# Package: qtl2 (via r-universe)

August 26, 2024

Version 0.37-2

Date 2024-08-26

Title Quantitative Trait Locus Mapping in Experimental Crosses

**Description** Provides a set of tools to perform quantitative trait locus (QTL) analysis in experimental crosses. It is a reimplementation of the 'R/qtl' package to better handle high-dimensional data and complex cross designs. Broman et al. (2019) <doi:10.1534/genetics.118.301595>.

Author Karl W Broman [aut, cre] (<https://orcid.org/0000-0002-4914-6671>), R Core Team [ctb]

Maintainer Karl W Broman <broman@wisc.edu>

**Copyright** Code for Brent's method for univariate function optimization was taken from R 3.2.2 (Copyright 1995, 1996 Robert Gentleman and Ross Ihaka, Copyright 2003-2004 The R Foundation, Copyright 1998-2014 The R Core Team).

**Depends** R (>= 3.1.0)

**Imports** Rcpp (>= 1.0.7), yaml (>= 2.1.13), jsonlite (>= 0.9.17), data.table (>= 1.10.4-3), parallel, stats, utils, graphics, grDevices, RSQLite

Suggests testthat, devtools, roxygen2, vdiffr, qtl

License GPL-3

URL https://kbroman.org/qtl2/, https://github.com/rqtl/qtl2

BugReports https://github.com/rqtl/qtl2/issues

LazyData true

Encoding UTF-8

ByteCompile true

LinkingTo Rcpp, RcppEigen

RoxygenNote 7.3.2

**Roxygen** list(markdown=TRUE)

Repository https://rqtl.r-universe.dev

Contents

RemoteUrl https://github.com/rqtl/qtl2RemoteRef HEADRemoteSha 0134eb1590f617f7b30cce69f434fe863c0e8b03

# Contents

add_threshold
basic_summaries
batch_cols
batch_vec
bayes_int
calc_entropy
calc_errorlod
calc_genoprob
calc_geno_freq
calc_grid
calc_het
calc_kinship
calc_raw_founder_maf 19
calc_raw_geno_freq
calc_raw_het
calc_raw_maf
calc_sdp
cbind.calc_genoprob
cbind.scan1
cbind.scan1perm
cbind.sim_geno
cbind.viterbi
cbind_expand
CCcolors
check_cross2
chisq_colpairs
chr_lengths
clean
clean_genoprob
clean_scan1
compare_geno
compare_genoprob
compare_maps
convert2cross2
count_xo
create_gene_query_func
create_snpinfo
create_variant_query_func
decomp_kinship
drop_markers
drop_nullmarkers

2

# Contents

est_herit	45
est_map	47
find_dup_markers	48
find_ibd_segments	49
find_index_snp	51
	52
find_marker	53
	54
- 1	55
	57
	60
	61
	62
	63
	65
0 = =	66
8 = =	66
	67
- 1	69
	70
	70 71
1 = 1	71 72
- 1	72 73
—	73 74
	74 75
I = -C	
	76 77
	77
e	78
$=$ 1 $=$ $\varepsilon$	79
—	80
- 0	81
	82
1 - 1 - c	85
1 -0	85
	87
	89
1 – 1	91
1 – 0	92
1 -1	94
I I 8	96
1 -=	98
I I	99
plot_snpasso	
predict_snpgeno	
print.cross2	
print.summary.scan1perm	
probs_to_grid	
pull_genoprobint	06

pull_genoprobpos
pull_markers
qtl2version
rbind.calc_genoprob
rbind.scan1
rbind.scan1perm
rbind.sim_geno
rbind.viterbi
read_cross2
read_pheno
recode_snps
reduce_map_gaps
reduce_markers
replace_ids
scale_kinship
scan1
scan1blup
scan1coef
scan1max
scan1perm
scan1snps
sdp2char
sim_geno
smooth_gmap
subset.calc_genoprob
subset.cross2
subset.sim_geno
subset.viterbi
subset_scan1
summary.cross2
summary_compare_geno
summary_scan1perm
top_snps
unsmooth_gmap
viterbi
write_control_file
xpos_scan1
zip_datafiles

# Index

add\_threshold

# Description

Draw line segments at significance thresholds for a genome scan plot

#### Usage

add\_threshold(map, thresholdA, thresholdX = NULL, chr = NULL, gap = NULL, ...)

# Arguments

map	Marker map used in the genome scan plot
thresholdA	Autosomal threshold. Numeric or a list. (If a list, the "A" component is taken to be thresholdA and the "X" component is taken to be thresholdX.)
thresholdX	X chromosome threshold (if missing, assumed to be the same as thresholdA)
chr	Chromosomes that were included in the plot
gap	Gap between chromosomes in the plot. Default is 1% of the total genome length.
	Additional arguments passed to graphics::segments()

# Value

None.

basic\_summaries

# Description

Basic summaries of a cross2 object.

# Usage

n\_ind(cross2)

n\_ind\_geno(cross2)

n\_ind\_pheno(cross2)

n\_ind\_covar(cross2)

n\_ind\_gnp(cross2)

ind\_ids(cross2)

ind\_ids\_geno(cross2)

ind\_ids\_pheno(cross2)

ind\_ids\_covar(cross2)

ind\_ids\_gnp(cross2)

n\_chr(cross2)

n\_founders(cross2)

founders(cross2)

chr\_names(cross2)

tot\_mar(cross2)

n\_mar(cross2)

marker\_names(cross2)

n\_pheno(cross2)

pheno\_names(cross2)

#### basic\_summaries

n\_covar(cross2)
covar\_names(cross2)
n\_phenocovar(cross2)
phenocovar\_names(cross2)

# Arguments

cross2 An object of class "cross2", as output by read\_cross2(). For details, see the R/qtl2 developer guide.

# Value

Variously a number, vector of numbers, or vector of character strings.

#### Functions

- n\_ind(): Number of individuals (either genotyped or phenotyped)
- n\_ind\_geno(): Number of genotyped individuals
- n\_ind\_pheno(): Number of phenotyped individuals
- n\_ind\_covar(): Number of individuals with covariate data
- n\_ind\_gnp(): Number of individuals with both genotype and phenotype data
- ind\_ids(): IDs of individuals (either genotyped or phenotyped)
- ind\_ids\_geno(): IDs of genotyped individuals
- ind\_ids\_pheno(): IDs of phenotyped individuals
- ind\_ids\_covar(): IDs of individuals with covariate data
- ind\_ids\_gnp(): IDs of individuals with both genotype and phenotype data
- n\_chr(): Number of chromosomes
- n\_founders(): Number of founder strains
- founders(): Names of founder strains
- chr\_names(): Chromosome names
- tot\_mar(): Total number of markers
- n\_mar(): Number of markers on each chromosome
- marker\_names(): Marker names
- n\_pheno(): Number of phenotypes
- pheno\_names(): Phenotype names
- n\_covar(): Number of covariates
- covar\_names(): Covariate names
- n\_phenocovar(): Number of phenotype covariates
- phenocovar\_names(): Phenotype covariate names

## See Also

```
summary.cross2()
```

batch\_cols

# Batch columns by pattern of missing values

# Description

Identify batches of columns of a matrix that have the same pattern of missing values.

# Usage

batch\_cols(mat, max\_batch = NULL)

# Arguments

mat	A numeric matrix
max_batch	Maximum batch size

### Value

A list containing the batches, each with two components: cols containing numeric indices of the columns in the corresponding batch, and omit containing a vector of row indices that have missing values in this batch.

### See Also

batch\_vec()

batch\_vec

# Description

Split a vector into batches, each no longer than batch\_size and creating at least n\_cores batches, for use in parallel calculations.

# Usage

```
batch_vec(vec, batch_size = NULL, n_cores = 1)
```

# Arguments

vec	A vector to be split into batches
batch_size	Maximum size for each batch
n_cores	Number of compute cores, to be used as a minimum number of batches.

# Value

A list of vectors, each no longer than batch\_size, and with at least n\_cores componenets.

# See Also

### batch\_cols()

# Examples

```
vec_split <- batch_vec(1:304, 50, 8)
vec_split2 <- batch_vec(1:304, 50)</pre>
```

bayes\_int

Calculate Bayes credible intervals

# Description

Calculate Bayes credible intervals for a single LOD curve on a single chromosome, with the ability to identify intervals for multiple LOD peaks.

# Usage

```
bayes_int(
  scan1_output,
  map,
  chr = NULL,
  lodcolumn = 1,
  threshold = 0,
  peakdrop = Inf,
  prob = 0.95,
  expand2markers = TRUE
)
```

# Arguments

scan1_output	An object of class "scan1" as returned by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
chr	Chromosome ID to consider (must be a single value).
lodcolumn	LOD score column to consider (must be a single value).
threshold	Minimum LOD score for a peak.
peakdrop	Amount that the LOD score must drop between peaks, if multiple peaks are to be defined on a chromosome.
prob	Nominal coverage for the interval.
expand2markers	If TRUE, QTL intervals are expanded so that their endpoints are at genetic markers.

# Details

We identify a set of peaks defined as local maxima that exceed the specified threshold, with the requirement that the LOD score must have dropped by at least peakdrop below the lowest of any two adjacent peaks.

At a given peak, if there are ties, with multiple positions jointly achieving the maximum LOD score, we take the average of these positions as the location of the peak.

The default is to use threshold=0, peakdrop=Inf, and prob=0.95. We then return results a single peak, no matter the maximum LOD score, and give a 95% Bayes credible interval.

# Value

A matrix with three columns:

- ci\_lo lower bound of interval
- pos peak position
- ci\_hi upper bound of interval

Each row corresponds to a different peak.

10

### calc\_entropy

# See Also

lod\_int(), find\_peaks(), scan1()

### Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)</pre>
```

```
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
```

```
# 95% Bayes credible interval for QTL on chr 7, first phenotype
bayes_int(out, map, chr=7, lodcolum=1)
```

calc\_entropy Calculate entropy of genotype probability distribution

### Description

For each individual at each genomic position, calculate the entropy of the genotype probability distribution, as a quantitative summary of the amount of missing information.

#### Usage

```
calc_entropy(probs, quiet = TRUE, cores = 1)
```

### Arguments

probs	Genotype probabilities, as calculated from calc_genoprob().
quiet	IF FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().)
	Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

# Details

We calculate  $-sum(p \log_2 p)$ , where we take  $0 \log 0 = 0$ .

### Value

A list of matrices (each matrix is a chromosome and is arranged as individuals x markers).

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
probs <- calc_genoprob(grav2, error_prob=0.002)
e <- calc_entropy(probs)
e <- do.call("cbind", e) # combine chromosomes into one big matrix
# summarize by individual
mean_ind <- rowMeans(e)
# summarize by marker
mean_marker <- colMeans(e)</pre>
```

calc\_errorlod Calculate genotyping error LOD scores

#### Description

Use the genotype probabilities calculated with calc\_genoprob() to calculate genotyping error LOD scores, to help identify potential genotyping errors (and problem markers and/or individuals).

#### Usage

```
calc_errorlod(cross, probs, quiet = TRUE, cores = 1)
```

### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
probs	Genotype probabilities as calculated from calc_genoprob().
quiet	If FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

#### Details

Let  $O_k$  denote the observed marker genotype at position k, and  $g_k$  denote the corresponding true underlying genotype.

Following Lincoln and Lander (1992), we calculate LOD =  $log_{10}[Pr(O_k|g_k = O_k)/Pr(O_k|g_k \neq O_K)]$ 

#### Value

A list of matrices of genotyping error LOD scores. Each matrix corresponds to a chromosome and is arranged as individuals x markers.

### calc\_genoprob

### References

Lincoln SE, Lander ES (1992) Systematic detection of errors in genetic linkage data. Genomics 14:604–610.

### See Also

calc\_genoprob()

# Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
probs <- calc_genoprob(iron, error_prob=0.002, map_function="c-f")
errorlod <- calc_errorlod(iron, probs)
# combine into one matrix</pre>
```

```
errorlod <- do.call("cbind", errorlod)</pre>
```

calc\_genoprob

#### Calculate conditional genotype probabilities

### Description

Uses a hidden Markov model to calculate the probabilities of the true underlying genotypes given the observed multipoint marker data, with possible allowance for genotyping errors.

### Usage

```
calc_genoprob(
    cross,
    map = NULL,
    error_prob = 0.0001,
    map_function = c("haldane", "kosambi", "c-f", "morgan"),
    lowmem = FALSE,
    quiet = TRUE,
    cores = 1
)
```

### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
map	Genetic map of markers. May include pseudomarker locations (that is, locations that are not within the marker genotype data). If NULL, the genetic map in cross is used.
error_prob	Assumed genotyping error probability
map_function	Character string indicating the map function to use to convert genetic distances to recombination fractions.

lowmem	If FALSE, split individuals into groups with common sex and crossinfo and then precalculate the transition matrices for a chromosome; potentially a lot faster but using more memory.
quiet	If FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

# Details

Let  $O_k$  denote the observed marker genotype at position k, and  $g_k$  denote the corresponding true underlying genotype.

We use the forward-backward equations to calculate  $\alpha_{kv} = \log Pr(O_1, \dots, O_k, g_k = v)$  and  $\beta_{kv} = \log Pr(O_{k+1}, \dots, O_n | g_k = v)$ 

We then obtain  $Pr(g_k|O_1, \ldots, O_n) = \exp(\alpha_{kv} + \beta_{kv})/s$  where  $s = \sum_v \exp(\alpha_{kv} + \beta_{kv})$ 

### Value

An object of class "calc\_genoprob": a list of three-dimensional arrays of probabilities, individuals x genotypes x positions. (Note that the arrangement is different from R/qtl.) Also contains four attributes:

- crosstype The cross type of the input cross.
- is\_x\_chr Logical vector indicating whether chromosomes are to be treated as the X chromosome or not, from input cross.
- alleles Vector of allele codes, from input cross.
- alleleprobs Logical value (FALSE) that indicates whether the probabilities are compressed to allele probabilities, as from genoprob\_to\_alleleprob().

# See Also

```
insert_pseudomarkers()
```

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
gmap_w_pmar <- insert_pseudomarkers(grav2$gmap, step=1)
probs <- calc_genoprob(grav2, gmap_w_pmar, error_prob=0.002)</pre>
```

# Description

Calculate genotype frequencies, by individual or by marker

#### Usage

```
calc_geno_freq(probs, by = c("individual", "marker"), omit_x = TRUE)
```

#### Arguments

probs	List of arrays of genotype probabilities, as calculated by calc_genoprob().
by	Whether to summarize by individual or marker
omit_x	If TRUE, results are just for the autosomes. If FALSE, results are a list of length two, containing the results for the autosomes and those for the X chromosome.

# Value

If omit\_x=TRUE, the result is a matrix of genotype frequencies; columns are genotypes and rows are either individuals or markers.

If necessary (that is, if omit\_x=FALSE, the data include the X chromosome, and the set of genotypes on the X chromosome are different than on the autosomes), the result is a list with two components (for the autosomes and for the X chromosome), each being a matrix of genotype frequencies.

# See Also

```
calc_raw_geno_freq(), calc_het()
```

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
p <- calc_genoprob(iron, err=0.002)
# genotype frequencies by marker
tab_g <- calc_geno_freq(p, "marker")
# allele frequencies by marker
ap <- genoprob_to_alleleprob(p)
tab_a <- calc_geno_freq(ap, "marker")</pre>
```

calc\_grid

# Description

Construct vectors of logical indicators that indicate which positions correspond to locations along a grid

### Usage

calc\_grid(map, step = 0, off\_end = 0, tol = 0.01)

#### Arguments

map	A list of numeric vectors; each vector gives marker positions for a single chro- mosome.
step	Distance between pseudomarkers and markers; if step=0 no pseudomarkers are inserted.
off_end	Distance beyond terminal markers in which to insert pseudomarkers.
tol	Tolerance for determining whether a pseudomarker would duplicate a marker position.

#### Details

The function insert\_pseudomarkers(), with stepwidth="fixed", will insert a grid of pseudomarkers, to a marker map. The present function gives a series of TRUE/FALSE vectors that indicate which positions fall on the grid. This is for use with probs\_to\_grid(), for reducing genotype probabilities, calculated with calc\_genoprob(), to just the positions on the grid. The main value of this is to speed up genome scan computations in the case of very dense markers, by focusing on just a grid of positions rather than on all marker locations.

# Value

A list of logical (TRUE/FALSE) vectors that indicate, for a marker/pseudomarker map created by insert\_pseudomarkers() with step>0 and stepwidth="fixed", which positions correspond to he locations along the fixed grid.

# See Also

```
insert_pseudomarkers(), probs_to_grid(), map_to_grid()
```

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap_w_pmar <- insert_pseudomarkers(iron$gmap, step=1)
grid <- calc_grid(iron$gmap, step=1)</pre>
```

calc\_het

# Description

Calculate heterozygosites, by individual or by marker

# Usage

```
calc_het(probs, by = c("individual", "marker"), omit_x = TRUE, cores = 1)
```

# Arguments

probs	List of arrays of genotype probabilities, as calculated by calc_genoprob().
by	Whether to summarize by individual or marker
omit_x	If TRUE, omit the X chromosome.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

### Details

calc\_het() looks at the genotype names (the 2nd dimension of the dimnames of the input probs), which must be two-letter names, and assumes that when the two letters are different it's a heterozy-gous genotype while if they're the same it's a homozygous genotype

# Value

The result is a vector of estimated heterozygosities

# See Also

calc\_raw\_het(), calc\_geno\_freq()

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
p <- calc_genoprob(iron, err=0.002)</pre>
```

```
# heterozygosities by individual
het_ind <- calc_het(p)</pre>
```

```
# heterozygosities by marker
het_mar <- calc_het(p, "marker")</pre>
```

```
calc_kinship
```

### Description

Calculate genetic similarity among individuals (kinship matrix) from conditional genotype probabilities.

# Usage

```
calc_kinship(
  probs,
  type = c("overall", "loco", "chr"),
  omit_x = FALSE,
  use_allele_probs = TRUE,
  quiet = TRUE,
  cores = 1
)
```

### Arguments

probs	Genotype probabilities, as calculated from calc_genoprob().	
type	Indicates whether to calculate the overall kinship ("overall", using all chro- mosomes), the kinship matrix leaving out one chromosome at a time ("loco"), or the kinship matrix for each chromosome ("chr").	
omit_x	If TRUE, only use the autosomes; ignored when type="chr".	
use_allele_probs		
	If TRUE, assess similarity with allele probabilities (that is, first run genoprob_to_alleleprob()); otherwise use the genotype probabilities.	
quiet	IF FALSE, print progress messages.	
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().	

# Details

If use\_allele\_probs=TRUE (the default), we first convert the genotype probabilities to allele probabilities (using genoprob\_to\_alleleprob()).

We then calculate  $\sum_{kl} (p_{ikl}p_{jkl})$  where k = position, l = allele, and i, j are two individuals.

For crosses with just two possible genotypes (e.g., backcross), we don't convert to allele probabilities but just use the original genotype probabilities.

#### Value

If type="overall" (the default), a matrix of proportion of matching alleles. Otherwise a list with one matrix per chromosome.

## Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
probs <- calc_genoprob(grav2, map, error_prob=0.002)
K <- calc_kinship(probs)
# using only markers/pseudomarkers on the grid
grid <- calc_grid(grav2$gmap, step=1)
probs_sub <- probs_to_grid(probs, grid)
K_grid <- calc_kinship(probs_sub)</pre>
```

calc\_raw\_founder\_maf Calculate founder minor allele frequencies from raw SNP genotypes

# Description

Calculate minor allele frequency from raw SNP genotypes in founders, by founder strain or by marker

# Usage

calc\_raw\_founder\_maf(cross, by = c("individual", "marker"))

#### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
by	Indicates whether to summarize by founder strain ("individual") or by marker.

# Value

A vector of minor allele frequencies, one for each founder strain or marker.

# See Also

recode\_snps(), calc\_raw\_maf()

### Examples

## End(Not run)

calc\_raw\_geno\_freq Calculate genotype frequencies from raw SNP genotypes

### Description

Calculate genotype frequencies from raw SNP genotypes, by individual or by marker

# Usage

```
calc_raw_geno_freq(cross, by = c("individual", "marker"), cores = 1)
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
by	Indicates whether to summarize by individual or by marker.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

### Value

A matrix of genotypes frequencies with 3 columns (AA, AB, and BB) and with rows being either individuals or markers.

# See Also

calc\_raw\_maf(), calc\_raw\_het(), recode\_snps(), calc\_geno\_freq()

# Examples

## End(Not run)

calc\_raw\_het

### Description

Calculate estimated heterozygosity for each individual from raw SNP genotypes

# Usage

```
calc_raw_het(cross, by = c("individual", "marker"))
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
by	Indicates whether to summarize by founder strain ("individual") or by marker.

# Value

A vector of heterozygosities, one for each individual or marker.

### See Also

recode\_snps(), calc\_raw\_maf(), calc\_raw\_founder\_maf(), calc\_raw\_geno\_freq(), calc\_het()

# Examples

## End(Not run)

calc_raw_maf	Calculate minor a	llele frequency	from raw SNP genotypes
--------------	-------------------	-----------------	------------------------

# Description

Calculate minor allele frequency from raw SNP genotypes, by individual or by marker

# Usage

```
calc_raw_maf(cross, by = c("individual", "marker"))
```

### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
by	Indicates whether to summarize by founder strain ("individual") or by marker.

# Value

A vector of minor allele frequencies, one for each individual or marker.

### See Also

```
recode_snps(), calc_raw_founder_maf(), calc_raw_het(), calc_raw_geno_freq()
```

#### Examples

## End(Not run)

calc\_sdp

Calculate strain distribution pattern from SNP genotypes

### Description

Calculate the strain distribution patterns (SDPs) from the strain genotypes at a set of SNPs.

### Usage

calc\_sdp(geno)

#### Arguments

geno

Matrix of SNP genotypes, markers x strains, coded as 1 (AA) and 3 (BB). Markers with values other than 1 or 3 are omitted, and monomorphic markers, are omitted.

# Value

A vector of strain distribution patterns: integers between 1 and  $2^n - 2$  where n is the number of strains, whose binary representation indicates the strain genotypes.

# See Also

invert\_sdp(), sdp2char()

cbind.calc\_genoprob

#### Examples

cbind.calc\_genoprob Join genotype probabilities for different chromosomes

#### Description

Join multiple genotype probability objects, as produced by calc\_genoprob(), for the same set of individuals but different chromosomes.

### Usage

```
## S3 method for class 'calc_genoprob'
cbind(...)
```

### Arguments

... Genotype probability objects as produced by calc\_genoprob(). Must have the same set of individuals.

# Value

An object of class "calc\_genoprob", like the input; see calc\_genoprob().

### See Also

rbind.calc\_genoprob()

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
probsA <- calc_genoprob(grav2[1:5,1:2], map, error_prob=0.002)
probsB <- calc_genoprob(grav2[1:5,3:4], map, error_prob=0.002)
probs <- cbind(probsA, probsB)</pre>
```

cbind.scan1

### Description

Join multiple scan1() results for different phenotypes; must have the same map.

### Usage

```
## S3 method for class 'scan1'
cbind(...)
```

#### Arguments

. . .

Genome scan objects of class "scan1", as produced by scan1(). Must have the same map.

# Details

If components addcovar(), Xcovar, intcovar, weights do not match between objects, we omit this information.

If hsq present but has differing numbers of rows, we omit this information.

### Value

An object of class '"scan1", like the inputs, but with the lod score columns from the inputs combined as multiple columns in a single object.

# See Also

rbind.scan1(), scan1()

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
probs <- calc_genoprob(grav2, map, error_prob=0.002)
phe1 <- grav2$pheno[,1,drop=FALSE]
phe2 <- grav2$pheno[,2,drop=FALSE]
out1 <- scan1(probs, phe1) # phenotype 1
out2 <- scan1(probs, phe2) # phenotype 2
out <- cbind(out1, out2)</pre>
```

cbind.scan1perm

#### Description

Column-bind multiple scan1perm objects with the same numbers of rows.

