# An introduction to the PopGenome package

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# 1 Introduction

PopGenome is a new package for population genomic analyses and method development. PopGenome includes, e.g., a wide range of polymorphism, neutrality statistics, and FST estimates; these can be applied to sequence data stored in alignment format, as well as to whole genome SNP data, e.g., from the 1000/1001 Genome projects. The full range of methods can be applied to whole alignments, sets of sub-sequences, and sliding windows based on either nucleotide positions or on SNP counts. PopGenome is also able to handle GFF/GTF annotation files and automatically specifies the SNPs located in, e.g., exon or intron regions. Those subsites can be analyzed together (e.g., all introns together) or each region seperately (e.g., one value per intron). The PopGenome framework is linked to Hudson's MS and Ewing's MSMS programs for significance tests using coalescent simulations.

The following sections explain how to use the PopGenome package. Detailed information about the functions and their parameters can be found in the PopGenome manual on CRAN.

# 2 Installing PopGenome

Installing the package via R

> install.packages("PopGenome")

Loading the PopGenome package

> library(PopGenome)

# 3 Reading data (alignments)

Reading three alignments in FASTA-format (4CL1tl.fas, C4Htl.fas and CADtl.fas) stored in the folder "FASTA". (An example FASTA-file can be found in the data subdirectory of the PopGenome package on CRAN. Other alignment formats – such as Phylip, MEGA, MAF – are also accepted.

**Note:** valid nucleotides are A,a,C,c,T,t,U,u,G,g,-(gap),N,n(unknown). Internally those nucleotides are re-coded into numerical values:

- $T, U \rightarrow 1$
- $C \rightarrow 2$
- $G \rightarrow 3$
- $A \rightarrow 4$
- $unknown \rightarrow 5$

```
• -\rightarrow 6
```

### > GENOME.class <- readData("FASTA")</pre>

GENOME.class is an object of class GENOME. When typing GENOME.class, we get some information about the main methods provided by PopGenome and how to access the results. The GENOME class is the input for every function printed below.

**Note:** GENOME.class is just a variable, you can choose an arbitrary variable name instead.

### > GENOME.class

# Modules:

#### Modules

_	_	_	_	_

	Calculation	Description	Get.the.Result
1	readData	Reading data	get.sum.data
2	neutrality.stats	Neutrality tests	get.neutrality
3	linkage.stats	Linkage disequilibrium	get.linkage
4	recomb.stats	Recombination	<pre>get.recomb</pre>
5	$F\_ST.stats$	Fixation index	<pre>get.F_ST,get.diversity</pre>
6	MKT	McDonald-Kreitman test	get.MKT
7	detail.stats	Mixed statistics	get.detail
8	MS	Coalescent simulation	@
9			
10	set.populations	Defines the populations	
11	sliding.window.transform	Sliding window	
12	splitting.data	Splits the data	
13	show.slots	?provided slots?	
14	get.status	Status of calculations	

The class GENOME contains all observed data and statistic values which are presentable in a multi-locus-scale (vector or matrix). Use the function <code>show.slots(GENOME.class)</code> to get an overview, or check out the PopGenome manual on CRAN. To access those values we use the @-operator.

How many sites were analyzed in each alignment?

```
> GENOME.class@n.sites
```

```
4CL1tl.fas C4Htl.fas CADtl.fas
2979 2620 2930
```

> GENOME.class@region.names

```
[1] "4CL1tl.fas" "C4Htl.fas" "CADtl.fas"
```

To get some summary information from the alignments, use the get.sum.data function. This function extracts the values from the class GENOME and puts them into a matrix. We can also look at those values seperately with the @-operator (GENOME.class@n.biallelic.sites).

# > get.sum.data(GENOME.class)

	n.sites	n.biallelic.site	s n.gaps	n.unknowns	n.valid.sites
4CL1tl.fas	2979	17	6 617	0	2362
C4Htl.fas	2620	8	4 1454	. 0	1161
CADtl.fas	2930	19	7 740	0	2189
n.polyallelic.sites trans.transv.ratio					
4CL1tl.fas		0	1.	120482	
C4Htl.fas		5	1.	470588	
CADtl.fas		1	0.	970000	

The Slot region.data contains some detailed (site specific) information that cannot be presented in a multi-locus-scale. region.data is another class, and its slots are also accessable with the @ operator. See also the figure in section PopGenome classes.

