The procedures in SAS/Genetics (Release 9.1.3)

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Presented by Tõnu Möls Journal Club: 24th March 2009 8 procedures:

Procedure

ALLELE CASECONTROL FAMILY HAPLOTYPE HTSNP INBREED

PSMOOTH TPLOT Macro TPLOT Results (Frame) TPLOT Results (SCL)

Implementation

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ALLELE procedure

Preliminary analyses on genetic marker data. Joint analyses on markers and traits Multinomial distribution of marker alleles Random sampling Indication of marker informativeness

ODS Tables Created by the ALLELE Procedure

ODS Table Name	Description	PROC ALLELE option
MarkerSumm	Marker summary	default
AlleleFreq	Allele frequencies	default
GenotypeFreq	Genotype frequencies	default
LDMeasures	Linkage disequilibrium measures	CORRCOEFF, DELTA, DPRIME, PROPDIFF, or YULESQ

data markers;

input (a1-a10) (\$);
datalines;

в	в	Α	в	в	в	Α	Α	в	в
Α	Α	в	в	Α	в	Α	в	С	С
В	в	Α	Α	в	в	в	в	Α	С
Α	в	Α	в	Α	в	Α	в	Α	в
Α	Α	Α	в	Α	в	в	в	С	С
В	в	Α	Α	Α	в	Α	в	С	С
Α	в	в	в	Α	в	Α	Α	Α	в
Α	в	Α	Α	Α	Α	Α	Α	Α	Α
В	в	Α	Α	Α	Α	Α	в	в	в
Α	в	Α	в	Α	в	в	в	Α	С
Α	Α	Α	в	Α	Α	Α	в	в	С
В	в	Α	в	Α	в	Α	в	Α	С
Α	в	в	в	Α	Α	Α	в	Α	С
В	в	в	в	Α	Α	Α	Α	Α	в
Α	в	Α	Α	Α	в	Α	Α	С	С
Α	в	Α	Α	Α	в	Α	в	С	С
В	в	Α	Α	Α	Α	Α	в	Α	Α
Α	Α	Α	в	Α	Α	Α	в	Α	в
Α	в	Α	Α	Α	Α	в	в	С	С
Α	Α	Α	Α	Α	Α	Α	Α	в	в
Α	в	в	в	Α	Α	Α	Α	С	С
Α	в	Α	в	Α	в	Α	Α	в	в
В	в	Α	в	Α	в	Α	Α	Α	С
Α	в	Α	Α	Α	в	Α	в	Α	С
A	в	в	в	в	в	Α	в	в	в

data markers; input (g1-g5) (\$); datalines;

B/B A/B B/B A/A B/B A/A B/B A/B A/B C/C B/B A/A B/B B/B A/C A/B A/B A/B A/B A/B A/A A/B A/B B/B C/C B/B A/A A/B A/B C/C A/B B/B A/B A/A A/B A/B A/A A/A A/A A/A B/B A/A A/A A/B B/B A/B A/B A/B B/B A/C A/A A/B A/A A/B B/C B/B A/B A/B A/B A/C A/B B/B A/A A/B A/C B/B B/B A/A A/A A/B A/B A/A A/B A/A C/C A/B A/A A/B A/B C/C B/B A/A A/A A/B A/A A/A A/B A/A A/B A/B A/B A/A A/A B/B C/C A/A A/A A/A A/A B/B A/B B/B A/A A/A C/C A/B A/B A/B A/A B/B B/B A/B A/B A/A A/C A/B A/A A/B A/B A/C A/B B/B B/B A/B B/B

,

;

Indiv Alleles Content Square DF ChiSq I Marker1 25 2 0.3714 0.4800 0.4928 0.0169 1 0.8967 1 Marker2 25 2 0.3685 0.3600 0.4872 1.7041 1 0.1918 0 Marker3 25 2 0.3546 0.4800 0.4608 0.0434 1 0.8350 1 Marker4 25 2 0.3648 0.4800 0.4800 0.0000 1 1.0000 1	The ALLELE Procedure									
LocusNumber of IndivNumber of AllelesPolymorph Info ContentHeterozygosityAllelic DiversityChi- SquareDF $Pr >$ ChiSqTMarker12520.37140.48000.49280.016910.89671Marker22520.36850.36000.48721.704110.19180Marker32520.35460.48000.46080.043410.83501Marker42520.36480.48000.48000.000011.00001	Marker Summary									
Locusof Indivof AllelesInto ContentHeterozygosity ContentDiversityChi- SquareDF $Pr >$ ChiSqTMarker12520.37140.48000.49280.016910.89671Marker22520.36850.36000.48721.704110.19180Marker32520.35460.48000.46080.043410.83501Marker42520.36480.48000.48000.000011.00001		Number	Number	Polymorph		Allalia		Test	for HWE	
Marker22520.36850.36000.48721.704110.19180Marker32520.35460.48000.46080.043410.83501Marker42520.36480.48000.48000.000011.00001	Locus				Heterozygosity			DF		Prob Exact
Marker3 25 2 0.3546 0.4800 0.4608 0.0434 1 0.8350 1 Marker4 25 2 0.3648 0.4800 0.4800 0.0000 1 1.0000 1	Marker1	25	2	0.3714	0.4800	0.4928	0.0169	1	0.8967	1.0000
Marker4 25 2 0.3648 0.4800 0.4800 0.0000 1 1.0000 1	Marker2	25	2	0.3685	0.3600	0.4872	1.7041	1	0.1918	0.2262
	Marker3	25	2	0.3546	0.4800	0.4608	0.0434	1	0.8350	1.0000
	Marker4	25	2	0.3648	0.4800	0.4800	0.0000	1	1.0000	1.0000
Marker5 25 3 0.5817 0.4400 0.6552 9.3537 3 0.0249 0	Marker5	25	3	0.5817	0.4400	0.6552	9.3537	3	0.0249	0.0106

$$\text{PIC} = 1 - \sum_{u=1}^{k} \tilde{p}_{u}^{2} - \sum_{u=1}^{k-1} \sum_{v=u+1}^{k} 2\tilde{p}_{u}^{2}\tilde{p}_{v}^{2} \qquad \text{Het} = 1 - \sum_{u=1}^{k} \tilde{P}_{uu} \qquad \text{Div} = 1 - \sum_{u=1}^{k} \tilde{p}_{u}^{2}$$

ODS table of allele frequencies

		Allele	Frequencies	ł	
Locus	Allele	Frequency	Standard Error	95% Confi	idence Limits
Marker1	А	0.4400	0.0711	0.3000	0.5800
Marker1	В	0.5600	0.0711	0.4200	0.7000
Marker2	Λ	0.5800	0.0784	0.4200	0.7400
Marker2	В	0.4200	0.0784	0.2600	0.5800
Marker3	А	0.6400	0.0665	0.5200	0.7600
Marker3	В	0.3600	0.0665	0.2400	0.4800
Marker4	А	0.6000	0.0693	0.4600	0.7400
Marker4	В	0.4000	0.0693	0.2600	0.5400
Marker5	А	0.2800	0.0637	0.1400	0.4200
Marker5	В	0.3000	0.0800	0.1600	0.4600
Marker5	С	0.4200	0.0833	0.2800	0.6000

