SPAGeDi 1.2

a program for Spatial Pattern Analysis of Genetic Diversity

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User's manual

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1. NOTE ABOUT SPAGeDi 1.2

SPAGeDi has been tested on several data sets and results were checked for consistency with alternative softwares whenever possible. It may nevertheless still contain bugs (corrected bugs are listed at the end of this manual). Some of these bugs are probably easy to detect by causing the program to crash or leading to obvious erroneous results for particular data sets and analyses. But others, more critical, may just cause biased results that appear plausible. Hence, it is advised to take much care checking the consistency of the information from the results file. The authors would appreciate being informed of any detected bug. The authors claim no responsibility if or whenever a bug causes a misinterpretation of the results given by SPAGeDi.

What's new in SPAGeDi?

Implementations in version 1.2:

- 1°) SPAGeDi 1.2 proposes new statistics (e.g. N_{ST}) to characterize differentiation among populations using "ordered alleles", i.e. considering the phylogenetic distance between alleles (or haplotypes), as proposed by Pons & Petit (1996). Permutation tests permit to assess whether the allele phylogeny contributes to the differentiation pattern, which can be used to test phylogeographic patterns.
- 2°) SPAGeDi 1.2 proposes an **estimator of the mean kinship coefficient** between populations (Gij) closely related to the autocorrelation of population allele frequencies (Barbujani 1987).
- 3°) SPAGeDi 1.2 proposes a new estimator of the relationship coefficient between individuals (Li et al. 1993).
- 4°) SPAGeDi 1.2 can use specific **reference allele frequencies** (to specify in a file) to compute relatedness coefficients between individuals.
- 5°) SPAGeDi 1.2 includes an iterative **procedure to estimate gene dispersal parameters** from isolation-by-distance patterns by regressing pairwise kinship coefficients on distance over a restricted distance range (this requires an estimate of the effective population density).
- 6°) SPAGeDi 1.2 provides better error messages. The most common data file errors are systematically listed in a file called "error.txt" when launching the program. As far as possible, error messages when problems occur were improved. These messages are not yet optimal so that suggestions to improve them are welcome. Empty lines in data files are now allowed. Problems when entering instructions with the keyboard under Windows 2000 and latter versions have been solved.

Implementations in version 1.1:

- 1°) SPAGeDi 1.1 can treat data from **dominant** genetic markers such as AFLP or RAPD to compute pairwise **relatedness** coefficients **between individuals**. Details about the statistics used can be found in Hardy (2003). The way to code phenotypes of dominant markers in the data file is explained in § 3.2.2.
- 2°) SPAGeDi 1.1 proposes an **allele size permutation test** indicating whether microsatellite allele sizes are informative with respect to genetic differentiation. Details about this test and its applications are given in Hardy et al. (2003).

How to cite SPAGeDi?

Hardy, O. J. & X. Vekemans (2002). SPAGeDi: a versatile computer program to analyse spatial genetic structure at the individual or population levels. *Molecular Ecology Notes* 2: 618-620.

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2. WHAT IS SPAGeDi?

2.1. PURPOSE

SPAGeDi is primarily designed to characterise the spatial genetic structure of mapped individuals and/or mapped populations using genotype data (e.g. isozymes, RFLP, microsatellites) of any ploidy level. For polyploids, analyses assume polysomic inheritance as in autopolyploids. Polyploids with disomic inheritance (allopolyploids) can be treated correctly only if alleles from different homeologous genomes can be distinguished so that genotypes are treated as diploid data. SPAGeDi can compute inbreeding coefficients as well as various statistics describing relatedness or differentiation between individuals or populations by pairwise comparisons. To analyse how values of pairwise comparisons are related to geographical distances, SPAGeDi computes 1°) average values for a set of predefined distance intervals, in a way similar to a spatial autocorrelation analysis, 2°) linear regressions of pairwise statistics on geographical distances (or their logarithm). The slopes of these regressions can potentially be used to obtain indirect estimates of gene dispersal distances parameters (e.g. neighbourhood size), and provide a synthetic measure of the strength of spatial structuring. SPAGeDi can also treat data without spatial information, providing global estimates of genetic differentiation and/or matrices of pairwise statistics between individuals or populations.

Different permutation procedures allow to test if there is significant inbreeding, population differentiation, spatial structure, or if microsatellite allele size or the phylogenetic distance between alleles carries relevant information about genetic structure.

Analyses can be carried out on data sets containing individuals with different ploidy levels, but not on data sets mixing loci corresponding to different ploidy levels within individuals (e.g. genotypes based nuclear and cytoplasmic DNA can not be analysed simultaneously, except for an haploid organism). Data from dominant markers (RAPD, AFLP) can be used to carry out analyses at the individual level with diploids (relatedness coefficients between individuals). Presently, there is no statistics adapted to dominant markers for analyses at the population level or for higher ploidy levels. One can always enter such data as haploid data (not mixed with data from codominant markers), but much caution must be taken in the interpretation of the results.

2.2. HOW TO USE SPAGEDI – SHORT OVERVIEW

SPAGeDi runs under Windows (9x or higher) but has no fancy windowing features. To launch the program just double click on the program icon or on its shortcut, or bring a data file icon on the program icon. A **single data** file must contain all individual characteristics (name, category, spatial coordinate(s), genotypes). Details of the analyses to be carried out (individual *versus* population level, population definition, statistics, permutation tests, various options) will be specified after the program has been launched. Results of the analyses are written to a **single results file**. Data and results files are text files with tab delimited pieces of information. Hence they are best opened and edited using a worksheet software such as Excel. Data files can be converted from and into FSTAT and GENEPOP formats. Although **error messages** are displayed when problems occur, typically because the data file is not properly formatted, they may not be sufficient to find out the errors. Therefore we urge users to read carefully the instructions for preparing data files (next chapter).

2.3. DATA TREATED BY SPAGeDi

SPAGeDi requires that the following information is provided for each individual: 1°) one to three spatial coordinates (facultative), 2°) value of a categorical variable (facultative), and 3°) its' genotype at each locus (missing data allowed). The categorical variable can be used to define populations or to restrict analyses within or among categories. The spatial coordinate(s) permit(s) SPAGeDi to compute pairwise distances between individuals or populations (Euclidian distances). Alternatively, pairwise distances can be defined in a separate matrix.

2.4. THREE WAYS TO SPECIFY POPULATIONS

Populations can be defined in three different ways: 1°) as categorical groups, where one population includes all individuals sharing the same categorical variable, 2°) as spatial groups, where a spatial group includes all individuals sharing the same spatial coordinates and following each other in the data file, 3°) as spatiocategorical groups, where a spatio-categorical group includes all individuals belonging to both the same spatial group and categorical group. When populations are defined using the categorical variable, each spatial coordinate of a given population is computed as the average coordinate of the individuals it contains.

2.5. STATISTICS COMPUTED

Statistics for pairwise comparisons between **populations** include:

 F_{ST} a measure of population differentiation (intra-class kinship coefficient) (Weir & Cockerham 1984)

 G_{ST} equivalent to F_{ST} but estimator with different statistical properties (Pons & Petit 1996)

 R_{ST} F_{ST} analogue based on allele size (Slatkin 1995, estimated as Michalakis & Excoffier 1996)

 $N_{\rm ST}$ $F_{\rm ST}$ analogue accounting for the genetic distances between alleles (Pons & Petit 1996)

Rho intra-class relatedness coefficient permitting among ploidy comparisons (Ronfort et al. 1998)

Gij mean kinship coefficient between populations (Barbujani 1987)

Ds Nei's 1978 standard genetic distance

 $(\delta \mu)^2$ Ds analogue based on allele size (Goldstein and Pollok 1997)

Global F- or R- statistics (inbreeding coefficients) are also provided.

Statistics designed for pairwise comparisons between **individuals** include

Kinship coefficients: 3 estimators including one for dominant markers (Loiselle et al. 1995, Ritland 1996, Hardy 2003)

Relationship coefficients: 6 estimators including one for dominant markers (Hardy & Vekemans 1999, Lynch & Ritland 1999, Queller & Goodnight 1989, Wang 2002, Li et al. 1993, Hardy 2003)

Fraternity coefficients: 2 estimators (Lynch & Ritland 1999, Wang 2002)

Rousset's distance between individuals (Rousset 2000)

A kinship analogue based on allele size (Streiff et al. 1998)

Inbreeding coefficients (computed as kinship coefficients between genes within individuals)

All statistics are computed for each locus and a multilocus weighted average. Note that an estimate of the inbreeding coefficient must be entered to compute kinship or relationship coefficients with dominant markers in diploids.

The **actual variance** of these coefficients (i.e. the remaining variance when sampling variance has been removed) can be estimated following the method of Ritland (2000). The actual variance of kinship (or relatedness) coefficients and of pairwise $F_{\rm ST}$ is necessary for *in situ*, genetic markers based inference of, respectively, the heritability and $Q_{\rm ST}$ of quantitative traits.

For pairwise coefficients, mean values per distance intervals and regression slopes on spatial distance are given (unless spatial information are lacking).

Jackknifying loci (i.e. deleting information from one locus at a time) provides approximate standard errors for the multilocus estimates.

Permutation tests whereby the statistics are computed again after that locations, individuals or genes are permuted provide ad hoc tests for spatial genetic structure, population differentiation or inbreeding coefficients, respectively. Note that permuting locations is equivalent to carrying out a Mantel test. Permutation of microsatellite allele sizes or of the phylogenetic distances between alleles also permit to test if the mutation rate is sufficient to affect the genetic structure (test of phylogeographic patterns) (Hardy et al. 2003; Pons & Petit 1996).

3. CREATING A DATA FILE

The data file is a text file. It is advised to create the data file using a worksheet program such as Excel and then save it as a "tab delimited text file". If you do not have this option, try "DOS text" ("Text Unicode" or "ASCI" formats might not work).

3.1 STRUCTURE OF THE DATA FILE

Comments lines: they are not read by the program and can be put anywhere in file. Comment lines must begin by the two characters // . Empty lines are allowed.

The data file must be in the following format, with each piece of information within a line being separated by a **tab** (i.e. each piece of information put in adjacent columns if using a worksheet program to generate the data file). Hereafter, first, second, third,... line refers to non-comment and non-empty lines.

- **first line**: 6 **format numbers** separated by a tab in the following order:
 - number of individuals
 - number of categories (0 if no category defined)
 - number of spatial coordinates (0 to 3)
 - number of loci (or the number you wish to use if the data set contains more)
 - number of digits used to code one allele (1 to 3); or set a value ≤0 to specify data from dominant markers
 - ploidy (2 = diploid; for data with several ploidy levels, give the largest)
- second line: definition of distance intervals:
 - number of distance intervals (n)
 - the *n* maximal distances corresponding to each interval

Note 1: alternatively you can enter only the desired number of intervals preceded by a **negative sign**; the program then defines the n maximal distances in such a way that the number of pairwise comparisons within each distance interval is approximately constant.