# > GENOME.class@region.data

# SLOTS:

	Slots	Description
1	populations	Samples of each population (rows)
2	populations2	Samples of each population (names)
3	outgroup	Samples of outgroup
4	transitions	Biallelic site transitions
5	biallelic.matrix	Biallelic matrix
6	${\tt n.singletons}$	Number of singletons
7	biallelic.sites	Position of biallelic sites
8	reference	SNP reference
9	n.nucleotides	Number of nucleotides per sequence
10	biallelic.compositions	Nucleotides per sequence (biallelic)
11	synonymous	Synonymous biallelic sites
12	$\verb biallelic.substitutions \\$	Biallelic substitutions
13	polyallelic.sites	Sites with >2 nucleotides
14	sites.with.gaps	Sites with gap positions
15	sites.with.unknowns	Sites with unknown positions
16	minor.alleles	Minor alleles
17	codons	Codons of biallelic substitutions
18	${\tt IntronSNPS}$	SNPs in intron region
19	UTRSNPS	SNPs in UTR region
20	${\tt CodingSNPS}$	SNPs in coding region

21 22 ExonSNPS GeneSNPS SNPs in exon region SNPs in gene region

\_\_\_\_\_

These are the Slots (class region.data)

The first 10 biallelic positions ([1:10]) of the first alignment ([[1]]):

> GENOME.class@region.data@biallelic.sites[[1]][1:10]

[1] 12 13 31 44 59 101 121 154 165 202

Which of those biallelic sites are transitions?

- > GENOME.class@region.data@transitions[[1]][1:10]
  - [1] TRUE TRUE TRUE TRUE TRUE FALSE TRUE FALSE FALSE

# 3.1 The slots of the class region.data

### populations

'list' of length n.populations. Contains the row identifiers (biallelic.matrix) of each individual

### populations2

list of length n.populations. Contains the character names of each individual **outgroup** 

contains the row identifiers (biallelic.matrix) of the outgroup individuals

# transitions

a boolean vector of length n.snps. TRUE if the substitution producing the SNP was a transition

### biallelic.matrix

all calculations are based on this matrix. It contains zeros (major alleles) and ones (minor alleles). rows=individuals. columns=SNPs (see get.biallelic.matrix in the manual) If the parameter include.unknown of the readData function is switched to TRUE, the unknown nucleotides are NA in the biallelic matrix.

# n.singletons

vector of length n.individuals. Number of SNPs where the minor allele occurs in exactly one individual.

#### biallelic.sites

positions of the single nucleotide polymorphisms (SNP)

#### n.nucleotides

number of valid nucleotides for each individual.

# biallelic.composition

the nucleotide distribution for each individual

### synonymous

 $\label{lem:non-coding} \mbox{ vector of length} = \mbox{n.snps.} \ \ \mbox{TRUE:synonymous, FALSE:non-synonymous, NA:non-coding region}$ 

#### biallelic.substitutions

The correspondig nucleotides of the SNPs:

first row: minor allele, second row: mayor allele

#### polyallelic.sites

position of poly-allelic sites (>2 nucleotides)

### sites.with.gaps

sites including gaps (those sites are excluded)

### sites.with.unknowns

sites with unknown positions (N,n,?). Those sites are included if the parameter include.unknown ist TRUE

### minor.alleles

The minor allele of the SNP represented as a numerical value

#### codons

a list of length=n.coding.snps. The codon changes are represented as numerical values. For SNP data we provide the function set.synnonsyn because of memory issues. See also get.codons for detailed information about the codon changes, and codontable() to define alternative genetic codes.

### <FEATURE>SNPS

boolean vector of length=n.snps, TRUE, if the SNP lies in a (coding, exon, intron or UTR) region. This slot will be present after reading data with the corresponding GFF-file.