Genotype frequencies for each marker with the associated disequilibrium coefficient, its standard error, and the 95% confidence limits

		Genotype	Frequen	eies		
Locus	Genotype	Frequency	HWD Coeff	Standard Error	95% Co Lin	
Marker1	Λ/Λ	0.2000	0.0064	0.0493	-0.0916	0.0956
Marker1	A/B	0.4800	0.0064	0.0493	-0.0916	0.0956
Marker1	B/B	0.3200	0.0064	0.0493	-0.0916	0.0956
Marker2	Λ/Λ	0.4000	0.0636	0.0477	-0.0336	0.1484
Marker2	A/B	0.3600	0.0636	0.0477	-0.0336	0.1484
Marker2	B/B	0.2400	0.0636	0.0477	-0.0336	0.1484
Marker3	A/A	0.4000	-0.0096	0.0457	-0.1044	0.0800
Marker3	Λ/B	0.4800	-0.0096	0.0457	-0.1044	0.0800
Marker3	B/B	0.1200	-0.0096	0.0457	-0.1044	0.0800
Marker4	A/A	0.3600	0.0000	0.0480	-0.0916	0.0864
Marker4	A/B	0.4800	0.0000	0.0480	-0.0916	0.0864
Marker4	B/B	0.1600	0.0000	0.0480	-0.0916	0.0864
Marker5	A/A	0.0800	0.0016	0.0405	-0.0756	0.0816
Marker5	A/B	0.1600	0.0040	0.0337	-0.0664	0.0636
Marker5	A/C	0.2400	-0.0024	0.0380	-0.0736	0.0680
Marker5	B/B	0.2000	0.1100	0.0445	0.0144	0.1884
Marker5	B/C	0.0400	0.1060	0.0282	0.0440	0.1564
Marker5	C/C	0.2800	0.1036	0.0453	0.0096	0.1884

Statistics for testing individual markers for HWE and marker pairs for linkage disequilibrium LD

Obs	Locus1	Locus2	NIndiv	Test	ChiSq	DF	ProbChi	ProbEx
1	Marker1	Marker1	25	HWE	0.01687	1	0.89667	1.0000
2	Marker1	Marker2	25	LD	1.05799	1	0.30367	0.6707
3	Marker1	Marker3	25	LD	1.42074	1	0.23328	0.6524
4	Marker1	Marker4	25	LD	0.33144	1	0.56481	0.9668
5	Marker1	Marker5	25	LD	2.29785	2	0.31698	0.8398
6	Marker2	Marker2	25	HWE	1.70412	1	0.19175	0.2262
7	Marker2	Marker3	25	LD	0.13798	1	0.71030	0.7242
8	Marker2	Marker4	25	LD	1.34100	1	0.24686	0.9015
9	Marker2	Marker5	25	LD	1.13574	2	0.56673	0.5503
10	Marker3	Marker3	25	HWE	0.04340	1	0.83497	1.0000
11	Marker3	Marker4	25	LD	0.46296	1	0.49624	0.9323
12	Marker3	Marker5	25	LD	0.95899	2	0.61909	0.2624
13	Marker4	Marker4	25	HWE	0.00000	1	1.00000	1.0000
14	Marker4	Marker5	25	LD	6.16071	2	0.04594	0.9235
15	Marker5	Marker5	25	HWE	9.35374	3	0.02494	0.0106

Measures of Marker Informativeness

Polymorphism Information Content The polymorphism information content (PIC) measures the probability of differentiating the allele transmitted by a given parent to its child given the marker genotype of father, mother, and child (Botstein et al. 1980). It is computed as

Heterozygosity

The heterozygosity, sometimes called the observed heterozygosity, is simply the proportion of heterozygous individuals in the data set and is calculated as *Allelic Diversity*

The allelic diversity, sometimes called the expected heterozygosity, is the expected proportion of heterozygous individuals in the data set when HWE holds and is calculated as

$$\text{PIC} = 1 - \sum_{u=1}^{k} \tilde{p}_{u}^{2} - \sum_{u=1}^{k-1} \sum_{v=u+1}^{k} 2\tilde{p}_{u}^{2} \tilde{p}_{v}^{2}$$

Het =
$$1 - \sum_{u=1}^{k} \tilde{P}_{uu}$$

$$\mathsf{Div} = 1 - \sum_{u=1}^{k} \tilde{p}_u^2$$

Testing for Hardy-Weinberg Equilibrium

Chi-Square Goodness-of-Fit Test

The chi-square goodness-of-fit test can be used to test markers for HWE. The chi-square statistic has k(k-1)/2 degrees of freedom where k is the number of alleles at the marker locus.

Permutation Version of Exact Test The permutation version of the exact test given by Guo and Thompson (1992) is based on the conditional probability of genotype counts given allelic counts and the hypothesis of allelic independence. The test statistic is where

is the number of heterozygous individuals. Significance levels are calculated by the Monte Carlo permutation procedure. The 2n alleles are randomly permuted the number of times indicated in the PERMS= option to form new sets of ngenotypes. The significance level is then calculated as the proportion of times the value of Tfor each set of permuted data exceeds the value of T for the actual data.

$$X_T^2 = \sum_u \frac{(n_{uu} - n\tilde{p}_u^2)^2}{n\tilde{p}_u^2} + \sum_u \sum_{v>u} \frac{(n_{uv} - 2n\tilde{p}_u\tilde{p}_v)^2}{2n\tilde{p}_u\tilde{p}_v}$$

$$T = \frac{n!}{(2n)!} \frac{2^h \prod_u n_u!}{\prod_{u,v} n_{uv}!}$$

$$h = \sum_{u} \sum_{v \neq u} n_{uv}$$

Example:	data snps; 2222211122
-	input s1-s10; 2 2 2 2 2 1 2 2 2 2
ComputingLinkage	datalines; 2222212122
Disequilibrium Measures for	2 2 2 1 2 1 1 2 2 2 2 2 2 2 1 2 1 2 2
SNP Data	2 2 2 2 2 1 1 1 2 2 2 2 2 2 1 2 2 2 2
The data set contains 44	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$
individuals' genotypes at	
five SNPs	
	2 2 2 2 2 1 2 2 2 2 2 2 2 1 1 2 2
	2 2 2 2 2 1 2 1 2 2 2 2 2 2 2 1 2 1 2 2
	2 2 2 2 2 1 2 2 2 2 2 2 1 2 1 2 2
	2 2 2 2 1 1 1 1 2 2 2 2 2 2 2 2 2 2 2 2
	2 2 1 1 2 2 2 1 2 2 2 2 2 2 2 2 2 2 2 2
	2 2 2 1 2 2 2 1 2 2 2 2 2 2 1 2 1 2 2
	2 2 2 2 1 1 1 2 2 2 2 2 2 2 2 2
	2 2 2 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
	2 2 2 2 1 1 2 1 2 2 ;

PROC ALLELE can be performed as follows:

This analysis produces summary statistics of the five SNPs as well as the Linkage Disequilibrium Measures table, which contains estimated two-locus haplotype frequencies and disequilibrium coefficients, and the linkage disequilibrium measures r, D', and Q. The allele and genotype frequency output tables are suppressed with the NOFREQ option.