#### Usage

```
## S3 method for class 'scan1perm'
cbind(...)
```

# Arguments

```
• • •
```

A set of permutation results from scan1perm() (objects of class "scan1perm". If different numbers of permutation replicates were used, those columns with fewer replicates are padded with missing values NA. However, if any include autosome/X chromosome-specific permutations, they must all be such.

# Details

The aim of this function is to concatenate the results from multiple runs of a permutation test with scan1perm(), generally with different phenotypes and/or methods, to be used in parallel with rbind.scan1perm().

### Value

The combined column-binded input, as an object of class "scan1perm"; see scan1perm().

# See Also

rbind.scan1perm(), scan1perm(), scan1()

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)</pre>
```

```
Xcovar <- get_x_covar(iron)
# permutations with genome scan (just 3 replicates, for illustration)
operm1 <- scan1perm(probs, pheno[,1,drop=FALSE], addcovar=covar, Xcovar=Xcovar, n_perm=3)
operm2 <- scan1perm(probs, pheno[,2,drop=FALSE], addcovar=covar, Xcovar=Xcovar, n_perm=3)
operm <- cbind(operm1, operm2)</pre>
```

cbind.sim\_geno Join genotype imputations for different chromosomes

### Description

Join multiple genotype imputation objects, as produced by sim\_geno(), for the same individuals but different chromosomes.

# Usage

```
## S3 method for class 'sim_geno'
cbind(...)
```

#### Arguments

• • •

Genotype imputation objects as produced by sim\_geno(). Must have the same set of individuals.

#### Value

An object of class "sim\_geno", like the input; see sim\_geno().

### See Also

```
rbind.sim_geno(), sim_geno()
```

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
drawsA <- sim_geno(grav2[1:5,1:2], map, error_prob=0.002, n_draws=4)
drawsB <- sim_geno(grav2[1:5,3:4], map, error_prob=0.002, n_draws=4)
draws <- cbind(drawsA, drawsB)</pre>
```

26

cbind.viterbi

### Description

Join multiple viterbi objects, as produced by viterbi(), for the same set of individuals but different chromosomes.

# Usage

## S3 method for class 'viterbi'
cbind(...)

# Arguments

• • •

Imputed genotype objects as produced by viterbi(). Must have the same set of individuals.

# Value

An object of class "viterbi", like the input; see viterbi().

# See Also

rbind.viterbi(), viterbi()

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
gA <- viterbi(grav2[1:5,1:2], map, error_prob=0.002)
gB <- viterbi(grav2[1:5,3:4], map, error_prob=0.002)
g <- cbind(gA, gB)</pre>
```

cbind_expand	Combine matrice.	s by columns,	expanding and aligning rows
			1

## Description

This is like base::cbind() but using row names to align the rows and expanding with missing values if there are rows in some matrices but not others.

#### Usage

cbind\_expand(...)

### Arguments

. . .

A set of matrices or data frames

# Value

The matrices combined by columns, using row names to align the rows, and expanding with missing values if there are rows in some matrices but not others.

### Examples

```
df1 <- data.frame(x=c(1,2,3,NA,4), y=c(5,8,9,10,11), row.names=c("A", "B", "C", "D", "E"))
df2 <- data.frame(w=c(7,8,0,9,10), z=c(6,NA,NA,9,10), row.names=c("A", "B", "F", "C", "D"))
cbind_expand(df1, df2)
```

CCcolors

Collaborative Cross colors

### Description

A vector of 8 colors for use with the mouse Collaborative Cross and Diversity Outbreds.

#### Details

CCorigcolors are the original eight colors for the Collaborative Cross founder strains. CCaltcolors are a slightly modified version, but still not color-blind friendly. CCcolors are derived from the the Okabe-Ito color blind friendly palette in Wong (2011) Nature Methods doi:10.1038/nmeth.1618.

# Source

https://csbio.unc.edu/CCstatus/index.py?run=AvailableLines.information

```
data(CCcolors)
data(CCaltcolors)
data(CCorigcolors)
```

check\_cross2

# Check a cross2 object

### Description

Check the integrity of the data within a cross2 object.

# Usage

```
check_cross2(cross2)
```

# Arguments

cross2 An object of class "cross2", as output by read\_cross2(). For details, see the R/qtl2 developer guide.

# Details

Checks whether a cross2 object meets the specifications. Problems are issued as warnings.

### Value

If everything is correct, returns TRUE; otherwise FALSE, with attributes that give the problems.

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
check_cross2(grav2)</pre>
```

chisq\_colpairs Chi-square test on all pairs of columns

# Description

Perform a chi-square test for independence for all pairs of columns of a matrix.

# Usage

```
chisq_colpairs(x)
```

#### Arguments

х

A matrix of positive integers. NAs and values <= 0 are treated as missing.

# Value

A matrix of size p x p, where p is the number of columns in the input matrix x, containing the chi-square test statistics for independence, applied to pairs of columns of x. The diagonal of the result will be all NAs.

# Examples

```
z <- matrix(sample(1:2, 500, replace=TRUE), ncol=5)
chisq_colpairs(z)</pre>
```

chr\_lengths

# Calculate chromosome lengths

# Description

Calculate chromosome lengths for a map object

#### Usage

```
chr_lengths(map, collapse_to_AX = FALSE)
```

# Arguments

mapA list of vectors, each specifying locations of the markers.

 $\verb|collapse_to_AX| If TRUE, collapse to the total lengths of the autosomes and X chromosome.$ 

# Details

We take diff(range(v)) for each vector, v.

# Value

A vector of chromosome lengths. If collapse\_to\_AX=TRUE, the result is a vector of length 2 (autosomal and X chromosome lengths).

# See Also

scan1perm()

clean

# Description

Clean an object by removing messy values

# Usage

clean(object, ...)

### Arguments

object	Object to be cleaned
	Other arguments

# Value

Input object with messy values cleaned up

### See Also

clean.scan1(), clean.calc\_genoprob()

clean\_genoprob Clean genotype probabilities

# Description

Clean up genotype probabilities by setting small values to 0 and for a genotype column where the maximum value is rather small, set all values in that column to 0.

# Usage

```
clean_genoprob(
   object,
   value_threshold = 0.000001,
   column_threshold = 0.01,
   ind = NULL,
   cores = 1,
   ...
)
## S3 method for class 'calc_genoprob'
clean(
```

```
object,
value_threshold = 0.000001,
column_threshold = 0.01,
ind = NULL,
cores = 1,
...
```

### Arguments

object	Genotype probabilities as calculated by calc_genoprob().	
value_threshol	d	
	Probabilities below this value will be set to 0.	
column_thresho	ld	
	For genotype columns where the maximum value is below this threshold, all values will be set to 0. This must be less than $1/k$ where k is the number of genotypes.	
ind	Optional vector of individuals (logical, numeric, or character). If provided, only the genotype probabilities for these individuals will be cleaned, though the full set will be returned.	
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().	
	Ignored at this point.	

# Details

In cases where a particular genotype is largely absent, scan1coef() and fit1() can give unstable estimates of the genotype effects. Cleaning up the genotype probabilities by setting small values to 0 helps to ensure that such effects get set to NA.

At each position and for each genotype column, we find the maximum probability across individuals. If that maximum is < column\_threshold, all values in that genotype column at that position are set to 0.

In addition, any genotype probabilities that are < value\_threshold (generally < column\_threshold) are set to 0.

The probabilities are then re-scaled so that the probabilities for each individual at each position sum to 1.

If ind is provided, the function is applied only to the designated subset of individuals. This may be useful when only a subset of individuals have been phenotyped, as you may want to zero out genotype columns where that subset of individuals has only negligible probability values.

# Value

A cleaned version of the input genotype probabilities object, object.

32

# clean\_scan1

### Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# calculate genotype probabilities
probs <- calc_genoprob(iron, error_prob=0.002)
# clean the genotype probabilities
# (doesn't really do anything in this case, because there are no small but non-zero values)
probs_clean <- clean(probs)
# clean only the females' genotype probabilities
probs_cleanf <- clean(probs, ind=names(iron$is_female)[iron$is_female])</pre>
```

clean\_scan1 Clean scan1 output

# Description

Clean scan1 output by replacing negative values with NA and remove rows where all values are NA.

# Usage

```
clean_scan1(object, ...)
## S3 method for class 'scan1'
clean(object, ...)
```

### Arguments

object	Output of scan1().
	Ignored at present

# Value

The input object with negative values replaced with NAs and then rows with all NAs removed.

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
pr <- calc_genoprob(iron)
out <- scan1(pr, iron$pheno)
out <- clean(out)</pre>
```

compare\_geno

# Description

Count the number of matching genotypes between all pairs of individuals, to look for unusually closely related individuals.

# Usage

```
compare_geno(cross, omit_x = FALSE, proportion = TRUE, quiet = TRUE, cores = 1)
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
omit_x	If TRUE, only use autosomal genotypes
proportion	If TRUE (the default), the upper triangle of the result contains the proportions of matching genotypes. If FALSE, the upper triangle contains counts of matching genotypes.
quiet	IF FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

# Value

A square matrix; diagonal is number of observed genotypes for each individual. The values in the lower triangle are the numbers of markers where both of a pair were genotyped. The values in the upper triangle are either proportions or counts of matching genotypes for each pair (depending on whether proportion=TRUE or =FALSE). The object is given class "compare\_geno".

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
cg <- compare_geno(grav2)
summary(cg)</pre>
```

compare\_genoprob

# Description

Compare two sets of genotype probabilities for one individual on a single chromosome.

# Usage

```
compare_genoprob(
  probs1,
  probs2,
  cross,
  ind = 1,
  chr = NULL,
  minprob = 0.95,
  minmarkers = 10,
  minwidth = 0,
  annotate = FALSE
)
```

# Arguments

probs1	Genotype probabilities (as produced by calc_genoprob()) or allele dosages (as produced by genoprob_to_alleleprob()).
probs2	A second set of genotype probabilities, just like probs1.
cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
ind	Individual to plot, either a numeric index or an ID.
chr	Selected chromosome; a single character string.
minprob	Minimum probability for inferring genotypes (passed to maxmarg()).
minmarkers	Minimum number of markers in results.
minwidth	Minimum width in results.
annotate	If TRUE, add some annotations to the geno1 and geno2 columns to indicate, where they differ, which one matches what appears to be the best genotype. (* = matches the best genotype; - = lower match).

# Details

The function does the following:

- Reduce the probabilities to a set of common locations that also appear in cross.
- Use maxmarg() to infer the genotype at every position using each set of probabilities.
- Identify intervals where the two inferred genotypes are constant.
- Within each segment, compare the observed SNP genotypes to the founders' genotypes.

A data frame with each row con

A data frame with each row corresponding to an interval over which probs1 and probs2 each have a fixed inferred genotype. Columns include the two inferred genotypes, the start and end points and width of the interval, and when founder genotypes are in cross, the proportions of SNPs where the individual matches each possible genotypes.

# See Also

plot\_genoprobcomp()

### Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
iron <- iron[1,"2"] # subset to first individual on chr 2
map <- insert_pseudomarkers(iron$gmap, step=1)
# in presence of a genotyping error, how much does error_prob matter?
iron$geno[[1]][1,3] <- 3
pr_e <- calc_genoprob(iron, map, error_prob=0.002)
pr_ne <- calc_genoprob(iron, map, error_prob=1e-15)</pre>
```

compare\_genoprob(pr\_e, pr\_ne, iron, minmarkers=1, minprob=0.5)

compare\_maps

Compare two marker maps

### Description

Compare two marker maps, identifying markers that are only in one of the two maps, or that are in different orders on the two maps.

### Usage

```
compare_maps(map1, map2)
```

### Arguments

map1	A list of numeric vectors; each vector gives marker positions for a single chro-
	mosome.
map2	A second map, in the same format as map1.

36

# Value

## convert2cross2

## Value

A data frame containing

- marker marker name
- chr\_map1 chromosome ID on map1
- pos\_map1 position on map1
- chr\_map2 chromosome ID on map2
- pos\_map2 position on map2

## Examples

```
# load some data
iron <- read_cross2( system.file("extdata", "iron.zip", package="qtl2") )
gmap <- iron$gmap
pmap <- iron$pmap
# omit a marker from each map
gmap[[7]] <- gmap[[7]][-3]
pmap[[8]] <- pmap[[8]][-7]
# swap order of a couple of markers on the physical map
names(pmap[[9]])[3:4] <- names(pmap[[9]])[4:3]
# move a marker to a different chromosome
pmap[[10]] <- c(pmap[[10]], pmap[[1]][2])[c(1,3,2)]
pmap[[1]] <- pmap[[1]][-2]</pre>
```

```
# compare these messed-up maps
compare_maps(gmap, pmap)
```

convert2cross2 Convert R/qtl cross object to new format

#### Description

Convert a cross object from the R/qtl format to the R/qtl2 format

### Usage

```
convert2cross2(cross)
```

## Arguments

cross An object of class "cross"; see qtl::read.cross() for details.

## Value

Object of class "cross2". For details, see the R/qtl2 developer guide.

## See Also

read\_cross2()

## Examples

library(qtl)
data(hyper)
hyper2 <- convert2cross2(hyper)</pre>

count\_xo

Count numbers of crossovers

## Description

Estimate the numbers of crossovers in each individual on each chromosome.

#### Usage

```
count_xo(geno, quiet = TRUE, cores = 1)
```

# Arguments

geno	List of matrices of genotypes (output of maxmarg() or viterbi()) or a list of 3d-arrays of genotypes (output of sim_geno()).
quiet	If FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

## Value

A matrix of crossover counts, individuals x chromosomes, or (if the input was the output of sim\_geno()) a 3d-array of crossover counts, individuals x chromosomes x imputations.

#### See Also

locate\_xo()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
map <- insert_pseudomarkers(iron$gmap, step=1)
pr <- calc_genoprob(iron, map, error_prob=0.002, map_function="c-f")
g <- maxmarg(pr)
n_xo <- count_xo(g)
# imputations
imp <- sim_geno(iron, map, error_prob=0.002, map_function="c-f", n_draws=32)</pre>
```

create\_gene\_query\_func

Create a function to query genes

## Description

Create a function that will connect to a SQLite database of gene information and return a data frame with gene information for a selected region.

## Usage

```
create_gene_query_func(
 dbfile = NULL,
 db = NULL,
 table_name = "genes",
 chr_field = "chr",
 start_field = "start",
 stop_field = "stop",
 name_field = "Name",
 strand_field = "strand",
 filter = NULL
)
```

### Arguments

dbfile	Name of database file
db	Optional database connection (provide one of file and db).
table_name	Name of table in the database
chr_field	Name of chromosome field
start_field	Name of field with start position (in basepairs)
stop_field	Name of field with stop position (in basepairs)
name_field	Name of field with gene name
strand_field	Name of field with strand (+/-)
filter	Additional SQL filter (as a character string).

#### Details

Note that this function assumes that the database has start and stop fields that are in basepairs, but the selection uses positions in Mbp, and the output data frame should have start and stop columns in Mbp.

Also note that a SQLite database of MGI mouse genes is available at figshare: doi:10.6084/m9.figshare.5286019.v7

## Value

Function with three arguments, chr, start, and end, which returns a data frame with the genes overlapping that region, with start and end being in Mbp. The output should contain at least the columns Name, chr, start, and stop, the latter two being positions in Mbp.

#### Examples

```
# create query function by connecting to file
dbfile <- system.file("extdata", "mouse_genes_small.sqlite", package="qtl2")
query_genes <- create_gene_query_func(dbfile, filter="(source=='MGI')")
# query_genes will connect and disconnect each time
genes <- query_genes("2", 97.0, 98.0)
# connect and disconnect separately
library(RSQLite)
db <- dbConnect(SQLite(), dbfile)
query_genes <- create_gene_query_func(db=db, filter="(source=='MGI')")
genes <- query_genes("2", 97.0, 98.0)
dbDisconnect(db)
```

create\_snpinfo Create snp information table for a cross

### Description

Create a table of snp information from a cross, for use with scan1snps().

## Usage

```
create_snpinfo(cross)
```

#### Arguments

cross

Object of class "cross2". For details, see the R/qtl2 developer guide.

#### Value

A data frame of SNP information with the following columns:

- chr Character string or factor with chromosome
- pos Position (in same units as in the "map" attribute in genoprobs.
- snp Character string with SNP identifier (if missing, the rownames are used).
- sdp Strain distribution pattern: an integer, between 1 and  $2^n 2$  where n is the number of strains, whose binary encoding indicates the founder genotypes SNPs with missing founder genotypes are omitted.

## See Also

index\_snps(), scan1snps(), genoprob\_to\_snpprob()

### Examples

```
## Not run:
# load example data and calculate genotype probabilities
file <- paste0("https://raw.githubusercontent.com/rqtl/",</pre>
                "qtl2data/main/DO_Recla/recla.zip")
recla <- read_cross2(file)</pre>
snpinfo <- create_snpinfo(recla)</pre>
# calculate genotype probabilities
pr <- calc_genoprob(recla, error_prob=0.002, map_function="c-f")</pre>
# index the snp information
snpinfo <- index_snps(recla$pmap, snpinfo)</pre>
# sex covariate
sex <- setNames((recla$covar$Sex=="female")*1, rownames(recla$covar))</pre>
# perform a SNP scan
out <- scan1snps(pr, recla$pmap, recla$pheno[,"bw"], addcovar=sex, snpinfo=snpinfo)
# plot the LOD scores
plot(out$lod, snpinfo, altcol="green3")
```

## End(Not run)

create\_variant\_query\_func Create a function to query variants

#### Description

Create a function that will connect to a SQLite database of founder variant information and return a data frame with variants for a selected region.

## Usage

```
create_variant_query_func(
  dbfile = NULL,
  db = NULL,
  table_name = "variants",
   chr_field = "chr",
   pos_field = "pos",
   id_field = "snp_id",
   sdp_field = "sdp",
   filter = NULL
)
```

#### Arguments

dbfile	Name of database file
db	Optional database connection (provide one of file and db).
table_name	Name of table in the database
chr_field	Name of chromosome field
pos_field	Name of position field
id_field	Name of SNP/variant ID field
sdp_field	Name of strain distribution pattern (SDP) field
filter	Additional SQL filter (as a character string)

#### Details

Note that this function assumes that the database has a pos field that is in basepairs, but the selection uses start and end positions in Mbp, and the output data frame should have pos in Mbp.

Also note that a SQLite database of variants in the founder strains of the mouse Collaborative Cross is available at figshare: doi:10.6084/m9.figshare.5280229.v3

#### Value

Function with three arguments, chr, start, and end, which returns a data frame with the variants in that region, with start and end being in Mbp. The output should contain at least the columns chr and pos, the latter being position in Mbp.

#### Examples

```
# create query function by connecting to file
dbfile <- system.file("extdata", "cc_variants_small.sqlite", package="qtl2")
query_variants <- create_variant_query_func(dbfile)
# query_variants will connect and disconnect each time
variants <- query_variants("2", 97.0, 98.0)
# create query function to just grab SNPs
query_snps <- create_variant_query_func(dbfile, filter="type=='snp'")
# query_variants will connect and disconnect each time
```

## decomp\_kinship

```
snps <- query_snps("2", 97.0, 98.0)
# connect and disconnect separately
library(RSQLite)
db <- dbConnect(SQLite(), dbfile)
query_variants <- create_variant_query_func(db=db)
variants <- query_variants("2", 97.0, 98.0)
dbDisconnect(db)</pre>
```

decomp\_kinship Calculate eigen decomposition of kinship matrix

# Description

Calculate the eigen decomposition of a kinship matrix, or of a list of such matrices.

## Usage

```
decomp_kinship(kinship, cores = 1)
```

## Arguments

kinship	A square matrix, or a list of square matrices.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().)
	Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

#### Details

The result contains an attribute "eigen\_decomp".

## Value

The eigen values and the **transposed** eigen vectors, as a list containing a vector values and a matrix vectors.

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
map <- insert_pseudomarkers(iron$gmap, step=1)
probs <- calc_genoprob(iron, map, error_prob=0.002)
K <- calc_kinship(probs)</pre>
```

Ke <- decomp\_kinship(K)</pre>

drop\_markers

## Description

Drop a vector of markers from a cross2 object.

## Usage

```
drop_markers(cross, markers)
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
markers	A vector of marker names.

## Value

The input cross with the specified markers removed.

### See Also

```
pull_markers(), drop_nullmarkers(), reduce_markers(), find_dup_markers()
```

## Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
markers2drop <- c("BH.342C/347L-Col", "GH.94L", "EG.357C/359L-Col", "CD.245L", "ANL2")
grav2_rev <- drop_markers(grav2, markers2drop)</pre>
```

drop\_nullmarkers Drop markers with no genotype data

## Description

Drop markers with no genotype data (or no informative genotypes)

### Usage

```
drop_nullmarkers(cross, quiet = FALSE)
```

### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
quiet	If FALSE, print information about how many markers were dropped.

## est\_herit

# Details

We omit any markers that have completely missing data, or if founder genotypes are present (e.g., for Diversity Outbreds), the founder genotypes are missing or are all the same.

#### Value

The input cross with the uninformative markers removed.

# See Also

drop\_markers(), pull\_markers()

### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
# make a couple of markers missing
grav2$geno[[2]][,c(3,25)] <- 0
grav2_rev <- drop_nullmarkers(grav2)</pre>
```

est\_herit

*Estimate heritability with a linear mixed model* 

#### Description

Estimate the heritability of a set of traits via a linear mixed model, with possible allowance for covariates.

#### Usage

```
est_herit(
   pheno,
   kinship,
   addcovar = NULL,
   weights = NULL,
   reml = TRUE,
   cores = 1,
   ...
)
```

#### Arguments

pheno	A numeric matrix of phenotypes, individuals x phenotypes.
kinship	A kinship matrix.
addcovar	An optional numeric matrix of additive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.

est\_herit

reml	If true, use REML; otherwise, use maximimum likelihood.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().
	Additional control parameters (see details).

## Details

We fit the model  $y = X\beta + \epsilon$  where  $\epsilon$  is multivariate normal with mean 0 and covariance matrix  $\sigma^2[h^2(2K) + I]$  where K is the kinship matrix and I is the identity matrix.

If weights are provided, the covariance matrix becomes  $\sigma^2[h^2(2K) + D]$  where D is a diagonal matrix with the reciprocal of the weights.

For each of the inputs, the row names are used as individual identifiers, to align individuals.

If reml=TRUE, restricted maximum likelihood (reml) is used to estimate the heritability, separately for each phenotype.

Additional control parameters include tol for the tolerance for convergence, quiet for controlling whether messages will be display, max\_batch for the maximum number of phenotypes in a batch, and check\_boundary for whether the 0 and 1 boundary values for the estimated heritability will be checked explicitly.

## Value

A vector of estimated heritabilities, corresponding to the columns in pheno. The result has attributes "sample\_size", "log10lik" and "resid\_sd".

#### Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

# calculate genotype probabilities
probs <- calc\_genoprob(iron, map, error\_prob=0.002)</pre>

```
# kinship matrix
kinship <- calc_kinship(probs)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)</pre>
```

```
# perform genome scan
hsq <- est_herit(pheno, kinship, covar)</pre>
```

est\_map

## Description

Uses a hidden Markov model to re-estimate the genetic map for an experimental cross, with possible allowance for genotyping errors.

## Usage

```
est_map(
    cross,
    error_prob = 0.0001,
    map_function = c("haldane", "kosambi", "c-f", "morgan"),
    lowmem = FALSE,
    maxit = 10000,
    tol = 0.000001,
    quiet = TRUE,
    save_rf = FALSE,
    cores = 1
)
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
error_prob	Assumed genotyping error probability
<pre>map_function</pre>	Character string indicating the map function to use to convert genetic distances to recombination fractions.
lowmem	If FALSE, precalculate initial and emission probabilities, and at each iteration calculate the transition matrices for a chromosome; potentially a lot faster but using more memory. Needs to be tailored somewhat to cross type. For example, multi-way RIL may need to reorder the transition matrix according to cross order, and AIL and DO need separate transition matrices for each generation.
maxit	Maximum number of iterations in EM algorithm.
tol	Tolerance for determining convergence
quiet	If FALSE, print progress messages.
save_rf	If TRUE, save the estimated recombination fractions as an attribute ("rf") of the result.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

## Details

The map is estimated assuming no crossover interference, but a map function (by default, Haldane's) is used to derive the genetic distances.

48

#### Value

A list of numeric vectors, with the estimated marker locations (in cM). The location of the initial marker on each chromosome is kept the same as in the input cross.

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))</pre>
```

```
gmap <- est_map(grav2, error_prob=0.002)</pre>
```

find\_dup\_markers Find markers with identical genotype data

#### Description

Identify sets of markers with identical genotype data.

