

# Package ‘mrmr’

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**Type** Package

**Title** Mixed Models for Repeated Measures

**Version** 0.3.10

**Description** Mixed models for repeated measures (MMRM) are a popular choice for analyzing longitudinal continuous outcomes in randomized clinical trials and beyond; see Cnaan, Laird and Slasor (1997) <[doi:10.1002/\(SICI\)1097-0258\(19971030\)16:20%3C2349::AID-SIM667%3E3.0.CO;2-E](https://doi.org/10.1002/(SICI)1097-0258(19971030)16:20%3C2349::AID-SIM667%3E3.0.CO;2-E)> for a tutorial and Mallinckrodt, Lane, Schnell, Peng and Mancuso (2008) <[doi:10.1177/009286150804200402](https://doi.org/10.1177/009286150804200402)> for a review. This package implements MMRM based on the marginal linear model without random effects using Template Model Builder ('TMB') which enables fast and robust model fitting. Users can specify a variety of covariance matrices, weight observations, fit models with restricted or standard maximum likelihood inference, perform hypothesis testing with Satterthwaite or Kenward-Roger adjustment, and extract least square means estimates by using 'emmeans'.

**License** Apache License 2.0

**URL** <https://openpharma.github.io/mrmr/>

**BugReports** <https://github.com/openpharma/mrmr/issues>

**Depends** R (>= 4.0)

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 'kenwardroger.R' 'mmrm-methods.R' 'mmrm-package.R' 'utils.R'  
 'residual.R' 'utils-nse.R' 'utils-formula.R' 'satterthwaite.R'  
 'skipping.R' 'tidiers.R' 'testing.R' 'tmb-methods.R' 'tmb.R'  
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mrmr-package

mrmr *Package*


---

## Description

mrmr implements mixed models with repeated measures (MMRM) in R.

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## See Also

Useful links:

- <https://openpharma.github.io/mmrn/>
- Report bugs at <https://github.com/openpharma/mmrn/issues>

---

as.cov\_struct

*Coerce into a Covariance Structure Definition*

---

## Description

[Maturing]

## Usage

```
as.cov_struct(x, ...)

## S3 method for class 'formula'
as.cov_struct(x, warn_partial = TRUE, ...)
```

## Arguments

x	an object from which to derive a covariance structure. See object specific sections for details.
...	additional arguments unused.
warn_partial	(flag) whether to emit a warning when parts of the formula are disregarded.

## Details

A covariance structure can be parsed from a model definition formula or call. Generally, covariance structures defined using non-standard evaluation take the following form:

```
type( (visit, )* visit | (group /)? subject )
```

For example, formulas may include terms such as

```
us(time | subject)
cp(time | group / subject)
sp_exp(coord1, coord2 | group / subject)
```

Note that only `sp_exp` (spatial) covariance structures may provide multiple coordinates, which identify the Euclidean distance between the time points.

**Value**

A `cov_struct()` object.

**Methods (by class)**

- `as.cov_struct(formula)`: When provided a formula, any specialized functions are assumed to be covariance structure definitions and must follow the form:

```
y ~ xs + type( (visit, )* visit | (group /)? subject )
```

Any component on the right hand side of a formula is considered when searching for a covariance definition.

**See Also**

Other covariance types: [cov\\_struct\(\)](#), [covariance\\_types](#)

**Examples**

```
# provide a covariance structure as a right-sided formula
as.cov_struct(~ csh(visit | group / subject))

# when part of a full formula, suppress warnings using `warn_partial = FALSE`
as.cov_struct(y ~ x + csh(visit | group / subject), warn_partial = FALSE)
```

---

bcva\_data

*Example Data on BCVA*

---

**Description**

[Experimental]

**Usage**

```
bcva_data
```

**Format**

A tibble with 10,000 rows and 7 variables:

- USUBJID: subject ID.
- VISITN: visit number (numeric).
- AVISIT: visit number (factor).
- ARMCD: treatment, TRT or CTL.
- RACE: 3-category race.
- BCVA\_BL: BCVA at baseline.
- BCVA\_CHG: Change in BCVA at study visits.

**Note**

Measurements of BCVA (best corrected visual acuity) is a measure of how many letters a person can read off of an eye chart using corrective lenses or contacts. This is a common endpoint in ophthalmology trials.

