

Package ‘getspres’

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Title SPRE Statistics for Exploring Heterogeneity in Meta-Analysis

Version 0.2.0

Description An implementation of SPRE (standardised predicted random-effects) statistics in R to explore heterogeneity in genetic association meta-analyses, as described by Magosi et al. (2019) <doi:10.1093/bioinformatics/btz590>. SPRE statistics are precision weighted residuals that indicate the direction and extent with which individual study-effects in a meta-analysis deviate from the average genetic effect. Overly influential positive outliers have the potential to inflate average genetic effects in a meta-analysis whilst negative outliers might lower or change the direction of effect. See the 'getspres' website for documentation and examples <<https://magosil86.github.io/getspres/>>.

Depends R (>= 3.1.0)

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URL <https://magosil86.github.io/getspres/>

BugReports <https://github.com/magosil86/getspres/issues>

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Author Lerato E Magosi [aut],
Jemma C Hopewell [aut],
Martin Farrall [aut],
Lerato E Magosi [cre]

Maintainer Lerato E Magosi <magosil86@gmail.com>

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getspres	<i>Exploring Heterogeneity in Meta-Analysis with SPRE Statistics.</i>
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Description

getspres computes *SPRE* (standardised predicted random-effects) statistics to identify outlier studies in genetic association meta-analyses which might have undue influence on the average genetic effect leading to inflated genetic signals.

Usage

```
getspres(beta_in, se_in, study_names_in, variant_names_in, ...)
```

```
## Default S3 method:
getspres(
  beta_in,
  se_in,
  study_names_in,
  variant_names_in,
  tau2_method = "DL",
  verbose_output = FALSE,
  ...
)
```

Arguments

beta_in	A numeric vector of study effect-sizes e.g. log odds-ratios.
se_in	A numeric vector of standard errors, genomically corrected at study-level.
study_names_in	A character vector of study names.
variant_names_in	A character vector of variant names e.g. rsIDs.
...	other arguments.
tau2_method	A character scalar, specifying the method that should be used to estimate heterogeneity either through DerSimonian and Laird's moment-based estimate "DL" or restricted maximum likelihood "REML". Note: The REML method uses the iterative Fisher scoring algorithm (step length = 0.5, maximum iterations = 10000) to estimate tau2. Default is "DL".
verbose_output	An optional boolean to display intermediate output. (Default is FALSE).

Details

SPRE statistics are precision-weighted residuals that summarise the direction and extent with which observed study effects in a meta-analysis differ from the summary (or average genetic) effect. See the getspres website for more information, documentation and examples.

getspres takes as input study effect-size estimates and their corresponding standard errors (i.e. summary data). Study effect estimates could be in the form of linear regression coefficients or log-transformed regression coefficients (per-allele log odds ratios) from logistic regression.

getspres uses inverse-variance weighted meta-analysis models in the metafor R package to calculate *SPRE* statistics.

Value

Returns a list containing:

- number_variants A numeric scalar for the number of variants
- number_studies A numeric scalar for the number of studies
- spre_dataset A dataframe that is a dataset of computed *SPRE* statistics and contains the following fields:
 - beta , study effect-size estimates
 - se , corresponding standard errors of the study effect-size estimates
 - variant_names , variant names
 - study_names , study names
 - study , study numbers
 - snp , snp numbers
 - tau2 , tau_squared, estimate of amount of between-study variance
 - I2 , I_squared, heterogeneity index (Higgins inconsistency metric) representing proportion of total observed variation due to between-study variance
 - Q , Q-statistic (Cochran's Q)
 - xb , prediction excluding random effects
 - xbse , standard error of prediction excluding random effects
 - xbu , predictions including random effects
 - stdxbu , standard error of prediction (fitted values) including random effects
 - hat , leverage a.k.a diagonal elements of the projection hat matrix
 - rawresid , raw residuals
 - uncondse , unconditional standard errors
 - spre , *SPRE* statistics (standardised predicted random effects) i.e. raw residuals divided by the unconditional standard errors

Methods (by class)

- default: Computes *SPRE* statistics in genetic association meta-analyses

See Also

<https://magosil86.github.io/getspres/> to the visit getspres website.