# 4 Reading data with GFF/GTF information

The GFF folder contains GFF-files for each alignment stored in the FASTA folder. The GFF files should have the same names (without any extensions like .fas or .gff) as the corresponding FASTA files (in this example: 4CL1tl, C4Htl and CADtl) to ensure that sequence and annotation are matched correctly.

> GENOME.class <- readData("FASTA",gffpath="GFF")</pre>

Which of the first 10 SNPs ([1:10]) of the second ([[2]]) alignment are part of a synonymous mutation?

- > GENOME.class@region.data@synonymous[[2]][1:10]
  - [1] TRUE TRUE TRUE TRUE TRUE TRUE NA NA NA NA

NA values indicate that the sites are not in a coding region

- > GENOME.class@region.data@CodingSNPS[[2]][1:10]
  - [1] 1413 1428 1446 1455 1482 1488 1744 1756 1798 1802

## 4.1 Splitting the data into subsites

PopGenome can subdivide the data based on features defined in the GFF file. In this example we split the alignment into coding (CDS) regions. The returned value is again an object of class GENOME.

```
> GENOME.class.split <- splitting.data(GENOME.class, subsites="coding")
```

Each region now contains the SNP information of each coding region defined in the GFF files. In case of whole-genome SNP data this mechanism can be very useful. (see manual:readSNP,readVCF and section (Reading data (SNP files)))

```
> GENOME.class.split@n.sites
[1] 1056 413 103 96 785 132 595 92 112 226 438 220
> GENOME.class.split <- neutrality.stats(GENOME.class.split)</pre>
```

Apply the methods in the neutrality module to all synonymous SNPs in the coding regions.

```
> GENOME.class.split <- neutrality.stats(GENOME.class.split, subsites="syn")
> GENOME.class.split@Tajima.D
```

The function get.gff.info provides additional features to extract annotation informations out of a GFF/GTF file.

# 5 Define the populations

Define two poulations as a list.

```
> GENOME.class <- set.populations(GENOME.class,list(
+ c("CON","KAS-1","RUB-1","PER-1","RI-0","MR-0","TUL-0"),
+ c("MH-0","Y0-0","ITA-0","CVI-0","COL-2","LA-0","NC-1")
+ ))</pre>
```

Individual names are returned by the function get.individuals(GENOME.class)

# 6 Define an outgroup

If one ore more outgroup sequences are defined, PopGenome will only consider SNPs where the outgroup is monomorphic; the monomorphic nucleotide is then automatically defined as the major allele (encoded by 0).

```
> GENOME.class <- set.outgroup(GENOME.class,c("Alyr-1","Alyr-2"))</pre>
```

# 7 Statistics

The methods and statistical tests provided by PopGenome are listed in the user manual. The corresponding references are in the references section.

# 7.1 Neutrality statistics

```
> GENOME.class <- neutrality.stats(GENOME.class)
```

Getting the results from the object of class GENOME.

```
> get.neutrality(GENOME.class)
```

```
neutrality stats
pop 1 Numeric,27
pop 2 Numeric,27
```

Let's look at the first population [[1]].

> get.neutrality(GENOME.class)[[1]]

```
Tajima.D n.segregating.sites Rozas.R_2
                                                         Fu.Li.F
                                                                    Fu.Li.D
4CL1tl.fas -1.1791799
                                        16
                                                  NA -0.9247377 -1.1331823
            0.6987394
                                        17
C4Htl.fas
                                                  NA 0.6742517
                                                                  0.4167836
CADtl.fas
            0.5503743
                                        14
                                                  NA 0.4458431 0.1590690
           Fu.F_S Fay.Wu.H Zeng.E Strobeck.S
4CL1tl.fas
               NA
                       NaN
                               NaN
C4Htl.fas
               NA
                                           NA
                       NaN
                               NaN
CADtl.fas
               NA
                               NaN
                                           NA
                       NaN
```

The NA values indicate that the statistics could not be calculated. This can have several reasons.

- the statistic needs an outgroup
- the statistic was not switched on
- there are no SNPs in the entire region

In each module you can switch on/off statistics (to accelerate calculations), and you can define an outgroup. Check out the PopGenome manual on CRAN for details. PopGenome also provides a population specific view of each statistics.

