Testing and Modeling Genotypic Disequilibria

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1 Introduction

In a diploid, sexually reproducing species, at a locus where there are two alleles A and a, the possible genotypes are AA, Aa and aa. In a population of size N, with p the frequency of the A allele and q the frequency of the a allele, the expected numbers under Hardy-Weinberg equilibrium are $NP_{AA} = Np^2$ for the AA genotype, $NP_{Aa} = 2Npq$ for the Aa genotype, and $NP_{aa} = Nq^2$ for the aa genotype. Writing $m = \log(Np^2)$ and $\log(q/p) = m_a$, the logarithms of the frequencies may be written:

$$\log(Np^2) = m \tag{1}$$

$$\log(2Npq) = m + \log(2) + m_a \tag{2}$$

$$\log(Nq^2) = m + 2m_a \tag{3}$$

Thus the model is loglinear, and can be fitted as a generalized linear model with poisson error and offset log(2) for the heterozygote. For example:

```
> obs <- c(AA=147, Aa=78, aa=17)
> oset <- c(0, log(2), 0)
> ma <- c(0,1,2)
> hw.glm <- glm(obs ~ ma, family=poisson, offset=oset)
> summary(hw.glm)
Call:
glm(formula = obs ~ ma, family = poisson, offset = oset)
Coefficients:
           Estimate Std. Error z value Pr(>|z|)
(Intercept) 4.96256
                       0.08137 60.99
                                          <2e-16
           -1.20039
                       0.10778 -11.14
                                          <2e-16
(Dispersion parameter for poisson family taken to be 1)
    Null deviance: 149.3527
                            on 2
                                  degrees of freedom
Residual deviance:
                    2.0436
                            on 1
                                  degrees of freedom
AIC: 23.751
Number of Fisher Scoring iterations: 4
```

The function hwde() may also be used to fit this model, at the same time introducing a further "disequilibrium" term. The default output is the analysis of deviance table.

```
> hwdat <- data.frame(Observed=c(147,78,17), locus1=c("AA","Aa","aa"))
```

Now call the function.

```
> library(hwde)
> hwde(data=hwdat)
[1] "Analysis of Deviance Table"
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1
                 149.353
+a
            1
                    2.044
                              147.309
                                         <2e-16
            0
                    0.000
                                2.044
                                         0.1528
+aa
```

The disequilibrium term has the form

$$m_{aa} = \log \frac{4P_{AA}P_{aa}}{P_{Aa}^2}$$

Notice that the parameters m_a and m_{aa} have been abbreviated, in the computer output, to a and aa respectively. The parameter m models the reference or baseline level, and is estimated by the intercept term.

To obtain estimates of parameters, including the disequilibrium parameter m_{aa} , do the following:

```
> data.df <- hwde(data = hwdat)$data.df
[1] "Analysis of Deviance Table"
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
            2
                 149.353
+a
            1
                   2.044
                          1
                              147.309
                                        <2e-16
            0
                   0.000
                           1
                                2.044
                                        0.1528
+aa
> names(data.df)
[1] "obs"
           "data" "a"
                          "aa"
                                 "oset"
> summary(glm(obs ~ a + aa, offset=oset, family=poisson, data=data.df))$coef
             Estimate Std. Error z value
                                                Pr(>|z|)
(Intercept) 1.8332133
                       0.2425356 7.558532 4.076429e-14
                       0.2676640 1.955793 5.048954e-02
a
            0.5234955
            1.1102283
                       0.3419186 3.247055 1.166060e-03
```

Note again that the intercept estimates m, and that aa is the additive version of the disequilibrium parameter.

We leave till later detailed information on the use of hwde(), including details on how to obtain fitted values and residuals.

1.1 Several different populations

If there several different populations, there must be a parameter (by default assumed to have the name Population), that accounts for different population sizes. In the code, this translates to a main effect gp in the log-linear model. Additionally, there may be different values for m_a and m_{aa} in the different populations.

A second locus requires the parameters m_b and m_{bb} for that locus. Additionally, parameters may be required that model quantities that, in the loglinear model, have the role of interactions between the two loci. Huttley and Wilson (2000) introduce the multiplicative versions of the following parameters:

 s_{ab} , the "sum of digenic disequilibria for the total sample"

 q_{ab} , the "product of digenic disequilibria for the total sample"

 m_{aab} and m_{abb} , which are "trigenic disequilibria terms for the total sample"

In the usual case where phase for double heterozygotes is unknown and only nine genotypic classes can be distinguished, no degrees of freedom remain that might be used to estimate a quadrigenic disequilibrium term.

