TDT vignette
Use of snpStats in family–based studies

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Pedigree data

The `snpStats` package contains some tools for analysis of family-based studies. These assume that a subject support file provides the information necessary to reconstruct pedigrees in the well-known format used in the *LINKAGE* package. Each line of the support file must contain an identifier of the *pedigree* to which the individual belongs, together with an identifier of subject within pedigree, and the within-pedigree identifiers for the subject's father and mother. Usually this information, together with phenotype data, will be contained in a dataframe with rownames which link to the rownames of the `SnpMatrix` containing the genotype data. The following commands read some illustrative data on 3,017 subjects and 43 (autosomal) SNPs[1]. The data consist of a dataframe containing the subject and pedigree information (`pedData`) and a `SnpMatrix` containing the genotype data (`genotypes`):

```
> require(snpStats)
> data(families)
> genotypes

A SnpMatrix with  3017 rows and  43 columns
Row names:  id02336 ... id02732
Col names:  rs91126 ... rs98918

> head(pedData)
```

	familyid	member	father	mother	sex	affected
id02336	fam0005	1	NA	NA	1	1
id00695	fam0005	2	NA	NA	2	1
id02750	fam0005	3	1	2	2	2
id01836	fam0005	4	1	2	2	2
id02533	fam0006	1	NA	NA	2	1
id01069	fam0006	2	NA	NA	1	1

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[1]These data are on a much smaller scale than would arise in genome-wide studies, but serve to illustrate the available tools. Note, however, that execution speeds are quite adequate for genome-wide data.

The first family comprises four individuals: two parents and two sibling offspring. The parents are "founders" in the pedigree, *i.e.* there is no data for their parents, so that their `father` and `mother` identifiers are set to `NA`. This differs from the convention in the *LINKAGE* package, which would code these as zero. Otherwise coding is as in *LINKAGE*: `sex` is coded 1 for male and 2 for female, and disease status (`affected`) is coded 1 for unaffected and 2 for affected.

Checking for mis-inheritances

The function `misinherits` counts non-Mendelian inheritances in the data. It returns a logical matrix with one row for each subject who has any mis-inheritances and one column for each SNP which was ever mis-inherited.

```
> mis <- misinherits(data=pedData, snp.data=genotypes)
> dim(mis)

[1] 114  37
```