The ALLELE Procedure								
	Marker Summary							
	Number	Number	Polymorph		Allelic	Tes	t for]	HWE
Locus	of Indiv	of Alleles	Info Content	Heterozygosity	Diversity	Chi- Square	DF	Pr > ChiSq
SNP1	44	1	0.0000	0.0000	0.0000	0.0000	0	
SNP2	44	2	0.1190	0.0909	0.1271	3.5627	1	0.0591
SNP3	41	2	0.3283	0.4390	0.4140	0.1493	1	0.6992
SNP4	43	2	0.3728	0.4884	0.4957	0.0093	1	0.9231
SNP5	44	1	0.0000	0.0000	0.0000	0.0000	0	•

$$\text{PIC} = 1 - \sum_{u=1}^{k} \tilde{p}_{u}^{2} - \sum_{u=1}^{k-1} \sum_{v=u+1}^{k} 2\tilde{p}_{u}^{2}\tilde{p}_{v}^{2} \qquad \qquad \text{Het} = 1 - \sum_{u=1}^{k} \tilde{P}_{uu} \qquad \qquad \text{Div} = 1 - \sum_{u=1}^{k} \tilde{p}_{u}^{2}$$

		Link	age Disequili	brium Mc	asures		
Locus1	Locus2	Haplotype	Frequency	LD Coeff	Corr Coeff	Lewontin's D'	Yule's Q
SNP1	SNP2	2-1	0.0682	-0.0000			
SNP1	SNP2	2-2	0.9318	-0.0000			
SNP1	SNP3	2-1	0.2927	-0.0000			
SNP1	SNP3	2-2	0.7073	-0.0000	•		
SNP1	SNP4	2-1	0.5465	-0.0000			
SNP1	SNP4	2-2	0.4535	-0.0000			
SNP1	SNP5	2-2	1.0000	0.0000			
SNP2	SNP3	1-2	0.0732	0.0214	0.1807	1.0000	1.0000
SNP2	SNP3	2-1	0.2927	0.0214	0.1807	1.0000	1.0000
SNP2	SNP3	2-2	0.6341	-0.0214	-0.1807	-1.0000	-1.0000
SNP2	SNP4	1-1	0.0331	-0.0050	-0.0398	-0.1322	-0.1546
SNP2	SNP4	1-2	0.0367	0.0050	0.0398	0.1322	0.1546
SNP2	SNP4	2-1	0.5134	0.0050	0.0398	0.1322	0.1546
SNP2	SNP4	2-2	0.4168	-0.0050	-0.0398	-0.1322	-0.1546
SNP2	SNP5	1-2	0.0682	-0.0000		•	
SNP2	SNP5	2-2	0.9318	-0.0000			
SNP3	SNP4	1-1	0.2221	0.0608	0.2661	0.4382	0.5529
SNP3	SNP4	1-2	0.0779	-0.0608	-0.2661	-0.4382	-0.5529
SNP3	SNP4	2-1	0.3154	-0.0608	-0.2661	-0.4382	-0.5529
SNP3	SNP4	2-2	0.3846	0.0608	0.2661	0.4382	0.5529
SNP3	SNP5	1-2	0.2927	-0.0000			•
SNP3	SNP5	2-2	0.7073	-0.0000			
SNP4	SNP5	1-2	0.5465	-0.0000			
SNP4	SNP5	2-2	0.4535	-0.0000	•		

Linkage Disequilibrium Measures

PROC ALLELE offers five linkage disequilibrium measures to be calculated for each pair of alleles M_u and N_v located at loci **M** and **N** respectively: the correlation coefficient *r*, the population attributable risk , Lewontin's *D*', the proportional difference *d*, and Yule's *Q*.

$$r = \frac{D}{(p_1 p_2 q_1 q_2)^{1/2}}$$

$$\delta = \frac{D}{q_1 p_{22}}$$

$$D' = \frac{D}{D_{\max}}, D_{\max} = \begin{cases} \min(p_1 q_2, q_1 p_2), & D > 0\\ \min(p_1 q_1, q_2 p_2), & D < 0 \end{cases}$$

$$d = \frac{D}{q_1 q_2}$$

$$Q = \frac{D}{p_{11} p_{22} + p_{12} p_{21}}$$

HTSNP procedure

Single nucleotide polymorphisms (SNP), roughly one SNP per 1kb in the human genome, accounts for about 90% of human DNA polymorphism

SNPs over large genomic regions suggest the presence of discrete blocks with limited haplotype diversity punctuated by recombination hot spots

Within each block, because of high LD, some allele(s) may always be coexistent with a particular allele at another locus such that

- (1) little haplotype diversity exists in the block, and
- (2) not all SNPs will be essential in characterizing the haplotype structure in the block

The most common haplotypes could usually be captured by a small subset of SNPs, termed haplotype tag SNPs (htSNPs) by Johnson et al. (2001).

Selection of such a SNP subset that distinguishes all haplotypes is known as the minimum test set problem.

The search space of choosing k SNPs out of m is $\binom{m}{k} = \frac{m!}{k!(m-k)!}$, for which enumerating all possible k-SNP combinations becomes impractical even for moderate numbers of *m* and *k*. HTSNP procedure implements some heuristic algorithms for fast identification of an optimal subset of SNPs without mining through all possible combinations.

Methods of finding Haplotype Tag SNPs

Incremental Search starts with finding a first marker of maximum locus richness and goes through the remaining markers to find each time that, which maximizes PDE. **Decremental Search** operates in an opposite manner.

Iterative Maximization Search (Gouesnard et al. 2001) is a fast algorithm for choosing an optimal k-subset from m accessions. It starts from a random selection of k markers for which all the core collections of size k-1 are tested. The subset with the highest PDE is retained. Among the other m-k markers, one that brings the greatest increase in the goodness criterion is selected and a new k-locus set is obtained. Exclusion and inclusion of one marker in the new k-locus set is repeated until convergence. Each iteration needs to evaluate the PDE k times for k-1 markers and m-k times for k markers.

Simulated Annealing Search (Kirkpatrick, Gelatt, and Vecchi 1983) has been adopted in many combinatorial optimization problems. /global optimization problem of applied mathematics, namely locating a good approximation to the global minimum of a given function in a large search space. It is often used when the search space is discrete (e.g., all tours that visit a given set of cities). For certain problems, simulated annealing may be more effective than exhaustive enumeration — provided that the goal is merely to find an acceptably good solution in a fixed amount of time, rather than the best possible solution./ Starting from a selection of k markers, one marker is randomly swapped with another from the unselected markers. The change of haplotype goodness is evaluated using an energy function for the marker exchange. Acceptance of the exchange is judged with the Metropolis criterion (Metropolis et al. 1953) using the change Δ of energy function and the annealing temperature T:

 $\Pr\{\text{new point is accepted}\} = \begin{cases} 1, & \Delta \le 0\\ \exp(-\Delta/T), & \Delta > 0 \end{cases}$

Exhaustive Search An exhaustive search of k markers from m involves traversal of all possible selections once and only once.