As noted above, the formulae in Huttley and Wilson (2000) give the multiplicative equivalents of these terms, using upper case letters. The additive versions used here (e.g., they have M_A where I have $m_a = \log(M_A)$) use the corresponding lower case letters. Note however that in the second column on p.2131 of Huttley and Wilson, in the equations for $\ln P_{Ab}^{AB}$ and $\ln P_{aB}^{AB}$, $\ln Q_{AB}^2$ should be, in each case, $\ln Q_{AB}$. The equations are given correctly in Weir and Wilson (1986), though with slight changes of notation. See also Weir (1996).

The function allows an arbitrary number of loci. Terms s_{ab} , q_{ab} , m_{abb} and m_{aab} are fitted for every pair of loci. Terms that correspond to second (or, with > 3 loci, higher order) interactions contribute, in the present version of the code, to the residual. Try

> hwde(data=mendelABC, loci=c("seedshape","cotylcolor","coatcolor"))

2 Details of Use of hwde()

First recall the simple example that was described above. The data were entered, from the keyboard, into a data frame hwdat that had the form:

Observed locus1

147 AA78 Aa17 aa

The coding used in the column headed locus1 can be varied; any two characters may be used for the alleles. With the column names that are shown, the corresponding parameter settings for the function hwde() can be left at their defaults.

An alternative is to enter the data, exactly as displayed above (though the spacing is immaterial), into a file. If the file is called **hw.txt** and is placed in the working directory, then it can be read in with:

```
> hwdat <- read.table("hw.txt", header=TRUE)</pre>
```

If there is a second locus, the default name is locus2. The default name for any third locus is locus3, etc. Where there is a column that has codes for different populations, the default name is Population.

Example - two populations and two loci

With this introduction, we move directly to data, with two populations and two loci, that are suited to fitting all the parameters that the function currently allows, i.e., m_{aa} , m_{bb} , m_{cc} , s_{ab} , s_{ac} , s_{bc} , q_{ab} , q_{ac} , q_{bc} , m_{abb} , m_{acc} , m_{bcc} , m_{aab} , m_{aac} , m_{bbc} .

Data (Mourant et al, 1976) are:

locus1	locus2	Observed
MM	SS	91
MM	Ss	147
MM	SS	85
MN	SS	32
MN	Ss	78
	MM MM MM MN	MM Ss MM ss MN SS

Indian	MN	SS	75
Indian	NN	SS	5
Indian	NN	Ss	17
Indian	NN	SS	7
Irish	MM	SS	121
Irish	MM	Ss	248
Irish	MM	SS	164
Irish	MN	SS	53
Irish	MN	Ss	422
Irish	MN	SS	375
Irish	NN	SS	9
Irish	NN	Ss	65
Irish	NN	SS	241

Assuming that this is stored in a file **IndianIrish.txt**, we can read in the data and do the analysis thus:

> IndianIrish <- read.table("IndianIrish.txt", header=TRUE)</pre>

> hwde(data=IndianIrish)

[1] "Analysis of Deviance Table"

	Resid.	Df	Resid.	Dev	Df	Deviance	Pr(>Chi)
1		17	1724	.07			
+gp		16	1090	.41	1	633.66	< 2.2e-16
+(a+b)		14	486	.72	2	603.69	< 2.2e-16
+(aa+bb)		12	480	.31	2	6.41	0.040653
+sab		11	463	3.76	1	16.55	4.738e-05
+qab		10	218	3.42	1	245.34	< 2.2e-16
+(abb+aab)		8	217	.15	2	1.28	0.527693
+gp:(a+b)		6	37	.94	2	179.21	< 2.2e-16
+gp:(aa+bb)		4	35	.46	2	2.48	0.288680
+gp:sab		3	26	. 29	1	9.16	0.002471
+gp:qab		2	5	.94	1	20.36	6.425e-06
+gp:(abb+aab)		0	C	00.0	2	5.94	0.051410

The above is the compact default output, in which terms that are at the same level of a hierarchy are grouped. For a first pass through the data, this may be the preferred output. A form of output in which each term correspods to a single degree of freedom is available by using the parameter setting group.terms=FALSE, i.e.,

> hwde(data=IndianIrish, group.terms=FALSE)

difference from the last previous Residual Deviance term that is marked with an r (= reference) as the first character in the row in which it appears.