Examples

```
library(getspres)

# Calculate SPRE statistics for a subset of variants in the heartgenes214 dataset.
# heartgenes214 is a case-control GWAS meta-analysis of coronary artery disease.
# To learn more about the heartgenes214 dataset ?heartgenes214

# Calculating SPRE statistics for 3 variants in heartgenes214

heartgenes3 <- subset(heartgenes214,
  variants %in% c("rs10139550", "rs10168194", "rs11191416"))

getspres_results <- getspres(beta_in = heartgenes3$beta_flipped,
  se_in = heartgenes3$gcse,
  study_names_in = heartgenes3$studies,
  variant_names_in = heartgenes3$variants)

# Explore results generated by the getspres function
str(getspres_results)

# Retrieve number of studies and variants
getspres_results$number_variants
getspres_results$number_studies

# Retrieve SPRE dataset
df_spres <- getspres_results$spre_dataset
head(df_spres)

# Extract SPREs from SPRE dataset
head(spres <- df_spres[, "spre"])

# Exploring available options in the getspres function:
# 1. Estimate heterogeneity using "REML", default is "DL"
# 2. Calculate SPRE statistics verbosely
getspres_results <- getspres(beta_in = heartgenes3$beta_flipped,
  se_in = heartgenes3$gcse,
  study_names_in = heartgenes3$studies,
  variant_names_in = heartgenes3$variants,
  tau2_method = "REML",
  verbose_output = TRUE)

str(getspres_results)
```

heartgenes214	<i>heartgenes214.</i>
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Description

heartgenes214 is a multi-ethnic GWAS meta-analysis dataset for coronary artery disease.

Usage

heartgenes214

Format

A data frame with seven variables:

beta_flipped Effect-sizes expressed as log odds ratios. Numeric

gcse Standard errors

studies Names of participating studies

variants Names of genetic variants/SNPs

cases Number of cases in each participating study

controls Number of controls in each participating study

fdr214_gwas46 Flag indicating GWAS significant variants, 1: Not GWAS-significant, 2: GWAS-significant

Details

It comprises summary data (effect-sizes and their corresponding standard errors) for 48 studies (68,801 cases and 123,504 controls), at 214 lead variants independently associated with coronary artery disease ($P < 0.00005$, $FDR < 5\%$). Of the 214 lead variants, 44 are genome-wide significant ($p < 5e-08$). The meta-analysis dataset is based on individuals of: African American, Hispanic American, East Asian, South Asian, Middle Eastern and European ancestry.

The study effect-sizes have been flipped to ensure alignment of the effect alleles.

Standard errors were genomically corrected at the study-level.

Source

Magosi LE, Goel A, Hopewell JC, Farrall M, on behalf of the CARDIoGRAMplusC4D Consortium (2017) Identifying systematic heterogeneity patterns in genetic association meta-analysis studies. *PLoS Genet* 13(5): e1006755. <https://doi.org/10.1371/journal.pgen.1006755>.

<https://magosil86.github.io/getmstatistic/>

plotspres

plotspres *generates forest plots showing SPRE statistics.*

Description

Forest plots showing *SPRE* (standardised predicted random-effects) statistics can be useful in highlighting overly influential outlier studies with the potential to inflate summary effect estimates in genetic association meta-analyses.

Usage

```
plotspres(beta_in, se_in, study_names_in, variant_names_in, spres_in, ...)
```

```
## Default S3 method:
```

```
plotspres(
  beta_in,
  se_in,
  study_names_in,
  variant_names_in,
  spres_in,
  spre_colour_palette = c("mono_colour", "black"),
  set_studyNOs_as_studyIDs = FALSE,
  set_study_field_width = "%02.0f",
  set_cex = 0.66,
  set_xlim,
  set_ylim,
  set_at,
  tau2_method = "DL",
  adjust_labels = 1,
  save_plot = TRUE,
  verbose_output = FALSE,
  ...
)
```

Arguments

beta_in	A numeric vector of observed study effects e.g. log odds-ratios.
se_in	A numeric vector of standard errors, genomically corrected at study-level.
study_names_in	A character vector of study names.
variant_names_in	A character vector of variant names e.g. rsIDs.
spres_in	A numeric vector of <i>SPRE</i> statistics.
...	other arguments.