Calculations

For *n* haplotypes diversity D_H records the weighted differences of all n^2 pairwise comparisons of two haplotypes (Clayton 2002).

Here h_i and h_j denote the *i*-th and the *j*-th haplotype, and w_i and w_j are the corresponding weights. The difference of two *m*-locus haplotypes, h_i and h_j , is computed as the total allele differences at the *m* loci:

where h_{ik} is the allele of the *i*-th haplotype observed at the *k*-th locus and

If only distinct haplotypes are recorded with their corresponding frequencies using the FREQ statement, then F_i , the weighted frequency of haplotype h_i , can be calculated as

where w_j and f_j are the weight and frequency. The estimate of haplotype diversity D_H is then proportional to the average of the gene diversity D_k at all *m* loci, computed as

where $D_k = 1 - \sum_{u=1}^{l_k} p_{ku}^2$ and l_k is the number of alleles at the *k*-th locus. The weighted allele frequency p_{ku} of the *u*-th allele at the *k*-th locus is recorded as

The diversity computed in this way measures the probability that two haplotypes sampled from the population differ at any locus.

$$D_H = \sum_{i=1}^{n} \sum_{j=1}^{n} w_i w_j (h_i - h_j)$$

$$h_i - h_j = \sum_{k=1}^m (h_{ik} - h_{jk})$$

$$h_{ik} - h_{jk} = \begin{cases} 0 & h_{ik} = h_{jk} \\ 1 & h_{ik} \neq h_{jk} \end{cases}$$

$$F_i = \frac{w_i f_i}{\sum_j w_j f_j}$$

$$D = \frac{\sum_{k=1}^{m} D_k}{m}$$

$$p_{ku} = \sum_{h_{ik}=u} F_i$$

Goodness of a Core Set Selection

For a core set of *k* SNPs, the *n* observed haplotypes can be classified into *G* groups of *k*-locus haplotypes. The haplotype residual diversity, R_D , is defined as the sum of the within-group diversities for the *G* groups: -----> where F_g is the weighted frequency of the *g*-th group calculated as -----> and F_{ig} is the within-group weighted frequency for the *i*-th haplotype in the *g*-th group with $F_{ig} = F_i/F_g$. Similarly, R_D can be calculated as the total within-group gene diversity: ----->

and

where

is calculated using the within-group allele frequencies $p_{k u g}$. The proportion of diversity explained (PDE) by a SNP set selection is used to evaluate the goodness of that selection. *PDE* is calculated as

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The selected search algorithm finds the optimal subset that maximizes *PDE*.

$$R_D = \sum_{g=1}^G F_g^2 \Big[\sum_{i \in G_g} \sum_{j \in G_g} F_{ig} F_{jg} (h_i - h_j) \Big]$$
$$F_g = \sum_{i \in G_g} F_i$$
$$R_D = \sum_{g=1}^G F_g^2 D_g$$
$$D_g = \frac{\sum_{k=1}^m D_{kg}}{m}$$

$$D_{kg} = 1 - \sum_{u} p_{kug}^2$$

 $PDE = 1 - [(R_D)/D]$

Using the HAPLOTYPE and HTSNP Procedures Together

Before using PROC HTSNP, you may need to first run PROC HAPLOTYPE if you have data with unknown phase in order to estimate the haplotype frequencies

	The HAPLOTY	PE Proced	lure		
	Haplotype F	requencie	S		
Number	Haplotype	Freq	Standard Error		onfidence nits
1	G-T-A-T-C-G-G-C-C-G-A-A-C	0.01988	0.00807	0.00406	0.03570
2	T-C-G-G-C-G-G-C-C-G-A-A-C	0.09173	0.01669	0.05902	0.12445
3	T-T-A-G-C-A-A-G-C-G-A-A-A	0,16666	0.02155	0.12442	0.20890
4	T-T-A-G-C-A-G-G-C-G-A-A-A	0.05667	0.01337	0.03046	0.08287
5	T-T-A-G-C-A-G-G-C-G-A-A	0,03663	0.01086	0.01534	0.05793
6	T-T-G-G-C-A-G-G-C-G-A-A-A	0.01579	0.00721	0.00166	0.02992
7	T-T-G-G-C-G-G-C-C-G-A-A-C	0.40576	0.02840	0.35011	0.46142
8	T-T-G-G-C-G-G-G-T-C-A-A-A	0.02667	0.00932	0.00841	0.04493
9	T-T-G-G-C-G-G-G-T-G-A-A-A	0.00861	0.00534	0.00000	0.01908
10	T-T-G-G-G-G-G-C-C-G-A-G-C	0.16250	0.02133	0.12069	0.20432

proc htsnp data=hapfreq

Size=4	/* Number of tag SNPs */
method=sa	/* Simulated Annealing Method */
Best=5	/* Five best SNPs tag sets */
cutoff=0.05	
Seed=123	/* Initializing the random selection of first the set */
outstat=out;	
var m1-m13;	/* SNPs in haplotypes */
freq freq;	/* Weights of different haplotypes */

run;

In this example, the Simulated Annealing Search method is specified for finding the best sets of size four. The output data set OUT that is created by PROC HTSNP is then printed out to show the best five sets of SNPs that were selected.

Obs	HTSNP1	HTSNP2	HTSNP3	HTSNP4	PDE
1	m2	m5	m7	m13	1
2	m2	m7	m8	m12	1
3	m2	m5	m7	m8	1
4	m2	m7	m12	m13	1
5	m2	m5	m6	m7	1

HAPLOTYPE procedure

The HAPLOTYPE procedure uses the expectation-maximization (EM) algorithm to generate maximum likelihood estimates of haplotype frequencies given a multilocus sample of genetic marker genotypes under the assumption of Hardy-Weinberg equilibrium (HWE). These estimates can then in turn be used to assign the probability that each individual possesses a particular haplotype pair.

PROC HAPLOTYPE performs a likelihood ratio test to test the hypothesis of no LD between marker loci.

Another application is association testing of disease susceptibility: haplotypes might include two or more causative sites show synergistic interaction.

PROC HAPLOTYPE can use case-control data to calculate test statistics for the hypothesis of no association between alleles comprising the haplotypes and disease status; such tests are carried out over all haplotypes at the loci specified, or for individual haplotypes.

