.89252017         0.81767393
## Lag 500            0.73419722         0.65250913
## Lag 1000           0.60685285         0.52632018
## Lag 5000           -0.05722557        -0.07523754
##
## , , traittarsus.1:traitbwt.2.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0              0.1843936          0.13991929
## Lag 100            0.1916063          0.14051238
## Lag 500            0.1997384          0.15350375
## Lag 1000           0.1779242          0.14333349
## Lag 5000           0.1059309          0.02408511
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0              0.2159001          0.18941930
## Lag 100            0.2292097          0.19756460
## Lag 500            0.2470830          0.21294438
## Lag 1000           0.2430281          0.21181509
## Lag 5000           0.1564420          0.09549646
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0              0.13991929         0.11864383
## Lag 100            0.14051238         0.11475981
## Lag 500            0.15350375         0.12191537
## Lag 1000           0.14333349         0.12719171
## Lag 5000           0.02408511         -0.00131242
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0              0.1778597          0.1473359
## Lag 100            0.1832967          0.1558541
## Lag 500            0.2035349          0.1829320

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## Lag 1000           0.2085231           0.1947599
## Lag 5000           0.1096782           0.1076230
##      traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0              0.2159001           0.1778597
## Lag 100             0.2292097           0.1832967
## Lag 500             0.2470830           0.2035349
## Lag 1000            0.2430281           0.2085231
## Lag 5000            0.1564420           0.1096782
##      traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0              0.2148002           0.2380759
## Lag 100             0.2260582           0.2490150
## Lag 500             0.2434250           0.2659527
## Lag 1000            0.2396319           0.2621966
## Lag 5000            0.2311987           0.2321063
##      traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0              0.18941930          0.1473359
## Lag 100             0.19756460          0.1558541
## Lag 500             0.21294438          0.1829320
## Lag 1000            0.21181509          0.1947599
## Lag 5000            0.09549646          0.1076230
##      traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0              0.2380759           0.2116292
## Lag 100             0.2490150           0.2231338
## Lag 500             0.2659527           0.2429222
## Lag 1000            0.2621966           0.2441661
## Lag 5000            0.2321063           0.2236298
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0              -0.032369970         0.05768879          0.06054337
## Lag 100             -0.031641858         0.05864059          0.06580478
## Lag 500             -0.001530924         0.04503326          0.07399288
## Lag 1000            -0.016005519         0.03161375          0.08585903
## Lag 5000             0.010619935         0.01730446          -0.02417570
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0              -0.032356483         0.1703875           0.012376673
## Lag 100             -0.034654911         0.1647608           0.013581085
## Lag 500             -0.052597308         0.1600416           0.005553001
## Lag 1000            -0.039591608         0.1872703           -0.009042770
## Lag 5000             0.004455343         0.1972374           -0.027185781
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0              -0.03846399          0.07575889
## Lag 100             -0.03390447          0.06853459
## Lag 500             -0.02862102          0.06274714
## Lag 1000            -0.01799649          0.06065063
## Lag 5000             -0.05555704          0.02202828
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0                  -0.2645685          -0.18013988
## Lag 100                 -0.2616234          -0.17607364

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## Lag 500           -0.2652457          -0.18034511
## Lag 1000          -0.2671219          -0.17702256
## Lag 5000          -0.2093233          -0.06028968
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.9603400          1.0000000
## Lag 100             0.9432077          0.9762869
## Lag 500             0.9124477          0.9252632
## Lag 1000            0.8838890          0.8792611
## Lag 5000            0.6117319          0.6007465
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0              -0.18013988         -0.15026999
## Lag 100             -0.17607364         -0.14749648
## Lag 500             -0.18034511         -0.15145709
## Lag 1000            -0.17702256         -0.15538707
## Lag 5000            -0.06028968         -0.02621414
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.9303986          0.9527160
## Lag 100             0.9147001          0.9321247
## Lag 500             0.8850543          0.8884276
## Lag 1000            0.8576398          0.8431799
## Lag 5000            0.5876791          0.5807055
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.9603400          0.9303986
## Lag 100             0.9432077          0.9147001
## Lag 500             0.9124477          0.8850543
## Lag 1000            0.8838890          0.8576398
## Lag 5000            0.6117319          0.5876791
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              -0.2006563          -0.2276707
## Lag 100             -0.2124295          -0.2354203
## Lag 500             -0.2379446          -0.2556157
## Lag 1000            -0.2389714          -0.2561306
## Lag 5000            -0.2130152          -0.2137306
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              1.0000000          0.9527160
## Lag 100             0.9762869          0.9321247
## Lag 500             0.9252632          0.8884276
## Lag 1000            0.8792611          0.8431799
## Lag 5000            0.6007465          0.5807055
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              -0.2276707          -0.2128094
## Lag 100             -0.2354203          -0.2168142
## Lag 500             -0.2556157          -0.2324758
## Lag 1000            -0.2561306          -0.2373888
## Lag 5000            -0.2137306          -0.1974646
##
## , , traitbwt.1:traitbwt.2.units

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##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                  0.1045202           0.10675486
## Lag 100                 0.1144301           0.10909010
## Lag 500                 0.1244192           0.13415361
## Lag 1000                0.1101292           0.11544191
## Lag 5000                0.1331335           0.04367713
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                  0.1751169           0.18455878
## Lag 100                 0.1936998           0.19497627
## Lag 500                 0.2199790           0.20989089
## Lag 1000                0.2292872           0.21238609
## Lag 5000                0.1693953           0.06254965
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  0.10675486          0.15466547
## Lag 100                 0.10909010          0.15753844
## Lag 500                 0.13415361          0.17187977
## Lag 1000                0.11544191          0.14889411
## Lag 5000                0.04367713          -0.01393939
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                  0.1618783           0.17860886
## Lag 100                 0.1807130           0.20406604
## Lag 500                 0.2315550           0.24587070
## Lag 1000                0.2433453           0.24000705
## Lag 5000                0.1313066           0.07959284
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                  0.1751169           0.1618783
## Lag 100                 0.1936998           0.1807130
## Lag 500                 0.2199790           0.2315550
## Lag 1000                0.2292872           0.2433453
## Lag 5000                0.1693953           0.1313066
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                  0.2339355           0.2581627
## Lag 100                 0.2476485           0.2736295
## Lag 500                 0.2790724           0.3078386
## Lag 1000                0.2842308           0.3166841
## Lag 5000                0.2101542           0.1923682
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                  0.18455878          0.17860886
## Lag 100                 0.19497627          0.20406604
## Lag 500                 0.20989089          0.24587070
## Lag 1000                0.21238609          0.24000705
## Lag 5000                0.06254965          0.07959284
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                  0.2581627           0.2416106
## Lag 100                 0.2736295           0.2608676
## Lag 500                 0.3078386           0.2963880

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## Lag 1000           0.3166841           0.3068141
## Lag 5000           0.1923682           0.1712560
##      traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            -0.023701193          0.04700881          0.05361674
## Lag 100          -0.021944414          0.05837761          0.06114321
## Lag 500          -0.004019018          0.04124133          0.05472308
## Lag 1000         -0.014592442          0.02897141          0.08299624
## Lag 5000         -0.001455363          0.01713092         -0.03330245
##      traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            -0.03458419           0.1501420          -0.030979909
## Lag 100          -0.02882752           0.1495980          -0.034573917
## Lag 500          -0.05044049           0.1542703          -0.037881298
## Lag 1000         -0.04326520           0.1753867          -0.049179995
## Lag 5000          0.01976572           0.1700994          -0.002621514
##      traittarsus.2.mother traitbwt.2.mother
## Lag 0            -0.05091781           0.08248472
## Lag 100          -0.04612708           0.07643625
## Lag 500          -0.04592155           0.07211789
## Lag 1000         -0.02428388           0.04810929
## Lag 5000          0.05454058           0.02230047
##      traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0              -0.1971901          -0.17008659
## Lag 100             -0.2006748          -0.16466175
## Lag 500             -0.2146434          -0.17764629
## Lag 1000            -0.2152829          -0.15594317
## Lag 5000            -0.2163189          -0.07951097
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0              0.9034487           0.9527160
## Lag 100             0.8867034           0.9318204
## Lag 500             0.8558412           0.8820862
## Lag 1000            0.8281770           0.8332104
## Lag 5000            0.5777198           0.5582022
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            -0.17008659          -0.1902055
## Lag 100           -0.16466175          -0.1881759
## Lag 500           -0.17764629          -0.1918480
## Lag 1000          -0.15594317          -0.1592987
## Lag 5000          -0.07951097          -0.0307547
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0              0.9457989          1.0000000
## Lag 100             0.9171312           0.9591397
## Lag 500             0.8650556           0.8846934
## Lag 1000            0.8286470           0.8252821
## Lag 5000            0.5520607           0.5384118
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0              0.9034487           0.9457989
## Lag 100             0.8867034           0.9171312

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## Lag 500          0.8558412          0.8650556
## Lag 1000         0.8281770          0.8286470
## Lag 5000         0.5777198          0.5520607
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0            -0.2187484          -0.2507033
## Lag 100          -0.2283894          -0.2593514
## Lag 500          -0.2686154          -0.2983604
## Lag 1000         -0.2818961          -0.3082824
## Lag 5000         -0.1951997          -0.1732268
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            0.9527160          1.0000000
## Lag 100          0.9318204          0.9591397
## Lag 500          0.8820862          0.8846934
## Lag 1000         0.8332104          0.8252821
## Lag 5000         0.5582022          0.5384118
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0            -0.2507033          -0.2443613
## Lag 100          -0.2593514          -0.2551879
## Lag 500          -0.2983604          -0.2884325
## Lag 1000         -0.3082824          -0.2942164
## Lag 5000         -0.1732268          -0.1455134
##
## , , traittarsus.2:traitbwt.2.units
##
##      traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0            -0.09066377         -0.12847847
## Lag 100          -0.06791265         -0.11660390
## Lag 500          -0.02813480         -0.08793510
## Lag 1000         -0.02464042         -0.10829057
## Lag 5000         0.09059977          0.04232726
##      traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0            -0.5208193          -0.6200400
## Lag 100          -0.4814739          -0.5760115
## Lag 500          -0.4160646          -0.4939845
## Lag 1000         -0.3309504          -0.3961937
## Lag 5000         0.1719378          0.1704323
##      traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0            -0.12847847         -0.16728073
## Lag 100          -0.11660390         -0.16431801
## Lag 500          -0.08793510         -0.14580635
## Lag 1000         -0.10829057         -0.17627597
## Lag 5000         0.04232726          0.02621324
##      traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0            -0.4729141          -0.49584021
## Lag 100          -0.4400289          -0.46178772
## Lag 500          -0.3667218          -0.38855190
## Lag 1000         -0.3075053          -0.32632756

```

```

## Lag 5000          0.1176403      0.08762877
## traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0            -0.5208193     -0.4729141
## Lag 100          -0.4814739     -0.4400289
## Lag 500          -0.4160646     -0.3667218
## Lag 1000         -0.3309504     -0.3075053
## Lag 5000          0.1719378      0.1176403
## traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0            -0.84107771    -0.93853010
## Lag 100          -0.80537663    -0.88900373
## Lag 500          -0.70011206    -0.74957513
## Lag 1000         -0.58965340    -0.62094832
## Lag 5000          -0.01326034    0.04348338
## traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0            -0.6200400     -0.49584021
## Lag 100          -0.5760115     -0.46178772
## Lag 500          -0.4939845     -0.38855190
## Lag 1000         -0.3961937     -0.32632756
## Lag 5000          0.1704323      0.08762877
## traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0            -0.93853010    -0.85484247
## Lag 100          -0.88900373    -0.80629890
## Lag 500          -0.74957513    -0.66699739
## Lag 1000         -0.62094832    -0.53588327
## Lag 5000          0.04348338    0.06798066
## traittarsus.1.byear traitbwt.1.byear traittarsus.2.byear
## Lag 0            -0.04533921    -0.001638125   -0.07410803
## Lag 100          -0.04526568    0.002752216   -0.05435921
## Lag 500          -0.02361959    -0.010266769   -0.04066778
## Lag 1000         -0.08661480    -0.055915794   -0.03404716
## Lag 5000          0.02028860    -0.015718402   0.05000555
## traitbwt.2.byear traittarsus.1.mother traitbwt.1.mother
## Lag 0            0.09224924    -0.08507034    0.117048388
## Lag 100          0.09712475    -0.09090158    0.119562799
## Lag 500          0.06704383    -0.06950074    0.093258667
## Lag 1000         0.06946025    -0.09719184    0.091569829
## Lag 5000          -0.05957284   -0.01506617    -0.004965446
## traittarsus.2.mother traitbwt.2.mother
## Lag 0            0.11599294    -0.06384447
## Lag 100          0.12644815    -0.06061567
## Lag 500          0.12233108    -0.05933391
## Lag 1000         0.11302925    -0.05409993
## Lag 5000          0.02623229    0.01443371
## traittarsus.1:traittarsus.1.units traitbwt.1:traittarsus.1.units
## Lag 0            0.11602826    0.13276682
## Lag 100          0.08412251    0.10350782
## Lag 500          0.06419072    0.10039070

```

```

## Lag 1000          0.08359941          0.11689682
## Lag 5000         -0.10317497         -0.07583069
##      traittarsus.2:traittarsus.1.units traitbwt.2:traittarsus.1.units
## Lag 0            -0.1801544           -0.2276707
## Lag 100          -0.1866591           -0.2347923
## Lag 500          -0.1769783           -0.2196866
## Lag 1000         -0.1617965           -0.1997778
## Lag 5000         -0.0380065           -0.0414611
##      traittarsus.1:traitbwt.1.units traitbwt.1:traitbwt.1.units
## Lag 0            0.13276682          0.13741353
## Lag 100          0.10350782          0.11641231
## Lag 500          0.10039070          0.12001620
## Lag 1000         0.11689682          0.15480317
## Lag 5000         -0.07583069         -0.03093518
##      traittarsus.2:traitbwt.1.units traitbwt.2:traitbwt.1.units
## Lag 0            -0.21678569          -0.2507033
## Lag 100          -0.22198433          -0.2562972
## Lag 500          -0.20689452          -0.2323026
## Lag 1000         -0.19261030          -0.2149386
## Lag 5000         -0.02955808          -0.0282441
##      traittarsus.1:traittarsus.2.units traitbwt.1:traittarsus.2.units
## Lag 0            -0.1801544           -0.21678569
## Lag 100          -0.1866591           -0.22198433
## Lag 500          -0.1769783           -0.20689452
## Lag 1000         -0.1617965           -0.19261030
## Lag 5000         -0.0380065           -0.02955808
##      traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0            0.926857338          1.00000000
## Lag 100          0.838930228          0.89252017
## Lag 500          0.714319171          0.73419722
## Lag 1000         0.601447081          0.60685285
## Lag 5000         -0.005767447          -0.05722557
##      traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0            -0.2276707           -0.2507033
## Lag 100          -0.2347923           -0.2562972
## Lag 500          -0.2196866           -0.2323026
## Lag 1000         -0.1997778           -0.2149386
## Lag 5000         -0.0414611           -0.0282441
##      traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0            1.00000000          0.92347712
## Lag 100          0.89252017          0.81767393
## Lag 500          0.73419722          0.65250913
## Lag 1000         0.60685285          0.52632018
## Lag 5000         -0.05722557          -0.07523754
##
## , , traitbwt.2:traitbwt.2.units
##

```

```

##          traittarsus.1:traittarsus.1.animal traitbwt.1:traittarsus.1.animal
## Lag 0                  -0.043059890      -0.10947233
## Lag 100                 -0.021006970      -0.08947374
## Lag 500                 -0.001331729      -0.06046211
## Lag 1000                -0.008677951      -0.10284705
## Lag 5000                0.111890868      0.03101996
##          traittarsus.2:traittarsus.1.animal traitbwt.2:traittarsus.1.animal
## Lag 0                  -0.3939711      -0.5649761
## Lag 100                 -0.3674489      -0.5248935
## Lag 500                 -0.3564907      -0.4716725
## Lag 1000                -0.3026255      -0.3959870
## Lag 5000                0.1392553      0.1506927
##          traittarsus.1:traitbwt.1.animal traitbwt.1:traitbwt.1.animal
## Lag 0                  -0.10947233     -0.187811354
## Lag 100                 -0.08947374     -0.167609258
## Lag 500                 -0.06046211     -0.153207032
## Lag 1000                -0.10284705     -0.191376283
## Lag 5000                0.03101996     -0.009266993
##          traittarsus.2:traitbwt.1.animal traitbwt.2:traitbwt.1.animal
## Lag 0                  -0.43960419     -0.56988609
## Lag 100                 -0.41008412     -0.52185804
## Lag 500                 -0.34705550     -0.43385778
## Lag 1000                -0.31152313     -0.36023626
## Lag 5000                0.07052962      0.07222864
##          traittarsus.1:traittarsus.2.animal traitbwt.1:traittarsus.2.animal
## Lag 0                  -0.3939711      -0.43960419
## Lag 100                 -0.3674489      -0.41008412
## Lag 500                 -0.3564907      -0.34705550
## Lag 1000                -0.3026255      -0.31152313
## Lag 5000                0.1392553      0.07052962
##          traittarsus.2:traittarsus.2.animal traitbwt.2:traittarsus.2.animal
## Lag 0                  -0.67910852     -0.85888343
## Lag 100                 -0.65466001     -0.81238323
## Lag 500                 -0.60647885     -0.70052835
## Lag 1000                -0.54835587     -0.61884830
## Lag 5000                -0.02772098     0.03373215
##          traittarsus.1:traitbwt.2.animal traitbwt.1:traitbwt.2.animal
## Lag 0                  -0.5649761      -0.56988609
## Lag 100                 -0.5248935      -0.52185804
## Lag 500                 -0.4716725      -0.43385778
## Lag 1000                -0.3959870      -0.36023626
## Lag 5000                0.1506927      0.07222864
##          traittarsus.2:traitbwt.2.animal traitbwt.2:traitbwt.2.animal
## Lag 0                  -0.85888343     -0.90703776
## Lag 100                 -0.81238323     -0.83971454
## Lag 500                 -0.70052835     -0.68420955
## Lag 1000                -0.61884830     -0.57020879

```

```

## Lag 5000          0.03373215          0.07361913
## traittarsus.1:byear traitbwt.1:byear traittarsus.2:byear
## Lag 0            -0.044815839         -0.007601342         -0.09539294
## Lag 100           -0.046632422         -0.009618828         -0.07733270
## Lag 500           -0.039749429         -0.026455434         -0.05466463
## Lag 1000          -0.096232932         -0.063248471         -0.04859295
## Lag 5000          -0.007307801         -0.011838332         0.07303267
## traitbwt.2:byear traittarsus.1:mother traitbwt.1:mother
## Lag 0             0.07926090          -0.11755479          0.1533639724
## Lag 100            0.09525987          -0.11193096          0.1505628703
## Lag 500            0.05998165          -0.06092327          0.1143539752
## Lag 1000           0.06572238          -0.09798980          0.1117926423
## Lag 5000           -0.05632018         -0.00703712          0.0007242137
## traittarsus.2:mother traitbwt.2:mother
## Lag 0              0.09798770          -0.07912857
## Lag 100             0.09530597          -0.05178227
## Lag 500             0.10285009          -0.03607903
## Lag 1000            0.08664286          -0.05475142
## Lag 5000            0.03594017          0.03608548
## traittarsus.1:traittarsus.1:units traitbwt.1:traittarsus.1:units
## Lag 0                0.10062986          0.11344391
## Lag 100               0.07039582          0.08507484
## Lag 500               0.04998573          0.08084299
## Lag 1000              0.07979248          0.11040825
## Lag 5000              -0.11441105         -0.06475361
## traittarsus.2:traittarsus.1:units traitbwt.2:traittarsus.1:units
## Lag 0                -0.13454205         -0.21280936
## Lag 100               -0.14136050         -0.21928247
## Lag 500               -0.12646869         -0.19382371
## Lag 1000              -0.12162967         -0.17751558
## Lag 5000              -0.02590274         -0.02880894
## traittarsus.1:traitbwt.1:units traitbwt.1:traitbwt.1:units
## Lag 0                0.11344391          0.131450832
## Lag 100               0.08507484          0.105602349
## Lag 500               0.08084299          0.125553514
## Lag 1000              0.11040825          0.165398766
## Lag 5000              -0.06475361          0.003312387
## traittarsus.2:traitbwt.1:units traitbwt.2:traitbwt.1:units
## Lag 0                -0.18001725         -0.24436125
## Lag 100               -0.18862947         -0.25750569
## Lag 500               -0.17962012         -0.23342183
## Lag 1000              -0.16368716         -0.20650410
## Lag 5000              -0.02638416         -0.02742457
## traittarsus.1:traittarsus.2:units traitbwt.1:traittarsus.2:units
## Lag 0                -0.13454205         -0.18001725
## Lag 100               -0.14136050         -0.18862947
## Lag 500               -0.12646869         -0.17962012

```

```

## Lag 1000           -0.12162967      -0.16368716
## Lag 5000           -0.02590274      -0.02638416
##          traittarsus.2:traittarsus.2.units traitbwt.2:traittarsus.2.units
## Lag 0              0.75421487       0.92347712
## Lag 100             0.68911416       0.81697064
## Lag 500             0.62438762       0.68932883
## Lag 1000            0.56071503       0.60055861
## Lag 5000            0.01204694      -0.04603239
##          traittarsus.1:traitbwt.2.units traitbwt.1:traitbwt.2.units
## Lag 0              -0.21280936      -0.24436125
## Lag 100             -0.21928247      -0.25750569
## Lag 500             -0.19382371      -0.23342183
## Lag 1000            -0.17751558      -0.20650410
## Lag 5000            -0.02880894      -0.02742457
##          traittarsus.2:traitbwt.2.units traitbwt.2:traitbwt.2.units
## Lag 0              0.92347712       1.00000000
## Lag 100            0.81697064       0.84520133
## Lag 500            0.68932883       0.66116525
## Lag 1000            0.60055861       0.55204450
## Lag 5000            -0.04603239      -0.08094849

```

3.5 brms

First load brms:

```

library(brms)
Amat <- as.matrix(nadiv::makeA(gryphonped))

```

3.5.1 Fitting the model

Fitting a multivariate model in brms involves several new consideration above those for fitting univariate models. First, we need to create two models/objects with the function `bf` fitting the desired univariate model structure for each response variable (here `bwt` and `tarsus`). It is the equivalent of writing `mvbf(bwt, tarsus)`, but the advantage to create two distinct model is to specific different model structure (fixed or random effect) for each response variable.

Then, the two objects/models are added into a third model to quantify all the estimates in addition to their covariance. Contrary to MCMCglmm or asreml-R, brms directly estimate the covariance and the correlation in its outputs. Our most basic model can be specified as:

```

bf_bwt <- bf(bwt ~ 1 + (1 | a | gr(animal, cov = Amat)))
bf_tarsus <- bf(tarsus ~ 1 + (1 | a | gr(animal, cov = Amat)))
brms_m2.1 <- brm(
  bf_bwt + bf_tarsus + set_rescor(TRUE),
  data = gryphon,

```

```

data2 = list(Amat = Amat),
chains = 2, cores = 2, iter = 1000
)
save(brms_m2.1, file = "data/brms_m2_1.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load("data/brms_m2_1.rda")
summary(brms_m2.1)

```

```

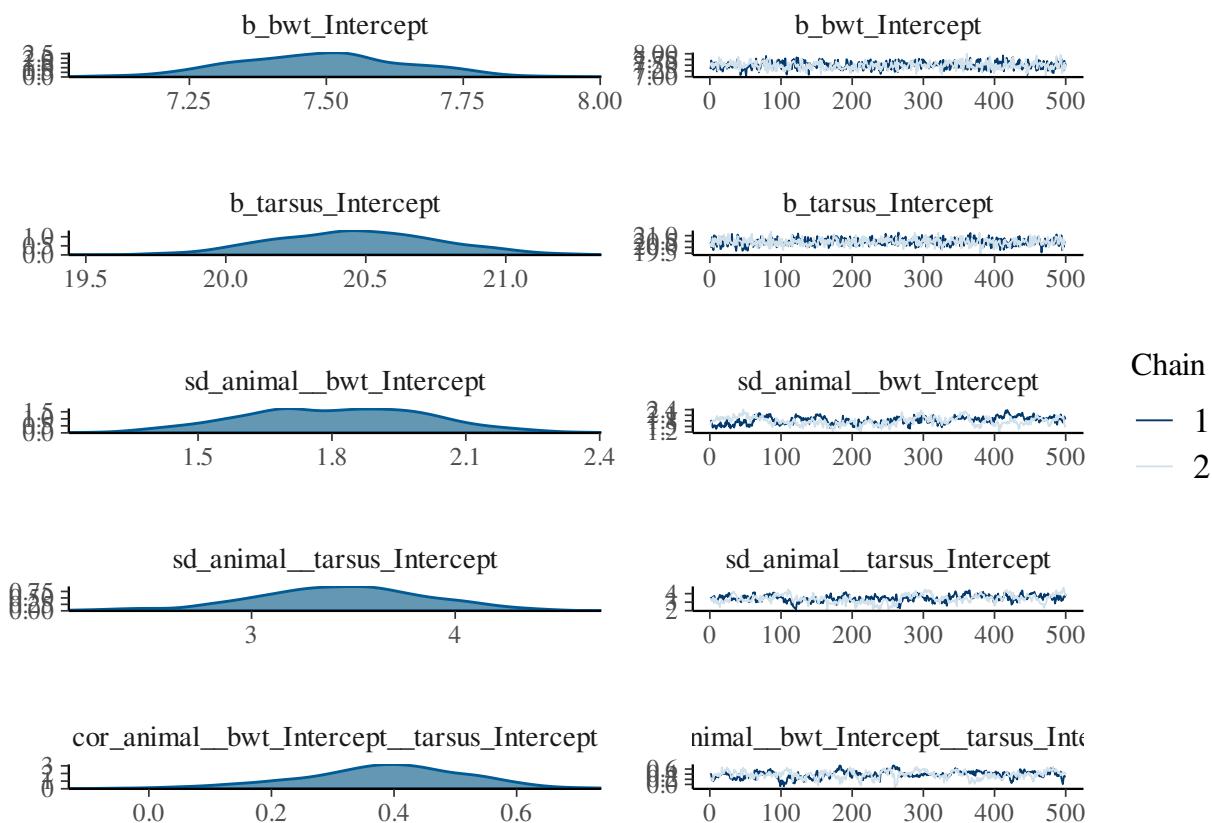
## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

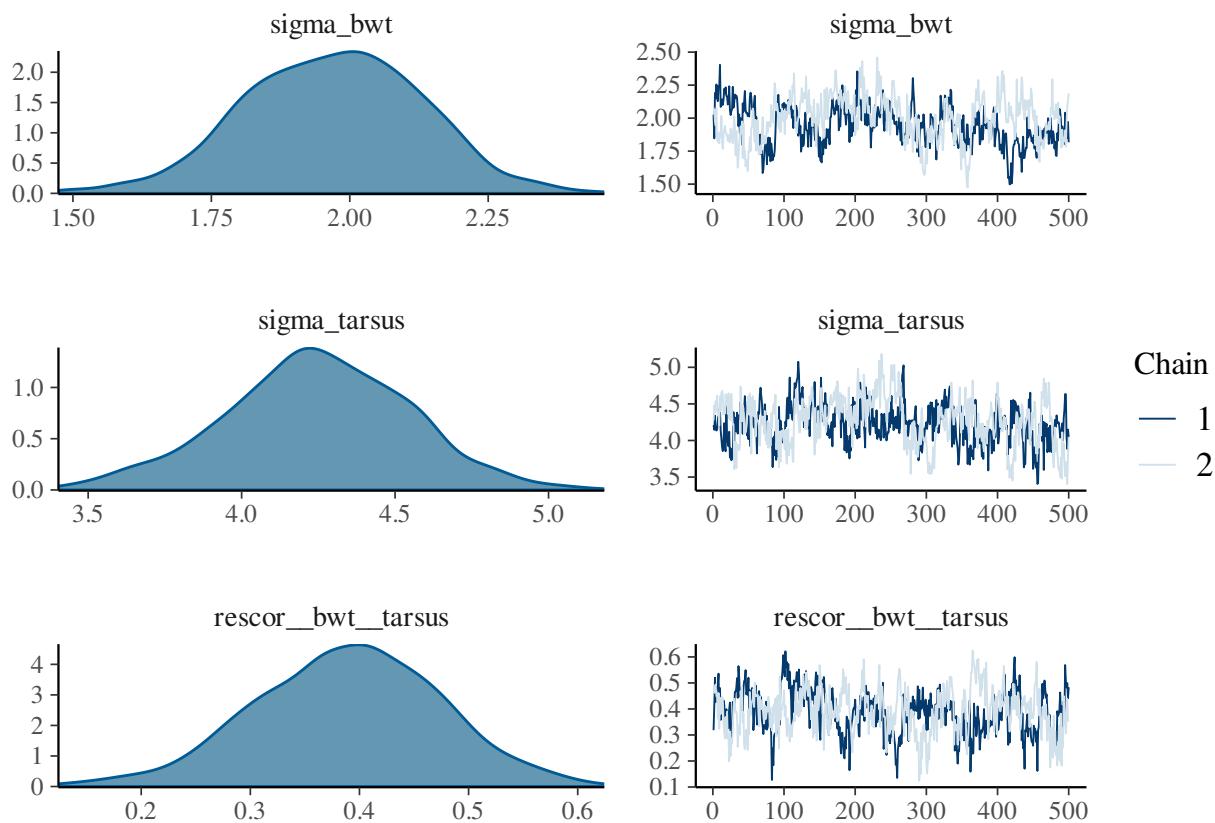
## Family: MV(gaussian, gaussian)
## Links: mu = identity; sigma = identity
##          mu = identity; sigma = identity
## Formula: bwt ~ 1 + (1 | p | gr(animal, cov = Amat))
##           tarsus ~ 1 + (1 | p | gr(animal, cov = Amat))
## Data: gryphon (Number of observations: 683)
## Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##        total post-warmup draws = 1000
##
## Group-Level Effects:
## ~animal (Number of levels: 683)
##                               Estimate Est.Error l-95% CI u-95% CI Rhat
## sd(bwt_Intercept)            1.81     0.21    1.41    2.20 1.06
## sd(tarsus_Intercept)         3.44     0.43    2.49    4.25 1.05
## cor(bwt_Intercept,tarsus_Intercept) 0.38     0.14    0.08    0.62 1.02
##                               Bulk_ESS Tail_ESS
## sd(bwt_Intercept)            31       192
## sd(tarsus_Intercept)         61       173
## cor(bwt_Intercept,tarsus_Intercept) 101      232
##
## Population-Level Effects:
##                               Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## bwt_Intercept      7.49     0.16    7.20    7.79 1.00      608     839
## tarsus_Intercept 20.47     0.30   19.92   21.03 1.00      868     803
##
## Family Specific Parameters:
##                               Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma_bwt         1.97     0.16    1.66    2.28 1.06      27      172
## sigma_tarsus      4.24     0.30    3.63    4.82 1.04      72      162
##
## Residual Correlations:
##                               Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS

```