**Source**

This is an artificial dataset.

---

 component

*Component Access for mrmr\_tmb Objects*


---

**Description**

**[Experimental]**

**Usage**

```
component(
  object,
  name = c("cov_type", "subject_var", "n_theta", "n_subjects", "n_timepoints", "n_obs",
    "beta_vcov", "beta_vcov_complete", "varcov", "formula", "dataset", "n_groups",
    "reml", "convergence", "evaluations", "method", "optimizer", "conv_message", "call",
    "theta_est", "beta_est", "beta_est_complete", "beta_aliased", "x_matrix", "y_vector",
    "neg_log_lik", "jac_list", "theta_vcov", "full_frame")
)
```

**Arguments**

object	(mrmr_tmb) the fitted MMRM.
name	(character) the component(s) to be retrieved.

**Details**

Available component() names are as follows:

- call: low-level function call which generated the model.
- formula: model formula.
- dataset: data set name.
- cov\_type: covariance structure type.
- n\_theta: number of parameters.
- n\_subjects: number of subjects.
- n\_timepoints: number of modeled time points.

- `n_obs`: total number of observations.
- `reml`: was REML used (ML was used if FALSE).
- `neg_log_lik`: negative log likelihood.
- `convergence`: convergence code from optimizer.
- `conv_message`: message accompanying the convergence code.
- `evaluations`: number of function evaluations for optimization.
- `method`: Adjustment method which was used (for `mrm` objects), otherwise NULL (for `mrm_tmb` objects).
- `beta_vcov`: estimated variance-covariance matrix of coefficients (excluding aliased coefficients). When Kenward-Roger/Empirical adjusted coefficients covariance matrix is used, the adjusted covariance matrix is returned (to still obtain the original asymptotic covariance matrix use `object$beta_vcov`).
- `beta_vcov_complete`: estimated variance-covariance matrix including aliased coefficients with entries set to NA.
- `varcor`: estimated covariance matrix for residuals. If there are multiple groups, a named list of estimated covariance matrices for residuals will be returned. The names are the group levels.
- `theta_est`: estimated variance parameters.
- `beta_est`: estimated coefficients (excluding aliased coefficients).
- `beta_est_complete`: estimated coefficients including aliased coefficients set to NA.
- `beta_aliased`: whether each coefficient was aliased (i.e. cannot be estimated) or not.
- `theta_vcov`: estimated variance-covariance matrix of variance parameters.
- `x_matrix`: design matrix used (excluding aliased columns).
- `y_vector`: response vector used.
- `jac_list`: Jacobian, see [h\\_jac\\_list\(\)](#) for details.
- `full_frame`: data.frame with `n` rows containing all variables needed in the model.

### Value

The corresponding component of the object, see details.

### See Also

In the `lme4` package there is a similar function `getME()`.

### Examples

```
fit <- mrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data
)
# Get all available components.
component(fit)
# Get convergence code and message.
```

```

component(fit, c("convergence", "conv_message"))
# Get modeled formula as a string.
component(fit, c("formula"))

```

---

covariance\_types      *Covariance Types*

---

## Description

[Maturing]

## Usage

```

cov_types(
  form = c("name", "abbr", "habbr"),
  filter = c("heterogeneous", "spatial")
)

```

## Arguments

**form** (character)  
 covariance structure type name form. One or more of "name", "abbr" (abbreviation), or "habbr" (heterogeneous abbreviation).

**filter** (character)  
 covariance structure type filter. One or more of "heterogeneous" or "spatial".

## Value

A character vector of accepted covariance structure type names and abbreviations.