Note 2: if you do not wish distance intervals, put 0.

- third line: the names used as column labels (up to 15 characters long, without space):
 - a generic name for individuals (e.g. "Ind")
 - a generic name for categories (e.g. "Cat"), only if categories are defined
 - a generic name for each spatial coordinates (e.g. "X", "Y")
 - the name of each locus (e.g. "Pgm", "Est", ...)
- <u>fourth line and next ones</u>: individual data (each line = 1 individual):
 - name of the individual (up to 15 characters)
 - name of the category (up to 15 characters), only if categories are defined
 - coordinate along each axis (integer or floating point, up to 10 digits)
 - genotype at each locus (also separated by a tab)
- <u>last line</u> (after the last individual): the word "END" (in uppercase)

3.2. How to code genotypes?

3.2.1. Codominant data

Single locus	genotypes are	represented b	v numbers in	either of	the following ways:

Single locu	s genotypes a	are represented by	numbers in eit	her of the fo	llowing w	ays:	
		omologous gene is ters other than a ta					y number of non
12	/45	1 12	99, 23	6.6		3601	
are co	rrect genotyp	es for a diploid w	ith up to 2 digit	ts per allele.			
2°) the alle	le of each ho	mologous gene is	exactly <i>n</i> digit	ts long and	alleles are	not separated by	other characters:
12	45	0112	9923	0606		3601	
are the	e same genot	ypes as above.					
In both case	es, non nume	rical characters <u>ca</u>	nnot follow the	righter mos	st digit.		
Notes:							
1°) missing	genotypes a	are represented by	giving the valu	ie 0 : e.g.,			
(00	000,000,000,	•	000		
all	represent a r	nissing genotype.					
2°) incomp	lete genotyp	es are represented	by giving the	v alue 0 to u	ndetermin	ed alleles on the	right: e.g
· •)5-00	05,0	500	0500			<u></u>
all	represent the	e same incomplete	genotype of a	diploid (2 d	igits per al	llele).	
3°) the first	t 0's are opti	onal so that 0112	and 0606 could	also be wri	tten as 112	2 and 606, respec	tively
,	-	rels can co-occur				•	-
		e necessary number					
İ	23 125 125	121	97 123	0 0 97	123		
are	e correct geno	otypes for a tetrape	oid and two dip	oloids, respe	ctively (3	digits per allele).	
5°) do not o	confound inco	omplete genotypes	s with genotype	s for a ploic	ly level lo	wer than annound	eed: e.g.,
2 3	3 4 0	4500	0234	45			
		present 2 tetraploi genotypes), respec			ypes), a tri	ploid and a diplo	oid (the two latter
3.2.2. <u>Dom</u>	inant data (s	set the 5 th format n	umber, "numbe	er of digits"	0 , ≤ 0 and t	he 6 th format nun	nber, ploidy = 2)
Single locu	s genotypes a	are represented by	numbers in eitl	her of the fo	llowing w	ays:	
1°) if the "r	number of dig	git" is set to 0, put					
0	for a m	nissing data					
1	for a re	ecessive genotype					
2	for a d	ominant genotype					
2°) if the "r	number of dig	git" is set to –X (i.	e. a negative nu	ımber), put			
-X	for a m	nissing data					
0	for a re	ecessive genotype					
1	for a d	ominant genotype					

3.3. EXAMPLE OF DATA FILE

// this an example (lines beginning by // are comment lines)

// #ind	#cat	#coord	#loci	#dig/loc	ploidy	
5	0	2	4	2	2	
4	10	20.5	50	100		
Ind	Lat	Long	adh	got	pgm	lap
ind1	7.3	21	0101	0303	0	0101
ind2	8.4	52	101	101	102	103
3	5.11	103	0101	0303	0102	0103
4	1.0	13.2	1,1	3 3	1-2	01 03
lastind	1.94	129	0	1701	0118	1799
END						

which specifies that 5 diploid individuals, not defined by categories, with location defined by 2 spatial coordinates, are scored at 4 loci where alleles are defined by 2 digits, and that 4 distance intervals will be considered as follow: [0 to 10],]10 to 20.5],]20.5 to 50],]50 to 100]. Note that individuals 3 and 4 share the same genotypes but written in different ways.

3.4. NOTE ABOUT DISTANCE INTERVALS

When specific distance intervals are defined in the data file, the program checks that the **maximal distance** between two individuals / populations is not greater than the maximal distance of the last distance interval. Otherwise, an **additional interval** is **created**. Additional classes are also created for analyses at the individual level: an *intra-individual* class containing inbreeding coefficients (only for kinship statistics), and an *intra-group* class if individuals are organised in spatial groups (see §3.5.).

Use a point to indicate decimals (as in American notation), using a coma (as in French notation) would cause distances to be misinterpreted.

3.5. NOTE ABOUT SPATIAL GROUPS

If individuals consist of **spatial groups** that should be recognized (e.g. sibs from a given family, individuals from a given population), **individuals** belonging to a same group must **follow each other** in the data file <u>and</u> they must be given the **same spatial coordinates**. For analysis carried at the individual level, the program will then add a distance class for the pairwise coefficients between members of the same group (*intra-group* class).

For **analyses at the individual level**, when each individual receives specific spatial coordinates (no spatial groups, i.e. no two adjacent individuals in the data file share the same location), individuals are considered as independent from one another. This is typically the kind of analysis focusing on one continuously distributed population. If instead individuals are organised in spatial groups, **individuals from a same group are treated as dependent**. In such case, regression analyses do not take into account pairwise comparisons between individuals from a same group. The procedures for location permutations is also affected, as spatial group locations rather than individual locations are permuted (see $\S 5.6$.). When asking for the matrices of pairwise spatial and genetic distances between individuals, the value of the spatial distance between members of the same group is set conventionally to -1.

3.6. NOTE ABOUT MICROSATELLITE ALLELE SIZES

Several statistics are based on microsatellite allele sizes (e.g. R-statistics, Goldstein and Pollok's (1997) $\delta\mu^2$, Streiff et al. (1998) kinship analogue) using the size specified in the genotypes of the data file. Ideally, this size should be the number of repeats of the microsatellite motif. The computed statistics will still be valid if the size correspond to a constant plus the number of repeats (but the *mean allele size* information, see § 5.2., will not give the mean number of repeats). Problems may occur if allele sizes are given in terms of number of nucleotides rather than repeats. For the $\delta\mu^2$ statistic, single locus estimates will be multiplied by the square of the motif size

(the same holds for the *Variance of allele size* information, § 5.2). For R-statistics and Streiff et al. (1998) kinship analogue, single locus estimates will not be affected, but multilocus estimates would be affected if the motif size vary among loci, in which case one should change the data file, dividing allele sizes per locus by the corresponding motif size.

3.7. USING A MATRIX TO DEFINE ARBITRARY PAIRWISE SPATIAL DISTANCES

Pairwise spatial distances between individuals or populations are normally computed as Euclidian distances using the spatial coordinates. However, you can also specify each pairwise distance in an arbitrary way using a matrix. This can be useful in three cases: 1°) If you wish to consider non Euclidian spatial distances, such as distances taking into account the earth curvature, or distances more closely related to the probability of gene movements between locations. 2°) If you are not interested in spatial distances but in some other kind of pairwise distances (e.g. a morphological distance between individuals or populations) that you wish to correlate with genetic distance. 3°) If you wish to compute average statistics for particular pairwise comparisons between individuals / populations (for this purpose, you can define "distance" intervals and pairwise "distances" using integers).

The matrix of pairwise distances can be put at the end of the data file (just after the word "END", see section 3.2.), or at the beginning of another text file. The use of such matrix and its location are specified while running the program (see §4.2.5. and 4.2.9.). The matrix can be written in two formats: a matrix format or a column format.

Matrix format:

This is a square matrix. The first line must begin with the letter **M**, followed by a number representing the matrix size (# of lines and columns). Then, individual or population names corresponding to each column must be written (separated by tab). Each of the next lines begin by the corresponding individual or population name followed by the pairwise distances attributed. The last line must contain the word "END". Example:

// This is an example of a pairwise distance matrix written in matrix format with 5 rows and columns

M5	pop1	pop2	pop3	pop4	pop
pop1	0	10.3	12	6	0
pop2	10.3	0	65	18	98
pop3	12	65	0	34	54
pop4	6	18	34	0	15
pop5	0	98	54	15	0
END					

Column format:

In column format, each line corresponds to a pairwise comparison. The first line must begin with the letter C, followed by the number of lines (# of pairwise distances defined). Each of the next lines begins by the two individual or population names, separated by a tab, followed by the pairwise distance attributed. The last line must contain the word "END". Example (the following matrix contains the same information as the one above except that self comparisons are left undefined):

// This is an example of a pairwise distance matrix written in column format with 15 pairwise distances defined C15

```
pop1
       pop2
               10.3
pop1
       pop3
               12
       pop4
pop1
               6
pop1
       pop5
               0
pop2
       pop3
               65
pop2
       pop4
               18
pop2
       pop5
               98
               34
pop3
       pop4
pop3
               54
       pop5
               15
pop4
       pop5
END
```

Notes:

- 1°) For both matrix and column formats, the **order** of individuals / populations is **unimportant** (i.e. does not need to follow that of the data file).
- 2°) Self-comparisons are not taken into account.
- 3°) The **names must match** exactly those of the **data file** (case also matters!). This is straightforward for analyses at the individual level. However, for analyses at population level, population names vary: A) If one population = one **categorical group**, its name is that of the category. B) If one population = one **spatial group**, its name is that of the first individual of the spatial group in the data file. C) If one population = one **spatio-categorical group**, its name is written by joining the name of the first individual of the spatial group (as found in the data file) with the name of the category, the two being separated by the character '-'.
 - In order to create a template of the arbitrary matrix with the correct individual / population names, it can be convenient to run the program a first time without defining a pairwise distance matrix but asking to write pairwise distances and statistics in matrix or column formats (see §4.2.5. and 4.2.7.).
- 4°) Each pairwise comparison does not need to be defined, so that a matrix that does not contain all individuals / populations, or a matrix incompletely filled, are also accepted.
- 5°) Symmetrical comparisons (e.g. i-j and j-i) can not contain different distances (but one can be undefined).