#### Usage

```
find_dup_markers(cross, chr, exact_only = TRUE, adjacent_only = FALSE)
```

#### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
chr	Optional vector specifying which chromosomes to consider. This may be a log- ical, numeric, or character string vector.
exact_only	If TRUE, look only for markers that have matching genotypes and the same pat- tern of missing data; if FALSE, also look for cases where the observed genotypes at one marker match those at another, and where the first marker has missing genotype whenever the genotype for the second marker is missing.
adjacent_only	If TRUE, look only for sets of markers that are adjacent to each other.

## Details

If exact.only=TRUE, we look only for groups of markers whose pattern of missing data and observed genotypes match exactly. One marker (chosen at random) is selected as the name of the group (in the output of the function).

If exact.only=FALSE, we look also for markers whose observed genotypes are contained in the observed genotypes of another marker. We use a pair of nested loops, working from the markers with the most observed genotypes to the markers with the fewest observed genotypes.

#### Value

A list of marker names; each component is a set of markers whose genotypes match one other marker, and the name of the component is the name of the marker that they match.

#### find\_ibd\_segments

## See Also

```
drop_markers(), drop_nullmarkers(), reduce_markers()
```

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
dup <- find_dup_markers(grav2)
grav2_nodup <- drop_markers(grav2, unlist(dup))</pre>
```

find\_ibd\_segments Find IBD segments for a set of strains

# Description

Find IBD segments (regions with a lot of shared SNP genotypes) for a set of strains

## Usage

find\_ibd\_segments(geno, map, min\_lod = 15, error\_prob = 0.001, cores = 1)

## Arguments

geno	List of matrices of founder genotypes. The matrices correspond to the genotypes on chromosomes and are arrayed as founders x markers.
map	List of vectors of marker positions
min_lod	Threshold for minimum LOD score for a segment
error_prob	Genotyping error/mutation probability
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

### Details

For each strain pair on each chromosome, we consider all marker intervals and calculate a LOD score comparing the two hypotheses: that the strains are IBD in the interval, vs. that they are not. We assume that the two strains are homozygous at all markers, and use the model from Broman and Weber (1999), which assumes linkage equilibrium between markers and uses a simple model for genotype frequencies in the presence of genotyping errors or mutations.

Note that inference of IBD segments is heavily dependent on how SNPs were chosen to be genotyped. (For example, were the SNPs ascertained based on their polymorphism between a particular strain pair?) A data frame whose rows are IBD segments and whose columns are:

- Strain 1
- Strain 2
- Chromosome
- Left marker
- Right marker
- Left position
- Right position
- Left marker index
- Right marker index
- Interval length
- Number of markers
- Number of mismatches
- LOD score

## References

Broman KW, Weber JL (1999) Long homozygous chromosomal segments in reference families from the Centre d'Étude du Polymorphisme Humain. Am J Hum Genet 65:1493–1500.

## Examples

```
## Not run:
# load D0 data from Recla et al. (2014) Mamm Genome 25:211-222.
recla <- read_cross2("https://raw.githubusercontent.com/rqtl/qtl2data/main/D0_Recla/recla.zip")
# grab founder genotypes and physical map
fg <- recla$founder_geno
pmap <- recla$pmap
# find shared segments
(segs <- find_ibd_segments(fg, pmap, min_lod=10, error_prob=0.0001))
## End(Not run)
```

find\_index\_snp Find name of indexed snp

#### Description

For a particular SNP, find the name of the corresponding indexed SNP.

## Usage

find\_index\_snp(snpinfo, snp)

### Arguments

snpinfo	Data frame with SNP information with the following columns:
	<ul> <li>chr - Character string or factor with chromosome</li> </ul>
	• index - Numeric index of equivalent, indexed SNP, as produced by index_snps().
	• snp - Character string with SNP identifier (if missing, the rownames are
	used).
snp	Name of snp to look for (can be a vector).

## Value

A vector of SNP IDs (the corresponding indexed SNPs), with NA if a SNP is not found.

#### See Also

### find\_marker()

### Examples

# update snp info by adding the SNP index column

```
snpinfo <- index_snps(recla$pmap, snpinfo)
# find indexed snp for a particular snp
find_index_snp(snpinfo, "m3")
## End(Not run)</pre>
```

find\_map\_gaps Find gaps in a genetic map

# Description

Find gaps between markers in a genetic map

### Usage

```
find_map_gaps(map, min_gap = 50)
```

## Arguments

map	Genetic map as a list of vectors (each vector is a chromosome and contains the marker positions).
min_gap	Minimum gap length to return.

# Value

Data frame with 6 columns: chromosome, marker to left of gap, numeric index of marker to left, marker to right of gap, numeric index of marker to right, and the length of the gap.

# See Also

```
reduce_map_gaps()
```

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
find_map_gaps(iron$gmap, 40)</pre>
```

find\_marker

### Description

Find markers closest to specified set of positions, or within a specified interval.

#### Usage

find\_marker(map, chr, pos = NULL, interval = NULL)

### Arguments

map	A map object: a list (corresponding to chromosomes) of vectors of marker po- sitions. Can also be a snpinfo object (data frame with columns chr and pos; marker names taken from column snp or if that doesn't exist from the row names)
chr	A vector of chromosomes
pos	A vector of positions
interval	A pair of positions (provide either pos or interval but not both)

#### Details

If pos is provided, interval should not be, and vice versa.

If pos is provided, then chr and pos should either be the same length, or one of them should have length 1 (to be expanded to the length of the other).

If interval is provided, then chr should have length 1.

#### Value

A vector of marker names, either closest to the positions specified by pos, or within the interval defined by interval.

## See Also

find\_markerpos(), find\_index\_snp(), pull\_genoprobpos(), pull\_genoprobint()

# Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# find markers by their genetic map positions
find_marker(iron$gmap, c(8, 11), c(37.7, 56.9))
# find markers by their physical map positions (two markers on chr 7)
find_marker(iron$pmap, 7, c(44.2, 108.9))</pre>
```

```
# find markers in an interval
find_marker(iron$pmap, 16, interval=c(35, 80))
```

find\_markerpos Find positions of markers

## Description

Find positions of markers within a cross object

## Usage

find\_markerpos(cross, markers, na.rm = TRUE)

## Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide. Can also be a map (as a list of vectors of marker positions).
markers	A vector of marker names.
na.rm	If TRUE, don't include not-found markers in the results (but issue a warning if some markers weren't found). If FALSE, include those markers with NA for chr and position.

## Value

A data frame with chromosome and genetic and physical positions (in columns "gmap" and "pmap"), with markers as row names. If the input cross is not a cross2 object but rather a map, the output contains chr and pos.

## See Also

find\_marker()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# find markers
find_markerpos(iron, c("D8Mit294", "D11Mit101"))</pre>
```

find\_peaks

# Description

Find peaks in a set of LOD curves (output from scan1()

## Usage

```
find_peaks(
    scan1_output,
    map,
    threshold = 3,
    peakdrop = Inf,
    drop = NULL,
    prob = NULL,
    thresholdX = NULL,
    peakdropX = NULL,
    dropX = NULL,
    probX = NULL,
    expand2markers = TRUE,
    sort_by = c("column", "pos", "lod"),
    cores = 1
)
```

## Arguments

scan1_output	An object of class "scan1" as returned by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers(). Can also be an indexed SNP info table, as from index_snps() or scan1snps().
threshold	Minimum LOD score for a peak (can be a vector with separate thresholds for each lod score column in scan1_output)
peakdrop	Amount that the LOD score must drop between peaks, if multiple peaks are to be defined on a chromosome. (Can be a vector with separate values for each lod score column in scan1_output.)
drop	If provided, LOD support intervals are included in the results, and this indicates the amount to drop in the support interval. (Can be a vector with separate values for each lod score column in scan1_output.) Must be $\leq$ peakdrop
prob	If provided, Bayes credible intervals are included in the results, and this indi- cates the nominal coverage. (Can be a vector with separate values for each lod score column in scan1_output.) Provide just one of drop and prob.
thresholdX	Separate threshold for the X chromosome; if unspecified, the same threshold is used for both autosomes and the X chromosome. (Like threshold, this can be a vector with separate thresholds for each lod score column.)

peakdropX	Like peakdrop, but for the X chromosome; if unspecified, the same value is used for both autosomes and the X chromosome. (Can be a vector with separate values for each lod score column in scan1_output.)
dropX	Amount to drop for LOD support intervals on the X chromosome. Ignored if drop is not provided. (Can be a vector with separate values for each lod score column in scan1_output.)
probX	Nominal coverage for Bayes intervals on the X chromosome. Ignored if prob is not provided. (Can be a vector with separate values for each lod score column in scan1_output.)
expand2markers	If TRUE (and if drop or prob is provided, so that QTL intervals are calculated), QTL intervals are expanded so that their endpoints are at genetic markers.
sort_by	Indicates whether to sort the rows by lod column, genomic position, or LOD score.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

## Details

For each lod score column on each chromosome, we return a set of peaks defined as local maxima that exceed the specified threshold, with the requirement that the LOD score must have dropped by at least peakdrop below the lowest of any two adjacent peaks.

At a given peak, if there are ties, with multiple positions jointly achieving the maximum LOD score, we take the average of these positions as the location of the peak.

## Value

A data frame with each row being a single peak on a single chromosome for a single LOD score column, and with columns

- lodindex lod column index
- lodcolumn lod column name
- chr chromosome ID
- · pos peak position
- lod lod score at peak

If drop or prob is provided, the results will include two additional columns: ci\_lo and ci\_hi, with the endpoints of the LOD support intervals or Bayes credible wintervals.

# See Also

scan1(), lod\_int(), bayes\_int()

### Examples

fit1

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno</pre>
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
Xcovar <- get_x_covar(iron)</pre>
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
# find just the highest peak on each chromosome
find_peaks(out, map, threshold=3)
# possibly multiple peaks per chromosome
find_peaks(out, map, threshold=3, peakdrop=1)
# possibly multiple peaks, also getting 1-LOD support intervals
find_peaks(out, map, threshold=3, peakdrop=1, drop=1)
# possibly multiple peaks, also getting 90% Bayes intervals
find_peaks(out, map, threshold=3, peakdrop=1, prob=0.9)
```

fit1

Fit single-QTL model at a single position

## Description

Fit a single-QTL model at a single putative QTL position and get detailed results about estimated coefficients and individuals contributions to the LOD score.

## Usage

```
fit1(
  genoprobs,
  pheno,
  kinship = NULL,
  addcovar = NULL,
  nullcovar = NULL,
  intcovar = NULL,
```

```
weights = NULL,
contrasts = NULL,
model = c("normal", "binary"),
zerosum = TRUE,
se = TRUE,
hsq = NULL,
reml = TRUE,
blup = FALSE,
....)
```

# Arguments

genoprobs	A matrix of genotype probabilities, individuals x genotypes. If NULL, we create a single intercept column, matching the individual IDs in pheno.
pheno	A numeric vector of phenotype values (just one phenotype, not a matrix of them)
kinship	Optional kinship matrix.
addcovar	An optional numeric matrix of additive covariates.
nullcovar	An optional numeric matrix of additional additive covariates that are used under the null hypothesis (of no QTL) but not under the alternative (with a QTL). This is needed for the X chromosome, where we might need sex as a additive covariate under the null hypothesis, but we wouldn't want to include it under the alternative as it would be collinear with the QTL effects.
intcovar	An optional numeric matrix of interactive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.
contrasts	An optional numeric matrix of genotype contrasts, size genotypes x genotypes. For an intercross, you might use $cbind(mu=c(1,1,1), a=c(-1, 0, 1), d=c(0, 1, 0))$ to get mean, additive effect, and dominance effect. The default is the identity matrix.
model	Indicates whether to use a normal model (least squares) or binary model (logistic regression) for the phenotype. If model="binary", the phenotypes must have values in $[0, 1]$ .
zerosum	If TRUE, force the genotype or allele coefficients sum to 0 by subtracting their mean and add another column with the mean. Ignored if contrasts is provided.
se	If TRUE, calculate the standard errors.
hsq	(Optional) residual heritability; used only if kinship provided.
reml	If kinship provided: if reml=TRUE, use REML; otherwise maximum likelihood.
blup	If TRUE, fit a model with QTL effects being random, as in scan1blup().
	Additional control parameters; see Details;

# Details

For each of the inputs, the row names are used as individual identifiers, to align individuals.

If kinship is absent, Haley-Knott regression is performed. If kinship is provided, a linear mixed model is used, with a polygenic effect estimated under the null hypothesis of no (major) QTL, and then taken as fixed as known in the genome scan.

If contrasts is provided, the genotype probability matrix, P, is post-multiplied by the contrasts matrix, A, prior to fitting the model. So we use  $P \cdot A$  as the X matrix in the model. One might view the rows of  $A^{-1}$  as the set of contrasts, as the estimated effects are the estimated genotype effects pre-multiplied by  $A^{-1}$ .

The ... argument can contain several additional control parameters; suspended for simplicity (or confusion, depending on your point of view). tol is used as a tolerance value for linear regression by QR decomposition (in determining whether columns are linearly dependent on others and should be omitted); default 1e-12. maxit is the maximum number of iterations for converence of the iterative algorithm used when model=binary. bintol is used as a tolerance for converence for the iterative algorithm used when model=binary. eta\_max is the maximum value for the "linear predictor" in the case model="binary" (a bit of a technicality to avoid fitted values exactly at 0 or 1).

#### Value

A list containing

- coef Vector of estimated coefficients.
- SE Vector of estimated standard errors (included if se=TRUE).
- lod The overall lod score.
- ind\_lod Vector of individual contributions to the LOD score (not provided if kinship is used).
- fitted Fitted values.
- resid Residuals. If blup==TRUE, only coef and SE are included at present.

## References

Haley CS, Knott SA (1992) A simple regression method for mapping quantitative trait loci in line crosses using flanking markers. Heredity 69:315–324.

Kang HM, Zaitlen NA, Wade CM, Kirby A, Heckerman D, Daly MJ, Eskin E (2008) Efficient control of population structure in model organism association mapping. Genetics 178:1709–1723.

#### See Also

pull\_genoprobpos(), find\_marker()

### Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=5)</pre>
```

fit1

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno[,1]</pre>
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
# scan chromosome 7 to find peak
out <- scan1(probs[,"7"], pheno, addcovar=covar)</pre>
# find peak position
max_pos <- max(out, map)</pre>
# genoprobs at max position
pr_max <- pull_genoprobpos(probs, map, max_pos$chr, max_pos$pos)</pre>
# fit QTL model just at that position
out_fit1 <- fit1(pr_max, pheno, addcovar=covar)</pre>
```

fread\_csv

Read a csv file

### Description

Read a csv file via data.table::fread() using a particular set of options, including the ability to transpose the result.

### Usage

```
fread_csv(
  filename,
  sep = ",",
  na.strings = c("NA", "-"),
  comment.char = "#",
  transpose = FALSE,
  rownames_included = TRUE
)
```

#### Arguments

filename	Name of input file
sep	Field separator
na.strings	Missing value codes
comment.char	Comment character; rest of line after this character is ignored
transpose	If TRUE, transpose the result
rownames_included	
	If TRUE, the first column is taken to be row names

If TRUE, the first column is taken to be row names.

## Details

Initial two lines can contain comments with number of rows and columns. Number of columns includes an ID column; number of rows does not include the header row.

The first column is taken to be a set of row names

## Value

Data frame

## See Also

fread\_csv\_numer()

## Examples

## Not run: mydata <- fread\_csv("myfile.csv", transpose=TRUE)</pre>

fread\_csv\_numer Read a csv file that has numeric columns

#### Description

Read a csv file via data.table::fread() using a particular set of options, including the ability to transpose the result. This version assumes that the contents other than the first column and the header row are strictly numeric.

## Usage

```
fread_csv_numer(
   filename,
   sep = ",",
   na.strings = c("NA", "-"),
   comment.char = "#",
   transpose = FALSE,
   rownames_included = TRUE
)
```

### Arguments

filename	Name of input file
sep	Field separator
na.strings	Missing value codes
comment.char	Comment character; rest of line after this character is ignored
transpose	If TRUE, transpose the result
rownames_included	
	If TRUE, the first column is taken to be row names

If TRUE, the first column is taken to be row names.

## Details

Initial two lines can contain comments with number of rows and columns. Number of columns includes an ID column; number of rows does not include the header row.

The first column is taken to be a set of row names

## Value

Data frame

## See Also

fread\_csv()

## Examples

```
## Not run: mydata <- fread_csv_numer("myfile.csv", transpose=TRUE)</pre>
```

genoprob\_to\_alleleprob

Convert genotype probabilities to allele probabilities

#### Description

Reduce genotype probabilities (as calculated by calc\_genoprob()) to allele probabilities.

#### Usage

```
genoprob_to_alleleprob(probs, quiet = TRUE, cores = 1)
```

### Arguments

probs	Genotype probabilities, as calculated from calc_genoprob().
quiet	IF FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().)
	Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

#### Value

An object of class "calc\_genoprob", like the input probs, but with probabilities collapsed to alleles rather than genotypes. See calc\_genoprob().

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap_w_pmar <- insert_pseudomarkers(iron, step=1)
probs <- calc_genoprob(iron, gmap_w_pmar, error_prob=0.002)
allele_probs <- genoprob_to_alleleprob(probs)</pre>
```

genoprob\_to\_snpprob Convert genotype probabilities to SNP probabilities

## Description

For multi-parent populations, convert use founder genotypes at a set of SNPs to convert founderbased genotype probabilities to SNP genotype probabilities.

## Usage

genoprob\_to\_snpprob(genoprobs, snpinfo)

## Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
snpinfo	Data frame with SNP information with the following columns (the last three are generally derived with index_snps()):
	chr - Character string or factor with chromosome
	• pos - Position (in same units as in the "map" attribute in genoprobs.
	• sdp - Strain distribution pattern: an integer, between 1 and $2^n - 2$ where $n$ is the number of strains, whose binary encoding indicates the founder genotypes
	<ul> <li>snp - Character string with SNP identifier (if missing, the rownames are used).</li> </ul>
	<ul> <li>index - Indices that indicate equivalent groups of SNPs, calculated by index_snps().</li> </ul>
	• intervals - Indexes that indicate which marker intervals the SNPs reside.
	<ul> <li>on_map - Indicate whether SNP coincides with a marker in the genoprobs</li> </ul>
	Alternatively, snpinfo can be a object of class "cross2", as output by read_cross2(), containing the data for a multi-parent population with founder genotypes, in which case the SNP information for all markers with complete founder genotype data is calculated and then used. But, in this case, the genotype probabilities must be at the markers in the cross.

## Details

We first split the SNPs by chromosome and use snpinfo\$index to subset to non-equivalent SNPs. snpinfo\$interval indicates the intervals in the genotype probabilities that contain each. For SNPs contained within an interval, we use the average of the probabilities for the two endpoints. We then collapse the probabilities according to the strain distribution pattern.

An object of class "calc\_genoprob", like the input genoprobs, but with imputed genotype probabilities at the selected SNPs indicated in snpinfo\$index. See calc\_genoprob().

If the input genoprobs is for allele probabilities, the probs output has just two probability columns (for the two SNP alleles). If the input has a full set of n(n + 1)/2 probabilities for n strains, the probs output has 3 probabilities (for the three SNP genotypes). If the input has full genotype probabilities for the X chromosome (n(n + 1)/2 genotypes for the females followed by n hemizygous genotypes for the males), the output has 5 probabilities: the 3 female SNP genotypes followed by the two male hemizygous SNP genotypes.

#### See Also

index\_snps(), calc\_genoprob(), scan1snps()

#### Examples

```
## Not run:
# load example data and calculate genotype probabilities
file <- paste0("https://raw.githubusercontent.com/rqtl/",</pre>
                "qtl2data/main/D0_Recla/recla.zip")
recla <- read_cross2(file)</pre>
recla <- recla[c(1:2,53:54), c("19","X")] # subset to 4 mice and 2 chromosomes
probs <- calc_genoprob(recla, error_prob=0.002)</pre>
# founder genotypes for a set of SNPs
snpgeno <- rbind(m1=c(3,1,1,3,1,1,1,1),</pre>
                 m2=c(1,3,1,3,1,3,1,3),
                  m3=c(1,1,1,1,3,3,3,3),
                  m4=c(1,3,1,3,1,3,1,3))
sdp <- calc_sdp(snpgeno)</pre>
snpinfo <- data.frame(chr=c("19", "19", "X", "X"),</pre>
                       pos=c(40.36, 40.53, 110.91, 111.21),
                       sdp=sdp,
                       snp=c("m1", "m2", "m3", "m4"), stringsAsFactors=FALSE)
# identify groups of equivalent SNPs
snpinfo <- index_snps(recla$pmap, snpinfo)</pre>
# collapse to SNP genotype probabilities
snpprobs <- genoprob_to_snpprob(probs, snpinfo)</pre>
# could also first convert to allele probs
aprobs <- genoprob_to_alleleprob(probs)</pre>
snpaprobs <- genoprob_to_snpprob(aprobs, snpinfo)</pre>
## End(Not run)
```

64

## Value

get\_common\_ids

## Description

For a set objects with IDs as row names (or, for a vector, just names), find the IDs that are present in all of the objects.

## Usage

```
get_common_ids(..., complete.cases = FALSE)
```

## Arguments

	A set of objects: vectors, lists, matrices, data frames, and/or arrays. If one is a character vector with no names attribute, it's taken to be a set of IDs, itself.
complete.cases	If TRUE, look at matrices and non-character vectors and keep only individuals with no missing values.

# Details

This is used (mostly internally) to align phenotypes, genotype probabilities, and covariates in preparation for a genome scan. The complete.cases argument is used to omit individuals with any missing covariate values.

## Value

A vector of character strings for the individuals that are in common.

## Examples

```
x <- matrix(0, nrow=10, ncol=5); rownames(x) <- LETTERS[1:10]
y <- matrix(0, nrow=5, ncol=5); rownames(y) <- LETTERS[(1:5)+7]
z <- LETTERS[5:15]
get_common_ids(x, y, z)
x[8,1] <- NA
get_common_ids(x, y, z)
get_common_ids(x, y, z, complete.cases=TRUE)
```

get\_x\_covar

#### Description

Get the matrix of covariates to be used for the null hypothesis when performing QTL analysis with the X chromosome.

#### Usage

get\_x\_covar(cross)

#### Arguments

cross Object of class "cross2". For details, see the R/qtl2 developer guide.

## Details

For most crosses, the result is either NULL (indicating no additional covariates are needed) or a matrix with a single column containing sex indicators (1 for males and 0 for females).

For an intercross, we also consider cross direction. There are four cases: (1) All male or all female but just one direction: no covariate; (2) All female but both directions: covariate indicating cross direction; (3) Both sexes, one direction: covariate indicating sex; (4) Both sexes, both directions: a covariate indicating sex and a covariate that is 1 for females from the reverse direction and 0 otherwise.

# Value

A matrix of size individuals x no. covariates.

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
xcovar <- get_x_covar(iron)</pre>
```

guess\_phase Guess

Guess phase of imputed genotypes

#### Description

Turn imputed genotypes into phased genotypes along chromosomes by attempting to pick the phase that leads to the fewest recombination events.

#### Usage

```
guess_phase(cross, geno, deterministic = FALSE, cores = 1)
```

#### index\_snps

### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
geno	Imputed genotypes, as a list of matrices, as from maxmarg().
deterministic	If TRUE, preferentially put smaller allele first when there's uncertainty. If FALSE, the order of alleles is random in such cases.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

# Details

We randomly assign the pair of alleles at the first locus to two haplotypes, and then work left to right, assigning alleles to haplotypes one locus at a time seeking the fewest recombination events. The results are subject to arbitrary and random choices. For example, to the right of a homozygous region, either orientation is equally reasonable.

#### Value

If input cross is phase-known (e.g., recombinant inbred lines), the output will be the input geno. Otherwise, the output will be a list of three-dimensional arrays of imputed genotypes, individual x position x haplotype (1/2).

## See Also

maxmarg()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap <- insert_pseudomarkers(iron$gmap, step=1)
probs <- calc_genoprob(iron, gmap, error_prob=0.002)
imp_geno <- maxmarg(probs)
ph_geno <- guess_phase(iron, imp_geno)</pre>
```

index\_snps

Create index of equivalent SNPs

## Description

For a set of SNPs and a map of marker/pseudomarkers, partition the SNPs into groups that are contained within common intervals and have the same strain distribution pattern, and then create an index to a set of distinct SNPs, one per partition.