## Abbreviations for Covariance Structures

### Common Covariance Structures::

Structure	Description	Parameters	$(i, j)$ element
ad	Ante-dependence	$m$	$\sigma^2 \prod_{k=i}^{j-1} \rho_k$
adh	Heterogeneous ante-dependence	$2m - 1$	$\sigma_i \sigma_j \prod_{k=i}^{j-1} \rho_k$
ar1	First-order auto-regressive	2	$\sigma^2 \rho^{ i-j }$
ar1h	Heterogeneous first-order auto-regressive	$m + 1$	$\sigma_i \sigma_j \rho^{ i-j }$
cs	Compound symmetry	2	$\sigma^2 [\rho I(i \neq j) + I(i = j)]$
csh	Heterogeneous compound symmetry	$m + 1$	$\sigma_i \sigma_j [\rho I(i \neq j) + I(i = j)]$
toep	Toeplitz	$m$	$\sigma_{ i-j +1}$
toeph	Heterogeneous Toeplitz	$2m - 1$	$\sigma_i \sigma_j \rho^{ i-j }$
us	Unstructured	$m(m + 1)/2$	$\sigma_{ij}$

where  $i$  and  $j$  denote  $i$ -th and  $j$ -th time points, respectively, out of total  $m$  time points,  $1 \leq i, j \leq m$



*m.*

### Note

The **ante-dependence** covariance structure in this package refers to homogeneous ante-dependence, while the ante-dependence covariance structure from SAS PROC MIXED refers to heterogeneous ante-dependence and the homogeneous version is not available in SAS.

For all non-spatial covariance structures, the time variable must be coded as a factor.

#### Spatial Covariance structures::

Structure	Description	Parameters	$(i, j)$ element
sp_exp	spatial exponential	2	$\sigma^2 \rho^{-d_{ij}}$

where  $d_{ij}$  denotes the Euclidean distance between time points  $i$  and  $j$ .

### See Also

Other covariance types: [as.cov\\_struct\(\)](#), [cov\\_struct\(\)](#)

---

cov\_struct

*Define a Covariance Structure*

---

### Description

[Maturing]

### Usage

```
cov_struct(
  type = cov_types(),
  visits,
  subject,
  group = character(),
  heterogeneous = FALSE
)
```

### Arguments

type	(string) the name of the covariance structure type to use. For available options, see <code>cov_types()</code> . If a type abbreviation is used that implies heterogeneity (e.g. <code>cph</code> ) and no value is provided to <code>heterogeneous</code> , then the heterogeneity is derived from the type name.
visits	(character) a vector of variable names to use for the longitudinal terms of the covariance structure. Multiple terms are only permitted for the "spatial" covariance type.

subject	(string) the name of the variable that encodes a subject identifier.
group	(string) optionally, the name of the variable that encodes a grouping variable for subjects.
heterogeneous	(flag)

**Value**

A cov\_struct object.

**See Also**

Other covariance types: [as.cov\\_struct\(\)](#), [covariance\\_types](#)

**Examples**

```
cov_struct("csh", "AVISITN", "USUBJID")
cov_struct("spatial", c("VISITA", "VISITB"), group = "GRP", subject = "SBJ")
```

---

df\_1d

---

*Calculation of Degrees of Freedom for One-Dimensional Contrast*


---

**Description**

**[Experimental]** Calculates the estimate, adjusted standard error, degrees of freedom, t statistic and p-value for one-dimensional contrast.

**Usage**

```
df_1d(object, contrast)
```

**Arguments**

object	(mmrm) the MMRM fit.
contrast	(numeric) contrast vector. Note that this should not include elements for singular coefficient estimates, i.e. only refer to the actually estimated coefficients.

**Value**

List with est, se, df, t\_stat and p\_val.

**Examples**

```

object <- mrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data
)
contrast <- numeric(length(object$beta_est))
contrast[3] <- 1
df_1d(object, contrast)

```

df\_md

*Calculation of Degrees of Freedom for Multi-Dimensional Contrast***Description**

**[Experimental]** Calculates the estimate, standard error, degrees of freedom, t statistic and p-value for one-dimensional contrast, depending on the method used in [mrm\(\)](#).

**Usage**

```
df_md(object, contrast)
```

**Arguments**

object	(mrm) the MMRM fit.
contrast	(matrix) numeric contrast matrix, if given a numeric then this is coerced to a row vector. Note that this should not include elements for singular coefficient estimates, i.e. only refer to the actually estimated coefficients.

**Value**

List with num\_df, denom\_df, f\_stat and p\_val (2-sided p-value).

