## Vignette for the package rehh (version 2+)

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This vignette is aimed at presenting additional information on the R package *rehh* by describing how to use it to perform whole genome scan for footprints of selection using statistics related to the Extended Haplotype Homozygosity (EHH) (SABETI *et al.* 2002). Importantly, the current implementation of tests assumes markers are bi-allelic.

The *rehh* package is currently available for most platforms (Linux, MS Windows and MacOSX) from the CRAN repository (http://cran.r-project.org/) and may be installed using standard procedure. Once the package has been successfully installed on your system, it can be loaded using the following command:

library(rehh)

## 1 Input Files

The package *rehh* basically requires as input:

- 1. haplotype data file(s) (in three possible format) for each population of interest (see 1.1)
- 2. a SNP information file (see 1.2)

Important Note: For a given chromosome, SNPs are assumed to be ordered in the same way in both the haplotype file (columns) and the SNP information file.

For illustration purposes, example files that originate from a previously published study on the Creole cattle breed from Guadeloupe (CGU) (GAUTIER and NAVES 2011) are provided in the package and can be copied in the working directory with the following command:

make.example.files()

In the following examples, the command make.example.file() is assumed to have been run (see above) so that example files are in the working directory.

#### 1.1 Haplotype data file

Three haplotype input file formats are supported:

- 1. a standard haplotype format. Each line represents a haplotype (the first element being an haplotype identifier) with SNP genotype in columns as in the example file bta12\_cgu.hap created when running the function make.example.files() and that consists of 280 haplotypes (identifier 1 to 280) of 1424 SNPs each. See 1.3.1 for a detailed example.
- 2. A (transposed) format with haplotype in columns and SNPs in row as in the example file bta12\_cgu.thap created when running the function make.example.files(). Note that this format is similar to the one produced by the phasing program SHAPEIT2 (O'CONNELL et al. 2014). See 1.3.2 for a detailed example.
- 3. the output file format from the phasing program fastPHASE (SCHEET and STEPHENS 2006) as in the bta12\_hapguess\_switch.out example file created when running the function make.example.files(). Note that haplotypes might originate from several different populations (i.e., if the -u fastPHASE option was used). See 1.3.3 for a detailed example.

By default alleles are assumed to be coded as 0 (missing data), 1 (ancestral allele) or 2 (derived allele). Recoding of the alleles in this format, according to the SNP information data file (see 1.2) can be performed with the recode.allele option of the function data2haplohh() (see 1.3).

## 1.2 SNP information data file

This data file should contain SNP information as in the map.inp example file created when running the function make.example.files(). Each line contains five columns corresponding to:

- 1. the SNP name
- 2. the SNP chromosome (or scaffold) of origin
- 3. the SNP position on the chromosome (or scaffold). Note that it is up to the user to choose either physical or genetic map positions (if available).
- 4. the SNP ancestral allele (as coded in the haplotype input file)
- 5. the SNP derived alleles (as coded in the haplotype input file)

The fourth and fifth columns (allele coding) should be filled in but the corresponding information is only used when activating the recode.allele option of the function data2haplohh() (see 1.3). In that case, for each SNP, the allele specified in the fourth (respectively fifth) column will be recoded as 1 (respectively 2), any other allele name will be recoded as 0 (i.e., missing data). More importantly, it should be noticed that the ancestral or derived allele information associated to this coding are only relevant for within population tests (based on iHS). In other words, if one is only interested in across-population tests (based on RSD or RSD or RSD alleles in the fourth and fifth column may be performed randomly.

As an illustration, the following R command displays the first five row of the map.inp example file created when running the function make.example.files():

#### head(read.table("map.inp"))

```
V1 V2
                    V3 V4 V5
 1 F0100190
              1 113642
 2 F0100220
              1 244699
 3 F0100250
              1 369419
> 4 F0100270
              1 447278
                           Т
> 5 F0100280
              1 487654
                        Τ
                            Δ
> 6 F0100290
              1 524507
```

## 1.3 Loading data files

The data2haplohh() function allows to convert data file into an R object of class haplohh subsequently used by the functions of the *rehh* package. The following main options are available to recode alleles or select SNPs (based on Minor Allele Frequency or percentage of missing data) and haplotypes (based on percentage of missing data):

- 1. Allele recoding option: This option is activated with recode.allele=TRUE and allows to recode haplotype according to the ancestral and derived allele definition available in the SNP information file (fourth and fifth columns) as: 0 (missing data), 1 (ancestral allele) or 2 (derived allele).
- 2. Discard haplotypes with a high amount of missing data. If haplotypes contain missing data (which is generally not the case since most phasing programs allow imputing missing genotypes), it is possible to discard those with less min\_perc\_geno.hap % of the SNPs genotyped. By default min\_perc\_geno.hap=100 meaning that only completely phased haplotypes are retained.
- 3. Discard SNPs with a high amount of missing data. It is possible to discard SNPs genotyped on less than min\_perc\_geno.snp % of the haplotypes. By default min\_perc\_geno.snp=100 meaning that only fully genotyped SNPs are retained.

4. **Discard SNPs with a low Minor Allele Frequency**. It is possible to discard SNPs with a MAF below min\_maf. This is generally not recommended and by default min\_maf=0 meaning that all SNPs are retained.<sup>1</sup>

More details about the different arguments of the function are available in the documentation accessible using the command:

```
?data2haplohh
```

In the following we detail three different examples based on the example data files provided with the package (see 1).

#### 1.3.1 Example 1: reading haplotype file in standard format

In this example, the example haplotype input file bta12\_cgu.hap (standard format) and SNP information input files map.inp are converted into an haplohh object named hap. Because the map file contains information for SNPs mapping to other chromosomes than the one of interest (BTA12), we use the option chr.name=12. Allele recoding is activated (recode.allele=TRUE) to allow recoding alleles in the rehh format (0,1 or 2).