3.8. DEFINING GENETIC DISTANCES BETWEEN ALLELES

When a statistic based on the genetic distances between alleles is request (e.g. N_{ST}), the program asks to specify the file containing the **distance matrix between alleles**. The latter can be put at the end of the data file or in another file, and must be a symmetrical square matrix with the following format:

First line: name of the locus followed by the allele names (numbers)

Next lines: allele name followed by the genetic distance between alleles

Example:

// This is an example of a distance matrix between alleles for a locus called "Hapl"

		1						
Hapl	1	2	3	4	15	26	7	18
1	0	6	5	2	4	4	3	3
2	6	0	1	4	2	2	3	3
3	5	1	0	3	1	1	2	2
4	2	4	3	0	2	2	1	1
15	4	2	1	2	0	2	1	3
26	4	2	1	2	2	0	3	1
7							0	2
18								0
END								

Notes:

- 1°) Locus names must match exactly those of the data file (case matters).
- 2°) The **order** of alleles must be the same along rows and columns.
- 3°) Each allele found in the data file must occur in the matrix but the latter can contain additional alleles.
- 3°) Self-comparisons are not taken into account.
- 4°) The distance between each allelic pair must be defined but it can be so only one time in the matrix (i.e. a half matrix is also accepted).

3.9. DEFINING REFERENCE ALLELE FREQUENCIES FOR RELATEDNESS COEFFICIENTS

Most statistics available for analyses at the individual level (coefficients of kinship, relationship,...) provide measures of genetic similarity between individuals that are **relative** to a sample of individuals (usually all individuals in the data set), which defines the "**reference allele frequencies**". However, specific reference allele frequencies **can be given in a distinct file** (see option § 4.3.3. - 6bis) with the following format:

First line: for consecutive loci, name of each locus followed by the total number of alleles

Next lines (one per allele): for consecutive loci, allele name followed by the allele frequency

Example:

// This is an example of a matrix with reference allele frequencies for 3 loci called "Loc1", "Loc2", "Loc3".

Loc1	5	Loc2	8	Loc3	3
1	0.3	120	0.01	2	0.67
2	0.1	122	0.04	43	0.32
3	0.05	124	0.35	3	0.01
4	0.15	130	0.13		
15	0.4	132	0.10		
		140	0.05		
		142	0.07		
		144	0.25		

Notes:

- 1°) These allele frequencies must be in a **distinct file** (the default name is "freq.txt"), not in the data file.
- 2°) Locus names must match exactly those of the data file (case matters).
- 3°) All loci in the data file must occur but additional loci may also occur (they will not be read).
- 4°) The order of alleles is unimportant.
- 5°) **All alleles found in the data file** must occur and be given a **non-null frequency**. Other alleles can also be present.
- 6°) The sum of allele frequencies at each locus must be one (sum between 0.999 and 1.001 accepted).

3.10. Present data size limitations

max. 20000 individuals

max. 2000 loci

max. 999 alleles per locus (i.e. max 3 digits per allele)

max. 30 characters for the individual, category and locus names

max. 20000 random permutations

max. ploidy = 8 (octoploid) (note that all analyses on polyploids assume polysomic inheritance)

max. 100 distance intervals

max. length of any line in the data file: 20000 characters

Please contact us if these limitations are a problem for you, we may be able to send you a recompiled version with other specifications.

4. RUNNING THE PROGRAM

The program runs on PC with Windows 9x or later versions, but has no fancy windowing features. It also runs on a Macintosh under *virtual PC*. It is written in C language using functions conforming to ANSI C standard (except for one console I/O function).

4.1. LAUNCHING THE PROGRAM

Launch the program by double-clicking on its icon; the data file must then reside in the same folder as the program file (this is also the procedure to follow if you need to import a data file). Alternatively, you can launch the program by dragging the data file (in format *SPAGeDi*) on its icon or on the icon of a shortcut to the program; the data file can then reside anywhere and the result file will be written in the directory of the data file.

Error messages are given when files cannot be opened, data files are not well formatted or contain inconsistent information. These messages are not yet optimal and you may have difficulties finding out what is wrong in your data file (suggestions to improve this are welcome). When launching SPAGeDi, an error file "error.txt" is opened (and its previous content erased) and common errors made when preparing data files are listed. Additional information is added in this file whenever a problem occurs. SPAGeDi checks that the number of individuals and the number of categories found are the one specified in the data file, but there is no check for the number of loci (analyses considering only the first loci listed can thus be done by adjusting the number of loci given in line containing the format numbers).

4.2. SPECIFYING THE DATA / RESULTS FILES

Once the program is launched, you are requested to enter the name of the data file (unless you launched the program by dragging the data file icon on the program) and the name of the results file.

If you just press *RETURN* to these questions, the default names "in.txt" and/or "out.txt" will be considered as data and results files, respectively (this can be useful if you wish to carry out many different analyses on the same data set without having to enter the file names each time).

You can also **import data** from a file in FSTAT (Goudet 1995) or GENEPOP (Raymont and Rousset 1995) format. Therefore, press *SPACE* and then *RETURN* when asked to enter the data file name, and select the format of the data file (FSTAT or GENEPOP). A new data file in *SPAGeDi* format will then be created, but it will not contain spatial information, so that you need to add them (as spatial coordinates per individual or as a matrix of pairwise distances), unless you don not need spatial analyses.

If a file with the same name as the results file already exists in the folder, the program will ask if you wish to: erase the existing file first (enter 'e'), add results to the end of this file (enter 'a' or simply press *RETURN*), or change the name of the ouput file (enter the new name).

Once the data and result files are specified, the program first displays the basic information from the data file on the screen and waits for user to hit the *RETURN* key. The first set of information displayed is: the number of individuals, the number of categories and their names, the number of spatial coordinates and their names, the number of loci and their names, the number of digits used to specify alleles, the specified ploidy of the data, and the number of individuals of each ploidy. At this stage, if some individuals have missing genotypes at all loci, a warning message is addressed (**but the analysis can go on anyway**), and if different loci suggest different ploidy levels within some individuals, a warning message is addressed and the data file must be modified (the program stops here). The second set of information displayed is the groups recognised (categorical, spatial and spatio-categorical ones) with the minimal and maximal numbers of individuals per group.

4.3. SELECTING THE APPROPRIATE OPTIONS

You define the analyses to carry out and the results to write down by selecting options in 4 successive panels: 1°) *Level of analyses*, 2°) *Statistics*, 3°) *Computational options*, 4°) *Output options*. Some of the options will not be available depending on the structure of the data. You can come back to the beginning at different stages if you made an error of selection.

4.3.1. Level of analyses: individual vs population

Analyses are carried out at the individual level or population level. When both categorical and spatial groups occur, you have also the choice among three different ways to define populations: as categorical, spatial, or spatio-categorical groups. If there are no categorical nor spatial groups in the data set, analyses are restricted to the individual level.

4.3.2. Statistics

You must select the statistics to be computed (you can select several simultaneously). These statistics are computed for each pair of individuals or populations and the average values per distance interval as well as the regression statistics are given in the results file. More details about those statistics are given in § 6.1 and 6.2.

For analyses at the **individual level** with <u>codominant</u> markers, 11 statistics for pairwise comparisons between individuals are available:

- 1°) A kinship coefficient estimated according to J. Nason (described in Loiselle et al. 1995).
- 2°) A kinship coefficient estimated according to Ritland (1996).
- 3°) A relationship coefficient computed as Moran's I statistic (Hardy and Vekemans 1999).
- 4°) A relationship coefficient estimated according to Queller and Goodnight (1989).
- 5°) A relationship coefficient estimated according to Lynch and Ritland (1999) (r coef).
- 6°) A relationship coefficient estimated according to Wang (2002) (r coef).
- 7°) A relationship coefficient estimated according to Li et al. (1993).
- 8°) A fraternity coefficient (4-genes coefficient) estimated according to Lynch and Ritland (1999) (△ coef).
- 9°) A fraternity coefficient (4-genes coefficient) estimated according to Wang (2002) (Δ coef).
- 10°) A distance measure described in Rousset (2000) (the one called \hat{a} by Rousset).
- 11°) A correlation coefficient between allele sizes for use with microsatellites (Streiff et al. 1998).

Note: statistic 10° can not be computed for haploid data, and statistics 4° , 5° , 6° , 7° , 8° and 9° can presently be computed only for diploid data (5° , 6° , 8° and 9° also assume a population with Hardy-Weinberg genotypic proportions).

For the kinship coefficients, intra-individual values are also computed (as kinship between genes within individuals), providing estimates of an inbreeding coefficient.

For analyses at the **individual level** with <u>dominant</u> markers in <u>diploids</u> (see §3.2.2), 2 statistics are available:

- 1°) A kinship coefficient (Hardy 2003).
- 2°) A relationship coefficient (Hardy 2003).

For analyses at the **population level** with <u>codominant</u> markers, there are 8 choices for global and pairwise statistics between populations:

Statistics based on allele identity / non-identity

- 1°) Global F-statistics and pairwise F_{ST}
- 2°) Global F-statistics and pairwise Rho
- 3°) Global G_{ST} and pairwise G_{ST}
- 4°) Global G_{ST} and pairwise Gij
- 5°) Global F-statistics and pairwise Ds (Nei's standard genetic distance, Nei 1978)

Statistics based on allele size for microsatellites

- 6°) Global R-statistics and pairwise R_{ST}
- 7°) Global R-statistics and pairwise dm2 (Goldstein's $(\delta u)^2$ distance, Goldstein and Pollock 1997)

Statistics based on distances between alleles

8°) Global $N_{\rm ST}$ and pairwise $N_{\rm ST}$

When a statistic based on distance between alleles is asked, the program will ask to specify the file containing the matrix of distances between alleles.

4.3.3. Computational options

Once the statistics are chosen, you can select among different options regarding computations (several options can be selected simultaneously):

1°) Use a matrix to define pairwise spatial distances.

This option allows to define pairwise spatial distances between individuals / populations in an arbitrary way (otherwise, Euclidian distances are computed from the spatial coordinates given in the data file). Therefore, you must enter the name of the file containing the matrix (if the matrix follows the genotype information in the data file, just press *Return*). Details of the format of the matrix are given in § 3.6.

2°) Make partial regression analyses (i.e. over restricted distance range).

This option allows to define a distance range within which the spatial regression is computed, a useful option for gene dispersal parameter estimations (§ 6.3.). If this option is not selected, the regressions are carried out using all pairwise comparisons, except those with a distance of zero for the regressions on ln(distance). Otherwise, minimal and maximal distances defining the range must be given. Entering no values (i.e. just pressing *RETURN*) means that the minimal or maximal distance is not bounded.

3°) Make permutation tests.

This option allows to test the significance of different statistics by random permutations of genes, individuals, locations, or allele sizes. More details in § 4.3.5.

4°) Jackknife over loci.