ODS Tables Created by the HAPLOTYPE Procedure

ODS Table Name	Description	Statement or Option
AnalysisInfo	Analysis information	default
IterationHistory	Iteration history	ITPRINT
ConvergenceStatus	Convergence status	default
HaplotypeFreq	Haplotype frequencies	default
LDTest	Test for allelic associations	LD
CCTest	Test for marker-trait association	TRAIT statement
HapTraitTest	Tests for haplotype-trait association	TRAIT / TESTALL

Example

```
data markers;
   input (m1-m8) ($);
   datalines;
   BAB
        в
         в
 B
           A A
   ABB
       A B
           Α
             в
 Α
   BAABBBB
 B
  вававав
 Α
  AABABBB
 A
  вааавав
 B
  ввваваа
 Α
  ваааааа
 Α
  ВАААААВ
 в
  ВΑ
      BABB
 Α
             в
  ВАВАВАА
 Α
   ВΑ
      ваваа
 В
  вааавав
 Α
  в
    в
      BBBAB
 Α
  AABAAAB
 Α
  вававав
 B
  вввааав
 Α
 в
   вв
      BAAAA
    АААВАА
 A
  в
      AABA
    Α
 А
   в
             В
    AAAAA
   в
 \mathbf{B}
             в
   A A
      BAAAB
 Α
 AB
    AAAAB
             в
 A A A A A A
   B
    B
      BAAAA
 Α
 ;
```

You can now use PROC HAPLOTYPE to infer the possible haplotypes and estimate the fourlocus haplotype frequencies in this sample. The following statements will perform these calculations:

```
proc haplotype data=markers
    out=hapout
    init=random
    prefix=SNP;
    var m1-m8;
```

run;

EM algorithm estimates the haplotype frequencies Option INIT=RANDOM indicates that initial haplotype frequencies are randomly generated Maximum number of iterations is set to 100.

The HAPLOTYPE Procedure

Analysis Information

Loci Used	SNP1 SNP2 SNP3 SNP4
Number of Individuals	25
Number of Starts	1
Convergence Criterion	0.00001
Iterations Checked for Conv.	1
Maximum Number of Iterations	100
Number of Iterations Used	15
Log Likelihood	-95.94742
Initialization Method	Random
Random Number Seed	51220
Standard Error Method	Binomial
Haplotype Frequency Cutoff	0

2 A-A-A-B 0.07527 0.03769 0.00140 0. 3 A-A-B-A 0.00000 0.00000 0.00000 0.0 4 A-A-B-B 0.00000 0.00010 0.00000 0.0 5 A-B-A 0.09307 0.04151 0.01173 0. 6 A-B-A-B 0.05335 0.03210 0.00000 0.0 7 A-B-A-A 0.00002 0.00061 0.00000 0.0 8 A-B-B-B 0.07526 0.03769 0.00140 0. 9 B-A-A-A 0.08638 0.04013 0.00772 0.	24105 14914
2 A-A-A-B 0.07527 0.03769 0.00140 0. 3 A-A-B-A 0.00000 0.00000 0.00000 0.0 4 A-A-B-B 0.00000 0.00010 0.00000 0.0 5 A-B-A 0.09307 0.04151 0.01173 0. 6 A-B-A-B 0.05335 0.03210 0.00000 0.0 7 A-B-B-A 0.00002 0.00061 0.00000 0.0 8 A-B-B-B 0.07526 0.03769 0.00140 0. 9 B-A-A-A 0.08638 0.04013 0.00772 0.	14914
3 A-A-B-A 0.00000 0.00000 0.00000 0.0 4 A-A-B-B 0.00000 0.00010 0.00000 0.0 5 A-B-A-A 0.09307 0.04151 0.01173 0.0 6 A-B-A-B 0.05335 0.03210 0.00000 0.0 7 A-B-B-A 0.00002 0.00061 0.00000 0.0 8 A-B-B-B 0.07526 0.03769 0.00140 0.0 9 B-A-A-A 0.08638 0.04013 0.00772 0.0	
4 A-A-B-B 0.00000 0.00010 0.00000 0.0 5 A-B-A-A 0.09307 0.04151 0.01173 0. 6 A-B-A-B 0.05335 0.03210 0.00000 0. 7 A-B-B-A 0.00002 0.00061 0.00000 0. 8 A-B-B-B 0.07526 0.03769 0.00140 0. 9 B-A-A-A 0.08638 0.04013 0.00772 0.	
5 A-B-A-A 0.09307 0.04151 0.01173 0. 6 A-B-A-B 0.05335 0.03210 0.00000 0. 7 A-B-B-A 0.00002 0.00061 0.00000 0.0 8 A-B-B-B 0.07526 0.03769 0.00140 0. 9 B-A-A-A 0.08638 0.04013 0.00772 0.	00000
6 A-B-A-B 0.05335 0.03210 0.00000 0.7 7 A-B-B-A 0.00002 0.00061 0.00000 0.6 8 A-B-B-B 0.07526 0.03769 0.00140 0.7 9 B-A-A-A 0.08638 0.04013 0.00772 0.7	00020
7 A-B-B-A 0.00002 0.00061 0.00000 0.0 8 A-B-B-B 0.07526 0.03769 0.00140 0.0 9 B-A-A-A 0.08638 0.04013 0.00772 0.0	17442
8 A-B-B-B 0.07526 0.03769 0.00140 0. 9 B-A-A-A 0.08638 0.04013 0.00772 0.	1627
9 B-A-A-A 0.08638 0.04013 0.00772 0.	00122
	14913
10 B-A-A-B 0.08792 0.04046 0.00863 0	16504
	16722
11 B-A-B-A 0.07921 0.03858 0.00359 0.	15482
12 B-A-B-B 0.10819 0.04437 0.02122 0.	19517
13 B-B-A-A 0.10098 0.04304 0.01662 0.	18534
14 B-B-A-B 0.00000 0.00001 0.00000 0.0	00002
15 B-B-A 0.09732 0.04234 0.01433 0.	18030
16 B-B-B-B 0.00000 0.00001 0.00000	

Output Data Set from the HAPLOTYPE Procedure Each individual's genotype with each of the possible haplotype pairs that can comprise the genotype, and the probability the genotype can be resolved into each of the possible haplotype pairs.