If no value is specified for chr.name argument and more than one chromosome is detected in the map file, the function asks interactively which chromosome to choose:

> 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29

12

> 1:

```
> Map file seems OK: 1424 SNPs declared for chromosome 12
> Standard rehh input file assumed
```

<sup>&</sup>gt; Alleles are being recoded according to map file as:

<sup>&</sup>gt; 0 (missing data), 1 (ancestral allele) or 2 (derived allele)

 $<sup>\</sup>overline{\ ^{1}{\rm The\ arguments\ min\_perc\_geno.hap,\ min\_perc\_geno.snp}\ {\rm and\ min\_maf}\ {\rm are\ evaluated\ in\ this\ order}.$ 

```
> Discard Haplotype with less than 100 % of genotyped SNPs
```

- > No haplotype discarded
- > Discard SNPs genotyped on less than 100 % of haplotypes
- > No SNP discarded
- > Data consists of 280 haplotypes and 1424 SNPs

Finally, as an example of error message, the following message is prompted if the number of SNPs in the chromosome (for instance when a wrong chromosome is declared) does not correspond to the one in the haplotype file:

- > Map file seems OK: 1123 SNPs declared for chromosome 18
- > Standard rehh input file assumed
- > The number of snp in the haplotypes 1424 is not equal
- > to the number of snps declared in the map file 1123
- > Error in data2haplohh(hap\_file = "bta12\_cgu.hap", map\_file = "map.inp", : Conversion stopped

#### 1.3.2 Example 2: reading haplotype file in transposed format (SHAPIT2-like)

In this example, the example haplotype input file bta12\_cgu.thap (transposed format) and SNP information input files map.inp are converted into an haplohh object named hap. Setting haplotype.in.columns=TRUE informs the function that the haplotype file is in transposed format:

```
> Map file seems OK: 1424 SNPs declared for chromosome 12
```

- > Haplotype are in columns with no header
- > Alleles are being recoded according to map file as:
- > 0 (missing data), 1 (ancestral allele) or 2 (derived allele)
- > Discard Haplotype with less than 100 % of genotyped SNPs
- > No haplotype discarded
- > Discard SNPs genotyped on less than 100 % of haplotypes
- > No SNP discarded
- > Data consists of 280 haplotypes and 1424 SNPs

#### 1.3.3 Example 3: reading haplotype file in fastPHASE output format

In this example, the example <code>fastPHASE</code> output file <code>bta12\_hapguess\_switch.out</code> and SNP information input files <code>map.inp</code> are converted into a <code>haplohh</code> object named <code>haplo</code>. As explained above we use the option <code>chr.name=12</code>. Because, haplotypes originate from several populations (the -u <code>fastPHASE</code> option was used), we specify the population of interest (in our example the 280 haplotypes from the CGU population, see above) using the option <code>popsel=7</code> (7 corresponding to the code of CGU in the example <code>fastPHASE</code> input files).

```
hap<-data2haplohh(hap_file="bta12_hapguess_switch.out",map_file="map.inp", recode.allele=TRUE,popsel=7,chr.name=12)
```

```
> Map file seems OK: 1424 SNPs declared for chromosome 12
> Looks like a FastPHASE haplotype file
> Haplotypes originate from 8 different populations in the fastPhase output file
> Alleles are being recoded according to map file as:
> 0 (missing data), 1 (ancestral allele) or 2 (derived allele)
> Discard Haplotype with less than 100 % of genotyped SNPs
> No haplotype discarded
> Discard SNPs genotyped on less than 100 % of haplotypes
> No SNP discarded
> Data consists of 280 haplotypes and 1424 SNPs
```

If no value is specified for the popsel argument and more than one population is detected in the *fastPHASE* output file, the function asks interactively which population to chose:

#### 7

```
> Map file seems OK: 1424 SNPs declared for chromosome 12
> Looks like a FastPHASE haplotype file
> Haplotypes originate from 8 different populations in the fastPhase output file
> Alleles are being recoded according to map file as:
> 0 (missing data), 1 (ancestral allele) or 2 (derived allele)
> Discard Haplotype with less than 100 % of genotyped SNPs
> No haplotype discarded
> Discard SNPs genotyped on less than 100 % of haplotypes
> No SNP discarded
> Data consists of 280 haplotypes and 1424 SNPs
```

## 2 Computing *EHH*-based statistics for individual markers with the calc\_ehh(), calc\_ehhs() and scan\_hh() functions

#### 2.1 Definition and Computation

## 2.1.1 The (allele-specific) Extended Haplotype Homozygosity (EHH)

For a given core allele (i.e., the ancestral or derived allele) at a focal SNP, the (allele–specific) extended haplotype homozygosity (EHH) is defined as the probability that two randomly chosen chromosomes (carrying the core allele considered) are identical by descent (as assayed by homozygosity at all SNPs) over a given surrounding chromosome region (SABETI et al. 2002). The EHH thus aims at measuring to which extent an extended haplotype is transmitted without recombination. In practice, the EHH (EHH<sub>s,t</sub>) of a tested core allele  $a_s$  ( $a_s = 1$  or  $a_s = 2$ ) for a focal SNP s over the chromosome interval extending to SNP t is computed as:

$$EHH_{s,t} = \frac{1}{n_{a_s}(n_{a_s} - 1)} \sum_{k=1}^{K_{a_s,t}} n_k(n_k - 1)$$
(1)

where  $n_{a_s}$  represents the number of haplotype carrying the core allele  $a_s$ ,  $K_{a_s,t}$  represents the number of different extended haplotypes (from SNP s to SNP t) carrying  $a_s$  and  $n_k$  is the number of the extended haplotype k (i.e.,  $n_{a_s} = \sum_{k=1}^{K_{a_s,t}} n_k$ ).

#### 2.1.2 The integrated (allele-specific) EHH (iHH)

By definition, irrespective of the allele considered, EHH starts at 1, and decays monotonically to 0 with increasing distance from the focal SNP. For a given core allele, the integrated EHH (iHH) is defined as the area under the EHH curve with respect to map position (Voight  $et\ al.\ 2006$ )<sup>2</sup>. In rehh, iHH is computed using the trapezoid method. In practice, the integral might only be computed for the regions of the curve above an arbitrarily small EHH value (e.g., EHH > 0.05).