**Examples**

```

object <- mrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data
)
contrast <- matrix(data = 0, nrow = 2, ncol = length(object$beta_est))
contrast[1, 2] <- contrast[2, 3] <- 1
df_md(object, contrast)

```

---

emmeans_support	<i>Support for emmeans</i>
-----------------	----------------------------

---

## Description

### [Experimental]

This package includes methods that allow mrm objects to be used with the emmeans package. emmeans computes estimated marginal means (also called least-square means) for the coefficients of the MMRM. We can also e.g. obtain differences between groups by applying `pairs()` on the object returned by `emmeans::emmeans()`.

## Examples

```
fit <- mrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data
)
if (require(emmeans)) {
  emmeans(fit, ~ ARMCD | AVISIT)
  pairs(emmeans(fit, ~ ARMCD | AVISIT), reverse = TRUE)
}
```

---

emp_start	<i>Empirical Starting Value</i>
-----------	---------------------------------

---

## Description

Obtain empirical start value for unstructured covariance

## Usage

```
emp_start(data, model_formula, visit_var, subject_var, subject_groups, ...)
```

## Arguments

data	(data.frame) data used for model fitting.
model_formula	(formula) the formula in mrm model without covariance structure part.
visit_var	(string) visit variable.
subject_var	(string) subject id variable.
subject_groups	(factor) subject group assignment.
...	not used.

**Details**

This `emp_start` only works for unstructured covariance structure. It uses linear regression to first obtain the coefficients and use the residuals to obtain the empirical variance-covariance, and it is then used to obtain the starting values.

**Value**

A numeric vector of starting values.

**Note**

`data` is used instead of `full_frame` because `full_frame` is already transformed if model contains transformations, e.g.  $\log(\text{FEV1}) \sim \exp(\text{FEV1\_BL})$  will drop `FEV1` and `FEV1_BL` but add  $\log(\text{FEV1})$  and  $\exp(\text{FEV1\_BL})$  in `full_frame`.

---

`fev_data`*Example Data on FEV1*

---

**Description**

[Experimental]

**Usage**

```
fev_data
```

**Format**

A tibble with 800 rows and 7 variables:

- `USUBJID`: subject ID.
- `AVISIT`: visit number.
- `ARMCD`: treatment, TRT or PBO.
- `RACE`: 3-category race.
- `SEX`: sex.
- `FEV1_BL`: FEV1 at baseline (%).
- `FEV1`: FEV1 at study visits.
- `WEIGHT`: weighting variable.

**Note**

Measurements of FEV1 (forced expired volume in one second) is a measure of how quickly the lungs can be emptied. Low levels of FEV1 may indicate chronic obstructive pulmonary disease (COPD).

**Source**

This is an artificial dataset.

fit\_mmrn

*Low-Level Fitting Function for MMRM***Description****[Experimental]**

This is the low-level function to fit an MMRM. Note that this does not try different optimizers or adds Jacobian information etc. in contrast to `mmrm()`.

**Usage**

```
fit_mmrn(
  formula,
  data,
  weights,
  reml = TRUE,
  covariance = NULL,
  tmb_data,
  formula_parts,
  control = mmrm_control()
)
```

**Arguments**

formula	(formula) model formula with exactly one special term specifying the visits within subjects, see details.
data	(data.frame) input data containing the variables used in formula.
weights	(vector) input vector containing the weights.
reml	(flag) whether restricted maximum likelihood (REML) estimation is used, otherwise maximum likelihood (ML) is used.
covariance	(cov_struct) A covariance structure type definition, or value that can be coerced to a covariance structure using <code>as.cov_struct()</code> . If no value is provided, a structure is derived from the provided formula.
tmb_data	(mmrm_tmb_data) object.
formula_parts	(mmrm_tmb_formula_parts) list with formula parts from <code>h_mmrn_tmb_formula_parts()</code> .
control	(mmrm_control) list of control options produced by <code>mmrm_control()</code> .

## Details

The formula typically looks like:

```
FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID)
```

which specifies response and covariates as usual, and exactly one special term defines which covariance structure is used and what are the visit and subject variables.

Always use only the first optimizer if multiple optimizers are provided.

## Value

List of class `mrm_tmb`, see `h_mrm_tmb_fit()` for details. In addition, it contains elements `call` and `optimizer`.