_ I												
	ID	ml	m2	m3	m4	m5	mб	m7	п18	HAPLOTYPE1	HAPLOTYPE2	PROB
	1	В	В	A	В	В	в	A	A	B-A-B-A	B-B-A	1.00
	2	A	А	B	в	A	в	A	B	A-B-A-A	A-B-B-B	1,00
	2	Α	А	В	В	Α	В	А	В	А-В-А-В	А-В-В-А	0.00
	3	В	В	Α	А	В	В	В	В	B-A-B-B	B-A-B-B	1.00
	4	٨	В	Λ	в	۸	В	Λ	в	Л-Л-Л-В	В-В-А	0,26
	4	А	В	Α	В	Α	В	А	В	A-B-A-A	B-A-B-B	0.36
	4	A	в	A	в	A	В	A	B	A-I3-A-I3	B-A-B-A	0,15
	4	Α	В	Α	В	Α	В	А	В	А-В-В-А	B-A-A-B	0.00
	4	Ā	В	A	в	Α	в	Ä	В	А-В-В-В	B-A-A-A	0.23
	5	Α	А	Α	В	Α	В	В	В	A-A-A-B	A-B-B-B	1.00
	6	В	В	Α	А	Α	В	А	В	B-A-A-A	B-A-B-B	0.57
	6	В	В	Α	А	Α	В	А	В	B-A-A-B	B-A-B-A	0.43
	7	A	В	В	В	Α	В	А	А	A-B-A-A	B-B-A	1.00
	7	A	В	В	в	A	в	А	А	A-B-A	B-B-A-A	0.00
	8	Α	В	Α	А	Α	А	А	А	A-A-A-A	B-A-A-A	1.00
	9	В	В	Α	А	Α	А	А	В	B-A-A-A	B-A-A-B	1.00
	10	Α	В	Α	В	Α	В	В	В	A-B-A-B	B-A-B-B	0.47
	10	А	В	А	В	А	В	В	В	A-B-B-B	B-A-A-B	0.53
	11	A	в	А	в	А	в	А	А	A-A-A-A	B-B-A	0.65
	11	A	В	А	В	А	в	А	А	A-B-A-A	B-A-B-A	0.35
	11	A	В	A	в	A	в	A	А	A-B-B-A	B-A-A-A	0,00
	12	В	В	٨	в	۸	В	Α	Λ	Β-Λ-Λ-Λ	В-В-В-А	0,51
	12	В	В	Α	В	Α	В	А	А	B-A-B-A	B-B-A-A	0.49
	13	A	В	A	А	A	B	А	B	A-A-A-A	B-A-B-B	0.72
	13	A	В	А	А	A	в	A	в	A-A-A-B	B-A-B-A	0.28
	14	A	В	В	В	В	В	A	В	A-B-B-B	B-B-A	1,00
	15	Λ	٨	Α	в	Α	Λ	Α	в	Λ-Λ-Α-Λ	А-В-А-В	0,52
	15	٨	Λ	Α	в	Δ	Α	Α	В	А-А-А-В	А-В-А-А	0,48
	16	В	В	A	В	A	в	A	в	B-A-A-B	B-B-A	0.44
	16	В	в	A	В	A	в	A	В	В-А-В-В	B-B-A-A	0,56

CUTOFF= option affects the "Haplotype Frequencies" table. <u>To view only the haplotypes with</u> <u>an estimated frequency of at least 0.10</u>:

proc haplotype data=ehdata se=jackknife cutoff=0.10 nlag=4; var m1-m6;

run;

Now, the "Haplotype Frequencies" table is displayed as:

	The HAPLOTYPE Procedure									
	Haplotype Frequencies									
Number	Number Haplotype Freq Standard 95% Confidence Error Limits									
1	1-1-3	0.11509	0.01766	0.08048	0.14971					
2	1-2-3	0.12788	0.02094	0.08685	0.16891					
3	2-1-2	0.11700	0.01782	0.08207	0.15193					
4	2-2-1	0.11766	0.01831	0.08177	0.15355					
5	2-2-3	0.10397	0.01833	0.06805	0.13989					

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PSMOOTH procedure

In the search for complex disease genes, linkage and/or association tests are often performed on markers from a genome-wide scan or SNPs from a finely scaled map. This means hundreds or even thousands of hypotheses are being simultaneously tested. Plotting the negative log *p*-values of all the marker tests will reveal many peaks that indicate significant test results, some of which are false positives. In order to reduce the number of false positives or improve power, <u>smoothing methods can be applied</u> that take into account *p*-values from neighboring, and possibly correlated, markers. That is, the peak length can be used to indicate significance in addition to the peak height. The PSMOOTH procedure offers smoothing methods that implement Simes' method (1986). Fisher's method (1932), and/or the truncated product method (TPM) (2002) for multiple hypothesis testing. These methods modify the *p*-value from each marker test using a function of its original *p*-value and the *p*-values of the tests on the nearest markers. Since the number of hypothesis tests being performed is not reduced, adjustments to correct the smoothed *p*-values for multiple testing are available as well.

PROC PSMOOTH can take any data set containing any number of columns of *p*-values as an input data set, including the output data sets from the CASECONTROL and FAMILY procedures (see Chapter 3 and Chapter 4 for more information).

data tests;			11	0.28561	12	0.80495	
input Mar	ker	Pvalue @@;		0.28501	42 44	0.80495	
datalines;				0.53831	46	0.78712	
1 0.72841	2	0.40271	43 47	0.88493	40 48	0.36260	
3 0.32147	4	0.91616	47	0.53310	40 50	0.65709	
5 0.27377	6	0.48943	49 51	0.26527	50 52	0.46860	
7 0.40131	8	0.25555	-	0.20327	52 54	0.54956	
9 0.57585	10	0.20925		0.33403	56	0.04933	
11 0.01531	12	0.23306		0.12016	58	0.76181	
13 0.69397	14	0.33040	59	0.80158	60	0.18244	
15 0.97265	16	0.53639	61	0.01382	62	0.15100	
17 0.88397	18	0.03188	63	0.04713	64	0.52655	
19 0.13570	20	0.79138		0.59368	66	0.94420	
21 0.99467	22	0.37831	67	0.60104	68	0.32848	
23 0.86459	24	0.97092	69	0.90195	70	0.21374	
25 0.19372	26	0.85339	71	0.95471	72	0.14145	
27 0.32078	28	0.31806	73	0.95215	74	0.70330	
29 0.00655	30	0.82401	75	0.19921	76	0.99086	
31 0.65339	32	0.36115	77	0.75736	78	0.23761	
33 0.92704	34	0.49558	79	0.87260	80	0.91472	
35 0.64842	36	0.43606	81	0.33650	82	0.26160	
37 0.67060	38	0.87520	83	0.41948	84	0.62817	
39 0.78006	40	0.27252	85	0.48721	86	0.67093	
			87	0.53089	88	0.13623	

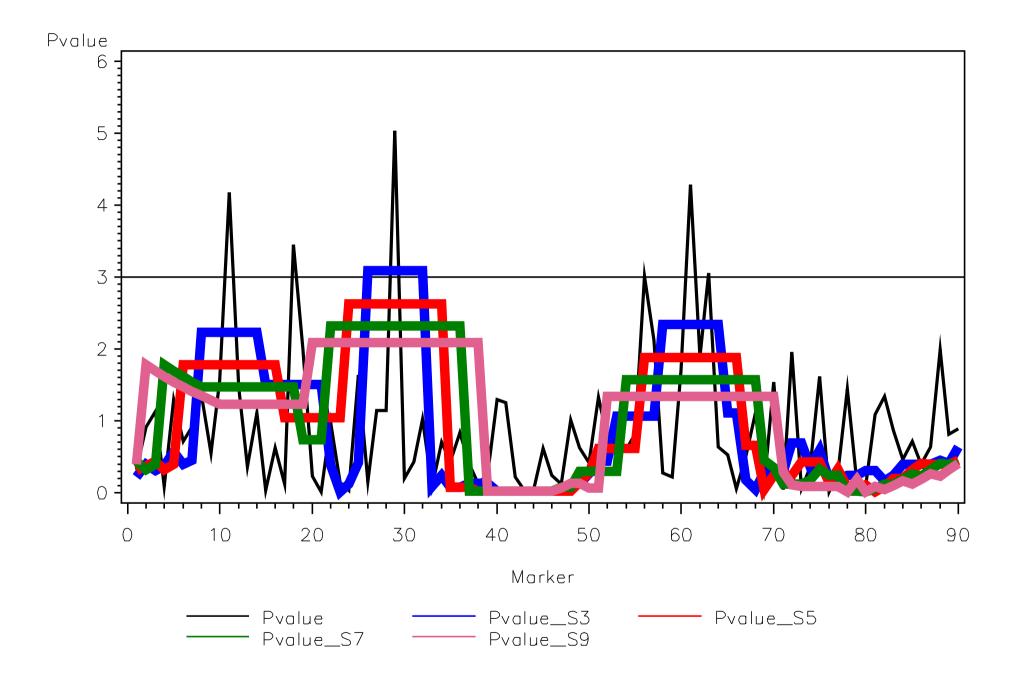
89 0.44344 90 0.41172

;

proc **psmooth**

data=tests
 out=pnew
 simes
 bandwidth=3 to 9 by 2
 neglog;
var Pvalue;
id Marker;

run;