```
## rescor(bwt,tarsus)      0.39      0.09      0.21      0.55 1.02      95     179
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
```

```
plot(brms_m2.1, ask = FALSE)
```





```
VarCorr(brms_m2.1)
```

```
## $animal
## $animal$sd
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 1.808171 0.2050233 1.412824 2.204805
## tarsus_Intercept 3.438368 0.4283612 2.491218 4.245264
##
## $animal$cor
## , , bwt_Intercept
##
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 1.0000000 0.0000000 1.00000000 1.0000000
## tarsus_Intercept 0.3814062 0.1380014 0.07581464 0.6209038
##
## , , tarsus_Intercept
##
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 0.3814062 0.1380014 0.07581464 0.6209038
## tarsus_Intercept 1.0000000 0.0000000 1.00000000 1.0000000
##
## $animal$cov
```

```
## , , bwt_Intercept
##
##           Estimate Est.Error      Q2.5     Q97.5
## bwt_Intercept    3.311473 0.7430185 1.9960721 4.861167
## tarsus_Intercept 2.440166 1.0901689 0.3870783 4.668720
##
## , , tarsus_Intercept
##
##           Estimate Est.Error      Q2.5     Q97.5
## bwt_Intercept    2.440166 1.090169 0.3870783 4.66872
## tarsus_Intercept 12.005688 2.918741 6.2061701 18.02226
##
##
##
## $residual__
## $residual__$sd
##           Estimate Est.Error      Q2.5     Q97.5
## bwt      1.970532 0.1597581 1.658782 2.276074
## tarsus   4.244704 0.2984518 3.632824 4.820109
##
## $residual__$cor
## , , bwt
##
##           Estimate Est.Error      Q2.5     Q97.5
## bwt      1.0000000 0.0000000 1.0000000 1.0000000
## tarsus   0.3888754 0.08510488 0.2127907 0.5526631
##
## , , tarsus
##
##           Estimate Est.Error      Q2.5     Q97.5
## bwt      0.3888754 0.08510488 0.2127907 0.5526631
## tarsus   1.0000000 0.0000000 1.0000000 1.0000000
##
##
## $residual__$cov
## , , bwt
##
##           Estimate Est.Error      Q2.5     Q97.5
## bwt      3.908493 0.6282892 2.751557 5.180511
## tarsus   3.289995 0.9305960 1.572647 5.147133
##
## , , tarsus
##
##           Estimate Est.Error      Q2.5     Q97.5
## bwt      3.289995 0.930596 1.572647 5.147133
## tarsus   18.106495 2.530138 13.197409 23.233452
```

It is also possible to calculate the heritability for each trait using the function ‘as.mcmc’

```
v_animal <- (VarCorr(brms_m2.1, summary = FALSE)$animal$sd)^2
v_r <- (VarCorr(brms_m2.1, summary = FALSE)$residual$sd)^2

h.bwt.2 <- as.mcmc(v_animal[, 1] / (v_animal[, 1] + v_r[, 1]))
h.tarsus.2 <- as.mcmc(v_animal[, 1] / (v_animal[, 1] + v_r[, 1]))

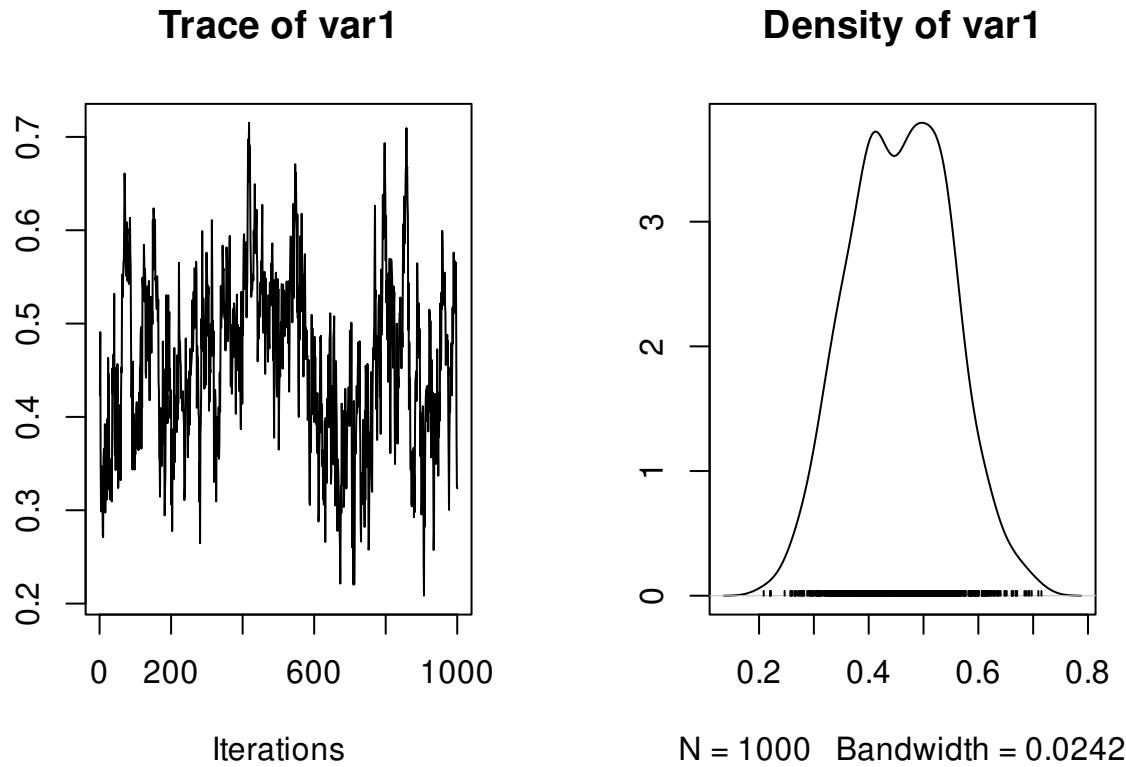
summary(h.bwt.2)
```

```
## 
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD   Naive SE Time-series SE
## 0.457051 0.090878 0.002874 0.011675
##
## 2. Quantiles for each variable:
##
##    2.5%    25%    50%    75%   97.5%
## 0.2878 0.3926 0.4596 0.5254 0.6297
```

```
summary(h.tarsus.2)
```

```
## 
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD   Naive SE Time-series SE
## 0.457051 0.090878 0.002874 0.011675
##
## 2. Quantiles for each variable:
##
##    2.5%    25%    50%    75%   97.5%
## 0.2878 0.3926 0.4596 0.5254 0.6297
```

```
plot(h.bwt.2)
plot(h.tarsus.2)
```



It is also possible to extract the correlation. Just to remember it is an example, the correlation distribution is skewed to 1 due to a weak prior and model parameters. Note, since

```
cor_g <- as.mcmc((VarCorr(brms_m2.1, summary = FALSE)$animal$cor[, 1, 2]))
cor_res <- as.mcmc((VarCorr(brms_m2.1, summary = FALSE)$residual$cor[, 1, 2]))
```

```
summary(cor_g)
```

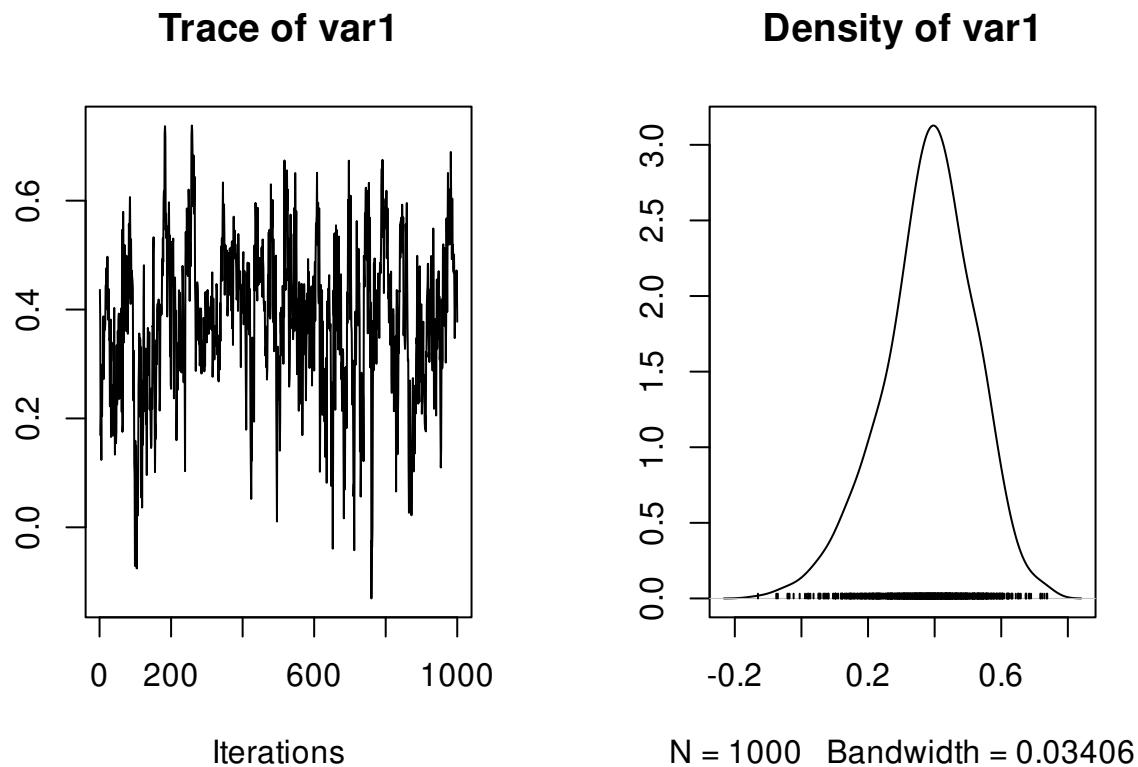
```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean          SD      Naive SE Time-series SE
## 0.381406   0.138001   0.004364    0.014946
```

```
##  
## 2. Quantiles for each variable:  
##  
##      2.5%     25%     50%     75%   97.5%  
## 0.07581 0.30354 0.39041 0.47497 0.62090
```

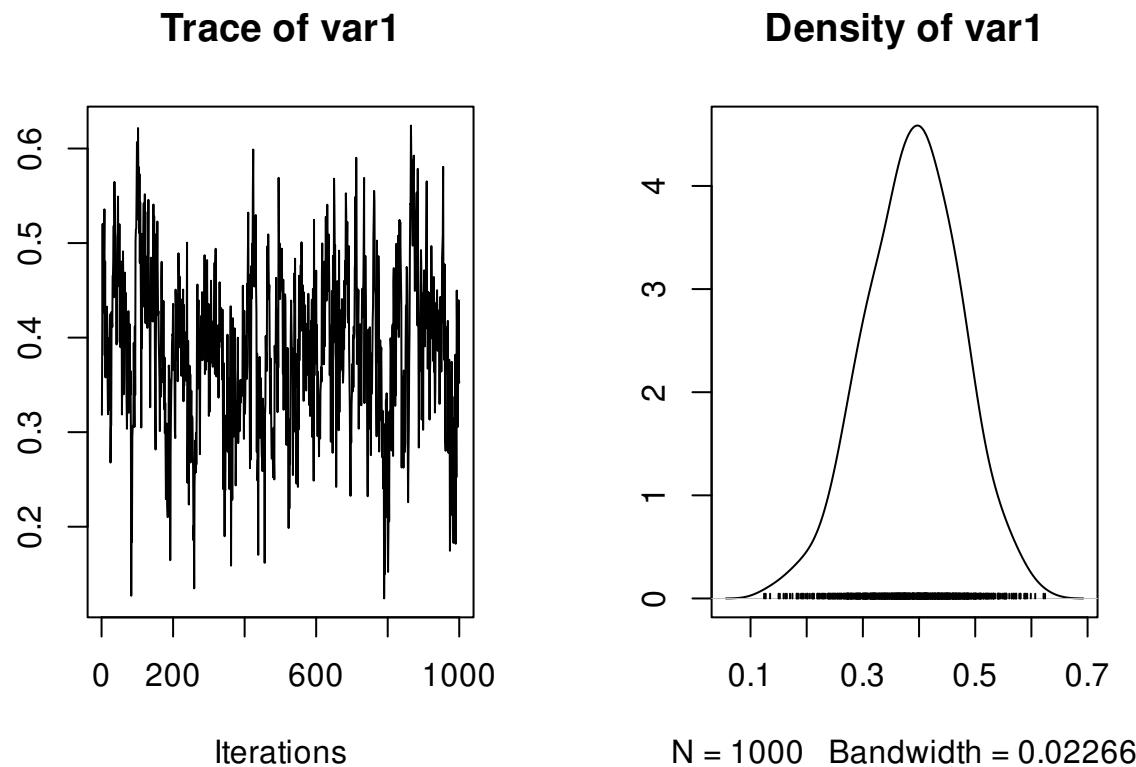
```
summary(cor_res)
```

```
##  
## Iterations = 1:1000  
## Thinning interval = 1  
## Number of chains = 1  
## Sample size per chain = 1000  
##  
## 1. Empirical mean and standard deviation for each variable,  
##    plus standard error of the mean:  
##  
##           Mean        SD    Naive SE Time-series SE  
## 0.388875 0.085105 0.002691 0.009215  
##  
## 2. Quantiles for each variable:  
##  
##      2.5%     25%     50%     75%   97.5%  
## 0.2128 0.3319 0.3913 0.4484 0.5527
```

```
plot(cor_g)
```



```
plot(cor_res)
```



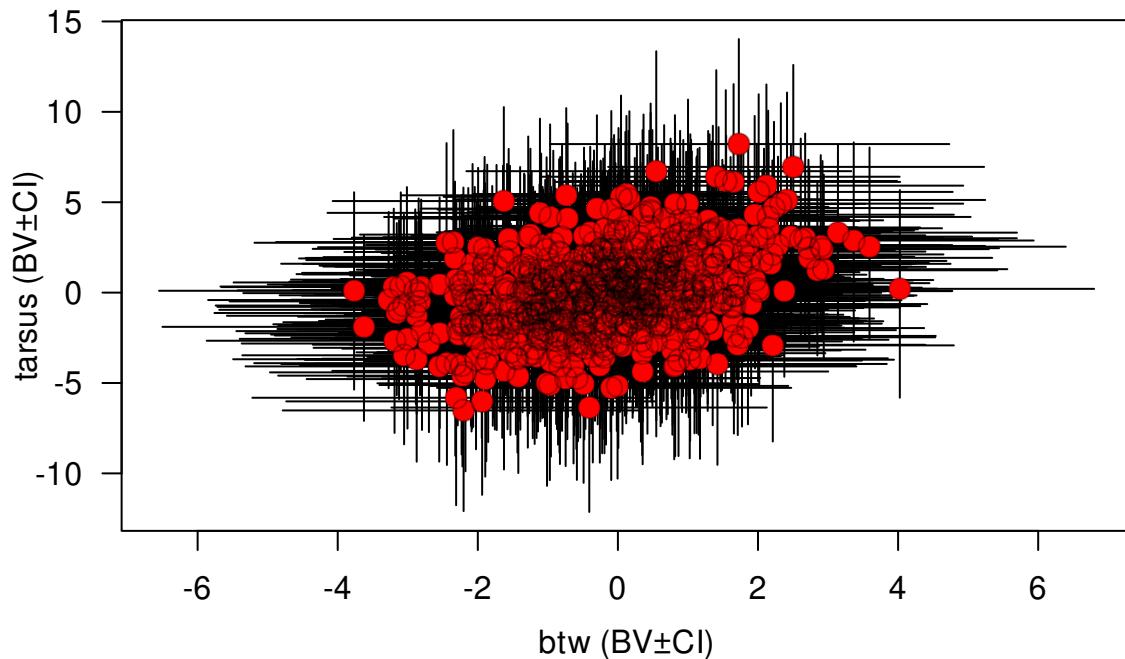
Here we can plot the genetic correlation by extraction the breeding values or BLUP.

```
bls_m2.1 <- ranef(brms_m2.1)$animal
bl_m2.1 <- as.data.frame(abind::abind(lapply(1:dim(bls_m2.1)[[3]], function(x) bls_m2.1[, colnames(bl_m2.1) <- paste0(rep(dimnames(bls_m2.1)[[3]], each = 3), c("", "_lo", "_up"))
bl_m2.1$id <- rownames(bl_m2.1)
```

Here, some simple code to plot the genetic correlation.

```
plot(tarsus_Intercept ~ bwt_Intercept, bl_m2.1,
  xlab = "", ylab = "",
  xlim = c(min(bl_m2.1$bwt_Intercept_lo), max(bl_m2.1$bwt_Intercept_up)),
  ylim = c(min(bl_m2.1$tarsus_Intercept_lo), max(bl_m2.1$tarsus_Intercept_up)),
  las = 1.2, type = "n")
)
with(
  bl_m2.1,
  segments(
    x0 = bwt_Intercept, y0 = tarsus_Intercept_lo,
    x1 = bwt_Intercept, y1 = tarsus_Intercept_up,
    col = "black"
)
```

```
)
with(bl_m2.1, segments(
  x0 = bwt_Intercept_lo, y0 = tarsus_Intercept,
  x1 = bwt_Intercept_up, y1 = tarsus_Intercept,
  col = "black"
))
points(tarsus_Intercept ~ bwt_Intercept, bl_m2.1, pch = 16, col = "red", cex = 1.5)
points(tarsus_Intercept ~ bwt_Intercept, bl_m2.1, pch = 1, col = rgb(0, 0, 0, 0.3), cex =
mtext("btw (BV±CI)", side = 1, line = 2.4)
mtext("tarsus (BV±CI)", side = 2, line = 2, las = 3)
```



3.5.2 Adding fixed and random effects

Fixed and random effects can be added just as for the univariate case. Given that our full model of bwt from tutorial 1 had sex as a fixed effect as well as random effects of byear and mother, we could specify a bivariate formulation of this using the following code (including a line to save the output):

```
bf_bwt_2 <- bf(bwt ~ 1 + sex + (1 | a | gr(animal, cov = Amat)) + (1 | b | byear) + (1 | c | gmat))
bf_tarsus_2 <- bf(tarsus ~ 1 + sex + (1 | a | gr(animal, cov = Amat)) + (1 | b | byear) + (1 | c | gmat))
```

```

brms_m2.2 <- brm(
  bf_bwt_2 + bf_tarsus_2 + set_rescor(TRUE),
  data = gryphon,
  data2 = list(Amat = Amat),
  chains = 2, cores = 2, iter = 1000
)

save(brms_m2.2, file = "data/brms_m2.2.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load("data/brms_m2.2.rda")
summary(brms_m2.2)

```

```

## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## Warning: There were 4 divergent transitions after warmup. Increasing adapt_delta
## above 0.8 may help. See http://mc-stan.org/misc/warnings.html#divergent-
## transitions-after-warmup

## Family: MV(gaussian, gaussian)
##   Links: mu = identity; sigma = identity
##          mu = identity; sigma = identity
## Formula: bwt ~ 1 + sex + (1 | a | gr(animal, cov = Amat)) + (1 | b | byear) + (1 | c |
##          tarsus ~ 1 + sex + (1 | a | gr(animal, cov = Amat)) + (1 | b | byear) + (1 | c |
##          Data: gryphon (Number of observations: 683)
##   Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##          total post-warmup draws = 1000
##
## Group-Level Effects:
##   ~animal (Number of levels: 683)
##                               Estimate Est.Error l-95% CI u-95% CI Rhat
##   ## sd(bwt_Intercept)           1.31     0.21     0.86     1.69 1.06
##   ## sd(tarsus_Intercept)       2.88     0.47     1.88     3.70 1.01
##   ## cor(bwt_Intercept,tarsus_Intercept) 0.60     0.17     0.18     0.89 1.09
##   ##                               Bulk_ESS Tail_ESS
##   ## sd(bwt_Intercept)           54        58
##   ## sd(tarsus_Intercept)       52        162
##   ## cor(bwt_Intercept,tarsus_Intercept) 25        30
##   ##
##   ~byear (Number of levels: 34)
##                               Estimate Est.Error l-95% CI u-95% CI Rhat
##   ## sd(bwt_Intercept)           0.99     0.17     0.71     1.39 1.00
##   ## sd(tarsus_Intercept)       2.02     0.34     1.45     2.78 1.00

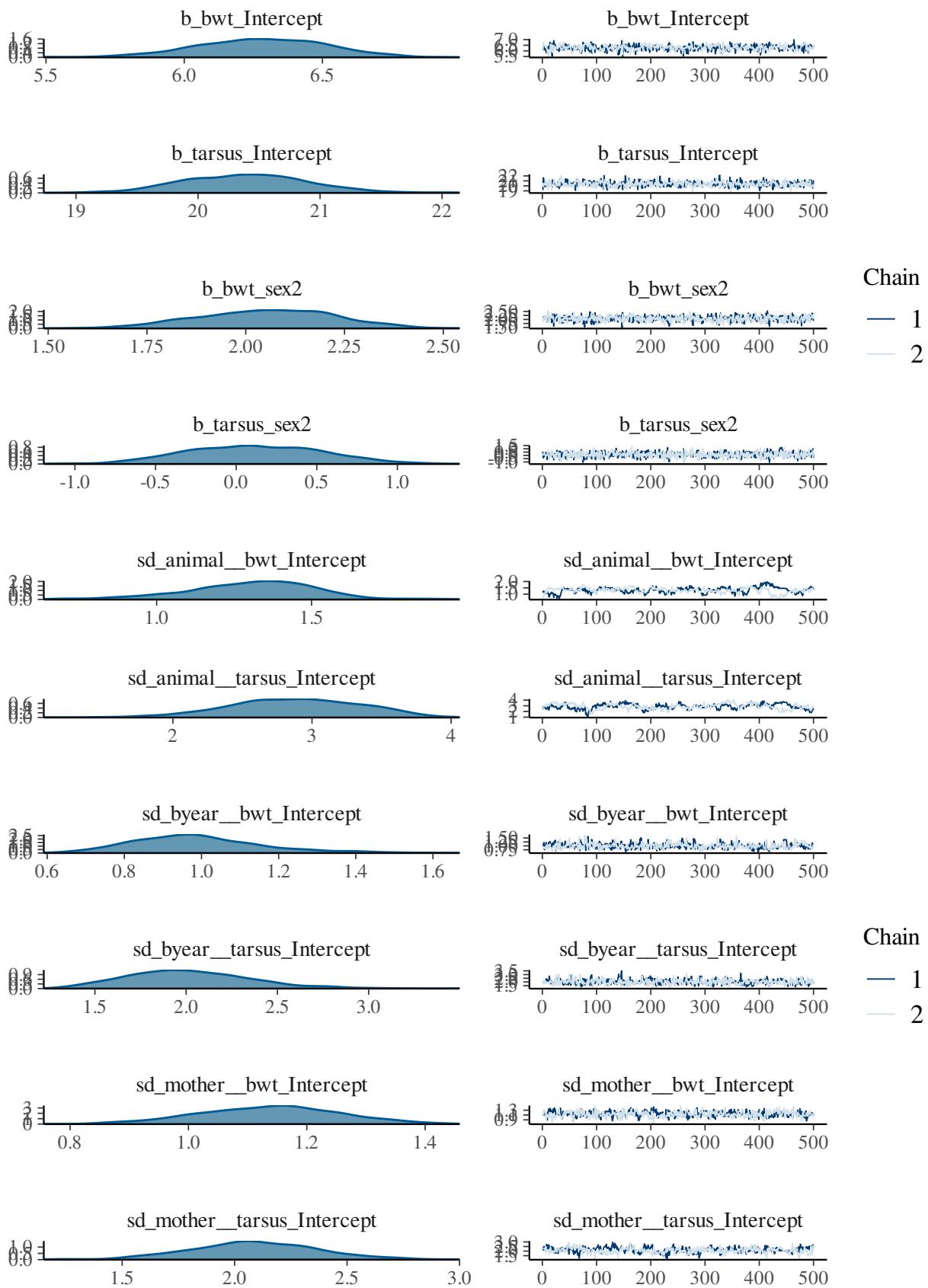
```

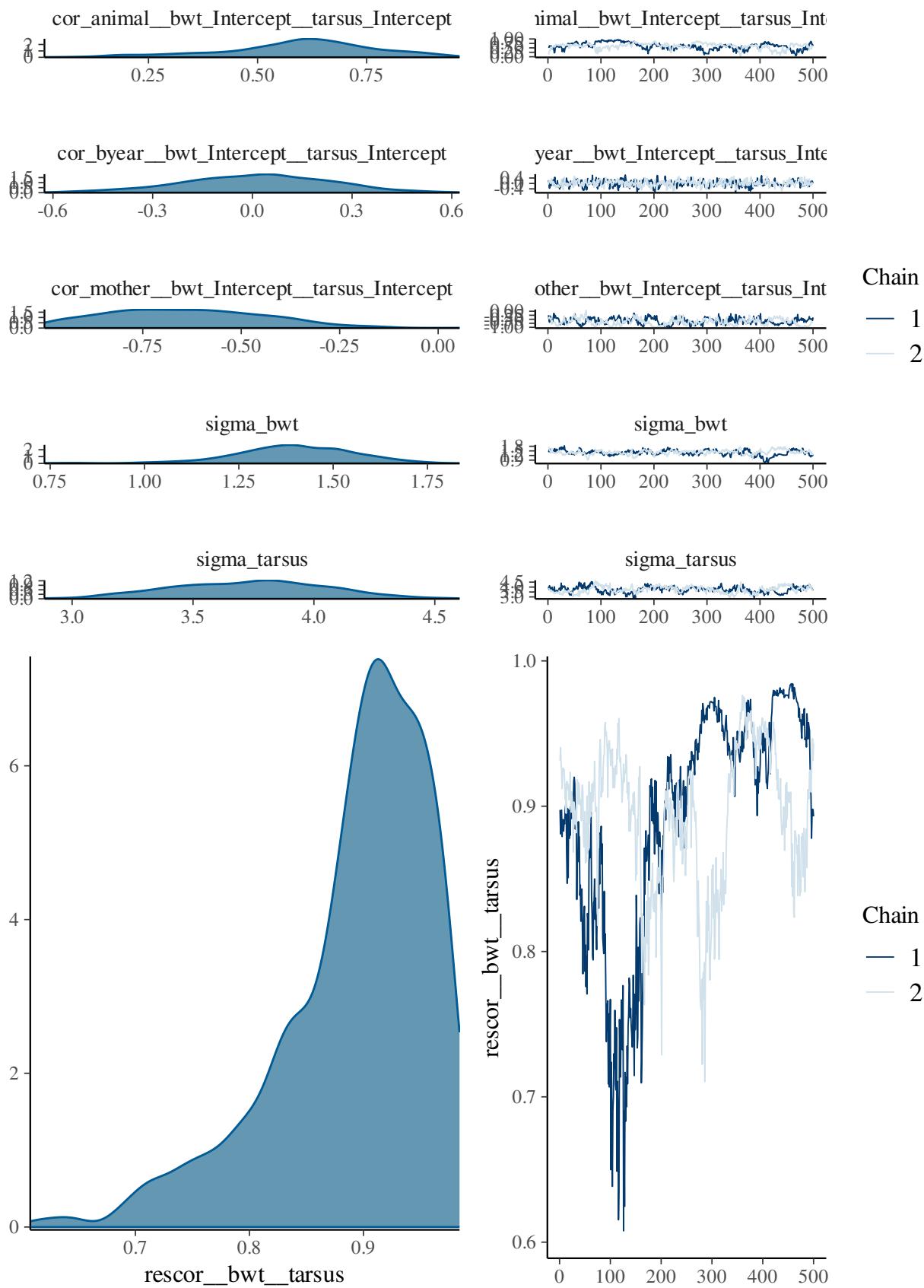
```

## cor(bwt_Intercept,tarsus_Intercept)      0.01    0.22   -0.44    0.44 1.01
##                                         Bulk_ESS Tail_ESS
## sd(bwt_Intercept)                      424     509
## sd(tarsus_Intercept)                   525     699
## cor(bwt_Intercept,tarsus_Intercept)    478     492
##
## ~mother (Number of levels: 352)
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat
## sd(bwt_Intercept)                      1.14     0.12    0.90    1.36 1.01
## sd(tarsus_Intercept)                   2.09     0.29    1.54    2.67 1.01
## cor(bwt_Intercept,tarsus_Intercept)    -0.64     0.20   -0.97   -0.24 1.02
##                                         Bulk_ESS Tail_ESS
## sd(bwt_Intercept)                      370     764
## sd(tarsus_Intercept)                   134     420
## cor(bwt_Intercept,tarsus_Intercept)    84      147
##
## Population-Level Effects:
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## bwt_Intercept                  6.28     0.24    5.80    6.73 1.00    453     703
## tarsus_Intercept                20.39    0.52   19.45   21.39 1.00    755     760
## bwt_sex2                       2.05     0.17    1.71    2.37 1.00   1097     715
## tarsus_sex2                     0.11     0.42   -0.67    0.90 1.00    780     578
##
## Family Specific Parameters:
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma_bwt                      1.40     0.16    1.05    1.68 1.04      59      60
## sigma_tarsus                    3.73     0.32    3.14    4.32 1.00      55     176
##
## Residual Correlations:
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## rescor(bwt,tarsus)               0.89     0.07    0.72    0.98 1.50       4      24
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

```

```
plot(brms_m2.2, ask = FALSE)
```





```
VarCorr(brms_m2.2)
```

```
## $animal
## $animal$sd
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 1.306921 0.2109780 0.8555702 1.687254
## tarsus_Intercept 2.876731 0.4718524 1.8827128 3.699873
##
## $animal$cor
## , , bwt_Intercept
##
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept 0.5961514 0.1727943 0.1849725 0.8929607
##
## , , tarsus_Intercept
##
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 0.5961514 0.1727943 0.1849725 0.8929607
## tarsus_Intercept 1.0000000 0.0000000 1.0000000 1.0000000
##
## $animal$cov
## , , bwt_Intercept
##
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 1.752510 0.5433004 0.7320013 2.846827
## tarsus_Intercept 2.357852 1.0462644 0.3960956 4.526731
##
## , , tarsus_Intercept
##
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 2.357852 1.046264 0.3960956 4.526731
## tarsus_Intercept 8.498003 2.672465 3.5446115 13.689062
##
## $byear
## $byear$sd
##             Estimate Est.Error    Q2.5    Q97.5
## bwt_Intercept 0.9904858 0.1696752 0.7102391 1.390260
## tarsus_Intercept 2.0166804 0.3366954 1.4463862 2.777005
##
## $byear$cor
## , , bwt_Intercept
```