## Examples

```
formula <- FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID)
data <- fev_data
system.time(result <- fit_mrm(formula, data, rep(1, nrow(fev_data))))
```

---

fit\_single\_optimizer    *Fitting an MMRM with Single Optimizer*

---

## Description

### [Experimental]

This function helps to fit an MMRM using TMB with a single optimizer, while capturing messages and warnings.

## Usage

```
fit_single_optimizer(
  formula,
  data,
  weights,
  reml = TRUE,
  covariance = NULL,
  tmb_data,
  formula_parts,
  ...,
  control = mrm_control(...)
)
```

**Arguments**

formula	(formula) the model formula, see details.
data	(data) the data to be used for the model.
weights	(vector) an optional vector of weights to be used in the fitting process. Should be NULL or a numeric vector.
reml	(flag) whether restricted maximum likelihood (REML) estimation is used, otherwise maximum likelihood (ML) is used.
covariance	(cov_struct) a covariance structure type definition as produced with <code>cov_struct()</code> , or value that can be coerced to a covariance structure using <code>as.cov_struct()</code> . If no value is provided, a structure is derived from the provided formula.
tmb_data	(mrmr_tmb_data) object.
formula_parts	(mrmr_tmb_formula_parts) object.
...	Additional arguments to pass to <code>mrmr_control()</code> .
control	(mrmr_control) object.

**Details**

`fit_single_optimizer` will fit the mrmr model using the control provided. If there are multiple optimizers provided in control, only the first optimizer will be used. If `tmb_data` and `formula_parts` are both provided, `formula`, `data`, `weights`, `reml`, and `covariance` are ignored.

**Value**

The `mrmr_fit` object, with additional attributes containing warnings, messages, optimizer used and convergence status in addition to the `mrmr_tmb` contents.

**Examples**

```
mod_fit <- fit_single_optimizer(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data,
  weights = rep(1, nrow(fev_data)),
  optimizer = "nlminb"
)
attr(mod_fit, "converged")
```



---

format.cov_struct	<i>Format Covariance Structure Object</i>
-------------------	---

---

### Description

Format Covariance Structure Object

### Usage

```
## S3 method for class 'cov_struct'  
format(x, ...)
```

### Arguments

x	(cov_struct) a covariance structure object.
...	Additional arguments unused.

### Value

A formatted string for x.

---

mrm	<i>Fit an MMRM</i>
-----	--------------------

---

### Description

#### [Experimental]

This is the main function fitting the MMRM.

### Usage

```
mrm(  
  formula,  
  data,  
  weights = NULL,  
  covariance = NULL,  
  reml = TRUE,  
  control = mrm_control(...),  
  ...  
)
```

**Arguments**

formula	(formula) the model formula, see details.
data	(data) the data to be used for the model.
weights	(vector) an optional vector of weights to be used in the fitting process. Should be NULL or a numeric vector.
covariance	(cov_struct) a covariance structure type definition as produced with <code>cov_struct()</code> , or value that can be coerced to a covariance structure using <code>as.cov_struct()</code> . If no value is provided, a structure is derived from the provided formula.
reml	(flag) whether restricted maximum likelihood (REML) estimation is used, otherwise maximum likelihood (ML) is used.
control	(mmrm_control) fine-grained fitting specifications list created with <code>mmrm_control()</code> .
...	arguments passed to <code>mmrm_control()</code> .

**Details**

The formula typically looks like:  $FEV1 \sim RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID)$  so specifies response and covariates as usual, and exactly one special term defines which covariance structure is used and what are the time point and subject variables. The covariance structures in the formula can be found in [covariance\\_types](#).

The time points have to be unique for each subject. That is, there cannot be time points with multiple observations for any subject. The rationale is that these observations would need to be correlated, but it is not possible within the currently implemented covariance structure framework to do that correctly. Moreover, for non-spatial covariance structures, the time variable must be a factor variable.

When optimizer is not set, first the default optimizer (L-BFGS-B) is used to fit the model. If that converges, this is returned. If not, the other available optimizers from `h_get_optimizers()`, including BFGS, CG and `nlmnmb` are tried (in parallel if `n_cores` is set and not on Windows). If none of the optimizers converge, then the function fails. Otherwise the best fit is returned.

