Package 'GEint'

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Type Package

Title Misspecified Models for Gene-Environment Interaction

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Description

The first major functionality is to compute the bias in misspecified linear gene-environment interaction models. The most generalized function for this objective is GE_bias(). However GE_bias() requires specification of many higher order moments of covariates in the model. If users are unsure about how to calculate/estimate these higher order moments, it may be easier to use GE_bias_normal_squaredmis(). This function places many more assumptions on the covariates (most notably that they are all jointly generated from a multivariate normal distribution) and is thus able to automatically calculate many of the higher order moments automatically, necessitating only that the user specify some covariances. There are also functions to solve for the bias through simulation and nonlinear equation solvers, these can be used to check your work. Second major functionality is to implement the Bootstrap Inference with Correct Sandwich (BICS) testing procedure, which we have found to provide better finite-sample performance than other inference procedures for testing GxE interaction. More details on these functions are available in Sun, Carroll, Christiani, and Lin (2018) <doi:10.1111/biom.12813>.

Imports mvtnorm, bindata, nleqslv, pracma, speedglm, rje, geepack,

stats

License GPL-3

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Suggests knitr, rmarkdown, testthat

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GE_bias

GE_bias.R

Description

A function to calculate the bias in testing for GxE interaction.

Usage

GE_bias(beta_list, cov_list, cov_mat_list, mu_list, HOM_list)

beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If G or Z or M is a vector, then beta_G/beta_Z/beta_M should be vectors. If Z and/or M/W do not exist in your model, then set beta_Z and/or beta_M = 0.
cov_list	A list of expectations (which happen to be covariances if all covariates are centered at 0) in the order specified by GE_enumerate_inputs(). If Z and/or M/W do not exist in your model, then treat them as constants 0. For example, if Z doesn't exist and W includes 2 covariates, then set $cov(EZ) = 0$ and $cov(ZW) = (0,0)$. If describing expectations relating two vectors, i.e. Z includes two covariates and W includes three covariates, sort by the first term and then the second. Thus in the example, the first three terms of $cov(ZW)$ are $cov(Z_1,W_1),cov(Z_1,W_2),cov(Z_1,W_3)$, and the last three terms are $cov(Z_3,W_1),cov(Z_3,W_2),cov(Z_3,W_3)$.
cov_mat_list	A list of matrices of expectations as specified by GE_enumerate_inputs().
mu_list	A list of means as specified by GE_enumerate_inputs().
HOM_list	A list of higher order moments as specified by GE_enumerate_inputs().

A list of the fitted coefficients alpha

Examples

```
solutions <- GE_bias_normal_squaredmis( beta_list=as.list(runif(n=6, min=0, max=1)),
rho_list=as.list(rep(0.3,6)), prob_G=0.3, cov_Z=1, cov_W=1)
GE_bias(beta_list=solutions$beta_list, solutions$cov_list, solutions$cov_mat_list,
solutions$mu_list, solutions$HOM_list)
```

GE_bias_normal_squaredmis

GE_bias_normal_squaredmis.R

Description

A function to calculate the bias in testing for GxE interaction, making many more assumptions than GE_bias(). The additional assumptions are added to simplify the process of calculating/estimating many higher order moments which the user may not be familiar with.

The following assumptions are made:

(1) All fitted covariates besides G (that is, E, all Z, and all W) have a marginal standard normal distribution with mean 0 and variance 1. This corresponds to the case of the researcher standardizing all of their fitted covariates.

(2) All G are generated by means of thresholding two independent normal RVs and are centered to have mean 0.

(3) The joint distributions of E, Z, W, and the thresholded variables underlying G can be described by a multivariate normal distribution.

(4) The misspecification is of the form $f(E)=h(E)=E^2$, and $M_j=W_j^2$ for all j. In particular, W always has the same length as M here.