## Usage

```
index_snps(map, snpinfo, tol = 0.0000001)
```

#### Arguments

map	Physical map of markers and pseudomarkers; generally created from insert_pseudomarkers() and used for a set of genotype probabilities (calculated with calc_genoprob()) that are to be used to interpolate SNP genotype probabilities (with genoprob_to_snpprob()).
snpinfo	Data frame with SNP information with the following columns:
	<ul> <li>chr - Character string or factor with chromosome</li> <li>pos - Position (in same units as in the "map").</li> <li>sdp - Strain distribution pattern: an integer, between 1 and 2<sup>n</sup> - 2 where n is the number of strains, whose binary encoding indicates the founder genotypes</li> <li>snp - Character string with SNP identifier (if missing, the rownames are used).</li> </ul>
tol	Tolerance for determining whether a SNP is exactly at a position at which geno- type probabilities were already calculated.

## Details

We split the SNPs by chromosome and identify the intervals in the map that contain each. For SNPs within tol of a position at which the genotype probabilities were calculated, we take the SNP to be at that position. For each marker position or interval, we then partition the SNPs into groups that have distinct strain distribution patterns, and choose a single index SNP for each partition.

## Value

A data frame containing the input snpinfo with three added columns: "index" (which indicates the groups of equivalent SNPs), "interval" (which indicates the map interval containing the SNP, with values starting at 0), and on\_map (which indicates that the SNP is within tol of a position on the map). The rows get reordered, so that they are ordered by chromosome and position, and the values in the "index" column are *by chromosome*.

#### See Also

genoprob\_to\_snpprob(), scan1snps(), find\_index\_snp()

## Examples

```
pos=c(40.36, 40.53, 110.91, 111.21),
sdp=sdp,
snp=c("m1", "m2", "m3", "m4"), stringsAsFactors=FALSE)
# update snp info by adding the SNP index column
snpinfo <- index_snps(recla$pmap, snpinfo)
## End(Not run)
```

insert\_pseudomarkers Insert pseudomarkers into a marker map

# Description

Insert pseudomarkers into a map of genetic markers

# Usage

```
insert_pseudomarkers(
  map,
  step = 0,
  off_end = 0,
  stepwidth = c("fixed", "max"),
  pseudomarker_map = NULL,
  tol = 0.01,
  cores = 1
)
```

# Arguments

map	A list of numeric vectors; each vector gives marker positions for a single chro- mosome.	
step	Distance between pseudomarkers and markers; if step=0 no pseudomarkers are inserted.	
off_end	Distance beyond terminal markers in which to insert pseudomarkers.	
stepwidth	Indicates whether to use a fixed grid (stepwidth="fixed") or to use the max- imal distance between pseudomarkers to ensure that no two adjacent mark- ers/pseudomarkers are more than step apart.	
pseudomarker_map		
	A map of pseudomarker locations; if provided the step, off_end, and stepwidth arguments are ignored.	
tol	Tolerance for determining whether a pseudomarker would duplicate a marker position.	
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().	

## Details

If stepwidth="fixed", a grid of pseudomarkers is added to the marker map.

If stepwidth="max", a minimal set of pseudomarkers are added, so that the maximum distance between adjacent markers or pseudomarkers is at least step. If two adjacent markers are separated by less than step, no pseudomarkers will be added to the interval. If they are more then step apart, a set of equally-spaced pseudomarkers will be added.

If pseudomarker\_map is provided, then the step, off\_end, and stepwidth arguments are ignored, and the input pseudomarker\_map is taken to be the set of pseudomarker positions.

## Value

A list like the input map with pseudomarkers inserted. Will also have an attribute "is\_x\_chr", taken from the input map.

## See Also

calc\_genoprob(), calc\_grid()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap_w_pmar <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

interp\_genoprob Interpolate genotype probabilities

## Description

Linear interpolation of genotype probabilities, mostly to get two sets onto the same map for comparison purposes.

## Usage

```
interp_genoprob(probs, map, cores = 1)
```

## Arguments

probs	Genotype probabilities, as calculated from calc_genoprob().
map	List of vectors of map positions.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

#### interp\_map

#### Details

We reduce probs to the positions present in map and then interpolate the genotype probabilities at additional positions in map by linear interpolation using the two adjacent positions. Off the ends, we just copy over the first or last value unchanged.

In general, it's better to use insert\_pseudomarkers() and calc\_genoprob() to get genotype probabilities at additional positions along a chromosome. This function is a **very** crude alternative that was implemented in order to compare genotype probabilities derived by different methods, where we first need to get them onto a common set of positions.

#### Value

An object of class "calc\_genoprob", like the input, but with additional positions present in map. See calc\_genoprob().

## See Also

calc\_genoprob()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
probs <- calc_genoprob(iron, iron$gmap, error_prob=0.002)
# you generally wouldn't want to do this, but this is an illustration
map <- insert_pseudomarkers(iron$gmap, step=1)
probs_map <- interp_genoprob(probs, map)</pre>
```

interp\_map

Interpolate between maps

### Description

Use interpolate to convert from one map to another

# Usage

interp\_map(map, oldmap, newmap)

#### Arguments

map	The map to be interpolated; a list of vectors.
oldmap	Map with positions in the original scale, as in map.
newmap	Map with positions in the new scale.

## Value

Object of same form as input map but in the units as in newmap.

## Examples

interp\_map(tointerp, iron\$gmap, iron\$pmap)

invert\_sdp

Calculate SNP genotype matrix from strain distribution patterns

## Description

Calculate the matrix of SNP genotypes from a vector of strain distribution patterns (SDPs).

#### Usage

```
invert_sdp(sdp, n_strains)
```

### Arguments

sdp	Vector of strain distribution patterns (integers between 1 and $2^n - 2$ where	
	the number of strains.	
n_strains	Number of strains	

## Value

Matrix of SNP genotypes, markers x strains, coded as 1 (AA) and 3 (BB). Markers with values other than 1 or 3 are omitted, and monomorphic markers, are omitted.

#### See Also

sdp2char(), calc\_sdp()

### Examples

sdp <- c(m1=1, m2=12, m3=240)
invert\_sdp(sdp, 8)</pre>

locate\_xo

Locate crossovers

# Description

Estimate the locations of crossovers in each individual on each chromosome.

### Usage

```
locate_xo(geno, map, quiet = TRUE, cores = 1)
```

# Arguments

geno	List of matrices of genotypes (output of maxmarg() or viterbi()).
map	List of vectors with the map positions of the markers.
quiet	If FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

# Value

A list of lists of estimated crossover locations, with crossovers placed at the midpoint of the intervals that contain them.

# See Also

count\_xo()

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
map <- insert_pseudomarkers(iron$gmap, step=1)
pr <- calc_genoprob(iron, map, error_prob=0.002, map_function="c-f")
g <- maxmarg(pr)
pos <- locate_xo(g, iron$gmap)</pre>
```

lod\_int

# Description

Calculate LOD support intervals for a single LOD curve on a single chromosome, with the ability to identify intervals for multiple LOD peaks.

#### Usage

```
lod_int(
   scan1_output,
   map,
   chr = NULL,
   lodcolumn = 1,
   threshold = 0,
   peakdrop = Inf,
   drop = 1.5,
   expand2markers = TRUE
)
```

# Arguments

scan1_output	An object of class "scan1" as returned by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
chr	Chromosome ID to consider (must be a single value).
lodcolumn	LOD score column to consider (must be a single value).
threshold	Minimum LOD score for a peak.
peakdrop	Amount that the LOD score must drop between peaks, if multiple peaks are to be defined on a chromosome.
drop	Amount to drop in the support interval. Must be $\leq$ peakdrop
expand2markers	If TRUE, QTL intervals are expanded so that their endpoints are at genetic markers.

# Details

We identify a set of peaks defined as local maxima that exceed the specified threshold, with the requirement that the LOD score must have dropped by at least peakdrop below the lowest of any two adjacent peaks.

At a given peak, if there are ties, with multiple positions jointly achieving the maximum LOD score, we take the average of these positions as the location of the peak.

The default is to use threshold=0, peakdrop=Inf, and drop=1.5. We then return results a single peak, no matter the maximum LOD score, and give a 1.5-LOD support interval.

### Value

A matrix with three columns:

- ci\_lo lower bound of interval
- pos peak position
- ci\_hi upper bound of interval

Each row corresponds to a different peak.

### See Also

```
bayes_int(), find_peaks(), scan1()
```

## Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)</pre>
```

# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>

```
# 1.5-LOD support interval for QTL on chr 7, first phenotype
lod_int(out, map, chr=7, lodcolum=1)
```

map\_to\_grid Subset a map to positions on a grid

## Description

Subset a map object to the locations on some grid.

## Usage

map\_to\_grid(map, grid)

#### Arguments

map	A list of vectors of marker positions.
grid	A list of logical vectors (aligned with map), with TRUE indicating the position is on the grid.

# Details

This is generally for the case of a map created with insert\_pseudomarkers() with step>0 and stepwidth="fixed", so that the pseudomarkers form a grid along each chromosome.

### Value

Same list as input, but subset to just include pseudomarkers along a grid.

# See Also

calc\_grid(), probs\_to\_grid()

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map_w_pmar <- insert_pseudomarkers(grav2$gmap, step=1)
sapply(map_w_pmar, length)
grid <- calc_grid(grav2$gmap, step=1)
map_sub <- map_to_grid(map_w_pmar, grid)
sapply(map_sub, length)</pre>
```

mat2strata Define strata based on rows of a matrix

# Description

Use the rows of a matrix to define a set of strata for a stratified permutation test

### Usage

mat2strata(mat)

#### Arguments

mat A covariate matrix, as individuals x covariates

# Value

A vector of character strings: for each row of mat, we use base::paste() with collapse="|".

### See Also

get\_x\_covar(), scan1perm()

## maxlod

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
Xcovar <- get_x_covar(iron)
perm_strata <- mat2strata(Xcovar)</pre>
```

```
maxlod
```

Overall maximum LOD score

### Description

Find overall maximum LOD score in genome scan results, across all positions and columns.

### Usage

```
maxlod(scan1_output, map = NULL, chr = NULL, lodcolumn = NULL)
```

# Arguments

scan1_output	An object of class "scan1" as returned by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
chr	Optional vector of chromosomes to consider.
lodcolumn	An integer or character string indicating the LOD score column, either as a numeric index or column name. If NULL, return maximum for all columns.

## Value

A single number: the maximum LOD score across all columns and positions for the selected chromosomes.

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)</pre>
```

# maxmarg

```
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)
# overall maximum
maxlod(out)
# maximum on chromosome 2
maxlod(out, map, "2")
```

maxmarg

Find genotypes with maximum marginal probabilities

# Description

For each individual at each position, find the genotype with the maximum marginal probability.

# Usage

```
maxmarg(
  probs,
  map = NULL,
  minprob = 0.95,
  chr = NULL,
  pos = NULL,
  return_char = FALSE,
  quiet = TRUE,
  cores = 1,
  tol = 0.000000000001
)
```

# Arguments

probs	Genotype probabilities, as calculated from calc_genoprob().
map	Map of pseudomarkers in probs. Used only if chr and pos are provided.
minprob	Minimum probability for making a call. If maximum probability is less then this value, give NA.
chr	If provided (along with pos), consider only the single specified position.
pos	If provided (along with chr), consider only the single specified position.
return_char	If TRUE, return genotype names as character strings.
quiet	IF FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().
tol	Tolerance value; genotypes with probability that are within tol of each other are treated as equivalent.

## Details

If multiple genotypes share the maximum probability, one is chosen at random.

### Value

If chr and pos are provided, a vector of genotypes is returned. In this case, map is needed.

Otherwise, the result is a object like that returned by viterbi(), A list of two-dimensional arrays of imputed genotypes, individuals x positions. Also includes these attributes:

- crosstype The cross type of the input cross.
- is\_x\_chr Logical vector indicating whether chromosomes are to be treated as the X chromosome or not, from input cross.
- alleles Vector of allele codes, from input cross.

#### See Also

```
sim_geno(), viterbi()
```

### Examples

```
# load data and calculate genotype probabilities
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
pr <- calc_genoprob(iron, error_prob=0.002)
# full set of imputed genotypes
ginf <- maxmarg(pr)
# imputed genotypes at a fixed position
g <- maxmarg(pr, iron$gmap, chr=8, pos=45.5)
# return genotype names rather than integers
g <- maxmarg(pr, iron$gmap, chr=8, pos=45.5, return_char=TRUE)</pre>
```

max\_compare\_geno Find pair with most similar genotypes

### Description

From results of compare\_geno(), show the pair with most similar genotypes.

### Usage

```
max_compare_geno(object, ...)
## S3 method for class 'compare_geno'
max(object, ...)
```

### Arguments

object	A square matrix with genotype comparisons for pairs of individuals, as output
	<pre>by compare_geno().</pre>
	Ignored

# Value

Data frame with individual pair, proportion matches, number of mismatches, number of matches, and total markers genotyped.

### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
cg <- compare_geno(grav2)
max(cg)</pre>
```

max\_scan1

Find position with maximum LOD score

### Description

Return data frame with the positions having maximum LOD score for a particular LOD score column

# Usage

```
max_scan1(
    scan1_output,
    map = NULL,
    lodcolumn = 1,
    chr = NULL,
    na.rm = TRUE,
    ...
)
## S3 method for class 'scan1'
max(scan1_output, map = NULL, lodcolumn = 1, chr = NULL, na.rm = TRUE, ...)
```

## Arguments

scan1_output	An object of class "scan1" as returned by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
	Can also be an indexed SNP info table, as from index_snps() or scan1snps().
lodcolumn	An integer or character string indicating the LOD score column, either as a nu-
	meric index or column name. If NULL, return maximum for all columns.
chr	Optional vector of chromosomes to consider.
na.rm	Ignored (take to be TRUE)
	Ignored

#### n\_missing

### Value

If map is NULL, the genome-wide maximum LOD score for the selected column is returned. If also lodcolumn is NULL, you get a vector with the maximum LOD for each column.

If map is provided, the return value is a data.frame with three columns: chr, pos, and lod score. But if lodcolumn is NULL, you get the maximum for each lod score column, in the format provided by find\_peaks(), so a data.frame with five columns: lodindex, lodcolumn, chr, pos, and lod.

# Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
Xcovar <- get_x_covar(iron)</pre>
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
# maximum of first column
max(out, map)
# maximum of spleen column
max(out, map, lodcolumn="spleen")
# maximum of first column on chr 2
max(out, map, chr="2")
```

n\_missing

Count missing genotypes

## Description

Number (or proportion) of missing (or non-missing) genotypes by individual or marker

#### Usage

```
n_missing(
  cross,
  by = c("individual", "marker"),
```

```
summary = c("count", "proportion")
)
n_typed(
  cross,
  by = c("individual", "marker"),
  summary = c("count", "proportion")
)
```

## Arguments

cross	An object of class "cross2", as output by read_cross2(). For details, see the R/qtl2 developer guide.
by	Whether to summarize by individual or marker
summary	Whether to take count or proportion

# Value

Vector of counts (or proportions) of missing (or non-missing) genotypes.

## Functions

- n\_missing(): Count missing genotypes
- n\_typed(): Count genotypes

# Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
nmis_ind <- n_missing(iron)
pmis_mar <- n_typed(iron, "mar", "proportion")
plot(nmis_ind, xlab="Individual", ylab="No. missing genotypes")
plot(pmis_mar, xlab="Markers", ylab="Prop. genotyped")</pre>
```

plot\_coef

Plot QTL effects along chromosome

# Description

Plot estimated QTL effects along a chromosomes.

# Usage

```
plot_coef(
    x,
    map,
    columns = NULL,
    col = NULL,
```

```
scan1_output = NULL,
  add = FALSE,
  gap = NULL,
  top_panel_prop = 0.65,
  legend = NULL,
  . . .
)
plot_coefCC(
  х,
  map,
  columns = 1:8,
  col = qtl2::CCcolors,
  scan1_output = NULL,
  add = FALSE,
  gap = NULL,
  top_panel_prop = 0.65,
  legend = NULL,
  . . .
)
## S3 method for class 'scan1coef'
plot(
  х,
  map,
  columns = 1,
  col = NULL,
  scan1_output = NULL,
  add = FALSE,
  gap = NULL,
  top_panel_prop = 0.65,
  legend = NULL,
  . . .
)
```

# Arguments

х	Estimated QTL effects ("coefficients") as obtained from scan1coef().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
columns	Vector of columns to plot
col	Vector of colors, same length as columns. If NULL, some default choices are made.
scan1_output	If provided, we make a two-panel plot with coefficients on top and LOD scores below. Should have just one LOD score column; if multiple, only the first is used.
add	If TRUE, add to current plot (must have same map and chromosomes).
gap	Gap between chromosomes. The default is 1% of the total genome length.

<pre>top_panel_prop</pre>	If scan1_output provided, this gives the proportion of the plot that is devoted
	to the top panel.
legend	Location of legend, such as "bottomleft" or "topright" (NULL for no legend) $% \left( \left( {{{\rm{NULL}}} \right)_{\rm{star}}} \right)$
	Additional graphics parameters.

#### Details

plot\_coefCC() is the same as plot\_coef(), but forcing columns=1:8 and using the Collaborative Cross colors, CCcolors.

### Value

None.

### Hidden graphics parameters

A number of graphics parameters can be passed via .... For example, bgcolor to control the background color, and things like ylab and ylim. These are not included as formal parameters in order to avoid cluttering the function definition.

In the case that scan1\_output is provided, col, ylab, and ylim all control the panel with estimated QTL effects, while col\_lod, ylab\_lod, and ylim\_lod control the LOD curve panel.

If legend is indicated so that a legend is shown, legend\_lab controls the labels in the legend, and legend\_ncol indicates the number of columns in the legend.

## See Also

CCcolors, plot\_scan1(), plot\_snpasso()

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno[,1]
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
# calculate coefficients for chromosome 7
coef <- scan1coef(probs[,7], pheno, addcovar=covar)</pre>
```

```
# plot QTL effects (note the need to subset the map object, for chromosome 7)
plot(coef, map[7], columns=1:3, col=c("slateblue", "violetred", "green3"))
```

plot\_compare\_geno Plot of compare\_geno object.

# Description

From results of compare\_geno(), plot histogram of

## Usage

```
plot_compare_geno(x, rug = TRUE, ...)
```

```
## S3 method for class 'compare_geno'
plot(x, rug = TRUE, ...)
```

### Arguments

x	A square matrix with genotype comparisons for pairs of individuals, as output by compare_geno().
rug	If true, use rug() to plot tick marks at observed values below the histogram.
	Additional graphics parameters passed to hist()

## Value

None.

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
cg <- compare_geno(grav2)
plot(cg)</pre>
```

```
plot_genes
```

Plot gene locations for a genomic interval

# Description

Plot gene locations for a genomic interval, as rectangles with gene symbol (and arrow indicating strand/direction) below.

# Usage

```
plot_genes(
  genes,
  minrow = 4,
  padding = 0.2,
  colors = c("black", "red3", "green4", "blue3", "orange"),
  scale_pos = 1,
  start_field = "start",
  stop_field = "start",
  stop_field = "stop",
  strand_field = "strand",
  name_field = "Name",
  ...
)
```

## Arguments

genes	Data frame containing start and stop in Mbp, strand (as "-", "+", or NA), and Name.
minrow	Minimum number of rows of genes in the plot
padding	Proportion to pad with white space around the genes
colors	Vectors of colors, used sequentially and then re-used.
scale_pos	Factor by which to scale position (for example, to convert basepairs to Mbp)
start_field	Character string with name of column containing the genes' start positions.
stop_field	Character string with name of column containing the genes' stop positions.
strand_field	Character string with name of column containing the genes' strands. (The values of the corresponding field can be character strings "+" or "-", or numeric +1 or -1.)
name_field	Character string with name of column containing the genes' names.
	Optional arguments passed to plot().

# Value

None.

## Hidden graphics parameters

Graphics parameters can be passed via .... For example, xlim to control the x-axis limits. These are not included as formal

# Examples

```
# use scale_pos=1e-6 because data in bp but we want the plot in Mbp
plot_genes(genes, xlim=c(140, 146), scale_pos=1e-6)
```

```
plot_genoprob
```

Plot genotype probabilities for one individual on one chromosome.

### Description

Plot the genotype probabilities for one individual on one chromosome, as a heat map.

### Usage

```
plot_genoprob(
    probs,
    map,
    ind = 1,
    chr = NULL,
    geno = NULL,
    color_scheme = c("gray", "viridis"),
    col = NULL,
    threshold = 0,
    swap_axes = FALSE,
    ...
)
## S3 method for class 'calc_genoprob'
```

```
plot(x, ...)
```

# Arguments

probs	Genotype probabilities (as produced by calc_genoprob()) or allele dosages (as produced by genoprob_to_alleleprob()).
map	Marker map (a list of vectors of marker positions).
ind	Individual to plot, either a numeric index or an ID.
chr	Selected chromosome to plot; a single character string.
geno	Optional vector of genotypes or alleles to be shown (vector of integers or character strings)
color_scheme	Color scheme for the heatmap (ignored if col is provided).
col	Optional vector of colors for the heatmap.
threshold	Threshold for genotype probabilities; only genotypes that achieve this value somewhere on the chromosome will be shown.

swap_axes	If TRUE, swap the axes, so that the genotypes are on the x-axis and the chromo- some position is on the y-axis.
	Additional graphics parameters passed to graphics::image().
x	Genotype probabilities (as produced by calc_genoprob()) or allele dosages (as produced by genoprob_to_alleleprob()). (For the S3 type plot function, this has to be called x.)

#### Value

None.

### Hidden graphics parameters

A number of graphics parameters can be passed via .... For example, hlines, hlines\_col, hlines\_lwd, and hlines\_lty to control the horizontal grid lines. (Use hlines=NA to avoid plotting horizontal grid lines.) Similarly vlines, vlines\_col, vlines\_lwd, and vlines\_lty for vertical grid lines. You can also use many standard graphics parameters like xlab and xlim. These are not included as formal parameters in order to avoid cluttering the function definition.

#### See Also

plot\_genoprobcomp()

```
# load data and calculate genotype probabilities
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
iron <- iron[,"2"] # subset to chr 2</pre>
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
pr <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# plot the probabilities for the individual labeled "262"
\# (white = 0, black = 1)
plot_genoprob(pr, map, ind="262")
# change the x-axis label
plot_genoprob(pr, map, ind="262", xlab="Position (cM)")
# swap the axes so that the chromosome runs vertically
plot_genoprob(pr, map, ind="262", swap_axes=TRUE, ylab="Position (cM)")
# This is more interesting for a Diversity Outbred mouse example
## Not run:
file <- paste0("https://raw.githubusercontent.com/rqtl/",</pre>
               "qtl2data/main/D0ex/D0ex.zip")
D0ex <- read_cross2(file)</pre>
# subset to chr 2 and X and individuals labeled "232" and "256"
D0ex <- D0ex[c("232", "256"), c("2", "X")]</pre>
pr <- calc_genoprob(D0ex, error_prob=0.002)</pre>
# plot individual "256" on chr 2 (default is to pick first chr in the probs)
plot_genoprob(pr, DOex$pmap, ind="256")
```

```
# omit states that never have probability >= 0.5
plot_genoprob(pr, D0ex$pmap, ind="256", threshold=0.05)
# X chr male 232: just show the AY-HY genotype probabilities
plot_genoprob(pr, D0ex$pmap, ind="232", chr="X", geno=paste0(LETTERS[1:8], "Y"))
# could also indicate genotypes by number
plot_genoprob(pr, D0ex$pmap, ind="232", chr="X", geno=37:44)
# and can use negative indexes
plot_genoprob(pr, D0ex$pmap, ind="232", chr="X", geno=-(1:36))
# X chr female 256: just show the first 36 genotype probabilities
plot_genoprob(pr, DOex$pmap, ind="256", chr="X", geno=1:36)
# again, can give threshold to omit genotypes whose probabilities never reach that threshold
plot_genoprob(pr, D0ex$pmap, ind="256", chr="X", geno=1:36, threshold=0.5)
# can also look at the allele dosages
apr <- genoprob_to_alleleprob(pr)</pre>
plot_genoprob(apr, DOex$pmap, ind="232")
## End(Not run)
```

plot\_genoprobcomp Plot comparison of two sets of genotype probabilities

## Description

Plot a comparison of two sets of genotype probabilities for one individual on one chromosome, as a heat map.

### Usage

```
plot_genoprobcomp(
    probs1,
    probs2,
    map,
    ind = 1,
    chr = NULL,
    geno = NULL,
    threshold = 0,
    n_colors = 256,
    swap_axes = FALSE,
    ...
)
```

## Arguments

probs1	Genotype probabilities (as produced by calc_genoprob()) or allele dosages (as produced by genoprob_to_alleleprob()).
probs2	A second set of genotype probabilities, just like probs1.
map	Marker map (a list of vectors of marker positions).
ind	Individual to plot, either a numeric index or an ID.
chr	Selected chromosome to plot; a single character string.
geno	Optional vector of genotypes or alleles to be shown (vector of integers or character strings)
threshold	Threshold for genotype probabilities; only genotypes that achieve this value somewhere on the chromosome (in one or the other set of probabilities) will be shown.
n_colors	Number of colors in each color scale.
swap_axes	If TRUE, swap the axes, so that the genotypes are on the x-axis and the chromo- some position is on the y-axis.
	Additional graphics parameters passed to graphics::image().

## Details

We plot the first set of probabilities in the range white to blue and the second set in the range white to red and attempt to combine them, for colors that are white, some amount of blue or red, or where both are large something like blackish purple.

## Value

None.

### See Also

plot\_genoprob()

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
iron <- iron[228,"1"]  # subset to one individual on chr 1
map <- insert_pseudomarkers(iron$gmap, step=5)
# introduce genotype error and look at difference in genotype probabilities
pr_ne <- calc_genoprob(iron, map, error_prob=0.002)
iron$geno[[1]][1,2] <- 3
pr_e <- calc_genoprob(iron, map, error_prob=0.002)
# image of probabilities + comparison
par(mfrow=c(3,1))
plot_genoprob(pr_ne, map, main="No error")
plot_genoprob(pr_e, map, main="With an error")</pre>
```

# plot\_lodpeaks

plot\_genoprobcomp(pr\_ne, pr\_e, map, main="Comparison")

plot\_lodpeaks

Plot LOD scores vs QTL peak locations

# Description

Create a scatterplot of LOD scores vs QTL peak locations (possibly with intervals) for multiple traits.