# > GENOME.class@Tajima.D

```
pop 1 pop 2
4CL1tl.fas -1.1791799 -0.0702101
C4Htl.fas 0.6987394 1.1819777
CADtl.fas 0.5503743 0.2682897
```

If we have read in the data together with the corresponding GFF files, PopGenome can also analyse subsites such as exon, coding, utr or intron regions.

```
> GENOME.class <- neutrality.stats(GENOME.class, subsites="coding")
```

> GENOME.class@Tajima.D

```
pop 1 pop 2
4CL1tl.fas -1.023785 0.2626617
C4Htl.fas 1.013372 1.9121846
CADtl.fas 1.981520 1.5191652
```

We can also analyse each subsite-region separately by splitting the data as described in section 2.1.

```
> GENOME.class.split <- splitting.data(GENOME.class, subsites="coding")
```

- > GENOME.class.split <- neutrality.stats(GENOME.class.split)
- > GENOME.class.split@Tajima.D

```
pop 1
                            pop 2
240 - 1295 -0.2749244 -0.3186974
1890 - 2302 -1.0062306 0.7546749
2679 - 2781 -1.0062306
                       0.5590170
2884 - 2979 -1.0062306
3465 - 4249
                    NA
                               NA
4337 - 4468
                   NaN
                              NaN
4696 - 5290 -1.6097384 2.1259529
6181 - 6272
                   NaN
                              NaN
6412 - 6523
                   NaN
                              NaN
7320 - 7545 0.2390231
                       1.8112198
7643 - 8080 -0.3018700
                       1.1684289
8176 - 8395
                   NaN
                              NaN
```

The splitting.data function transforms the class into another object of class GENOME. Thus, we can apply all methods easily to the transformed class GENOME.class.split. Let's, for example, analyse all non-synonymous SNPs in the coding regions.

```
> GENOME.class.split <- neutrality.stats(GENOME.class.split, subsites="nonsyn")
```

The PopGenome framework provides several modules to calculate statistics. All methods will work in the same way as the neutrality.stats() function described above. The input is always an object of class GENOME.

# 7.2 The slot region.stats

The slot region.stats includes some site-specific statistics or values that cannot be shown in a multi-locus-scale. See also the section PopGenome classes.

```
> GENOME.class@region.stats
SLOTS:
____
                   Slots
                                             Description Module
                                    Nucleotide diversity
                                                              FST
1
   nucleotide.diversity
2
     haplotype.diversity
                                     Haplotype diversity
                                                              FST
3
        haplotype.counts
                                  Haplotype distribution
                                                              FST
     minor.allele.freqs
                                Minor allele frequencies Detail
4
                                  Linkage disequilibrium Linkage
5 linkage.disequilibrium
     biallelic.structure Shared and fixed polymorphisms
These are the Slots (class region.data)
> GENOME.class <- F_ST.stats(GENOME.class)</pre>
> GENOME.class <- diversity.stats(GENOME.class)</pre>
> GENOME.class@region.stats@nucleotide.diversity
[[1]]
         pop 1
                  pop 2
pop 1 5.142857
                     NA
pop 2 6.163265 5.238095
[[2]]
         pop 1 pop 2
pop 1 7.809524
                  NA
pop 2 8.816327
[[3]]
         pop 1
                  pop 2
pop 1 6.285714
pop 2 5.836735 4.285714
```

### nucleotide.diversity

The nucleotide diversity (average pairwise nucleotide differences) within and between the populations. Have to be divided by the slot GENOME.class@n.sites to obtain diversity

```
per site (see also diversity.stats).
```

### haplotype.diversity

The haplotype diversity (average pairwise haplotype differences) within and between the populations. (see also: diversity.stats)

# haplotype.counts

A vector of length=n.indivuals. Number of times the sequence of a specific indivual appears in the whole population

# minor.allele.freqs

The minor allele (0) frequencies for each SNP calculated with the function detail.stats.