Thus, 114 of the subjects and 37 of the SNPs had at least one mis-inheritance. The following commands count mis-inheritances per subject and plot its frequency distribution, and similarly, for mis-inheritances per SNP:

```
> per.subj <- apply(mis, 1, sum, na.rm=TRUE)
> per.snp <- apply(mis, 2, sum, na.rm=TRUE)
> par(mfrow = c(1, 2))
> hist(per.subj,main='Histogram per Subject', xlab='Subject')
> hist(per.snp,main='Histogram per SNP', xlab='SNP')
```
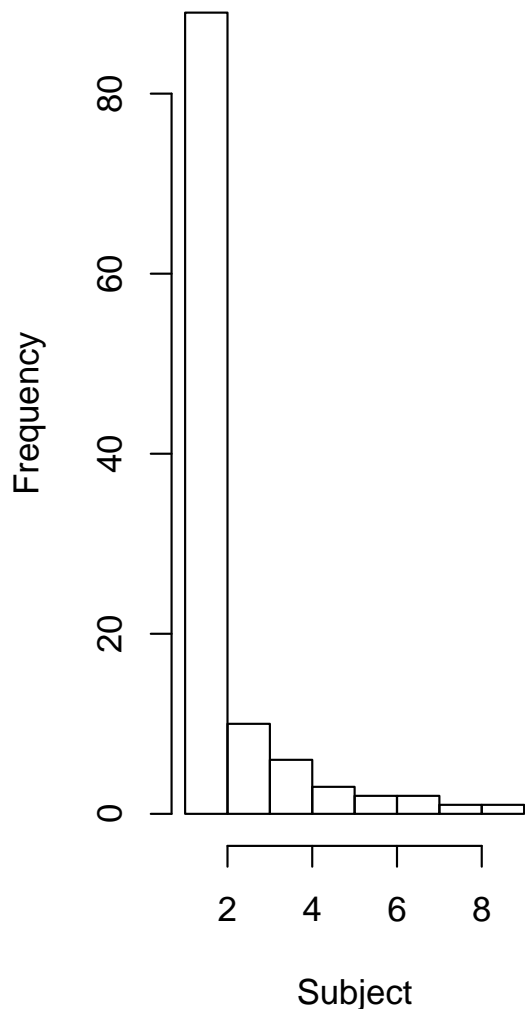
Histogram per Subject   ## Histogram per SNP
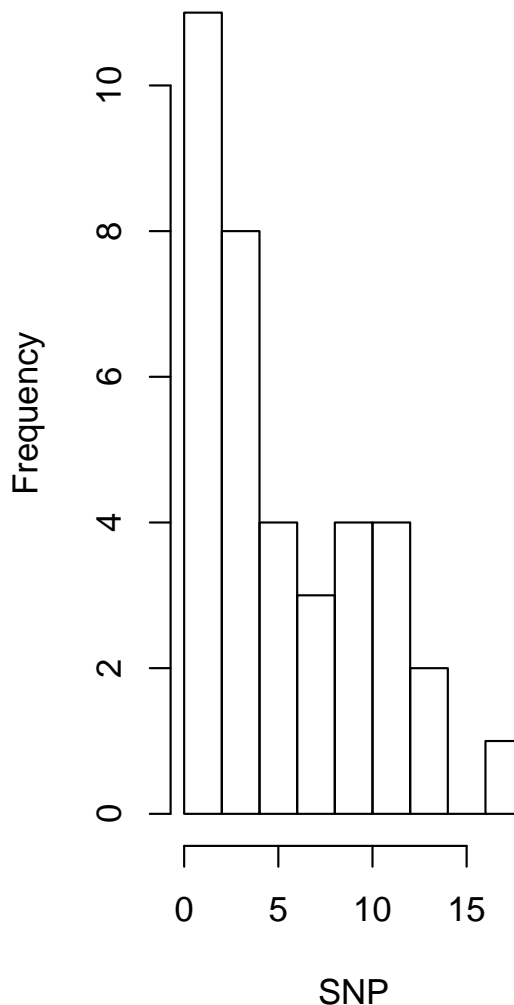
Note that mis-inheritances must be ascribed to offspring, although the error may lie with the parent data. The following commands first extract the pedigree identifiers for mis-inheriting subjects and go on to chart the numbers of mis-inheritances per family:

```
> fam <- pedData[rownames(mis), "familyid"]
> per.fam <- tapply(per.subj, fam, sum)
> par(mfrow = c(1, 1))
> hist(per.fam, main='Histogram per Family', xlab='Family')
```

Histogram per Family
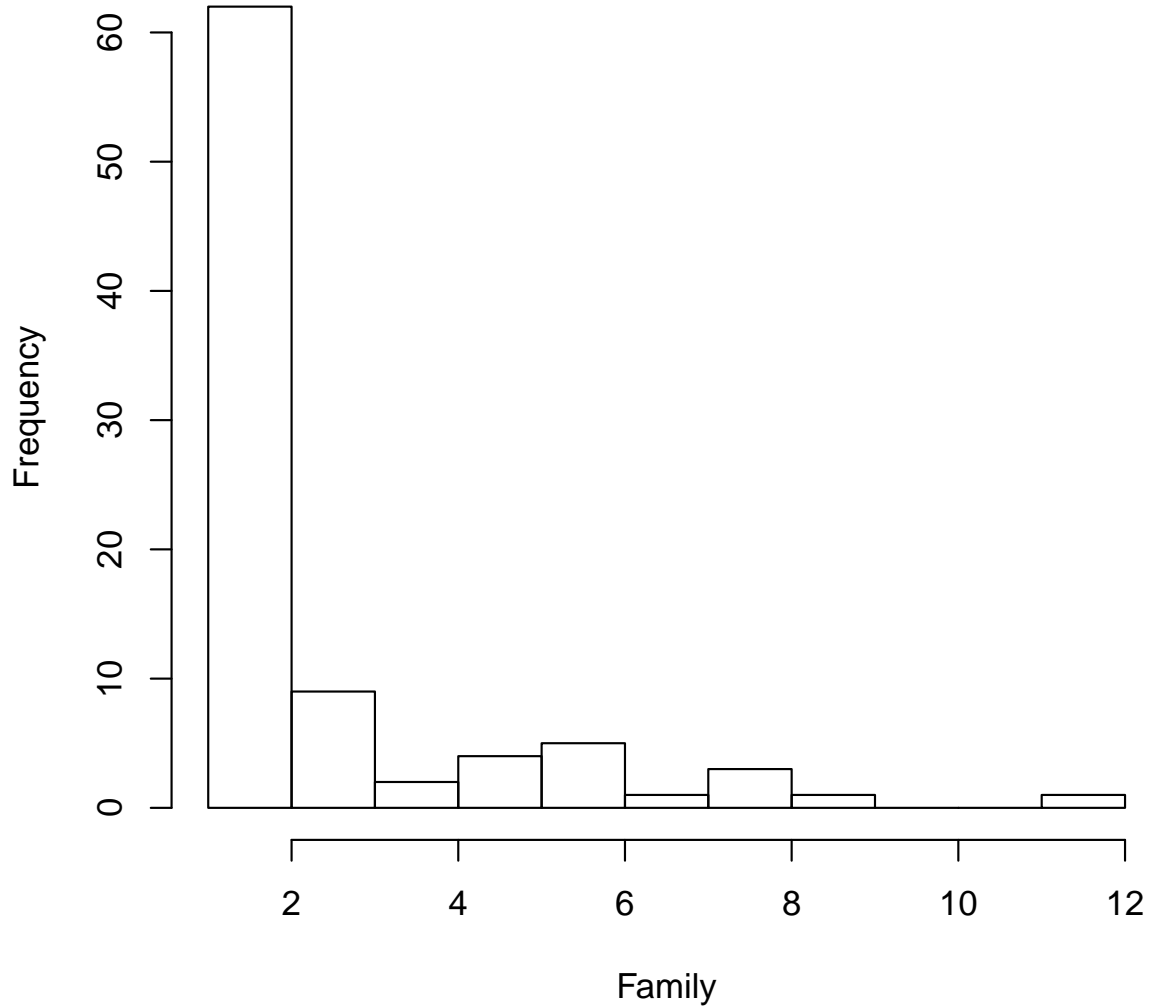
None of the above analyses suggest serious problems with the data, although there are clearly a few genotyping errors.