The estimates of parameters in the maximal (or, with appropriate modification, any other) model can be extracted thus:

[1] "Analysis of Deviance Table"

	Resid.	Df	Resid.	Dev	Df	Deviance	Pr(>Chi)
1		26	15.3	3272			
+(a+b+c)		23	14.5	5604	3	0.7669	0.8574
+(aa+bb+cc)		20	13.5	5445	3	1.0158	0.7974
+(sab+sac+sbc)		17	9.7	7381	3	3.8064	0.2831
+(qab+qac+qbc)		14	7.4	4493	3	2.2889	0.5147
+(abb+acc+bcc+aab+aac+bbc)		8	5.3	1362	6	2.3130	0.8888

```
> models <- II.hwde$models
> maxmodel <- models[[length(models)]]</pre>
> summary(maxmodel)$coef
                Estimate Std. Error
                                         z value
                                                     Pr(>|z|)
(Intercept)
             2.119196819
                          0.2546858 8.32082671 8.735188e-17
             0.015218967
                          0.2645433
                                     0.05752922 9.541236e-01
a
             0.245726216
                          0.2659182
                                      0.92406689 3.554515e-01
b
             0.143067467
                          0.2650647
                                      0.53974548 5.893726e-01
С
                                     0.05077273 9.595066e-01
             0.021854471
                          0.4304371
aa
                          0.4172846 -0.42654686 6.697094e-01
bb
            -0.177991441
            -0.157633137
                          0.4284390 -0.36792438 7.129296e-01
СС
             0.075610484 0.2422532
                                     0.31211346 7.549543e-01
            -0.031663422
                          0.2357458 -0.13431172 8.931561e-01
sac
            -0.247841900
                          0.2445746 -1.01335891 3.108888e-01
sbc
qab
            -0.245328249
                          0.3637890 -0.67436970 5.000763e-01
qac
             0.186752307
                          0.3524097
                                     0.52992949 5.961608e-01
            -0.105419665
                          0.3673842 -0.28694664 7.741532e-01
qbc
             0.020721843
                          0.2250188
                                     0.09208938 9.266270e-01
abb
acc
            -0.083962005
                          0.2255440 -0.37226444 7.096960e-01
                          0.2300915 -0.50911316 6.106729e-01
bcc
            -0.117142586
aab
             0.230215504
                          0.2296021
                                     1.00267155 3.160194e-01
                          0.2259632 -0.87995213 3.788852e-01
aac
            -0.198836824
bbc
            -0.008483307
                          0.2267579 -0.03741130 9.701571e-01
```

3 Obtaining Additional Output

By default, the function returns (invisibly) a list with two elements. The first holds the analysis of variance table. The second holds the data and contrast terms that are required for fitting the various models. For example:

```
> hwdat.hw <- hwde(data=hwdat)
[1] "Analysis of Deviance Table"
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1
            2
                  149.353
+a
             1
                    2.044
                              147.309
                                         <2e-16
+aa
            0
                    0.000
                                 2.044
                                         0.1528
> names(hwdat)
[1] "Observed" "locus1"
> hwdat.hw$data.df
   obs data a aa oset
AA 147
         AA 2
               1
                     1
Aa 78
         Aa 1
               0
                     2
   17
         aa O
                     1
```

The following illustrates the direct use of the information in hwdat.hw\$data.df, giving the user complete control over the models that are fitted.

```
> data.df <- hwdat.hw$data.df
> model1 <- glm(obs ~ a, family=poisson, data=data.df, offset=log(oset))
> model2 <- glm(obs ~ a+aa, family=poisson, data=data.df, offset=log(oset))
> model1
```

```
Call: glm(formula = obs ~ a, family = poisson, data = data.df, offset = log(oset))
Coefficients:
(Intercept)
                       a
      2.562
                   1.200
Degrees of Freedom: 2 Total (i.e. Null); 1 Residual
Null Deviance:
                           149.4
Residual Deviance: 2.044
                                  AIC: 23.75
  Here is the output data frame for the IndianIrish data.
> II.hw <- hwde(data=IndianIrish, aovtable.print=FALSE)
> dataII.df <- II.hw$data.df</pre>
> dataII.df
   obs
           gp locus1 locus2 a b aa bb sab qab abb aab oset
1
   91 Indian
                         SS 2 2 1
                                    1
                                         0
                                             2
2
  147 Indian
                         Ss 2 1
                                             1
                                                     1
3
   85 Indian
                         ss 2 0
                                             0
                                                 0
                                                     0
                  MM
                                 1
                                    0
                                         0
                                                          1
                                                          2
   32 Indian
                  MN
                         SS 1 2 0
                                                     0
4
                                     1
                                         0
                                             1
                                                 1
5
   78 Indian
                  MN
                         Ss 1 1
                                  0
                                     0
                                             0
                                                 0
                                                     0
6
   75 Indian
                  MN
                         ss 1 0
                                  0
                                             0
7
                  NN
                         SS 0 2
                                 0
                                         0
                                             0
                                                 0
                                                     0
    5 Indian
                                     1
                                                          1
                         Ss 0 1
                                                 0
    17 Indian
                  NN
                                 0
                                    0
                                         0
                                             0
                                                     0
8
9
    7 Indian
                  NN
                         ss 0 0
                                 0
                                    0
                                         0
                                             0
                                                 0
                                                     0
10 121 Irish
                  MM
                         SS 2 2
                                  1
                                     1
                                         0
                                             2
                                                 2
                                                     2
                                                          1
                         Ss 2 1
11 248 Irish
                  MM
                                    0
                                         0
                                             1
                                                 0
                                                     1
                                                          2
                                  1
                         ss 2 0
12 164
        Irish
                  MM
                                  1
                                     0
                                         0
                                             0
                                                 0
                                                     0
                                                          1
13 53
        Irish
                  MN
                         SS 1 2
                                 0
                                         0
                                                     0
                                                          2
                                             1
                                                 1
14 422
        Irish
                  MN
                         Ss 1 1
                                  0
                                    0
                                             0
                                                 0
                                                     0
                                                          4
                                         1
15 375 Irish
                  MN
                                                 0
                                                          2
                         ss 1 0
                                  0
                                    0
                                         0
                                             0
                                                     0
16
   9 Irish
                  NN
                         SS 0 2 0
                                     1
                                         Ω
                                             0
                                                 0
                                                     Λ
                                                          1
17 65 Irish
                  NN
                         Ss 0 1
                                             0
                                                          2
```