# linkage.disequilibrium

The function linkage.stats(...,detail=TRUE) calculates some linkage disequilibrium measurenments for each pair of SNPs  $(r^2, D'...)$ . See also: R2.stats

### biallelic.structure

```
Can be calculated with the function
```

```
detail.stats(GENOME.class, biallelic.structure=TRUE).
```

To extract the results use the function

```
get.detail(GENOME.class,biallelic.structure=TRUE)
```

The returned values (for each SNP) are described in the user manual.

# 8 Sliding Window Analyses

The function sliding.window.transform() transforms an object of class GENOME into another object of class GENOME, where now regions correspond to individual windows. This mechanism enables the user to apply all methods that exist in the PopGenome framework.

PopGenome tries to concatenate the data if the parameter whole.data is set to TRUE. This mechanism enables the user to work with very large datasets, which can be split into smaller chunks that are stored in the input folder. PopGenome is able to concatenate these chunks for analysis. Functions like readVCF and readSNP will do this automatically (see also concatenate.regions) If whole.data=FALSE, the regions are scanned seperately.

```
type=1: Define windows based on SNP countstype=2: Define windows based on nucleotide counts
```

# 8.1 Scanning the whole data

```
> GENOME.class.slide <- linkage.stats(GENOME.class.slide)
> get.linkage(GENOME.class.slide)[[1]]
               Wall.B
                         Wall.Q
                                  Rozas.ZA
                                               Rozas.ZZ Kelly.Z_nS
1 - 50 :
            0.6666667 0.7500000 0.66666667 0.29166667 0.375000000
                            NaN 0.00000000 0.00000000 0.000000000
51 - 100 :
                  NaN
101 - 150 : 0.0000000 0.0000000 0.01851852 -0.05266204 0.071180556
151 - 200 : 0.6250000 0.6666667 0.37847222 0.10206619 0.276406036
201 - 250 : 0.5833333  0.6923077  5.40972222  1.05354208  4.356180145
251 - 300 : 0.0000000 0.0000000 0.01388889 -0.17860000 0.192488889
301 - 350 : 0.0000000 0.0000000 0.01388889 0.00462963 0.009259259
351 - 400 : 0.4000000 0.5000000 3.95688889 2.19704321 1.759845679
401 - 450 : 0.5000000 0.6000000 1.81250000 1.31916667 0.493333333
  The slot GENOME.class.slide@region.names can be used to generate the positions
on the x-axis for, e.g., a plot along the chromosome. See also the function PopGplot.
> xaxis <- strsplit(GENOME.class.slide@region.names,split=" ; ")
> xaxis <- sapply(GENOME.class.slide@region.names,function(x){
   return(mean(as.numeric(x)))
   })
> plot(xaxis,GENOME.class.slide@Wall.B)
8.2 Scanning the regions separately
> GENOME.class.slide <- sliding.window.transform(GENOME.class,width=50,</pre>
                         jump=50,type=1,whole.data=FALSE)
> GENOME.class.slide@region.names
[1] "1:4CL1tl.fas" "2:4CL1tl.fas" "3:4CL1tl.fas" "4:C4Htl.fas" "5:CADtl.fas"
[6] "6:CADtl.fas" "7:CADtl.fas"
> GENOME.class.slide <- linkage.stats(GENOME.class.slide)
> get.linkage(GENOME.class.slide)[[1]]
                Wall.B Wall.Q
                                Rozas.ZA
                                            Rozas.ZZ Kelly.Z_nS
1:4CL1tl.fas 0.6666667
                         0.75 0.66666667 0.29166667 0.37500000
                          NaN 0.00000000 0.00000000 0.00000000
2:4CL1tl.fas
3:4CL1tl.fas 0.0000000
                         0.00 0.01851852 -0.05266204 0.07118056
                         0.80 0.54086420 -0.09315802 0.63402222
4:C4Htl.fas 0.6666667
5:CADtl.fas 0.0000000
                         0.00 2.09259259 -0.04456019 2.13715278
6:CADtl.fas 0.0000000
                         0.00\ 0.01388889\ -1.37808642\ 1.39197531
```