TDT tests

At present, the package only allows testing of discrete disease phenotypes in case–parent trios — basically the Transmission/Disequilibrium Test (TDT). This is carried out by the function `tdt.snp`, which returns the same class of object as that returned by `single.snp.tests`; allelic (1 df) and genotypic (2 df) tests are computed. The following commands compute

the tests, display the *p*-values, and plot quantile–quantile plots of the 1 df tests chi-squared statistics:

```
> tests <- tdt.snp(data = pedData, snp.data = genotypes)

Analysing 1466 potentially complete trios in 733 different pedigrees

> cbind(p.values.1df = p.value(tests, 1),
+       p.values.2df = p.value(tests, 2))

        p.values.1df p.values.2df
rs91126    0.3034837    1.385e-02
rs62927    0.1113713    2.810e-01
rs79960    0.6942913    1.506e-05
rs19348    0.0895551    2.013e-01
rs99786    0.0072618    2.713e-02
rs36984    0.1434326    5.476e-03
rs52628    0.9178502    2.469e-01
rs6699     0.0001807    4.812e-05
rs12373    0.4590596    5.772e-01
rs35215    0.2115224    4.463e-01
rs41229    0.0159203    3.931e-02
rs86267    0.1344540    5.129e-04
rs23261    0.6174657    5.535e-03
rs69208    0.0854324    1.671e-01
rs16483    0.6603136    8.925e-01
rs8558     0.4961518    5.590e-01
rs55762    0.0689901    1.913e-01
rs8124     0.2336604    1.364e-01
rs72056    0.0298914    7.391e-02
rs82369    0.0813984    2.131e-01
rs97686    0.5612452    8.410e-01
rs77065    0.7236736           NA
rs53106    0.9586501    5.387e-02
rs37378    0.2194916    8.937e-03
rs83832    0.8755190    8.936e-01
rs35431    0.4226781    4.597e-01
rs61158    0.5343400    5.050e-01
rs32410    0.0387410    5.317e-02
rs85906    0.2319977    4.759e-01
rs83977    0.2807488    2.069e-01
rs24527    0.2963307    3.894e-01
rs73721    0.0729240    9.018e-03
rs36088    0.0324330    3.061e-02
```

```
rs32998    0.5571397    7.510e-01
rs5566     0.0672924    3.919e-02
rs98256    0.5858278    7.995e-01
rs29479    0.8193228    2.904e-01
rs42938    0.0611009    1.147e-05
rs32018    0.7280652    4.997e-02
rs39483    0.2304232    4.391e-03
rs42367    0.2674484    2.496e-01
rs87640    0.0223276    3.507e-02
rs98918    0.0581770    2.983e-04

> qq.chisq(chi.squared(tests, 1), df = 1)

     N omitted  lambda
 43.000   0.000   3.401
```