```
##  
##  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept 1.00000000 0.0000000 1.0000000 1.0000000  
## tarsus_Intercept 0.01351405 0.2206186 -0.4367809 0.4412319  
##  
## , , tarsus_Intercept  
##  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept 0.01351405 0.2206186 -0.4367809 0.4412319  
## tarsus_Intercept 1.00000000 0.0000000 1.0000000 1.0000000  
##  
##  
##  
## $byear$cov  
## , , bwt_Intercept  
##  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept 1.00982294 0.3583697 0.5044397 1.932823  
## tarsus_Intercept 0.06412895 0.4880116 -0.8559486 1.092317  
##  
## , , tarsus_Intercept  
##  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept 0.06412895 0.4880116 -0.8559486 1.092317  
## tarsus_Intercept 4.18025049 1.4249595 2.0920330 7.711755  
##  
##  
##  
##  
## $mother  
## $mother$sd  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept 1.137249 0.1175495 0.8968349 1.358078  
## tarsus_Intercept 2.088602 0.2865291 1.5425683 2.673791  
##  
##  
## $mother$cor  
## , , bwt_Intercept  
##  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept 1.0000000 0.0000000 1.0000000 1.0000000  
## tarsus_Intercept -0.6413329 0.1968102 -0.9745925 -0.2367393  
##  
## , , tarsus_Intercept  
##  
##  
## bwt_Intercept Estimate Est.Error Q2.5 Q97.5  
## bwt_Intercept -0.6413329 0.1968102 -0.9745925 -0.2367393  
## tarsus_Intercept 1.0000000 0.0000000 1.0000000 1.0000000  
##  
##
```

```
## $mother$cov
## , , bwt_Intercept
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept    1.307139 0.2669748  0.8043132  1.8443762
## tarsus_Intercept -1.467511 0.3650113 -2.1491979 -0.6718737
##
## , , tarsus_Intercept
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept    -1.467511 0.3650113 -2.149198 -0.6718737
## tarsus_Intercept   4.444274 1.2018090  2.379517  7.1491575
##
## 
## 
## $residual__
## $residual__$sd
##           Estimate Est.Error      Q2.5      Q97.5
## bwt      1.396815 0.1597393  1.052042  1.684810
## tarsus   3.733614 0.3170549  3.140204  4.319446
##
## $residual__$cor
## , , bwt
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt      1.0000000 0.0000000  1.0000000  1.0000000
## tarsus   0.8925119 0.0672286  0.7167862  0.9774545
##
## , , tarsus
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt      0.8925119 0.0672286  0.7167862  0.9774545
## tarsus   1.0000000 0.0000000  1.0000000  1.0000000
##
## 
## 
## $residual__$cov
## , , bwt
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt      1.976584 0.4347085  1.106793  2.838585
## tarsus   4.683451 0.9103808  2.846560  6.440550
##
## , , tarsus
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt      4.683451 0.9103808  2.846560  6.44055
## tarsus  14.040300 2.3674563  9.860883 18.65761
```

Evaluation of the statistical support for these genetic and maternal correlations is straightforward. Because we imposed no constraint on their estimation, we can evaluate the extent to which the posterior distributions overlap zero:

```
cor_g <- as.mcmc((VarCorr(brms_m2.2, summary = FALSE)$animal$cor[, 1, 2]))
cor_res <- as.mcmc((VarCorr(brms_m2.2, summary = FALSE)$residual$cor[, 1, 2]))
cor_mother <- as.mcmc((VarCorr(brms_m2.2, summary = FALSE)$mother$cor[, 1, 2]))
cor_byear <- as.mcmc((VarCorr(brms_m2.2, summary = FALSE)$byear$cor[, 1, 2]))

summary(cor_g)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD   Naive SE Time-series SE
## 0.596151  0.172794  0.005464  0.028178
##
## 2. Quantiles for each variable:
##
##    2.5%    25%    50%    75%  97.5%
## 0.1850  0.5065  0.6117  0.7064  0.8930
```

```
summary(cor_mother)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD   Naive SE Time-series SE
## -0.641333  0.196810  0.006224  0.021355
##
## 2. Quantiles for each variable:
##
##    2.5%    25%    50%    75%  97.5%
## -0.9746 -0.7933 -0.6507 -0.5027 -0.2367
```

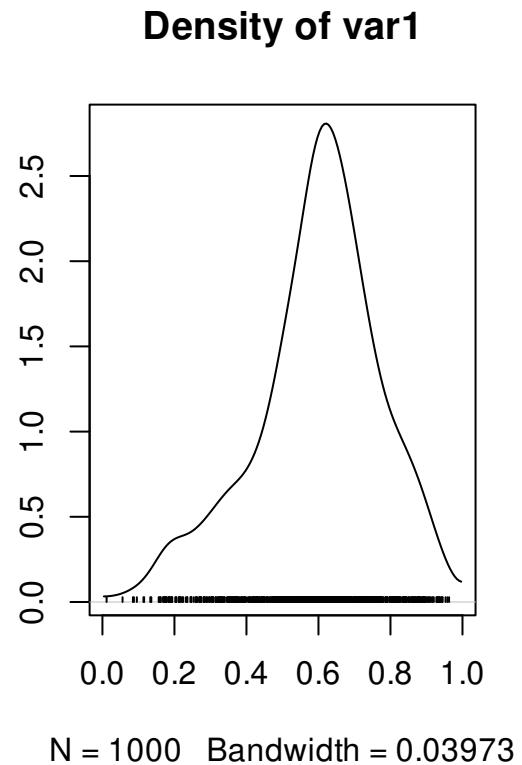
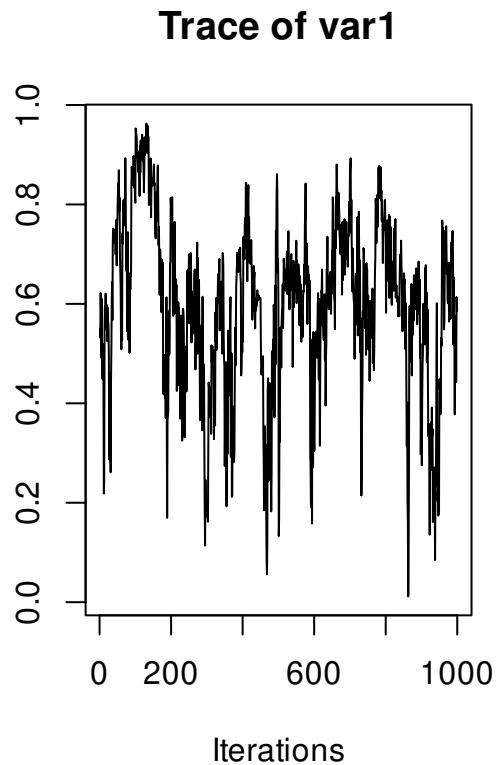
```
summary(cor_byear)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD      Naive SE Time-series SE
## 0.013514   0.220619   0.006977   0.009633
##
## 2. Quantiles for each variable:
##
##    2.5%     25%     50%     75%   97.5%
## -0.43678 -0.13505  0.02408  0.16947  0.44123
```

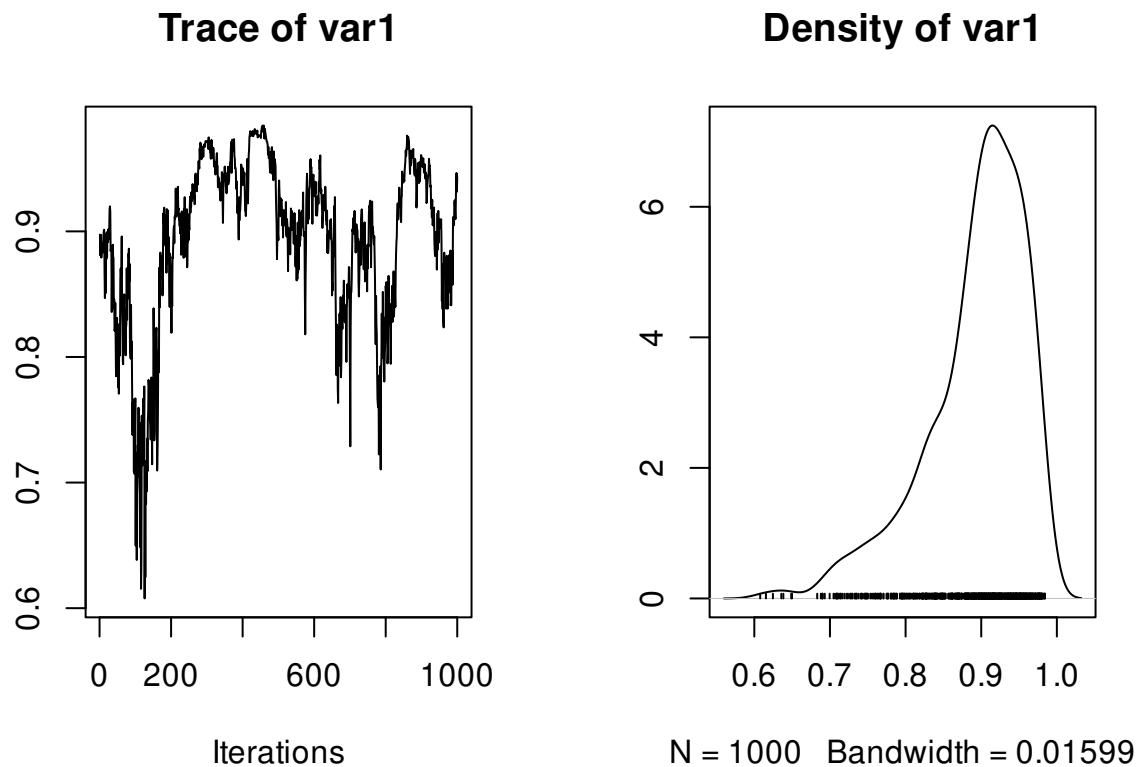
```
summary(cor_res)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD      Naive SE Time-series SE
## 0.892512   0.067229   0.002126   0.026496
##
## 2. Quantiles for each variable:
##
##    2.5%     25%     50%     75%   97.5%
## 0.7168  0.8620  0.9069  0.9425  0.9775
```

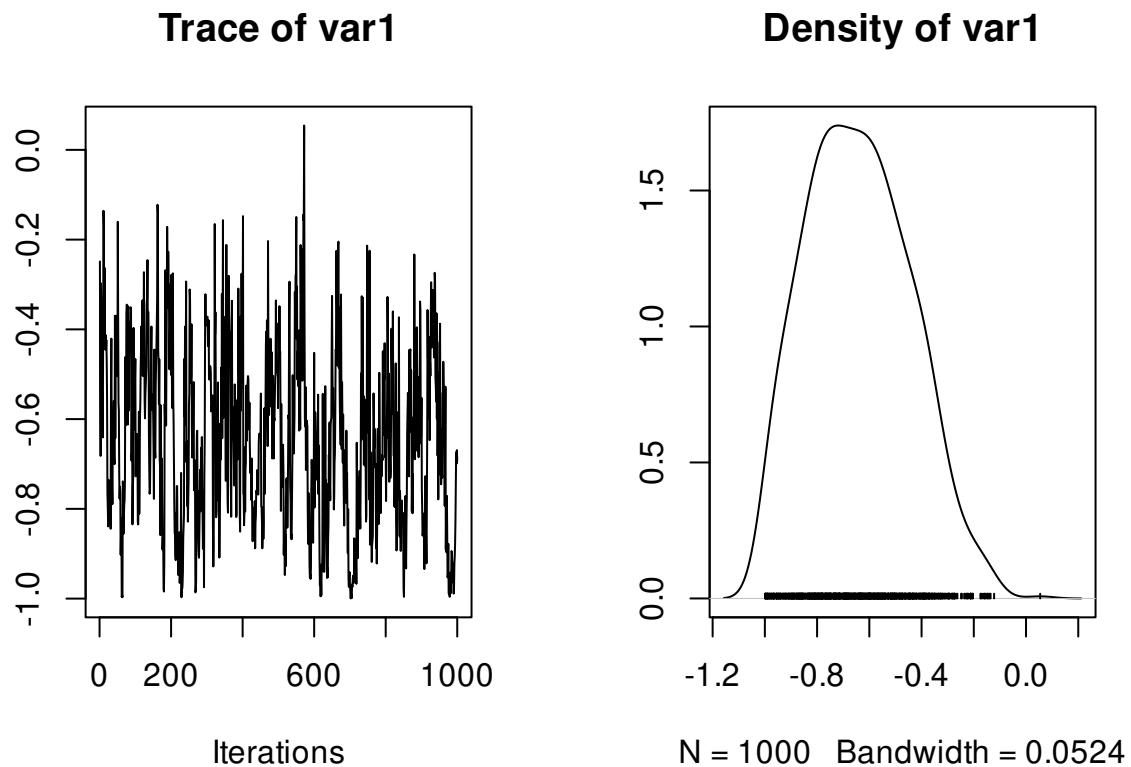
```
plot(cor_g)
```



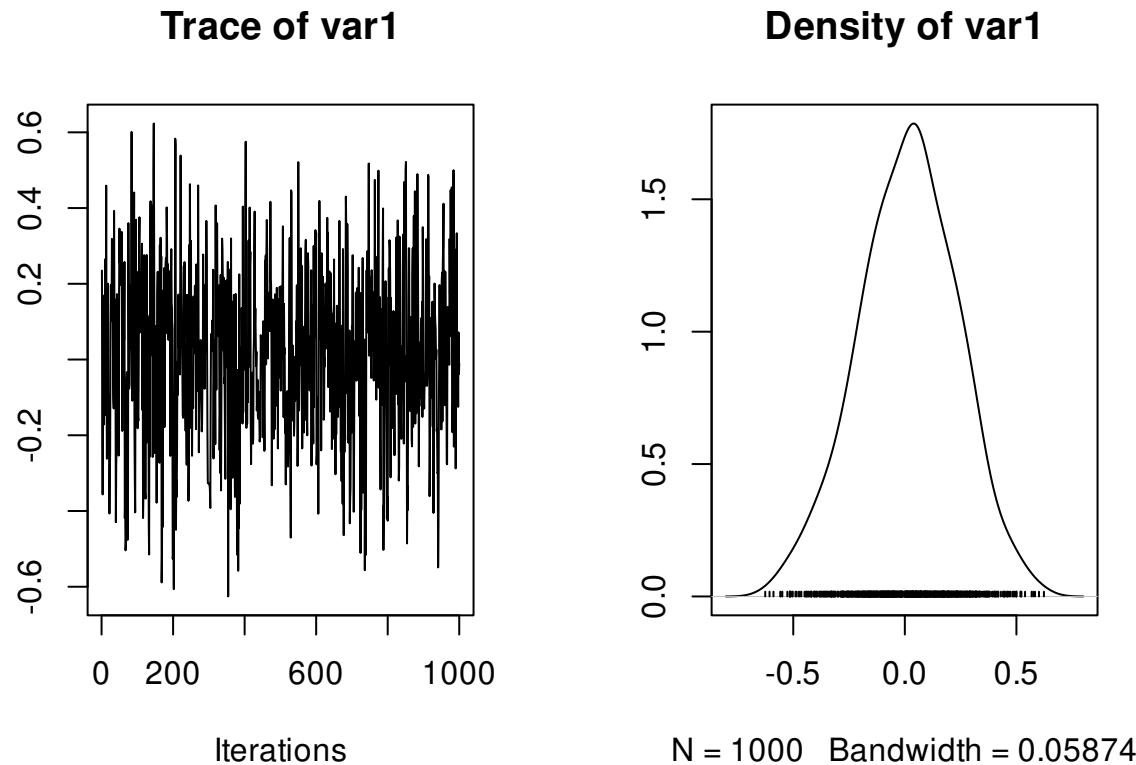
```
plot(cor_res)
```



```
plot(cor_mother)
```



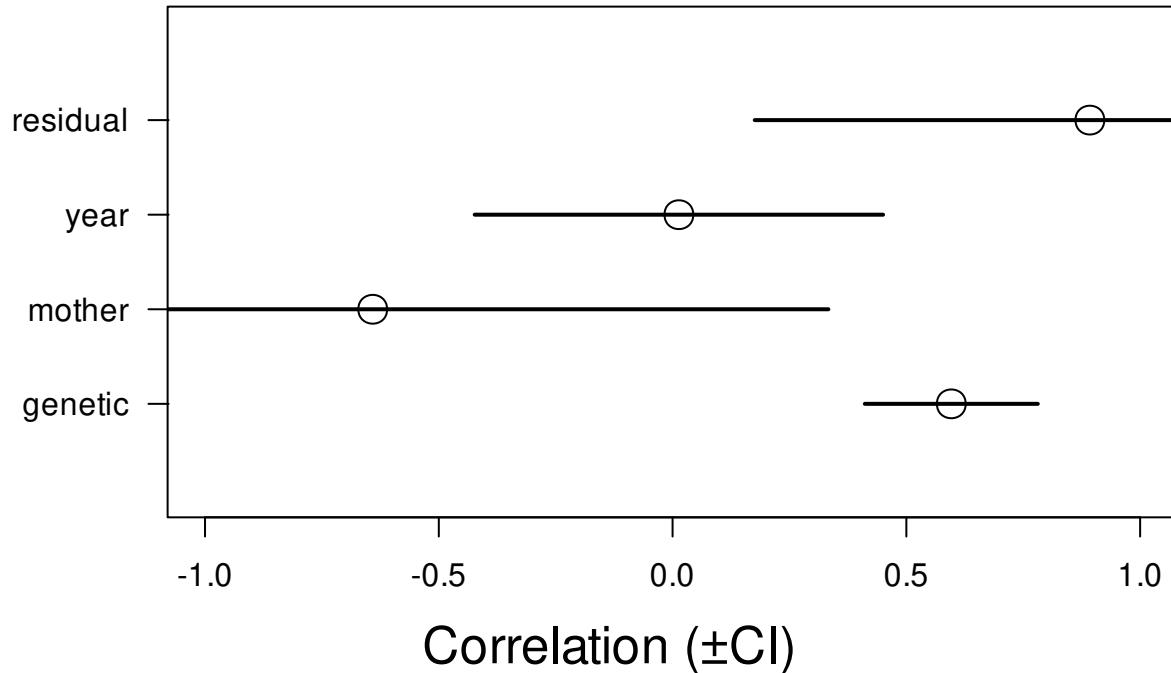
```
plot(cor_byear)
```



Neither of these posterior distributions overlaps zero, so we can consider them both statistically supported.

```
cor.est <- rbind(
  cbind(summary(cor_g)$statistics[1], summary(cor_g)$quantiles[1], summary(cor_g)$quantile,
  cbind(summary(cor_mother)$statistics[1], summary(cor_mother)$quantiles[1], summary(cor_m
  cbind(summary(cor_byear)$statistics[1], summary(cor_byear)$quantiles[1], summary(cor_bye
  cbind(summary(cor_res)$statistics[1], summary(cor_res)$quantiles[1], summary(cor_res)$qu

)
plot(c(1, 2, 3, 4) ~ cor.est[, 1], xlim = c(-1, 1), ylim = c(0, 5), xlab = "", ylab = "",
segments(y0 = 1, x0 = cor.est[1, 1] - cor.est[1, 2], y1 = 1, x1 = cor.est[1, 1] + cor.est[
segments(y0 = 2, x0 = cor.est[2, 1] - cor.est[2, 2], y1 = 2, x1 = cor.est[2, 1] + cor.est[
segments(y0 = 3, x0 = cor.est[3, 1] - cor.est[3, 2], y1 = 3, x1 = cor.est[3, 1] + cor.est[
segments(y0 = 4, x0 = cor.est[4, 1] - cor.est[4, 2], y1 = 4, x1 = cor.est[4, 1] + cor.est[
mtext("Correlation (\u00b1CI)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6)
axis(2, at = 1, labels = c("genetic"), las = 2, cex.axis = 1)
axis(2, at = 2, labels = c("mother"), las = 2, cex.axis = 1)
axis(2, at = 3, labels = c("year"), las = 2, cex.axis = 1)
axis(2, at = 4, labels = c("residual"), las = 2, cex.axis = 1)
```



Note, brms estimates the correlation and also the covariance. We can also recalculate the correlation directly from the covariance. To facilitate the extraction of the different parameter, we can the fucntion `as_draws_df`

```

cov_g <- (VarCorr(brms_m2.2, summary = FALSE)$animal$cov)[, 1, 2]
cov_res <- (VarCorr(brms_m2.2, summary = FALSE)$residual$cov)[, 1, 2]
cov_mother <- (VarCorr(brms_m2.2, summary = FALSE)$mother$cov)[, 1, 2]
cov_byear <- (VarCorr(brms_m2.2, summary = FALSE)$byear$cov)[, 1, 2]

var.est <- as_draws_df(brms_m2.2, variable = c("sd", "sigma"), regex = TRUE)
var.est <- var.est^2

cor_g_2 <- as.mcmc(cov_g / sqrt(var.est[1] * var.est[2]))
cor_byear_2 <- as.mcmc(cov_byear / sqrt(var.est[3] * var.est[4]))
cor_mother_2 <- as.mcmc(cov_g / sqrt(var.est[5] * var.est[6]))
cor_res_2 <- as.mcmc(cov_res / sqrt(var.est[7] * var.est[8]))

summary(cor_g_2)

```

```

## 
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1

```

```

## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##     plus standard error of the mean:
##
##           Mean          SD      Naive SE Time-series SE
##           0.596151    0.172794    0.005464    0.028178
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## 0.1850  0.5065  0.6117  0.7064  0.8930

```

```
summary(cor_byear_2)
```

```

##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##     plus standard error of the mean:
##
##           Mean          SD      Naive SE Time-series SE
##           0.013514    0.220619    0.006977    0.009633
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## -0.43678 -0.13505  0.02408  0.16947  0.44123

```

```
summary(cor_mother_2)
```

```

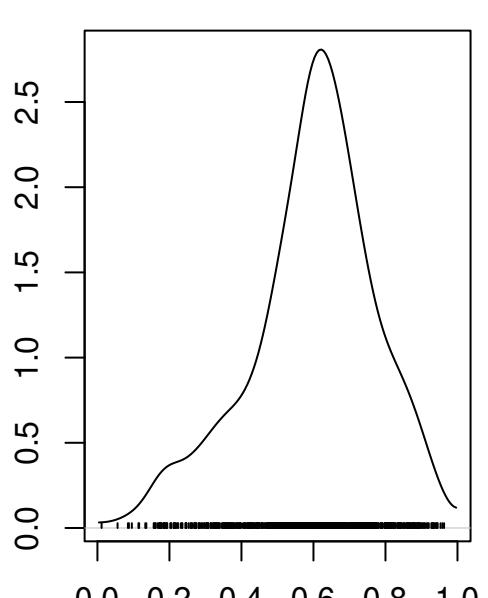
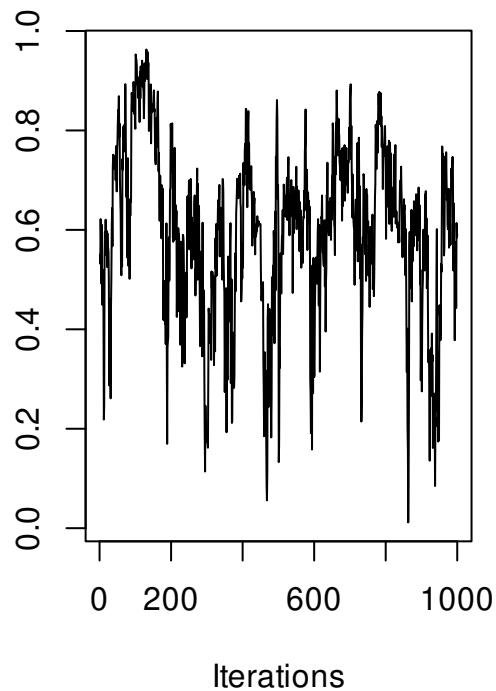
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##     plus standard error of the mean:
##
##           Mean          SD      Naive SE Time-series SE
##           1.01862     0.46337     0.01465     0.06126
##
```

```
## 2. Quantiles for each variable:  
##  
##    2.5%     25%     50%     75%   97.5%  
## 0.1514  0.6969  1.0310  1.3241  2.0032
```

```
summary(cor_res_2)
```

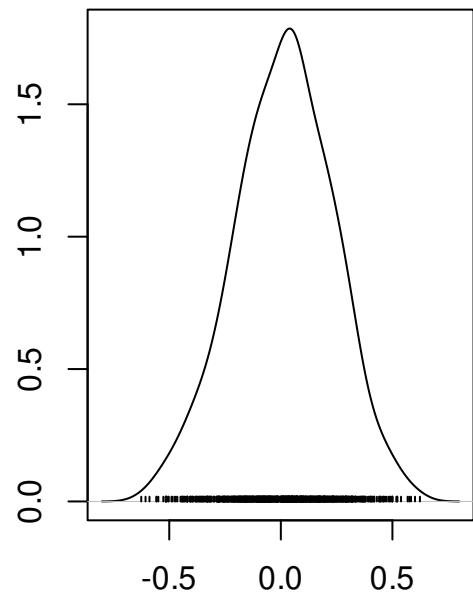
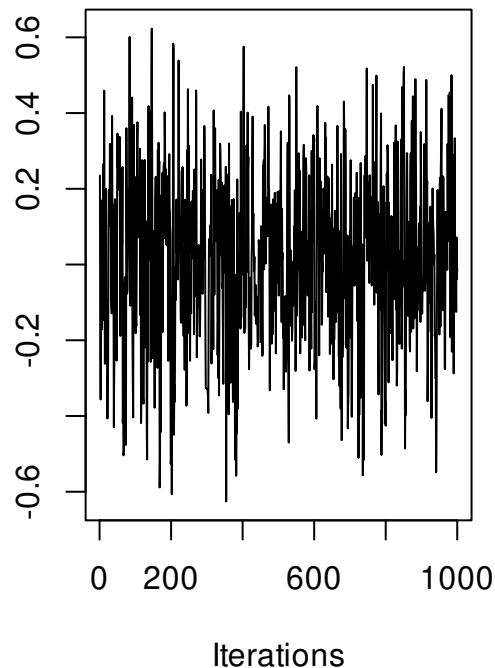
```
##  
## Iterations = 1:1000  
## Thinning interval = 1  
## Number of chains = 1  
## Sample size per chain = 1000  
##  
## 1. Empirical mean and standard deviation for each variable,  
##    plus standard error of the mean:  
##  
##           Mean        SD    Naive SE Time-series SE  
## 0.892512  0.067229  0.002126  0.026496  
##  
## 2. Quantiles for each variable:  
##  
##    2.5%     25%     50%     75%   97.5%  
## 0.7168  0.8620  0.9069  0.9425  0.9775
```

```
plot(cor_g_2)
```

Trace of sd_animal_bwt_Intercept

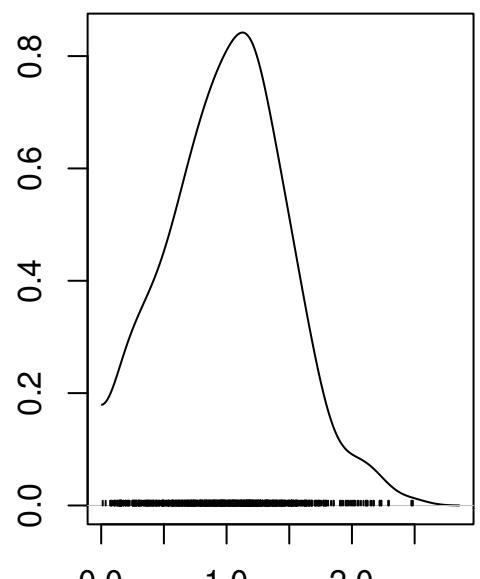
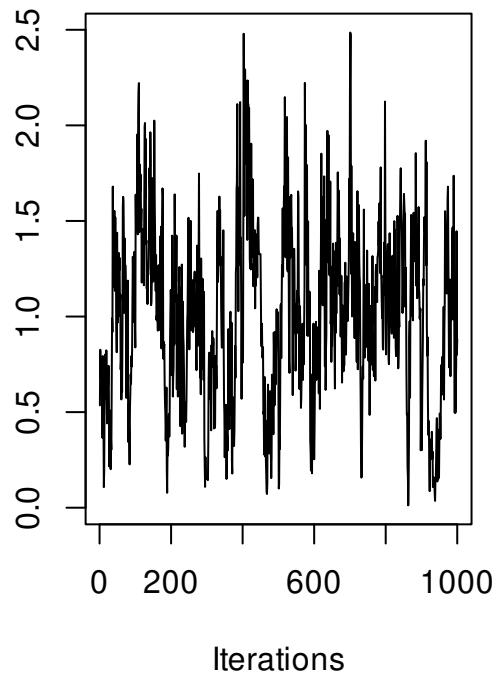
N = 1000 Bandwidth = 0.03973

```
plot(cor_byear_2)
```

Trace of sd_byear_bwt_Interce Density of sd_byear_bwt_Interc

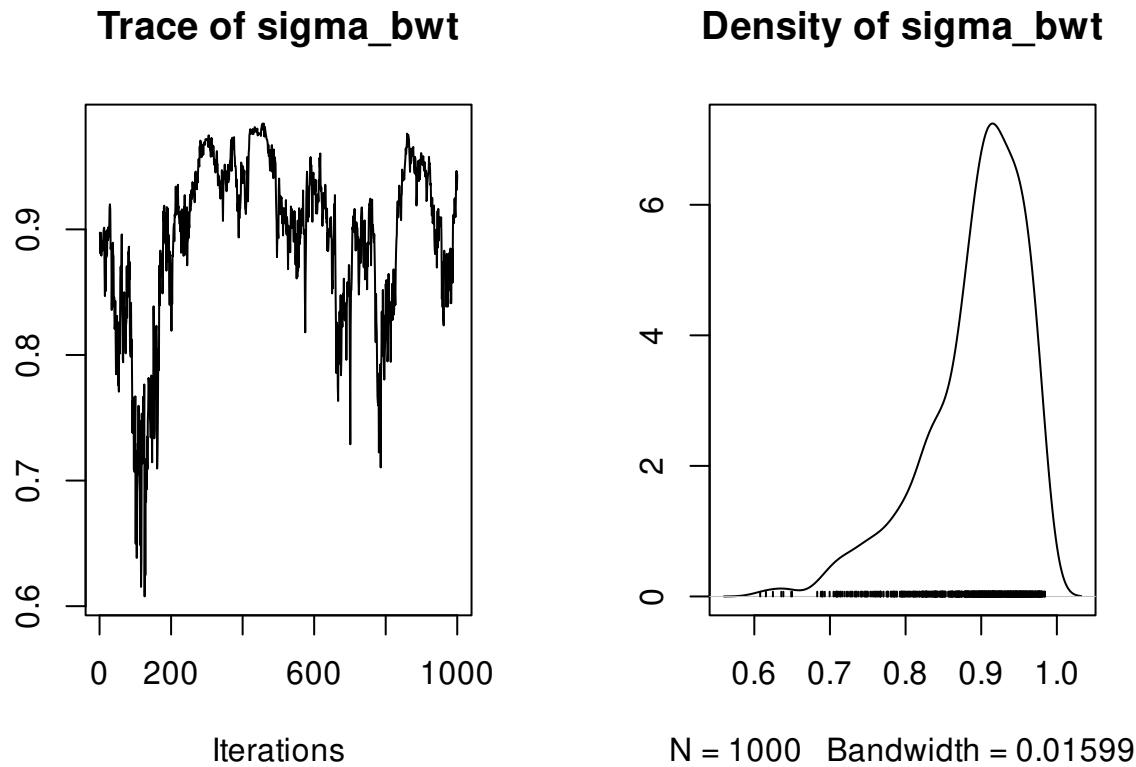
N = 1000 Bandwidth = 0.05874