With this option, mean jackknifed estimators and jackknife standard errors are computed for multilocus average statistics. Jackknifying necessitates at least 2 polymorphic loci, but at least 6 polymorphic loci should be necessary for reliable estimates.

5°) Restrict pairwise comparisons within or among (selected) categories.

If the data are organised in categorical groups and analyses are carried out at the level of individuals or populations defined as spatio-categorical groups, you can select the type of pairwise comparisons for which the pairwise statistics are to be computed:

- 1°) All pairs (i.e. irrespective of categorical groups) = default option
- 2°) Only pairs within categories
- 3°) Only pairs among categories
- 4°) Only pairs within a specified category
- 5°) Only pairs between two specified categories

When 4° or 5° is selected, the name(s) of the category(ies) is(are) to be given.

When 2° or 4° is selected and analyses are carried out at the individual level, you must select between two reference allele frequencies to compute the statistics (see § 6.1.1. for explanations):

- 1°) whole sample (i.e. pairwise coefficients are computed relative to the whole sample)
- 2°) sample within category (i.e. pairwise coefficients are computed relative to the sample to which the pair of individuals belongs)

6°) Pairwise Fst (or Rst, or Rho) provided as Fst/(1-Fst) ratio.

When this option is selected, pairwise differentiation between population will be estimated using $F_{\rm ST}/(1-F_{\rm ST})$ ratios. This is useful to analyse isolation-by-distance patterns because $F_{\rm ST}/(1-F_{\rm ST})$ is expected to vary linearly with the distance (or its logarithm). See §6.3.

6bis°) Define reference allele frequencies to compute relatedness coefficients.

When this option is selected, pairwise relatedness coefficients will be computed relative to reference allele frequencies given in a separate file (see §3.9. for the format). *SPAGeDi* will ask the name of this file. This option cannot be applied for the statistics developed for dominant markers in diploids, the relationship coefficient computed as a Moran's *I* statistic, and Rousset's (2000) *a* coefficient.

4.3.4. Output options

A second set of options concerns the information given in the results file:

1°) Report allele frequencies for each population / category (otherwise only averages reported).

In the results file, global allele frequencies and gene diversities are reported. Activating this option means that this information will also be given for each population (or for each categorical group in the case of analyses at the individual level including categories).

2°) Report all stat of regression analyses (otherwise only slopes reported).

When this option is activated, the following statistics of the regressions of pairwise statistics on spatial distances are provided: slope, intercept, determination coefficient, number of pairs, mean and variance of values of (log) distance and statistics.

3°) Report matrices with pairwise spatial distances and genetic coefficients.

With this option, pairwise spatial distances and pairwise statistics are given at the end of the results file. You must also specify whether the pairwise statistics are to be given for each locus or only the multilocus estimates, and whether pairwise values are to be written only in columnar form or also in matrix form. You can also select Phylip format which gives a square matrix of genetic distances that can be copied directly to a text file for further analyses (there is no tab delimitations). Note that in Phylip format, negative genetic distances are given the value -0.0000 Estimates of the inbreeding coefficient for each individual are given in the columnar format if you asked to compute a kinship coefficient between individuals (the inbreeding coefficients given are computed as kinship coefficient between homologous genes within individual).

4°) Report actual variance of pairwise genetic coefficients (Ritland 2000).

With this option activated, the actual variance (i.e. excluding sampling variance) of pairwise statistics is given for each distance class following the approach described in Ritland (2000), which requires independent loci (at least two). An estimate of the standard error by jackknifing over loci is also given with at least 3 loci. This variance is useful to compute marker based estimates of the heritability (h^2) or population differentiation (Ost) at quantitative traits (Ritland 1996, 2000).

5°) Convert data file into GENEPOP or FSTAT format.

This option allows to create a data file that can be used by the software FSTAT (Goudet 1995) or GENEPOP (Raymond and Rousset 1995), and it is available only with diploid data. If analyses were asked at the population level, the GENEPOP or FSTAT file codes data for the same populations as selected. For analyses selected at the individual level, the FSTAT file code data as a single population, whereas the GENEPOP file code data as if each individual constituted a single population (this is the necessary format to use Rousset's pairwise distance between individuals in GENEPOP).

6°) Estimate gene dispersal sigma.

For analyses at the individual level, this option can be used to estimate the gene dispersal distance parameter sigma from the regression of pairwise kinship coefficients on the logarithm of the distance. You must assume that genotypes come from a two-dimensional population at drift-dispersal equilibrium so that theoretical expectations of isolation-by-distance models hold (§ 6.3.). You will be asked to enter the effective population density. *SPAGeDi* will then apply an iterative procedure to estimate the sigma from the genetic structure on a restricted distance range (see § 6.3.).

4.3.5. Permutation tests

If permutation tests are selected, you have two sets of additional options (you can select several at once): Firstly (only if statistics based on allele size or distance between alleles have been selected),

$1^\circ)$ Test of genetic structuring (permuting genes, individuals and/or locations)

To test individual inbreeding, population differentiation, and/or spatial structure.

2°) Test of mutation effect on genetic structure (permuting alleles)

To test if the allele size (microsatellites) or the phylogenetic distance between alleles is informative with respect to genetic structuring.

3°) Test of mutation effect on genetic differentiation for each pair of populations

To test, for each pair of populations, if the allele size or the phylogenetic distance between alleles is informative with respect to differentiation.

Secondly,

1°) Report only P-values (otherwise details of permutation tests are reported)

If this option is selected, only P-values for 2-sided tests are reported. Otherwise, the following details are given: object permuted, # permutations, # of different values of the statistic after permutation, observed values before permutation, mean values after permutation, standard errors of mean values after permutation, 95% confidence intervals, P-values of 1- and 2-sided tests.

2°) Define # of permutations for each randomised unit (otherwise same #)

Allows to define a high number of permutations for the statistics that most interest you, and no or few permutations for the ones that are not of interest for you or that would take a lot of computation time.

3°) Initialise random number generator (otherwise initialisation on clock)

Define initial seed for random number generator, otherwise the latter is defined according to the computer's internal clock (this option is useful for debugging).

You must then enter the number(s) of permutations you wish. On large data sets, resampling can be time consuming, hence there is a compromise between computation time and precision of the probability (*P*-values). It is advisable to enter <u>at least</u> 200 if you are satisfied with a 5% significance level, 1000 for a 1% level, 10000 for a 0.1% level. Enter "0" if you do not need tests.

4.4. INFORMATION DISPLAYED DURING COMPUTATIONS

Once the program proceeds to the calculations, it displays the computational stage: computation of allele frequencies, of distance intervals, of pairwise statistics, permutation tests. The program can be stopped anytime by pressing "Ctrl" + "c". When the computations are finished, a message will appear on the screan and pressing any key will close the window. You can proceed to examination of the results file. If the program crashed, do not forget to open the file "error.txt", because this may give you some information on the origin of the problem.

Details relative to distance intervals are displayed once computed, and computations proceed unless there are more than 20 intervals, in which case you must press *RETURN* to view them in turn. Each interval (class) is characterised by

1°) <i>max d</i>	its maximal distance (the minimal distance is the maximal distance of the preceding interval)
2°) mean d	the average distance between individuals / populations for the pairs belonging to the interval
3°) mean ln(d)	idem but using the ln(distance) between individuals / populations
4°) # pairs	the number of pairwise comparisons belonging to the interval
5°) % partic	the proportion (%) of all individuals / populations represented at least once in the interval
6°) CV partic	the coefficient of variation of the number of times each individual / population is represented

Notes:

- 1°) If analyses are restricted to pairwise comparisons within or among (specified) category(ies), the information per distance intervals considers only pairs satisfying these conditions.
- 2°) Information on distance intervals can be useful for fine-tuning them. For example, low % partic and/or high CV partic means that the statistics computed for the corresponding interval involve data from only a fraction of the individuals / populations. Hence, as a rule of thumb, we advise that for each distance interval: % partic > 50%, and CV partic <= 1. For analyses at the individual level we also advise that # pairs > 100, given the large standard errors typically observed for pairwise coefficients between individuals (with many loci or highly polymorphic loci this number could be reduced, but with a low level of polymorphism it might be better to consider # pairs > 500).

5. INTERPRET THE RESULTS FILE

All the results are found in a single results file. The results file can be read as a text file or as an EXCEL worksheet; in the latter case you can change the extension into .xls and open the file by double-clicking on its icon. The results appear in the following order.

5.1. BASIC INFORMATION

First, the basic information as it appeared on the screen when running the program is written: names of data and results files, numbers of individuals, categories, spatial coordinates and loci, names of categories, spatial coordinates and loci, ploidy, numbers of individuals for each ploidy level, number of categorical, spatial and spatio-categorical groups (see § 4.2.).

5.2. ALLELE FREQUENCY ANALYSIS

Second, for each locus are written: the number of missing genotypes (# missing genotypes), the number of incomplete genotypes(# incomplete genotypes), the total number of defined genes (# of defined genes), the number of alleles with non zero frequency (# alleles), the gene diversity corrected for sample size (He), the name (or size) of each allele (allele names or allele size) (i.e. the number given in the data file), and the allele frequencies (allele frequencies). When a statistic based on allele size (e.g. R-statistics) has been selected, the mean (Mean allele size) and variance (Variance of allele size) of allele sizes are also given. This information is given for the whole sample and, if asked when selecting the options, for each population (analysis at population level) or each category (analysis at individual level).

If relatedness coefficient were computed using specified reference allele frequencies (individual level analyses), the latter will be written.

5.3. Type of analyses

After the allele frequencies information, it is specified whether the analyses are carried out at the individual or population level, if pairwise comparisons are restricted to pairs within or among category(ies), and, for analyses at the individual level, if statistics are computed on basis of the global (whole sample) or local (within category) allele frequencies (for comparisons within (a) category) or relative to given reference allele frequencies.

5.4. DISTANCE INTERVALS

Next, for each distance interval corresponding to a column, are written:

- Dist classes: the names of the distance classes (1, 2, ...)
- Max distance: the maximum distance defining the interval: distance interval c = 1 Max dist(c-1), Max dist(c-1)
- Number of pairs: the number of pairs of individuals separated by the given distance interval
- % partic: the percentage of individuals participating at least once in a pairwise comparison within the interval
- *CV partic*: the coefficient of variation (i.e. the ratio of the standard deviation over the average) of the number of times each individual participate in pairwise comparisons within the interval
- Mean distance: the average distance separating pairs of individuals within the interval
- Mean ln(distance): the average natural logarithm of the distance separating pairs of individuals within the interval

Note: For analyses at the individual level, an *intra individual* class is added for comparison of genes within individual (only defined for kinship statistics when ploidy is larger than one), and this class actually corresponds to an inbreeding coefficient. When individuals consist of groups, the distance class "1" corresponds to *intra group* comparisons.