#### 2.1.3 The site-specific Extended Haplotype Homozygosity (EHHS)

For a given core SNP, the (site–specific) extended haplotype homozygosity (*EHHS*) is defined as the probability that two randomly chosen chromosomes are identical by descent (as assayed by homozygosity at all SNPs) over a given surrounding chromosome region. *EHHS* might roughly be viewed as linear combination of the *EHH*'s for the two alternative alleles with weights function of the corresponding allele frequencies. Two different *EHHS* estimators further denoted as EHHS<sup>Sab</sup> (SABETI *et al.* 2007) and EHHS<sup>Tang</sup> (TANG *et al.* 2007) have been proposed. For a focal SNP s over a chromosome interval extending to SNP t, these are computed as (following the same notation as above):

EHHS<sup>Sab</sup><sub>s,t</sub> = 
$$\frac{1}{n_s(n_s - 1)} \sum_{a_s = 1}^{a_s = 2} \left( \sum_{k=1}^{K_{a_s,t}} n_k(n_k - 1) \right)$$
 (2)

where  $n_s = \sum_{a_s=1}^{a_s=2} n_{a_s}$  and

$$EHHS_{s,t}^{Tang} = \frac{1 - h_{hap}^{(s,t)}}{1 - h_{all}^{(s)}}$$
(3)

where:

1. 
$$h_{all}^{(s)} = \frac{n_s}{n_s-1} \left(1 - \frac{1}{n_s^2} \sum_{a_s=1}^{a_s=2} n_{a_s}^2\right)$$
 is an estimator of the focal SNP heterozygosity

2. 
$$h_{hap}^{(s,t)} = \frac{n_s}{n_s - 1} \left( 1 - \frac{1}{n_s^2} \sum_{a_s = 1}^{a_s = 2} \left( \sum_{k=1}^{K_{a_s,t}} n_k^2 \right) \right)$$
 is an estimator of haplotype heterozygosity across the chromosome region extending from SNP  $s$  to SNP  $t$ .

 $<sup>^2</sup>$ In their seminal paper, Voight et al. considered genetic distances and apply a penalty (proportional to physical distances) for successive SNPs separated by more than 20 kb. In addition, they did not compute iHH if any physical distance between a pair of neighboring SNPs was above 200 kb. We did not implement such an approach in rehh although this might easily be done (providing relevant information is available for the genome of interest) by modifying the positions of the markers in SNP information input file.

#### 2.1.4 The integrated EHHS (iES)

As for the EHH (see 2.1.2), EHHS starts at 1 and decays monotonically to 0 with increasing distance from the focal SNP. For a given focal SNP, and in a similar fashion as the iHH, iES is defined as the integrated EHHS (Tang et~al.~2007). Depending on the EHHS estimator considered (respectively,  $EHHS^{Sab}$  and  $EHHS^{Tang}$ ), two different iES estimators, that we further denoted as  $iES^{Sab}$  and  $iES^{Tang}$  respectively, can be computed. As for iHH, the iES integral is computed using the trapezoid method and might only be computed for the regions of the curve above an arbitrarily small EHHS value (e.g., EHHS > 0.05).

#### 2.1.5 Dealing with missing data

In the computation of both EHH and EHHS from a focal SNP s to a SNP t, only extended haplotypes with no missing data are considered. As a consequence, the number of extended haplotypes retained to compute these two statistics might decrease with distance from the focal SNP. However if the number of available extended haplotypes falls below a threshold (e.g., limhaplo=5), EHH and EHHS are not computed further. Note however that most phasing programs (such as fastPHASE or SHAPEIT2) allow to impute missing genotypes resulting in phased haplotypes with no missing data.

## 2.2 The function calc\_ehh()

The calc\_ehh() function allows to compute EHH for both the ancestral  $(a_s = 1)$  and derived  $(a_s = 2)$  alleles at the  $s^{th}$  SNP relative to each SNP (t) upstream and downstream and corresponding iHH. The two options limehh and limhaplo allow to specify condition to stop computing EHH (see 2.1.1). By default limehh=0.05 and limhaplo=2. Finally, if plotehh=TRUE, the decay of EHH for both the ancestral and derived alleles is plotted against SNP map position (main\_leg allows to change the plot legend). More details are available in the R documentation by using the command:

```
?calc_ehh
```

In the following example, the *EHH* statistics are computed for both the ancestral and derived allele of the 456<sup>th</sup> focal SNP. Note that the haplohh\_cgu\_bta12 object was generated using the data2haplohh() function with the example input files (1.3.1). For convenience, it is stored as an example object (accessible with the R function data) as shown below:

```
#example haplohh object (280 haplotypes, 1424 SNPs) see ?haplohh_cgu_bta12 for details
data(haplohh_cgu_bta12)
#computing EHH statistics for the focal SNP at position 456
#which display a strong signal of selection
res.ehh<-calc_ehh(haplohh_cgu_bta12,mrk=456)</pre>
```

The five different elements of the resulting res.ehh object are as follows:

```
> F1205380 F1205390 F1205400 F1205420 F1205440
> Ancestral allele 85 85 85 85
> Derived allele 195 195 195 195 195
```

res.ehh\$freq all1

> [1] 0.3035714

#### res.ehh\$ihh

```
> Ancestral allele Derived allele
> 284633 2057152
```

In addition, as plotehh=TRUE by default, we obtain the following plot (Figure 1):

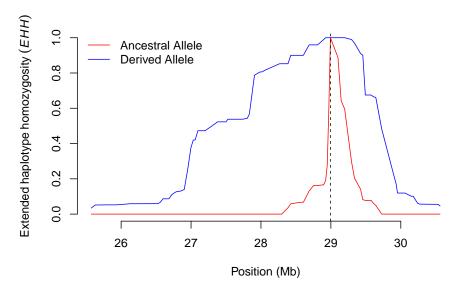


Figure 1: Graphical output for the calc\_ehh() function

## 2.3 The function calc\_ehhs()

The calc\_ehhs() function allows to compute the EHHS (both the EHHS<sup>Sab</sup> and EHHS<sup>Tang</sup> estimators) at the  $s^{th}$  SNP relative to each SNP (t) upstream and downstream. This function also compute the corresponding iES ( $iES^{Sab}$  and  $iES^{Tang}$  estimators respectively). The two options limehhs and limhaplo allow to specify condition to stop computing EHHS (see 2.1.3). By default limehhs=0.05 and limhaplo=2. Finally, if plotehhs=TRUE, the decay of EHHS is plotted against SNP map position (main\_leg allows to change the plot legend). More details are available in the R documentation by using the command:

#### ?calc\_ehhs

In the following example, the *EHHS* statistics are computed for the 456<sup>th</sup> focal SNP on the haplohh\_cgu\_bta12 object defined above (see 2.2) was generated using the data2haplohh() function with the example input files (see 1.3.1) described above. For convenience, it is stored as an example object (accessible with the R function data).

```
#example haplohh object (280 haplotypes, 1424 SNPs) see ?haplohh_cgu_bta12 for details
data(haplohh_cgu_bta12)
#computing EHH statistics for the focal SNP at position 456
#which display a strong signal of selection
res.ehhs<-calc_ehhs(haplohh_cgu_bta12,mrk=456)</pre>
```