QQ plot

Expected
Expected distribution: chi–squared (1 df)
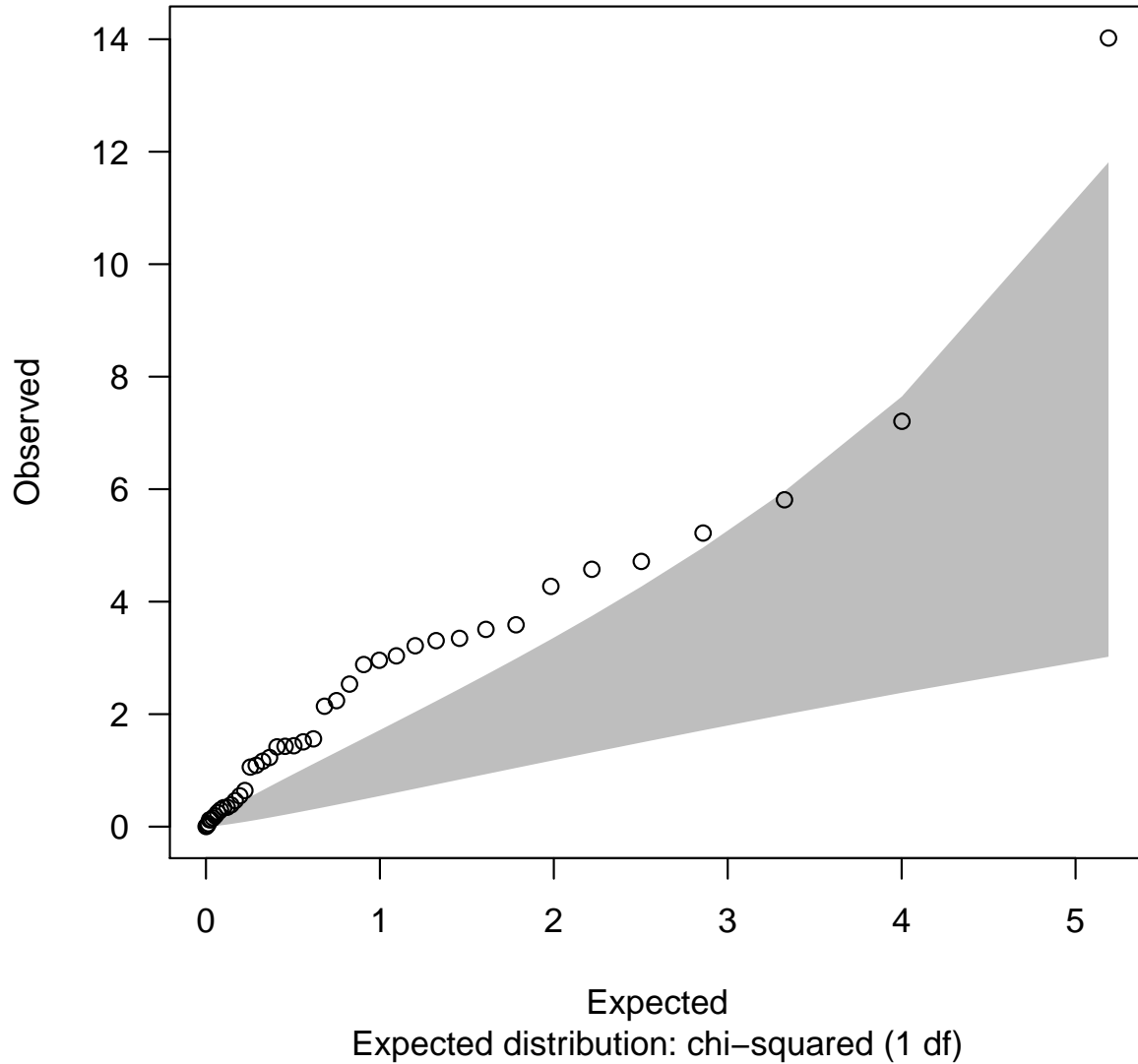
Since these SNPs were all in a region of known association, the overdispersion of test statistics is not surprising. Note that, because each family had two affected offspring, there were twice as many parent-offspring trios as families. In the above tests, the contribution of the two trios in each family to the test statistic have been assumed to be independent. When there is *linkage* between the genetic locus and disease trait, this assumption is incorrect and an alternative variance estimate can be used by specifying `robust=TRUE` in the call. However, in practice, linkage is very rarely strong enough to require this correction.