5.5. COMPUTED STATISTICS

For each selected statistic, the following results are given for the multilocus estimate and each locus:

- in columns labelled $F_{\rm IT}$, $F_{\rm IS}$, $F_{\rm ST}$ or $R_{\rm IT}$, $R_{\rm IS}$, $R_{\rm ST}$ or $G_{\rm ST}$ or
- in columns corresponding to each distance class: the average value of the pairwise coefficients computed over all pairs of individuals or populations within the distance interval (all pairs of genes within individuals in the case of the "intra individual" class, for analyses at the individual level).
- <u>under the column "average"</u>: the average value of the coefficients computed over all pairs of individuals or populations, whatever the distance (for analyses at individual level, it includes *intra group* class but not *intra individual* class).
- <u>under "distance range for regression analyses"</u>: the distance range used to compute regressions of pairwise statistics on spatial distance or ln(distance).

The next columns report the results of the regression analyses, first with the linear distance, then with the ln(distance). If the option "Report details of regression analyses" has not been selected (see §4.2.5), only the slopes (b-lin and b-log) are given; otherwise the following statistics are reported for each regression analysis:

- the slope b
- the intercept a
- the coefficient of determination r2 (i.e. squared correlation coefficient)
- the number of pairwise comparisons N (taking account of missing data)
- the mean (Md) and variance (Vd) of pairwise distances or $\ln(\text{distances})$
- the mean (Mv) and variance (Vv) of pairwise statistics

If the option *Jackknifing over loci* has been selected (see §4.2.5), results of a jackknife procedure deleting each locus at a time are given on the two lines following the information of the last locus: the first line gives the jackknifed estimates, the second one gives their standard errors. Calculations follow Sokal and Rohlf (1995, p.821).

Notes:

- 1°) For analyses at the individual level, the *intra individual* kinship coefficient is an inbreeding coefficient expressing the departure from Hardy-Weinberg genotypic proportions (cf. F_{IS}). When individuals consist of spatial groups corresponding to different populations, this is equivalent to F_{IT} (not F_{IS}). Kinship statistics for the *intra group* class provides an estimator similar to F_{ST} if groups correspond to different populations.
- 2°) For analyses at the individual level, the slopes of the regressions do not include the pairs of individuals within spatial groups (*intra group* class). As slopes do not depend on an arbitrary choice of distance, they offer a convenient measure of the degree of spatial genetic structuring. Moreover, under some conditions, these slopes can be related to population genetic parameters like neighbourhood size (see §6.3.).

5.6. PERMUTATION TESTS

If permutation tests are selected as option (see §4.2.5), results of these tests are written after the pairwise coefficients. These tests are based on the comparison of the observed values with the corresponding frequency distributions when random permutations of the data are performed. For each locus and the multilocus estimates, tests are given for global statistics (population level analyses), each distance class, and the slopes of the regressions analyses.

The following information is reported (unless the option "Report only P-values" has not been selected - see §4.3.5 – in which case only P-values for the two-sided tests are given):

- the object (genes, individuals or location) permuted (and how): Object permuted

- the number of valid permutations (i.e. for which the statistic was computable): N valid permut

- the number of different values obtained for the different permutations: N different permut val

- the observed value (i.e. before permutation): Obs val

- the average value after permutation:

Mean permut val - the standard error of the distribution of values after permutation: SD permut val

- the lower 95% confidence interval value: 95%CI-inf

- the upper 95% confidence interval value:

95%CI-sup - the P-value for the 1-sided test observed value < permuted value: P(1-sided test, H1: obs < exp)

- the P-value for the 1-sided test observed value > permuted value: P(1-sided test, H1: obs > exp)

- the P-value for the 2-sided test observed value different from permuted value: P(2-sided test, H1: obs!=exp)

The following **code** is used to designate the object permuted and how it is permuted (*Objected permuted*):

GaI permutation of Genes among all Individuals

GaIwC permutation of Genes among Individuals within Category GaIwP permutation of Genes among Individuals within Population

permutation of Individuals among Spatial Groups IaSG

permutation of Individuals among Spatial Groups within Category **IaSGwC**

permutation of Individuals among all Populations IaP

IaPwC permutation of Individuals among Populations within Category ILaI permutation of Individual Locations among all Individuals

ILaIwC permutation of Individual Locations among Individuals within Category **SGLaSG** permutation of Spatial Group Locations among all Spatial Groups

SGLaSGwC permutation of Spatial Group Locations among Spatial Groups within Category

PLaP permutation of Population Locations among all Populations

PLaPwC permutation of Population Locations among Population within Category

ASaAwL permutation of <u>Allele Sizes among Alleles within Locus</u>

RCoDMbA permutation of Rows and Columns of Distance Matrices between Alleles

When permutation of an object is done within category, it means that the permuted objects remain in their original categorical group after permutation. This is done when pairwise comparisons are restricted to within category(ies) (see §4.2.3.).

As the preceding code shows, the **object permuted** varies:

- Genes are permuted among individuals, each locus independently, for tests on $F_{\rm IS}$, $F_{\rm IT}$, $R_{\rm IS}$, $R_{\rm IT}$ and intra individual coefficients. Missing data are not permuted (i.e. permutation concerns only defined genes). For $F_{\rm IS}$ and $R_{\rm IS}$, genes are permuted only within population.
- Individuals (i.e. whole genotypes) are permuted among populations or spatial groups for tests on global $F_{\rm ST}$, $R_{\rm ST}$, Rho, $G_{\rm ST}$, $N_{\rm ST}$ and *intra group* coefficients.
- Individual Locations (for analyses at the individual level without spatial groups), Spatial Group Locations (for analyses at the individual level with spatial groups), or <u>Population Locations</u> (for analyses at the population level) are permuted among the available locations for tests on each distance class (except the intra individual and *intra group* ones), and tests on the regression slopes. This is equivalent to a Mantel test between a matrix of genetic distances and a matrix of geographic distances.
- Allele Sizes represented within each locus are permuted among allelic states to test if allele sizes are informative (assuming stepwise mutations) on global R-statistics ($R_{\rm IS}$, $R_{\rm IT}$, $R_{\rm ST}$), pairwise $R_{\rm ST}$ (and regression slope), or the correlation coefficient between allele sizes (Streiff et al. 1998) for individual level analyses (cf. Hardy et al. 2003).
- Rows and Columns of Distances Matrices between Alleles are permute to test if the genetic distances between alleles are informative on global or pairwise N_{ST} (and regression slope) (cf. Pons & Petit 1996; Burban et al. 1999).

Notes:

- 1°) Tests based on individual permutations indicate whether population or spatial groups are genetically differentiated, whereas tests based on location permutations indicate whether the degree of differentiation or relatedness between individuals, spatial groups or populations depends on the geographical distance.
- 2°) When using an **arbitrary matrix** to define **pairwise distances** (§3.6.), location permutations correspond to permutations of the rows and columns of this matrix (as in a Mantel test).
- 3°) For tests based on individual or location permutations, the presence of missing data is not taken into account, i.e. even if there is no genes defined at some loci for a given individual (location), this individual (location) will be permuted with the other ones. Hence, these tests can be biased for loci with a significant proportion of missing data. In such case, it might be preferable to make the tests on each locus separately using single locus data files in which missing data are removed.
- 4°) When analyses are restricted to **comparisons among categories** (see §4.2.3.), care must be taken in the **interpretation of the tests** based on location permutations (i.e. tests on pairwise coefficients and regression slopes). It can be tempting to interpret these tests as indicating whether the spatial structures within each category have developed independently or not, because if gene flow occurs or had occurred recently among categories, one would indeed expect a spatial correlation between the patterns of genetic variation of the different categories. However, these tests are **biased** for such purpose because they are based on random permutations that not only make the spatial structures of the different categories independent from one another, but also break down the structure within each category (ideally the level of structuring within category should be kept intact). Therefore, a test may be significant because the patterns of spatial genetic variation within category match for different categories just by chance, whereas these structures developed truly independently. Nevertheless, a test of the independence of the spatial structures of the different categories can be done if different independent loci (i.e. in linkage phase equilibrium within category) are available, using conventional non parametric methods (e.g. sign tests) on the regression slopes per locus.

5.7. MATRICES OF PAIRWISE DISTANCES AND STATISTICS

When asked as option (see §4.2.5), matrices of pairwise geographical distances between individuals and pairwise coefficients (for the multilocus estimates, optionally for every locus) are provided in two possible formats (defined as options, see §4.2.7): as square matrices (lines correspond to one individual / population, column to another), or in columns (first and second individuals / populations are figured in two columns, spatial distances, multilocus estimates and/or per locus estimates are given in the next columns). You can also select Phylip format which gives a square matrix of genetic distances that can be copied directly to a text file for further analyses (there is no tab delimitations). Note that in Phylip format, negative genetic distances are given the value –0.0000. Estimates of the inbreeding coefficient for each individual are given in the columnar format if you asked to compute a kinship coefficient between individuals. Note that the value reported for the geographical distance between individuals belonging to the same spatial group is –1.

6. TECHNICAL NOTES

6.1. STATISTICS FOR INDIVIDUAL LEVEL ANALYSES

Analyses at the individual level are carried out by computing measures of genetic relatedness or genetic distance between individuals for each possible pair (unless stated differently, see §4.3.3). These pairwise coefficients are computed for each locus and a multilocus weighted average. They are regressed on pairwise spatial distances and they are averaged to compute mean values per distance interval. Hence, a multilocus estimate for a distance interval is computed by first averaging pairwise coefficients over loci (weighted average), then averaging multilocus pairwise coefficients over all pairs included in the distance interval.

For codominant data, *SPAGeDi* allows the user to compute five types of "relatedness" coefficients between individuals: "kinship", "relationship" and "fraternity" coefficients, plus a distance measure based on allele identity, and a kinship analogue based on allele size. For some of these coefficients, several estimators are available, so that a total of 13 different statistics can be estimated. Comparisons of the statistical properties of different estimators can be found in Lynch & Ritland (1999), Van de Casteel et al. (2001), Wang (2002), Vekemans & Hardy (2004). The fraternity coefficient is a "4-genes" coefficient, in the sense that it is based on the simultaneous comparison of all of the 4 homologous genes of two diploid individuals. The other coefficients are "2-genes" coefficients, because they are ultimately based on comparisons between 2 homologous genes. For dominant data, *SPAGeDi* allows to compute two types of "relatedness" coefficients between individuals: "kinship" and "relationship" coefficients. There is no unified terminology for these different coefficients so that we attempt to define them below.