### Usage

plot\_lodpeaks(peaks, map, chr = NULL, gap = NULL, intervals = FALSE, ...)

## Arguments

peaks	Data frame such as that produced by find_peaks()) containing columns chr, pos, lodindex, and lodcolumn. May also contain columns ci_lo and ci_hi, in which case intervals will be plotted.
map	Marker map, used to get chromosome lengths (and start and end positions).
chr	Selected chromosomes to plot; a vector of character strings.
gap	Gap between chromosomes. The default is 1% of the total genome length.
intervals	If TRUE and peaks contains QTL intervals, plot the intervals.
	Additional graphics parameters

# Value

None.

# Hidden graphics parameters

A number of graphics parameters can be passed via . . . . For example, bgcolor to control the background color and altbgcolor to control the background color on alternate chromosomes. These are not included as formal parameters in order to avoid cluttering the function definition.

# See Also

find\_peaks(), plot\_peaks()

# Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno</pre>
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
Xcovar <- get_x_covar(iron)</pre>
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
# find peaks above lod=3.5 (and calculate 1.5-LOD support intervals)
peaks <- find_peaks(out, map, threshold=3.5, drop=1.5)</pre>
plot_lodpeaks(peaks, map)
```

plot_onegeno	Plot one individual's genome-wide genotypes

# Description

Plot one individual's genome-wide genotypes

# Usage

```
plot_onegeno(
  geno,
  map,
  ind = 1,
  chr = NULL,
  col = NULL,
  na_col = "white",
  swap_axes = FALSE,
  border = "black",
  shift = FALSE,
  chrwidth = 0.5,
  ...
)
```

### plot\_onegeno

### Arguments

geno	Imputed phase-known genotypes, as a list of matrices (as produced by maxmarg()) or a list of three-dimensional arrays (as produced by guess_phase()).
map	Marker map (a list of vectors of marker positions).
ind	Individual to plot, either a numeric index or an ID.
chr	Selected chromosomes to plot; a vector of character strings.
col	Vector of colors for the different genotypes.
na_col	Color for missing segments.
swap_axes	If TRUE, swap the axes, so that the chromosomes run horizontally.
border	Color of outer border around chromosome rectangles.
shift	If TRUE, shift the chromosomes so they all start at 0.
chrwidth	Total width of rectangles for each chromosome, as a fraction of the distance between them.
	Additional graphics parameters

## Value

None.

#### Hidden graphics parameters

A number of graphics parameters can be passed via .... For example, bgcolor to control the background color. These are not included as formal parameters in order to avoid cluttering the function definition.

```
# load data and calculate genotype probabilities
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
iron <- iron["146", ] # subset to individual 146</pre>
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
pr <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# infer genotypes, as those with maximal marginal probability
m <- maxmarg(pr)</pre>
# guess phase
ph <- guess_phase(iron, m)</pre>
# plot phased genotypes
plot_onegeno(ph, map, shift=TRUE, col=c("slateblue", "Orchid"))
# this is more interesting for Diversity Outbred mouse data
## Not run:
file <- paste0("https://raw.githubusercontent.com/rqtl/",</pre>
                "qtl2data/main/D0ex/D0ex.zip")
D0ex <- read_cross2(file)</pre>
```

```
# subset to individuals labeled "232" and "256"
D0ex <- D0ex[c("232", "256"), ]
pr <- calc_genoprob(D0ex, error_prob=0.002)
# infer genotypes, as those with maximal marginal probability
m <- maxmarg(pr, minprob=0.5)
# guess phase
ph <- guess_phase(D0ex, m)
# plot phased genotypes
plot_onegeno(ph, D0ex$gmap, shift=TRUE)
plot_onegeno(ph, D0ex$gmap, ind="256", shift=TRUE)
## End(Not run)
```

plot\_peaks

Plot QTL peak locations

# Description

Plot QTL peak locations (possibly with intervals) for multiple traits.

# Usage

```
plot_peaks(
    peaks,
    map,
    chr = NULL,
    tick_height = 0.3,
    gap = NULL,
    lod_labels = FALSE,
    ...
)
```

# Arguments

peaks	Data frame such as that produced by find_peaks()) containing columns chr, pos, lodindex, and lodcolumn. May also contain columns ci_lo and ci_hi, in which case intervals will be plotted.
map	Marker map, used to get chromosome lengths (and start and end positions).
chr	Selected chromosomes to plot; a vector of character strings.
tick_height	Height of tick marks at the peaks (a number between 0 and 1).
gap	Gap between chromosomes. The default is 1% of the total genome length.

### plot\_peaks

lod_labels	If TRUE, plot LOD scores near the intervals. Uses three hidden graphics param-
	eters, label_gap (distance between CI and LOD text label), label_left (vec-
	tor that indicates whether the labels should go on the left side; TRUE=on left,
	FALSE=on right, NA=put into larger gap on that chromosome), and label_cex
	that controls the size of these labels
	Additional graphics parameters

## Value

None.

## Hidden graphics parameters

A number of graphics parameters can be passed via . . . . For example, bgcolor to control the background color and altbgcolor to control the background color on alternate chromosomes. These are not included as formal parameters in order to avoid cluttering the function definition.

#### See Also

find\_peaks(), plot\_lodpeaks()

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno</pre>
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
Xcovar <- get_x_covar(iron)</pre>
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
# find peaks above lod=3.5 (and calculate 1.5-LOD support intervals)
peaks <- find_peaks(out, map, threshold=3.5, drop=1.5)</pre>
plot_peaks(peaks, map)
# show LOD scores
plot_peaks(peaks, map, lod_labels=TRUE)
# show LOD scores, controlling whether they go on the left or right
```

plot\_pxg

plot\_pxg

Plot phenotype vs genotype

# Description

Plot phenotype vs genotype for a single putative QTL and a single phenotype.

# Usage

```
plot_pxg(
  geno,
  pheno,
  sort = TRUE,
  SEmult = NULL,
  pooledSD = TRUE,
  swap_axes = FALSE,
  jitter = 0.2,
  force_labels = TRUE,
  alternate_labels = FALSE,
  omit_points = FALSE,
  ...
)
```

# Arguments

geno	Vector of genotypes, for example as produced by maxmarg() with specific chr and pos.
pheno	Vector of phenotypes.
sort	If TRUE, sort genotypes from largest to smallest.
SEmult	If specified, interval estimates of the within-group averages will be displayed, as mean +/- SE $\star$ SEmult.
pooledSD	If TRUE and SEmult is specified, calculated a pooled within-group SD. Otherwise, get separate estimates of the within-group SD for each group.
swap_axes	If TRUE, swap the axes, so that the genotypes are on the y-axis and the pheno- type is on the x-axis.
jitter	Amount to jitter the points horizontally, if a vector of length $> 0$ , it is taken to be the actual jitter amounts (with values between -0.5 and 0.5).
force_labels	If TRUE, force all genotype labels to be shown.
alternate_label	S
	If TRUE, place genotype labels in two rows
omit_points	If TRUE, omit the points, just plotting the averages (and, potentially, the +/- SE intervals).
	Additional graphics parameters, passed to plot().

### plot\_pxg

### Value

(Invisibly) A matrix with rows being the genotype groups and columns for the means and (if SEmult is specified) the SEs.

### Hidden graphics parameters

A number of graphics parameters can be passed via .... For example, bgcolor to control the background color, and seg\_width, seg\_lwd, and seg\_col to control the lines at the confidence intervals. Further, hlines, hlines\_col, hlines\_lwd, and hlines\_lty to control the horizontal grid lines. (Use hlines=NA to avoid plotting horizontal grid lines.) Similarly vlines, vlines\_col, vlines\_lwd, and vlines\_lty for vertical grid lines. These are not included as formal parameters in order to avoid cluttering the function definition.

## See Also

## plot\_coef()

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# inferred genotype at a 28.6 cM on chr 16
geno <- maxmarg(probs, map, chr=16, pos=28.6, return_char=TRUE)</pre>
# plot phenotype vs genotype
plot_pxg(geno, log10(iron$pheno[,1]), ylab=expression(log[10](Liver)))
# include +/- 2 SE intervals
plot_pxg(geno, log10(iron$pheno[,1]), ylab=expression(log[10](Liver)),
         SEmult=2)
# plot just the means
plot_pxg(geno, log10(iron$pheno[,1]), ylab=expression(log[10](Liver)),
         omit_points=TRUE)
# plot just the means +/- 2 SEs
plot_pxg(geno, log10(iron$pheno[,1]), ylab=expression(log[10](Liver)),
         omit_points=TRUE, SEmult=2)
```

plot\_scan1

## Description

Plot LOD curves for a genome scan

## Usage

```
plot_scan1(x, map, lodcolumn = 1, chr = NULL, add = FALSE, gap = NULL, ...)
## S3 method for class 'scan1'
plot(x, map, lodcolumn = 1, chr = NULL, add = FALSE, gap = NULL, ...)
```

## Arguments

х	An object of class "scan1", as output by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
lodcolumn	LOD score column to plot (a numeric index, or a character string for a column name). Only one value allowed.
chr	Selected chromosomes to plot; a vector of character strings.
add	If TRUE, add to current plot (must have same map and chromosomes).
gap	Gap between chromosomes. The default is 1% of the total genome length.
	Additional graphics parameters.

## Value

None.

# Hidden graphics parameters

A number of graphics parameters can be passed via .... For example, bgcolor to control the background color and altbgcolor to control the background color on alternate chromosomes. col controls the color of lines/curves; altcol can be used if you want alternative chromosomes in different colors. These are not included as formal parameters in order to avoid cluttering the function definition.

# See Also

plot\_coef(), plot\_coefCC(), plot\_snpasso()

### plot\_sdp

### Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno</pre>
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
Xcovar <- get_x_covar(iron)</pre>
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
# plot the results for selected chromosomes
ylim <- c(0, maxlod(out)*1.02) # need to strip class to get overall max LOD
chr <- c(2,7,8,9,15,16)
plot(out, map, chr=chr, ylim=ylim)
plot(out, map, lodcolumn=2, chr=chr, col="violetred", add=TRUE)
legend("topleft", lwd=2, col=c("darkslateblue", "violetred"), colnames(out),
       bg="gray90")
# plot just one chromosome
plot(out, map, chr=8, ylim=ylim)
plot(out, map, chr=8, lodcolumn=2, col="violetred", add=TRUE)
# lodcolumn can also be a column name
plot(out, map, lodcolumn="liver", ylim=ylim)
plot(out, map, lodcolumn="spleen", col="violetred", add=TRUE)
```

plot\_sdp

plot strain distribution patterns for SNPs

### Description

plot the strain distribution patterns of SNPs using tracks of tick-marks for each founder strain

#### Usage

```
plot_sdp(pos, sdp, strain_labels = names(qtl2::CCcolors), ...)
```

### Arguments

pos vector of SNP positions

# Details

Additional arguments, such as xlab, ylab, xlim, and main, are passed via ...; also bgcolor to control the color of the background, and col and lwd to control the color and thickness of the tick marks.

## Value

None.

## See Also

calc\_sdp(), invert\_sdp()

## Examples

```
n_tick <- 50
plot_sdp(runif(n_tick, 0, 100), sample(0:255, n_tick, replace=TRUE))</pre>
```

plot\_snpasso Plot SNP associations

## Description

Plot SNP associations, with possible expansion from distinct snps to all snps.

# Usage

```
plot_snpasso(
  scanloutput,
  snpinfo,
  genes = NULL,
  lodcolumn = 1,
  show_all_snps = TRUE,
  chr = NULL,
  add = FALSE,
  drop_hilit = NA,
  col_hilit = "violetred",
  col = "darkslateblue",
  gap = NULL,
 minlod = 0,
  sdp_panel = FALSE,
  strain_labels = names(qtl2::CCcolors),
  . . .
)
```

# plot\_snpasso

# Arguments

scan1output	Output of scan1() using SNP probabilities derived by genoprob_to_snpprob().
snpinfo	Data frame with SNP information with the following columns (the last three are generally derived from with index_snps()):
	chr - Character string or factor with chromosome
	• pos - Position (in same units as in the "map" attribute in genoprobs.
	• sdp - Strain distribution pattern: an integer, between 1 and $2^n - 2$ where $n$ is the number of strains, whose binary encoding indicates the founder genotypes
	• snp - Character string with SNP identifier (if missing, the rownames are used).
	• index - Indices that indicate equivalent groups of SNPs.
	• intervals - Indexes that indicate which marker intervals the SNPs reside.
	• on_map - Indicate whether SNP coincides with a marker in the genoprobs
genes	Optional data frame containing gene information for the region, with columns start and stop in Mbp, strand (as "-", "+", or NA), and Name. If included, a two-panel plot is produced, with SNP associations above and gene locations below.
lodcolumn	LOD score column to plot (a numeric index, or a character string for a column name). Only one value allowed.
show_all_snps	If TRUE, expand to show all SNPs.
chr	Vector of character strings with chromosome IDs to plot.
add	If TRUE, add to current plot (must have same map and chromosomes).
drop_hilit	SNPs with LOD score within this amount of the maximum SNP association will be highlighted.
col_hilit	Color of highlighted points
col	Color of other points
gap	Gap between chromosomes. The default is 1% of the total genome length.
minlod	Minimum LOD to display. (Mostly for GWAS, in which case using minlod=1 will greatly increase the plotting speed, since the vast majority of points would be omitted.
sdp_panel	Include a panel with the strain distribution patterns for the highlighted SNPs
strain_labels	Labels for the strains, if sdp_pane1=TRUE.
•••	Additional graphics parameters.

# Value

None.

#### **Hidden graphics parameters**

A number of graphics parameters can be passed via .... For example, bgcolor to control the background color,altbgcolor to control the background color on alternate chromosomes, altcol to control the point color on alternate chromosomes, cex for character expansion for the points (default 0.5), pch for the plotting character for the points (default 16), and ylim for y-axis limits. If you are including genes and/or SDP panels, you can use panel\_prop to control the relative heights of the panels, from top to bottom.

## See Also

plot\_scan1(), plot\_coef(), plot\_coefCC()

### Examples

```
## Not run:
# load example DO data from web
file <- paste0("https://raw.githubusercontent.com/rqtl/",</pre>
                "qtl2data/main/D0ex/D0ex.zip")
D0ex <- read_cross2(file)</pre>
# subset to chr 2
D0ex <- D0ex[,"2"]</pre>
# calculate genotype probabilities and convert to allele probabilities
pr <- calc_genoprob(D0ex, error_prob=0.002)</pre>
apr <- genoprob_to_alleleprob(pr)</pre>
# query function for grabbing info about variants in region
snp_dbfile <- system.file("extdata", "cc_variants_small.sqlite", package="qtl2")</pre>
query_variants <- create_variant_query_func(snp_dbfile)</pre>
# SNP association scan
out_snps <- scan1snps(apr, D0ex$pmap, D0ex$pheno, query_func=query_variants,
                       chr=2, start=97, end=98, keep_all_snps=TRUE)
# plot results
plot_snpasso(out_snps$lod, out_snps$snpinfo)
# can also just type plot()
plot(out_snps$lod, out_snps$snpinfo)
# plot just subset of distinct SNPs
plot(out_snps$lod, out_snps$snpinfo, show_all_snps=FALSE)
# highlight the top snps (with LOD within 1.5 of max)
plot(out_snps$lod, out_snps$snpinfo, drop_hilit=1.5)
# query function for finding genes in region
gene_dbfile <- system.file("extdata", "mouse_genes_small.sqlite", package="qtl2")</pre>
query_genes <- create_gene_query_func(gene_dbfile)</pre>
genes <- query_genes(2, 97, 98)</pre>
```

```
# plot SNP association results with gene locations
plot(out_snps$lod, out_snps$snpinfo, drop_hilit=1.5, genes=genes)
# plot SNP asso results with genes plus SDPs of highlighted SNPs
plot(out_snps$lod, out_snps$snpinfo, drop_hilit=2, genes=genes, sdp_panel=TRUE)
## End(Not run)
```

predict\_snpgeno Predict SNP genotypes

# Description

Predict SNP genotypes in a multiparent population from inferred genotypes plus founder strains' SNP alleles.

# Usage

predict\_snpgeno(cross, geno, cores = 1)

#### Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
geno	Imputed genotypes, as a list of matrices, as from maxmarg().
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

### Value

A list of matrices with inferred SNP genotypes, coded 1/2/3.

## See Also

maxmarg(), viterbi(), calc\_errorlod()

## Examples

m <- maxmarg(probs, minprob=0.5)</pre>

```
# inferred SNP genotypes
inferg <- predict_snpgeno(DOex, m)
## End(Not run)</pre>
```

print.cross2 Print a cross2 object

# Description

Print a summary of a cross2 object

### Usage

## S3 method for class 'cross2'
print(x, ...)

# Arguments

x An object of class "cross2", as output by read\_cross2(). For details, see the R/qtl2 developer guide.
 ... Ignored.

## Value

None.

print.summary.scan1perm

Print summary of scan1perm permutations

### Description

Print summary of scan1perm permutations

# Usage

## S3 method for class 'summary.scan1perm'
print(x, digits = 3, ...)

# Arguments

Х	Object of class "summary.scan1perm", as produced by summary_scan1perm().
digits	Number of digits in printing significance thresholds; passed to base::print().
	Ignored.

### probs\_to\_grid

## Details

This is to go with summary\_scan1perm(), so that the summary output is printed in a nice format. Generally not called directly, but it can be in order to control the number of digits that appear.

## Value

Invisibly returns the input, x.

## Examples

probs\_to\_grid Subset genotype probability array to pseudomarkers on a grid

#### Description

Subset genotype probability array (from calc\_genoprob() to a grid of pseudomarkers along each chromosome.

### Usage

```
probs_to_grid(probs, grid)
```

## Arguments

probs	Genotype probabilities as output from calc_genoprob() with stepwidth="fixed".
grid	List of logical vectors that indicate which positions are on the grid and should
	be retained.

## Details

This only works if calc\_genoprob() was run with stepwidth="fixed", so that the genotype probabilities were calculated at a grid of markers/pseudomarkers. When this is the case, we omit all but the probabilities on this grid. Use calc\_grid() to find the grid positions.

# Value

An object of class "calc\_genoprob", like the input, subset to just include pseudomarkers along a grid. See calc\_genoprob().

## See Also

```
calc_grid(), map_to_grid()
```

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map_w_pmar <- insert_pseudomarkers(grav2$gmap, step=1)
probs <- calc_genoprob(grav2, map_w_pmar, error_prob=0.002)
sapply(probs, dim)
grid <- calc_grid(grav2$gmap, step=1)
probs_sub <- probs_to_grid(probs, grid)
sapply(probs_sub, dim)</pre>
```

pull\_genoprobint *Pull genotype probabilities for an interval* 

## Description

Pull out the genotype probabilities for a given genomic interval

#### Usage

```
pull_genoprobint(genoprobs, map, chr, interval)
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
map	The marker map for the genotype probabilities
chr	Chromosome ID (single character sting)
interval	Interval (pair of numbers)

## Value

A list containing a single 3d array of genotype probabilities, like the input genoprobs but for the designated interval.

# pull\_genoprobpos

## See Also

find\_marker(), pull\_genoprobpos()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap <- insert_pseudomarkers(iron$gmap, step=1)
pr <- calc_genoprob(iron, gmap, error_prob=0.002)
pr_sub <- pull_genoprobint(pr, gmap, "8", c(25, 35))</pre>
```

pull\_genoprobpos Pull genotype probabilities for a particular position

# Description

Pull out the genotype probabilities for a particular position (by name)

### Usage

```
pull_genoprobpos(genoprobs, map = NULL, chr = NULL, pos = NULL, marker = NULL)
```

## Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
map	A map object: a list (corresponding to chromosomes) of vectors of marker po- sitions. Can also be a snpinfo object (data frame with columns chr and pos; marker names taken from column snp or if that doesn't exist from the row names)
chr	A chromosome ID
pos	A numeric position
marker	A single character string with the name of the position to pull out.

# Details

Provide either a marker/pseudomarker name (with the argument marker) or all of map, chr, and pos.

# Value

A matrix of genotype probabilities for the specified position.

# See Also

find\_marker(), fit1(), pull\_genoprobint()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap <- insert_pseudomarkers(iron$gmap, step=1)
pr <- calc_genoprob(iron, gmap, error_prob=0.002)
pmar <- find_marker(gmap, 8, 40)
pr_8_40 <- pull_genoprobpos(pr, pmar)
pr_8_40_alt <- pull_genoprobpos(pr, gmap, 8, 40)</pre>
```

pull\_markers Drop all but a specified set of markers

# Description

Drop all markers from a cross2 object expect those in a specified vector.

## Usage

pull\_markers(cross, markers)

## Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
markers	A vector of marker names.

# Value

The input cross with only the specified markers.