```
plot(cor_mother_2)
```

Trace of sd_mother__bwt_Interc

N = 1000 Bandwidth = 0.1234

```
plot(cor_res_2)
```



3.5.3 Partitioning (co)variances

As in the tutorial 1, it is possible to partition the variance-covariance matrix between groups (here sex)

```
bf_bwt_3 <- bf(bwt ~ 1 + sex + ((1 | a | gr(animal, cov = Amat, by = sex))) + (1 | b | bye
bf_tarsus_3 <- bf(tarsus ~ 1 + sex + (1 | a | gr(animal, cov = Amat, by = sex)) + (1 | b | b

brms_m2.3 <- brm(
  bf_bwt_3 + bf_tarsus_3 + set_rescor(TRUE),
  data = gryphon,
  data2 = list(Amat = Amat),
  chains = 2, cores = 2, iter = 1000
)

save(brms_m2.3, file = "data/brms_m2_3.rda")
```

Again we have provided the data from one such run. It can be accessed using the code:

```
load("data/brms_m2_3.rda")
summary(brms_m2.3)
```

```

## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

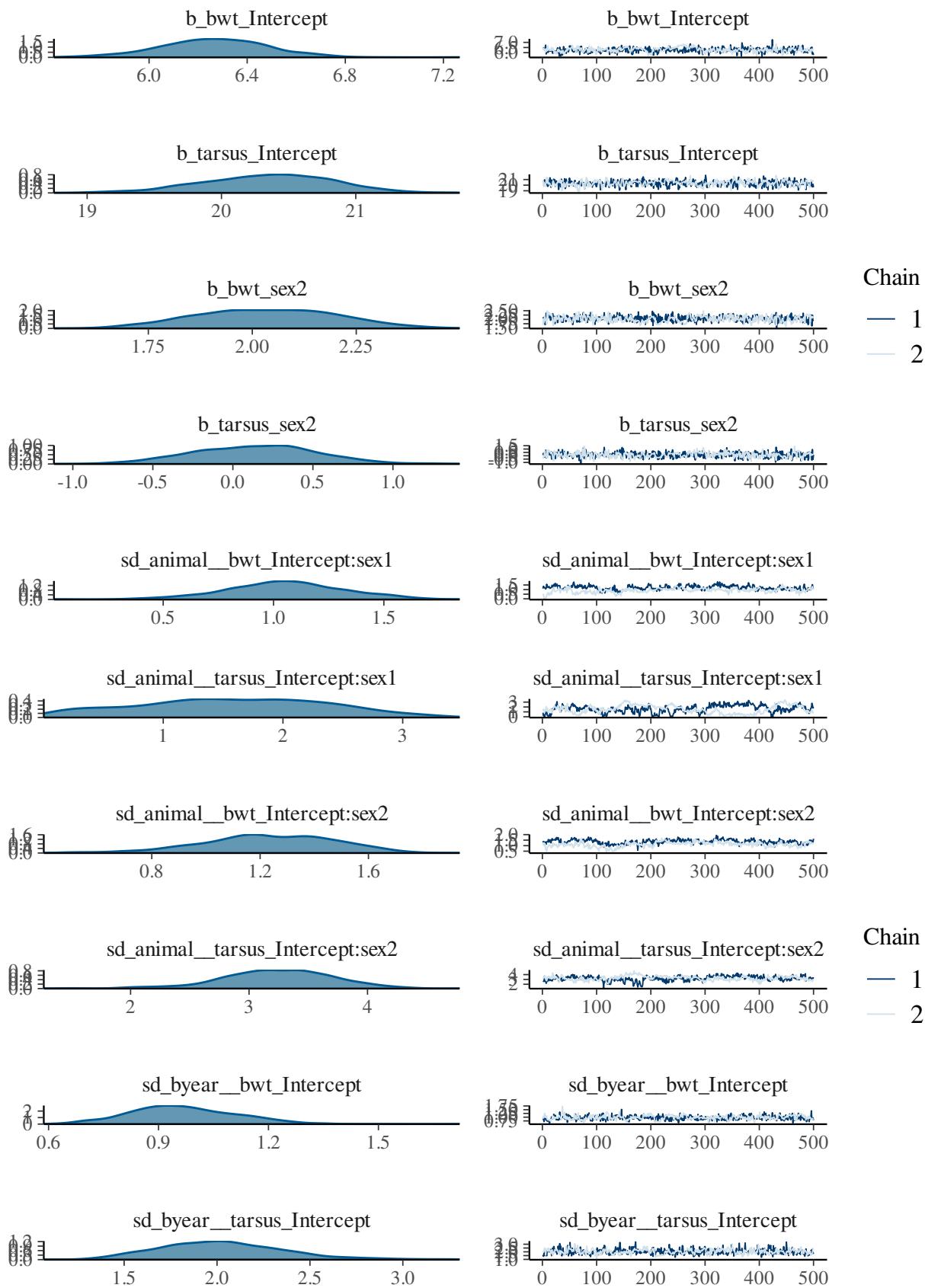
## Warning: There were 6 divergent transitions after warmup. Increasing adapt_delta
## above 0.8 may help. See http://mc-stan.org/misc/warnings.html#divergent-
## transitions-after-warmup

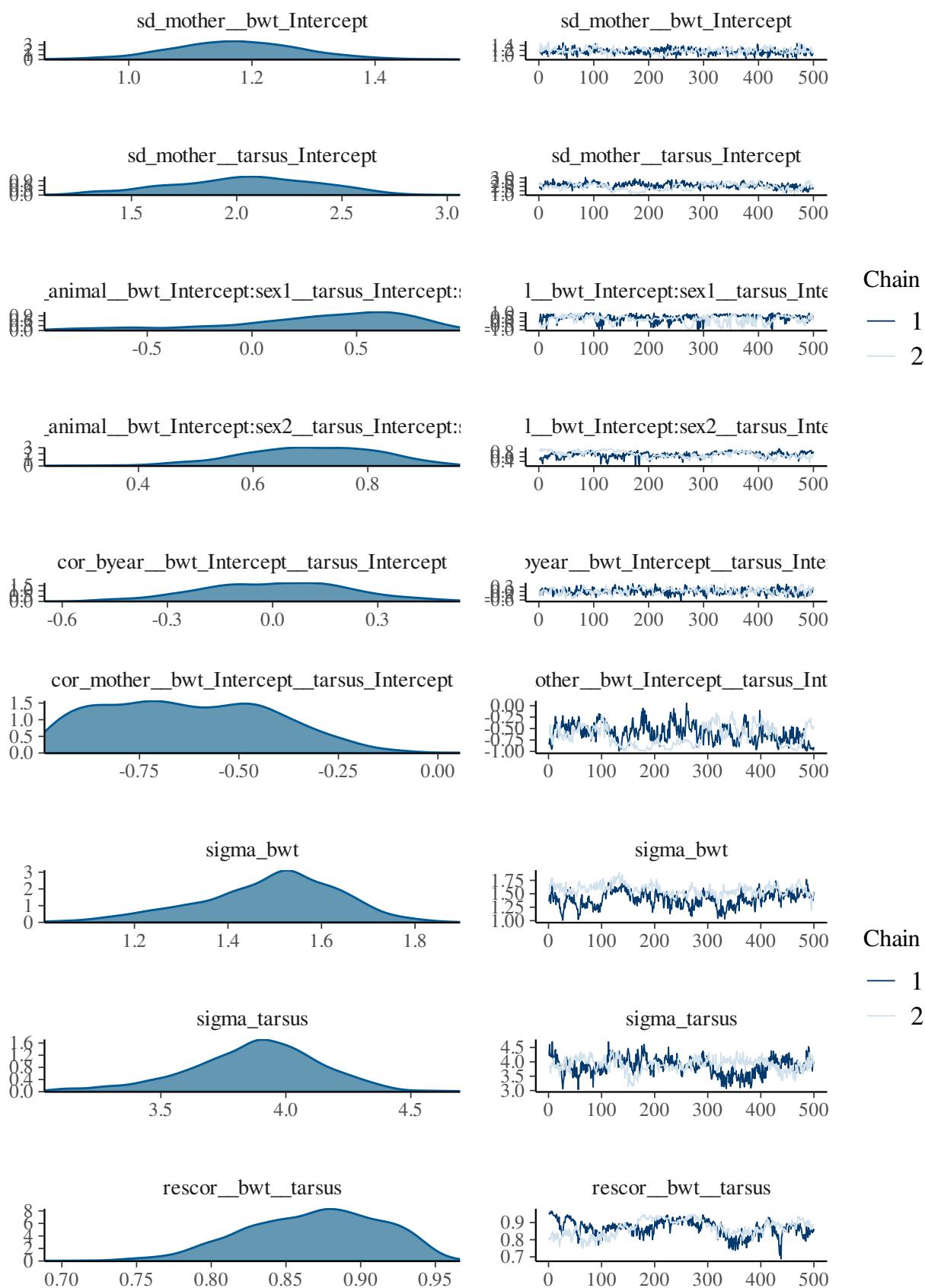
## Family: MV(gaussian, gaussian)
##   Links: mu = identity; sigma = identity
##          mu = identity; sigma = identity
## Formula: bwt ~ 1 + sex + ((1 | a | gr(animal, cov = Amat, by = sex))) + (1 | b | byear)
##           tarsus ~ 1 + sex + (1 | a | gr(animal, cov = Amat, by = sex)) + (1 | b | byear)
## Data: gryphon (Number of observations: 683)
## Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##        total post-warmup draws = 1000
##
## Group-Level Effects:
## ~animal (Number of levels: 683)
##                               Estimate Est.Error l-95% CI
## sd(bwt_Intercept:sex1)            1.05    0.28    0.48
## sd(tarsus_Intercept:sex1)         1.57    0.78    0.14
## sd(bwt_Intercept:sex2)            1.23    0.24    0.71
## sd(tarsus_Intercept:sex2)         3.24    0.48    2.17
## cor(bwt_Intercept:sex1,tarsus_Intercept:sex1)  0.32    0.43   -0.78
## cor(bwt_Intercept:sex2,tarsus_Intercept:sex2)  0.70    0.12    0.45
##                               u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(bwt_Intercept:sex1)            1.58  1.27      6     39
## sd(tarsus_Intercept:sex1)         3.03  1.08     24    117
## sd(bwt_Intercept:sex2)            1.65  1.24      7     46
## sd(tarsus_Intercept:sex2)         4.13  1.05     37     79
## cor(bwt_Intercept:sex1,tarsus_Intercept:sex1)  0.90  1.07     25    202
## cor(bwt_Intercept:sex2,tarsus_Intercept:sex2)  0.90  1.09     28     44
##
## ~byear (Number of levels: 34)
##                               Estimate Est.Error l-95% CI u-95% CI Rhat
## sd(bwt_Intercept)                0.97    0.14    0.70    1.26 1.01
## sd(tarsus_Intercept)              2.03    0.34    1.47    2.80 1.00
## cor(bwt_Intercept,tarsus_Intercept) 0.01    0.21   -0.41    0.41 1.01
##                               Bulk_ESS Tail_ESS
## sd(bwt_Intercept)                  282     292
## sd(tarsus_Intercept)                 324     361
## cor(bwt_Intercept,tarsus_Intercept) 183     393
##
## ~mother (Number of levels: 352)
##                               Estimate Est.Error l-95% CI u-95% CI Rhat
## sd(bwt_Intercept)                1.18    0.11    0.96    1.39 1.02
## sd(tarsus_Intercept)              2.04    0.34    1.33    2.65 1.09

```

```
## cor(bwt_Intercept,tarsus_Intercept) -0.63 0.21 -0.97 -0.23 1.08
##                                     Bulk_ESS Tail_ESS
## sd(bwt_Intercept)                170    352
## sd(tarsus_Intercept)             24     60
## cor(bwt_Intercept,tarsus_Intercept) 22    100
##
## Population-Level Effects:
##                         Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## bwt_Intercept      6.27     0.22    5.82    6.68 1.00    195    123
## tarsus_Intercept  20.35     0.48   19.35   21.21 1.00    494    578
## bwt_sex2          2.04     0.17    1.71    2.36 1.01    265    384
## tarsus_sex2       0.14     0.39   -0.60    0.88 1.00    691    621
##
## Family Specific Parameters:
##                         Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma_bwt         1.49     0.15    1.16    1.74 1.33      5     81
## sigma_tarsus      3.88     0.27    3.27    4.35 1.04     42     85
##
## Residual Correlations:
##                         Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## rescor(bwt,tarsus)  0.87     0.05    0.78    0.94 1.04     22     94
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
```

```
plot(brms_m2.3, ask = FALSE)
```





```
VarCorr(brms_m2.3)
```

```
## $animal
## $animal$sd
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     1.051791 0.2757021 0.4836926 1.575837
## tarsus_Intercept:sex1 1.573820 0.7777373 0.1389929 3.025426
## bwt_Intercept:sex2     1.232722 0.2436415 0.7075357 1.650628
## tarsus_Intercept:sex2 3.237363 0.4773725 2.1667846 4.126560
##
## $animal$cor
## , , bwt_Intercept:sex1
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept:sex1 0.3232502 0.4324034 -0.7811983 0.9007145
## bwt_Intercept:sex2     0.0000000 0.0000000 0.0000000 0.0000000
## tarsus_Intercept:sex2 0.0000000 0.0000000 0.0000000 0.0000000
##
## , , tarsus_Intercept:sex1
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.3232502 0.4324034 -0.7811983 0.9007145
## tarsus_Intercept:sex1 1.0000000 0.0000000 1.0000000 1.0000000
## bwt_Intercept:sex2     0.0000000 0.0000000 0.0000000 0.0000000
## tarsus_Intercept:sex2 0.0000000 0.0000000 0.0000000 0.0000000
##
## , , bwt_Intercept:sex2
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.0000000 0.0000000 0.0000000 0.0000000
## tarsus_Intercept:sex1 0.0000000 0.0000000 0.0000000 0.0000000
## bwt_Intercept:sex2     1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept:sex2 0.6975925 0.1213692 0.4474181 0.9047623
##
## , , tarsus_Intercept:sex2
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.0000000 0.0000000 0.0000000 0.0000000
## tarsus_Intercept:sex1 0.0000000 0.0000000 0.0000000 0.0000000
## bwt_Intercept:sex2     0.6975925 0.1213692 0.4474181 0.9047623
## tarsus_Intercept:sex2 1.0000000 0.0000000 1.0000000 1.0000000
##
## 
## 
## $animal$cov
```

```

## , , bwt_Intercept:sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1    1.1821990 0.5768155  0.2339592 2.483263
## tarsus_Intercept:sex1 0.7957577 0.9482719 -0.4879668 3.035395
## bwt_Intercept:sex2    0.0000000 0.0000000  0.0000000 0.000000
## tarsus_Intercept:sex2 0.0000000 0.0000000  0.0000000 0.000000
##
## , , tarsus_Intercept:sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1    0.7957577 0.9482719 -0.48796685 3.035395
## tarsus_Intercept:sex1 3.0811806 2.5183190  0.01931907 9.153208
## bwt_Intercept:sex2    0.0000000 0.0000000  0.0000000 0.000000
## tarsus_Intercept:sex2 0.0000000 0.0000000  0.0000000 0.000000
##
## , , bwt_Intercept:sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1    0.000000 0.0000000 0.0000000 0.000000
## tarsus_Intercept:sex1 0.000000 0.0000000 0.0000000 0.000000
## bwt_Intercept:sex2    1.578907 0.5865683 0.5006069 2.724572
## tarsus_Intercept:sex2 2.842372 1.0153859 1.1593760 5.086448
##
## , , tarsus_Intercept:sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1    0.000000 0.000000 0.000000 0.000000
## tarsus_Intercept:sex1 0.000000 0.000000 0.000000 0.000000
## bwt_Intercept:sex2    2.842372 1.015386 1.159376 5.086448
## tarsus_Intercept:sex2 10.708178 3.017245 4.694964 17.028497
##
##
##
## $byear
## $byear$sd
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept    0.9676965 0.1434955 0.7018779 1.258518
## tarsus_Intercept 2.0290144 0.3382466 1.4650518 2.801932
##
## $byear$cor
## , , bwt_Intercept
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept    1.000000000 0.0000000 1.0000000 1.000000
## tarsus_Intercept 0.009103073 0.2077977 -0.4096021 0.410902
##

```

```

## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 0.009103073 0.2077977 -0.4096021 0.410902
## tarsus_Intercept 1.000000000 0.0000000 1.0000000 1.000000
##
## 
## 
## $byear$cov
## , , bwt_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 0.95700691 0.2875804 0.4926327 1.583869
## tarsus_Intercept 0.04233908 0.4457863 -0.8475401 1.014920
##
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 0.04233908 0.4457863 -0.8475401 1.014920
## tarsus_Intercept 4.23119588 1.4453420 2.1463767 7.850826
##
## 
## 
## 
## $mother
## $mother$sd
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 1.175168 0.1083817 0.958963 1.388926
## tarsus_Intercept 2.038471 0.3445355 1.327549 2.648997
##
## 
## $mother$cor
## , , bwt_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept -0.6301934 0.2113836 -0.9652044 -0.2283133
##
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept -0.6301934 0.2113836 -0.9652044 -0.2283133
## tarsus_Intercept 1.0000000 0.0000000 1.0000000 1.0000000
##
## 
## 
## $mother$cov
## , , bwt_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 1.392755 0.2562907 0.9196105 1.929117

```

```

## tarsus_Intercept -1.434727 0.3664843 -2.0769067 -0.674943
##
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept     -1.434727 0.3664843 -2.076907 -0.674943
## tarsus_Intercept   4.273951 1.3914013  1.762387  7.017188
##
## 
## 
## $residual__
## $residual__$sd
##             Estimate Est.Error      Q2.5      Q97.5
## bwt      1.489342 0.1489641 1.162192 1.741559
## tarsus  3.875633 0.2650604 3.268802 4.345400
##
## $residual__$cor
## , , bwt
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt      1.0000000 0.0000000 1.0000000 1.0000000
## tarsus  0.8685184 0.04534008 0.7772864 0.9414488
##
## , , tarsus
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt      0.8685184 0.04534008 0.7772864 0.9414488
## tarsus  1.0000000 0.0000000 1.0000000 1.0000000
##
## 
## $residual__$cov
## , , bwt
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt      2.240307 0.4341334 1.350691 3.033029
## tarsus  5.034739 0.7852137 3.272088 6.400539
##
## , , tarsus
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt      5.034739 0.7852137 3.272088 6.400539
## tarsus 15.090721 2.0309529 10.685067 18.882505

```

However, this model is lacking an important and essential group-specific partitioning (we do with the asreml-R and MCMCglmm). We need to partition the residual variance (or sigma) as well. Doing so, we will use the argument ‘sigma’ to partition the model by sex. To avoid an estimation of the difference between sexes, we need to remove the estimate of the intercept at the sigma level.

```

bf_bwt_4 <- bf(bwt ~ 1 + sex + ((1 | a | gr(animal, cov = Amat, by = sex))) + (1 | b | bye
bf_tarsus_4 <- bf(tarsus ~ 1 + sex + (1 | a | gr(animal, cov = Amat, by = sex)) + (1 | b | b

brms_m2.4 <- brm(
  bf_bwt_4 + bf_tarsus_4 + set_rescor(TRUE),
  data = gryphon,
  data2 = list(Amat = Amat),
  chains = 2, cores = 2, iter = 1000
)
save(brms_m2.4, file = "data/brms_m2.4.rda")

```

Again we have provided the data from one such run. It can be accessed using the code:

```

load("data/brms_m2.4.rda")
summary(brms_m2.4)

```

```

## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## Warning: There were 6 divergent transitions after warmup. Increasing adapt_delta
## above 0.8 may help. See http://mc-stan.org/misc/warnings.html#divergent-
## transitions-after-warmup

## Family: MV(gaussian, gaussian)
##   Links: mu = identity; sigma = log
##          mu = identity; sigma = log
## Formula: bwt ~ 1 + sex + ((1 | a | gr(animal, cov = Amat, by = sex))) + (1 | b | byear)
##          sigma ~ sex - 1
##          tarsus ~ 1 + sex + (1 | a | gr(animal, cov = Amat, by = sex)) + (1 | b | byear)
##          sigma ~ sex - 1
## Data: gryphon (Number of observations: 683)
## Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##        total post-warmup draws = 1000
##
## Group-Level Effects:
## ~animal (Number of levels: 683)
##                               Estimate Est.Error l-95% CI
## sd(bwt_Intercept:sex1)           0.86      0.35     0.14
## sd(tarsus_Intercept:sex1)        1.40      0.82     0.10
## sd(bwt_Intercept:sex2)           1.40      0.23     0.94
## sd(tarsus_Intercept:sex2)        3.49      0.72     1.96
## cor(bwt_Intercept:sex1,tarsus_Intercept:sex1)  0.28      0.51    -0.85
## cor(bwt_Intercept:sex2,tarsus_Intercept:sex2)  0.78      0.10     0.56
##                               u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(bwt_Intercept:sex1)           1.48    1.00       59      101

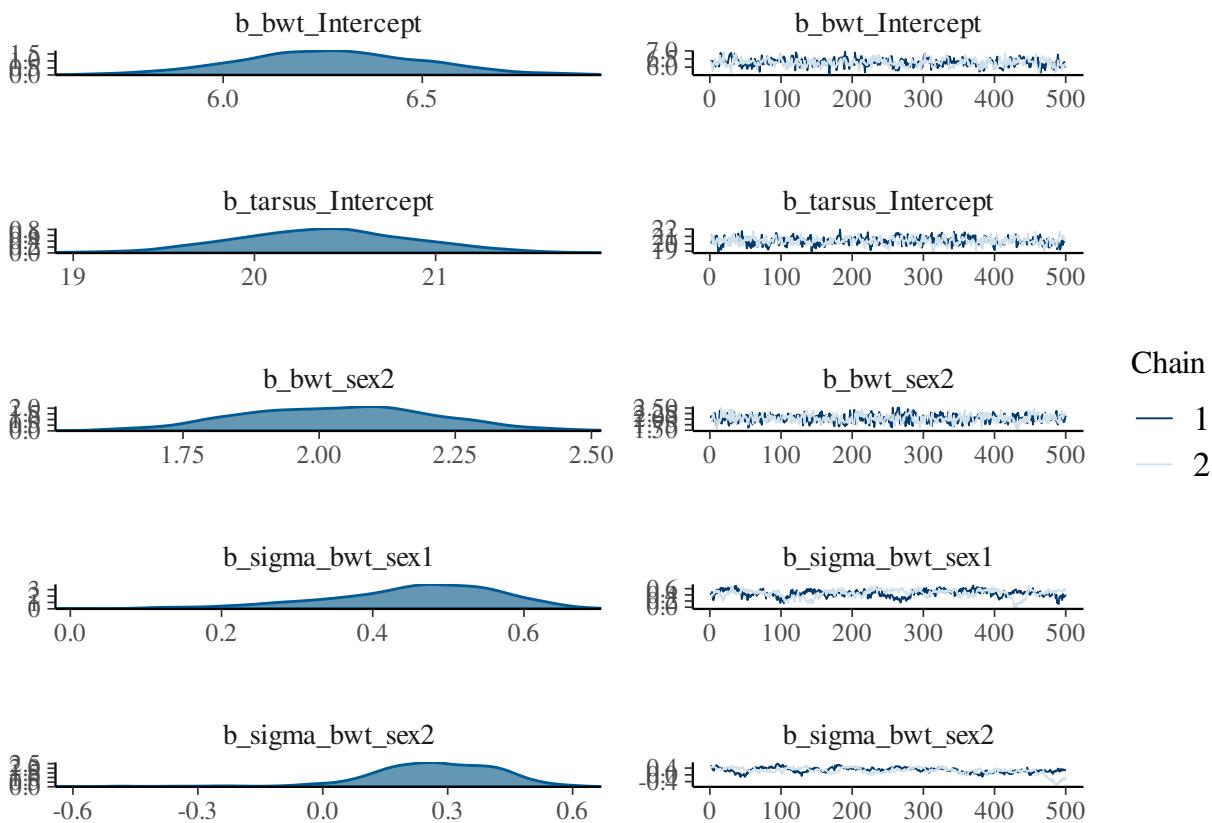
```

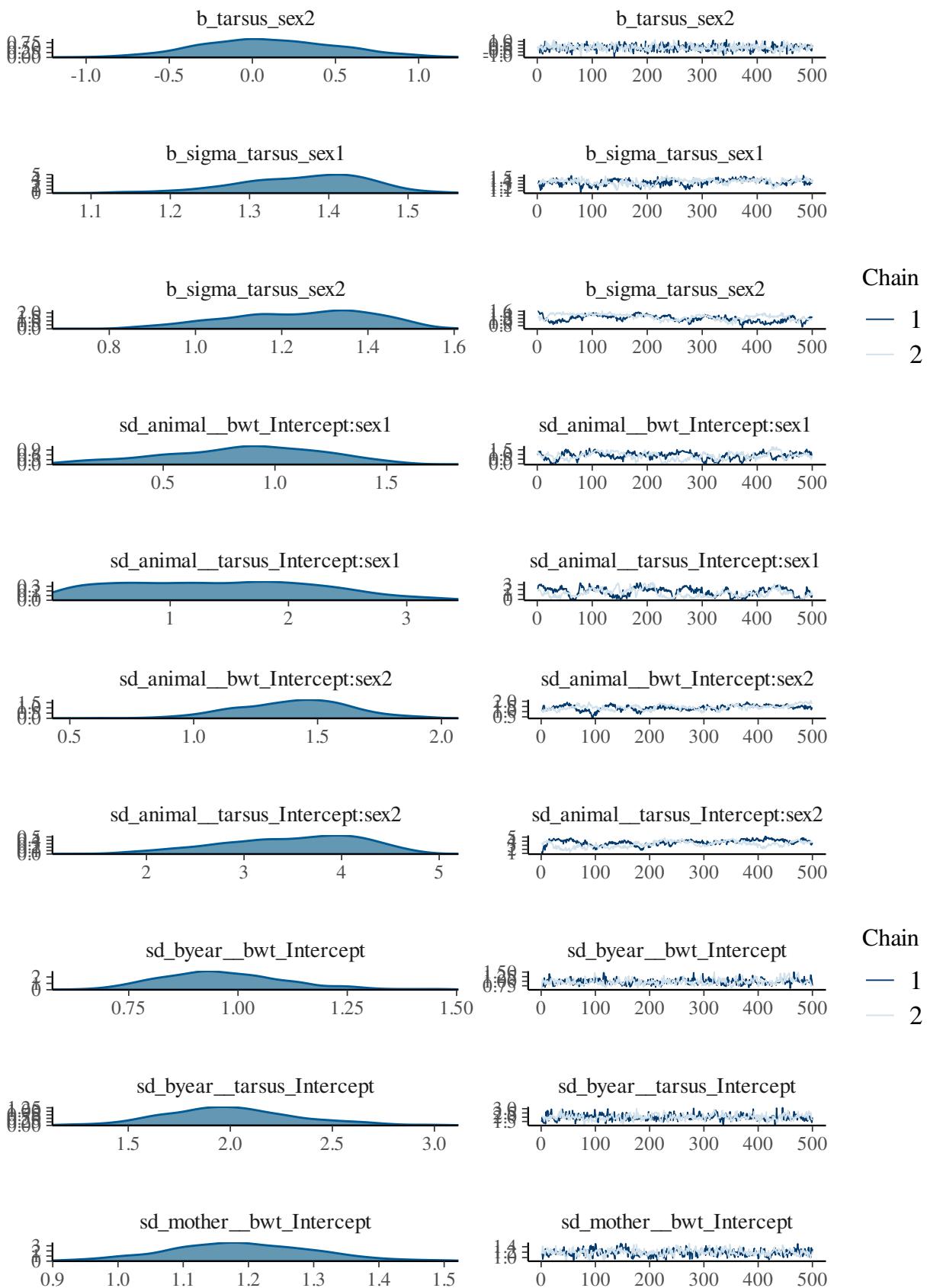
```