0.60 0.88888889 -0.27527778 1.16416667

7:CADtl.fas 0.5000000

# 9 Reading data (SNP files)

PopGenome can handle SNP data formats such as VCF (1000 human genomes project), HapMap, and .SNP (1001 Arabidopsis genomes project). VCF files can be read in with the function readData(,format="VCF"). Just as with alignments, the VCF files have to be stored in a folder that is given as an input parameter. To study whole genomes, VCFs can be split into chunks (by position), which should be numbered consecutively and stored in the same folder. PopGenome can concatenate them afterwards internally. Alternatively, use the function readVCF, which can read in a tabix-indexed VCF-file like those published from the 1000 Genome project. readVCF supports fast access of defined subregions of the file and automatically splits the data into chunks in cases when the region of interest is too big to fit into the available computer memory (RAM).

The function readSNP reads data published from the 1001 Genomes project (Arabidopsis), where the *quality-variant.txt* files, which include variant calls from every single individual, have to be stored together in one folder. The readData function can also read HapMap data. (readData(,format="HapMap")) example files can be found in the subdirectory *data* of the PopGenome package.

### 9.1 Example

Reading data from the 1001 Genomes project (Arabidopsis)

```
# reading chromosome 1
> GENOME.class <- readSNP("Arabidopsis", CHR=1)
# scan the data with consecutive windows
# window size: 1000 nucleotides (type=2)
# jump size: 1000 nucleotides (type=2)
> GENOME.class.slide <- sliding.window.transform(GENOME.class,1000,1000,type=2)
# calculate diversity statistics for all individuals
> GENOME.class.slide <- diversity.stats(GENOME.class.slide)
# Get the results ([[1]], because only one pop is defined)
> get.diversity(GENOME.class.slide)[[1]]
# alternative: directly access the nucleotide diversity
> plot(GENOME.class.slide@nuc.diversity.within)
```

readSNP and readVCF also accept a GFF-file as an input. To scan alle exons of chromosome 1 and only calculate the diversity of the nonsynonymous sites, do the following:

```
# read chromosome 1 with the corresponding GFF-file
> GENOME.class <- readSNP("Arabidopsis", CHR=1, gffpath="Ara.gff")
# verify the nonsyn/syn SNPs (we need the reference sequence as a FASTA file!)
> GENOME.class <- set.synnonsyn(GENOME.class, ref.chr="chr1.fas")
# split the data into exon regions
> GENOME.class.exons <- splitting.data(GENOME.class,subsites="exon")
# calculate the nonsynonymous diversities
> GENOME.class.exons <- diversity.stats(GENOME.class.exons, subsites="nonsyn")</pre>
```

We can split the data into genes, exons, introns, UTRs, and coding regions if these features are annotated in the GFF file. See also get.gff.info in the manual.

# 10 Coalescent simulation

PopGenome supports the Coalescent simulation program MS from Richard Hudson, as well as the MSMS simulation tool from Greg Ewing. The observed statistics are compared to the simulated values. You have to specify the  $\theta$  value and the PopGenome module you want to apply to the simulated data. An new object of class cs.stats will be created. The main input is an object of class GENOME.

```
> MS.class <- MS(GENOME.class,thetaID="Tajima",neutrality=TRUE)
> MS.class
SLOTS:
----
       Slots
                                                           Description
1 prob.less
                         Prob. that sim.val <= obs.val P(sim <= obs)</pre>
2 prob.equal
                            Prob. that sim.val = obs.val P(sim = obs)
3 valid.iter
                         number of valid iter. for each test and loci
4
     obs.val
                                              obs.values for each test
                                             number of loci considered
5
      n.loci
6
     n.iter
                                   number of iterations for each loci
7
    average
                  average values of each statistic (across all loci)
                  variance values of each statistic (across all loci)
8
    variance
       locus list of loc.stats objects, (detail stats for each locus)
```

Lets look at the data of the first region

> MS.class@locus[[1]]

```
Length Class Mode
1 loc.stats S4
----
SLOTS:
----
Slots
```

	Slots	Description
1	n.sam	number of samples for each iteration
2	n.iter	number of iteration
3	theta	mutation parameter