Note that fine-grained control specifications can either be passed directly to the `mmrm` function, or via the `control` argument for bundling together with the `mmrm_control()` function. Both cannot be used together, since this would delete the arguments passed via `mmrm`.

**Value**

An `mmrm` object.

**Note**

The `mmrm` object is also an `mmrm_fit` and an `mmrm_tmb` object, therefore corresponding methods also work (see [mmrm\\_tmb\\_methods](#)).

Additional contents depend on the choice of the adjustment method:

- If Satterthwaite adjustment is used, the Jacobian information `jac_list` is included.
- If Kenward-Roger adjustment is used, `kr_comp` contains necessary components and `beta_vcov_adj` includes the adjusted coefficients covariance matrix.

Use of the package `emmeans` is supported, see [emmeans\\_support](#).

NA values are always omitted regardless of `na.action` setting.

When the number of visit levels is large, it usually requires large memory to create the covariance matrix. By default, the maximum allowed visit levels is 100, and if there are more visit levels, a confirmation is needed if run interactively. You can use `options(mmrm.max_visits = <target>)` to increase the maximum allowed number of visit levels. In non-interactive sessions the confirmation is not raised and will directly give you an error if the number of visit levels exceeds the maximum.

## Examples

```
fit <- mmrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data
)

# Direct specification of control details:
fit <- mmrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data,
  weights = fev_data$WEIGHTS,
  method = "Kenward-Roger"
)

# Alternative specification via control argument (but you cannot mix the
# two approaches):
fit <- mmrm(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data,
  control = mmrm_control(method = "Kenward-Roger")
)
```

---

mmrm\_control

*Control Parameters for Fitting an MMRM*

---

## Description

**[Experimental]** Fine-grained specification of the MMRM fit details is possible using this control function.

**Usage**

```
mmrm_control(
  n_cores = 1L,
  method = c("Satterthwaite", "Kenward-Roger", "Residual", "Between-Within"),
  vcov = NULL,
  start = std_start,
  accept_singular = TRUE,
  drop_visit_levels = TRUE,
  ...,
  optimizers = h_get_optimizers(...)
)
```

**Arguments**

<code>n_cores</code>	(count) number of cores to be used.
<code>method</code>	(string) adjustment method for degrees of freedom.
<code>vcov</code>	(string) coefficients covariance matrix adjustment method.
<code>start</code>	(NULL, numeric or function) optional start values for variance parameters. See details for more information.
<code>accept_singular</code>	(flag) whether singular design matrices are reduced to full rank automatically and additional coefficient estimates will be missing.
<code>drop_visit_levels</code>	(flag) whether to drop levels for visit variable, if visit variable is a factor, see details.
<code>...</code>	additional arguments passed to <code>h_get_optimizers()</code> .
<code>optimizers</code>	(list) optimizer specification, created with <code>h_get_optimizers()</code> .

**Details**

For example, if the data only has observations at visits VIS1, VIS3 and VIS4, by default they are treated to be equally spaced, the distance from VIS1 to VIS3, and from VIS3 to VIS4, are identical. However, you can manually convert this visit into a factor, with `levels = c("VIS1", "VIS2", "VIS3", "VIS4")`, and also use `drop_visits_levels = FALSE`, then the distance from VIS1 to VIS3 will be double, as VIS2 is a valid visit. However, please be cautious because this can lead to convergence failure when using an unstructured covariance matrix and there are no observations at the missing visits.

- The `method` and `vcov` arguments specify the degrees of freedom and coefficients covariance matrix adjustment methods, respectively.
  - Allowed `vcov` includes: "Asymptotic", "Kenward-Roger", "Kenward-Roger-Linear", "Empirical" (CR0), "Empirical-Jackknife" (CR3), and "Empirical-Bias-Reduced" (CR2).

- Allowed method includes: "Satterthwaite", "Kenward-Roger", "Between-Within" and "Residual".
- If method is "Kenward-Roger" then only "Kenward-Roger" or "Kenward-Roger-Linear" are allowed for vcov.
- The vcov argument can be NULL to use the default covariance method depending on the method used for degrees of freedom, see the following table:

method	Default vcov
Satterthwaite	Asymptotic
Kenward-Roger	Kenward-Roger
Residual	Empirical
Between-Within	Asymptotic

- Please note that "Kenward-Roger" for "Unstructured" covariance gives different results compared to SAS; Use "Kenward-Roger-Linear" for vcov instead for better matching of the SAS results.
- The argument start is used to facilitate the choice of initial values for fitting the model. If function is provided, make sure its parameter is a valid element of mrmr\_tmb\_data or mrmr\_tmb\_formula\_parts and it returns a numeric vector. By default or if NULL is provided, std\_start will be used. Other implemented methods include emp\_start.

