## PopGenome Session

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## 1 Introduction

PopGenome is a new package for population genomic analysis and method development. PopGenome includes, e.g., a wide range of polymorphism, neutrality statistics and FST estimates, which are applicable to sequence data stored in alignment format, as well as whole genome SNP data from the 1000/1001 Genome projects. The full range of methods can be applied to sliding-windows based on either the whole genome or only the SNPs. 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 at once (e.g., all introns together) or each region seperately (e.g., one value per intron). The PopGenome framework is linked to Hudson's MS program for significance tests using coalescent simulations.

The following sections explain how to use the PopGenome package. Detail informations about the functions and their corresponding parameter can be found in the PopGenome manual on CRAN.

# 2 Install PopGenome

Install 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. Alignment formats like 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 coded into numeric values:

•  $T, U \rightarrow 1$ 

- $\bullet$   $C \rightarrow 2$
- $G \rightarrow 3$
- $\bullet$   $A \rightarrow 4$
- $unknown \rightarrow 5$
- $\bullet$   $\rightarrow 6$
- > GENOME.class <- readData("FASTA")

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

**Note:** GENOME.class ist just a variable, you can choose here whatever you want.

> GENOME.class

#### ----

# Modules:

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

The class <code>GENOME</code> contains all observed data and statistic values which are presentable in a multi-locus-scale. 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 <code>@-operator</code>.

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.biall	elic.sites	n.gaps	n.unknowns	n.valid.sites
4CL1tl.fas	2979	176	617	0	2362
C4Htl.fas	2620	84	1454	0	1161
CADtl.fas	2930	197	740	0	2189
	n.polyallelic.s	ites trans	.transv	.ratio	
4CL1tl.fas		0	1.3	120482	
C4Htl.fas		5	1.4	170588	
CADtl.fas		1	0.9	970000	

The Slot region.data contains some detail (site specific) informations, which are not presentable in a multi-locus-scale. region.data is another class and its slots are 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	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	biallelic.substitutions	Biallelic substitutions			
13	polyallelic.sites	es 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	IntronSNPS	SNPs in intron region			
19	UTRSNPS	SNPs in UTR region			
20	CodingSNPS	SNPs in coding region			
21	ExonSNPS	SNPs in exon region			
22	GeneSNPS	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 (mayor allele) and ones (minor allele). 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 exactly one individual causes a SNP.

#### 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

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

### biallelic.substitutions

The correspondig nucleotides of the SNPs:

first row: minor allele, second row: mayor allele

## polyallelic.sites

position of polyallelic 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.snps. The codon changes are represented as numerical values. This slot is only available for data in alignment format.

For SNP data we provide the function set.synnonsyn because of memory issues. See also get.codons for some detail informations about the codon-changes and codontable() to define your own genetic code.

#### <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 as the corresponding FASTA-files (in this example: 4CL1tl.gff, C4Htl.gff and CADtl.gff) to ensure the right matching.

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

Which of the first 10 SNPs ([1:10]) of the second ([[2]]) alignment are part of an 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 in subsites

PopGenome can scan the data based on the features defined in the GFF file. In this example we are splitting 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 contains now the SNP-informations 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)

Apply 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")
+ ))
```

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 monomorh. The monomorhic nucleotide is than automatically defined as the major allele 0 (non mutant).

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

## 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

Lets look at the first population [[1]].

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

Getting the result from the object of class GENOME.

> get.neutrality(GENOME.class)

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

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

```
Tajima.D n.segregating.sites Rozas.R_2
                                                        Fu.Li.F
                                                                   Fu.Li.D
4CL1tl.fas -1.1791799
                                                  NA -0.9247377 -1.1331823
                                        16
C4Htl.fas
            0.6987394
                                        17
                                                  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
                              NaN
4CL1tl.fas
               NA
                       NaN
                              NaN
C4Htl.fas
               NA
                       NaN
                                          NΑ
CADtl.fas
               NA
                       NaN
                              NaN
                                          NA
```

The NA values indicates 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 and define an outgroup. (check out the PopGenome manual on CRAN). PopGenome also provides a population specific view of each statistic value.

> 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 with the corresponding GFF files PopGenome can also analyse subsites like 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
```

Or 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
                               NaN
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
                   {\tt NaN}
                               NaN
```

The splitting.data function just transforms the class into another object of class GENOME, as a consequence we can apply all methods to the transformed class GENOME.class.split. Lets 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. Every time the input is an object of class GENOME.

## 7.2 The slot region.stats

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

> GENOME.class@region.stats

SLOTS:

Slots Description Module nucleotide.diversity Nucleotide diversity FST 1 2 haplotype.diversity Haplotype diversity FST 3 haplotype.counts Haplotype distribution **FST** minor.allele.freqs Minor allele frequencies Detail 5 linkage.disequilibrium Linkage disequilibrium Linkage

5 biallelic.structure Shared and fixed polymorphisms Detai

These are the Slots (class region.data)

```
> GENOME.class <- F_ST.stats(GENOME.class)
or
> GENOME.class <- diversity.stats(GENOME.class)</pre>
```

<sup>&</sup>gt; 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 4

[[3]]

pop 1 pop 2

pop 1 6.285714 NA

pop 2 5.836735 4.285714
```

### nucleotide.diversity

The nucleotide diversity (average pairwaise nucleotide differences) within and between the populations. Have to be devided by the slot GENOME.class@n.sites (see also diversity.stats)

### haplotype.diversity

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

### haplotype.counts

A vector of length=n.indivuals. Number of times a sequence of a specific indivual-sequence 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 SNP  $(r^2, D'...)$ . 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 Analysis

The sliding.window.transform() transforms an object of class GENOME in another object of class GENOME. This mechanism enables the user to apply all methods existing in the PopGenome framework.