The five different elements of the resulting res.ehhs object are as follows:

```
res.ehhs$EHHS Sabeti et al 2007[453:459]
> F1205370 F1205380 F1205390 F1205400 F1205420 F1205440 F1205450
> 0.5017153 0.5095238 0.5347926 1.0000000 0.5654122 0.5429595 0.5386841
res.ehhs$EHHS_Tang_et_al_2007[453:459]
 F1205370 F1205380 F1205390 F1205400 F1205420 F1205440 F1205450
> 0.8715588 0.8851234 0.9290193 1.0000000 0.9822104 0.9432066 0.9357794
res.ehhs$nhaplo_eval[453:459]
> F1205370 F1205380 F1205390 F1205400 F1205420 F1205440 F1205450
       280
                280
                         280
                                  280
                                           280
                                                   280
                                                            280
res.ehhs$IES_Tang_et_al_2007
> [1] 1760565
res.ehhs$IES_Sabeti_et_al_2007
> [1] 964698
```

In addition, as plotehh=TRUE by default, we obtain the following plot (Figure 2):

### 2.4 The function scan\_hh()

The scan\_hh() function allows to efficiently compute *IHH* (for both the ancestral and derived alleles) and *IES* (both the iES<sup>Sab</sup> and iES<sup>Tang</sup> estimators) for all the SNPs in the haplohh object considered. The options limehh, limehhs and limhaplo specify conditions to stop computing *EHH* and *EHHS*. By default limehh=limehhs=0.05 and limhaplo=2. Finally, the option threads, set by dafault to threads=1, allows to specify the number of available threads to parallelize computation (parallelization being carried out over SNPs). For instance to scan the haplohh\_cgu\_bta12 object (containing data on 1424 SNPs for 280 haplotypes), one may use the following command:

```
data(haplohh_cgu_bta12)
res.scan<-scan_hh(haplohh_cgu_bta12)</pre>
```

The resulting object res.scan is a data frame with haplohh\_cgu\_bta12nsnp (number of SNPs declared in the haplohh object) and seven columns giving for each SNP:

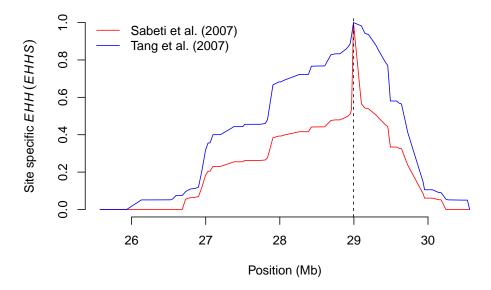


Figure 2: Graphical output for the calc\_ehhs() function

- 1. The SNP chromosome of origin
- 2. The SNP position
- 3. The SNP ancestral allele frequency
- 4. The estimated IHH for the ancestral allele (IHH\_A)
- 5. The estimated IHH for the derived allele (IHH\_D)
- 6. The estimated  $iES^{Tang}$
- 7. The estimated iES<sup>Sab</sup>.

As an example, the following R codes provide the dimension and the first five rows of the res.scan data frame obtained above:

```
dim(res.scan)
```

> [1] 1424 7

#### head(res.scan)

```
iHH_D iES_Tang_et_al_2007
           CHR POSITION
                            freq_A
                                      iHH A
> F1200140
            12
                  79823 0.1500000 135102.2
                                              68522.91
                                                                   69776.85
> F1200150
            12
                  125974 0.4071429 161680.3 107183.15
                                                                  123607.13
> F1200170
            12
                  175087 0.3571429 157333.1 155777.56
                                                                  156021.90
                 219152 0.2214286 250037.4 159839.73
> F1200180
            12
                                                                  166214.75
> F1200190
            12
                 256896 0.1750000 466071.8 173269.33
                                                                  184453.42
                  316254 0.3892857 292077.5 228681.21
> F1200210
            12
                                                                  246572.65
           iES_Sabeti_et_al_2007
> F1200140
                         53669.39
> F1200150
                         76287.51
> F1200170
                         92770.96
> F1200180
                        110712.37
> F1200190
                        134092.34
> F1200210
                        130156.22
```

Note that running scan\_hh() is more efficient than running calc\_ehh() and calc\_ehhs() in turn as shown in the example below (scan\_hh():.

```
system.time(res.scan<-scan_hh(haplohh_cgu_bta12))</pre>
```

```
> user system elapsed
> 0.284  0.000  0.284
```

```
foo<-function(haplo){
  res.ihh=res.ies=matrix(0,haplo@nsnp,2)
  for(i in 1:length(haplo@position)){
    res.ihh[i,]=calc_ehh(haplo,mrk=i,plotehh=FALSE)$ihh
    tmp=calc_ehhs(haplo,mrk=i,plotehhs=FALSE)
    res.ies[i,1]=tmp$IES_Tang_et_al_2007
    res.ies[i,2]=tmp$IES_Sabeti_et_al_2007
}
list(res.ies=res.ies,res.ihh=res.ihh)
}
system.time(res.scan2<-foo(haplohh_cgu_bta12))</pre>
```

```
> user system elapsed
> 13.340  0.000  13.363
```

Note however that the same results are obtained (since the same options were used) as illustrated by the following R code:

```
sum(res.scan2$res.ihh[,1]!=res.scan[,4]) + sum(res.scan2$res.ihh[,2]!=res.scan[,5]) +
sum(res.scan2$res.ies[,1]!=res.scan[,6]) + sum(res.scan2$res.ies[,2]!=res.scan[,7])
```

> [1] 0

# 3 Computing *iHS*, *Rsb* and *xpEHH*: the ihh2ihs(),ies2rsb() and ies2xpehh() functions

## 3.1 Within population test: the *iHS*

## 3.1.1 Definition

Let UniHS represent the log-ratio of the iHH for its ancestral  $(iHH_a)$  and derived  $(iHH_d)$  alleles:

$$\text{UniHS} = \log \left( \frac{\text{iHH}_a}{\text{iHH}_d} \right)$$

The iHS of a given focal SNP s (iHS(s)) is then defined as its standardized UniHS (UniHS(s)) following (VOIGHT  $et\ al.\ 2006$ ):

$$\mathrm{iHS}(s) = \frac{\mathrm{UniHS}(s) - \mu_{\mathrm{UniHS}}^{p_s}}{\sigma_{\mathrm{UniHS}}^{p_s}}$$

where  $\mu_{\text{UniHS}}^{p_s}$  and  $\sigma_{\text{UniHS}}^{p_s}$  represent the average and standard deviation of the UniHS computed over all the SNPs with a derived allele frequency  $p_s$  similar to that of the core SNP s. In practice, the derived allele

frequencies are generally binned so that each bin are large enough (e.g., >10 SNPs) to obtain reliable estimate of  $\mu_{\text{UniHS}}^{p_s}$  and  $\sigma_{\text{UniHS}}^{p_s}$ .