# See Also

drop\_markers(), drop\_nullmarkers()

### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
markers2drop <- c("BH.342C/347L-Col", "GH.94L", "EG.357C/359L-Col", "CD.245L", "ANL2")
grav2_rev <- pull_markers(grav2, markers2drop)</pre>
```

qtl2version

#### Description

Get installed version of R/qtl2

# Usage

qtl2version()

#### Value

A character string with the installed version of the R/qtl2 package.

# Examples

qtl2version()

rbind.calc\_genoprob Join genotype probabilities for different individuals

# Description

Join multiple genotype probability objects, as produced by calc\_genoprob(), for the same set of markers and genotypes but for different individuals.

# Usage

## S3 method for class 'calc\_genoprob'
rbind(...)

# Arguments

... Genotype probability objects as produced by calc\_genoprob(). Must have the same set of markers and genotypes.

# Value

An object of class "calc\_genoprob", like the input; see calc\_genoprob().

# See Also

cbind.calc\_genoprob()

## Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
probsA <- calc_genoprob(grav2[1:5,], map, error_prob=0.002)
probsB <- calc_genoprob(grav2[6:12,], map, error_prob=0.002)
probs <- rbind(probsA, probsB)</pre>
```

rbind.scan1

Join genome scan results for different chromosomes.

# Description

Join multiple scan1() results for different chromosomes; must have the same set of lod score column.

### Usage

```
## S3 method for class 'scan1'
rbind(...)
```

#### Arguments

• • •

Genome scan objects of class "scan1", as produced by scan1(). Must have the same lod score columns.

# Details

If components addcovar, Xcovar, intcovar, weights, sample\_size do not match between objects, we omit this information.

If hsq present, we simply rbind() the contents.

## Value

An object of class '"scan1", like the inputs, but with the results for different sets of chromosomes combined.

# See Also

cbind.scan1(), scan1()

# 110

#### rbind.scan1perm

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
probs <- calc_genoprob(grav2, map, error_prob=0.002)
phe <- grav2$pheno[,1,drop=FALSE]
out1 <- scan1(probs[,1], phe) # chr 1
out2 <- scan1(probs[,5], phe) # chr 5
out <- rbind(out1, out2)</pre>
```

rbind.scan1perm Combine data from scan1perm objects

## Description

Row-bind multiple scan1perm objects with the same set of columns

#### Usage

```
## S3 method for class 'scan1perm'
rbind(...)
## S3 method for class 'scan1perm'
c(...)
```

#### Arguments

... A set of permutation results from scan1perm() (objects of class "scan1perm"). They must have the same set of columns. If any include autosome/X chromosomespecific permutations, they must all be such.

# Details

The aim of this function is to concatenate the results from multiple runs of a permutation test with scan1perm(), to assist in the case that such permutations are done on multiple processors in parallel.

## Value

The combined row-binded input, as an object of class "scan1perm"; see scan1perm().

## See Also

cbind.scan1perm(), scan1perm(), scan1()

# Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)
# permutations with genome scan (just 3 replicates, for illustration)
operm1 <- scan1perm(probs, pheno, addcovar=covar, Xcovar=Xcovar, n_perm=3)
operm2 <- scan1perm(probs, pheno, addcovar=covar, Xcovar=Xcovar, n_perm=3)
operm <- rbind(operm1, operm2)</pre>
```

rbind.sim\_geno Join genotype imputations for different individuals

## Description

Join multiple genotype imputation objects, as produced by sim\_geno(), for the same set of markers but for different individuals.

#### Usage

## S3 method for class 'sim\_geno'
rbind(...)

#### Arguments

... Genotype imputation objects as produced by sim\_geno(). Must have the same set of markers and genotypes.

# Value

An object of class "sim\_geno", like the input; see sim\_geno().

#### See Also

```
cbind.sim_geno(), sim_geno()
```

# 112

## rbind.viterbi

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
drawsA <- sim_geno(grav2[1:5,], map, error_prob=0.002, n_draws=4)
drawsB <- sim_geno(grav2[6:12,], map, error_prob=0.002, n_draws=4)
draws <- rbind(drawsA, drawsB)</pre>
```

rbind.viterbi Join Viterbi results for different individuals

# Description

Join multiple imputed genotype objects, as produced by viterbi(), for the same set of markers but for different individuals.

#### Usage

```
## S3 method for class 'viterbi'
rbind(...)
```

#### Arguments

• • •

Imputed genotype objects as produced by viterbi(). Must have the same set of markers.

# Value

An object of class "viterbi", like the input; see viterbi().

## See Also

cbind.viterbi(), viterbi()

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
gA <- viterbi(grav2[1:5,], map, error_prob=0.002)
gB <- viterbi(grav2[6:12,], map, error_prob=0.002)
g <- rbind(gA, gB)</pre>
```

read\_cross2

## Description

Read QTL data from a set of files

#### Usage

read\_cross2(file, quiet = TRUE)

#### Arguments

file	Character string with path to the YAML or JSON file containing all of the control
	information. This could instead be a zip file containing all of the data files, in
	which case the contents are unzipped to a temporary directory and then read.
quiet	If FALSE, print progress messages.

## Details

A control file in YAML or JSON format contains information about basic parameters as well as the names of the series of data files to be read. See the sample data files and the vignette describing the input file format.

# Value

Object of class "cross2". For details, see the R/qtl2 developer guide.

# See Also

```
read_pheno(), write_control_file(), sample data files at https://kbroman.org/qtl2/pages/
sampledata.html and https://github.com/rqtl/qtl2data
```

```
## Not run:
yaml_file <- "https://kbroman.org/qtl2/assets/sampledata/grav2/grav2.yaml"
grav2 <- read_cross2(yaml_file)
## End(Not run)
zip_file <- system.file("extdata", "grav2.zip", package="qtl2")
grav2 <- read_cross2(zip_file)</pre>
```

read\_pheno

# Description

Read phenotype data from a CSV file (and, optionally, phenotype covariate data from a separate CSV file). The CSV files may be contained in zip files, separately or together.

#### Usage

```
read_pheno(
   file,
   phenocovarfile = NULL,
   sep = ",",
   na.strings = c("-", "NA"),
   comment.char = "#",
   transpose = FALSE,
   quiet = TRUE
)
```

# Arguments

file	Character string with path to the phenotype data file (or a zip file containing both the phenotype and phenotype covariate files).
phenocovarfile	Character string with path to the phenotype covariate file. This can be a separate CSV or zip file; if a zip file, it must contain exactly one CSV file. Alternatively, if the file argument indicates a zip file that contains two files (phenotypes and phenotype covariates), then this phenocovarfile argument must indicate the base name for the phenotype covariate file.
sep	the field separator character
na.strings	a character vector of strings which are to be interpreted as NA values.
comment.char	A character vector of length one containing a single character to denote comments within the CSV files.
transpose	If TRUE, the phenotype data will be transposed. The phenotype covariate infor- mation is <b>never</b> transposed.
quiet	If FALSE, print progress messages.

# Value

Either a matrix of phenotype data, or a list containing pheno (phenotype matrix) and phenocovar (phenotype covariate matrix).

#### See Also

```
read_cross2(), sample data files at https://kbroman.org/qtl2/pages/sampledata.html and
https://github.com/rqtl/qtl2data
```

## Examples

recode\_snps Recode SNPs by major allele

# Description

For multi-parent populations with founder genotypes, recode the raw SNP genotypes so that 1 means homozygous for the major allele in the founders.

## Usage

```
recode_snps(cross)
```

#### Arguments

cross Object of class "cross2". For details, see the R/qtl2 developer guide.

# Value

The input cross object with the raw SNP genotypes recoded so that 1 is homozygous for the major alleles in the founders.

#### See Also

calc\_raw\_founder\_maf(), calc\_raw\_maf()

#### Examples

## End(Not run)

116

reduce\_map\_gaps Reduce the lengths of gaps in a map

# Description

Reduce the lengths of gaps in a map

#### Usage

```
reduce_map_gaps(map, min_gap = 50)
```

#### Arguments

map	Genetic map as a list of vectors (each vector is a chromosome and contains the marker positions).
min_gap	Minimum gap length to return.

# Value

Input map with any gaps greater than min\_gap reduced to min\_gap.

## See Also

find\_map\_gaps()

# Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
rev_map <- reduce_map_gaps(iron$gmap, 30)</pre>
```

reduce\_markers Reduce markers to a subset of more-evenly-spaced ones

# Description

Find the largest subset of markers such that no two adjacent markers are separated by less than some distance.

# Usage

```
reduce_markers(
  map,
  min_distance = 1,
  weights = NULL,
  max_batch = 10000,
  batch_distance_mult = 1,
  cores = 1
)
```

# Arguments

map	A list with each component being a vector with the marker positions for a chro- mosome.	
<pre>min_distance</pre>	Minimum distance between markers.	
weights	A (optional) list of weights on the markers; same size as map.	
max_batch	Maximum number of markers to consider in a batch	
batch_distance_mult		
	If working with batches of markers, reduce min_distance by this multiple.	
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().	

# Details

Uses a dynamic programming algorithm to find, for each chromosome, the subset of markers for with max(weights) is maximal, subject to the constraint that no two adjacent markers may be separated by more than min\_distance.

The computation time for the algorithm grows with like the square of the number of markers, like 1 sec for 10k markers but 30 sec for 50k markers. If the number of markers on a chromosome is greater than max\_batch, the markers are split into batches and the algorithm applied to each batch with min\_distance smaller by a factor min\_distance\_mult, and then merged together for one last pass.

## Value

A list like the input map, but with the selected subset of markers.

# References

Broman KW, Weber JL (1999) Method for constructing confidently ordered linkage maps. Genet Epidemiol 16:337–343

# See Also

find\_dup\_markers(), drop\_markers()

118

# replace\_ids

# Examples

```
# read data
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
# grab genetic map
gmap <- grav2$gmap
# subset to markers that are >= 1 cM apart
gmap_sub <- reduce_markers(gmap, 1)
# drop all of the other markers from the cross
markers2keep <- unlist(lapply(gmap_sub, names))
grav2_sub <- pull_markers(grav2, markers2keep)</pre>
```

replace\_ids Replace individual IDs

## Description

Replace the individual IDs in an object with new ones

## Usage

```
replace_ids(x, ids)
## S3 method for class 'cross2'
replace_ids(x, ids)
## S3 method for class 'calc_genoprob'
replace_ids(x, ids)
## S3 method for class 'viterbi'
replace_ids(x, ids)
## S3 method for class 'sim_geno'
replace_ids(x, ids)
## S3 method for class 'matrix'
replace_ids(x, ids)
## S3 method for class 'data.frame'
replace_ids(x, ids)
```

#### Arguments

Х	Object whose IDs will be replaced
ids	Vector of character strings with the new individual IDs, with the names being
	the original IDs.

## Value

The input x object, but with individual IDs replaced.

#### Methods (by class)

- replace\_ids(cross2): Replace IDs in a "cross2" object
- replace\_ids(calc\_genoprob): Replace IDs in output from calc\_genoprob()
- replace\_ids(viterbi): Replace IDs in output from viterbi()
- replace\_ids(sim\_geno): Replace IDs in output from sim\_geno()
- replace\_ids(matrix): Replace IDs in a matrix
- replace\_ids(data.frame): Replace IDs in a data frame

#### Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
ids <- as.numeric(ind_ids(iron))</pre>
```

```
# replace the numeric IDs with IDs like "mouse003"
new_ids <- setNames( sprintf("mouse%03d", as.numeric(ids)), ids)</pre>
```

```
iron <- replace_ids(iron, new_ids)</pre>
```

scale\_kinship Scale kinship matrix

# Description

Scale kinship matrix to be like a correlation matrix.

#### Usage

```
scale_kinship(kinship)
```

## Arguments

kinship A kinship matrix, or a list of such in the case of the "leave one chromosome out" method, as calculated by calc\_kinship().

# Details

We take  $c_{ij} = k_{ij} / \sqrt{k_{ii} k_{jj}}$ 

#### Value

A matrix or list of matrices, as with the input, but with the matrices scaled to be like correlation matrices.

# scan1

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map <- insert_pseudomarkers(grav2$gmap, step=1)
probs <- calc_genoprob(grav2, map, error_prob=0.002)
K <- calc_kinship(probs)
Ka <- scale_kinship(K)</pre>
```

scan1

Genome scan with a single-QTL model

# Description

Genome scan with a single-QTL model by Haley-Knott regression or a linear mixed model, with possible allowance for covariates.

#### Usage

```
scan1(
  genoprobs,
  pheno,
  kinship = NULL,
  addcovar = NULL,
  Xcovar = NULL,
  intcovar = NULL,
  weights = NULL,
  reml = TRUE,
  model = c("normal", "binary"),
  hsq = NULL,
  cores = 1,
  ...
)
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
pheno	A numeric matrix of phenotypes, individuals x phenotypes.
kinship	Optional kinship matrix, or a list of kinship matrices (one per chromosome), in order to use the LOCO (leave one chromosome out) method.
addcovar	An optional numeric matrix of additive covariates.
Xcovar	An optional numeric matrix with additional additive covariates used for null hypothesis when scanning the X chromosome.
intcovar	An numeric optional matrix of interactive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.
reml	If kinship provided: if reml=TRUE, use REML; otherwise maximum likelihood.

scan1

model	Indicates whether to use a normal model (least squares) or binary model (logistic regression) for the phenotype. If model="binary", the phenotypes must have values in $[0, 1]$ .
hsq	Considered only if kinship is provided, in which case this is taken as the as- sumed value for the residual heritability. It should be a vector with length corre- sponding to the number of columns in pheno, or (if kinship corresponds to a list of LOCO kinship matrices) a matrix with dimension length(kinship) x ncol(pheno).
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster(). Additional control parameters; see Details.
	1 '

#### Details

We first fit the model  $y = X\beta + \epsilon$  where X is a matrix of covariates (or just an intercept) and  $\epsilon$  is multivariate normal with mean 0 and covariance matrix  $\sigma^2[h^2(2K) + I]$  where K is the kinship matrix and I is the identity matrix.

We then take  $h^2$  as fixed and then scan the genome, at each genomic position fitting the model  $y = P\alpha + X\beta + \epsilon$  where P is a matrix of genotype probabilities for the current position and again X is a matrix of covariates  $\epsilon$  is multivariate normal with mean 0 and covariance matrix  $\sigma^2[h^2(2K)+I]$ , taking  $h^2$  to be known.

Note that if weights are provided, the covariance matrix becomes  $\sigma^2[h^2(2K) + D]$  where D is a diagonal matrix with the reciprocal of the weights.

For each of the inputs, the row names are used as individual identifiers, to align individuals. The genoprobs object should have a component "is\_x\_chr" that indicates which of the chromosomes is the X chromosome, if any.

The ... argument can contain several additional control parameters; suspended for simplicity (or confusion, depending on your point of view). tol is used as a tolerance value for linear regression by QR decomposition (in determining whether columns are linearly dependent on others and should be omitted); default 1e-12. intcovar\_method indicates whether to use a high-memory (but potentially faster) method or a low-memory (and possibly slower) method, with values "highmem" or "lowmem"; default "lowmem". max\_batch indicates the maximum number of phenotypes to run together; default is unlimited. maxit is the maximum number of iterations for converence of the iterative algorithm used when model=binary. bintol is used as a tolerance for converence for the iterative algorithm used when model=binary. eta\_max is the maximum value for the "linear predictor" in the case model="binary" (a bit of a technicality to avoid fitted values exactly at 0 or 1).

If kinship is absent, Haley-Knott regression is performed. If kinship is provided, a linear mixed model is used, with a polygenic effect estimated under the null hypothesis of no (major) QTL, and then taken as fixed as known in the genome scan.

If kinship is a single matrix, then the hsq in the results is a vector of heritabilities (one value for each phenotype). If kinship is a list (one matrix per chromosome), then hsq is a matrix, chromosomes x phenotypes.

#### Value

An object of class "scan1": a matrix of LOD scores, positions x phenotypes. Also contains one or more of the following attributes:

scan1

- sample\_size Vector of sample sizes used for each phenotype
- hsq Included if kinship provided: A matrix of estimated heritabilities under the null hypothesis of no QTL. Columns are the phenotypes. If the "loco" method was used with calc\_kinship() to calculate a list of kinship matrices, one per chromosome, the rows of hsq will be the heritabilities for the different chromosomes (well, leaving out each one). If Xcovar was not NULL, there will at least be an autosome and X chromosome row.

#### References

Haley CS, Knott SA (1992) A simple regression method for mapping quantitative trait loci in line crosses using flanking markers. Heredity 69:315–324.

Kang HM, Zaitlen NA, Wade CM, Kirby A, Heckerman D, Daly MJ, Eskin E (2008) Efficient control of population structure in model organism association mapping. Genetics 178:1709–1723.

# See Also

scan1perm(), scan1coef(), cbind.scan1(), rbind.scan1(), scan1max()

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno</pre>
covar <- match(iron$covar$sex, c("f", "m")) # make numeric</pre>
names(covar) <- rownames(iron$covar)</pre>
Xcovar <- get_x_covar(iron)</pre>
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
# leave-one-chromosome-out kinship matrices
kinship <- calc_kinship(probs, "loco")</pre>
# genome scan with a linear mixed model
out_lmm <- scan1(probs, pheno, kinship, covar, Xcovar)</pre>
```

#### scan1blup

# Description

Calculate BLUPs of QTL effects in scan along one chromosome, with a single-QTL model treating the QTL effects as random, with possible allowance for covariates and for a residual polygenic effect.

# Usage

```
scan1blup(
  genoprobs,
  pheno,
  kinship = NULL,
  addcovar = NULL,
  nullcovar = NULL,
  contrasts = NULL,
  se = FALSE,
  reml = TRUE,
  tol = 0.00000000001,
  cores = 1,
  quiet = TRUE
)
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
pheno	A numeric vector of phenotype values (just one phenotype, not a matrix of them)
kinship	Optional kinship matrix, or a list of kinship matrices (one per chromosome), in order to use the LOCO (leave one chromosome out) method.
addcovar	An optional numeric matrix of additive covariates.
nullcovar	An optional numeric matrix of additional additive covariates that are used under the null hypothesis (of no QTL) but not under the alternative (with a QTL). This is needed for the X chromosome, where we might need sex as a additive covariate under the null hypothesis, but we wouldn't want to include it under the alternative as it would be collinear with the QTL effects. Only used if kinship is provided but hsq is not, to get estimate of residual heritability.
contrasts	An optional numeric matrix of genotype contrasts, size genotypes x genotypes. For an intercross, you might use $cbind(mu=c(1,0,0), a=c(-1, 0, 1), d=c(0, 1, 0))$ to get mean, additive effect, and dominance effect. The default is the identity matrix.
se	If TRUE, also calculate the standard errors.
reml	If rem1=TRUE, use REML to estimate variance components; otherwise maximum likelihood.

#### Details

For each of the inputs, the row names are used as individual identifiers, to align individuals.

If kinship is provided, the linear mixed model accounts for a residual polygenic effect, with a the polygenic variance estimated under the null hypothesis of no (major) QTL, and then taken as fixed as known in the scan to estimate QTL effects.

If contrasts is provided, the genotype probability matrix, P, is post-multiplied by the contrasts matrix, A, prior to fitting the model. So we use  $P \cdot A$  as the X matrix in the model. One might view the rows of  $A^{-1}$  as the set of contrasts, as the estimated effects are the estimated genotype effects pre-multiplied by  $A^{-1}$ .

#### Value

An object of class "scan1coef": a matrix of estimated regression coefficients, of dimension positions x number of effects. The number of effects is n\_genotypes + n\_addcovar. May also contain the following attributes:

- SE Present if se=TRUE: a matrix of estimated standard errors, of same dimension as coef.
- sample\_size Vector of sample sizes used for each phenotype

#### References

Haley CS, Knott SA (1992) A simple regression method for mapping quantitative trait loci in line crosses using flanking markers. Heredity 69:315–324.

Kang HM, Zaitlen NA, Wade CM, Kirby A, Heckerman D, Daly MJ, Eskin E (2008) Efficient control of population structure in model organism association mapping. Genetics 178:1709–1723.

Robinson GK (1991) That BLUP is a good thing: The estimation of random effects. Statist Sci 6:15–32.

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# convert to allele probabilities
aprobs <- genoprob_to_alleleprob(probs)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno[,1]
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
# calculate BLUPs of coefficients for chromosome 7
blup <- scan1blup(aprobs[,"7"], pheno, addcovar=covar)
# leave-one-chromosome-out kinship matrix for chr 7
kinship7 <- calc_kinship(probs, "loco")[["7"]]
# calculate BLUPs of coefficients for chromosome 7, adjusting for residual polygenic effect</pre>
```

# calculate BLUPs of coefficients for chromosome 7, adjusting for residual polygenic effect blup\_pg <- scan1blup(aprobs[,"7"], pheno, kinship7, addcovar=covar)</pre>

scan1coef

Calculate QTL effects in scan along one chromosome

#### Description

Calculate QTL effects in scan along one chromosome with a single-QTL model using Haley-Knott regression or a linear mixed model (the latter to account for a residual polygenic effect), with possible allowance for covariates.

## Usage

```
scan1coef(
  genoprobs,
  pheno,
  kinship = NULL,
  addcovar = NULL,
  nullcovar = NULL,
  intcovar = NULL,
  weights = NULL,
  contrasts = NULL,
  model = c("normal", "binary"),
  zerosum = TRUE,
  se = FALSE,
  hsq = NULL,
  reml = TRUE,
  ...
)
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
pheno	A numeric vector of phenotype values (just one phenotype, not a matrix of them)

scan1coef

kinship	Optional kinship matrix, or a list of kinship matrices (one per chromosome), in order to use the LOCO (leave one chromosome out) method.
addcovar	An optional numeric matrix of additive covariates.
nullcovar	An optional numeric matrix of additional additive covariates that are used under the null hypothesis (of no QTL) but not under the alternative (with a QTL). This is needed for the X chromosome, where we might need sex as a additive covariate under the null hypothesis, but we wouldn't want to include it under the alternative as it would be collinear with the QTL effects. Only used if kinship is provided but hsq is not, to get estimate of residual heritability.
intcovar	An optional numeric matrix of interactive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.
contrasts	An optional numeric matrix of genotype contrasts, size genotypes x genotypes. For an intercross, you might use $cbind(mu=c(1,1,1), a=c(-1, 0, 1), d=c(0, 1, 0))$ to get mean, additive effect, and dominance effect. The default is the identity matrix.
model	Indicates whether to use a normal model (least squares) or binary model (logistic regression) for the phenotype. If model="binary", the phenotypes must have values in $[0, 1]$ .
zerosum	If TRUE, force the genotype or allele coefficients sum to 0 by subtracting their mean and add another column with the mean. Ignored if contrasts is provided.
se	If TRUE, also calculate the standard errors.
hsq	(Optional) residual heritability; used only if kinship provided.
reml	If kinship provided: if reml=TRUE, use REML; otherwise maximum likelihood.
	Additional control parameters; see Details;

# Details

For each of the inputs, the row names are used as individual identifiers, to align individuals.

If kinship is absent, Haley-Knott regression is performed. If kinship is provided, a linear mixed model is used, with a polygenic effect estimated under the null hypothesis of no (major) QTL, and then taken as fixed as known in the genome scan.

If contrasts is provided, the genotype probability matrix, P, is post-multiplied by the contrasts matrix, A, prior to fitting the model. So we use  $P \cdot A$  as the X matrix in the model. One might view the rows of  $A^{-1}$  as the set of contrasts, as the estimated effects are the estimated genotype effects pre-multiplied by  $A^{-1}$ .

The ... argument can contain several additional control parameters; suspended for simplicity (or confusion, depending on your point of view). tol is used as a tolerance value for linear regression by QR decomposition (in determining whether columns are linearly dependent on others and should be omitted); default 1e-12. maxit is the maximum number of iterations for converence of the iterative algorithm used when model=binary. bintol is used as a tolerance for converence for the iterative algorithm used when model=binary. eta\_max is the maximum value for the "linear predictor" in the case model="binary" (a bit of a technicality to avoid fitted values exactly at 0 or 1).

#### Value

An object of class "scan1coef": a matrix of estimated regression coefficients, of dimension positions x number of effects. The number of effects is n\_genotypes + n\_addcovar + (n\_genotypes-1)\*n\_intcovar. May also contain the following attributes:

- SE Present if se=TRUE: a matrix of estimated standard errors, of same dimension as coef.
- sample\_size Vector of sample sizes used for each phenotype

#### References

Haley CS, Knott SA (1992) A simple regression method for mapping quantitative trait loci in line crosses using flanking markers. Heredity 69:315–324.