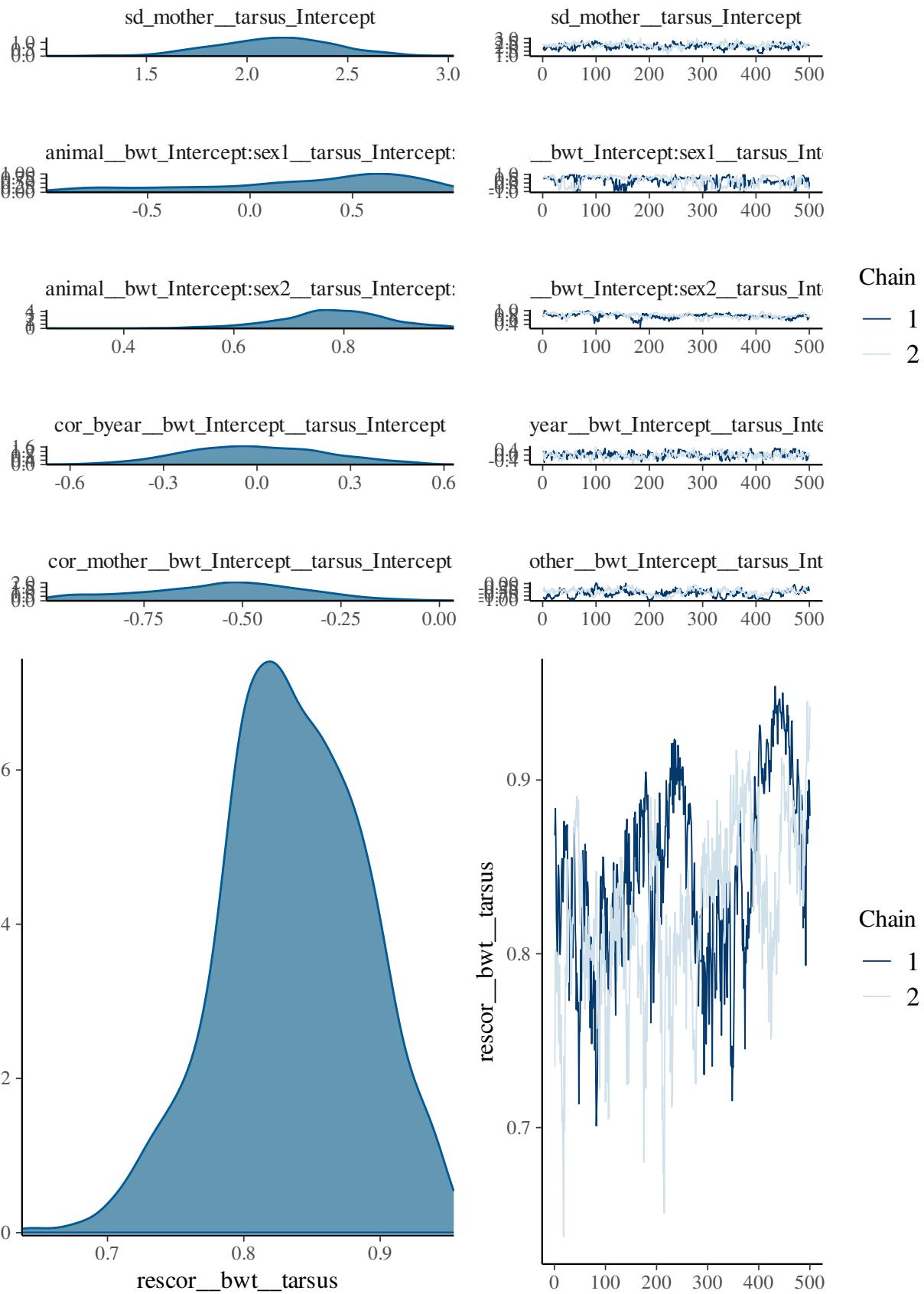
## sd(tarsus_Intercept:sex1)           3.00 1.07    45    111
## sd(bwt_Intercept:sex2)            1.83 1.04    42     40
## sd(tarsus_Intercept:sex2)          4.62 1.22     8     50
## cor(bwt_Intercept:tarsus_Intercept) 0.94 1.05    60    287
## cor(bwt_Intercept:tarsus_Intercept) 0.97 1.09    30     94
##
## ~byear (Number of levels: 34)
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat
## sd(bwt_Intercept)                  0.97      0.15     0.73    1.30 1.01
## sd(tarsus_Intercept)              2.01      0.33     1.41    2.70 1.00
## cor(bwt_Intercept,tarsus_Intercept) -0.00      0.22    -0.42    0.45 1.02
##                                         Bulk_ESS Tail_ESS
## sd(bwt_Intercept)                  283       412
## sd(tarsus_Intercept)              349       554
## cor(bwt_Intercept,tarsus_Intercept) 225       256
##
## ~mother (Number of levels: 352)
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat
## sd(bwt_Intercept)                  1.19      0.11     0.98    1.42 1.00
## sd(tarsus_Intercept)              2.14      0.30     1.57    2.71 1.05
## cor(bwt_Intercept,tarsus_Intercept) -0.55      0.21    -0.96   -0.16 1.04
##                                         Bulk_ESS Tail_ESS
## sd(bwt_Intercept)                  279       434
## sd(tarsus_Intercept)              46        227
## cor(bwt_Intercept,tarsus_Intercept) 46        122
##
## Population-Level Effects:
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## bwt_Intercept                   6.27      0.23     5.82    6.73 1.01    336    556
## tarsus_Intercept                20.39     0.49    19.42   21.37 1.00    384    509
## bwt_sex2                        2.04      0.17     1.70    2.38 1.00    483    507
## sigma_bwt_sex1                 0.45      0.12     0.18    0.63 1.00     68    128
## sigma_bwt_sex2                 0.26      0.17    -0.17    0.52 1.06     38     33
## tarsus_sex2                     0.10      0.41    -0.69    0.92 1.00    658    659
## sigma_tarsus_sex1               1.37      0.08     1.19    1.50 1.04     65    215
## sigma_tarsus_sex2               1.24      0.16     0.91    1.50 1.22      7     67
##
## Residual Correlations:
##                                         Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## rescor(bwt,tarsus)                0.84      0.05     0.73    0.93 1.15     11     29
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

```

```
plot(brms_m2.4, ask = FALSE)
```







```
VarCorr(brms_m2.4)
```

```
## $animal
## $animal$sd
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.8628389 0.3531696 0.14360929 1.481121
## tarsus_Intercept:sex1 1.3985239 0.8163731 0.09977045 3.002609
## bwt_Intercept:sex2     1.4023202 0.2325863 0.93896619 1.833688
## tarsus_Intercept:sex2 3.4858243 0.7167230 1.96177030 4.620105
##
## $animal$cor
## , , bwt_Intercept:sex1
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     1.000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept:sex1 0.277338 0.5119501 -0.8479996 0.9398158
## bwt_Intercept:sex2     0.000000 0.0000000 0.0000000 0.0000000
## tarsus_Intercept:sex2 0.000000 0.0000000 0.0000000 0.0000000
##
## , , tarsus_Intercept:sex1
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.277338 0.5119501 -0.8479996 0.9398158
## tarsus_Intercept:sex1 1.000000 0.0000000 1.0000000 1.0000000
## bwt_Intercept:sex2     0.000000 0.0000000 0.0000000 0.0000000
## tarsus_Intercept:sex2 0.000000 0.0000000 0.0000000 0.0000000
##
## , , bwt_Intercept:sex2
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.0000000 0.00000000 0.000000 0.0000000
## tarsus_Intercept:sex1 0.0000000 0.00000000 0.000000 0.0000000
## bwt_Intercept:sex2     1.0000000 0.00000000 1.000000 1.0000000
## tarsus_Intercept:sex2 0.7781659 0.09850178 0.564036 0.9699765
##
## , , tarsus_Intercept:sex2
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.0000000 0.00000000 0.000000 0.0000000
## tarsus_Intercept:sex1 0.0000000 0.00000000 0.000000 0.0000000
## bwt_Intercept:sex2     0.7781659 0.09850178 0.564036 0.9699765
## tarsus_Intercept:sex2 1.0000000 0.00000000 1.000000 1.0000000
##
## 
## $animal$cov
```

```

## , , bwt_Intercept:sex1
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.8690949 0.6002446  0.0206250 2.193721
## tarsus_Intercept:sex1 0.6483841 0.9143700 -0.4481203 2.989048
## bwt_Intercept:sex2     0.0000000 0.0000000  0.0000000 0.000000
## tarsus_Intercept:sex2 0.0000000 0.0000000  0.0000000 0.000000
##
## , , tarsus_Intercept:sex1
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.6483841 0.914370 -0.448120294 2.989048
## tarsus_Intercept:sex1 2.6216677 2.500989  0.009954235 9.015666
## bwt_Intercept:sex2     0.0000000 0.000000  0.000000000 0.000000
## tarsus_Intercept:sex2 0.0000000 0.000000  0.000000000 0.000000
##
## , , bwt_Intercept:sex2
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.000000 0.000000 0.0000000 0.000000
## tarsus_Intercept:sex1 0.000000 0.000000 0.0000000 0.000000
## bwt_Intercept:sex2     2.020544 0.639550 0.8816577 3.362416
## tarsus_Intercept:sex2 3.875299 1.298562 1.4247624 6.415724
##
## , , tarsus_Intercept:sex2
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept:sex1     0.000000 0.000000 0.000000 0.000000
## tarsus_Intercept:sex1 0.000000 0.000000 0.000000 0.000000
## bwt_Intercept:sex2     3.875299 1.298562 1.424762 6.415724
## tarsus_Intercept:sex2 12.664149 4.814957 3.848544 21.345370
##
##
##
## $byear
## $byear$sd
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept     0.9707654 0.1478361 0.7256973 1.298668
## tarsus_Intercept 2.0073203 0.3290043 1.4102530 2.699770
##
## $byear$cor
## , , bwt_Intercept
##
##                               Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept     1.000000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept -0.001551923 0.2236193 -0.4237849 0.4526618
##

```

```
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept -0.001551923 0.2236193 -0.4237849 0.4526618
## tarsus_Intercept 1.000000000 0.0000000 1.0000000 1.0000000
##
## 
## 
## $byear$cov
## , , bwt_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 0.9642191 0.3021042 0.5266366 1.686538
## tarsus_Intercept 0.0252866 0.4713288 -0.8080314 1.069567
##
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 0.0252866 0.4713288 -0.8080314 1.069567
## tarsus_Intercept 4.1374703 1.3676889 1.9888135 7.288761
##
## 
## 
## 
## $mother
## $mother$sd
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 1.189902 0.1113270 0.9791374 1.423333
## tarsus_Intercept 2.139290 0.2985875 1.5714814 2.708199
##
## 
## $mother$cor
## , , bwt_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_Intercept -0.5501934 0.2066985 -0.9589938 -0.1591737
##
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept -0.5501934 0.2066985 -0.9589938 -0.1591737
## tarsus_Intercept 1.0000000 0.0000000 1.0000000 1.0000000
##
## 
## 
## $mother$cov
## , , bwt_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept 1.428247 0.2672833 0.958710 2.0258758
```

```
## tarsus_Intercept -1.335450 0.3932707 -2.051866 -0.5146954
##
## , , tarsus_Intercept
##
##             Estimate Est.Error      Q2.5      Q97.5
## bwt_Intercept     -1.335450 0.3932707 -2.051866 -0.5146954
## tarsus_Intercept    4.665626 1.2688941  2.469554  7.3343400
```

Evaluation of the statistical support for these sex-specific correlations is straightforward. Because we imposed no constraint on their estimation, we can evaluate the extent to which the posterior distributions overlap zero or overlap each other:

```
cor_g_F <- as.mcmc((VarCorr(brms_m2.4, summary = FALSE)$animal$cor[, 1, 2]))
cor_g_M <- as.mcmc((VarCorr(brms_m2.4, summary = FALSE)$animal$cor[, 3, 4]))
```

```
summary(cor_g_F)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##             Mean           SD       Naive SE Time-series SE
## 0.27734        0.51195      0.01619      0.05981
##
## 2. Quantiles for each variable:
##
##    2.5%     25%     50%     75%   97.5%
## -0.84800 -0.02476  0.43226  0.67124  0.93982
```

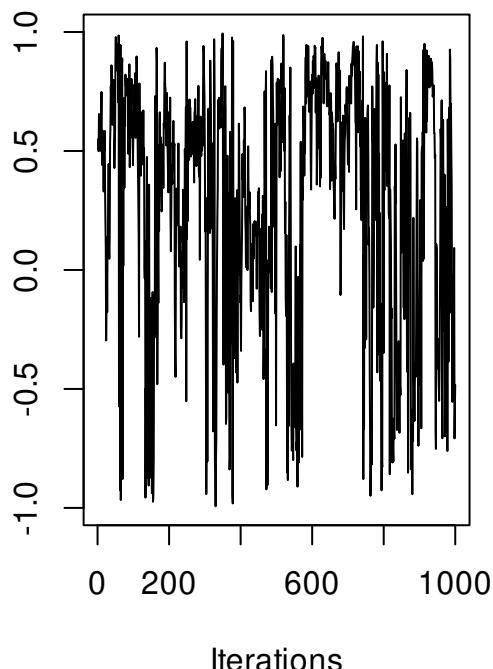
```
summary(cor_g_M)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##             Mean           SD       Naive SE Time-series SE
```

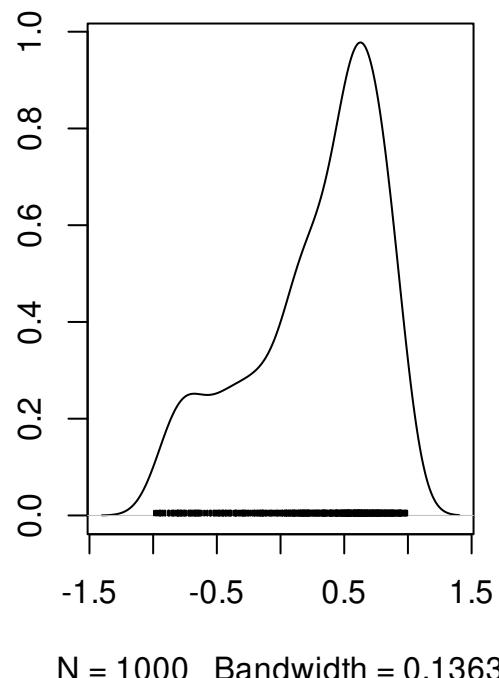
```
##      0.778166      0.098502      0.003115      0.013827
##
## 2. Quantiles for each variable:
##
##   2.5%    25%    50%    75%  97.5%
## 0.5640 0.7240 0.7805 0.8424 0.9700
```

```
plot(cor_g_F)
```

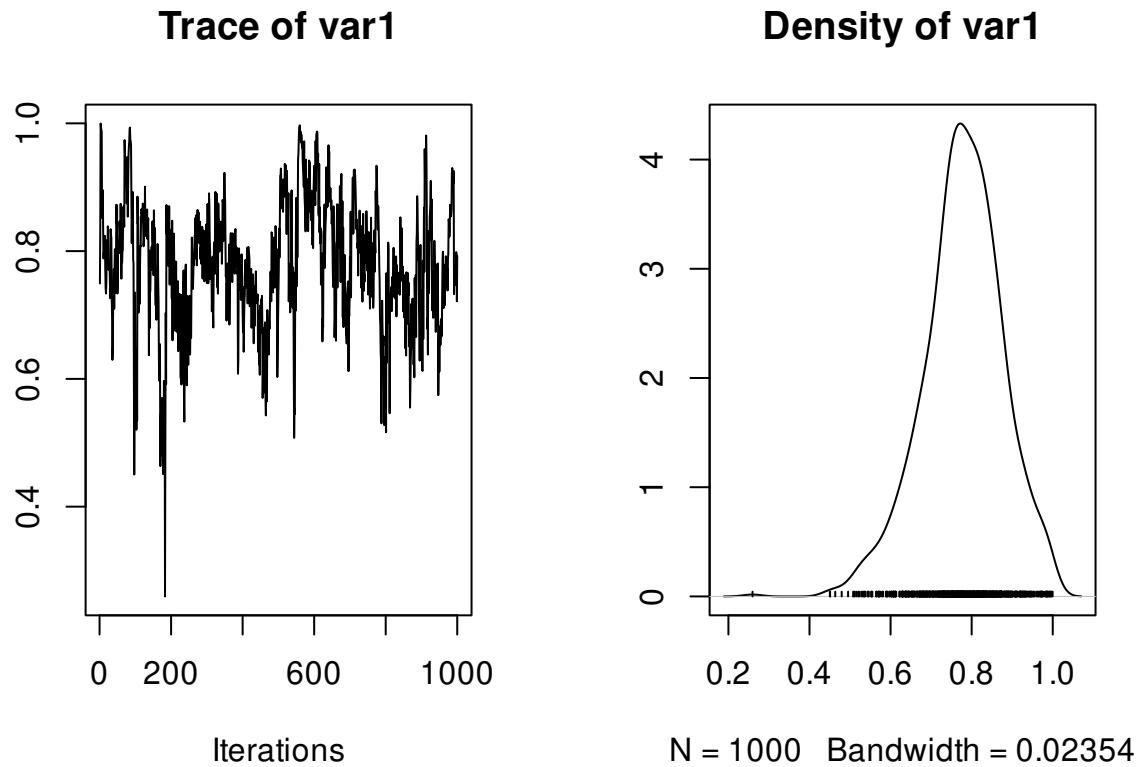
Trace of var1



Density of var1

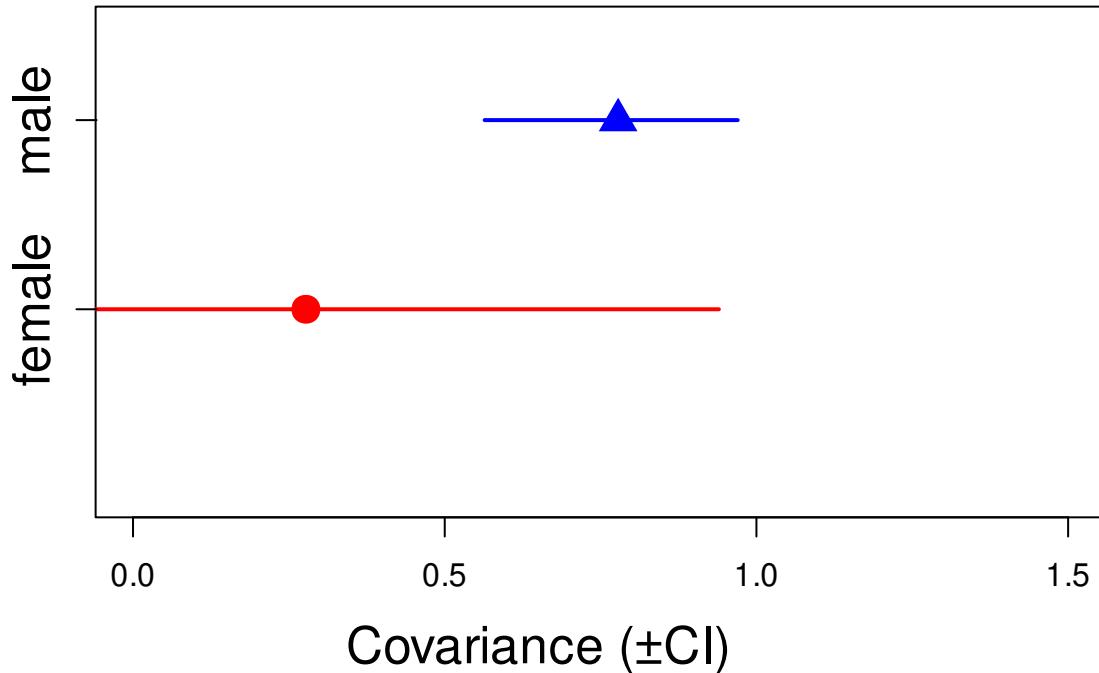


```
plot(cor_g_M)
```



Here a plot to visualize the overlaps of covariances.

```
cor.est <- rbind(
  cbind(summary(cor_g_F)$statistics[1], summary(cor_g_F)$quantiles[1], summary(cor_g_F)$qu
  cbind(summary(cor_g_M)$statistics[1], summary(cor_g_M)$quantiles[1], summary(cor_g_M)$qu
)
plot(c(1, 2) ~ cor.est[, 1], xlim = c(0, 1.5), ylim = c(0, 2.5), xlab = "", ylab = "", col
segments(y0 = 1, x0 = cor.est[1, 2], y1 = 1, x1 = cor.est[1, 3], col = c("red"), lwd = 2)
segments(y0 = 2, x0 = cor.est[2, 2], y1 = 2, x1 = cor.est[2, 3], col = c("blue"), lwd = 2)
mtext("Covariance (\u00b1CI)", side = 1, las = 1, adj = 0.4, line = 3, cex = 1.6)
axis(2, at = 1, labels = c("female"), las = 3, cex.axis = 1.6)
axis(2, at = 2, labels = c("male"), las = 3, cex.axis = 1.6)
```



Here a simple plot of the sex-specific genetic correlation using the BLUPs

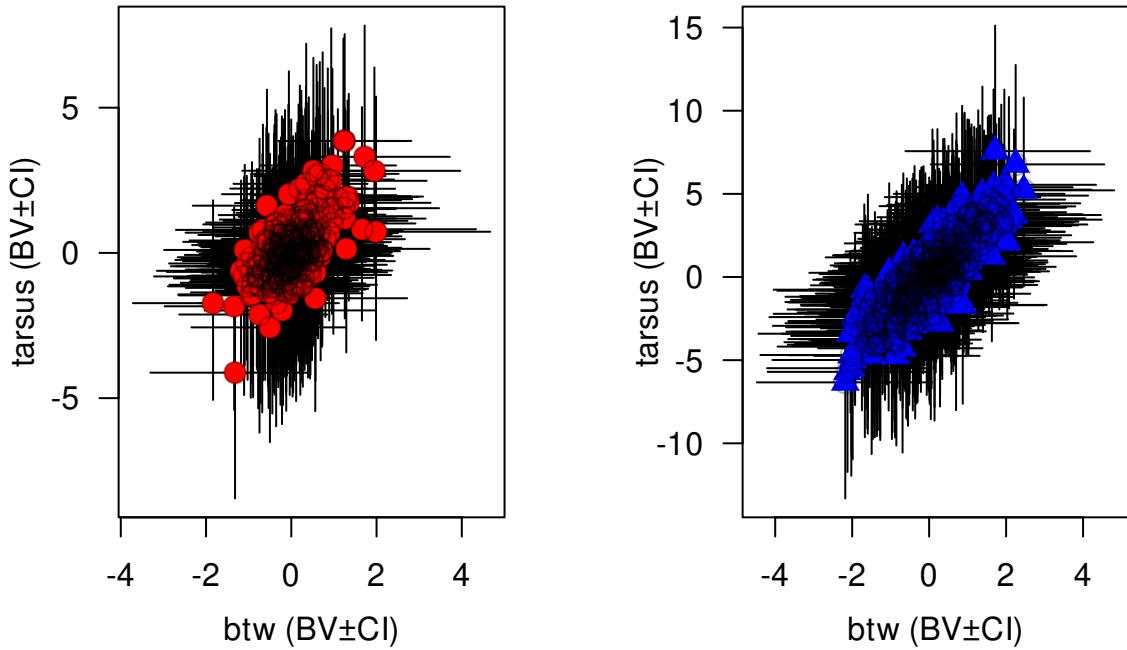
```

bls_m2.4 <- ranef(brms_m2.4)$animal
bl_m2.4 <- as.data.frame(abind::abind(lapply(1:dim(bls_m2.4)[3], function(x) bls_m2.4[, c
colnames(bl_m2.4) <- paste0(rep(dimnames(bls_m2.4)[[3]], each = 3), c("", "_lo", "_up"))
bl_m2.4$id <- rownames(bl_m2.4)
bl_m2.4$sex <- attr(dimnames(bls_m2.4)[[1]], "by")
FEM <- subset(bl_m2.4, sex == "1")
MAL <- subset(bl_m2.4, sex == "2")

#
par(mfrow = c(1, 2))
plot(tarsus_Intercept ~ bwt_Intercept, FEM,
      xlab = "", ylab = "",
      xlim = c(min(FEM$bwt_Intercept_lo), max(FEM$bwt_Intercept_up)),
      ylim = c(min(FEM$tarsus_Intercept_lo), max(FEM$tarsus_Intercept_up)),
      las = 1.2, type = "n")
)
segments(
  x0 = FEM$bwt_Intercept, y0 = FEM$tarsus_Intercept_lo,
  x1 = FEM$bwt_Intercept, y1 = FEM$tarsus_Intercept_up,
  col = "black"
)

```

```
)  
segments(  
  x0 = FEM$bwt_Intercept_lo, y0 = FEM$tarsus_Intercept,  
  x1 = FEM$bwt_Intercept_up, y1 = FEM$tarsus_Intercept,  
  col = "black"  
)  
points(tarsus_Intercept ~ bwt_Intercept, FEM, pch = 16, col = "red", cex = 1.5)  
points(tarsus_Intercept ~ bwt_Intercept, FEM, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.  
mtext("btw (BV±CI)", side = 1, line = 2.4)  
mtext("tarsus (BV±CI)", side = 2, line = 2, las = 3)  
#  
plot(tarsus_Intercept ~ bwt_Intercept, MAL,  
  xlab = "", ylab = "",  
  xlim = c(min(MAL$bwt_Intercept_lo), max(MAL$bwt_Intercept_up)),  
  ylim = c(min(MAL$tarsus_Intercept_lo), max(MAL$tarsus_Intercept_up)),  
  las = 1.2, type = "n")  
)  
segments(  
  x0 = MAL$bwt_Intercept, y0 = MAL$tarsus_Intercept_lo,  
  x1 = MAL$bwt_Intercept, y1 = MAL$tarsus_Intercept_up, col = "black")  
)  
segments(  
  x0 = MAL$bwt_Intercept_lo, y0 = MAL$tarsus_Intercept,  
  x1 = MAL$bwt_Intercept_up, y1 = MAL$tarsus_Intercept, col = "black")  
)  
points(tarsus_Intercept ~ bwt_Intercept, MAL, pch = 17, col = "blue", cex = 1.5)  
points(tarsus_Intercept ~ bwt_Intercept, MAL, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.  
mtext("btw (BV±CI)", side = 1, line = 2.4)  
mtext("tarsus (BV±CI)", side = 2, line = 2, las = 3)
```



3.5.4 Between groups (co)variances and the B-matrix

Animal models are amazing model. With different group within a population, it is also possible to estimate how much the different groups shared the same genetic via the cross-group genetic covariance. This covariance is essential to understand ontogenetic or sexual conflict, which can constraint or enhanced response to evolution. As an example, we estimate the cross-sex genetic correlation $r_{\{fm\}}$

It is important to keep in mind the covariance matrix at the residual level is zero and it is important to avoid estimating the cross-sex residual covariance because no individual switched sex during the experiment.

Note: the way of partitionning variance per sex is a bit different then the previous code “,by=sex”. This code is faster and also easier to understand. Note, it is possible to play with the | or || to estimate or not covariance between sexes.

```
bf_bwt_5 <- bf(
  bwt ~ 1 + sex + (0 + sex | a | gr(animal, cov = Amat)) + (0 + sex | b | mother) + (0 + sex | b | mother)
  sigma ~ sex - 1
)
bf_tarsus_5 <- bf(
  tarsus ~ 1 + sex + (0 + sex | a | gr(animal, cov = Amat)) + (0 + sex | b | mother) + (0 + sex | b | mother)
  sigma ~ sex - 1
```

```
)
brms_m2.5 <- brm(
  bf_bwt_5 + bf_tarsus_5 + set_rescor(TRUE),
  data = gryphon,
  data2 = list(Amat = Amat),
  chains = 2, cores = 2, iter = 1000
)
save(brms_m2.5, file = "data/brms_m2_5.rda")
```

Again we have provided the data from one such run. It can be accessed using the code:

```
load("data/brms_m2_5.rda")
summary(brms_m2.5)
```

```
## Warning: Parts of the model have not converged (some Rhats are > 1.05). Be
## careful when analysing the results! We recommend running more iterations and/or
## setting stronger priors.

