Package ‘r2mlm’

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Type Package

Title R-Squared Measures for Multilevel Models

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Description Generates both total- and level-specific R-squared measures from Rights and Sterba’s (2019) <doi:10.1037/met0000184> framework of R-squared measures for multilevel models with random intercepts and/or slopes, which is based on a complete decomposition of variance. Additionally generates graphical representations of these R-squared measures to allow visualizing and interpreting all measures in the framework together as an integrated set. This framework subsumes 10 previously-developed R-squared measures for multilevel models as special cases of 5 measures from the framework, and it also includes several newly-developed measures. Measures in the framework can be used to compute R-squared differences when comparing multilevel models (following proce-dures in Rights & Sterba (2020) <doi:10.1080/00273171.2019.1660605>).

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Encoding UTF-8

LazyData true

Imports dplyr (>= 0.8.5), magrittr (>= 1.5), methods (>= 3.6.3), rlang (>= 0.4.6), rockchalk (>= 1.8.144), stringr (>= 1.4.0), tidyselect (>= 1.0.0)

Depends lme4 (>= 1.1.23), nlme (>= 3.1.14), R (>= 3.2.0)

RoxygenNote 7.1.1

Suggests testthat, Matrix

URL https://github.com/mkshaw/r2mlm

BugReports https://github.com/mkshaw/r2mlm/issues

NeedsCompilation no
r2mlm

Compute R-squared values for multilevel models, automatically inputting parameter estimates.

Description

r2mlm reads in a multilevel model (MLM) object generated using lmer or nlme, and outputs all relevant R-squared measures from the Rights and Sterba (2019) framework of multilevel model R-squared measures, which can be visualized together as a set using the outputted bar chart decompositions of outcome variance. That is, when predictors are cluster-mean-centered, all R-squared measures from Rights & Sterba (2019) Table 1 and decompositions from Rights & Sterba (2019) Figure 1 are outputted. When predictors are not cluster-mean-centered, the total R-squared measures from Rights & Sterba (2019) Table 5, as well as bar chart decompositions are outputted. Any number of level-1 and/or level-2 predictors is supported. Any of the level-1 predictors can have random slopes.

Usage

r2mlm(model, bargraph = TRUE)

Arguments

model A model generated using lmer or nlme. Note that models using lmer must specify random effects at the end of the model, like so: outcome ~ 1 + fixed_effects + (random_effects | cluster_variable). Anything else (e.g., outcome ~ 1 + (random_effects | cluster_variable) + fixed_effects) will not work.

bargraph Optional bar graph output, default is TRUE.
Details

r2mlm first determines whether a given model was generated using `lmer` or `nlme`, then passes the model to helper functions that pull the raw data and parameter estimates from the model, and pass that information to `r2mlm_manual`.

Previous MLM literature has offered two perspectives on how to treat variance attributable to random intercepts and slopes, called the “marginal” and “conditional” approaches (e.g., Edwards et al., 2008; Orelien & Edwards, 2008; Vonesh & Chinchilli, 1997; Wang & Schaalje, 2009; Xu, 2003). In the marginal approach, all variance attributable to predictors via random slope variation and attributable to cluster means via random intercept variation (i.e., sources “v” and “m”) is treated as unexplained. In the conditional approach, variance attributable to predictors via random slope variation and/or attributable to cluster mean variation is treated as explained. This package offers researchers access to both the marginal and conditional approaches. There are 5 marginal measures: f1_total, f2_total, f_total, f1_within, and f2_between. The other 7 measures are conditional: v_total, m_total, fv_total, fvm_total, v_within, f1v_within, and m_between.

Value

If the input is a valid model, then the output will be a list and associated graphical representation of R-squared decompositions. If the model is not valid, it will return an error prompting the user to input a valid model.