<code>spre_colour_palette</code>	An optional character vector specifying the colour palette that should be used for observed study effects. There are 3 types of colour palettes available, namely: "mono_colour", "dual_colour" and "multi_colour"; with the "dual_colour" palette, observed study effects with negative <i>SPRE</i> statistics are coloured differently from those with positive <i>SPRE</i> statistics, and with the "multi_colour" palette observed study effects are colored in a gradient according to the <i>SPRE</i> statistic values. Default palette option is <code>spre_colour_palette = c("mono_colour", "black")</code> .
<code>set_studyNOs_as_studyIDs</code>	An optional boolean specifying whether study numbers should be used as study IDs in the forest plot. Default is FALSE.
<code>set_study_field_width</code>	An optional character vector of format strings, akin to the <code>fmt</code> character vector in the <code>sprintf</code> function. (Default is <code>set_study_field_width = "%02.0f"</code>).
<code>set_cex</code>	An optional character scalar and symbol expansion factor indicating the percentage by which text and symbols should be scaled relative to the reference; e.g. 1=reference, 1.3 is 30% larger, 0.3 is 30% smaller. (Default is <code>cex = 0.66</code>).
<code>set_xlim</code>	An optional numeric vector of length 2 indicating the horizontal limits of the plot region.
<code>set_ylim</code>	An optional numeric vector of length 2 indicating the y-axis limits of the plot.
<code>set_at</code>	An optional numeric vector indicating position of the x-axis tick marks and corresponding labels.
<code>tau2_method</code>	An optional character scalar, specifying the method that should be used to estimate heterogeneity either through DerSimonian and Laird's moment-based estimate "DL" or restricted maximum likelihood "REML". Note: The REML method uses the iterative Fisher scoring algorithm (step length = 0.5, maximum iterations = 10000) to estimate tau2. Default is "DL".
<code>adjust_labels</code>	An optional numeric scalar value that tweaks label (column header) positions. (Default is <code>adjust_labels = 1</code>).
<code>save_plot</code>	An optional boolean to save forestplot as a tiff file. Default is TRUE.
<code>verbose_output</code>	An optional boolean to display intermediate output. (Default is FALSE).

Details

`plotspres` takes as input *SPRE* statistics, observed study effects and corresponding standard errors (i.e. summary data). The observed study effects (i.e. study effect-size estimates) could be association statistics from either quantitative or binary trait meta-analyses, for instance, linear regression coefficients might be employed for quantitative traits and log-transformed logistic regression coefficients (per-allele log odds ratios) used for case-control meta-analyses. *SPRE* statistics can be calculated using the `getspres` function.

`plotspres` uses inverse-variance weighted fixed and random-effects meta-analysis models in the `metafor` R package to generate forestplots.

Value

Returns a list containing:

- `number_variants` A numeric scalar indicating the number of variants
- `number_studies` A numeric scalar indicating the number of studies
- `fixed_effect_results` A list of fixed-effect meta-analysis results for each variant examined
- `random_effects_results` A list of random-effects meta-analysis results for each variant examined
- `spre_forestplot_dataset` A dataframe of the data provided by the user for analysis which contains the following fields:
 - `beta` , study effect-size estimates
 - `se` , corresponding standard errors of study effect-size estimates
 - `variant_names` , variant names
 - `study_names` , study names
 - `spre` , *SPRE* (standardised predicted random-effects) statistics
 - `study_numbers` , study numbers
 - `variant_numbers` , variant numbers

Methods (by class)

- `default`: Generates forest plots showing *SPRE* statistics

See Also

`getspres` to calculate *SPRE* statistics and the `metafor` package to explore implementations of fixed and random-effects meta-analysis models in R. To access more information and examples visit the `getspres` website at: <https://magosil86.github.io/getspres/>.