Note that the iHS is constructed to have an approximately standard Gaussian distribution and to be comparable across SNPs regardless of their underlying allele frequencies. Hence, one may further transform iHS into  $p_{iHS}$  (Gautier and Naves 2011):

$$p_{iHS} = -\log_{10} (1 - 2|\Phi(iHS) - 0.5|)$$

where  $\Phi(x)$  represents the Gaussian cumulative distribution function. Assuming most of the genotyped SNPs behave neutrally (i.e., the genome-wide empirical iHS distribution is a fair approximation of the neutral distribution),  $p_{iHS}$  might thus be interpreted as a two-sided P-value (on a  $-\log_{10}$  scale) associated to the neutral hypothesis of no selection.

#### 3.1.2 The function ihh2ihs()

The ihh2ihs() function allows to compute *iHS* using a matrix of *iHH* statistics (for both the ancestral and derived alleles) in the same format as obtained from the scan\_hh() function (see 2.4). The argument minmaf allows to remove SNPs according to their MAF (by default SNPs with a MAF<minmaf=0.05 are discarded from the standardization). The argument freqbin controls the size of the allele frequency bins used to perform standardization (see 3.1.1). More precisely allele frequency bins vary from minmaf to 1-minmaf per step of size freqbin (by default freqbin=0.025). Note that if freqbin is set to 0 (e.g., with a large number of SNPs and few haplotypes), standardization is performed considering each observed frequency as a frequency class.

For instance, to perform a whole genome scan one might run  $scan_hh()$  in turn on haplotype data from each chromosome and concatenate the resulting matrices before standardization. In the following example, we assume that the haplotype files are named as  $hap_chr_i.pop1$  where the chromosome number i goes from 1 to 29 and the SNP information file is named snp.info. The R code below then generates a matrix wg.res with  $iHH_a$  and  $iHH_d$  estimates for all SNPs in an appropriate format to perform standardization with the ihh2ihs function:

```
for(i in 1:29){
  hap_file=paste("hap_chr_",i,".pop1",sep="")
  data<-data2haplohh(hap_file="hap_file","snp.info",chr.name=i)
  res<-scan_hh(data)
  if(i==1){wg.res<-res}else{wg.res<-rbind(wg.res,res)}
  }
  wg.ihs<-ihh2ihs(wg.res)</pre>
```

As a matter of illustration, results of a similar genome scan (Gautier and Naves 2011) are provided as example data sets. The following R code allows to compute the *iHS* for the CGU population:

```
data(wgscan.cgu)
## results from a genome scan (44,057 SNPs) see ?wgscan.eut and ?wgscan.cgu for details
ihs.cgu<-ihh2ihs(wgscan.cgu)</pre>
```

The corresponding object ihs.cgu is a list with two elements corresponding to

1. a matrix of SNP iHS and the corresponding  $p_{iHS}$  (P-values in a  $-\log_1 0$  scale assuming the iHS are normally distributed under the neutral hypothesis). For instance, the five first rows of the ihs.cgu\$iHS data frame are displayed below using the following R command:

#### head(ihs.cgu\$iHS)

```
CHR POSITION
                               iHS -log10(p-value)
> F0100190
                 113642 -0.5582992
                                         0.2390952
> F0100220
                 244699
                         0.2723337
                                         0.1049282
                 369419 0.4810736
> F0100250
             1
                                         0.2003396
> F0100270
                 447278 1.0618710
             1
                                         0.5401640
> F0100280
                 487654 0.8184060
                                         0.3839181
             1
> F0100290
                 524507 -0.3897024
                                         0.1569189
```

2.a matrix summarizing the allele frequency bins. For instance, the five first rows of the ihs.cgu\$frequency.class data frame are displayed below using the following R command:

#### head(ihs.cgu\$frequency.class)

```
Number of SNPs mean of the log(iHHA/iHHD) ratio
> 0.05 - 0.075
                          1635
                                                       0.7286087
> 0.075 - 0.1
                                                       0.5804760
                          1316
> 0.1 - 0.125
                         1478
                                                       0.4710504
> 0.125 - 0.15
                         1593
                                                       0.3720585
> 0.15 - 0.175
                         1078
                                                       0.3263215
> 0.175 - 0.2
                         1325
                                                       0.2721166
               sd of the log(iHHA/iHHD) ratio
> 0.05 - 0.075
                                     0.6457742
> 0.075 - 0.1
                                     0.5556798
> 0.1 - 0.125
                                     0.5079392
> 0.125 - 0.15
                                     0.4708235
> 0.15 - 0.175
                                     0.4524270
> 0.175 - 0.2
                                     0.4533404
```

## 3.1.3 Manhattan plot of the results: the function ihsplot()

The ihsplot() function allows to draw a Manhattan plot of the Whole Genome scan results as stored in the list object produced by the function ihh2ies(). Various options are available to modify the graphical display (see ?ihsplot).

```
ihsplot(ihs.cgu,plot.pval=TRUE,ylim.scan=2,main="iHS (CGU cattle breed)")
```

#### 3.2 The Rsb-based pairwise population test

#### 3.2.1 Definition

For a given SNP s, let

$$LRiES(s)^{Tang} = \log \left( \frac{iES_{pop1}(s)^{Tang}}{iES_{pop2}(s)^{Tang}} \right)$$

represent the log-ratio of the  $iES_{pop1}(s)^{Tang}$  and  $iES_{pop2}(s)^{Tang}$  computed in the pop1 and pop2 populations (see 2.1.4).