See Also


Other r2mlm single model functions: `r2mlm3_manual()`, `r2mlm_long_manual()`, `r2mlm_manual()`

Examples

# Using lme4 for your model

# The "bobyqa" optimizer is required for this particular model to converge

model_lme4 <- lmer(satisfaction ~ 1 + salary_c + control_c + salary_m + control_m + s_t_ratio + (1 + salary_c + control_c| schoolID), data = teachsat, REML = TRUE, control = lmerControl(optimizer = "bobyqa"))

r2mlm(model_lme4)

# Using nlme for your model

model_nlme <- lme(satisfaction ~ 1 + salary_c + control_c + salary_m + control_m + s_t_ratio, random = ~ 1 + salary_c + control_c | schoolID, data = teachsat, method = "REML", control = lmeControl(opt = "optim"))

r2mlm(model_nlme)
\textit{r2mlm3\_manual} takes as input raw data and three-level multilevel model (MLM) parameter estimates and outputs all relevant R-squared measures as well as an accompanying bar chart.

\textbf{Usage}

\begin{verbatim}
r2mlm3_manual(
data, l1\_covs, l2\_covs, l3\_covs, random\_covs12, random\_covs13, random\_covs23, gamma\_1, gamma\_2, gamma\_3, Tau12, Tau13, Tau23, sigma2, clustermeancentered = TRUE, Tau2\_noncmc = NULL, Tau3\_noncmc = NULL, l2clusterID\_noncmc = NULL, l3clusterID\_noncmc = NULL, bargraph = TRUE
)
\end{verbatim}

\textbf{Arguments}

- \texttt{data} Dataset with rows denoting observations and columns denoting variables
- \texttt{l1\_covs} Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-1 predictors used in the MLM (if none used, set to NULL)
- \texttt{l2\_covs} Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-2 predictors used in the MLM (if none used, set to NULL)
- \texttt{l3\_covs} Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-3 predictors used in the MLM (if none used, set to NULL)
random_covs12 Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-1 predictors that have random slopes across level-2 units in the MLM (if no such random slopes, set to NULL)

gamma_1 Vector of fixed slope estimates for all level-1 predictors, to be entered in the order of the predictors listed by l1_covs (if none, set to NULL)

random_covs13 Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-1 predictors that have random slopes across level-3 units in the MLM (if no such random slopes, set to NULL)

random_covs23 Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-2 predictors that have random slopes across level-3 units in the MLM (if no such random slopes, set to NULL)

gamma_2 Vector of fixed slope estimates for all level-2 predictors, to be entered in the order of the predictors listed by l2_covs (if none, set to NULL)

random_covs12 Vector of numbers (or variable names) corresponding to the columns in the dataset of the level-1 predictors that have random slopes across level-2 units in the MLM (if no such random slopes, set to NULL)

gamma_3 Vector of fixed slope estimates for all level-3 predictors, to be entered in the order of the predictors listed by l3_covs (if none, set to NULL)

Tau12 For cluster-mean-centered model results (set to NULL if entering non-cluster-mean-centered model results), this is the random effect covariance matrix with the first row/column denoting the intercept variance and covariances across level-2 units and each subsequent row/column denotes a given level-1 predictor’s random slope variance and covariances across level-2 units (to be entered in the order listed by random_covs12; if none, set to NULL)

Tau13 For cluster-mean-centered model results (set to NULL if entering non-cluster-mean-centered model results), this is the random effect covariance matrix with the first row/column denoting the intercept variance and covariances across level-3 units and each subsequent row/column denotes a given level-1 predictor’s random slope variance and covariances across level-3 units (to be entered in the order listed by random_covs13; if none, set to NULL)

sigma2 Level-1 residual variance

clustermeancentered By default, this argument is set to TRUE, indicating that cluster-mean-centered model results are being inputted. When instead entering non-cluster-mean-centered model results, set this argument to FALSE. Additionally, for non-cluster-mean-centered model results, random effect variances/covariances are to be entered in arguments Tau2_noncmc and Tau3_noncmc (defined below), rather than in the Tau12, Tau13, and Tau23 arguments used for cluster-mean-centered model results. Additionally, when entering non-cluster-mean-centered model results, user must specify l2clusterID_noncmc and l3clusterID_noncmc (neither of which are necessary for cluster-mean-centered model results). Function input is otherwise the same for cluster-mean-centered and non-cluster-mean-centered model results.
For non-cluster-mean-centered model results, this is the level-2 random effect covariance matrix; the first row/column denotes the intercept variance and covariances across level-2 units and each subsequent row/column denotes a given predictor’s random slope variance and covariances across level-2 units (to be entered in the order listed by randomcovs12_noncmc; by default, this argument is set to NULL)