Usage

```
GE_bias_normal_squaredmis(beta_list, rho_list, prob_G, cov_Z = NULL,
    cov_W = NULL, corr_G = NULL)
```

beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_beta_I, beta_Z, beta_M. If G or Z or M is a vector, then beta_G/beta_Z/beta_should be vectors. If Z and/or M/W do not exist in your model, then set beta and/or beta $M = 0$.	
rho_list	A list of expectations (which happen to be covariances if all covariates are centered at 0) in the order specified by GE_enumerate_inputs(). If Z and/or M/W do not exist in your model, then treat them as constants 0. For example, if Z doesn't exist and W includes 2 covariates, then set $cov(EZ) = 0$ and	

	$cov(ZW) = (0,0)$. If describing expectations relating two vectors, i.e. Z includes two covariates and W includes three covariates, sort by the first term and then the second. Thus in the example, the first three terms of $cov(ZW)$ are $cov(Z_1,W_1),cov(Z_1,W_2), cov(Z_1,W_3)$, and the last three terms are $cov(Z_3,W_1), cov(Z_3,W_2), cov(Z_3,W_3)$.
prob_G	Probability that each allele is equal to 1. Since each SNP has two alleles, the expectation of G is 2*prob_G. Should be a d*1 vector.
cov_Z	Should be a matrix equal to cov(Z) or NULL if no Z.
cov_W	Should be a matrix equal to cov(W) or NULL if no W.
corr_G	Should be a matrix giving the *pairwise correlations* between each SNP in the set, or NULL. Must be specified if G is a vector. For example, the [2,3] element of the matrix would be the pairwise correlation between SNP2 and SNP3.

A list with the elements:

alpha_list	The asymptotic values of the fitted coefficients alpha.	
beta_list	The same beta_list that was given as input.	
cov_list	The list of all covariances (both input and calculated) for use with GE_nleqslv() and GE_bias().	
mu_list	List of calculated means for f(E), h(E), Z, M, and W for use with GE_nleqslv() and GE_bias().	
HOM_list	$List \ of \ calculated \ Higher \ Order \ Moments \ for \ use \ with \ GE_nleqslv() \ and \ GE_bias().$	

Examples

GE_bias_normal_squaredmis(beta_list=as.list(runif(n=6, min=0, max=1)), rho_list=as.list(rep(0.3,6)), cov_Z=1, cov_W=1, prob_G=0.3)

GE_bias_normal_squaredmis_old

GE_bias_normal_squaredmis_old.R

Description

A function to calculate the bias in testing for GxE interaction, making many more assumptions than GE_bias_old(). The additional assumptions are added to simplify the process of calculating/estimating many higher order moments which the user may not be familiar with. The following assumptions are made:

(1) All fitted covariates besides G (that is, E, all Z, and all W) have a marginal standard normal distribution with mean 0 and variance 1. This corresponds to the case of the researcher standardizing all of their fitted covariates.

(2) G is generated by means of thresholding two independent normal RVs and is centered to have mean 0. (3) The joint distributions of E, Z, W, and the thresholded variables underlying G can be

described by a multivariate normal distribution.

(4) The misspecification is of the form $f(E)=h(E)=E^2$, and $M_j=W_j^2$ for all j. In particular, W always has the same length as M here.

Usage

Arguments

beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If Z or M is a vector, then beta_Z and beta_M should be vectors. If Z or M is not in the model (i.e. all covariates other than G+E have been specified incorrectly, or all covariates other than G+E have been specified correctly, or the only covariates are G+E), then set beta_Z=0 and/or beta_M=0.
rho_list	A list of the 6 pairwise covariances between the covariates. These should be in the order (1) cov_GE (2) cov_GZ (3) cov_EZ (4) cov_GW (5) cov_EW (6) cov_ZW. Again if Z or W are vectors then terms like cov_GZ should be vectors (in the order cov(G,Z_1),,cov(G,Z_p)) where Z is of dimension p, and simi- larly for W. If Z or M are vectors, then cov_ZW should be a vector in the order (cov(Z_1,W_1),,cov(Z_1,W_q), cov(Z_2,W_1),,cov(Z_p,W_q) where Z is a vector of length p and W is a vector of length q. If Z or M are not in the model then treat them as the constant 0. So for example if Z is not in the model and M (and therefore W) is a vector of length 2, we would have cov_EZ=0 and cov(ZW) = (0,0).
prob_G	Probability that each allele is equal to 1. Since each SNP has two alleles, the expectation of G is $2*$ prob_G.
cov_Z	Only specify this if Z is a vector, gives the covariance matrix of Z (remember by assumption Z has mean 0 and variance 1). The (i,j) element of the matrix should be the $(i-1)(i-2)/2+j$ element of the vector.
cov_W	Only specify this if W is a vector, gives the covariance matrix of W (remember by assumption W has mean 0 and variance 1). The (i,j) element of the matrix should be the $(i-1)(i-2)/2+j$ element of the vector.