Most statistics available are relative measures of genetic similarity that depend on the definition of a reference sample or reference allele frequencies (see below). Specific reference allele frequencies can be defined in a distinct file ($\S 3.9$, $\S 4.3.3$.) but they will not be taken into account for the relationship estimator based on Moran's I statistic ($\S 6.1.2$), Rousset's distance measure ($\S 6.1.4$), and statistics developed for dominant markers in diploids. Note that sampling bias corrections that are normally applied for some statistics (see below) are not applied when specific reference allele frequencies are defined.

Synthetic table of the statistics proposed by SPAGeDi for individual level analyses and their properties.

Coefficient	Intra-indiv.	Assu	Assumptions		Statistical properties ³	
Estimator (ref)	estimate	Ploidy	Inbreeding ²	Accuracy	Precision	
	(inbreeding)			(low bias)	(low variance)	
Kinship						
Loiselle et al. 1995	+	1 to 8		+++	++	
Ritland 1996	+	1 to 8		++	+++	
Hardy 2003 (for dominant marker)		2	$F_{\rm I}$ to give	++	++	
Relationship						
Hardy & Vekemans 1999 (Moran's I)		1 to 8		+++	++	
Lynch & Ritland 1999		2	H-W	+++	++	
Queller & Goodnight 1989		2^{1}		+++	++	
Wang 2002		2	H-W	+++	++	
Li et al. 1993		2	H-W	+++	+++	
Hardy 2003 (for dominant marker)		2	$F_{\rm I}$ to give	++	++	
Fraternity						
Lynch & Ritland 1999		2	H-W	++	++	
Wang 2002		2	H-W	+++	+++	
Rousset's 'a'						
Rousset 2000		2 to 8	no selfing	+++	+	
Kinship analogue based on allele size						
Streiff et al. 1998	+	1 to 8		+++	+	

¹ Queller & Goodnight's estimator is defined for any ploidy level but *SPAGeDi* computes it only for diploids.

 $^{^{2}}$ *H-W* means that the estimator was derived assuming Hardy-Weinberg proportions and may be biased under inbreeding. F_{I} to give means that an independent estimate of the individual inbreeding coefficient must be provided.

³ Statistical properties are based on literature results (Lynch & Ritland 1999, Van de Casteel et al. 2001, Wang 2002, Vekemans & Hardy 2004) and personal experience. These indications must be considered with caution because the actual ranking of the performances of the statistics depend on the data set.

6.1.1. Kinship coefficient

Definition and interpretation

In a generic way, kinship coefficients, also called coancestry coefficients, are based on the probability of identity of alleles for two homologous genes sampled in some particular way. In the case of a kinship coefficient between two individuals, the two genes are randomly sampled within each of the two individuals. F-statistics are also kinship coefficients but for genes sampled in different ways (see §6.2.).

A kinship coefficient (F) is often defined as the probability of identity by descent of the genes compared (e.g. Ritland 1996) but estimators based on genetic markers actually estimate a "relative kinship", that can be defined as ratios of differences of probabilities of identity in state (Rousset 2002; Vekemans & Hardy 2004). Thus, equating these kinship coefficients with probability of identity by descent is not true in general (Rousset 2002). In the case of two individuals i and j, the kinship coefficient between them can be **defined as Fij=(Qij-Qm)/(1-Qm)**, where Qij is the probability of identity in state for random genes from i and j, and Qm is the average probability of identity by state for genes coming from random individuals from the sample (i.e. "reference population" = sample). As defined here, kinship is not really a population genetics parameter as it depends on an arbitrary sample. Note also that with this definition, **negative** relative kinship **coefficients** naturally occur between some individuals, it simply means that these are less related than random individuals (a definition equating kinship and probability of identity by descent would not allow negative values).

Changing the reference population

In some contexts, one wishes to compare estimates of kinship coefficients with some expected values derived from pedigree information. For example, in the case of sib families coming from non-inbred diploid parents, the kinship between sibs is expected to be 0.125 for half-sibs and 0.25 for full-sibs according to standard computations (e.g. Lynch and Walsh 1998). These are actually the expected values of a kinship coefficient relative to the parental generation (i.e. where Qm is for random genes from the parental generation, which is here the "reference population"). Thus, kinship is not relative to the same "reference population" when computing it from a data set containing some sib-families ("reference population" = sample) and when considering expected values from pedigree information ("reference population" = ancestors of the genealogy, which are assumed to be "unrelated"). One can however switch between these different references if one can find pairs of individuals in the data set that are expected to be "unrelated" in the sense of the putative pedigree (cf. Hardy 2003). For sibfamilies, this would be the case of pairs of individuals belonging to different families. Let F° be the kinship between individuals from different sib families as computed from the sample reference, you can then compute kinship coefficients relative to the pedigree reference, Fij', as Fij'=(Fij-F°)/(1-F°). These Fij' are expected to be 0.125 or 0.25 in case of half and full sibs, respectively. When allele frequencies of a reference population (e.g. the parental population) can be assessed precisely, an alternative approach consists in estimating kinship (or other relatedness) coefficients using specified reference allele frequencies corresponding to this reference population (see §3.9, §4.3.3). All alleles found in the individuals being compared must then have a non-null frequency (SPAGeDi tests this).

Estimators

For <u>codominant</u> markers, *SPAGeDi* proposes two estimators of kinship (coefficients relative to the sample): 1°) a kinship coefficient computed as a correlation coefficient between allelic states proposed by J. Nason (Loiselle *et al.* 1995), 2°) a kinship coefficient estimated according to Ritland (1996).

1°) is computed as $F_{ij} = \Sigma_l \left[\sum_a (\sum_{ci} \sum_{cj} (x_{lcia} - p_{la}) (x_{lcja} - p_{la}) / \sum_{ci} \sum_{cj} 1 \right] + \sum_a (p_{la} (1 - p_{la}) / (n_l - 1)) \right] / \sum_l \sum_a (p_{la} (1 - p_{la}))$ where x_{lcia} is an indicator variable $(x_{lcia} = 1)$ if the allele on chromosome c at locus l for individual i is a, otherwise $x_{lcia} = 0$), p_{la} is the frequency of allele a at locus l in the reference sample, n_l is the number of genes defined in the sample at locus l (the number of individuals times the ploidy level minus the number missing alleles), and Σ_{ci} stands for the sum over the homologous chromosomes of individual i. Here, the term involving $(n_l - 1)$ is a sampling bias correction. This estimator should be equivalent to the one computed by John Nason's FijAnal software, except that the bias correction might differ slightly.

Note that this formula is identical to:

$$F_{ij} = \sum_{l} \left[\sum_{a} (p_{ila} - p_{la})(p_{jla} - p_{la}) + \sum_{a} (p_{la}(1 - p_{la})/(n_{l} - 1)) \right] / \sum_{l} \sum_{a} (p_{la}(1 - p_{la}))$$
 where p_{ila} is the frequency of allele a at locus l in individual i .

2°) is computed as $F_{ij} = \Sigma_{l}((\Sigma_{a}\Sigma_{ci}\Sigma_{cj}(x_{lcia}x_{lcja}/p_{la})/\Sigma_{ci}\Sigma_{cj}1) - 1)/\Sigma_{l}(m_{l} - 1)$

where m_l is the number of different alleles found in the sample at locus l. Note that the bias correction consisting in removing the compared individuals when computing p_{la} , as suggested by Ritland (1996), is not applied.

The two estimators differ mainly by the way information from the different alleles and different loci are combined to provide average estimates per locus or multilocus estimates. Basically, Ritland's estimator weights allele contributions by $1/p_{la}$, giving more weight to rare alleles, and this estimator usually shows lower sampling variance, especially for unrelated individuals (Vekemans & Hardy 2004). Hence it is more powerful to detect genetic structure. However, it suffers downward bias as soon as one allele in the data set occurs at a low frequency (e.g. <5%). The estimator described in Loiselle *et al.* (1995) weights allele contribution by $p_{la}(1-p_{lal})$ and does not suffer particular bias in the presence of low frequency alleles.

For <u>dominant</u> markers, *SPAGeDi* proposes one estimator of kinship defined in Hardy (2003). To compute this estimator, the inbreeding coefficient must be given (the estimator is robust to moderate errors made on the assumed inbreeding coefficient).

6.1.2. Relationship coefficient

Definition

Relationship coefficients can be defined as the proportion of genes in one individual with alleles identical to these of a reference individual (in several papers (e.g. Queller and Goodnight 1989), the so called "relatedness" coefficient is what is here called "relationship" coefficient). As for kinship coefficients, relationship coefficients depend on a reference population or on reference allele frequencies that can be specified (except for estimator 1° based on Moran's I statistic). Relationship coefficient is the r in Hamilton's (1964) famous rule for altruistic behaviour: rb > c (b = fitness benefit, c = fitness cost).

The expected value of the relationship coefficient (r_{ij}) between two k-ploid individuals (i and j) with inbreeding coefficient F can be expressed in term of the kinship coefficient (F_{ij}) : $r_{ij} = F_{ij} \, k/(1+(k-1).F)$, reducing to $r_{ij} = 2F_{ij}$ for two non-inbred diploids. However, contrary to the kinship coefficient, the relatedness coefficient is not always symmetric (i.e. r_{ij} and r_{ji} have not necessarily the same expectations), in particular when comparing individuals with different ploidy levels as in haplo-diploid organisms. Presently, SPAGeDi considers only symmetrical relatedness coefficients (for asymmetric coefficients, see the program Relatedness by Goodnight and Queller, at http://www.bioc.rice.edu/~kfg/GSoft.html).

One advantage of the relationship coefficient when investigating the genetic structure due to gene flow and drift, is that, at constant gene flow parameters, it is not influenced by the ploidy level or the selfing rate (Hardy and Vekemans 1999). Hence, it is useful to compare the level of genetic structuring among ploidy levels (Hardy and Vekemans 2001).

Estimators

For codominant markers, SPAGeDi proposes 5 estimators.

1°) A first estimator of the relationship coefficient is computed as the correlation between individual allele frequencies (e.g. for a diploid, frequencies can take the following discrete values: 0, ½, 1):

$$r_{ij} = \sum_{l} \left[\sum_{a} (p_{ila} - p_{la})(p_{ila} - p_{la}) + \sum_{a} Var(p_{ila})/(n_{l} - 1) \right] / \sum_{l} \sum_{a} Var(p_{ila})$$

with $Var(p_{ila})$, the variance of individual allele frequencies. The term $\Sigma_a Var(p_{ila})/(n_l-1)$ is a sampling bias correction.

Averaging this estimator over distance classes give mean values of Moran's *I* statistic computed in the way proposed by Dewey and Heywood (1988) (Hardy and Vekemans 1999), except for the bias correction.