For non-cluster-mean-centered model results, this is the level-3 random effect covariance matrix; the first row/column denotes the intercept variance and covariances across level-3 units and each subsequent row/column denotes a given predictor’s random slope variance and covariances across level-3 units (to be entered in the order listed by randomcovs12_noncmc; by default, this argument is set to NULL)

For non-cluster-mean-centered model results, this is the number (or variable name) corresponding to the column in the dataset containing the level-2 cluster identification (function assumes that each level-2 cluster ID is unique; by default, this argument is set to NULL)

For non-cluster-mean-centered model results, this is the number (or variable name) corresponding to the column in the dataset containing the level-3 cluster identification (function assumes that each level-3 cluster ID is unique; by default, this argument is set to NULL)

Optional bar graph output, default is TRUE.

This function can also accommodate two-level models. To input results for two-level models, set the following arguments equal to NULL: l3_covs, random_covs13, random_covs23, gamma_3, Tau13, Tau23.

If the input is valid, then the output will be a list and associated graphical representation of R-squared decompositions. If the input is not valid, it will return an error.

Other r2mlm single model functions: `r2mlm_long_manual()`, `r2mlm_manual()`, `r2mlm()`
Description

r2mlm_comp reads in two multilevel models (MLMs) (generated using lmer or nlme) under comparison (designated Model A and Model B), and outputs all R-squared measures in the Rights and Sterba (2019) framework for both models, as well as R-squared differences between the two models. Definitions of these R-squared difference measures are provided in Rights & Sterba (2020) Table 1; importantly, to detect the impact of a specific kind of term (e.g., the kind of term added to Model A to form Model B), a particular target single-source R-squared difference measure from this framework is used. For instructions on how to identify which target single-source R-squared difference measure to interpret to detect the impact of which kind of term that distinguishes Model A from B, see Rights and Sterba (2020) Table 2. Additionally, this function produces side-by-side graphical comparisons of the R-squared measures for Model A vs. Model B that can be used to visualize changes in each measure across models. This function assumes all level-1 predictors are cluster-mean-centered, for reasons described in Rights & Sterba (2020). Any number of level-1 and/or level-2 predictors is supported and any of the level-1 predictors can have random slopes. This function can be used with either the hierarchical or the simultaneous model-building approach described in Rights and Sterba (2020). This function can be used with either nested or non-nested model comparisons (in which R-squared estimates for Model A are subtracted from those for Model B).

Usage

r2mlm_comp(modelA, modelB, data = NULL, bargraph = TRUE)

Arguments

modelA, modelB Models generated using lmer or nlme. Note that models using lmer must specify random effects at the end of the model, like so: outcome ~ 1 + fixed_effects + (random_effects | cluster_variable). Anything else (e.g., outcome ~ 1 + (random_effects | cluster_variable) + fixed_effects) will not work.

data Optional argument, only needed if models are not hierarchical. Dataset with rows denoting observations and columns denoting variables.
bargraph Optional bar graph output, default is TRUE.

Details

Assumes that both models are fit with lmer or both models are fit with nlme.

Value

If the inputs are valid models, then the output will be a list and associated graphical representation of R-squared decompositions. If the models are not valid, the function will return an error prompting the user to input valid models.

See Also


Other r2mlm model comparison functions: \texttt{r2mlm\_comp\_manual()}

**Examples**

# Using lme4 for your model
# The "bobyqa" optimizer is required for these particular models to converge
## Not run: # Model A, no "salary" components included
modelA\_lme4 <- lmer(satisfaction ~ 1 + control\_c + control\_m + s\_t\_ratio + (1 + control\_c | schoolID), data = teachsat, REML = TRUE, control = lmerControl(optimizer = "bobyqa"))