Value

A list with the elements:

alpha_list	The asymptotic values of the fitted coefficients alpha.	
beta_list	The same beta_list that was given as input.	
cov_list	The list of all covariances (both input and calculated) for use with GE_nleqslv() and GE_bias().	
cov_mat_list	List of additionally calculated covariance matrices for use with GE_nleqslv() and GE_bias().	

mu_list	List of calculated means for f(E), h(E), Z, M, and W for use with GE_nleqslv() and GE_bias().
HOM_list	$List of calculated Higher Order Moments for use with GE_nleqslv() and GE_bias().$

Examples

```
GE_bias_normal_squaredmis_old( beta_list=as.list(runif(n=6, min=0, max=1)),
rho_list=as.list(rep(0.3,6)), prob_G=0.3)
```

GE_bias_old GE_bias_old.R

Description

A function to calculate the bias in testing for GxE interaction.

Usage

```
GE_bias_old(beta_list, cov_list, cov_mat_list, mu_list, HOM_list)
```

Arguments

beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If Z or M is a vector, then beta_Z and beta_M should be vectors. If Z and/or M/W do not exist in your model, then set beta_Z and/or beta_M = 0.
cov_list	A list of expectations (which happen to be covariances if all covariates are centered at 0) in the order specified by GE_enumerate_inputs_old(). If Z and/or M/W do not exist in your model, then treat them as constants 0. For example, if Z doesn't exist and W includes 2 covariates, then set $cov(EZ) = 0$ and $cov(ZW) = (0,0)$.
cov_mat_list	A list of matrices of expectations as specified by GE_enumerate_inputs_old().
mu_list	A list of means as specified by GE_enumerate_inputs_old().
HOM_list	A list of higher order moments as specified by GE_enumerate_inputs_old().

Value

A list of the fitted coefficients alpha

Examples

```
solutions <- GE_bias_normal_squaredmis_old( beta_list=as.list(runif(n=6, min=0, max=1)),
rho_list=as.list(rep(0.3,6)), prob_G=0.3)
GE_bias_old(beta_list=solutions$beta_list, solutions$cov_list, solutions$cov_mat_list,
solutions$mu_list, solutions$HOM_list)
```

GE_BICS

Description

A function to perform inference on the GxE interaction regression coefficient. Shows better small sample performance than comparable methods.

Usage

```
GE_BICS(outcome, design_mat, num_boots = 1000, desired_coef,
    outcome_type, check_singular = FALSE)
```

Arguments

outcome	The outcome vector
design_mat	The design matrix of covariates
num_boots	The number of bootstrap resamples to perform - we suggest 1000
desired_coef	The column in the design matrix holding the interaction covariate
outcome_type	Either 'D' for dichotomous outcome or 'C' for continuous outcome
check_singular	Make sure the design matrix can be inverted for variance estimation

Value

The p-value for the interaction effect

Examples

```
E <- rnorm(n=500)
G <- rbinom(n=500, size=2, prob=0.3)
design_mat <- cbind(1, G, E, G*E)
outcome <- rnorm(500)
GE_BICS(outcome=outcome, design_mat=design_mat, desired_coef=4, outcome_type='C')</pre>
```

GE_BICS_old GE_BICS_old.R

Description

A function to perform inference on the GxE interaction regression coefficient. Shows better small sample performance than comparable methods.

Usage

```
GE_BICS_old(outcome, design_mat, num_boots = 1000, desired_coef,
    outcome_type)
```

Arguments

outcome	The outcome vector
design_mat	The design matrix of covariates
num_boots	The number of bootstrap resamples to perform - we suggest 1000
desired_coef	The column in the design matrix holding the interaction covariate
outcome_type	Either 'D' for dichotomous outcome or 'C' for continuous outcome

Value

The p-value for the interaction effect

Examples

```
E <- rnorm(n=500)
G <- rbinom(n=500, size=2, prob=0.3)
design_mat <- cbind(1, G, E, G*E)
outcome <- rnorm(500)
GE_BICS_old(outcome=outcome, design_mat=design_mat, desired_coef=4, outcome_type='C')</pre>
```

GE_enumerate_inputs GE_enumerate_inputs.R

Description

Call this function to display the necessary inputs for GE_bias_set. For terms in cov_mat_list like E[ZZ], these should be matrices where the (i,j) element is $E[Z_i^*Z_j]$.