CASECONTROL procedure

Marker information can be used to help locate the genes that affect susceptibility to a disease. The CASECONTROL procedure is designed for the interpretation of marker data when random samples are available from the populations of unrelated individuals who are either affected or unaffected by the disease. Several tests are available in PROC CASECONTROL that compare marker allele and/or genotype frequencies in the two populations, with frequency differences indicating an association of the marker with the disease. Although such an association may point to the proximity of the marker and disease genes in the genome, it may also reflect population structure, so care is needed in interpreting the results; association does not necessarily imply linkage.

The three chi-square tests available for testing case-control genotypic data are the genotype case-control test, which tests for dominant allele effects on the disease penetrance, and the allele case-control test and linear trend test, which test for additive allele effects on the disease penetrance. Since the allele case-control test requires the assumption of Hardy-Weinberg equilibrium (HWE), it may be desirable to run the ALLELE procedure on the data to perform the HWE test on each marker (see Chapter 2, " The ALLELE Procedure," for more information) prior to applying PROC CASECONTROL.

OUTSTAT= Data Set

The output data set specified in the OUTSTAT= option of the PROC CASECONTROL statement contains the following variables:

- BY variables, if any
- · Locus

•

- Counts of genotyped individuals for the two values of the TRAIT variable: NumTrait1 and NumTrait2, where 1 and 2
 - are replaced by the values of the TRAIT variable
 - Chi-square statistic for each test performed: ChiSqAllele,

ChiSqGenotype, and ChiSqTrend

Degrees of freedom for each test performed: dfAllele,

dfGenotype, and dfTrend

p-value for each test performed: ProbAllele, ProbGenotype, and ProbTrend

Example

```
data founders;
  input id disease al-a4 00;
  datalines;
   16437
4
             17
                  24727
39 2 6 8 7 7
              41
                  2 4 4
              50 2 4 2 3 7
46 1 8 4 1 5
54 24876
             56 2 7 4 7 7
             69 2 6 8 2 7
62 2 4 1 7 3
79 1 6 6 8 7 80 2 6 4 7 3
              85 1 5 6 6 2
83 2 8 4 2 7
95 1 3 2 3 7
              101 1 4 6 7 7
106 1 2 1 7 2
              107 1 1 2 7 7
115 2 4 2 7 5
              116 1 4 1 7 3
120 2 1 6 2 7 123 2 4 4 7 2
130 1 5 2 3 7 133 1 8 6 3 6
134 1 8 4 2 2
             13926476
142 2 3 6 7 7
              151 1 4 6 4 3
152 1 6 7 6 7
              153 1 5 1 7 6
154 1 4 6 6 6
              168 1 1 4 3 7
178 2 4 1 7 1
              187 1 1 8 1 2
189 2 6 4 5 7
              190 2 4 4 3 7
195 2 4 4 7 2
              207 2 1 6 7 7
216 1 7 4 1 5
              222 2 4 2 7 3
225 2 8 7 7 6
              234 1 6 4
                        22
244 1 4 4 7 6
              249 2 6 8 7 2
263 1 8 2 3 7
              267 2 2 2 2 7
276 2 1 6 7 1
              284 2 4 8 2 2
286 1 8 8 2 1
              289 1 2 6 6 3
290 1 2 4 5 7
              294 2 1 8 6 7
297 2 5 4 7 6
             313 1 1 7 7 2
337 1 2 6 7 6
             366 2 2 2 7 7
368 2 3 1 7 2 381 1 6 4 5 3
384 1 6 2 2 7 396 1 4 5 7 2
;
```

The multiallelic versions of the association tests are performed since each marker has more than two alleles.

The following code invokes the three case-control tests to find out whether there is a significant association between either of the markers and disease status.

```
proc casecontrol data=founders
    genotype allele trend;
    trait disease;
    var a1-a4;
run;
```

proc print noobs heading=h; run;

Output 3.1.1: Output Data Set from PROC CASECONTROL for Multiallelic Markers

Locus	NumTrait1	Num Trait2	ChiSqGenotype	ChiSqAllele	ChiSqTrend	dfGenotype	dfAllele	dfTrend	ProbGenotype	ProbAllele	ProbTrend
M1	30	30	27.333	4.441	5.039	24	7	7	0.2892	0.7278	0.6552
M2	30	30	18.077	8.772	13.244	15	7	7	0.2586	0.2694	0.0664

INBREED procedure

The INBREED procedure calculates the covariance or inbreeding coefficients for a pedigree. PROC INBREED is unique in that it handles very large populations.

The INBREED procedure has two modes of operation. One mode carries out analysis on the assumption that all the individuals belong to the same generation.

The other mode divides the population into nonoverlapping generations and analyzes each generation separately, assuming that the parents of individuals in the current generation are defined in the previous generation.

PROC INBREED also computes averages of the covariance or inbreeding coefficients within sex categories if the sex of individuals is known.

ODS Tables Produced in PROC INBREED

ODS Table Name	Description	Statement	Option
AvgCovCoef	Averages of covariance coefficient matrix	GENDER	COVAR and AVERAGE
AvgInbreedingCoef	Averages of inbreeding coefficient matrix	GENDER	AVERAGE
CovarianceCoefficient	Covariance coefficient table	PROC	COVAR and MATRIX
InbreedingCoefficient	Inbreeding coefficient table	PROC	MATRIX
IndividualCovCoef	Covariance coefficients of individuals	PROC	IND and COVAR
IndividualInbreedingCoef	Inbreeding coefficients of individuals	PROC	IND
MatingCovCoef	Covariance coefficients of matings	MATINGS	COVAR
MatingInbreedingCoef	Inbreeding coefficients of matings	MATINGS	
NumberOfObservations	Number of observations	PROC	



Often provide a more effective way of testing markers for association with disease status than case-control data. Case-control data may uncover significant associations between markers and a disease that could be caused by factors other than linkage, such as population structure.