# Model B, full model with "salary" components included
modelB\_lme4 <- lmer(satisfaction ~ 1 + salary\_c + control\_c + salary\_m + control\_m + s\_t\_ratio + (1 + salary\_c + control\_c | schoolID), data = teachsat, REML = TRUE, control = lmerControl(optimizer = "bobyqa"))

# Compare models
r2mlm\_comp(modelA\_lme4, modelB\_lme4)

# Compare models, optional data argument specified
r2mlm\_comp(modelA\_lme4, modelB\_lme4, teachsat)

# Using nlme for your model
# Model A, no "salary" components included
modelA\_nlme <- lme(satisfaction ~ 1 + control\_c + control\_m + s\_t\_ratio, random = ~ 1 + control\_c | schoolID, data = teachsat, method = "REML", control = lmeControl(opt = "optim"))

# Model B, full model with "salary" components included
modelB\_nlme <- lme(satisfaction ~ 1 + salary\_c + control\_c + salary\_m + control\_m + s\_t\_ratio, random = ~ 1 + salary\_c + control\_c | schoolID, data = teachsat, method = "REML", control = lmeControl(opt = "optim"))

# Compare models
r2mlm\_comp(modelA\_nlme, modelB\_nlme)

# Compare models, optional data argument specified
r2mlm\_comp(modelA\_nlme, modelB\_nlme, teachsat)
r2mlm_comp_manual

Compute R-squared differences between two multilevel models, manually inputting parameter estimates.

Description

r2mlm_comp_manual reads in raw data and multilevel model (MLM) parameter estimates from two separate models under comparison (designated Model A and Model B), and outputs all R-squared measures in the Rights and Sterba (2019) framework for both models, as well as R-squared differences between the two models. Definitions of these R-squared difference measures are provided in Rights & Sterba (2020) Table 1; importantly, to detect the impact of a specific kind of term (e.g., the kind of term added to Model A to form Model B), a particular target single-source R-squared difference measure from this framework is used. For instructions on how to identify which target single-source R-squared difference measure to interpret to detect the impact of which kind of term that distinguishes Model A from B, see Rights and Sterba (2020) Table 2. Additionally, this function produces side-by-side graphical comparisons of the R-squared measures for Model A vs. Model B that can be used to visualize changes in each measure across models. This function assumes all level-1 predictors are cluster-mean-centered for reasons described in Rights & Sterba (2020). Any number of level-1 and/or level-2 predictors is supported and any of the level-1 predictors can have random slopes. This function can be used with either the hierarchical or the simultaneous model-building approach described in Rights and Sterba (2020). This function can also be used with either nested or non-nested model comparisons (in which R-squared estimates for Model A are subtracted from those for Model B).

Usage

r2mlm_comp_manual(
  data,
  within_covs_modA,  
  between_covs_modA, 
  random_covs_modA,  
  gamma_w_modA,     
  gamma_b_modA,     
  Tau_modA,         
  sigma2_modA,      
  within_covs_modB, 
  between_covs_modB, 
  random_covs_modB, 
  gamma_w_modB,     
  gamma_b_modB,     
  Tau_modB,         
  sigma2_modB,      
  bargraph = TRUE   
)
Arguments

data
Dataset with rows denoting observations and columns denoting variables.

within_covs_modA, within_covs_modB
List of numbers corresponding to the columns in the dataset of the level-1 predictors used in the MLM (if none used, set to NULL).

between_covs_modA, between_covs_modB
List of numbers corresponding to the columns in the dataset of the level-2 predictors used in the MLM (if none used, set to NULL).

random_covs_modA, random_covs_modB
List of numbers corresponding to the columns in the dataset of the level-1 predictors that have random slopes in the MLM (if no random slopes, set to NULL).

gamma_w_modA, gamma_w_modB
Vector of fixed slope estimates for all level-1 predictors, to be entered in the order of the predictors listed by within_covs (if none, set to NULL).

gamma_b_modA, gamma_b_modB
Vector of fixed intercept estimate (if applicable; see has_intercept below) and fixed slope estimates for all level-2 predictors, to be entered intercept first (if applicable) followed by level-2 slopes in the order listed by between_covs (if none, set to NULL).