PopGenome tries to concatenate the data if the parameter whole.data is switched to TRUE. This mechanism enables the user to calculate really big data which can be splitted into smaller chunks stored in the input folder. PopGenome is able to concatenate them afterwards. Functions like readVCF and readSNP will do this automatically. (see also concatenate.regions) If whole.data=FALSE the

regions are scanned seperately.

```
type=1: Scanning the SNPs
type=2: Scanning the nucleotides
     Scanning the whole data
> GENOME.class.slide <- sliding.window.transform(GENOME.class,width=50,
                        jump=50, type=1, whole.data=TRUE)
> GENOME.class.slide@region.names
[1] "1 - 50 :"
                 "51 - 100 :" "101 - 150 :" "151 - 200 :" "201 - 250 :"
[6] "251 - 300 :" "301 - 350 :" "351 - 400 :" "401 - 450 :"
> GENOME.class.slide <- linkage.stats(GENOME.class.slide)</pre>
> get.linkage(GENOME.class.slide)[[1]]
              Wall.B
                        Wall.Q
                               Rozas.ZA
                                             Rozas.ZZ Kelly.Z_nS
           0.6666667 0.7500000 0.66666667 0.29166667 0.375000000
1 - 50 :
51 - 100 :
                           NaN 0.00000000 0.00000000 0.000000000
                 {\tt 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
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 x-
values for e.g, a plot along the chromosome.
> xaxis <- strsplit(GENOME.class.slide@region.names,split=" ; ")</pre>
> xaxis <- sapply(GENOME.class.slide@region.names,function(x){
                  return(mean(as.numeric(x)))
                  })
> plot(xaxis,GENOME.class.slide@Wall.B)
     Scanning the regions separately
> GENOME.class.slide <- sliding.window.transform(GENOME.class,width=50,
                        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)</pre>

> get.linkage(GENOME.class.slide)[[1]]

```
Rozas.ZZ Kelly.Z_nS
                Wall.B Wall.Q
                                Rozas, ZA
1:4CL1tl.fas 0.6666667
                         0.75 0.66666667 0.29166667 0.37500000
2:4CL1tl.fas
                   {\tt NaN}
                         NaN 0.00000000 0.00000000 0.00000000
3:4CL1tl.fas 0.0000000
                         0.00 0.01851852 -0.05266204 0.07118056
4:C4Htl.fas 0.6666667
                         0.80 0.54086420 -0.09315802 0.63402222
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 like VCF (1000 Genome project), HapMap and .SNP (1001 Genome project). VCF files can be read in with the function readData(,format="VCF"). The VCF file (same as with alignments) have to be stored in a folder. To study whole genomes, VCFs could be splitted in fairly sized chunks (by position) which should be numbered consecutively and stored in one folder. PopGenome can concatenate them afterwards in the PopGenome framework. Alternatively use the function readVCF which can read in a tabixed VCF-file like it is published from the 1000 Genome project. read-VCF supports fast access of defined subregions of the file and automatically splits the data in chunks in cases when the region of interest is too big for the available RAM.

The function readSNP reads data published from the 1001 Genome project (Arabidopsis), where the *quality-variant.txt* files, which include variant calls from every single individual, have to be stored in a folder. The readData function can also read HapMap data. (readData(,format="HapMap")) In case of using SNP-data the FAST parameter can be switched to TRUE. readData(, format="VCF", FAST=TRUE)!!! Example files can be found in the subdirectory data of the PopGenome package.

### 9.1 Example

Reading data from the 1001 Genome 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 in 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 the features are present in the GFF file. See also get.gff.info in the manual.

## 10 Coalescent simulation

PopGenome supports the Coalescent simulation program MS from Hudson as well as the MSMS simulation tool from Greg Ewing. The observed statistics are tested against the simulated values. You have to specify the  $\theta$  value and the 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>
                            Prob. that sim.val = obs.val P(sim = obs)
2 prob.equal
3 valid.iter
                         number of valid iter. for each test and loci
    obs.val
                                              obs.values for each test
5
     n.loci
                                             number of loci considered
6
     n.iter
                                   number of iterations for each loci
7
     average
                  average values of each statistic (across all loci)
8
                  variance values of each statistic (across all loci)
   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
```

n.sam

Description 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. $P(Sim \le Obs)$
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

\_\_\_\_\_

## 10.1 The function readMS

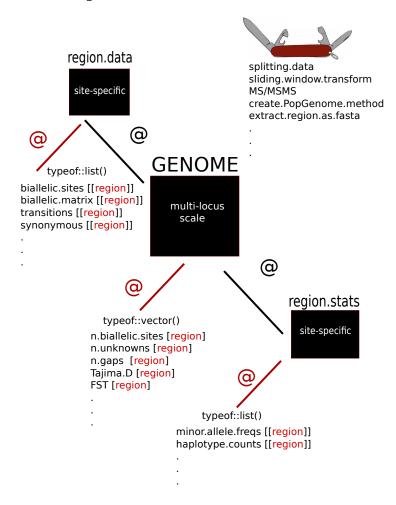
Reading data produced from the coalescent simulation programs MS (Hudson) and MSMS (Ewing).

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

After reading in 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 Synonymous & Non-Synonymous Sites

PopGenome will consider every single nucleotide polymorphism (SNP) seperately an verify if the SNP is part of a synonymous or nonsynonymous change. When there is a unknown or gap position in the entire site 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 nonsynonymous change PopGenome will set this SNP as a nonsynonymous SNP, also when there are additional synonymous changes. However, the slot GENOME.class@region.data@codons includes all codon changes and the function get.codons will also give more inside. If neccessary the user can redefine the syn/nonsyn changes by manipulating the GENOME.class@region.data@synonymous 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()