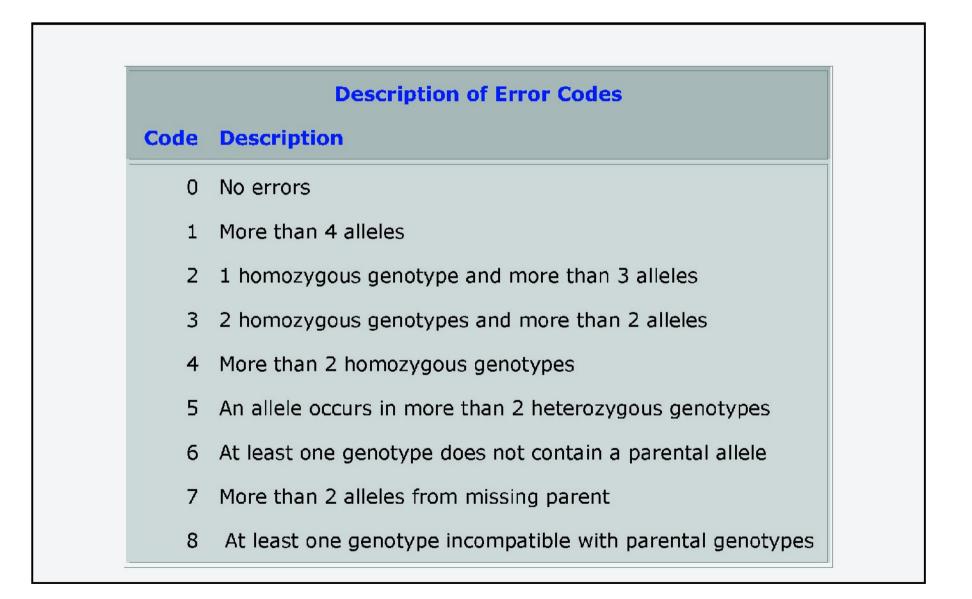
FAMILY procedure ensures that any significant associations found between a marker and disease status are due to linkage between the marker and disease locus. This is accomplished by using the transmission/disequilibrium test (TDT) and several variations of it that can accommodate different types of family data.

One type of family consists of parents, at least one heterozygous, and an affected child who have all been genotyped. This family structure is suitable for the original TDT.

Families containing at least one affected and one unaffected sibling from a sibship that have both been genotyped can be analyzed using the sibling tests: the sib TDT (S-TDT) or the nonparametric sibling disequilibrium test (SDT).

Both types of families can be jointly analyzed using the combined versions of the S-TDT and SDT and the reconstruction-combined TDT (RC-TDT). The RC-TDT can additionally accommodate families with no unaffected children and missing parental genotypes in certain situations.

		Fa	amily Sum	mary		
Parent1	Parent2	Locus	Number of Typed Parents	Number of Affected Children	Number of Unaffected Children	Error Code
1	2	M1	2	1	1	8
101	102	M1	1	2	0	6
201	202	M1	1	2	1	7
301	302	M1	0	1	2	5
401	402	M1	0	2	1	4
501	502	M1	0	1	2	3
601	602	M1	0	1	2	2
701	702	M1	0	2	2	1
801	802	M1	1	2	0	0



TPLOT macro

The %TPLOT macro creates a triangular plot that graphically displays genetic marker test results. The plot has colors and shapes representing

p-value ranges for tests of

the following quantities:

linkage disequilibrium between pairs of markers,

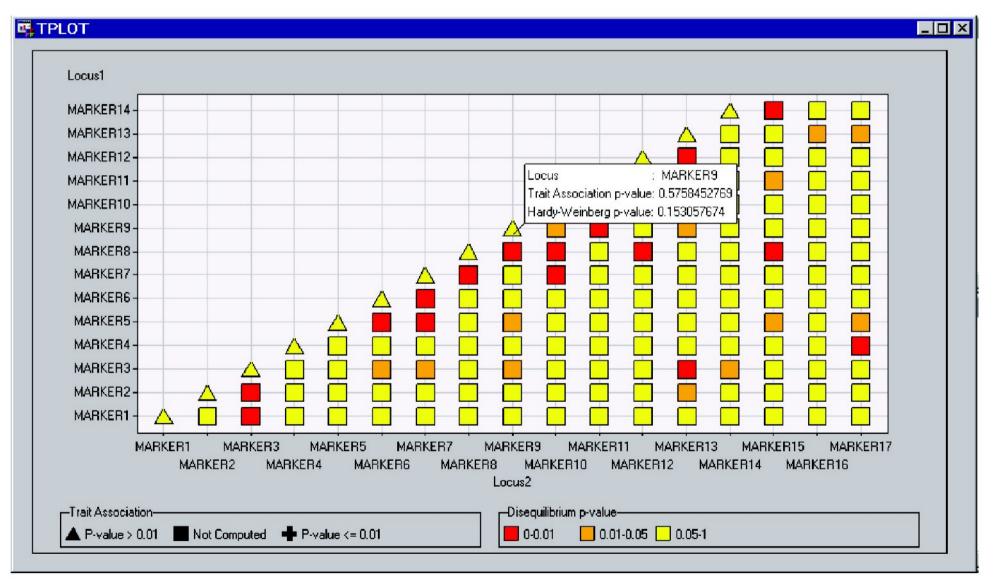
Hardy-Weinberg equilibrium (HWE) for individual markers, and

associations between markers and a dichotomous trait (such as disease status). This is a convenient way of combining information contained in output data sets from two separate SAS/Genetics procedures and summarizing it in an easily interpretable plot.

Thus, insights can be gleaned by simply studying a plot rather than by having to search through many rows of data or writing code to attempt to summarize the results.

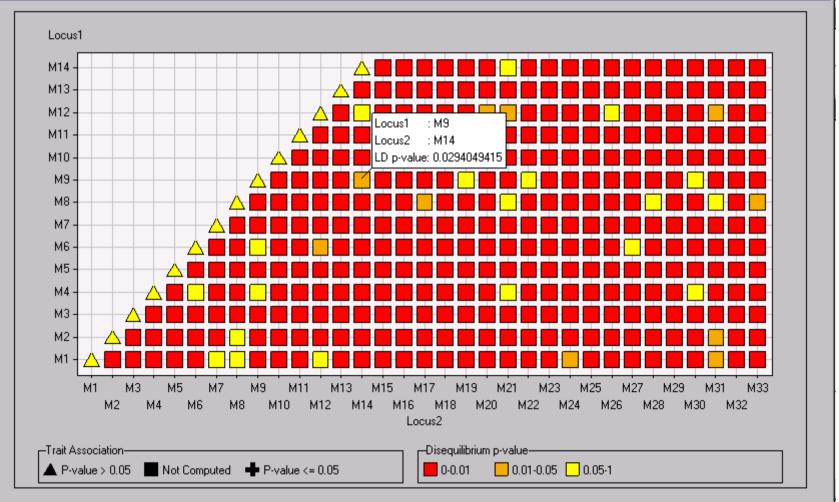
The %TPLOT macro is a part of the SAS Autocall library, and is automatically available for use in your SAS program provided that the SAS system option MAUTOSOURCE is in effect.

Colors and shapes of the data points are used to symbolize *p*-value ranges. The button in the toolbar enables the *p*-values to be displayed.



Results Window for TPLOT Macro

TPLOT



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Aitäh kuulamast!