The Rsb for a given focal SNP is then defined as the standardized LRiES(s)<sup>Tang</sup> (TANG et al. 2007):

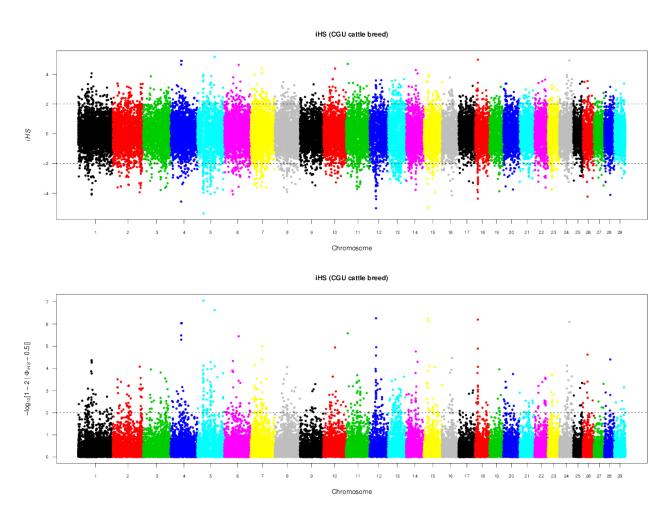


Figure 3: Graphical output for the ihsplot() function

$$rSB(s) = \frac{LRiES(s)^{Tang} - med_{LRiES^{Tang}}}{\sigma_{LRiES^{Tang}}}$$
(4)

where  $\text{med}_{\text{LRiES}^{\text{Tang}}}$  and  $\sigma_{\text{LRiES}^{\text{Tang}}}$  represent the median and standard deviation of the LRiES(s)<sup>Tang</sup> computed over all the analyzed SNPs. Note that the median is used instead of the mean because it is less sensitive to extreme data points (TANG et al. 2007). More importantly, it should be noticed that the information about the ancestral and derived status of alleles at the focal SNP is not needed.

As for the *iHS* (see 3.1.1), Rsb is constructed to have an approximately standard Gaussian distribution and may further be transformed into  $p_{rSB}$ :

$$p_{\text{rSB}} = -\log_{10} \left( 1 - 2|\Phi \left( \text{rSB} \right) - 0.5| \right) \tag{5}$$

where  $\Phi(x)$  represents the Gaussian cumulative distribution function. Assuming most of the genotyped SNPs behave neutrally (i.e., the genome-wide empirical Rsb distribution is a fair approximation of their corresponding neutral distributions),  $p_{\rm rSB}$  might thus be interpreted as a two-sided P-value (in a  $-\log_{10}$  scale) associated to the neutral hypothesis of no selection. Alternatively,  $p_{\rm rSB}$  might also be computed (GAUTIER and NAVES 2011):

$$pt_{rSB} = -\log_{10}(|\Phi(rSB)|) \tag{6}$$

 $p_{rSB}$  and  $p_{rSB}$  might then be interpreted as a one-sided P-value (in a  $-\log_{10}$  scale) allowing the identification of those sites displaying a significantly high extended haplotype homozygosity in population pop2 (represented in the denominator of the corresponding LRiES) relatively to the pop1 reference population.

#### 3.2.2 The function ies2rsb()

The ies2rsb() function allows to compute Rsb using two data frames containing the iES statistics for each of the two populations considered in the same format as the one obtained after running the scan\_hh() function (see 2.4). For instance, to perform a genome scan one might first run for each population scan\_hh() in turn on haplotype data from each chromosome and concatenate the resulting matrices. In the following example, we assume that the haplotype files are named as hap\_chr\_i.pop1 and hap\_chr\_i.pop2 where i is the chromosome number (going from 1 to 29), the suffixes pop1 and pop2 indicate the population of origin and the SNP information file is named snp.info. The R code below then generates two data frames (wg.res.pop1 and wg.res.pop2) containing the results from all SNPs in the appropriate format to compute Rsb with the ies2rsb() function:

```
for(i in 1:29){
  hap_file=paste("hap_chr_",i,".pop1",sep="")
  data<-data2haplohh(hap_file="hap_file","snp.info",chr.name=i)
  res<-scan_hh(data)
  if(i==1){wg.res.pop1<-res}else{wg.res.pop1<-rbind(wg.res.pop1,res)}
  hap_file=paste("hap_chr_",i,".pop2",sep="")
  data<-data2haplohh(hap_file="hap_file","snp.info",chr.name=i)
  res<-scan_hh(data)
  if(i==1){wg.res.pop2<-res}else{wg.res.pop2<-rbind(wg.res.pop2,res)}
}
wg.rsb<-ies2rsb(wg.res.pop1,wg.res.pop2)</pre>
```

As a matter of illustration, one may consider results from a similar genome scan (Gautier and Naves 2011) provided as example data sets and compute for each SNP the Rsb between the CGU and EUT populations as follows:

```
data(wgscan.cgu) ; data(wgscan.eut)
## results from a genome scan (44,057 SNPs) see ?wgscan.eut and ?wgscan.cgu for details
cguVSeut.rsb<-ies2rsb(wgscan.cgu,wgscan.eut,"CGU","EUT")</pre>
```

The resulting object <code>cguVSeut.rsb</code> is a data frame with of SNP Rsb (and corresponding P-Values assuming Rsb are normally distributed under the neutral hypothesis). Note that either bilateral (default) or unilateral might be performed (<code>method</code> argument). The five first rows of the <code>cguVSeut.rsb</code> data frame are displayed below using the following R command:

#### head(cguVSeut.rsb)

```
CHR POSITION Rsb (CGU vs. EUT) -log10(p-value) [bilateral]
> F0100190
             1
                  113642
                                 -0.3398574
                                                               0.13432529
> F0100220
                  244699
                                 -1.0566283
                                                               0.53658299
             1
> F0100250
                                 -0.1468326
                                                               0.05390941
                  369419
> F0100270
                  447278
                                 -1.8191608
                                                               1.16186336
                                 -0.2193069
> F0100280
             1
                  487654
                                                               0.08280392
> F0100290
                  524507
                                 -0.7941300
                                                               0.36945032
```

#### 3.2.3 Manhattan plot of the results: the function rsbplot()

The rsbplot() function allows to draw a Manhattan plot of the Whole Genome scan results as stored in the data frame produced by the function ies2rsb(). Various options are available to modify the graphical display (see ?rsbplot). As an example, the Figure 4 below provides the output of the function rsbplot for the xpEHH computed above across the CGU and EUT populations (see 3.2.2). Figure 4 was drawn using the following R code:

```
rsbplot(cguVSeut.rsb,plot.pval=TRUE)
```