Kang HM, Zaitlen NA, Wade CM, Kirby A, Heckerman D, Daly MJ, Eskin E (2008) Efficient control of population structure in model organism association mapping. Genetics 178:1709–1723.

#### Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)
# grab phenotypes and covariates; ensure that covariates have names attribute
```

```
pheno <- iron$pheno[,1]
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)</pre>
```

```
# calculate coefficients for chromosome 7
coef <- scan1coef(probs[,"7"], pheno, addcovar=covar)</pre>
```

```
# leave-one-chromosome-out kinship matrix for chr 7
kinship7 <- calc_kinship(probs, "loco")[["7"]]</pre>
```

```
# calculate coefficients for chromosome 7, adjusting for residual polygenic effect
coef_pg <- scan1coef(probs[,"7"], pheno, kinship7, addcovar=covar)</pre>
```

scan1max

Maximum LOD score from genome scan with a single-QTL model

#### Description

Maximum LOD score from genome scan with a single-QTL model by Haley-Knott regression or a linear mixed model, with possible allowance for covariates.

# scan1max

# Usage

```
scan1max(
  genoprobs,
  pheno,
  kinship = NULL,
  addcovar = NULL,
  Xcovar = NULL,
  intcovar = NULL,
  weights = NULL,
  weights = NULL,
  reml = TRUE,
  model = c("normal", "binary"),
  hsq = NULL,
  by_chr = FALSE,
  cores = 1,
  ...
)
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
pheno	A numeric matrix of phenotypes, individuals x phenotypes.
kinship	Optional kinship matrix, or a list of kinship matrices (one per chromosome), in order to use the LOCO (leave one chromosome out) method.
addcovar	An optional numeric matrix of additive covariates.
Xcovar	An optional numeric matrix with additional additive covariates used for null hypothesis when scanning the X chromosome.
intcovar	An numeric optional matrix of interactive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.
reml	If kinship provided: if rem1=TRUE, use REML; otherwise maximum likelihood.
model	Indicates whether to use a normal model (least squares) or binary model (logistic regression) for the phenotype. If model="binary", the phenotypes must have values in $[0, 1]$ .
hsq	Considered only if kinship is provided, in which case this is taken as the as- sumed value for the residual heritability. It should be a vector with length corre- sponding to the number of columns in pheno, or (if kinship corresponds to a list of LOCO kinship matrices) a matrix with dimension length(kinship) x ncol(pheno).
by_chr	If TRUE, save the individual chromosome maxima.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().
	Additional control parameters; see Details.

# Details

Equivalent to running scan1() and then saving the column maxima, with some savings in memory usage.

#### Value

Either a vector of genome-wide maximum LOD scores, or if by\_chr is TRUE, a matrix with the chromosome-specific maxima, with the rows being the chromosomes and the columns being the phenotypes.

# See Also

scan1(), scan1perm()

## Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)
# perform genome scan
out <- scan1max(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
```

scan1perm

Permutation test for genome scan with a single-QTL model

# Description

Permutation test for a enome scan with a single-QTL model by Haley-Knott regression or a linear mixed model, with possible allowance for covariates.

#### Usage

```
scan1perm(
  genoprobs,
  pheno,
  kinship = NULL,
  addcovar = NULL,
  Xcovar = NULL,
  intcovar = NULL,
```

130

# scan1perm

```
weights = NULL,
reml = TRUE,
model = c("normal", "binary"),
n_perm = 1,
perm_Xsp = FALSE,
perm_strata = NULL,
chr_lengths = NULL,
cores = 1,
...
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
pheno	A numeric matrix of phenotypes, individuals x phenotypes.
kinship	Optional kinship matrix, or a list of kinship matrices (one per chromosome), in order to use the LOCO (leave one chromosome out) method.
addcovar	An optional numeric matrix of additive covariates.
Xcovar	An optional numeric matrix with additional additive covariates used for null hypothesis when scanning the X chromosome.
intcovar	An optional numeric matrix of interactive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.
reml	If kinship provided: if reml=TRUE, use REML; otherwise maximum likelihood.
model	Indicates whether to use a normal model (least squares) or binary model (logistic regression) for the phenotype. If model="binary", the phenotypes must have values in $[0, 1]$ .
n_perm	Number of permutation replicates.
perm_Xsp	If TRUE, do separate permutations for the autosomes and the X chromosome.
perm_strata	Vector of strata, for a stratified permutation test. Should be named in the same way as the rows of pheno. The unique values define the strata.
chr_lengths	Lengths of the chromosomes; needed only if perm_Xsp=TRUE. See chr_lengths().
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().
	Additional control parameters; see Details.

# Details

If kinship is not provided, so that analysis proceeds by Haley-Knott regression, we permute the rows of the phenotype data; the same permutations are also applied to the rows of the covariates (addcovar, Xcovar, and intcovar) are permuted.

If kinship is provided, we instead permute the rows of the genotype data and fit an LMM with the same residual heritability (estimated under the null hypothesis of no QTL).

If Xcovar is provided and perm\_strata=NULL, we do a stratified permutation test with the strata defined by the rows of Xcovar. If a simple permutation test is desired, provide perm\_strata that is a vector containing a single repeated value.

The ... argument can contain several additional control parameters; suspended for simplicity (or confusion, depending on your point of view). tol is used as a tolerance value for linear regression by QR decomposition (in determining whether columns are linearly dependent on others and should be omitted); default 1e-12. maxit is the maximum number of iterations for converence of the iterative algorithm used when model=binary. bintol is used as a tolerance for converence for the iterative algorithm used when model=binary. eta\_max is the maximum value for the "linear predictor" in the case model="binary" (a bit of a technicality to avoid fitted values exactly at 0 or 1).

#### Value

If perm\_Xsp=FALSE, the result is matrix of genome-wide maximum LOD scores, permutation replicates x phenotypes. If perm\_Xsp=TRUE, the result is a list of two matrices, one for the autosomes and one for the X chromosome. The object is given class "scan1perm".

#### References

Churchill GA, Doerge RW (1994) Empirical threshold values for quantitative trait mapping. Genetics 138:963–971.

Manichaikul A, Palmer AA, Sen S, Broman KW (2007) Significance thresholds for quantitative trait locus mapping under selective genotyping. Genetics 177:1963–1966.

Haley CS, Knott SA (1992) A simple regression method for mapping quantitative trait loci in line crosses using flanking markers. Heredity 69:315–324.

Kang HM, Zaitlen NA, Wade CM, Kirby A, Heckerman D, Daly MJ, Eskin E (2008) Efficient control of population structure in model organism association mapping. Genetics 178:1709–1723.

#### See Also

scan1(), chr\_lengths(), mat2strata()

#### Examples

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)</pre>
```

132

## scan1snps

scan1snps

Single-QTL genome scan at imputed SNPs

## Description

Perform a single-QTL scan across the genome or a defined region at SNPs genotyped in the founders, by Haley-Knott regression or a liear mixed model, with possible allowance for covariates.

#### Usage

```
scan1snps(
  genoprobs,
 map,
 pheno,
 kinship = NULL,
  addcovar = NULL,
 Xcovar = NULL,
  intcovar = NULL,
 weights = NULL,
  reml = TRUE,
 model = c("normal", "binary"),
  query_func = NULL,
  chr = NULL,
  start = NULL,
  end = NULL,
  snpinfo = NULL,
  batch_length = 20,
```

```
keep_all_snps = FALSE,
cores = 1,
...
```

# Arguments

genoprobs	Genotype probabilities as calculated by calc_genoprob().
map	Physical map for the positions in the genoprobs object: A list of numeric vec- tors; each vector gives marker positions for a single chromosome.
pheno	A numeric matrix of phenotypes, individuals x phenotypes.
kinship	Optional kinship matrix, or a list of kinship matrices (one per chromosome), in order to use the LOCO (leave one chromosome out) method.
addcovar	An optional numeric matrix of additive covariates.
Xcovar	An optional numeric matrix with additional additive covariates used for null hypothesis when scanning the X chromosome.
intcovar	An optional numeric matrix of interactive covariates.
weights	An optional numeric vector of positive weights for the individuals. As with the other inputs, it must have names for individual identifiers.
reml	If kinship provided: if reml=TRUE, use REML; otherwise maximum likelihood.
model	Indicates whether to use a normal model (least squares) or binary model (logistic regression) for the phenotype. If model="binary", the phenotypes must have values in $[0, 1]$ .
query_func	Function for querying SNP information; see create_variant_query_func()). Takes arguments chr, start, end, (with start and end in the units in map, generally Mbp), and returns a data frame containing the columns snp, chr, pos, and sdp. (See snpinfo below.)
chr	Chromosome or chromosomes to scan
start	Position defining the start of an interval to scan. Should be a single number, and if provided, chr should also have length 1.
end	Position defining the end of an interval to scan. Should be a single number, and if provided, chr should also have length 1.
snpinfo	Optional data frame of SNPs to scan; if provided, query_func, chr, start, and end are ignored. Should contain the following columns:
	<ul> <li>chr - Character string or factor with chromosome</li> <li>pos - Position (in same units as in the "map").</li> <li>sdp - Strain distribution pattern: an integer, between 1 and 2<sup>n</sup> - 2 where n is the number of strains, whose binary encoding indicates the founder genotypes</li> <li>snp - Character string with SNP identifier (if missing, the rownames are used).</li> </ul>
batch_length	Interval length (in units of map, generally Mbp) to scan at one time.

## scan1snps

keep_all_snps	SNPs are grouped into equivalence classes based on position and founder geno- types; if keep_all_snps=FALSE, the return value will contain information only on the indexed SNPs (one per equivalence class).
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().
	Additional control parameters passed to scan1()

#### Details

The analysis proceeds as follows:

- Call query\_func() to grab all SNPs over a region.
- Use index\_snps() to group SNPs into equivalence classes.
- Use genoprob\_to\_snpprob() to convert genoprobs to SNP probabilities.
- Use scan1() to do a single-QTL scan at the SNPs.

#### Value

A list with two components: lod (matrix of LOD scores) and snpinfo (a data frame of SNPs that were scanned, including columns index which indicates groups of equivalent SNPs)

## See Also

scan1(), genoprob\_to\_snpprob(), index\_snps(), create\_variant\_query\_func(), plot\_snpasso()

sdp2char

## Description

Convert a vector of numeric codes for strain distribution patterns to character strings.

#### Usage

sdp2char(sdp, n\_strains = NULL, strains = NULL)

## Arguments

sdp	Vector of strain distribution patterns (integers between 1 and $2^n - 2$ where n is the number of strains.
n_strains	Number of founder strains (if missing but strains is provided, we use the length of strains)
strains	Vector of single-letter codes for the strains

## Value

Vector of character strings with the two groups of alleles separated by a vertical bar (I).

#### See Also

invert\_sdp(), calc\_sdp()

# Examples

```
sdp <- c(m1=1, m2=12, m3=240)
sdp2char(sdp, 8)
sdp2char(sdp, strains=c("A", "B", "1", "D", "Z", "C", "P", "W"))</pre>
```

•	
C 1 m	aono
STIII	geno

Simulate genotypes given observed marker data

#### Description

Uses a hidden Markov model to simulate from the joint distribution  $Pr(g \mid O)$  where g is the underlying sequence of true genotypes and O is the observed multipoint marker data, with possible allowance for genotyping errors.

#### sim\_geno

## Usage

```
sim_geno(
    cross,
    map = NULL,
    n_draws = 1,
    error_prob = 0.0001,
    map_function = c("haldane", "kosambi", "c-f", "morgan"),
    lowmem = FALSE,
    quiet = TRUE,
    cores = 1
)
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
map	Genetic map of markers. May include pseudomarker locations (that is, locations that are not within the marker genotype data). If NULL, the genetic map in cross is used.
n_draws	Number of simulations to perform.
error_prob	Assumed genotyping error probability
<pre>map_function</pre>	Character string indicating the map function to use to convert genetic distances to recombination fractions.
lowmem	If FALSE, split individuals into groups with common sex and crossinfo and then precalculate the transition matrices for a chromosome; potentially a lot faster but using more memory.
quiet	If FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

## Details

After performing the backward equations, we draw from  $Pr(g_1 = v|O)$  and then  $Pr(g_{k+1} = v|O, g_k = u)$ .

# Value

An object of class "sim\_geno": a list of three-dimensional arrays of imputed genotypes, individuals x positions x draws. Also contains three attributes:

- crosstype The cross type of the input cross.
- is\_x\_chr Logical vector indicating whether chromosomes are to be treated as the X chromosome or not, from input cross.
- alleles Vector of allele codes, from input cross.

# See Also

```
cbind.sim_geno(), rbind.sim_geno()
```

## Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map_w_pmar <- insert_pseudomarkers(grav2$gmap, step=1)
draws <- sim_geno(grav2, map_w_pmar, n_draws=4, error_prob=0.002)</pre>
```

smooth\_gmap

# Smooth genetic map

# Description

Smooth a genetic map by mixing it with a bit of constant recombination (using a separate recombination rate for each chromosome), to eliminate intervals that have exactly 0 recombination.

#### Usage

```
smooth_gmap(gmap, pmap, alpha = 0.02)
```

## Arguments

gmap	Genetic map, as a list of numeric vectors; each vector gives marker positions for a single chromosome.
pmap	Physical map, as a list of numeric vectors; each vector gives marker positions for a single chromosome, with the same chromosomes and markers as gmap.
alpha	Proportion of mixture to take from constant recombination.

# Details

An interval of genetic length  $d_g$  and physical length  $d_p$  is changed to have length  $(1-\alpha)d_g + \alpha d_p r$ where  $r = L_g/L_p$  is the chromosome-specific recombination rate.

## Value

A genetic map like the input gmap, but smoothed by mixing it with a proportion alpha of constant recombination on each chromosome.

#### See Also

unsmooth\_gmap()

## Examples

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap_adj <- smooth_gmap(iron$gmap, iron$pmap)</pre>
```

138

subset.calc\_genoprob Subsetting genotype probabilities

#### Description

Pull out a specified set of individuals and/or chromosomes from the results of calc\_genoprob().

# Usage

## S3 method for class 'calc\_genoprob'
subset(x, ind = NULL, chr = NULL, ...)

## S3 method for class 'calc\_genoprob'
x[ind = NULL, chr = NULL]

## Arguments

х	Genotype probabilities as output from calc_genoprob().
ind	A vector of individuals: numeric indices, logical values, or character string IDs
chr	A vector of chromosomes: logical values, or character string IDs. Numbers are interpreted as character string IDs.
	Ignored.

#### Value

An object of class "calc\_genoprob", like the input, with the selected individuals and/or chromsomes; see calc\_genoprob().

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))</pre>
```

```
pr <- calc_genoprob(grav2)
# keep just individuals 1:5, chromosome 2
prsub <- pr[1:5,2]
# keep just chromosome 2
prsub2 <- pr[,2]</pre>
```

subset.cross2

#### Description

Pull out a specified set of individuals and/or chromosomes from a cross2 object.

#### Usage

```
## S3 method for class 'cross2'
subset(x, ind = NULL, chr = NULL, ...)
## S3 method for class 'cross2'
x[ind = NULL, chr = NULL]
```

#### Arguments

Х	An object of class "cross2", as output by read_cross2(). For details, see the R/qtl2 developer guide.
ind	A vector of individuals: numeric indices, logical values, or character string IDs.
chr	A vector of chromosomes: numeric indices, logical values, or character string IDs
	Ignored.

#### Details

When subsetting by individual, if ind is numeric, they're assumed to be numeric indices; if character strings, they're assumed to be individual IDs. ind can be numeric or logical only if the genotype, phenotype, and covariate data all have the same individuals in the same order.

When subsetting by chromosome, chr is *always* converted to character strings and treated as chromosome IDs. So if there are three chromosomes with IDs "18", "19", and "X", mycross[,18] will give the first of the chromosomes (labeled "18") and mycross[,3] will give an error.

When using character string IDs for ind or chr, you can use "negative" subscripts to indicate exclusions, for example mycross[,c("-18", "-X")] or mycross["-Mouse2501",]. But you can't mix "positive" and "negative" subscripts, and if any of the individuals has an ID that begins with "-", you can't use negative subscripts like this.

#### Value

The input cross2 object, with the selected individuals and/or chromsomes.

#### Warning

The order of the two arguments is reversed relative to the related function in R/qtl.

#### subset.sim\_geno

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
# keep individuals 1-20 and chromosomes 3 and 4
grav2sub <- grav2[1:20, c(3,4)]
# keep just chromosome 1
grav2_c1 <- grav2[,1]</pre>
```

subset.sim\_geno Subsetting imputed genotypes

# Description

Pull out a specified set of individuals and/or chromosomes from the results of sim\_geno().

# Usage

```
## S3 method for class 'sim_geno'
subset(x, ind = NULL, chr = NULL, ...)
## S3 method for class 'sim_geno'
x[ind = NULL, chr = NULL]
```

# Arguments

x	Imputed genotypes as output from sim_geno().
ind	A vector of individuals: numeric indices, logical values, or character string IDs
chr	A vector of chromosomes: logical values, or character string IDs. Numbers are interpreted as character string IDs.
	Ignored.

# Value

An object of class "sim\_geno", like the input with the selected individuals and/or chromsomes; see sim\_geno().

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
dr <- sim_geno(grav2, n_draws=4)
# keep just individuals 1:5, chromosome 2
drsub <- dr[1:5,2]
# keep just chromosome 2
drsub2 <- dr[,2]</pre>
```

subset.viterbi

## Description

Pull out a specified set of individuals and/or chromosomes from the results of viterbi()

# Usage

```
## S3 method for class 'viterbi'
subset(x, ind = NULL, chr = NULL, ...)
## S3 method for class 'viterbi'
x[ind = NULL, chr = NULL]
```

## Arguments

х	Imputed genotypes as output from viterbi().
ind	A vector of individuals: numeric indices, logical values, or character string IDs
chr	A vector of chromosomes: logical values, or character string IDs. Numbers are interpreted as character string IDs.
	Ignored.

#### Value

An object of class "viterbi", like the input, with the selected individuals and/or chromosomes; see viterbi().

## Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
g <- viterbi(grav2)
# keep just individuals 1:5, chromosome 2
gsub <- g[1:5,2]
# keep just chromosome 2</pre>
```

gsub2 <- g[,2]

subset\_scan1

#### Description

Subset the output of scan1() by chromosome or column

#### Usage

```
subset_scan1(x, map = NULL, chr = NULL, lodcolumn = NULL, ...)
## S3 method for class 'scan1'
subset(x, map = NULL, chr = NULL, lodcolumn = NULL, ...)
```

## Arguments

х	An object of class "scan1" as returned by scan1().
map	A list of vectors of marker positions, as produced by insert_pseudomarkers().
chr	Vector of chromosomes.
lodcolumn	Vector of integers or character strings indicating the LOD score columns, either as a numeric indexes or column names.
	Ignored

## Value

Object of class "scan1", like the input, but subset by chromosome and/or column. See scan1().

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)
# pull out chromosome 8
```

```
out_c8 <- subset(out, map, chr="8")
# just the second column on chromosome 2
out_c2_spleen <- subset(out, map, "2", "spleen")
# all positions, but just the "liver" column
out_spleen <- subset(out, map, lodcolumn="spleen")</pre>
```

summary.cross2 Summary of cross2 object

# Description

Summarize a cross2 object

# Usage

## S3 method for class 'cross2'
summary(object, ...)

# Arguments

object	An object of class "cross2", as output by read_cross2(). For details, see the R/qtl2 developer guide.
	Ignored.

## Value

None.

# See Also

basic\_summaries

summary\_compare\_geno Basic summary of compare\_geno object

#### Description

From results of compare\_geno(), show pairs of individuals with similar genotypes.

#### summary\_scan1perm

## Usage

```
summary_compare_geno(object, threshold = 0.9, ...)
```

## S3 method for class 'compare\_geno'
summary(object, threshold = 0.9, ...)

```
## S3 method for class 'summary.compare_geno'
print(x, digits = 2, ...)
```

### Arguments

object	A square matrix with genotype comparisons for pairs of individuals, as output by compare_geno().
threshold	Minimum proportion matches for a pair of individuals to be shown.
	Ignored
x	Results of summary.compare_geno()
digits	Number of digits to print

### Value

Data frame with names of individuals in pair, proportion matches, number of mismatches, number of matches, and total markers genotyped. Last two columns are the numeric indexes of the individuals in the pair.

# Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
cg <- compare_geno(grav2)
summary(cg)</pre>
```

summary\_scan1perm Summarize scan1perm results

# Description

Summarize permutation test results from scan1perm(), as significance thresholds.

# Usage

```
summary_scan1perm(object, alpha = 0.05)
## S3 method for class 'scan1perm'
summary(object, alpha = 0.05, ...)
```

#### Arguments

object	An object of class "scanoneperm", as output by scan1perm()
alpha	Vector of significance levels
	Ignored

#### Details

In the case of X-chromosome-specific permutations (when scan1perm() was run with perm\_Xsp=TRUE, we follow the approach of Broman et al. (2006) to get separate thresholds for the autosomes and X chromosome, using

Let  $L_A$  and  $L_X$  be total the genetic lengths of the autosomes and X chromosome, respectively, and let  $L_T = L_A + L_X$  Then in place of  $\alpha$ , we use

$$\alpha_A = 1 - (1 - \alpha)^{L_A/L_T}$$

as the significance level for the autosomes and

$$\alpha_X = 1 - (1 - \alpha)^{L_X/L_T}$$

as the significance level for the X chromosome.

#### Value

An object of class summary.scan1perm. If scan1perm() was run with perm\_Xsp=FALSE, this is a single matrix of significance thresholds, with rows being signicance levels and columns being the columns in the input. If scan1perm() was run with perm\_Xsp=TRUE, this is a list of two matrices, with the significance thresholds for the autosomes and X chromosome, respectively.

The result has an attribute "n\_perm" that has the numbers of permutation replicates (either a matrix or a list of two matrices).