2°) A second estimator is defined in Queller and Goodnight (1989):

$$r_{\rm ij} = \sum_{\rm l} \sum_{\rm a} \sum_{\rm ci} \chi_{\rm lcia} (p_{\rm jla} - p_{\rm la}) / \sum_{\rm l} \sum_{\rm a} \sum_{\rm ci} \chi_{\rm lcia} (p_{\rm ila} - p_{\rm la})$$

SPAGeDi actually computes the average $(r_{ij} + r_{ji})/2$. Note that the estimator currently computed in SPAGeDi does not exclude related individuals to calculate p_{la} (a bias correction suggested by Queller and Goodnight 1989).

3°) Two additional estimators are defined in Lynch and Ritland (1999) and Wang (2002), respectively. These estimators can only be computed for diploids without inbreeding (genotypes in Hardy-Weinberg proportions). See these references for definitions and statistical properties in regard to other estimators. There is a sampling bias correction in Wang (2002) estimator which is not applied when reference allele frequencies are given.

4°) A fifth estimator is derived from Li et al. (1993) with a sample size correction by Wang (2002):

$$r_{ii} = \sum_{l} [\omega_{l}(S_{iil}-S_{0l})/(1-S_{0l})]/\sum_{l} \omega_{l}$$

where S_{ijl} is the average proportion of alleles in i found in j and vice versa at locus l [$S_{ijl} = 1$ (for i = aa, j = aa or i = ab, j = ab), $S_{ijl} = 0.75$ (for i = aa, j = ab), $S_{ijl} = 0.5$ (for i = ab, j = ac), $S_{ijl} = 0$ (for i = ab, j = ac), where a, b, c, d indicate alleles]

$$S_{0l} = 2[n_{l}[(\Sigma_{a}p_{la}^{2}) - 1]/(n_{l} - 1)] - [n_{l}^{2}(\Sigma_{a}p_{la}^{3}) - 3[n_{l}(\Sigma_{a}p_{la}^{2}) - 1] - 1]/[(n_{l} - 1)(n_{l} - 2)]$$

or $S_{0l} = \Sigma_a [2p_{la}^2 - p_{la}^3]$ when reference allele frequencies are given (no sample size correction).

 ω_l is the empirically determined locus weight defined as the inverse of the variance of single locus r_{ij} estimates over all i-j pairs (Van de Casteel et al. 2001).

This estimator often shows a low variance (high precision) compared to other ones.

For <u>dominant</u> markers, *SPAGeDi* proposes one estimator defined in Hardy (2003). To compute this estimator, the individual inbreeding coefficient must be given (the estimator is robust to moderate errors made on the assumed inbreeding coefficient).

6.1.3. Kinship type coefficient based on allele size

For microsatellite loci undergoing stepwise mutations, difference of allele sizes (not just the alleles identity vs non-identity information) contain information on coalesence time (Slatkin 1995). This information is taken into account in the R_{ij} coefficient, computed as an average correlation coefficient between allele sizes for homologous genes from two individuals (Streiff et al. 1998). This coefficient is to kinship coefficient what R-statistics are to F-statistics.

$$R_{ij} = \sum_{l} \left[\left(\sum_{ci} \sum_{cj} (s_{lci} - s_l) (s_{lcj} - s_l) / \sum_{ci} \sum_{cj} 1 \right) + \operatorname{Var}(s_l) / (n_l - 1) \right] / \sum_{l} \operatorname{Var}(s_l)$$

where s_{lci} is the size of the allele at locus l on chromosome c from individual i, s_1 is the mean allele size at locus l in the sample, and $Var(s_1)$ is the variance of allele size in the sample. The term $Var(s_1)/(n_1-1)$ is a sampling bias correction (removed when reference allele frequencies are defined).

6.1.4. Rousset's distance measure

Definition

Rousset (2000) proposed a genetic distance measure between individuals (a) analogous of the $F_{\rm ST}/(1-F_{\rm ST})$ ratio using pairs individuals instead of populations. In terms of probabilities of identity by state of genes (see §6.1.1.), this coefficient can be defined as $a_{ij} = ({\rm Qo-Qij})/(1-{\rm Qo})$, where Qo refers to genes within individuals. The advantage of this distance measure over kinship coefficient is that it is not relative to a "reference" population (the distance is calibrated on the distance between genes within individuals). However, this measure is undefined for haploid organisms, and it is much dependent on the selfing rate. It also suffer higher sampling variance than kinship coefficients (Vekemans & Hardy 2004). Rousset (2000) showed that the slope of the regression of this estimator with the distance can be used to provide an estimate of gene dispersal distances (see §6.3.).

Estimator

The estimator is the one called \hat{a} in Rousset (2000).

6.1.5. Fraternity coefficient

Definition

The fraternity coefficient, Δ_{ij} , defined for two diploids (i and j), is a function of the probability that the two genes of i are identical by descent to each of the genes of j (Lynch and Walsh 1998, Lynch and Ritland 1999), hence it depends on the states of all four genes. It can be expressed as a function of the kinship coefficients between the parents of i and j: $\Delta_{ij} = F_{mi,mj} \cdot F_{fi,fj} + F_{mi,fj} \cdot F_{fi,mj}$, where the subscripts mi, mj and fi, fj refer to the mother and father of i and j, respectively (Lynch and Walsh 1998). Hence, a positive Δ_{ij} coefficient means there is a double genetic link between i and j.

The use of both 2-genes and 4-genes coefficients can help assessing the type of parentage relationship linking two individuals; for example, in a random mating population, $\Delta_{ij} = 0$, 0.25, 0, and $F_{ij} = 0$, 0.25, 0, for i and j being parent-offspring, full sibs, or half sibs, respectively (when parents constitute the "reference population").

Estimators

Two estimators available are described in Lynch and Ritland (1999) and Wang (2002). These estimators can only be computed for diploids without inbreeding (genotypes in Hardy-Weinberg proportions). Practically, they perform well only with highly polymorphic loci (with at least 4 or 5 alleles). There is a sampling bias correction in Wang (2002) estimator which is not applied when reference allele frequencies are given.

6.2. STATISTICS FOR POPULATION LEVEL ANALYSES

For analyses at population level, **global** and **pairwise statistics** are computed. Global statistics are based on allele identity (F-statistics and Gst), microsatellite allele size (R-statistics), or the phylogenic distances between alleles (Nst). Similarly, pairwise statistics are based on allele identity (F_{ST} , Rho, G_{ST} , G_{ij} or Ds), microsatellite allele size (R_{ST} or $\delta \mu^2$), or the phylogenic distances between alleles (N_{ST}). Pairwise statistics are first computed for each pair of populations. Then they are regressed on pairwise spatial distances (regression analyses), and they are averaged over all pairs belonging to each predefined distance interval.

6.2.1. F-statistics and G_{ST}

Definition

F-statistics are based on allele identity and are types of kinship coefficients. In terms of probabilities of identity by state, they can be defined as $F_{\text{IT}}=(Q_0-Q_2)/(1-Q_2)$, $F_{\text{IS}}=(Q_0-Q_1)/(1-Q_1)$, and $F_{\text{ST}}=(Q_1-Q_2)/(1-Q_2)$, where Q_0 , Q_1 , Q_2 , refer to probabilities of identity of homologous genes within individuals, among individuals within population, and among individuals among populations, respectively. For global F-statistics, Q_2 refers to all populations, whereas for pairwise F_{ST} , Q_2 refers only to the two populations being compared. Equivalently, these statistics can be defined as intra-class correlation coefficients of allelic states for genes within individuals relative to all populations (F_{IT}), genes within individuals relative to a population (F_{IS}), and genes within populations relative to all populations (F_{ST}).

Estimators

For F-statistics, the estimation procedure is based on a nested ANOVA following Weir and Cockerham (1984), where populations are weighted according to their sample size.

 $G_{\rm ST}$ is an alternative estimator of $F_{\rm ST}$, based on a decomposition of diversity indices following Pons and Petit (1996), where populations have equal weight, irrespective of the sample size.

Note that both F_{ST} and G_{ST} assume "random population effects" in statistical terms, contrary to Nei's G_{ST} (not available in SPAGeDi) which assumes "fixed population effects".

6.2.2. Rho statistic

Definition

The *Rho* statistic is defined by Ronfort et al. (1998). It is to F_{ST} what the relationship coefficient is to the kinship coefficient (§6.1.2.), as it can be interpreted as an average relationship coefficient between individuals within population. *Rho* is equivalent to *Relat* in FSTAT software (Goudet 1995). This is a convenient statistic to compare the level of genetic structuring among ploidy levels (Ronfort et al. 1998). It relates to F-statistics in the following way: for a k-ploid, $Rho = k.F_{ST} / (1+-1) F_{IT}$, reducing to $Rho = 2 F_{ST} / (1+F_{IT})$ for a diploid.

Estimator

The estimator is computed as an intra-class correlation coefficient of individual allele frequencies, using an ANOVA framework (Ronfort et al. 1998) equivalent to that of Weir and Cockerham (1984) for F-statistics.

6.2.3. Gij statistic

Definition

Gij is an average kinship coefficient between the individuals of two populations (i, j), relative to a sample of populations (it expresses thus the genetic similarity rather than the distance between populations). It is equivalent

the mean correlation coefficient between the allele frequencies of the two populations multiplied by the global $F_{\rm ST}$ among populations (Barbujani 1987).

Estimator

$$Gij = 1 - h_{ii}/h_T$$

with $h_{ij} = \sum_{l} \left[\sum_{a \neq a'} (p_{ila} \, p_{jla'}) \right] / L$, where p_{ila} is the frequency of allele a at locus l in population i, and L is the number of loci, and h_T is the average h_{ij} over all population pairs $(i \neq j)$.

6.2.4. Ds – Nei's standard genetic distance

Definition

Nei's (1972) standard genetic distance is a measure of genetic differentiation between two populations often used in phylogenetic reconstruction. Under an infinite allele model (IAM), its expected value is approximately $Ds = 2\mu t$, where μ is the mutation rate, and t is the number of generations since population divergence.

Estimator

Nei's unbiased estimate is computed according to Nei (1978).

6.2.5. R-statistics

Definition

R-statistics are equivalent to F-statistics but based on allele sizes rather than allele identity (Slatkin 1995, Rousset 1996). They can be defined as intra class correlation coefficients of allelic sizes for genes within individuals relative to all populations ($R_{\rm IT}$), genes within individuals relative to a population ($R_{\rm IS}$), and genes within populations relative to all populations ($R_{\rm ST}$). They were developed for loci undergoing a stepwise mutation process. Under a random mutation process (IAM, KAM), expectations for R-statistics are equivalent to corresponding F-statistics, but they suffer higher sampling variances (Balloux and Goudet 2002). To test for the impact of stepwise mutations on genetic structuring, R-statistics can be compared to the corresponding F-statistics estimated following Weir and Cockerham (1984) (see §4.3.5. and §5.6. for permutation tests).