Tau_modA, Tau_modB
Random effect covariance matrix; note that the first row/column denotes the intercept variance and covariances (if intercept is fixed, set all to 0) and each subsequent row/column denotes a given random slope’s variance and covariances (to be entered in the order listed by random_covs).

sigma2_modA, sigma2_modB
Level-1 residual variance.

bargraph
Optional bar graph output, default is TRUE.

Value

If the inputs are valid models, then the output will be a list and associated graphical representation of R-squared decompositions.

See Also


Other r2mlm model comparison functions: r2mlm_comp()

Examples

# Model A: no "salary" components included
The provided text contains R code and documentation for the `r2mlm_long_manual` function. The function calculates R-squared values for longitudinal multilevel models, manually inputting parameter estimates. The description of the function explains its purpose and usage, which involves inputting raw data and multilevel model parameter estimates and outputs relevant R-squared measures along with a bar chart. The function extends the `r2mlm_manual` function by allowing researchers to input heteroscedastic variance estimates and by providing level-specific measures for non-cluster-mean-centered models. The usage example demonstrates how to call the function with necessary arguments.
Arguments

data: Dataset with rows denoting observations and columns denoting variables
covs: list of predictors in the dataset that have fixed components of slopes included in the model (if none, set to NULL)
random_covs: list of predictors in the dataset that have random components of slopes included in the model (if none, set to NULL)
clusterID: variable name in dataset corresponding to cluster (e.g., person) identification
gammas: vector containing estimated fixed components of all slopes, listed in the order specified in covs (if none, set to NULL)
Tau: random effect covariance matrix; the first row and the first column denote the intercept variance and covariances and each subsequent row/column denotes a given random slope’s variance and covariances (to be entered in the order listed by random_covs)
sigma2: level-1 residual variance; can be entered as a single number, or as a set of numbers, for example corresponding to different residual variances at individual timepoints; if entered as a set of numbers, function will assume equal weights and take the raw average of these to estimate the expectation of the error variance
bargraph: Optional bar graph output, default is TRUE.

Details

This function reads in raw data as well as parameter estimates from the researcher’s previously fit longitudinal growth model (hence, any software program can have been used to fit the researcher’s longitudinal growth model prior to the use of this R function, so long as parameter estimates from the fitted model are recorded; note that this function accommodates non-longitudinal models as well). This function then outputs R-squared measures as well as variance decompositions and associated bar charts outlined in Rights & Sterba (2021). This function allows researchers to input heteroscedastic residual variance by including multiple estimates, for example, corresponding to individual timepoints. Users need not specify if predictors are person-mean-centered or not—the function will automatically output total, within-person, and between-person variance attributable to each potential source of explained variance (f1, f2, v1, v2, and m). Note, however, that the interpretations of these sources differ for person-mean-centered versus non-person-mean-centered models and that variance attributable to v2 will necessarily be 0 for person-mean-centered models.

Value

If the input is valid, then the output will be a list and associated graphical representation of R-squared decompositions. If the input is not valid, it will return an error.

See Also


Other r2mlm single model functions: r2mlm3_manual(), r2mlm_manual(), r2mlm()
Examples

# Removing cluster-mean-centering from the teachsat dataset, for
# demonstration purposes

teachsat$salary <- teachsat$salary_c + 2
uncentered_model <- lmer(satisfaction ~ salary + (1 | schoolID), data = teachsat)

r2mlm_long_manual(data = teachsat,
covs = c("salary"),
random_covs = NULL,
clusterID = "schoolID",
gammas = c(0.07430),
Tau = as.matrix(Matrix::bdiag(VarCorr(uncentered_model))),
sigma2 = getME(uncentered_model, "sigma")^2,
bargraph = TRUE)

Description

r2mlm_manual takes as input raw data and parameter estimates from a multilevel model, and outputs all relevant R-squared measures from the Rights and Sterba (2019) framework of R-squared measures for multilevel models, which can be visualized together as a set using the outputted bar chart decompositions of outcome variance. That is, when predictors are cluster-mean-centered, all R-squared measures from Rights & Sterba (2019) Table 1 and decompositions from Rights & Sterba (2019) Figure 1 are outputted. When predictors are not cluster-mean-centered, the total R-squareds from Rights & Sterba (2019) Table 5, as well as bar chart decompositions are outputted. Any number of level-1 and/or level-2 predictors is supported. Any of the level-1 predictors can have random slopes.