#### 3.3 The *xpEHH*-based pairwise population test

#### 3.3.1 Definition

The xpEHH statistics (SABETI et al. 2007) is similar to the Rsb except that it is based on the  $iES_{pop2}(s)^{Sab}$  (instead of  $iES_{pop2}(s)^{Tang}$ ) estimator of the iES (see 2.1.4). Hence, for or a given SNP s, let

$$LRiES(s)^{Sab} = \log \left( \frac{iES_{pop1}(s)^{Sab}}{iES_{pop2}(s)^{Sab}} \right)$$

represent the log-ratio of the  $iES_{pop1}(s)^{Sab}$  and  $iES_{pop2}(s)^{Sab}$  computed in the pop1 and pop2 populations (see 2.1.4).

The xpEHH for a given focal SNP is then defined as the standardized LRiES(s)<sup>Sab</sup> (SABETI et al. 2007):

$$rSB(s) = \frac{LRiES(s)^{Sab} - med_{LRiES^{Sab}}}{\sigma_{LRiES^{Sab}}}$$
(7)

where  $\text{med}_{\text{LRiES}^{\text{Sab}}}$  and  $\sigma_{\text{LRiES}^{\text{Sab}}}$  represent the median and standard deviation of the LRiES(s)<sup>Sab</sup> computed over all the analyzed SNPs. More importantly, it should be noticed that the information about the ancestral and derived status of alleles at the focal SNP is not needed.

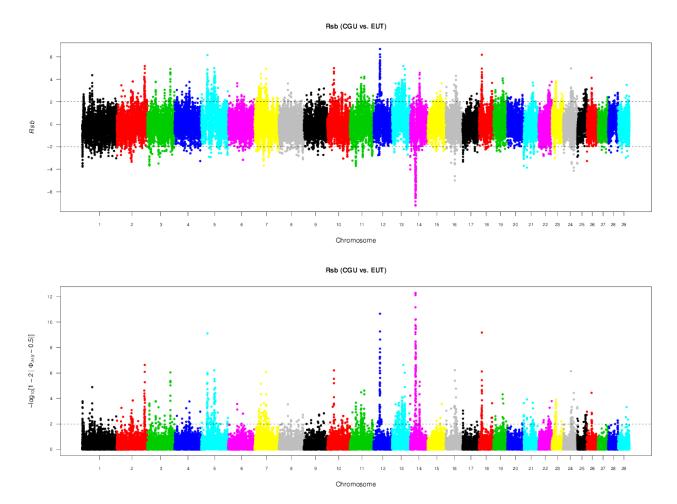


Figure 4: Graphical output for the rsbplot() function

As for the *iHS* (see 3.1.1) and *Rsb*, xpEHH is constructed to have an approximately standard Gaussian distribution and may further be transformed into  $p_{xpEHH}$ :

$$p_{\text{xpEHH}} = -\log_{10} \left( 1 - 2|\Phi\left(\text{xpEHH}\right) - 0.5| \right)$$
 (8)

where  $\Phi\left(x\right)$  represents the Gaussian cumulative distribution function. Assuming most of the genotyped SNPs behave neutrally (i.e., the genome-wide empirical xpEHH distribution is a fair approximation of their corresponding neutral distributions),  $p_{\text{xpEHH}}$  might thus be interpreted as a two-sided P-value (in a  $-\log_{10}$  scale) associated to the neutral hypothesis of no selection. Alternatively,  $p_{\text{xpEHH}}$  might also be computed (Gautier and Naves 2011):

$$p_{\text{xpEHH}} = -\log_{10} \left( |\Phi \left( \text{xpEHH} \right)| \right) \tag{9}$$

 $p'_{\text{xpEHH}}$  and  $p'_{\text{xpEHH}}$  might then be interpreted as a one-sided P-value (in a  $-\log_{10}$  scale) allowing the identification of those sites displaying a significantly high extended haplotype homozygosity in population pop2 (represented in the denominator of the corresponding LRiES) relatively to the pop1 reference population.

#### 3.3.2 The function ies2xpehh()

The ies2xpehh() function allows to compute xpEHH using two data frames containing the iES statistics for each of the two populations considered in the same format as the one obtained after running the scan\_hh()

function (see 2.4). For instance, to perform a genome scan one might first run for each population scan\_hh() in turn on haplotype data from each chromosome and concatenate the resulting matrices. In the following example, we assume that the haplotype files are named as hap\_chr\_i.pop1 and hap\_chr\_i.pop2 where i is the chromosome number (going from 1 to 29), the suffixes pop1 and pop2 indicate the population of origin and the SNP information file is named snp.info. The R code below then generates two data frames (wg.res.pop1 and wg.res.pop2) containing the results from all SNPs in the appropriate format to compute Rsb with the ies2rsb() function:

```
for(i in 1:29){
   hap_file=paste("hap_chr_",i,".pop1",sep="")
   data<-data2haplohh(hap_file="hap_file","snp.info",chr.name=i)
   res<-scan_hh(data)
   if(i==1){wg.res.pop1<-res}else{wg.res.pop1<-rbind(wg.res.pop1,res)}
   hap_file=paste("hap_chr_",i,".pop2",sep="")
   data<-data2haplohh(hap_file="hap_file","snp.info",chr.name=i)
   res<-scan_hh(data)
   if(i==1){wg.res.pop2<-res}else{wg.res.pop2<-rbind(wg.res.pop2,res)}
}
wg.xpehh<-ies2xpehh(wg.res.pop1,wg.res.pop2)</pre>
```

As a matter of illustration, one may consider results from a similar genome scan (Gautier and Naves 2011) provided as example data sets and compute for each SNP the *xpEHH* between the CGU and EUT populations as follows:

```
data(wgscan.cgu) ; data(wgscan.eut)
## results from a genome scan (44,057 SNPs) see ?wgscan.eut and ?wgscan.cgu for details
cguVSeut.xpehh<-ies2xpehh(wgscan.cgu,wgscan.eut,"CGU","EUT")</pre>
```