## Warning: There were 45 divergent transitions after warmup. Increasing
## adapt_delta above 0.8 may help. See http://mc-stan.org/misc/
## warnings.html#divergent-transitions-after-warmup

## Family: MV(gaussian, gaussian)
##   Links: mu = identity; sigma = log
##          mu = identity; sigma = log
## Formula: bwt ~ 1 + sex + (0 + sex | a | gr(animal, cov = Amat)) + (0 + sex | b | mother
##          sigma ~ sex - 1
##          tarsus ~ 1 + sex + (0 + sex | a | gr(animal, cov = Amat)) + (0 + sex | b | mother
##          sigma ~ sex - 1
## Data: gryphon (Number of observations: 683)
## Draws: 2 chains, each with iter = 1000; warmup = 500; thin = 1;
##        total post-warmup draws = 1000
##
## Group-Level Effects:
## ~animal (Number of levels: 683)
##                               Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS
## sd(bwt_sex1)                1.26     0.30    0.63    1.73 1.06      21
## sd(bwt_sex2)                1.08     0.42    0.20    1.77 1.08      18
## sd(tarsus_sex1)              2.26     0.72    0.61    3.57 1.04      40
## sd(tarsus_sex2)              2.74     1.05    0.61    4.47 1.13      12
## cor(bwt_sex1,bwt_sex2)       0.48     0.29   -0.24    0.87 1.02      84
## cor(bwt_sex1,tarsus_sex1)    0.57     0.25   -0.07    0.89 1.14      10
## cor(bwt_sex2,tarsus_sex1)    0.38     0.38   -0.53    0.91 1.25       7
## cor(bwt_sex1,tarsus_sex2)    0.17     0.31   -0.49    0.75 1.04      60
## cor(bwt_sex2,tarsus_sex2)    0.52     0.33   -0.37    0.87 1.20       8
```

```

## cor(tarsus_sex1,tarsus_sex2)      0.44    0.29    -0.30     0.87 1.03    47
##                                         Tail_ESS
## sd(bwt_sex1)                      104
## sd(bwt_sex2)                      25
## sd(tarsus_sex1)                   99
## sd(tarsus_sex2)                   42
## cor(bwt_sex1,bwt_sex2)            112
## cor(bwt_sex1,tarsus_sex1)          145
## cor(bwt_sex2,tarsus_sex1)          67
## cor(bwt_sex1,tarsus_sex2)          94
## cor(bwt_sex2,tarsus_sex2)          50
## cor(tarsus_sex1,tarsus_sex2)       44
##
## ~byear (Number of levels: 34)
##                                         Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS
## sd(bwt_sex1)                      0.80    0.16    0.53    1.16 1.00    394
## sd(bwt_sex2)                      1.14    0.19    0.81    1.55 1.01    358
## sd(tarsus_sex1)                   2.23    0.46    1.50    3.18 1.01    297
## sd(tarsus_sex2)                   2.34    0.49    1.56    3.41 1.01    229
## cor(bwt_sex1,bwt_sex2)            0.74    0.15    0.35    0.96 1.01    266
## cor(bwt_sex1,tarsus_sex1)          -0.11   0.24    -0.55   0.35 1.01    190
## cor(bwt_sex2,tarsus_sex1)          -0.39   0.20    -0.73   0.00 1.01    410
## cor(bwt_sex1,tarsus_sex2)          0.29    0.23    -0.17   0.71 1.00    256
## cor(bwt_sex2,tarsus_sex2)          0.29    0.21    -0.16   0.66 1.01    327
## cor(tarsus_sex1,tarsus_sex2)       0.52    0.19    0.12    0.84 1.00    285
##                                         Tail_ESS
## sd(bwt_sex1)                      619
## sd(bwt_sex2)                      653
## sd(tarsus_sex1)                   559
## sd(tarsus_sex2)                   239
## cor(bwt_sex1,bwt_sex2)            433
## cor(bwt_sex1,tarsus_sex1)          603
## cor(bwt_sex2,tarsus_sex1)          656
## cor(bwt_sex1,tarsus_sex2)          319
## cor(bwt_sex2,tarsus_sex2)          474
## cor(tarsus_sex1,tarsus_sex2)       600
##
## ~mother (Number of levels: 352)
##                                         Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS
## sd(bwt_sex1)                      1.08    0.15    0.79    1.39 1.01    281
## sd(bwt_sex2)                      1.33    0.15    1.03    1.62 1.00    233
## sd(tarsus_sex1)                   2.21    0.40    1.36    2.95 1.01     72
## sd(tarsus_sex2)                   2.31    0.49    1.38    3.34 1.05     50
## cor(bwt_sex1,bwt_sex2)            0.83    0.11    0.57    0.98 1.01     68
## cor(bwt_sex1,tarsus_sex1)          -0.50   0.24    -0.91   -0.07 1.01     57
## cor(bwt_sex2,tarsus_sex1)          -0.64   0.17    -0.93   -0.28 1.06     50
## cor(bwt_sex1,tarsus_sex2)          -0.51   0.22    -0.88   -0.08 1.05     67

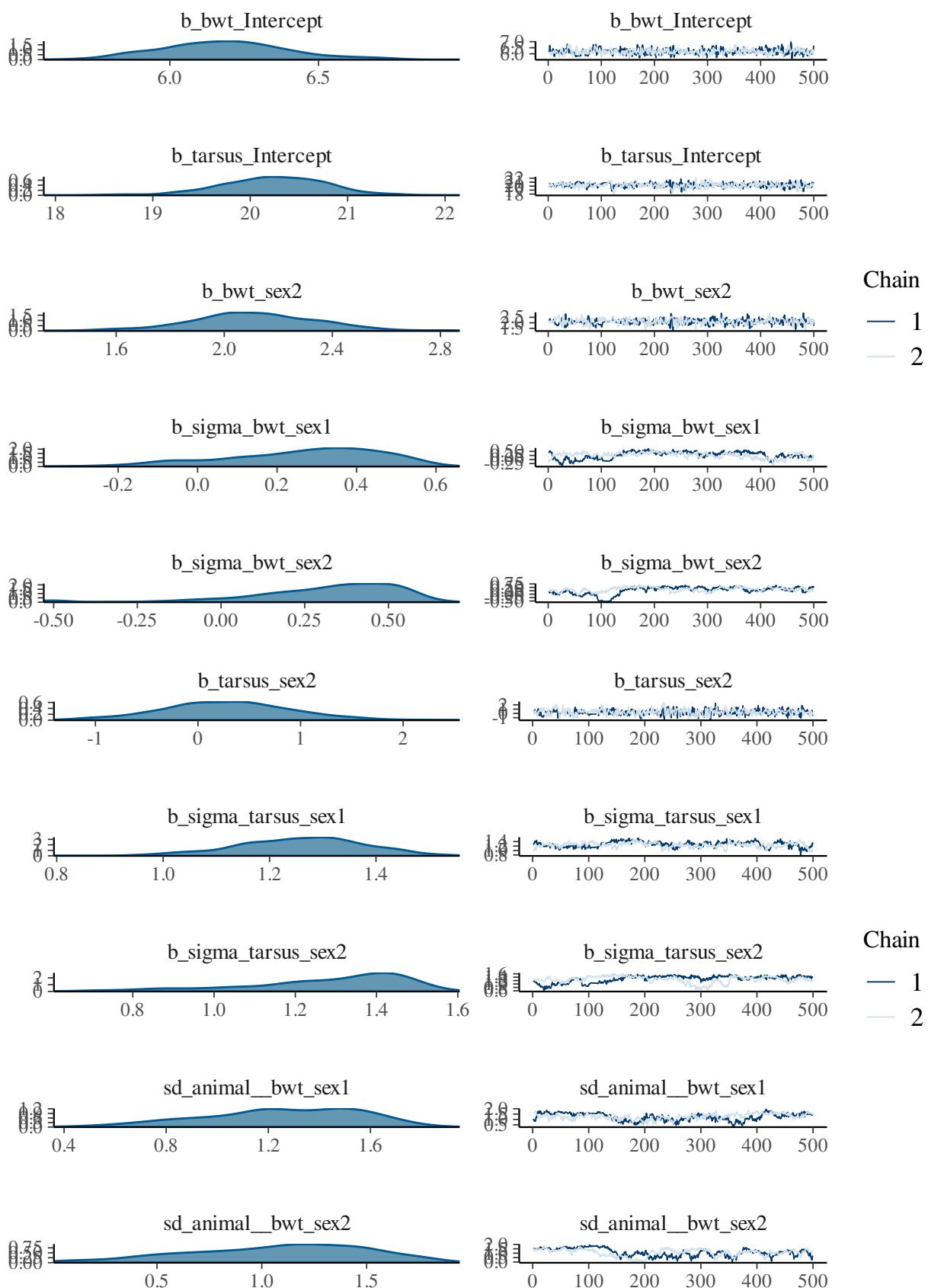
```

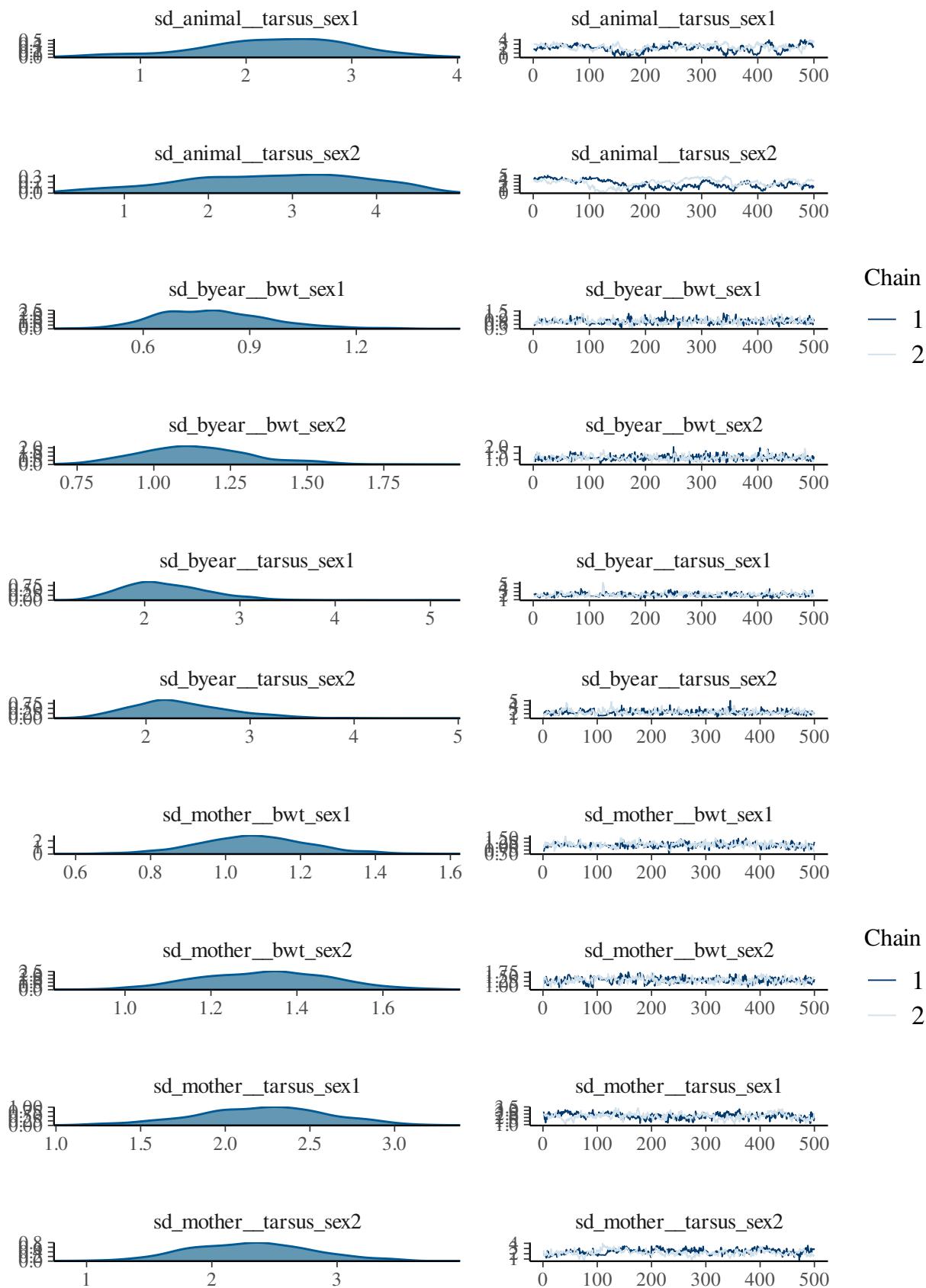
```

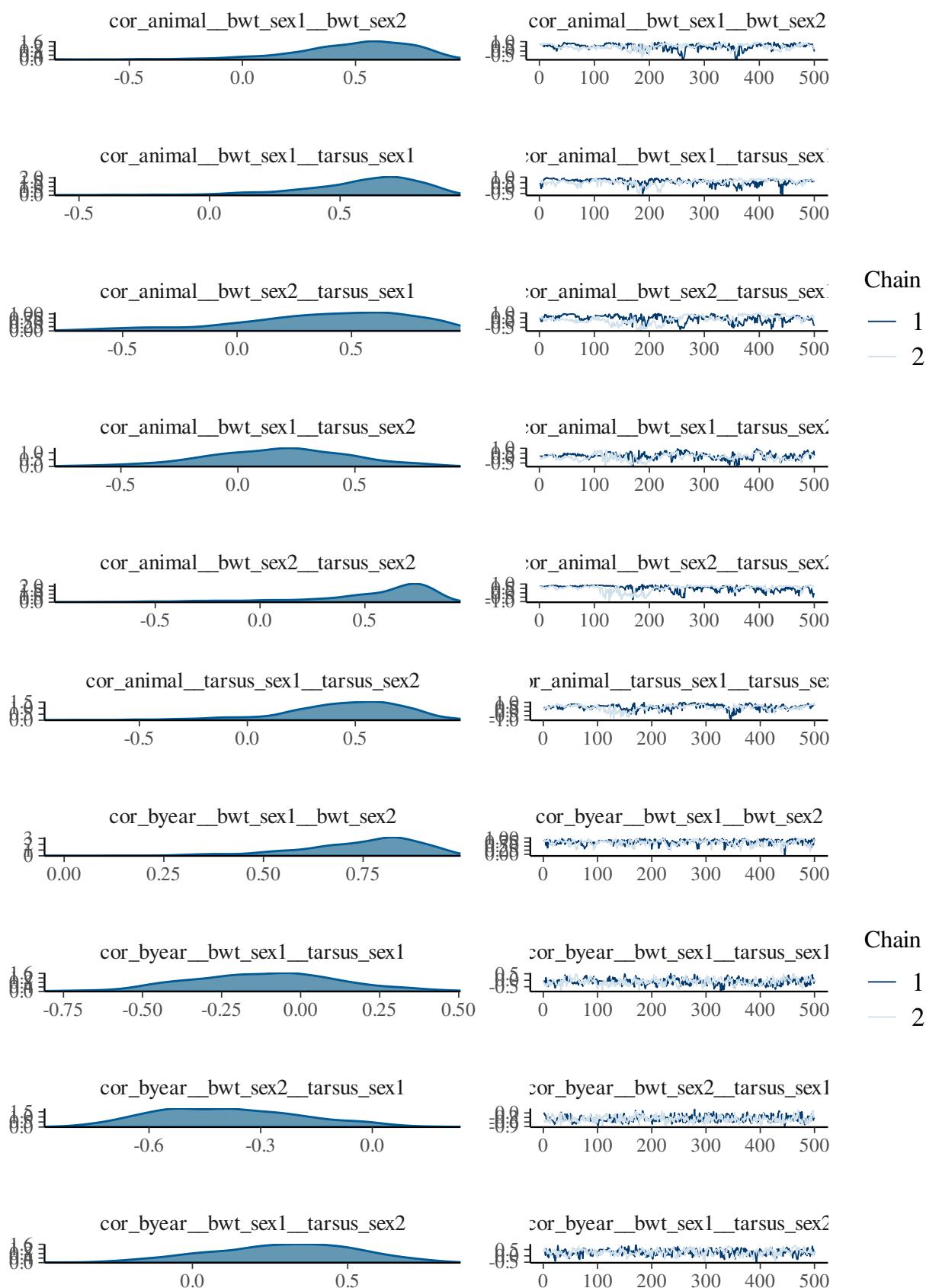
## cor(bwt_sex2,tarsus_sex2)      -0.36      0.26     -0.84      0.11 1.05      54
## cor(tarsus_sex1,tarsus_sex2)   0.72      0.16      0.37      0.95 1.01     249
##                                     Tail_ESS
## sd(bwt_sex1)                  518
## sd(bwt_sex2)                  585
## sd(tarsus_sex1)                224
## sd(tarsus_sex2)                191
## cor(bwt_sex1,bwt_sex2)        296
## cor(bwt_sex1,tarsus_sex1)     156
## cor(bwt_sex2,tarsus_sex1)     268
## cor(bwt_sex1,tarsus_sex2)     124
## cor(bwt_sex2,tarsus_sex2)     297
## cor(tarsus_sex1,tarsus_sex2)  596
##
## Population-Level Effects:
##             Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## bwt_Intercept      6.18     0.22    5.81    6.66 1.01     421     413
## tarsus_Intercept   20.24    0.55   19.12   21.25 1.01     497     527
## bwt_sex2           2.11     0.24    1.61    2.57 1.01     541     569
## sigma_bwt_sex1    0.27     0.20   -0.16    0.56 1.08      18      66
## sigma_bwt_sex2    0.31     0.23   -0.35    0.59 1.11      18      22
## tarsus_sex2         0.29     0.63   -0.96    1.56 1.00     490     535
## sigma_tarsus_sex1  1.26     0.12    1.01    1.47 1.02      50     109
## sigma_tarsus_sex2  1.28     0.20    0.80    1.53 1.13      12      42
##
## Residual Correlations:
##             Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## rescor(bwt,tarsus)    0.88     0.05    0.71    0.95 1.32      5     33
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

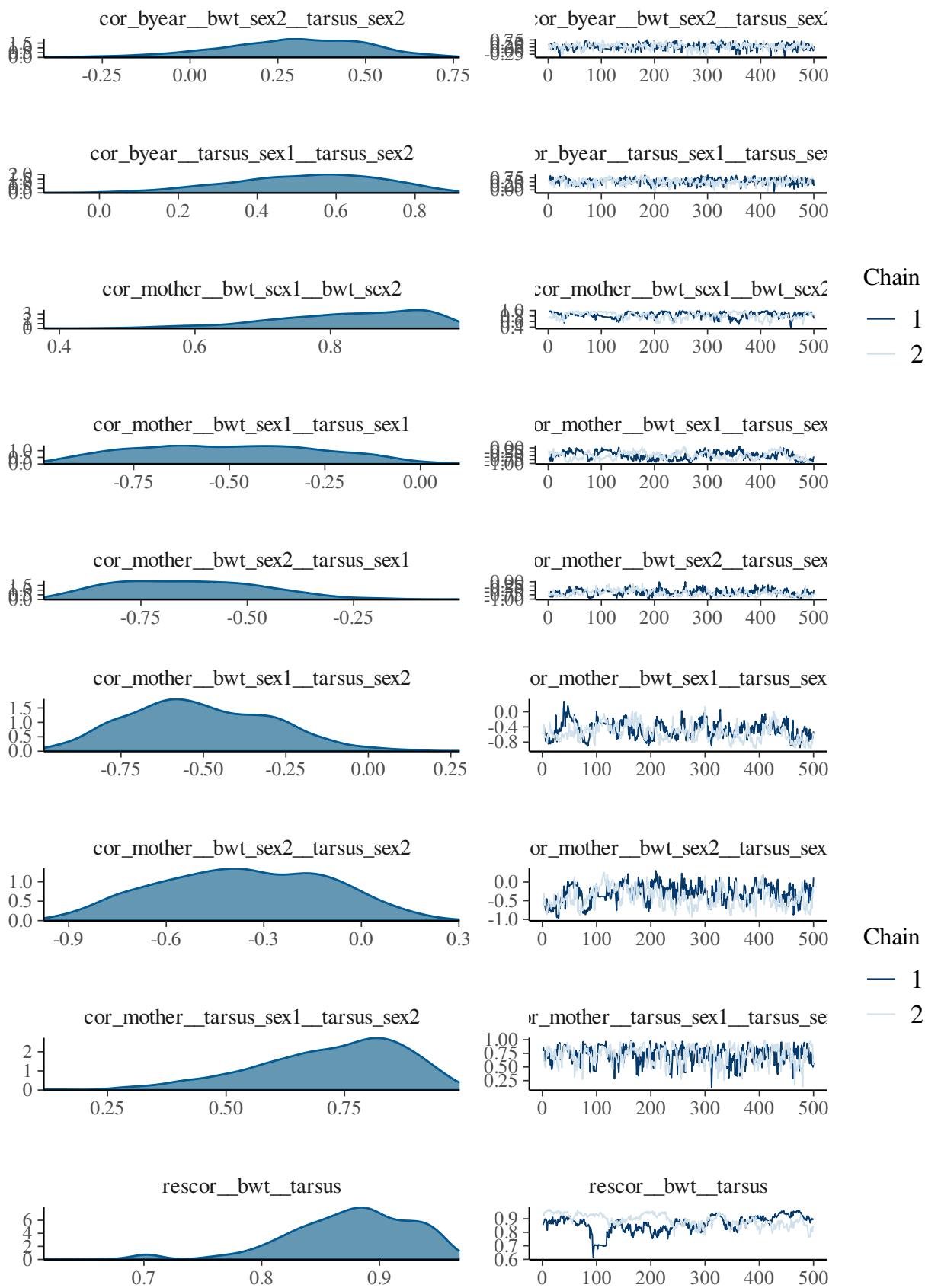
```

```
plot(brms_m2.5, ask = FALSE)
```









```
VarCorr(brms_m2.5)
```

```
## $animal
## $animal$sd
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.256797 0.3027123 0.6293789 1.733715
## bwt_sex2    1.077842 0.4168427 0.1958640 1.767136
## tarsus_sex1 2.259727 0.7239967 0.6135135 3.567762
## tarsus_sex2 2.744785 1.0537686 0.6051135 4.468542
##
## $animal$cor
## , , bwt_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.0000000 0.0000000 1.00000000 1.0000000
## bwt_sex2    0.4823461 0.2872049 -0.23539931 0.8699222
## tarsus_sex1 0.5711397 0.2473353 -0.06616842 0.8936581
## tarsus_sex2 0.1685774 0.3102320 -0.48914110 0.7539686
##
## , , bwt_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    0.4823461 0.2872049 -0.2353993 0.8699222
## bwt_sex2    1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_sex1 0.3772862 0.3765302 -0.5312556 0.9085651
## tarsus_sex2 0.5246223 0.3336679 -0.3678213 0.8692250
##
## , , tarsus_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    0.5711397 0.2473353 -0.06616842 0.8936581
## bwt_sex2    0.3772862 0.3765302 -0.53125561 0.9085651
## tarsus_sex1 1.0000000 0.0000000 1.00000000 1.0000000
## tarsus_sex2 0.4401433 0.2929178 -0.29616405 0.8720453
##
## , , tarsus_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    0.1685774 0.3102320 -0.4891411 0.7539686
## bwt_sex2    0.5246223 0.3336679 -0.3678213 0.8692250
## tarsus_sex1 0.4401433 0.2929178 -0.2961641 0.8720453
## tarsus_sex2 1.0000000 0.0000000 1.0000000 1.0000000
##
## 
## $animal$cov
```

```

## , , bwt_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.6710810 0.7317405  0.39611874 3.005768
## bwt_sex2    0.7428820 0.5579922 -0.10600955 2.049110
## tarsus_sex1 1.8733421 1.2159805 -0.05815667 4.340959
## tarsus_sex2 0.6471034 1.0646022 -1.25297349 2.930019
##
## , , bwt_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    0.742882 0.5579922 -0.10600955 2.049110
## bwt_sex2    1.335327 0.8557561  0.03836336 3.122771
## tarsus_sex1 1.105102 1.0991534 -0.68104784 3.388813
## tarsus_sex2 2.171388 1.8198507 -0.18718074 5.946047
##
## , , tarsus_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.873342 1.215980 -0.05815667 4.340959
## bwt_sex2    1.105102 1.099153 -0.68104784 3.388813
## tarsus_sex1 5.630014 3.143741  0.37639882 12.728924
## tarsus_sex2 3.150235 2.476892 -0.67170548 8.755673
##
## , , tarsus_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    0.6471034 1.064602 -1.2529735 2.930019
## bwt_sex2    2.1713876 1.819851 -0.1871807 5.946047
## tarsus_sex1 3.1502347 2.476892 -0.6717055 8.755673
## tarsus_sex2 8.6431609 5.649764  0.3661935 19.967865
##
## 
## 
## 
## $byear
## $byear$sd
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    0.7989572 0.1620838  0.5318383 1.156162
## bwt_sex2    1.1420876 0.1912205  0.8090083 1.549360
## tarsus_sex1 2.2286834 0.4609182  1.4995560 3.183107
## tarsus_sex2 2.3428101 0.4941316  1.5596304 3.405106
##
## 
## $byear$cor
## , , bwt_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.0000000 0.0000000  1.0000000 1.0000000

```

```
## bwt_sex2      0.7404024 0.1542690  0.3534945 0.9558481
## tarsus_sex1 -0.1137836 0.2357386 -0.5534464 0.3543949
## tarsus_sex2  0.2922708 0.2345550 -0.1749420 0.7113421
##
## , , bwt_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     0.7404024 0.1542690  0.3534945 0.955848060
## bwt_sex2     1.0000000 0.0000000  1.0000000 1.000000000
## tarsus_sex1 -0.3874250 0.1987378 -0.7323699 0.004258363
## tarsus_sex2  0.2937354 0.2059346 -0.1577533 0.659531338
##
## , , tarsus_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    -0.1137836 0.2357386 -0.5534464 0.354394929
## bwt_sex2    -0.3874250 0.1987378 -0.7323699 0.004258363
## tarsus_sex1  1.0000000 0.0000000  1.0000000 1.000000000
## tarsus_sex2  0.5226217 0.1897518  0.1183176 0.839911077
##
## , , tarsus_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     0.2922708 0.2345550 -0.1749420 0.7113421
## bwt_sex2     0.2937354 0.2059346 -0.1577533 0.6595313
## tarsus_sex1  0.5226217 0.1897518  0.1183176 0.8399111
## tarsus_sex2  1.0000000 0.0000000  1.0000000 1.0000000
##
## 
## $byear$cov
## , , bwt_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     0.6645776 0.2761781  0.2828520 1.3367115
## bwt_sex2     0.6843681 0.2597687  0.2749460 1.2955568
## tarsus_sex1 -0.1581409 0.4589758 -1.0281983 0.8339776
## tarsus_sex2  0.5456796 0.5013040 -0.3608892 1.5951028
##
## , , bwt_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     0.6843681 0.2597687  0.2749460 1.295556807
## bwt_sex2     1.3408929 0.4593417  0.6544951 2.400516888
## tarsus_sex1 -1.0167438 0.6693667 -2.4775184 0.009371017
## tarsus_sex2  0.8646682 0.7242669 -0.3384814 2.623722863
##
## , , tarsus_sex1
```

```

##                                     Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     -0.1581409 0.4589758 -1.0281983  0.833977585
## bwt_sex2     -1.0167438 0.6693667 -2.4775184  0.009371017
## tarsus_sex1  5.1792626 2.3047474  2.2486683 10.132170288
## tarsus_sex2  2.7818157 1.5318128  0.5297591  5.970346660
##
## , , tarsus_sex2
##
##                                     Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     0.5456796 0.5013040 -0.3608892  1.595103
## bwt_sex2     0.8646682 0.7242669 -0.3384814  2.623723
## tarsus_sex1  2.7818157 1.5318128  0.5297591  5.970347
## tarsus_sex2  5.7326811 2.5639312  2.4324504 11.594758
##
## 
## 
## 
## $mother
## $mother$sd
##                                     Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     1.076447 0.1526711 0.7866154 1.390273
## bwt_sex2     1.325206 0.1540539 1.0256350 1.621175
## tarsus_sex1  2.214033 0.3976449 1.3647309 2.946910
## tarsus_sex2  2.310902 0.4940802 1.3795156 3.339175
##
## $mother$cor
## , , bwt_sex1
##
##                                     Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     1.0000000 0.0000000 1.0000000 1.00000000
## bwt_sex2     0.8260360 0.1110784 0.5676601 0.97826511
## tarsus_sex1 -0.5024557 0.2385053 -0.9137423 -0.06798800
## tarsus_sex2 -0.5073494 0.2162931 -0.8806394 -0.07677131
##
## , , bwt_sex2
##
##                                     Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     0.8260360 0.1110784 0.5676601 0.9782651
## bwt_sex2     1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_sex1 -0.6373889 0.1744852 -0.9272366 -0.2775698
## tarsus_sex2 -0.3593672 0.2577845 -0.8364419  0.1135921
##
## , , tarsus_sex1
##
##                                     Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1     -0.5024557 0.2385053 -0.9137423 -0.0679880
## bwt_sex2     -0.6373889 0.1744852 -0.9272366 -0.2775698

```

```

## tarsus_sex1 1.0000000 0.0000000 1.0000000 1.0000000
## tarsus_sex2 0.7206209 0.1561300 0.3688228 0.9518425
##
## , , tarsus_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    -0.5073494 0.2162931 -0.8806394 -0.07677131
## bwt_sex2    -0.3593672 0.2577845 -0.8364419  0.11359208
## tarsus_sex1  0.7206209 0.1561300 0.3688228 0.95184250
## tarsus_sex2  1.0000000 0.0000000 1.0000000 1.00000000
##
##
## $mother$cov
## , , bwt_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.182023 0.3330334  0.6187659  1.9328578
## bwt_sex2    1.179468 0.2744898  0.6821359  1.7479435
## tarsus_sex1 -1.110068 0.4557029 -1.9379754 -0.2103052
## tarsus_sex2 -1.223308 0.5582766 -2.3558658 -0.1900476
##
## , , bwt_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    1.1794683 0.2744898  0.6821359  1.7479435
## bwt_sex2    1.7798788 0.4088241  1.0519272  2.6282093
## tarsus_sex1 -1.8437371 0.5678929 -2.9129058 -0.7723013
## tarsus_sex2 -0.9438083 0.6393327 -2.0044262  0.4581811
##
## , , tarsus_sex1
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    -1.110068 0.4557029 -1.937975 -0.2103052
## bwt_sex2    -1.843737 0.5678929 -2.912906 -0.7723013
## tarsus_sex1  5.059904 1.7495725  1.862491  8.6842757
## tarsus_sex2  3.692391 1.3309441  1.432767  6.5202798
##
## , , tarsus_sex2
##
##           Estimate Est.Error      Q2.5      Q97.5
## bwt_sex1    -1.2233079 0.5582766 -2.355866 -0.1900476
## bwt_sex2    -0.9438083 0.6393327 -2.004426  0.4581811
## tarsus_sex1  3.6923914 1.3309441  1.432767  6.5202798
## tarsus_sex2  5.5841373 2.3245515  1.903063 11.1500916

```

The cross-sex genetic correlation can estimate form the output of the model. For tarsus length at fledging, sexes shared a lot of genetic variance which is commun for a trait with low sexual

dimorphism. If the selection is antagonistic between males and females, sexes can not evolve freely from the other sexes and a intralocus sexual conflict can appeared.

```
cross_sex.cor.btw <- as.mcmc((VarCorr(brms_m2.5, summary = FALSE)$animal$cor[, 1, 2]))
cross_sex.cor.tarsus <- as.mcmc((VarCorr(brms_m2.5, summary = FALSE)$animal$cor[, 3, 4]))
```

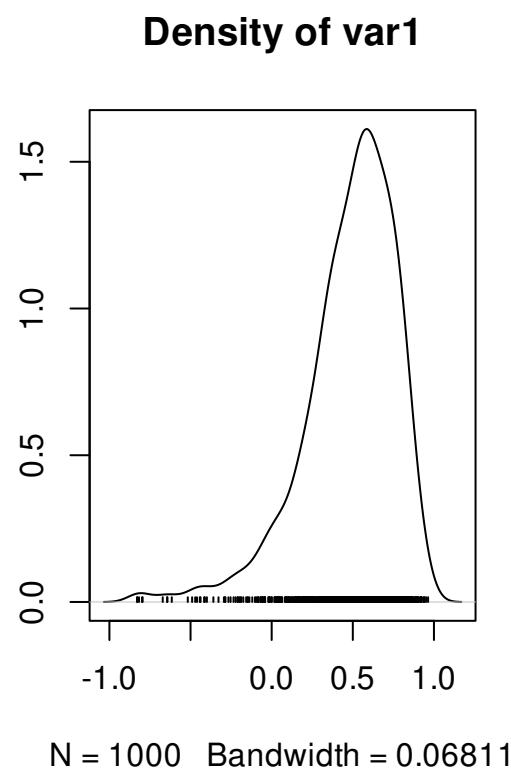
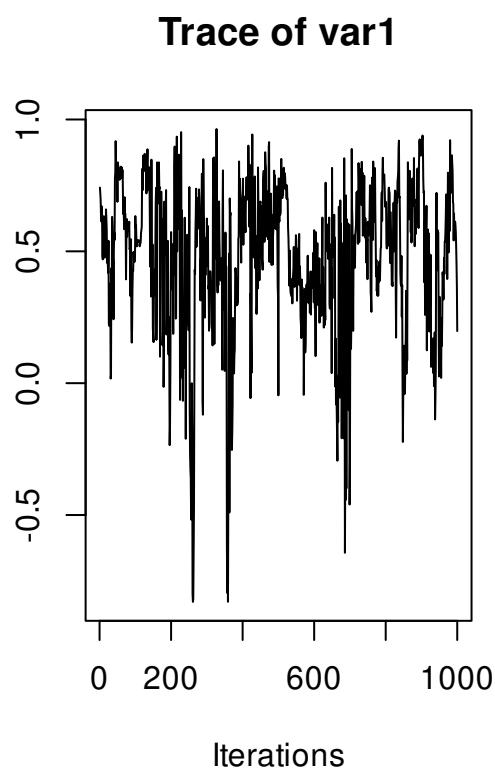
```
summary(cross_sex.cor.btw)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD     Naive SE Time-series SE
##       0.482346   0.287205   0.009082   0.032430
##
## 2. Quantiles for each variable:
##
##    2.5%     25%     50%     75%   97.5%
## -0.2354  0.3433  0.5365  0.6861  0.8699
```

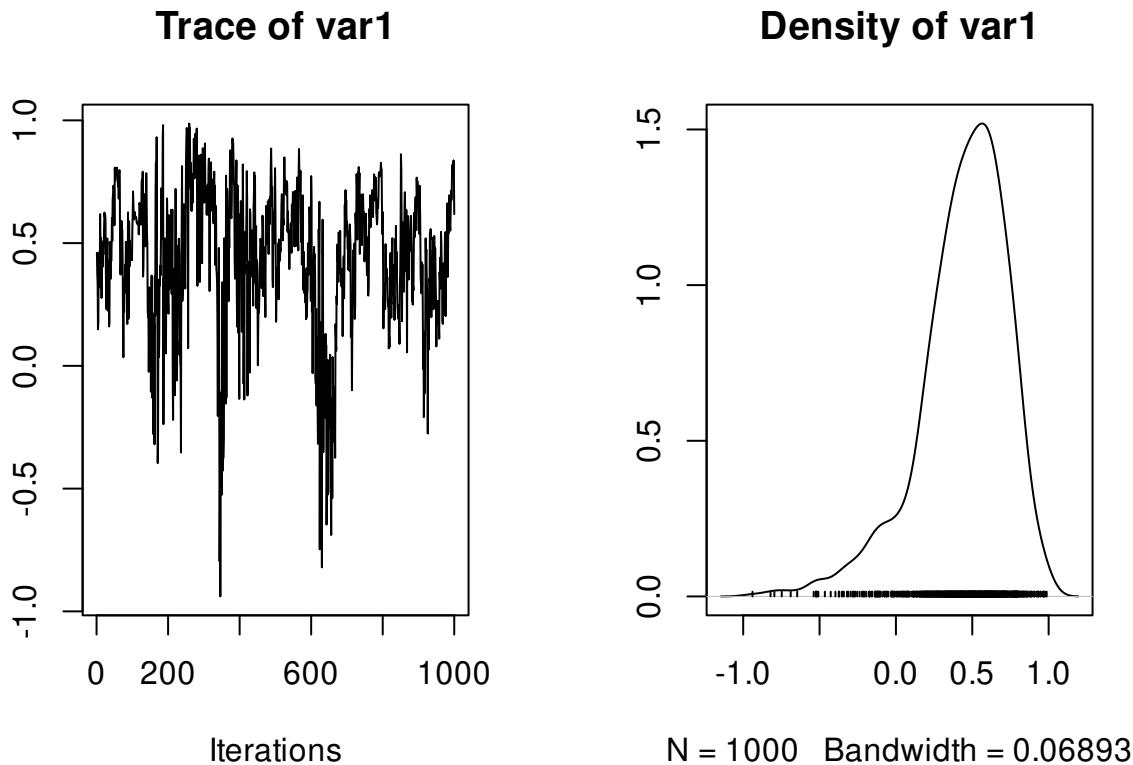
```
summary(cross_sex.cor.tarsus)
```

```
##
## Iterations = 1:1000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 1000
##
## 1. Empirical mean and standard deviation for each variable,
##    plus standard error of the mean:
##
##           Mean        SD     Naive SE Time-series SE
##       0.440143   0.292918   0.009263   0.048863
##
## 2. Quantiles for each variable:
##
##    2.5%     25%     50%     75%   97.5%
## -0.2962  0.2950  0.4846  0.6419  0.8720
```

```
plot(cross_sex.cor.btw)
```



```
plot(cross_sex.cor.tarsus)
```



Here, some simple code to extract the BLUP.