The user can now fit any sequence of models that may be required. For example, the user may wish to a sequence of models that is different from the sequence fitted by hwde().

Further control is available by supplying values for the parameters termlist and refmodel. For example, the default action with the data frame hwdat is equivalent to:

```
> hwde(termlist=c("+a","+aa"), refmodel=c(1,2), data=hwdat)
```

ss 0 0 0

```
[1] "Analysis of Deviance Table"
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1
            2
                 149.353
                    2.044
                              147.309
+a
            1
                           1
                                         <2e-16
            0
                    0.000
                          1
                                2.044
                                         0.1528
+aa
```

NN

18 241 Irish

In refmodel, 1 refers to the model that has constant term only.

The first six models can be fitted to the data frame IndianIrish by setting:

```
> hwde(termlist=c("+gp","+a","+b","+a+b","+aa"), refmodel=c(1,2,2,2,5),
+ data=IndianIrish)
```

```
[1] "Analysis of Deviance Table"
     Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1
             17
                   1724.07
                   1090.41
                                  633.66
             16
                             1
                                            <2e-16
+gp
             15
                    853.73
                                  236.68
                                            <2e-16
                             1
+a
             15
                    723.40
                                  367.02
+b
                             1
+a+b
             14
                    486.72
                             2
                                  603.69
                                            <2e-16
             13
                    485.59
                                           0.2872
+aa
                                    1.13
```

Extraction of the sequence of fitted models

A further possibility, with the parameter setting keep.models=TRUE, is to include the full sequence of models that have been fitted in the list that is returned by the function. For example:

```
> hwdat.hw <- hwde(data=hwdat, keep.models=TRUE)
[1] "Analysis of Deviance Table"
    Resid. Df Resid. Dev Df Deviance Pr(>Chi)
            2
                 149.353
1
+a
            1
                   2.044
                          1
                             147.309
                                        <2e-16
            0
                   0.000
                                2.044
                                        0.1528
+aa
> hwdat.hw$models[[2]]
                                 # The Hardy-Weinberg model
     glm(formula = obs ~ a, family = poisson, data = data.df, offset = log(oset))
Coefficients:
(Intercept)
                       a
      2.562
                   1.200
Degrees of Freedom: 2 Total (i.e. Null); 1 Residual
Null Deviance:
                           149.4
Residual Deviance: 2.044
                                  AIC: 23.75
> fitted(hwdat.hw$models[[2]])
       AA
                 Aa
142.95868
          86.08264 12.95868
```

The function fitted() can be replaced by any of the functions (coef(), resid(), predict(), etc.) that are available for use with a glm model object. Note that there are several different choices of residuals, with deviance residuals as the default. For the IndianIrish data there are, with the parameter setting group.terms=FALSE, 24 models from which to choose. Choose carefully!

4 Exact Hardy-Weinberg Test

The function hwexact(), supplied by Randall Johnson, does an exact test for Hardy-Weinberg equilibrium, conditional on the observed relative numbers of the two alleles. The only case implemented is for a single population and single locus. The algorithm is described in Wigginton et al (2005).

5 References

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