The resulting object cguVSeut.xpehh is a data frame with of SNP xpEHH (and corresponding P-values assuming xpEHH are normally distributed under the neutral hypothesis). Note that either bilateral (default) or unilateral might be performed (method argument). The five first rows of this data frame are displayed below using the following R command:

```
head(cguVSeut.xpehh)
```

```
CHR POSITION XPEHH (CGU vs. EUT) -log10(p-value) [bilateral]
> F0100190
             1
                 113642
                                  -0.5555841
                                                                 0.2377002
> F0100220
                 244699
                                  -0.7516166
                                                                 0.3445910
             1
                                  -0.8885736
> F0100250
                 369419
                                                                 0.4268588
> F0100270
                 447278
                                  -0.3470522
                                                                 0.1375394
             1
> F0100280
                 487654
                                  -0.9182772
                                                                 0.4455426
> F0100290
                 524507
                                  -0.7521031
                                                                 0.3448721
```

#### 3.3.3 Manhattan plot of the results: the function xpehhplot()

The xpehhplot() function allows to draw a Manhattan plot of the Whole Genome scan results as stored in the data frame produced by the function <code>ies2xpehh()</code>. Various options are available to modify the graphical display (see <code>?xpehhplot</code>). As an example, the Figure 5 below provides the output of the function <code>xpehhplot</code> for the <code>xpEHH</code> computed above across the CGU and EUT populations (see 3.3.2). Figure 5 was drawn using the following R code:

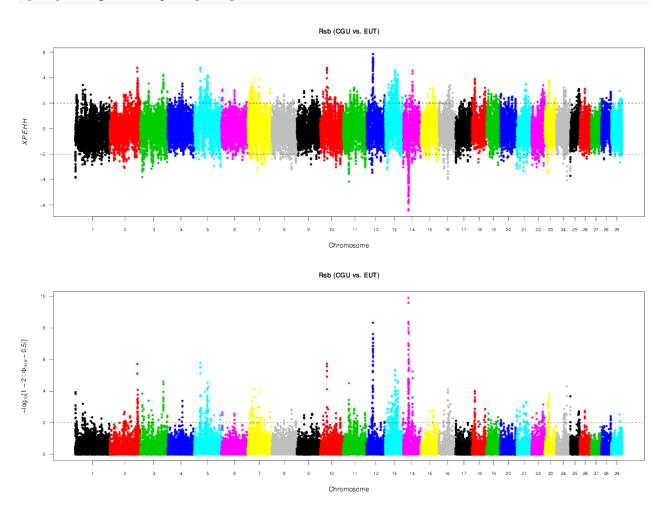


Figure 5: Graphical output for the xpehhplot() function

#### 3.3.4 xpEHH vs. Rsb comparison:

A plot of the xpEHH against Rsb estimates across the CGU and EUT populations (see 3.3.2 and 3.2.2 respectively) is represented in the Figure 6 below. This figure was generated using the following R code:

## 3.4 Visual inspection of the standardized scores distribution: the function distribution()

The distribution allows to easily visualize the distributions of the standardized scores (either iHS, Rsb or xpEHH) and compare them to the standard Gaussian distribution. As an example, the Figure 7 below provides the output the function distribution when considering the iHS estimates obtained for the CGU population (see 3.1.2) using the following R code:

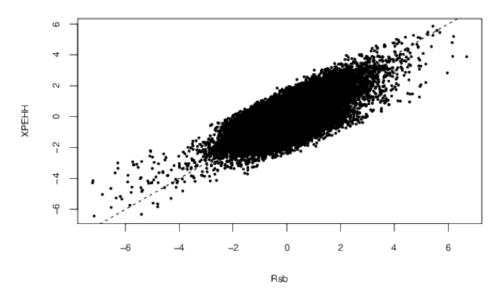


Figure 6: Comparison of the XPEHH and Rsb estimates across the CGU and EUT populations

distribplot(ihs.cgu\$iHS[,3],xlab="iHS")

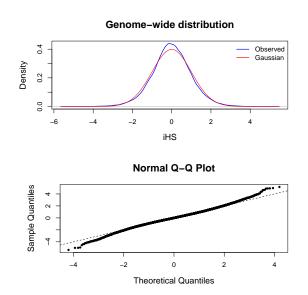
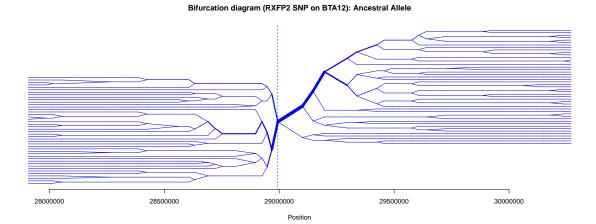


Figure 7: Graphical output for the function distribplot

# 4 Visualizing haplotype structure around a core allele: the function bifurcation.diagram()

The function bifurcation.diagram() function draws haplotype bifurcation diagrams (SABETI et al. 2002) that allow to better understand the origin of an observed footprints of selection. Such diagrams indeed consist in plotting the breakdown of LD at increasing distances from the core allele at the selected focal SNPs. The root (focal SNP) of each diagram is the core allele and is here identified by a vertical dashed line. The diagram is bi-directional, portraying both centromere-proximal and centromere-distal LD. Moving in

one direction, each marker is an opportunity for a node; the diagram either divides or not based on whether both or only one allele is present. Thus the breakdown of LD on the core haplotype background is portrayed at progressively longer distances. The thickness of the lines corresponds to the number of samples with the indicated long-distance haplotype. Several options are available to modify the aspect of the plots (see command ?bifurcation.diagram) As a matter of illustration, Figure 8 shows the bifurcation diagrams for both the derived and ancestral alleles at the  $456^{\rm th}$  SNP on BTA12 CGU haplotypes. This SNP displayed a strong signal of selection (using both iHS and Rsb statistics) and is located closed (<5kb) to a strong candidate genes involved in horn development (Gautier and Naves 2011). Figure 8 was obtained with the following R code:



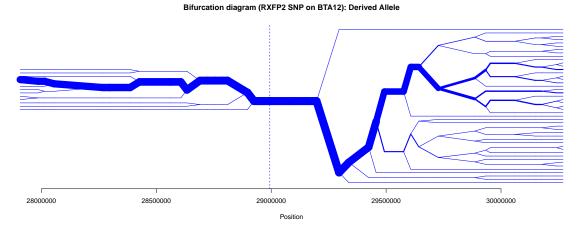


Figure 8: Graphical output for the function bifurcation.diagram()

## References

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