4	obs.val	vector with observed values for each test
5	positions	position of each polymorphic site
6	trees	if printtree=1, gene tree in Newick format
7	seeds	random numbers used to generate samples
8	halplotypes	haplotypes in each iteration
9	stats	variety of test stats compiled a matrix
10	<pre>loc.prob.less</pre>	Prob. that simulated val. <= to observed val. $P(Sim <= Obs)$
11	<pre>loc.prob.equal</pre>	Prob. that simulated val = to observed val. P(Sim = Obs)
12	${\tt loc.valid.iter}$	number of valid iteration for each test
13	quantiles	13 quantiles for each test

<sup>-----</sup>

# 10.1 The function readMS

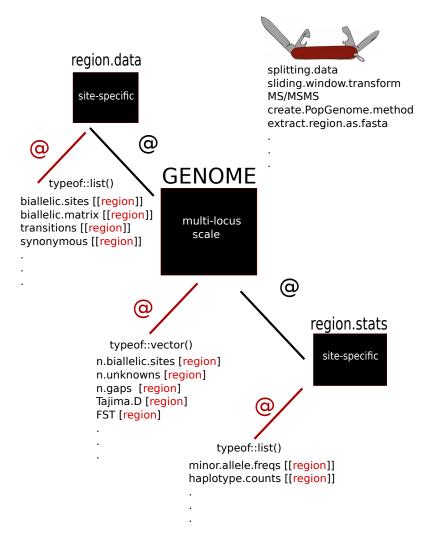
Reading data produced from the coalescent simulation programs MS (Hudson) and MSMS (Ewing).

```
> GENOME.class <- readMS(file="...")</pre>
```

After reading the output file of the coalescent simulations, the full range of methods can be applied to this data

<sup>[1] &</sup>quot;These are the Slots"

# 11 PopGenome classes



# 12 PopGenome internals

# 12.1 How PopGenome does handle missing data.

To include unknown positions (e.g. ./.,N,?) the parameter include.unknown have to be set in the corresponding reading-functions. PopGenome will code those positions into NaN in the biallelic.matrix (get.biallelic.matrix()). In case of nucleotide diversity measurenments and statistics which can be calculated site by site (e.g. neutrality.stats) PopGenome will ignore the missing positions and will apply the algorithms to the valid nucleotides. Lets consider the following bi-allelic vector:

```
bvector \leftarrow c(0,1,NaN,0)
```

To calculate the average nucleotide diversity PopGenome will do the following:

```
ones <- sum(bvector==1, na.rm=TRUE)
zeros <- sum(bvector==0, na.rm=TRUE)
sample.size <- ones + zeros
n.comparisons <- (sample.size*(sample.size-1))/2
nuc.diversity <- (ones * zeros)/n.comparisons</pre>
```

In case of haplotype based methods (e.g haplotype FST) sites including unknown positions are completely deleted.

# 12.2 Synonymous & Non-Synonymous Sites

PopGenome will consider every single nucleotide polymorphism (SNP) seperately and verify if the SNP is part of a synonymous or nonsynonymous change. When there is an unknown or gap position in the corresponding codon (nucleotide-triplet) of a specific individual, PopGenome will ignore those sequences and will try to find a valid codon and will interpret this change. If there is one non-synonymous change, PopGenome will set this SNP as a non-synonymous SNP, even when there are additional synonymous changes. However, the slot <code>GENOME.class@region.data@codons</code> includes all codon changes, and the function <code>get.codons</code> will also give more information. If neccessary, the user can redefine the synonymous/non-synonymous changes by manipulating the <code>GENOME.class@region.data@synonymous</code> slot or define subpositions of interest with the

splitting.data function.

When typing codontable in R, the codon table is printed, where the rows of the second matrix of the list corresponds to the numerical values of the slot GENOME.class@region.data@codons.

```
> codonTable <- codontable()
> codonTable[[2]]
```

The first matrix of this list (codonTable[[1]]) codes the corresponding Proteins of the nucleotide Triplets. PopGenome will always use the first row of this matrix (standard code) to interpret whether a change is synonymous or nonsynonymous. Here you can change the coding in the first row and load your own file in the R-environment

```
# change the file codontable.R
> library(PopGenome)
> source(".../codontable.R")
```

The function have to be codontable()