Examples

```
library(getspres)

# Generate a forest plot showing SPRE statistics for variants in heartgenes214.
# heartgenes214 is a case-control GWAS meta-analysis of coronary artery disease.
# To learn more about the heartgenes214 dataset ?heartgenes214

# Calculating SPRE statistics for 3 variants in heartgenes214

heartgenes3 <- subset(heartgenes214,
  variants %in% c("rs10139550", "rs10168194", "rs11191416"))

getspres_results <- getspres(beta_in = heartgenes3$beta_flipped,
  se_in = heartgenes3$gcse,
  study_names_in = heartgenes3$studies,
  variant_names_in = heartgenes3$variants)

# Explore results generated by the getspres function
str(getspres_results)

# Retrieve number of studies and variants
getspres_results$number_variants
```



```
getspres_results$number_studies

# Retrieve SPRE dataset
df_spres <- getspres_results$spre_dataset
head(df_spres)

# Extract SPREs from SPRE dataset
head(spres <- df_spres[, "spre"])

# Generating forest plots showing SPREs for variants in heartgenes3

# Forest plot with default settings
# Tip: To store plots set save_plot = TRUE (useful when generating multiple plots)
plotspres_res <- plotspres(beta_in = df_spres$beta,
                          se_in = df_spres$se,
                          study_names_in = as.character(df_spres$study_names),
                          variant_names_in = as.character(df_spres$variant_names),
                          spres_in = df_spres$spre,
                          save_plot = FALSE)

# Explore results generated by the plotspres function

# Retrieve number of studies and variants
plotspres_res$number_variants
plotspres_res$number_studies

# Retrieve fixed and random-effects meta-analysis results
fixed_effect_res <- plotspres_res$fixed_effect_results
random_effects_res <- plotspres_res$random_effects_results

# Retrieve dataset that was used to generate forest plots
df_plotspres <- plotspres_res$spre_forestplot_dataset

# Retrieve more detailed meta-analysis output
str(plotspres_res)

# Explore available options for plotspres forest plots:
# 1. Colorize study-effect estimates according to SPRE statistic values
# 2. Label studies by study number instead of study names
# 3. Format study labels (useful when using study numbers as study labels)
# 4. Change text size
# 5. Adjust x and y axes limits
# 6. Change method used to estimate amount of heterogeneity from "DL" to "REML"
# 7. Run verbosely to show intermediate results
# 8. Adjust label (i.e. column header) positions
# 9. Save plot as a tiff file (useful when generating multiple plots)

# Colorize study-effect estimates according to SPRE statistic values
```



```
        set_cex = 0.75, set_xlim = c(-2,2), set_ylim = c(-1.5,51),
        set_at = c(-0.6, -0.4, -0.2, 0.0, 0.2, 0.4, 0.6),
tau2_method = "REML", verbose_output = TRUE,
        save_plot = FALSE)

# Adjust label (i.e. column header) position, also keep plot in graphics window rather
#   than save as tiff file
df_rs10139550_3studies <- subset(df_rs10139550, as.numeric(df_rs10139550$study_names) <= 3)

# Before adjusting label positions
plotspres_res <- plotspres(beta_in = df_rs10139550_3studies$beta,
                          se_in = df_rs10139550_3studies$se,
                          study_names_in = as.character(df_rs10139550_3studies$study_names),
                          variant_names_in = as.character(df_rs10139550_3studies$variant_names),
                          spres_in = df_rs10139550_3studies$spres,
                          spre_colour_palette = c("dual_colour", c("blue","black")),
                          save_plot = FALSE)

# After adjusting label positions
plotspres_res <- plotspres(beta_in = df_rs10139550_3studies$beta,
                          se_in = df_rs10139550_3studies$se,
                          study_names_in = as.character(df_rs10139550_3studies$study_names),
                          variant_names_in = as.character(df_rs10139550_3studies$variant_names),
                          spres_in = df_rs10139550_3studies$spres,
                          spre_colour_palette = c("dual_colour", c("blue","black")),
                          adjust_labels = 1.7, save_plot = FALSE)
```

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