```
bls_m2.5 <- ranef(brms_m2.5)$animal
bl_m2.5 <- as.data.frame(abind::abind(lapply(1:4, function(x) bls_m2.5[, c(1, 3, 4), x])))
colnames(bl_m2.5) <- paste0(rep(dimnames(bls_m2.5)[[3]], each = 3), c("", "_lo", "_up"))
bl_m2.5$id <- rownames(bl_m2.5)
```

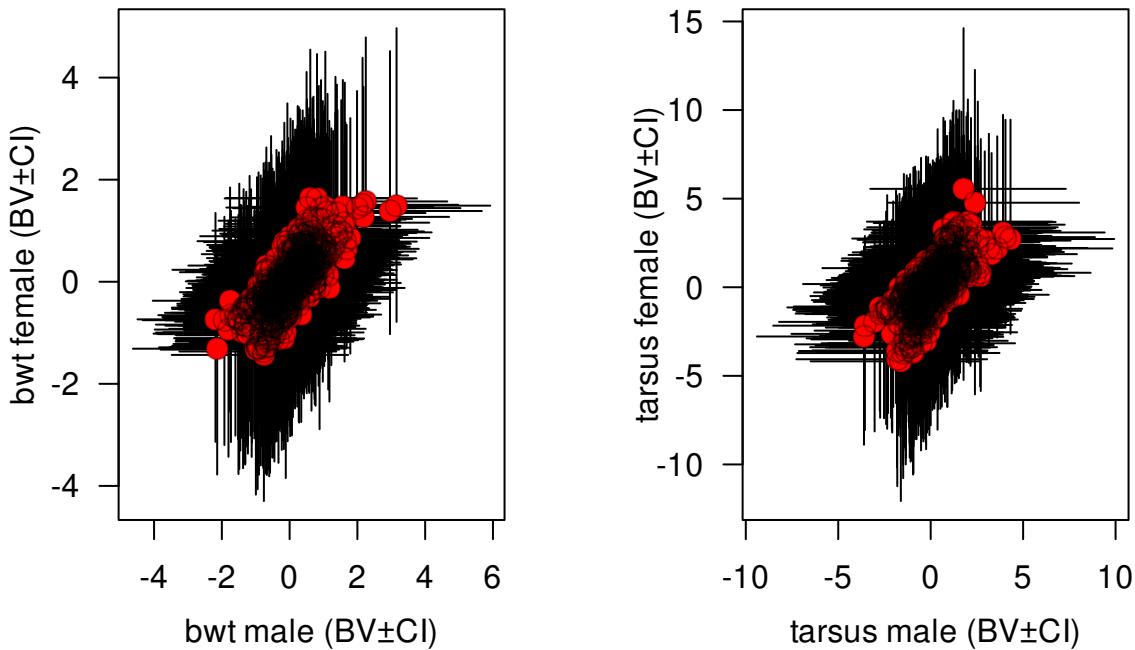
Here, some simple code to plot the cross-sex genetic correlation.

```
par(mfrow = c(1, 2))
plot(bwt_sex2 ~ bwt_sex1, bl_m2.5,
  xlab = "", ylab = "", las = 1.2, type = "n",
  xlim = c(min(bl_m2.5$bwt_sex1_lo), max(bl_m2.5$bwt_sex1_up)),
  ylim = c(min(bl_m2.5$bwt_sex2_lo), max(bl_m2.5$bwt_sex2_up)))
)
with(bl_m2.5, segments(x0 = bwt_sex1, y0 = bwt_sex2_lo, x1 = bwt_sex1, y1 = bwt_sex2_up,
  with(bl_m2.5, segments(x0 = bwt_sex1_lo, y0 = bwt_sex2, x1 = bwt_sex1_up, y1 = bwt_sex2,
  points(bwt_sex2 ~ bwt_sex1, bl_m2.5, pch = 16, col = "red", cex = 1.5)
  points(bwt_sex2 ~ bwt_sex1, bl_m2.5, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
  mtext("bwt male (BV±CI)", side = 1, line = 2.4)
  mtext("bwt female (BV±CI)", side = 2, line = 2, las = 3)
```

```

plot(tarsus_sex2 ~ tarsus_sex1, bl_m2.5,
  xlab = "", ylab = "", las = 1.2, type = "n",
  xlim = c(min(bl_m2.5$tarsus_sex1_lo), max(bl_m2.5$tarsus_sex1_up)),
  ylim = c(min(bl_m2.5$tarsus_sex2_lo), max(bl_m2.5$tarsus_sex2_up))
)
with(bl_m2.5, segments(x0 = tarsus_sex1, y0 = tarsus_sex2_lo, x1 = tarsus_sex1, y1 = tarsus_sex2_up))
with(bl_m2.5, segments(x0 = tarsus_sex1_lo, y0 = tarsus_sex2, x1 = tarsus_sex1_up, y1 = tarsus_sex2))
points(tarsus_sex2 ~ tarsus_sex1, bl_m2.5, pch = 16, col = "red", cex = 1.5)
points(tarsus_sex2 ~ tarsus_sex1, bl_m2.5, pch = 1, col = rgb(0, 0, 0, 0.3), cex = c(1.5))
mtext("tarsus male (BV±CI)", side = 1, line = 2.4)
mtext("tarsus female (BV±CI)", side = 2, line = 2, las = 3)

```



Within this model, we also have access to the rest of the B-matrix. Note, the cross-sex genetic correlation is just the diagonal of the B matrix. For now on, you can explore this matrix and estimate the cross-sex-cross-trait genetic correlation.

3.6 stan

to do

Chapitre 4

A repeated measures animal model

This tutorial will demonstrate how to run a univariate animal model for a trait with repeated observations using different R packages with an example data files provided.

4.1 Scenario and data

4.1.1 scenario

Since gryphons are iteroparous, multiple observations of reproductive traits are available for some individuals. Here we have repeated measures of lay date (measured in days after January 1) for individual females varying in age from 2 (age of sexual maturation) up until age 6. Not all females lay every year so the number of observations per female is variable (between 1 to 5). We want to know how repeatable the trait is, and (assuming it is repeatable) how heritable it is.

4.1.2 Data files

The pedigree file `gryphonped.csv` is that used in the preceding tutorials but we now use a new data file `gryphonRM.csv`. Columns correspond to individual identity (`animal`), birth year (`byear`), age in years (`age`), year of measurement (`year`) and lay date (`laydate`). Each row of the data file corresponds to a single phenotypic observation. Here the data is sorted by identity and then age so that the repeated observations on individuals are apparent. However this is not a requirement for analysis - data could equally be sorted by some other variable (*e.g.*, measurement year) or be in a random order.

```
str(gryphonRM)
```

```
## 'data.frame': 1607 obs. of 5 variables:
## $ animal : Factor w/ 469 levels "1","2","3","8",...: 1 1 1 1 1 2 2 2 3 3 ...
## $ byear  : Factor w/ 34 levels "968","970","971",...: 22 22 22 22 22 22 22 22 22 ...
## $ age    : Factor w/ 5 levels "2","3","4","5",...: 1 2 3 4 5 1 2 3 1 2 ...
## $ year   : Factor w/ 39 levels "970","971","972",...: 23 24 25 26 27 23 24 25 23 24 ...
## $ laydate: num 19 23 24 23 29 21 17 21 20 20 ...
```

```
summary(gryphonRM)
```

```
##      animal       byear      age      year    laydate
## 1      : 5  1000 : 109  2:308  1004 : 79 Min.   : 0.00
## 3      : 5  1001 : 98   3:322  1005 : 78 1st Qu.:20.00
## 9      : 5  999  : 86   4:339  1003 : 69 Median :24.00
## 17     : 5  1002 : 85   5:315  1006 : 64 Mean   :23.54
## 42     : 5  987  : 70   6:323  1002 : 60 3rd Qu.:27.00
## 50     : 5  989  : 66          988 : 54 Max.   :41.00
## (Other):1577 (Other):1093           (Other):1203
```

```
head(gryphonRM)
```

```
##   animal byear age year laydate
## 1      1  990  2  992     19
## 2      1  990  3  993     23
## 3      1  990  4  994     24
## 4      1  990  5  995     23
## 5      1  990  6  996     29
## 6      2  990  2  992     21
```

4.2 Asreml-R

First we need to load the `asreml` library:

```
library(asreml)
```

4.2.1 Estimating repeatability

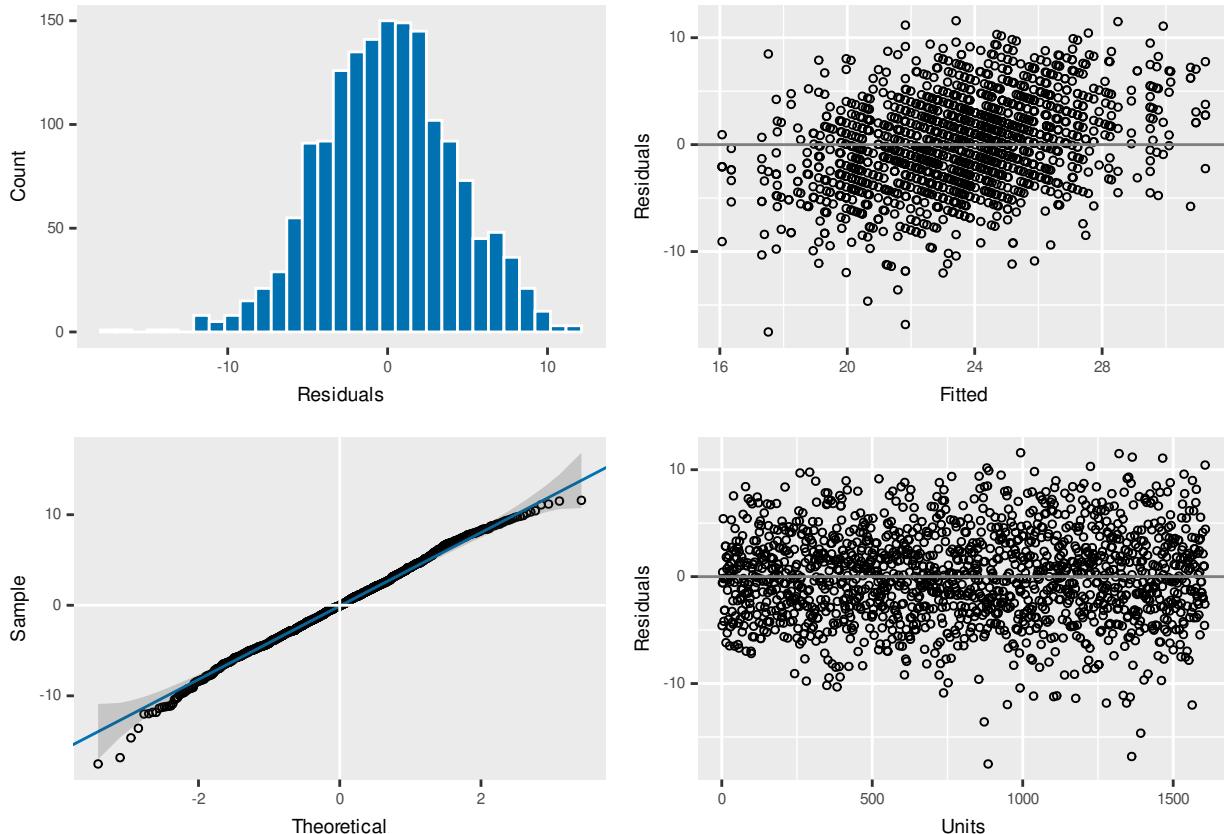
With repeated measures on individuals it is often of interest to see how repeatable a trait is. We can estimate the repeatability of a trait as the proportion of phenotypic variance V_P explained by individual variance V_{ind} ; $R = V_{ind}/V_P = V_{ind}/(V_{ind} + V_R)$.

```
modelv <- asreml(
  fixed = laydate ~ 1,
  random = ~animal,
  residual = ~ idv(units),
  data = gryphonRM,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:53 2022
```

```
##          LogLik     Sigma2      DF    wall      cpu
## 1 -10182.83 1.0 1606 16:56:53 0.0
## 2 -8266.10 1.0 1606 16:56:53 0.0
## 3 -6145.01 1.0 1606 16:56:53 0.0
## 4 -4651.57 1.0 1606 16:56:53 0.0
## 5 -3819.31 1.0 1606 16:56:53 0.0
## 6 -3554.22 1.0 1606 16:56:53 0.0
## 7 -3501.56 1.0 1606 16:56:53 0.0
## 8 -3497.58 1.0 1606 16:56:53 0.0
## 9 -3497.54 1.0 1606 16:56:53 0.0
## 10 -3497.54 1.0 1606 16:56:53 0.0
```

```
plot(modelv)
```



The model assumption seems correct, so we can look at the different estimates. Note that since we want to estimate the amount of variance explained by individual identity (rather than by additive genetic effects), we fit `animal` as a normal random effect and we don't associate it with the pedigree. Here, we also ask the model to remove any NA in `laydate`.

This model partitions the phenotypic variance in `laydate` as follows:

```
summary(modelv)$varcomp
```

```
##          component std.error   z.ratio bound %ch
```

The

```
## animal      11.08634 1.1794319  9.399728      P    0
## units!units 21.29643 0.8896196 23.938798      P    0
## units!R     1.00000        NA        NA      F    0
```

Between-individual (or among-individual) variance is given by the `animal` component, while the residual component (`units!units`) represents within-individual variance. Here then the repeatability of the trait can be determined by hand as 0.34 (*i.e.*, as $11.086/(11.086 + 21.296)$).

Mean lay date might change with age, so we could ask what the repeatability of lay date is after conditioning on age. This would be done by adding `age` into the model as a fixed effect.

```
modelw <- asreml(
  fixed = laydate ~ age,
  random = ~animal,
  residual = ~ idv(units),
  data = gryphonRM,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:54 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1      -8402.968      1.0 1602 16:56:54      0.0
## 2      -6912.361      1.0 1602 16:56:54      0.0
## 3      -5274.379      1.0 1602 16:56:54      0.0
## 4      -4143.634      1.0 1602 16:56:54      0.0
## 5      -3541.895      1.0 1602 16:56:54      0.0
## 6      -3372.909      1.0 1602 16:56:54      0.0
## 7      -3347.670      1.0 1602 16:56:54      0.0
## 8      -3346.655      1.0 1602 16:56:54      0.0
## 9      -3346.652      1.0 1602 16:56:54      0.0
```

```
summary(modelw)$varcomp
```

```
##           component std.error z.ratio bound %ch
## animal      12.28982  1.156115 10.63027      P    0
## units!units 16.37989  0.686619 23.85586      P    0
## units!R     1.00000        NA        NA      F    0
```

The repeatability of lay date, after accounting for age effects, is now estimated as 0.43 (*i.e.*, as $12.29/(12.29 + 16.38)$). So, just as we saw when estimating h^2 in Tutorial 1, the inclusion of fixed effects will alter the estimated effect size if we determine total phenotypic variance as the sum of the variance components. Thus, proper interpretation is vital.

```
summary(modelw, coef = TRUE)$coef.fixed
wald.asreml(modelw, ssType = "conditional", denDF = "numeric")
```

```

##          solution std.error z.ratio
## age_2        0.000000      NA       NA
## age_3      2.577777 0.3355253 7.682811
## age_4      4.247276 0.3309028 12.835418
## age_5      6.094490 0.3375537 18.054872
## age_6      3.132675 0.3371074  9.292811
## (Intercept) 20.305073 0.2899515 70.029214

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:54 2022
##          LogLik     Sigma2      DF    wall      cpu
## 1      -3346.652      1.0 1602 16:56:54      0.0
## 2      -3346.652      1.0 1602 16:56:54      0.0
## 3      -3346.652      1.0 1602 16:56:54      0.0

##
##          Df  denDF   F.inc   F.con Margin          Pr
## (Intercept) 1  460.2 14880.0 14880.0           0.000000e+00
## age         4 1225.3    88.7    88.7      A 2.89474e-66

```

Here age is modeled as a 5-level factor (specified using the function `as.factor()` at the beginning of the analysis). We could equally have fitted it as a continuous variable, in which case, given potential for a late life decline, we would probably also include a quadratic term. In addition, using `age` as continuous variable can help in saving some degree of freedom in the analysis.

4.2.2 Partitioning additive and permanent environment effects

Generally we expect that the repeatability will set the upper limit for heritability since among individual variation can be decomposed in the additive genetic variation and non additive genetic variation. In other word, the additive genetic variation is a subcomponent of the difference between individuals. Non-additive contributions to fixed among-individual differences are normally referred to as *permanent environment effects*. If a trait has repeated measures then it is necessary to model permanent environment effects in an animal model to prevent upward bias in V_A .

To illustrate it, we first fit the animal model:

```

gryphonped <- read.csv("data/gryphonped.csv")
gryphonped$id <- as.factor(gryphonped$id)
gryphonped$father <- as.factor(gryphonped$father)
gryphonped$mother <- as.factor(gryphonped$mother)

ainv <- ainverse(gryphonped)

modelx <- asreml(
  fixed = laydate ~ age,
  random = ~ vm(animal, ainv),
  residual = ~ idv(units),
  data = gryphonRM,

```

```

na.action = na.method(x = "omit", y = "omit")
)

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:54 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1     -8751.390      1.0  1602 16:56:54      0.0
## 2     -7169.205      1.0  1602 16:56:54      0.0
## 3     -5427.604      1.0  1602 16:56:54      0.0
## 4     -4219.598      1.0  1602 16:56:54      0.0
## 5     -3569.815      1.0  1602 16:56:54      0.0
## 6     -3382.341      1.0  1602 16:56:54      0.0
## 7     -3352.867      1.0  1602 16:56:54      0.0
## 8     -3351.565      1.0  1602 16:56:54      0.0
## 9     -3351.560      1.0  1602 16:56:54      0.0

```

Variance components are almost unchanged if we compare the previous model:

```
summary(modelx)$varcomp
```

```

##           component std.error   z.ratio bound %ch
## vm(animal, ainv) 13.91784  1.443968  9.638607    P  0
## units!units       16.84008  0.707365 23.806768    P  0
## units!R           1.00000        NA        NA      F  0

```

```
summary(modelw)$varcomp
```

```

##           component std.error   z.ratio bound %ch
## animal       12.28982  1.156115 10.63027    P  0
## units!units   16.37989  0.686619 23.85586    P  0
## units!R       1.00000        NA        NA      F  0

```

This suggests that most of the among-individual variance is – rightly or wrongly – being partitioned as V_A here. To instead obtain an unbiased estimate of V_A , we need to partition for both additive genetic *and* non-genetic sources of individual variation. We do it by fitting `animal` twice, once with a pedigree, and once without a pedigree (using `ide()`). Here, the command `ide` allow to create a second effect using a similar variable.

```

modely <- asreml(
  fixed = laydate ~ age,
  random = ~ vm(animal, ainv) + ide(animal),
  residual = ~ idv(units),
  data = gryphonRM,
  na.action = na.method(x = "omit", y = "omit")
)

```

```
## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:54 2022
##          LogLik      Sigma2      DF    wall    cpu
## 1     -7731.394      1.0  1602 16:56:54    0.0
## 2     -6426.548      1.0  1602 16:56:54    0.0
## 3     -4997.252      1.0  1602 16:56:54    0.0
## 4     -4018.486      1.0  1602 16:56:54    0.0
## 5     -3504.988      1.0  1602 16:56:54    0.0
## 6     -3363.160      1.0  1602 16:56:54    0.0
## 7     -3341.611      1.0  1602 16:56:54    0.0
## 8     -3340.682      1.0  1602 16:56:54    0.0
## 9     -3340.679      1.0  1602 16:56:54    0.0
```

```
summary(modely)$varcomp
```

	component	std.error	z.ratio	bound	%ch
## vm(animal, ainv)	4.876101	1.8087709	2.695809	P	0
## ide(animal)	7.400983	1.7280113	4.282948	P	0
## units!units	16.380188	0.6866189	23.856300	P	0
## units!R	1.000000	NA	NA	F	0

The estimate of V_A is now much lower since the additive and permanent environment effects are being properly separated. We can estimate h^2 and the repeatability from this model:

```
vpredict(modely, h2 ~ V1 / (V1 + V2 + V3))
```

	Estimate	SE
## h2	0.1701523	0.06073974

```
vpredict(modely, repeatability ~ (V1 + V2) / (V1 + V2 + V3))
```

	Estimate	SE
## repeatability	0.4284108	0.02741602

4.2.3 Adding additional effects and testing significance

Models of repeated measures can be extended to include other fixed or random effects. For example try including year of measurement (`year`) and birth year (`byear`) as random effects.

```
modelz <- asreml(
  fixed = laydate ~ age,
  random = ~ vm(animal, ainv) + ide(animal) +
    year + byear,
  residual = ~ idv(units),
  data = gryphonRM,
```

```

na.action = na.method(x = "omit", y = "omit")
)

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:54 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1    -4650.748      1.0  1602 16:56:54      0.0
## 2    -4088.264      1.0  1602 16:56:54      0.0
## 3    -3494.147      1.0  1602 16:56:54      0.0
## 4    -3127.161      1.0  1602 16:56:54      0.0 (1 restrained)
## 5    -2976.449      1.0  1602 16:56:54      0.0 (1 restrained)
## 6    -2955.785      1.0  1602 16:56:54      0.0 (1 restrained)
## 7    -2955.097      1.0  1602 16:56:54      0.0 (1 restrained)
## 8    -2955.095      1.0  1602 16:56:54      0.0 (1 restrained)
## 9    -2955.095      1.0  1602 16:56:54      0.0

```

```
summary(modelz)$varcomp
```

	component	std.error	z.ratio	bound	%ch
## byear	1.650876e-07	NA	NA	B	0
## year	7.938576e+00	1.9344619	4.103765	P	0
## vm(animal, ainv)	4.815136e+00	1.6682351	2.886365	P	0
## ide(animal)	8.433325e+00	1.5495778	5.442337	P	0
## units!units	7.795560e+00	0.3324411	23.449443	P	0
## units!R	1.000000e+00	NA	NA	F	0

This model will return additional variance components corresponding to variation in lay dates between years of measurement and between birth cohorts of females. V_{byear} is very low and B appeared which tell us that the model had fixed the variance as a boundary. If you compare this model to a reduced model with byear excluded the log-likelihood remains unchanged.

```

modelz_2 <- asreml(
  fixed = laydate ~ age,
  random = ~ vm(animal, ainv) + ide(animal) +
    year,
  residual = ~ idv(units),
  data = gryphonRM,
  na.action = na.method(x = "omit", y = "omit")
)

```

```

## Model fitted using the sigma parameterization.
## ASReml 4.1.0 Tue Nov 29 16:56:54 2022
##          LogLik      Sigma2      DF      wall      cpu
## 1    -4665.606      1.0  1602 16:56:54      0.0
## 2    -4097.928      1.0  1602 16:56:54      0.0

```

```
## 3 -3498.611 1.0 1602 16:56:54 0.0
## 4 -3128.789 1.0 1602 16:56:54 0.0
## 5 -2976.883 1.0 1602 16:56:54 0.0
## 6 -2955.806 1.0 1602 16:56:54 0.0
## 7 -2955.096 1.0 1602 16:56:54 0.0
## 8 -2955.095 1.0 1602 16:56:54 0.0
```

```
summary(modelz_2)$varcomp
```

	component	std.error	z.ratio	bound	%ch
## year	7.938576	1.9344829	4.103720	P	0
## vm(animal, ainv)	4.815137	1.6682366	2.886364	P	0
## ide(animal)	8.433324	1.5495828	5.442319	P	0
## units!units	7.795560	0.3324384	23.449637	P	0
## units!R	1.000000	NA	NA	F	0

```
modelz$loglik
```

```
## [1] -2955.095
```

```
modelz_2$loglik
```

```
## [1] -2955.095
```

```
1 - pchisq(2 * (modelz_2$loglik - modelz$loglik), 1)
```

```
## [1] 0.9990425
```

year effects could alternatively be included as fixed effects (try it!). This will reduce V_R and increase the estimates of heritability and repeatability, which must now be interpreted as proportions of phenotypic variance after conditioning on both age and year of measurement effects.