#### References

Broman KW, Sen Ś, Owens SE, Manichaikul A, Southard-Smith EM, Churchill GA (2006) The X chromosome in quantitative trait locus mapping. Genetics 174:2151-2158

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))</pre>
```

```
# insert pseudomarkers into map
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno</pre>
```

top\_snps

Create table of top snp associations

# Description

Create a table of the top snp associations

# Usage

```
top_snps(
   scan1_output,
   snpinfo,
   lodcolumn = 1,
   chr = NULL,
   drop = 1.5,
   show_all_snps = TRUE
)
```

# Arguments

scan1_output	Output of scan1(). Should contain a component "snpinfo", as when scan1() is run with SNP probabilities produced by genoprob_to_snpprob().
snpinfo	Data frame with SNP information with the following columns (the last three are generally derived with index_snps()):
	chr - Character string or factor with chromosome
	• pos - Position (in same units as in the "map" attribute in genoprobs.
	• sdp - Strain distribution pattern: an integer, between 1 and $2^n - 2$ where $n$ is the number of strains, whose binary encoding indicates the founder genotypes
	• snp - Character string with SNP identifier (if missing, the rownames are used).
	• index - Indices that indicate equivalent groups of SNPs, calculated by index_snps().
	<ul><li>intervals - Indexes that indicate which marker intervals the SNPs reside.</li><li>on_map - Indicate whether SNP coincides with a marker in the genoprobs</li></ul>

lodcolumn	Selected LOD score column to (a numeric index, or a character string for a column name). Only one value allowed.
chr	Selected chromosome; only one value allowed.
drop	Show all SNPs with LOD score within this amount of the maximum SNP association.
show_all_snps	If TRUE, expand to show all SNPs.

#### Value

Data frame like the input snpinfo with just the selected subset of rows, and with an added column with the LOD score.

#### See Also

index\_snps(), genoprob\_to\_snpprob(), scan1snps(), plot\_snpasso()

```
## Not run:
# load example DO data from web
file <- paste0("https://raw.githubusercontent.com/rqtl/",</pre>
                "qtl2data/main/DOex/DOex.zip")
D0ex <- read_cross2(file)</pre>
# subset to chr 2
DOex <- DOex[,"2"]
# calculate genotype probabilities and convert to allele probabilities
pr <- calc_genoprob(D0ex, error_prob=0.002)</pre>
apr <- genoprob_to_alleleprob(pr)</pre>
# query function for grabbing info about variants in region
dbfile <- system.file("extdata", "cc_variants_small.sqlite", package="qtl2")</pre>
query_variants <- create_variant_query_func(dbfile)</pre>
# SNP association scan, keep information on all SNPs
out_snps <- scan1snps(apr, D0ex$pmap, D0ex$pheno, query_func=query_variants,</pre>
                       chr=2, start=97, end=98, keep_all_snps=TRUE)
# table with top SNPs
top_snps(out_snps$lod, out_snps$snpinfo)
# top SNPs among the distinct subset at which calculations were performed
top_snps(out_snps$lod, out_snps$snpinfo, show_all_snps=FALSE)
# top SNPs within 0.5 LOD of max
top_snps(out_snps$lod, out_snps$snpinfo, drop=0.5)
## End(Not run)
```

unsmooth\_gmap

# Description

Performs the reverse operation of smooth\_gmap(), in case one wants to go back to the original genetic map.

#### Usage

```
unsmooth_gmap(gmap, pmap, alpha = 0.02)
```

#### Arguments

gmap	Genetic map, as a list of numeric vectors; each vector gives marker positions for a single chromosome.
pmap	Physical map, as a list of numeric vectors; each vector gives marker positions for a single chromosome, with the same chromosomes and markers as gmap.
alpha	Proportion of mixture to take from constant recombination.

#### Details

An interval of genetic length  $d_g$  and physical length  $d_p$  is changed to have length  $(d_g - \alpha d_p r)/(1 - \alpha)$  where  $r = L_g/L_p$  is the chromosome-specific recombination rate.

# Value

A genetic map like the input gmap, but with the reverse operation of smooth\_gmap() applied, provided that exactly the same physical map and alpha are used.

# See Also

```
smooth_gmap()
```

```
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
gmap_adj <- smooth_gmap(iron$gmap, iron$pmap)
gmap_back <- unsmooth_gmap(gmap_adj, iron$pmap)</pre>
```

#### viterbi

#### Description

Uses a hidden Markov model to calculate arg max  $Pr(g \mid O)$  where g is the underlying sequence of true genotypes and O is the observed multipoint marker data, with possible allowance for genotyping errors.

#### Usage

```
viterbi(
  cross,
  map = NULL,
  error_prob = 0.0001,
  map_function = c("haldane", "kosambi", "c-f", "morgan"),
  lowmem = FALSE,
  quiet = TRUE,
  cores = 1
)
```

# Arguments

cross	Object of class "cross2". For details, see the R/qtl2 developer guide.
map	Genetic map of markers. May include pseudomarker locations (that is, locations that are not within the marker genotype data). If NULL, the genetic map in cross is used.
error_prob	Assumed genotyping error probability
<pre>map_function</pre>	Character string indicating the map function to use to convert genetic distances to recombination fractions.
lowmem	If FALSE, split individuals into groups with common sex and crossinfo and then precalculate the transition matrices for a chromosome; potentially a lot faster but using more memory.
quiet	If FALSE, print progress messages.
cores	Number of CPU cores to use, for parallel calculations. (If 0, use parallel::detectCores().) Alternatively, this can be links to a set of cluster sockets, as produced by parallel::makeCluster().

#### Details

We use a hidden Markov model to find, for each individual on each chromosome, the most probable sequence of underlying genotypes given the observed marker data.

Note that we break ties at random, and our method for doing this may introduce some bias.

Consider the results with caution; the most probable sequence can have very low probability, and can have features that are quite unusual (for example, the number of recombination events can be too small). In most cases, the results of a single imputation with sim\_geno() will be more realistic.

#### Value

An object of class "viterbi": a list of two-dimensional arrays of imputed genotypes, individuals x positions. Also contains three attributes:

- crosstype The cross type of the input cross.
- is\_x\_chr Logical vector indicating whether chromosomes are to be treated as the X chromosome or not, from input cross.
- alleles Vector of allele codes, from input cross.

# See Also

```
sim_geno(), maxmarg(), cbind.viterbi(), rbind.viterbi()
```

#### Examples

```
grav2 <- read_cross2(system.file("extdata", "grav2.zip", package="qtl2"))
map_w_pmar <- insert_pseudomarkers(grav2$gmap, step=1)
g <- viterbi(grav2, map_w_pmar, error_prob=0.002)</pre>
```

write\_control\_file Write a control file for QTL data

#### Description

Write the control file (in YAML or JSON) needed by read\_cross2() for a set of QTL data.

#### Usage

```
write_control_file(
  output_file,
  crosstype = NULL,
  geno_file = NULL,
  founder_geno_file = NULL,
  gmap_file = NULL,
  pmap_file = NULL,
  pheno_file = NULL,
  covar_file = NULL,
  phenocovar_file = NULL,
  sex_file = NULL,
  sex_covar = NULL,
  sex_codes = NULL,
  crossinfo_file = NULL,
  crossinfo_covar = NULL,
  crossinfo_codes = NULL,
  geno_codes = NULL,
  alleles = NULL,
  xchr = NULL,
```

```
sep = ",",
na.strings = c("-", "NA"),
comment.char = "#",
geno_transposed = FALSE,
founder_geno_transposed = FALSE,
pheno_transposed = FALSE,
covar_transposed = FALSE,
phenocovar_transposed = FALSE,
description = NULL,
comments = NULL,
overwrite = FALSE
```

# Arguments

output_file	File name (with path) of the YAML or JSON file to be created, as a character string. If extension is . json, JSON format is used; otherwise, YAML is used.		
crosstype	Character string with the cross type.		
geno_file	File name for genotype data.		
founder_geno_fi	le		
	File name for the founder genotype data.		
gmap_file	File name for genetic map.		
pmap_file	File name for the physical map.		
pheno_file	File name for the phenotype data.		
covar_file phenocovar_file	File name for the covariate data.		
<u>-</u>	File name for the phenotype covariate data (i.e., metadata about the phenotypes).		
sex_file	File name for the individuals' sex. (Specify just one of sex_file or sex_covar.)		
sex_covar	Column name in the covariate data that corresponds to sex. (Specify just one of sex_file or sex_covar.)		
sex_codes	Named vector of character strings specifying the encoding of sex. The names attribute should be the codes used in the data files; the values within the vector should be "female" and "male".		
crossinfo_file	File name for the cross_info data. (Specify just one of crossinfo_file or crossinfo_covar.)		
crossinfo_covar			
	Column name in the covariate data that corresponds to the cross_info data. (Specify just one of crossinfo_file or crossinfo_covar.)		
crossinfo_codes			
	In the case that there is a single cross info column (whether in a file or as a covariate), you can provide a named vector of character strings specifying the encoding of cross_info. The names attribute should be the codes used; the values within the vector should be the codes to which they will be converted (for example, $\emptyset$ and 1 for an intercross).		

geno_codes	Named vector specifying the encoding of genotypes. The names attribute has the codes used within the genotype and founder genotype data files; the values within the vector should be the integers to which the genotypes will be con- verted.	
alleles	Vector of single-character codes for the founder alleles.	
xchr	Character string with the ID for the X chromosome.	
sep	Character string that separates columns in the data files.	
na.strings	Vector of character strings with codes to be treated as missing values.	
comment.char	Character string that is used as initial character in a set of leading comment lines in the data files.	
geno_transpose	d	
	If TRUE, genotype file is transposed (with markers as rows).	
founder_geno_t	ransposed If TRUE, founder genotype file is transposed (with markers as rows).	
pheno_transposed		
	If TRUE, phenotype file is transposed (with phenotypes as rows).	
covar_transpos	ed	
	If TRUE, covariate file is transposed (with covariates as rows).	
phenocovar_tra	nsposed If TRUE, phenotype covariate file is transposed (with phenotype covariates as rows).	
description	Optional character string describing the data.	
comments	Vector of character strings to be inserted as comments at the top of the file (in the case of YAML), with each string as a line. For JSON, the comments are instead included within the control object.	
overwrite	If TRUE, overwrite file if it exists. If FALSE (the default) and the file exists, stop with an error.	

# Details

This function takes a set of parameters and creates the control file (in YAML or JSON format) needed for the new input data file format for R/qtl2. See the sample data files and the vignette describing the input file format.

# Value

(Invisibly) The data structure that was written.

# See Also

read\_cross2(), sample data files at https://kbroman.org/qtl2/pages/sampledata.html

### Examples

```
# Control file for the sample dataset, grav2
grav2_control_file <- file.path(tempdir(), "grav2.yaml")</pre>
write_control_file(grav2_control_file,
                   crosstype="riself",
                   geno_file="grav2_geno.csv",
                   gmap_file="grav2_gmap.csv",
                   pheno_file="grav2_pheno.csv",
                   phenocovar_file="grav2_phenocovar.csv",
                   geno_codes=c(L=1L, C=2L),
                   alleles=c("L", "C"),
                   na.strings=c("-", "NA"))
# Control file for the sample dataset, iron
iron_control_file <- file.path(tempdir(), "iron.yaml")</pre>
write_control_file(iron_control_file,
                   crosstype="f2",
                   geno_file="iron_geno.csv",
                   gmap_file="iron_gmap.csv",
                   pheno_file="iron_pheno.csv"
                   covar_file="iron_covar.csv",
                   phenocovar_file="iron_phenocovar.csv",
                   geno_codes=c(SS=1L, SB=2L, BB=3L),
                   sex_covar="sex",
                   sex_codes=c(f="female", m="male"),
                   crossinfo_covar="cross_direction",
                   crossinfo_codes=c("(SxB)x(SxB)"=0L, "(BxS)x(BxS)"=1L),
                   xchr="X",
                   alleles=c("S", "B"),
                   na.strings=c("-", "NA"))
# Remove these files, to clean up temporary directory
unlink(c(grav2_control_file, iron_control_file))
```

```
xpos_scan1
```

Get x-axis position for genomic location

#### Description

For a plot of scan1() results, get the x-axis location that corresponds to a particular genomic location (chromosome ID and position).

#### Usage

xpos\_scan1(map, chr = NULL, gap = NULL, thechr, thepos)

#### Arguments

map

A list of vectors of marker positions, as produced by insert\_pseudomarkers().

chr	Selected chromosomes that were plotted (if used in the call to plot_scan1().
gap	The gap between chromosomes used in the call to plot_scan1().
thechr	Vector of chromosome IDs
thepos	Vector of chromosomal positions

#### Details

thechr and thepos should be the same length, or should have length 1 (in which case they are expanded to the length of the other vector).

#### Value

A vector of x-axis locations.

```
# read data
iron <- read_cross2(system.file("extdata", "iron.zip", package="qtl2"))
# insert pseudomarkers into map</pre>
```

```
map <- insert_pseudomarkers(iron$gmap, step=1)</pre>
```

```
# calculate genotype probabilities
probs <- calc_genoprob(iron, map, error_prob=0.002)</pre>
```

```
# grab phenotypes and covariates; ensure that covariates have names attribute
pheno <- iron$pheno
covar <- match(iron$covar$sex, c("f", "m")) # make numeric
names(covar) <- rownames(iron$covar)
Xcovar <- get_x_covar(iron)</pre>
```

```
# perform genome scan
out <- scan1(probs, pheno, addcovar=covar, Xcovar=Xcovar)</pre>
```

```
# Use xpos_scan1 to add points at the peaks
# first find the peaks with LOD > 3
peaks <- find_peaks(out, map)</pre>
```

```
# keep just the peaks for chromosomes that were plotted
peaks <- peaks[peaks$chr %in% chr,]</pre>
```

```
# find x-axis positions
xpos <- xpos_scan1(map, chr=chr, thechr=peaks$chr, thepos=peaks$pos)</pre>
```

```
# point colors
ptcolor <- c("darkslateblue", "violetred")[match(peaks$lodcolumn, c("liver", "spleen"))]
# plot points</pre>
```

points(xpos, peaks\$lod, pch=21, bg=ptcolor)

zip\_datafiles Zip a set of data files

# Description

Zip a set of data files (in format read by read\_cross2()).

# Usage

```
zip_datafiles(control_file, zip_file = NULL, overwrite = FALSE, quiet = TRUE)
```

# Arguments

<pre>control_file</pre>	Character string with path to the control file (YAML or JSON) containing all of the control information.
zip_file	Name of zip file to use. If NULL, we use the stem of control_file but with a .zip extension.
overwrite	If TRUE, overwrite file if it exists. If FALSE (the default) and the file exists, stop with an error.
quiet	If FALSE, print progress messages.

#### Details

The input control\_file is the control file (in YAML or JSON format) to be read by read\_cross2(). (See the sample data files and the vignette describing the input file format.)

The utils::zip() function is used to do the zipping.

The files should all be contained within the directory where the control\_file sits, or in a subdirectory of that directory. If file paths use ..., these get stripped by zip, and so the resulting zip file may not work with read\_cross2().

### Value

Character string with the file name of the zip file that was created.

# See Also

read\_cross2(), sample data files at https://kbroman.org/qtl2/pages/sampledata.html

# zip\_datafiles

# Examples

```
## Not run:
zipfile <- file.path(tempdir(), "grav2.zip")
zip_datafiles("grav2.yaml", zipfile)
```

## End(Not run)

# Index

\* **IO** read\_cross2, 114 read\_pheno, 115 zip\_datafiles, 156 \* datasets CCcolors, 28 \* graphics plot\_compare\_geno, 85 \* hgraphics plot\_genes, 85 plot\_sdp, 99 \* htest chisq\_colpairs, 29 \* utilities basic\_summaries, 6 calc\_entropy, 11 calc\_errorlod, 12 calc\_genoprob, 13 calc\_grid, 16 calc\_kinship, 18 calc\_raw\_founder\_maf, 19 calc\_raw\_geno\_freq, 20 calc\_raw\_het, 21 calc\_raw\_maf, 21 compare\_geno, 34 convert2cross2, 37 est\_map, 47 find\_dup\_markers, 48 genoprob\_to\_alleleprob, 62 get\_x\_covar, 66 guess\_phase, 66 insert\_pseudomarkers, 69 map\_to\_grid, 75 max\_compare\_geno, 79 predict\_snpgeno, 103 print.cross2, 104 probs\_to\_grid, 105 recode\_snps, 116 scale\_kinship, 120

sim\_geno, 136 smooth\_gmap, 138 subset.calc\_genoprob, 139 subset.cross2, 140 subset.sim\_geno, 141 subset.viterbi, 142 summary.cross2, 144 summary\_compare\_geno, 144 unsmooth\_gmap, 149 write\_control\_file, 151 [.calc\_genoprob(subset.calc\_genoprob), 139 [.cross2(subset.cross2), 140 [.sim\_geno(subset.sim\_geno), 141 [.viterbi(subset.viterbi), 142 add\_threshold, 5 base::cbind(), 27 base::paste(), 76 base::print(), 104 basic\_summaries, 6, 144 batch\_cols, 8 batch\_cols(), 9 batch\_vec, 9  $batch_vec(), 8$ bayes\_int, 9 bayes\_int(), 56, 75 c.scan1perm(rbind.scan1perm), 111 calc\_entropy, 11 calc\_errorlod, 12 calc\_errorlod(), 103 calc\_geno\_freq, 15 calc\_geno\_freq(), 17, 20 calc\_genoprob, 13 calc\_genoprob(), 11-13, 15-18, 23, 32, 35, 62-64, 68, 70, 71, 78, 87, 88, 90, 105–107, 109, 120, 121, 124, 126, 129, 131, 134, 139

# INDEX

calc\_grid, 16 calc\_grid(), 70, 76, 106 calc\_het, 17 calc\_het(), 15, 21 calc\_kinship, 18 calc\_kinship(), *120*, *123* calc\_raw\_founder\_maf, 19 calc\_raw\_founder\_maf(), 21, 22, 116 calc\_raw\_geno\_freq, 20 calc\_raw\_geno\_freq(), 15, 21, 22 calc\_raw\_het, 21 calc\_raw\_het(), 17, 20, 22 calc\_raw\_maf, 21 calc\_raw\_maf(), 19-21, 116 calc\_sdp, 22 calc\_sdp(), 72, 100, 136 cbind.calc\_genoprob, 23 cbind.calc\_genoprob(), 109 cbind.scan1, 24 cbind.scan1(), 110, 123 cbind.scan1perm, 25 cbind.scan1perm(), 111 cbind.sim\_geno, 26 cbind.sim\_geno(), 112, 137 cbind.viterbi, 27 cbind.viterbi(), *113*, *151* cbind\_expand, 27 CCaltcolors (CCcolors), 28 CCcolors, 28, 84 CCorigcolors (CCcolors), 28 check\_cross2, 29 chisq\_colpairs, 29  $chr_lengths, 30$ chr\_lengths(), 131, 132 chr\_names (basic\_summaries), 6 clean, 31 clean.calc\_genoprob (clean\_genoprob), 31 clean.calc\_genoprob(), 31 clean.scan1 (clean\_scan1), 33 clean.scan1(), 31 clean\_genoprob, 31 clean\_scan1, 33 compare\_geno, 34 compare\_geno(), 79, 80, 85, 144, 145 compare\_genoprob, 35 compare\_maps, 36 convert2cross2, 37 count\_xo, 38

 $count_xo(), 73$ covar\_names(basic\_summaries), 6 create\_gene\_query\_func, 39 create\_snpinfo, 40 create\_variant\_query\_func, 41 create\_variant\_query\_func(), 134, 135 data.table::fread(), 60, 61 decomp\_kinship, 43 drop\_markers, 44 drop\_markers(), 45, 49, 108, 118 drop\_nullmarkers, 44 drop\_nullmarkers(), 44, 49, 108 est\_herit, 45 est\_map, 47 find\_dup\_markers, 48 find\_dup\_markers(), 44, 118 find\_ibd\_segments, 49 find\_index\_snp, 51 find\_index\_snp(), 53, 68 find\_map\_gaps, 52 find\_map\_gaps(), 117 find\_marker, 53 find\_marker(), 51, 54, 59, 107 find\_markerpos, 54 find\_markerpos(), 53 find\_peaks, 55 find\_peaks(), 11, 75, 81, 91, 94, 95 fit1, 57 fit1(), 107 founders(basic\_summaries), 6 fread\_csv, 60 fread\_csv(), 62 fread\_csv\_numer, 61 fread\_csv\_numer(), 61 genoprob\_to\_alleleprob, 62 genoprob\_to\_alleleprob(), 14, 18, 35, 87, 88.90 genoprob\_to\_snpprob, 63 genoprob\_to\_snpprob(), 41, 68, 101, 135, 147, 148  $\texttt{get\_common\_ids}, 65$ get\_x\_covar, 66 get\_x\_covar(), 76 graphics::image(), 88, 90 graphics::segments(), 5

```
guess_phase, 66
guess_phase(), 93
```

hist(), 85

locate\_xo, 73
locate\_xo(), 38
lod\_int, 74
lod\_int(), 11, 56

```
map_to_grid, 75
map_to_grid(), 16, 106
marker_names (basic_summaries), 6
mat2strata, 76
mat2strata(), 132
max.compare_geno (max_compare_geno), 79
max.scan1 (max_scan1), 80
max_compare_geno, 79
max_scan1, 80
maxlod, 77
maxmarg, 78
maxmarg(), 35, 38, 67, 73, 93, 96, 103, 151
```

```
n_chr (basic_summaries), 6
n_covar (basic_summaries), 6
n_founders (basic_summaries), 6
n_ind (basic_summaries), 6
n_ind_geno (basic_summaries), 6
n_ind_gnp (basic_summaries), 6
n_ind_pheno (basic_summaries), 6
n_mar (basic_summaries), 6
n_missing, 81
```

n\_pheno (basic\_summaries), 6 n\_phenocovar (basic\_summaries), 6 n\_typed (n\_missing), 81 parallel::detectCores(), 11, 12, 14, 17, 18, 20, 32, 34, 38, 43, 46, 47, 49, 56, 62, 67, 69, 70, 73, 78, 103, 118, 122, 125, 129, 131, 135, 137, 150 parallel::makeCluster(), 11, 12, 14, 17, 18, 20, 32, 34, 38, 43, 46, 47, 49, 56, 62, 67, 69, 70, 73, 78, 103, 118, 122, 125, 129, 131, 135, 137, 150 pheno\_names(basic\_summaries), 6 phenocovar\_names (basic\_summaries), 6 plot.calc\_genoprob (plot\_genoprob), 87 plot.compare\_geno (plot\_compare\_geno), 85 plot.scan1 (plot\_scan1), 98 plot.scan1coef(plot\_coef), 82 plot\_coef, 82 plot\_coef(), 97, 98, 102 plot\_coefCC (plot\_coef), 82 plot\_coefCC(), 98, 102 plot\_compare\_geno, 85 plot\_genes, 85 plot\_genoprob, 87 plot\_genoprob(), 90 plot\_genoprobcomp, 89 plot\_genoprobcomp(), 36, 88 plot\_lodpeaks, 91 plot\_lodpeaks(), 95 plot\_onegeno, 92 plot\_peaks, 94 plot\_peaks(), 91 plot\_pxg, 96 plot\_scan1, 98 plot\_scan1(), 84, 102, 155 plot\_sdp, 99 plot\_snpasso, 100 plot\_snpasso(), 84, 98, 135, 148 predict\_snpgeno, 103 print.cross2, 104 print.summary.compare\_geno (summary\_compare\_geno), 144 print.summary.scan1perm, 104 probs\_to\_grid, 105 probs\_to\_grid(), 16, 76 pull\_genoprobint, 106 pull\_genoprobint(), 53, 107

# INDEX

```
pull_genoprobpos, 107
pull_genoprobpos(), 53, 59, 107
pull_markers, 108
pull_markers(), 44, 45
qtl2version, 109
qtl::read.cross(), 37
rbind.calc_genoprob, 109
rbind.calc_genoprob(), 23
rbind.scan1, 110
rbind.scan1(), 24, 123
rbind.scan1perm, 111
rbind.scan1perm(), 25
rbind.sim_geno, 112
rbind.sim_geno(), 26, 137
rbind.viterbi, 113
rbind.viterbi(), 27, 151
read_cross2, 114
read_cross2(), 7, 29, 38, 63, 82, 104, 115,
         140, 144, 151, 153, 156
read_pheno, 115
read_pheno(), 114
recode_snps, 116
recode_snps(), 19-22
reduce_map_gaps, 117
reduce_map_gaps(), 52
reduce_markers, 117
reduce_markers(), 44, 49
replace_ids, 119
rug(), 85
scale_kinship, 120
scan1, 121
scan1(), 10, 11, 24, 25, 33, 55, 56, 74, 75, 77,
        80, 98, 101, 110, 111, 130, 132, 135,
         143, 147, 154
scan1blup, 124
scan1blup(), 58
scan1coef, 126
scan1coef(), 83, 123
scan1max, 128
scan1max(), 123
scan1perm, 130
scan1perm(), 25, 30, 76, 111, 123, 130, 145,
         146
scan1snps, 133
scan1snps(), 40, 41, 55, 64, 68, 80, 148
sdp2char, 136
```

sdp2char(), 22, 72 sim\_geno, 136 sim\_geno(), 26, 38, 79, 112, 120, 141, 150, 151 smooth\_gmap, 138 smooth\_gmap(), 149 subset.calc\_genoprob, 139 subset.cross2, 140 subset.scan1 (subset\_scan1), 143 subset.sim\_geno, 141 subset.viterbi, 142 subset\_scan1, 143 summary.compare\_geno (summary\_compare\_geno), 144 summary.compare\_geno(), 145 summary.cross2, 144 summary.cross2(), 8 summary.scan1perm (summary\_scan1perm), 145 summary\_compare\_geno, 144 summary\_scan1perm, 145 summary\_scan1perm(), 104, 105

top\_snps, 147
tot\_mar (basic\_summaries), 6

```
unsmooth_gmap, 149
unsmooth_gmap(), 138
utils::zip(), 156
```

```
viterbi, 150
viterbi(), 27, 38, 73, 79, 103, 113, 120, 142
```

```
write_control_file, 151
write_control_file(), 114
```

```
xpos_scan1, 154
```

```
zip_datafiles, 156
```