# 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)</pre>
> 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
                       0.10778 -11.14
ma
            -1.20039
                                          <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"))</pre>
```

Now call the function.

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

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	AA
78	Aa
17	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:

Population	locus1	locus2	Observed	l
------------	--------	--------	----------	---

Indian	MM	SS	91
Indian	MM	Ss	147
Indian	MM	SS	85
Indian	MN	SS	32
Indian	MN	Ss	78

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 Dev:	ianc	e Table	э"		
	Resid.	Df 1	Resid.	Dev	Df	Deviance
1		17	1724	1.07		
+gp		16	1090	0.41	1	633.66
+(a+b)		14	486	5.72	2	603.69
+(aa+bb)		12	480	0.31	2	6.41
+sab		11	463	3.76	1	16.55
+qab		10	218	3.42	1	245.34
+(abb+aab)		8	217	7.15	2	1.28
+gp:(a+b)		6	37	7.94	2	179.21
+gp:(aa+bb)		4	35	5.46	2	2.48
+gp:sab		3	26	5.29	1	9.16
+gp:qab		2	Ę	5.94	1	20.36
+gp:(abb+aab)		0	(	0.00	2	5.94

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 corresponds 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
1		26	15.3272		
+(a+b+c)		23	14.5604	3	0.7669
+(aa+bb+cc)		20	13.5445	3	1.0158
+(sab+sac+sbc)		17	9.7381	3	3.8064
+(qab+qac+qbc)		14	7.4493	3	2.2889
+(abb+acc+bcc+aab+aac+bbc)		8	5.1362	6	2.3130

```
> models <- II.hwde$models
> maxmodel <- models[[length(models)]]
> summary(maxmodel)$coef
```

	Estimate	Std. Error	z value	$\Pr( z )$
(Intercept)	2.119196819	0.2546858	8.32082671	8.735188e-17
a	0.015218967	0.2645433	0.05752922	9.541236e-01
b	0.245726216	0.2659182	0.92406689	3.554515e-01
С	0.143067467	0.2650647	0.53974548	5.893726e-01
aa	0.021854471	0.4304371	0.05077273	9.595066e-01
bb	-0.177991441	0.4172846	-0.42654686	6.697094e-01
сс	-0.157633137	0.4284390	-0.36792438	7.129296e-01
sab	0.075610484	0.2422532	0.31211346	7.549543e-01
sac	-0.031663422	0.2357458	-0.13431172	8.931561e-01
sbc	-0.247841900	0.2445746	-1.01335891	3.108888e-01
qab	-0.245328249	0.3637890	-0.67436970	5.000763e-01
qac	0.186752307	0.3524097	0.52992949	5.961608e-01
qbc	-0.105419665	0.3673842	-0.28694664	7.741532e-01
abb	0.020721843	0.2250188	0.09208938	9.266270e-01
acc	-0.083962005	0.2255440	-0.37226444	7.096960e-01
bcc	-0.117142586	0.2300915	-0.50911316	6.106729e-01
aab	0.230215504	0.2296021	1.00267155	3.160194e-01
aac	-0.198836824	0.2259632	-0.87995213	3.788852e-01
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)</pre>
[1] "Analysis of Deviance Table"
    Resid. Df Resid. Dev Df Deviance
            2
                 149.353
1
                             147.309
+a
            1
                   2.044
                          1
+aa
            0
                   0.000 1
                                2.044
> 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
aa 17
         aa 0 0
                     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
> dataII.df
```

	obs	gp	locus1	locus2	a	b	aa	bb	sab	qab	abb	aab	oset
1	91	Indian	MM	SS	2	2	1	1	0	2	2	2	1
2	147	Indian	MM	Ss	2	1	1	0	0	1	0	1	2
3	85	Indian	MM	SS	2	0	1	0	0	0	0	0	1
4	32	Indian	MN	SS	1	2	0	1	0	1	1	0	2
5	78	Indian	MN	Ss	1	1	0	0	1	0	0	0	4
6	75	Indian	MN	SS	1	0	0	0	0	0	0	0	2
7	5	Indian	NN	SS	0	2	0	1	0	0	0	0	1
8	17	Indian	NN	Ss	0	1	0	0	0	0	0	0	2
9	7	Indian	NN	SS	0	0	0	0	0	0	0	0	1
10	121	Irish	MM	SS	2	2	1	1	0	2	2	2	1
11	248	Irish	MM	Ss	2	1	1	0	0	1	0	1	2
12	164	Irish	MM	SS	2	0	1	0	0	0	0	0	1
13	53	Irish	MN	SS	1	2	0	1	0	1	1	0	2
14	422	Irish	MN	Ss	1	1	0	0	1	0	0	0	4
15	375	Irish	MN	SS	1	0	0	0	0	0	0	0	2
16	9	Irish	NN	SS	0	2	0	1	0	0	0	0	1
17	65	Irish	NN	Ss	0	1	0	0	0	0	0	0	2
18	241	Irish	NN	SS	0	0	0	0	0	0	0	0	1

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)

```
[1] "Analysis of Deviance Table"
```

	Resid.	$\mathtt{Df}$	Resid.	Dev	$\mathtt{Df}$	Deviance
1		2	149	.353		
+a		1	2	.044	1	147.309
+aa		0	0	.000	1	2.044

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]	"Analys	Tal	ole"		
	Resid.	$\mathtt{Df}$	Resid. Dev	$\mathtt{Df}$	Deviance
1		17	1724.07		
+gp		16	1090.41	1	633.66
+a		15	853.73	1	236.68
+b		15	723.40	1	367.02
+a+b	þ	14	486.72	2	603.69
+aa		13	485.59	1	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)</pre> [1] "Analysis of Deviance Table" Resid. Df Resid. Dev Df Deviance 2 149.353 1 +a 1 2.044 1 147.309 0 0.000 2.044 +aa 1 > hwdat.hw\$models[[2]] # The Hardy-Weinberg model Call: glm(formula = obs ~ a, family = poisson, data = data.df, offset = log(oset)) Coefficients: (Intercept) а 2.562 1.200 Degrees of Freedom: 2 Total (i.e. Null); 1 Residual Null Deviance: 149.4Residual Deviance: 2.044 AIC: 23.75 > fitted(hwdat.hw\$models[[2]]) AA 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

Huttley, G.A. and Wilson, S.R. 2000. Testing for concordant equilibria between population samples. Genetics 156: 2127-2135.

Mourant, A.E., Kopec, A.C. and Domaniewska-Sobczak, K. 1976. The Distribution of the Human Blood Groups and Other Polymorphisms. Oxford University Press.

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Wigginton, J.E., Cutler, D.J. and Abecasis, G.R. 2000. A note on exact tests of Hardy-Weinberg equilibrium. *American Journal of Human Genetics* 76: 887-893.