Usage

```
GE_enumerate_inputs()
```

Value

Nothing

Examples

GE_enumerate_inputs()

GE_enumerate_inputs_old

GE_enumerate_inputs_old.R

Description

Call this function to display the necessary inputs for GE_bias. If you see a term like mu_Gf, that means it is the scalar $E[G^{*}f(E)]$. If you see a term like MU_GM, that means it is the vector $c(E[G^{*}M_{1}], E[G^{*}M_{2}], ..., E[G^{*}M_{q}])$ where M is of dimension q. For terms in cov_mat_list like E[ZZ], these should be matrices where the (i,j) element is $E[Z_{i}^{*}Z_{j}]$.

Usage

```
GE_enumerate_inputs_old()
```

Value

Nothing

Examples

GE_enumerate_inputs_old()

GE_nleqslv	GE_nleqslv.R #' Uses package nleqslv to get a numerical solution to
	the score equations, which we can use to check our direct solution from
	$GE_bias().$

Description

GE_nleqslv.R #' Uses package nleqslv to get a numerical solution to the score equations, which we can use to check our direct solution from GE_bias().

Usage

```
GE_nleqslv(beta_list, cov_list, cov_mat_list, mu_list, HOM_list)
```

Arguments

beta_list A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If G or Z or M is a vector, then beta_G/beta_Z/beta_M should be vectors. If Z and/or M/W do not exist in your model, then set beta_Z and/or beta_M = 0.

cov_list	A list of expectations (which happen to be covariances if all covariates are centered at 0) in the order specified by GE_enumerate_inputs(). If Z and/or M/W do not exist in your model, then treat them as constants 0. For example, if Z doesn't exist and W includes 2 covariates, then set $cov(EZ) = 0$ and $cov(ZW) = (0,0)$. If describing expectations relating two vectors, i.e. Z includes two covariates and W includes three covariates, sort by the first term and then the second. Thus in the example, the first three terms of $cov(ZW)$ are $cov(Z_1,W_1), cov(Z_1,W_2), cov(Z_1,W_3)$, and the last three terms are
cov_mat_list	cov(Z_3,W_1), cov(Z_3,W_2), cov(Z_3,W_3). A list of matrices of expectations as specified by GE_enumerate_inputs().
mu_list	A list of means as specified by GE_enumerate_inputs().
HOM_list	A list of higher order moments as specified by GE_enumerate_inputs().

A list of the fitted coefficients alpha

Examples

```
solutions <- GE_bias_normal_squaredmis( beta_list=as.list(runif(n=6, min=0, max=1)),
rho_list=as.list(rep(0.3,6)), prob_G=0.3, cov_Z=1, cov_W=1)
GE_nleqslv(beta_list=solutions$beta_list, solutions$cov_list, solutions$cov_mat_list,
solutions$mu_list, solutions$HOM_list)
```

GE_nleqslv_old	GE_nleqslv_old.R #' Uses package nleqslv to get a numerical solution
	to the score equations, which we can use to check our direct solution
	from GE_bias_old().

Description

 $GE_nleqslv_old.R$ #' Uses package nleqslv to get a numerical solution to the score equations, which we can use to check our direct solution from $GE_bias_old()$.

Usage

```
GE_nleqslv_old(beta_list, cov_list, cov_mat_list, mu_list, HOM_list)
```

beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If Z or M is a vector, then beta_Z and beta_M should be vectors. If Z and/or M/W do not exist in your model, then set beta_Z and/or beta_M = 0.
cov_list	A list of expectations (which happen to be covariances if all covariates are cen- tered at 0) in the order specified by GE_enumerate_inputs_old(). If Z and/or M/W do not exist in your model, then treat them as constants 0. For example, set $cov(EZ) = 0$ and $cov(ZW) = 0$.

GE_scoreeq_sim

cov_mat_list	A list of matrices of expectations as specified by GE_enumerate_inputs_old().
mu_list	A list of means as specified by GE_enumerate_inputs_old().
HOM_list	A list of higher order moments as specified by GE_enumerate_inputs_old().

Value

A list of the fitted coefficients alpha

Examples

```
solutions <- GE_bias_normal_squaredmis_old( beta_list=as.list(runif(n=6, min=0, max=1)),
rho_list=as.list(rep(0.3,6)), prob_G=0.3)
GE_nleqslv_old(beta_list=solutions$beta_list, solutions$cov_list, solutions$cov_mat_list,
solutions$mu_list, solutions$HOM_list)
```

GE_scoreeq_sim GE_scoreeq_sim.R

Description

Here we perform simulation to verify that we have solved for the correct alpha values in GE_bias_norm_squaredmis(). Make the same assumptions as in GE_bias_norm_squaredmis().