Usage

r2mlm_manual(
  data,
  within_covs,
  between_covs,
  random_covs,
  gamma_w,
  gamma_b,
  Tau,
  sigma2,
  has_intercept = TRUE,
  clustermeancentered = TRUE,
  bargraph = TRUE)

Compute R-squared values for multilevel models, manually inputting parameter estimates.
Arguments

- `data`: Dataset with rows denoting observations and columns denoting variables.
- `within_covs`: List of numbers corresponding to the columns in the dataset of the level-1 predictors used in the MLM (if none used, set to NULL).
- `between_covs`: List of numbers corresponding to the columns in the dataset of the level-2 predictors used in the MLM (if none used, set to NULL).
- `random_covs`: List of numbers corresponding to the columns in the dataset of the level-1 predictors that have random slopes in the MLM (if no random slopes, set to NULL).
- `gamma_w`: Vector of fixed slope estimates for all level-1 predictors, to be entered in the order of the predictors listed by `within_covs` (if none, set to NULL).
- `gamma_b`: Vector of fixed intercept estimate (if applicable; see `has_intercept` below) and fixed slope estimates for all level-2 predictors, to be entered intercept first (if applicable) followed by level-2 slopes in the order listed by `between_covs` (if none, set to NULL).
- `Tau`: Random effect covariance matrix; note that the first row/column denotes the intercept variance and covariances (if intercept is fixed, set all to 0) and each subsequent row/column denotes a given random slope’s variance and covariances (to be entered in the order listed by `random_covs`).
- `sigma2`: Level-1 residual variance.
- `has_intercept`: If set to TRUE, the first element of `gamma_b` is assumed to be the fixed intercept estimate; if set to FALSE, the first element of `gamma_b` is assumed to be the first fixed level-2 predictor slope; set to TRUE by default.
- `clustermeancentered`: If set to TRUE, all level-1 predictors (indicated by the `within_covs` list) are assumed to be cluster-mean-centered and function will output all decompositions; if set to FALSE, function will output only total decompositions (see Description above); set to TRUE by default.
- `bargraph`: Optional bar graph output, default is TRUE.

Value

If the input is valid, then the output will be a list and associated graphical representation of R-squared decompositions. If the input is not valid, it will return an error.

See Also


Other r2mlm single model functions: `r2mlm3_manual()`, `r2mlm_long_manual()`, `r2mlm()`

Examples

```r
# The bobyqa optimizer is required for this model to converge in lme4
```
teachsat

model <- lmer(satisfaction ~ 1 + salary_c + control_c + salary_m + control_m + s_t_ratio + (1 + salary_c + control_c | schoolID), data = teachsat, REML = TRUE, control = lmerControl(optimizer = "bobyqa"))

r2mlm_manual(data = teachsat, within_covs = c(5, 4), between_covs = c(7, 6, 8), random_covs = c(5, 4), gamma_w = c(0.074485, 0.310800), gamma_b = c(4.352652, 0.036759, 0.027532, -0.035250), Tau = matrix(c(0.387, 0.0000646, 0.00625, 0.0000646, 0.00277, -0.000333, 0.00625, -0.000333, 0.0285), 3, 3), sigma2 = 0.55031, has_intercept = TRUE, clustermeancentered = TRUE)

teachsat  

Description

A simulated dataset containing information about teacher job satisfaction. Teachers clustered within schools.

Usage

teachsat

Format

A data frame with 9000 rows and 8 columns:

schoolID  school identification number

teacherID  teacher identification number

satisfaction  teacher job satisfaction, 1-10 scale

control_c  school-mean-centered rating of teacher’s reported control over curriculum

salary_c  school-mean-centered teacher’s salary (in thousands of dollars)

control_m  school mean rating of teacher’s reported control over curriculum

salary_m  school mean teacher’s salary (in thousands of dollars)

s_t_ratio  student to teacher ratio for the school (number of students per teacher)
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