```
modelz_3 <- asreml(
  fixed = laydate ~ age + byear,
  random = ~ vm(animal, ainv) + ide(animal) +
    year,
  residual = ~ idv(units),
  data = gryphonRM,
  na.action = na.method(x = "omit", y = "omit")
)
```

```
## Model fitted using the sigma parameterization.
```

```
## ASReml 4.1.0 Tue Nov 29 16:56:55 2022
```

	LogLik	Sigma2	DF	wall	cpu
## 1	-4623.985	1.0	1569	16:56:55	0.0

```

## 2 -4063.535      1.0 1569 16:56:55 0.0
## 3 -3471.618      1.0 1569 16:56:55 0.0
## 4 -3105.972      1.0 1569 16:56:55 0.0
## 5 -2955.436      1.0 1569 16:56:55 0.0
## 6 -2934.435      1.0 1569 16:56:55 0.0
## 7 -2933.721      1.0 1569 16:56:55 0.0
## 8 -2933.720      1.0 1569 16:56:55 0.0

```

```
summary(modelz_3)$varcomp
```

```

##           component std.error z.ratio bound %ch
## year          8.029139 1.9920127 4.030666 P  0
## vm(animal, ainv) 5.060775 1.7855255 2.834334 P  0
## ide(animal)    8.412539 1.6494894 5.100087 P  0
## units!units     7.805139 0.3331474 23.428484 P  0
## units!R        1.000000         NA         NA     F  0

```

```

summary(modelw, coef = TRUE)$coef.fixed
wald.asreml(modelz_3, ssType = "conditional", denDF = "numeric")

```

```

##           solution std.error z.ratio
## age_2        0.000000      NA      NA
## age_3       2.577777 0.3355253 7.682811
## age_4       4.247276 0.3309028 12.835418
## age_5       6.094490 0.3375537 18.054872
## age_6       3.132675 0.3371074  9.292811
## (Intercept) 20.305073 0.2899515 70.029214

```

```
## Model fitted using the sigma parameterization.
```

```

## Warning in asreml(fixed = laydate ~ age + byear, random = ~vm(animal, ainv) + :
## Algebraic derivatives for denominator df not available.

```

```

## ASReml 4.1.0 Tue Nov 29 16:56:55 2022
##           LogLik      Sigma2      DF      wall      cpu
## 1      -2933.720      1.0 1569 16:56:55 0.0
## 2      -2933.720      1.0 1569 16:56:55 0.0
## Calculating denominator DF

##
##           Df denDF   F.inc   F.con Margin      Pr
## (Intercept) 1 55.3 1894.00 1894.00      0.00000
## age         4 845.2 152.70 132.90      A 0.00000
## byear       33 466.5 0.77    0.77      A 0.81646

```

4.3 gremlin

TODO (maybe just bother Matthew to do it)

Meanwhile



Figure 4.1: Keep it dry and do no feed after midnight.

4.4 MCMCglmm

4.4.1 Estimating repeatability

With repeated measures on individuals it is often of interest to see how repeatable a trait is. We can estimate the repeatability of a trait as the proportion of phenotypic variance V_P explained by individual variance V_{ind} ; $R = V_{ind}/V_P = V_{ind}/(V_{ind} + V_R)$. As you already know, bayesian modelisation requires prior. Here, we create a uninformative prior with one estimate for the G matrix and one estimate for the Residual matrix, in addition

```
# p.var <- var(gryphonRM$laydate, na.rm = TRUE)
prior3.1 <- list(G = list(G1 = list(V = 1, nu = 0.002)), R = list(
  V = 1,
  nu = 0.002
))
model3.1 <- MCMCglmm(laydate ~ 1,
  random = ~animal, data = gryphonRM,
```

```

prior = prior3.1, verbose = FALSE
)
posterior.mode(model3.1$VCV)

```

```

##     animal      units
## 11.14136 21.22028

```

Note the use of the term `animal` as random allowed to partition the phenotypic variance V_P into among individual variance V_{ind} associated with `animal` and residual variance V_R associated with `units`.

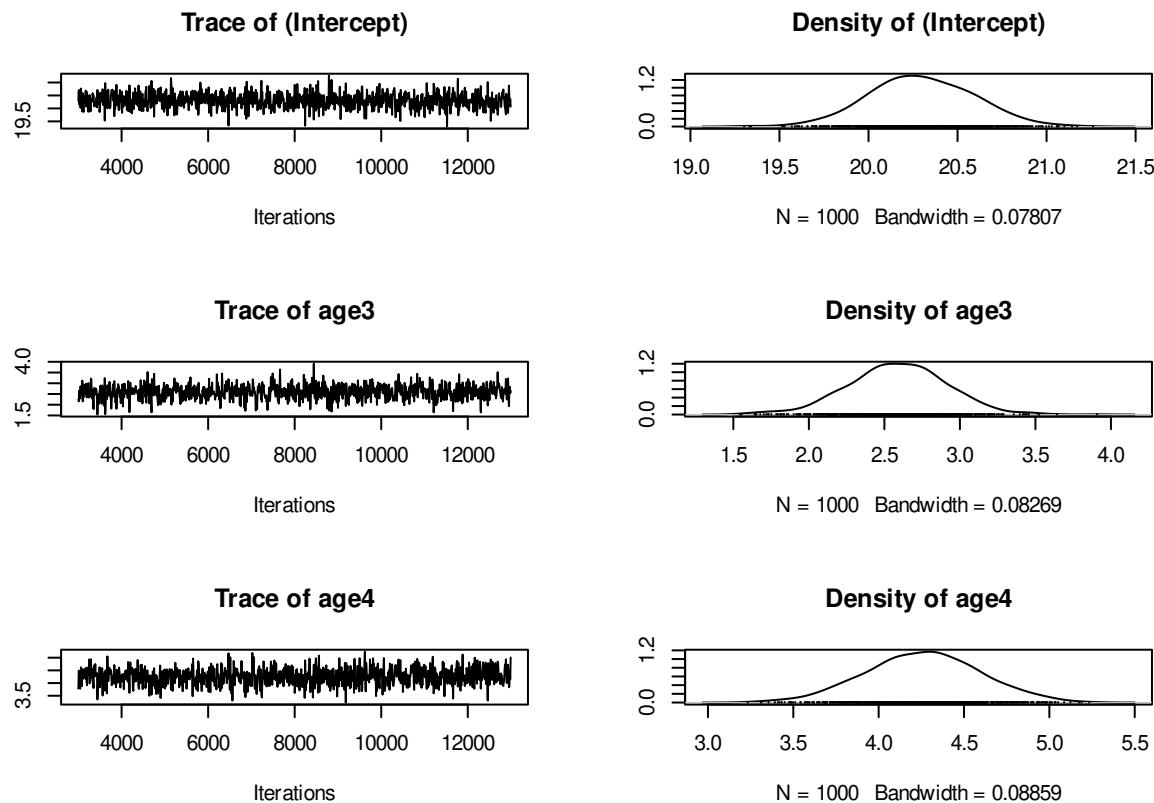
Here then the repeatability of the `laydate` can be determined as: 22.22 (*i.e.*, as $11.141/(11.141 + 21.22)$). Just a friendly remember, we work with Monte Carlo chain with model iteration, so the point estimate can be different (but very similar) each time you run the model.

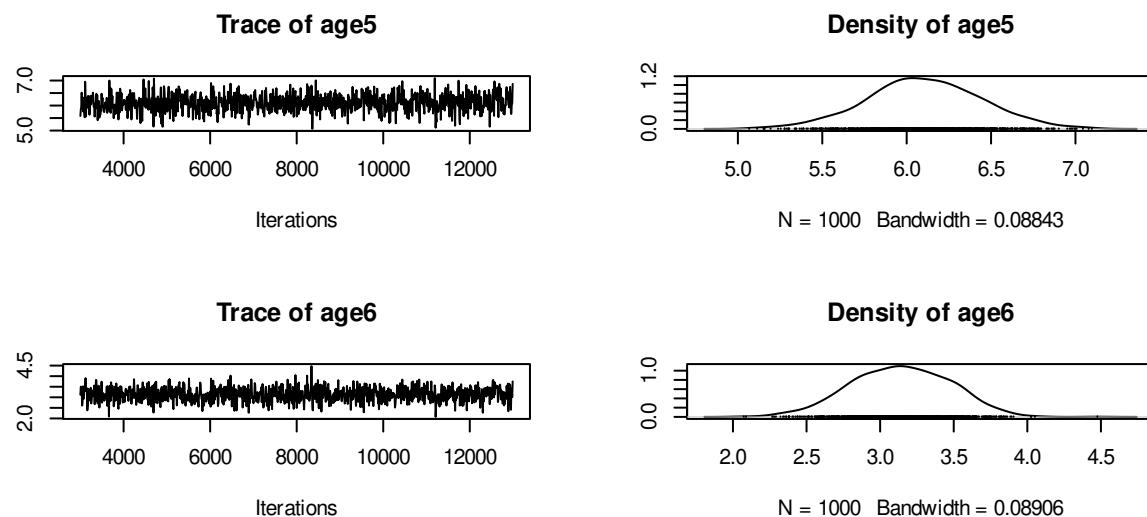
Mean lay date might change with age, so we could ask what the repeatability of lay date is after conditioning on age. This would be done by adding `age` into the model as a fixed effect.

```

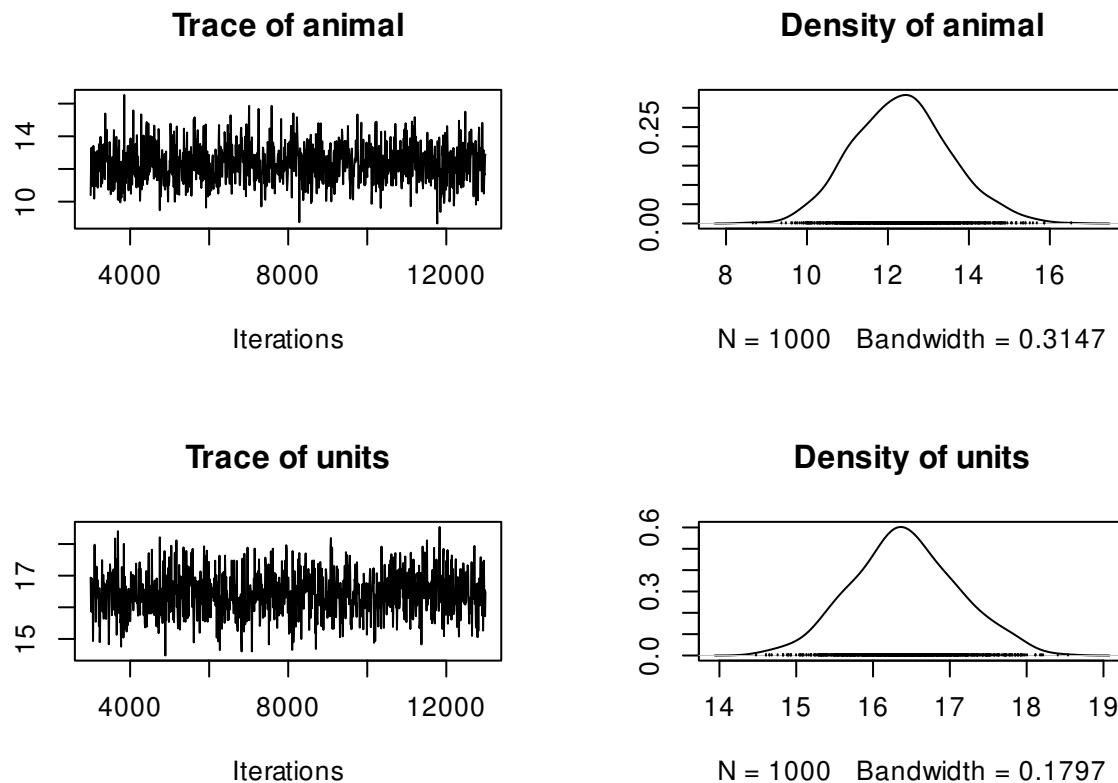
model3.2 <- MCMCglmm(laydate ~ age,
  random = ~animal, data = gryphonRM,
  prior = prior3.1, verbose = FALSE
)
plot(model3.2$Sol)

```





```
plot(model3.2$VCV)
```

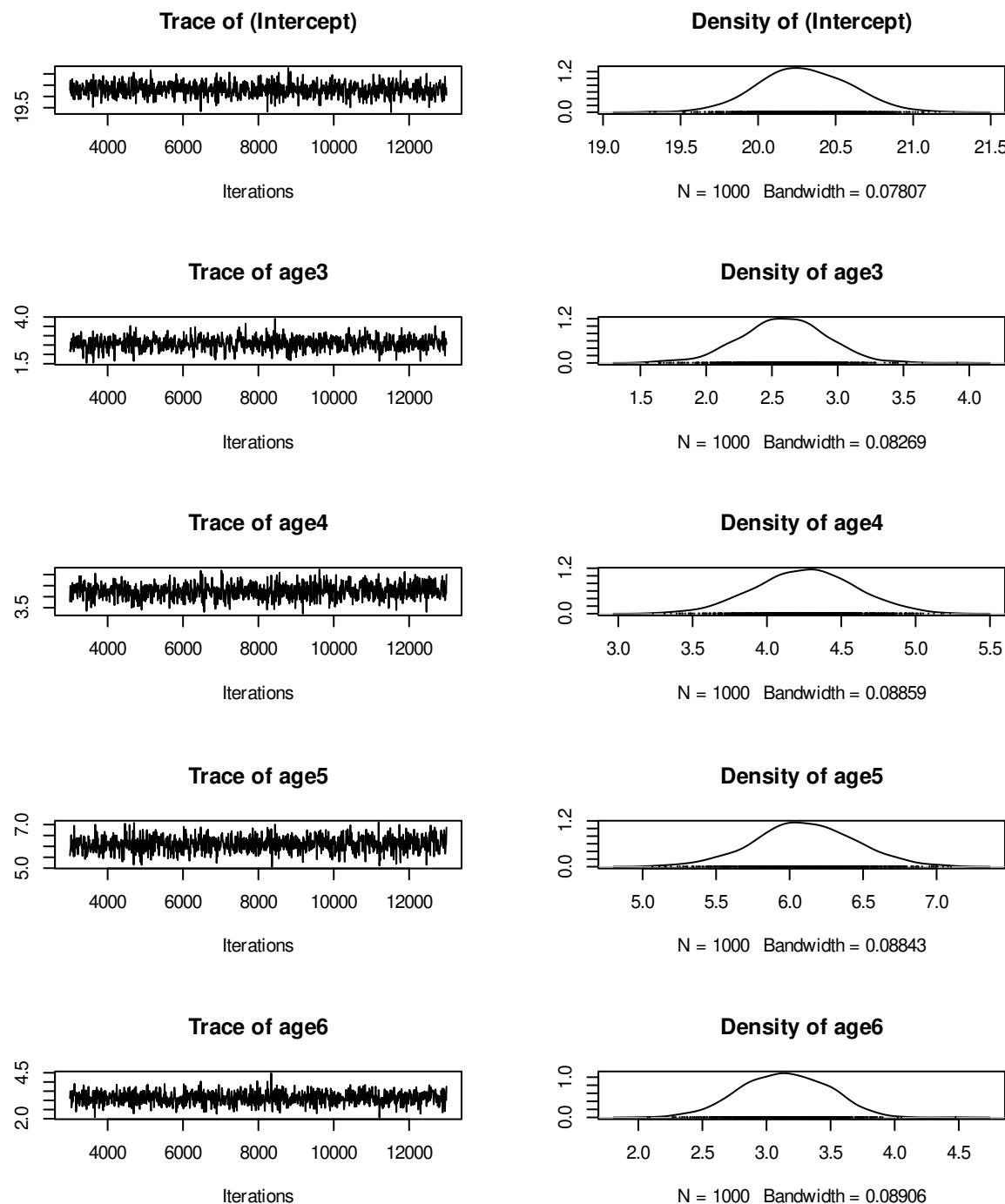


```
posterior.mode(model3.2$VCV)
```

```
##    animal     units
## 12.74460 16.27617
```

The model assumption seems correct, so we can look at the different estimates. Note that the random effect structure has remained unchanged because we did not modify the prior `prior3.1`. The repeatability of `laydate`, after accounting for age effects, is now estimated as 22.22 (*i.e.*, as $11.141/(11.141 + 21.22)$). Just as we saw when estimating h_2 in tutorial 1, the inclusion of fixed effects will alter the estimated effect size if we determine total phenotypic variance as the sum of the variance components. Thus, proper interpretation is vital.

```
plot(model3.2$Sol)
```



```
posterior.mode(model3.2$Sol)
```

```
## (Intercept)      age3       age4       age5       age6
## 20.169385    2.492732   4.327622   5.998775   3.123463
```

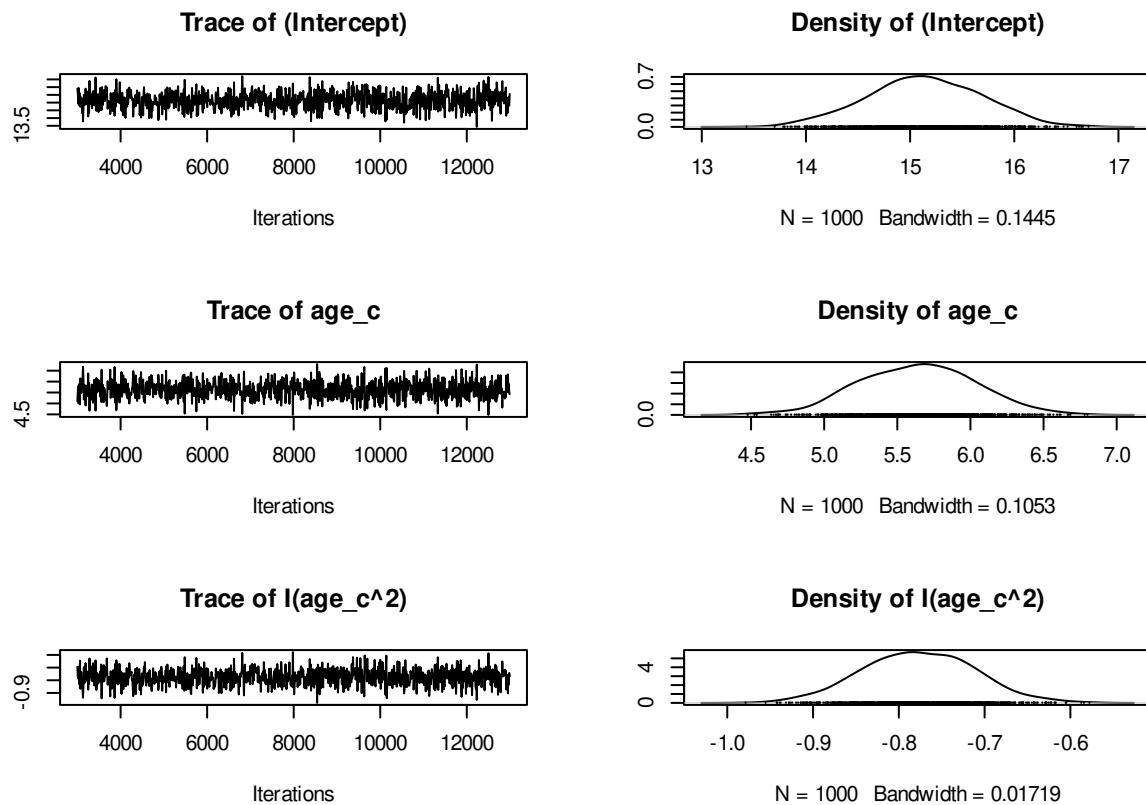
```
HPDinterval(model3.2$Sol, 0.95)
```

```
##           lower      upper
## (Intercept) 19.728159 20.862694
## age3        1.928464 3.201445
## age4        3.640339 4.940057
## age5        5.444036 6.778223
## age6        2.554508 3.825944
## attr(,"Probability")
## [1] 0.95
```

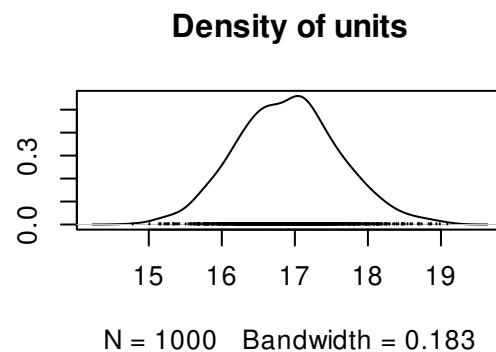
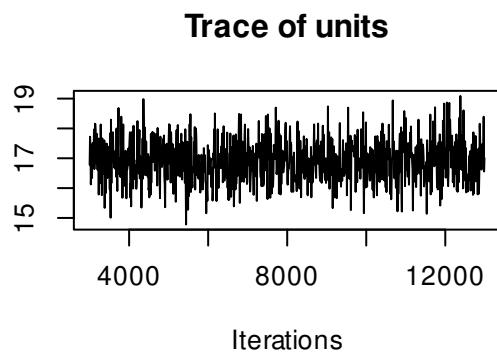
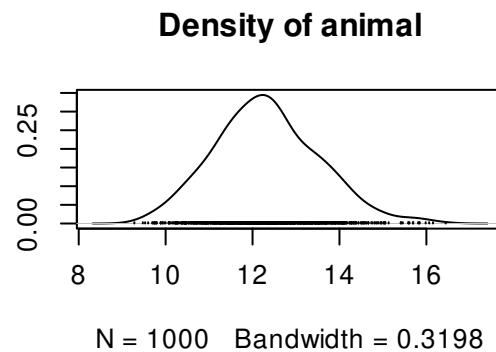
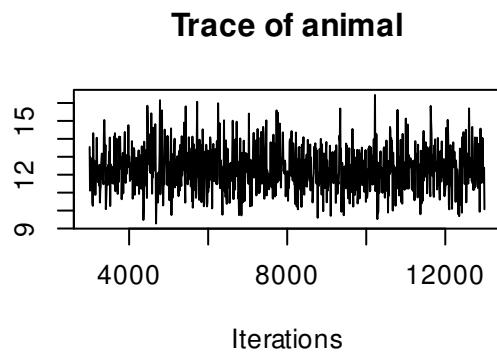
Here age is modeled as a 5-level factor (specified using the function `as.factor()` at the beginning of the analysis). We could equally have fitted it as a continuous variable, in which case, given potential for a late life decline, we would probably also include a quadratic term. In addition, using `age` as continuous variable can help in saving some degree of freedom in the analysis.

```
gryphonRM$age_c <- as.numeric(gryphonRM$age)

model3.2_2 <- MCMCglmm(laydate ~ age_c + I(age_c^2),
  random = ~animal, data = gryphonRM,
  prior = prior3.1, verbose = FALSE
)
plot(model3.2_2$Sol)
```



```
plot(model3.2_2$VCV)
```



```
posterior.mode(model3.2_2$VCV)
```

```
##    animal      units
## 11.47883 17.00732
```

```
posterior.mode(model3.2_2$Sol)
```

```
## (Intercept)      age_c   I(age_c^2)
## 15.126910     5.696353 -0.783416
```

```
HPDinterval(model3.2_2$Sol, 0.95)
```

```
##           lower       upper
## (Intercept) 13.986362 16.0715658
## age_c        4.976011  6.5004344
## I(age_c^2)  -0.907706 -0.6549909
## attr(,"Probability")
## [1] 0.95
```

4.4.2 Partitioning additive and permanent environment effects

Generally we expect that the repeatability will set the upper limit for heritability since among individual variation can be decomposed in the additive genetic variation and non additive genetic variation. In other word, the additive genetic variation is a subcomponent of the difference between individuals. Non-additive contributions to fixed among-individual differences are normally referred to as *permanent environment effects*. If a trait has repeated measures then it is necessary to model permanent environment effects in an animal model to prevent upward bias in V_A .

To illustrate it, we first fit the animal model:

```
Ainv <- inverseA(gryphonped)$Ainv
model3.3 <- MCMCglmm(laydate ~ 1 + age,
  random = ~animal, ginv = list(animal = Ainv),
  data = gryphonRM, prior = prior3.1, verbose = FALSE
)
```

Variance components are almost unchanged if we compare the previous model:

```
posterior.mode(model3.3$VCV)
```

```
##   animal     units
## 13.36499 16.93964
```

```
posterior.mode(model3.2$VCV)
```

```
##   animal     units
## 12.74460 16.27617
```

This suggests that most of the among-individual variance is – rightly or wrongly – being partitioned as V_A here. In fact here the partition is wrong since the simulation included both additive genetic effects and additional fixed heterogeneity that was not associated with the pedigree structure (i.e. permanent environment effects). In order to obtain an unbiased estimate of V_A , we need to fit the individual identity twice in the model: once linked to the pedigree (genetic effect) and once not linked to the pedigree (permanent environment effect). To do so, we need to duplicate the variable containing the individual identity `animal` and give it a new name. In addition, the prior need to be modified to integrate a seconf random effect. An more appropriate estimate of V_A is given by the model:

```
gryphonRM$animal_pe <- gryphonRM$animal
# p.var <- var(gryphonRM$laydate, na.rm = TRUE)
prior3.4 <- list(G = list(G1 = list(V = 1, nu = 0.002), G2 = list(
  V = 1,
  nu = 0.002
)), R = list(V = 1, nu = 0.002))
model3.4 <- MCMCglmm(laydate ~ 1 + age,
```

```

random = ~ animal + animal_pe,
ginv = list(animal = Ainv), data = gryphonRM, prior = prior3.4, verbose = FALSE
)
posterior.mode(model3.4$VCV)

```

```

##      animal animal_pe      units
##  4.159477  7.576943 16.219255

```

The estimate of V_A is now much lower (reduced from 13.6735 to 5.1238) due to a proper separation in the additive and permanent environment effects. We can estimate h^2 and the repeatability from this model:

```

model3.4.VP <- model3.4$VCV[, "animal"] + model3.4$VCV[, "animal_pe"] + model3.4$VCV[, "un"
model3.4.PE_VA <- model3.4$VCV[, "animal"] + model3.4$VCV[, "animal_pe"]
posterior.mode(model3.4.PE_VA / model3.4.VP)

```

```

##      var1
## 0.4135483

```

```

posterior.mode(model3.4$VCV[, "animal"] / model3.4.VP)

```

```

##      var1
## 0.169229

```

4.4.3 Adding additional effects and testing significance

Models of repeated measures can be extended to include other fixed or random effects. For example we can try including year of measurement (`year`) and birth year (`byear`) as other random effects.

```

# p.var <- var(gryphonRM$laydate, na.rm = TRUE)
prior3.5 <- list(G = list(G1 = list(V = 1, nu = 0.002), G2 = list(
  V = 1,
  nu = 0.002
), G3 = list(V = 1, nu = 0.002), G4 = list(
  V = 1,
  nu = 0.002
)), R = list(V = 1, nu = 0.002))

model3.5 <- MCMCglmm(laydate ~ 1 + age,
  random = ~ animal + animal_pe +
    year + byear, ginv = list(animal = Ainv), data = gryphonRM, prior = prior3.5,
  verbose = FALSE
)
posterior.mode(model3.5$VCV)

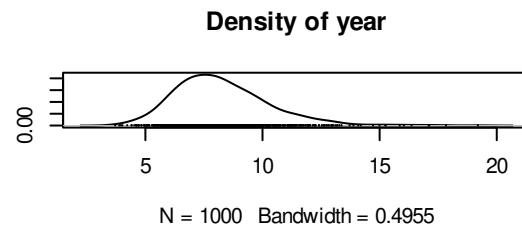
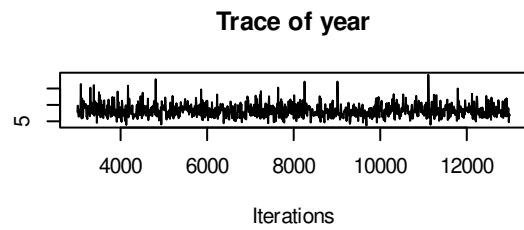
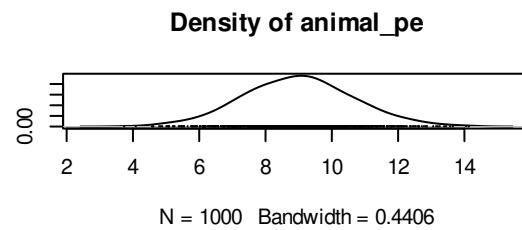
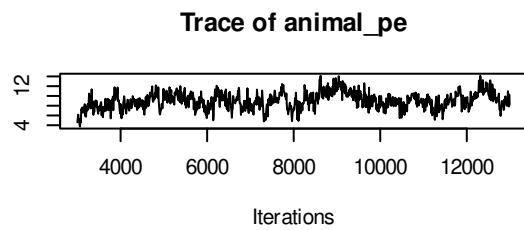
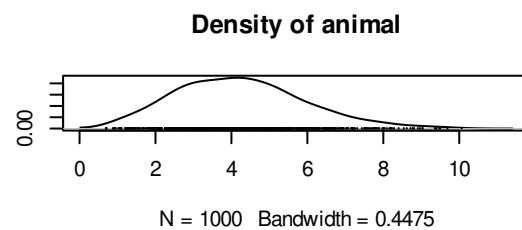
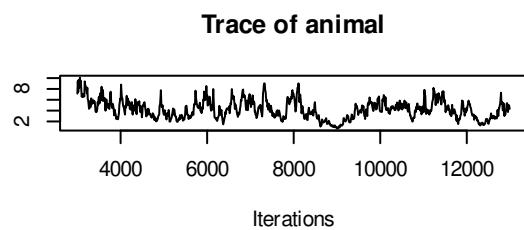
```

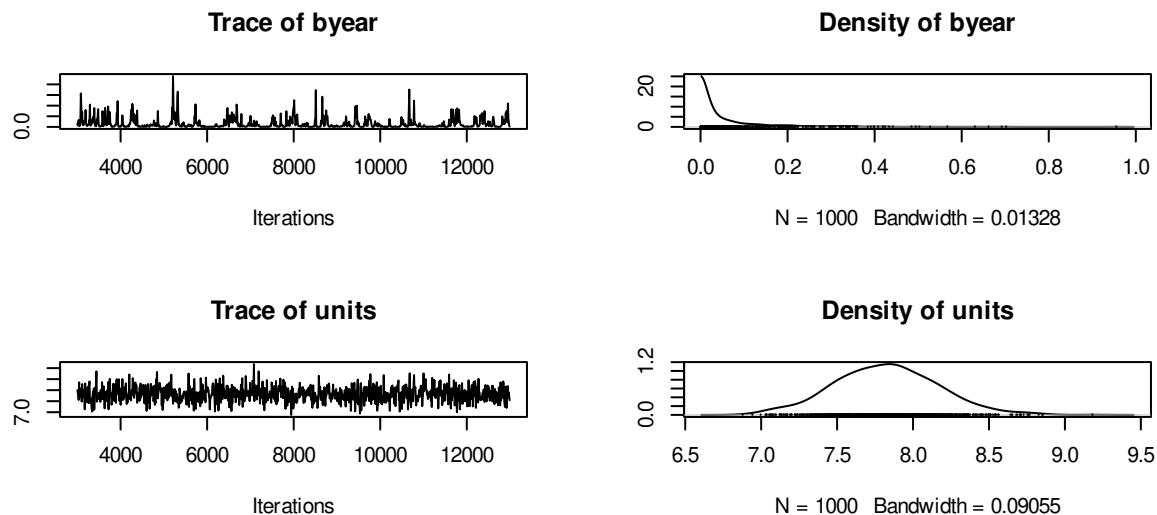
```
##      animal    animal_pe      year      byear      units
## 3.265010495 8.924558498 7.557087451 0.002605666 7.884707995
```

```
HPDinterval(model3.5$VCV, 0.95)
```

```
##           lower       upper
## animal     0.8798128012 7.3118109
## animal_pe 5.7782140304 12.3829408
## year       4.5928409666 12.1586746
## byear      0.0003427203 0.2808403
## units      7.1211681107 8.4819224
## attr(,"Probability")
## [1] 0.95
```

```
plot(model3.5$VCV)
```





This model will return additional variance components corresponding to year of measurement effects and birth year of the female effects.

V_{byear} is very low and its posterior distribution (via the function HPDinterval or plot) is very close to zero indicating its not significant. You have to remember bayesian model never estimate variable to 0 or passing zero, so you will never see a credible interval CI crossing zero for a variance. If you compared the DIC of model3.5 to a reduced model without byear, it should be very similar.

```

prior3.5_2 <- list(
  G = list(G1 = list(V = 1, nu = 0.002), G2 = list(
    V = 1,
    nu = 0.002
  ), G3 = list(V = 1, nu = 0.002)),
  R = list(V = 1, nu = 0.002)
)

model3.5_2 <- MCMCglmm(laydate ~ 1 + age,
  random = ~ animal + animal_pe +
  year, ginv = list(animal = Ainv), data = gryphonRM, prior = prior3.5_2,
  verbose = FALSE
)
posterior.mode(model3.5_2$VCV)

##      animal animal_pe      year      units

```

```
## 3.774539 8.762062 7.215465 7.813281
```

```
model3.5$DIC
```

```
## [1] 8291.123
```

```
model3.5_2$DIC
```

```
## [1] 8290.812
```

`year` effects could alternatively be included as fixed effects (try it!, you should be able to handle the new prior specification at this point). This will reduce V_R and increase the estimates of heritability and repeatability, which must now be interpreted as proportions of phenotypic variance after conditioning on both age and year of measurement effects.

4.5 brms

```
library(brms)

Amat <- as.matrix(nadiv::makeA(gryphonped))
gryphonRM$animal_pe <- gryphonRM$animal

model_simple1.1 <- brm(
  laydate ~ 1 + (1 | gr(animal, cov = Amat)) + (1 | animal_pe),
  data = gryphonRM,
  family = gaussian(),
  data2 = list(Amat = Amat),
  chains = 2, cores = 2, iter = 1000
)
summary(model_simple1.1)
plot(model_simple1.1)
```

4.6 stan

to do

Chapitre 5

Quick comparison of codes

5.1 Univariate model with repeated measures

5.1.1 Asreml-R

5.1.2 gremlin

5.1.3 MCMCglmm

5.1.4 brms

5.2 bivariate model

5.2.1 Asreml-R

5.2.2 gremlin

5.2.3 MCMCglmm

5.2.4 brms