Usage

GE_scoreeq_sim(num_sims = 5000, num_sub = 2000, beta_list, rho_list, prob_G, cov_Z = NULL, cov_W = NULL, corr_G = NULL)

num_sims	The number of simulations to run, we suggest 5000.
num_sub	The number of subjects to generate in every simulation, we suggest 2000.
beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If G or Z or M is a vector, then beta_G/beta_Z/beta_M should be vectors. If Z and/or M/W do not exist in your model, then set beta_Z and/or beta_M = 0.
rho_list	A list of expectations (which happen to be covariances if all covariates are centered at 0) in the order specified by GE_enumerate_inputs(). If Z and/or M/W do not exist in your model, then treat them as constants 0. For example, if Z doesn't exist and W includes 2 covariates, then set $cov(EZ) = 0$ and $cov(ZW) = (0,0)$. If describing expectations relating two vectors, i.e. Z includes two covariates and W includes three covariates, sort by the first term and then the second. Thus in the example, the first three terms of $cov(ZW)$ are $cov(Z_1,W_1),cov(Z_1,W_2), cov(Z_1,W_3)$, and the last three terms are $cov(Z_3,W_1), cov(Z_3,W_2), cov(Z_3,W_3)$.

prob_G	Probability that each allele is equal to 1. Since each SNP has two alleles, the expectation of G is 2*prob_G. Should be a d*1 vector.
cov_Z	Should be a matrix equal to cov(Z) or NULL if no Z.
cov_W	Should be a matrix equal to cov(W) or NULL if no W.
corr_G	Should be a matrix giving the *pairwise correlations* between each SNP in the set, or NULL. Must be specified if G is a vector. For example, the [2,3] element of the matrix would be the pairwise correlation between SNP2 and SNP3.

A list of the fitted values alpha

Examples

```
GE_scoreeq_sim( num_sims=10, num_sub=1000, beta_list=as.list(runif(n=6, min=0, max=1)),
rho_list=as.list(rep(0.3,6)), prob_G=0.3, cov_Z=1, cov_W=1)
```

GE_scoreeq_sim_old GE_scoreeq_sim_old.R

Description

Here we perform simulation to verify that we have solved for the correct alpha values in GE_bias_norm_squaredmis_old(). Make the same assumptions as in GE_bias_norm_squaredmis_old().

Usage

```
GE_scoreeq_sim_old(num_sims = 5000, num_sub = 2000, beta_list, prob_G,
    rho_list, cov_Z = NULL, cov_W = NULL)
```

num_sims	The number of simulations to run, we suggest 5000.
num_sub	The number of subjects to generate in every simulation, we suggest 2000.
beta_list	A list of the effect sizes in the true model. Use the order beta_0, beta_G, beta_E, beta_I, beta_Z, beta_M. If Z or M is a vector, then beta_Z and beta_M should be vectors.
prob_G	Probability that each allele is equal to 1. Since each SNP has two alleles, the expectation of G is 2*prob_G.
rho_list	A list of the 6 pairwise covariances between the covariates. These should be in the order (1) cov_GE (2) cov_GZ (3) cov_EZ (4) cov_GW (5) cov_EW (6) cov_ZW . Again if Z or W are vectors then terms like cov_GZ should be vectors (in the order $cov(G,Z_1),,cov(G,Z_p)$) where Z is of dimension p, and simi- larly for W. If Z or M are vectors, then cov_ZW should be a vector in the order $(cov(Z_1,W_1),,cov(Z_1,W_q), cov(Z_2,W_1),,cov(Z_p,W_q)$ where Z is a vector of length p and W is a vector of length q.

cov_Z	Only used if Z is a vector, gives the covariance matrix of Z (remember by assumption Z has mean 0 and variance 1). The (i,j) element of the matrix should be the $(i-1)(i-2)/2+j$ element of the vector.
cov_W	Only used if W is a vector, gives the covariance matrix of W (remember by assumption W has mean 0 and variance 1). The (i,j) element of the matrix should be the $(i-1)(i-2)/2+j$ element of the vector.

A list of the fitted values alpha

Examples

GE_scoreeq_sim_old(num_sims=10, beta_list=as.list(runif(n=6, min=0, max=1)), rho_list=as.list(rep(0.3,6)), prob_G=0.3)

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