Estimator

R-statistics are estimated using a nested ANOVA (Michalakis and Excoffier 1996).

6.2.6. <u>dm2 – Goldstein's genetic distance</u>

Definition

Goldstein et al. (1995) defined a distance $\delta \mu^2$ comparable to Nei's (1972) standard genetic distance but adapted for loci undergoing stepwise mutations (microsatellites). Under the stepwise mutation model, its expected value is approximately $dm2 = 2\mu t$, where μ is the mutation rate, and t is the number of generations since population divergence.

Estimator

The unbiased $\delta \mu^2$ estimator is defined in Goldstein and Pollok (1997).

6.2.7. <u>N</u>ST

Definition

 $N_{\rm ST}$ is an equivalent to $F_{\rm ST}$ (or $G_{\rm ST}$) but accounting for the phylogenetic distances between alleles ("ordered alleles"). To test for the impact of the allele phylogeny on genetic structuring, $N_{\rm ST}$ should be compared with $G_{\rm ST}$ (see §4.3.5. and §5.6. for permutation tests).

Estimators

The $N_{\rm ST}$ estimator is described in Pons and Petit (1996).

6.3. INFERENCE OF GENE DISPERSAL DISTANCES

Theoretical models of isolation by distance show that, if some conditions are met, the kinship and relationship coefficients between individuals and the pairwise $F_{\rm ST}$, Rho and $R_{\rm ST}$ coefficients between populations are expected to vary approximately linearly (at least within some distance range) with the logarithm of the distance in a two-dimensional space, and with the linear distance in a one-dimensional space (Rousset 1997, 2000; Hardy and Vekemans 1999; Hardy 2003; Vekemans & Hardy 2004; for an application see e.g. Fenster et al. 2003). The slope of the corresponding regressions can be used to estimate gene dispersal distances in terms of a product between population density and mean squared distance of gene movements:

In a **two-dimensional space**, defining $Nb = 4\text{Pi}.Ds^2$, where D is the "effective" population density (i.e. taking into account the variance of reproductive success among individuals), s^2 is $\frac{1}{2}$ the mean squared distance of gene dispersal, and Pi = 3.1415, Nb can be inferred in the following way for **diploids** using

(1) pairwise $F_{ST}/(1-F_{ST})$: $Nb \approx 1/blog$ (2) Rousset's a coefficient: $Nb \approx 1/blog$ (3) kinship coefficient: $Nb \approx -(1-F)/blog$

where blog is the regression slope based on the logarithm of spatial distance, and F is the inbreeding coefficient. (2) and (3) are correct in the absence of selfing. With selfing, (2) is biased, but good estimates can be obtained from (3) if F is replaced by the kinship coefficient between adjacent individuals.

These relationships hold best within a distance range which is approximately s to 20s. At shorter distances, the details of the gene dispersal distribution (not just s^2) matter (Rousset 2001; Heuertz et al. 2003). At large distances, mutation rate can also matter. SPAGeDi allows to **define a restricted distance range** to compute the regression slope (§ 4.3.3.).

For analyses at the individual level (with diploids) and assuming a two-dimensional population at drift-dispersal equilibrium, SPAGeDi can use an **iterative procedure to determine s and** Nb by regressing pairwise kinship coefficients on ln(distance) over a restricted distance range (§ 4.3.4.). The procedure requires an estimate of the effective population density, D. Starting from a global regression slope, the procedure consists in estimating Nb as $Nb = -(1-F_{(1)})/blog$, where $F_{(1)}$ is the kinship coefficient between individuals for the first distance class (assumed to correspond to adjacent individuals), and s is estimated as $s=\sqrt{(Nb/4D.Pi)}$. Then, restricting the regression (blog) to distances between s and 20s, Nb and s are estimated again. This step is repeated until s converges, with up to 100 iterations (Fenster et al. 2003; Vekemans & Hardy 2004). Successive s estimates are displayed on the screen. Convergence is not ensured, in which case no estimate is provided. The procedure can also cycle periodically around a set of values, in which case the 100^{th} value is given with a minus sign.

In a **one-dimensional space**, relationships (1), (2) and (3) also hold but with the regression slope based on linear distance instead of logarithmic distance, and with Nb defined as $4Ds^2$ where s^2 is the mean squared distance of gene dispersal (Rousset 1997). The iterative procedure to determine s and Nb should not be used for one-dimensional populations.

Note that the conditions necessary for valid inferences might not be met in general within natural populations (Hardy & Vekemans 1999). More discussions can be found in Rousset (1997, 2000, 2001). In any case, when a relatively linear relationship is observed, the slope expresses the degree of genetic structuring and contains most of the information regarding intra-locus structure.

6.4. Estimating the actual variance of pairwise coefficients for marker-based heritability and $Q_{\rm ST}$ estimates

Ritland (1996, 2000) proposed methods to estimate heritability and $Q_{\rm ST}$ using genetic markers. These methods require an estimate of the "actual" variance V (i.e. excluding sampling variance) of pairwise kinship coefficients between individuals (inference of heritability) or pairwise $F_{\rm ST}$ between populations (inference of $Q_{\rm ST}$):

$$V = \sum_{p} (((\sum_{l} W_{l} R_{pl})^{2} - \sum_{l} (W_{l} R_{pl})^{2}) / (1 - \sum_{l} W_{l}^{2})) / N - (\sum_{p} \sum_{l} W_{l} R_{pl} / N)^{2}$$

Where Σ_p stands for the sum over the considered pairs (i.e. within a distance class) of individuals or populations (*N* being the number of pairs), R_{pl} is the value of the pairwise statistic (kinship, relatedness, F_{ST} , ...) for pair p at locus l, and W_l is the locus l specific weight when computing multilocus R_p averages.

For heritability inference, the advantage of the approach is that quantitative characters can be measured in situ, avoiding the problem of having an heritability measure valid only for some experimental conditions. For $Q_{\rm ST}$ inference, the advantage of the Ritland's approach is that there is no need to estimate the heritability of the characters.

The method to obtain the actual variance of pairwise coefficients follows Ritland (2000) and requires at least two loci. A jackknife procedure over loci (at least 3 loci necessary) provides approximate standard errors of the variance estimate. The estimates are given under request (§ 4.3.4.) for pairs of individuals or populations belonging to each distance interval as well as for all pairs (under "average"). Note that a large sample size and a high number of loci and/or very polymorphic loci are required for reliable heritability inference.

6.5. TESTING PHYLOGEOGRAPHIC PATTERNS

A phylogeographic pattern occurs when gene copies sampled at nearby locations (e.g. within the same population) carry alleles that are more related on average than for gene copies sampled further apart. Under neutrality, such pattern is expected when the mutation rate is non negligible compared to the migration rate (Hardy et al. 2003).

Phylogeographic patterns can be tested only for 'ordered' alleles, such as microsatellites where differences in allele sizes informs on genetic distance (if stepwise mutations occur), or sequence data (or other multiple site polymorphisms at non recombinant DNA) where genetic distances between alleles can be attributed, for example as the number of mutations differentiating two alleles. SPAGeDi proposes several statistics that account for 'ordered' alleles, such as R_{ST} and $(\delta \mu)^2$ for microsatellites, and N_{ST} for alleles for which a matrix of genetic distances is provided (cf. § 3.8.). A phylogeographic structure appears when R_{ST} or N_{ST} is significantly larger than F_{ST} .

Testing for a phylogeographic pattern can be done using $R_{\rm ST}$ by permuting allele sizes among alleles (Hardy et al. 2003), or using $N_{\rm ST}$ by permuting genetic distances among alleles (Burban et al. 1999; cf. § 4.3.5., § 5.6.). The expected value after such permutation is equal to an $F_{\rm ST}$ because only the structure due to allele identity remains (note that considering the statistical properties of the different estimators computed by SPAGeDi, the $R_{\rm ST}$ statistic should be compared with the $F_{\rm ST}$ statistic, whereas the $N_{\rm ST}$ statistic should be compared with the $G_{\rm ST}$ statistic). The permutation procedures permit to assess the distribution of $R_{\rm ST}$ or $N_{\rm ST}$ under the null hypothesis that there is no phylogeographic pattern. Therefore the unilateral test corresponding to the alternative hypothesis that the observed $R_{\rm ST}$ or $N_{\rm ST}$ is superior to the corresponding value after permutation should be considered.

Testing the global R_{ST} or N_{ST} (or the average pairwise values) tell us whether there is a **phylogeographic signal** within **populations**, answering the question: Are distinct alleles more related within populations than among populations?

Testing the slope (b-lin or b-log values) of pairwise $R_{\rm ST}$ or $N_{\rm ST}$ tell us whether there is a **phylogeographic signal among populations**, answering the question: Are distinct alleles more related between nearby populations than between distant populations?

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8. BUG REPORTS

Ver. 1.1

- Two computational bugs occurred in version 1.1 and earlier and were corrected in version 1.1b:
 - 1°) Bug leading to erroneous jackknife estimates (mean and standard error) for Wang(2002) relationship estimator when missing data occur.
 - 2°) Bug leading occasionally to erroneous estimates for the Rousset's a distance between individuals for the last locus, and consequently for the multilocus and jackknife estimators.

Ver. 1.2

- A bug causing occasional crash when reading the file with reference allele frequencies has been corrected in version 1.2b (released on 11 Nov 2005).
- When selecting both options "Test of mutation effect on genetic differentiation for each population pair" and "Report only P-values", a bug caused inconsistent results regarding these tests in the output file. It has been corrected in version 1.2c (released on 23 Dec 2005).
- WARNING: For analyses at the individual level, when selecting the "allele size correlation coefficient" (Streiff et al. 1998), a problematic bug caused erroneous multilocus and jackknife estimates for this statistic (single-locus estimates were correctly computed). The bug has been corrected in version 1.2d (released on 17 Jan 2006) but all previous versions are affected!
- WARNING: For analyses at the individual level using dominant markers (kinship or relationship coefficient), a bug could cause the program to crash (it may also have lead to inconsistent results). The problem was resolved in version 1.2e.
- Two bugs causing occasional crash when reading the file with reference allele frequencies and when writing down results have been corrected in version 1.2f (released on 11 Apr 2007).
- Bugs occurring when importing large data sets in Genepop format have been corrected in version 1.2g (released on 24 Apr 2007).

FINAL NOTE

The program is regularly modified for further improvements. Any suggestion for improvement is welcome. Also if you have trouble with some data sets you can send us the data set by e mail and we could try to fix the problem.

Good luck!