## Value

List of class `mrmr_control` with the control parameters.

## Examples

```
mrmr_control(
  optimizer_fun = stats::optim,
  optimizer_args = list(method = "L-BFGS-B")
)
```

---

mrmr\_tidiers

*Tidying Methods for mrmr Objects*

---

## Description

### [Experimental]

These methods tidy the estimates from an `mrmr` object into a summary.

## Usage

```
## S3 method for class 'mrmr'
tidy(x, conf.int = FALSE, conf.level = 0.95, ...)
```

```
## S3 method for class 'mrmr'
glance(x, ...)

## S3 method for class 'mrmr'
augment(
  x,
  newdata = NULL,
  interval = c("none", "confidence", "prediction"),
  se_fit = (interval != "none"),
  type.residuals = c("response", "pearson", "normalized"),
  ...
)
```

### Arguments

<code>x</code>	(mrmr) fitted model.
<code>conf.int</code>	(flag) if TRUE columns for the lower ( <code>conf.low</code> ) and upper bounds ( <code>conf.high</code> ) of coefficient estimates are included.
<code>conf.level</code>	(number) defines the range of the optional confidence interval.
<code>...</code>	only used by <code>augment()</code> to pass arguments to the <code>predict.mrmr_tmb()</code> method.
<code>newdata</code>	(data.frame or NULL) optional new data frame.
<code>interval</code>	(string) type of interval calculation.
<code>se_fit</code>	(flag) whether to return standard errors of fit.
<code>type.residuals</code>	(string) passed on to <code>residuals.mrmr_tmb()</code> .

### Functions

- `tidy(mrmr)`: derives tidy tibble from an mrmr object.
- `glance(mrmr)`: derives glance tibble from an mrmr object.
- `augment(mrmr)`: derives augment tibble from an mrmr object.

### See Also

[mrmr\\_methods](#), [mrmr\\_tmb\\_methods](#) for additional methods.

### Examples

```
fit <- mrmr(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data
```

```

)
# Applying tidy method to return summary table of covariate estimates.
fit |> tidy()
fit |> tidy(conf.int = TRUE, conf.level = 0.9)
# Applying glance method to return summary table of goodness of fit statistics.
fit |> glance()
# Applying augment method to return merged `tibble` of model data, fitted and residuals.
fit |> augment()
fit |> augment(interval = "confidence")
fit |> augment(type.residuals = "pearson")

```

---

mrmr\_tmb\_methods      *Methods for mrmr\_tmb Objects*

---

## Description

**[Experimental]**

## Usage

```

## S3 method for class 'mrmr_tmb'
coef(object, complete = TRUE, ...)

## S3 method for class 'mrmr_tmb'
fitted(object, ...)

## S3 method for class 'mrmr_tmb'
predict(
  object,
  newdata,
  se.fit = FALSE,
  interval = c("none", "confidence", "prediction"),
  level = 0.95,
  nsim = 1000L,
  ...
)

## S3 method for class 'mrmr_tmb'
model.frame(
  formula,
  data,
  include = c("subject_var", "visit_var", "group_var", "response_var"),
  full,
  na.action = "na.omit",
  ...
)

## S3 method for class 'mrmr_tmb'

```

```

model.matrix(object, data, include = NULL, ...)

## S3 method for class 'mmrm_tmb'
terms(x, include = "response_var", ...)

## S3 method for class 'mmrm_tmb'
logLik(object, ...)

## S3 method for class 'mmrm_tmb'
formula(x, ...)

## S3 method for class 'mmrm_tmb'
vcov(object, complete = TRUE, ...)

## S3 method for class 'mmrm_tmb'
VarCorr(x, sigma = NA, ...)

## S3 method for class 'mmrm_tmb'
deviance(object, ...)

## S3 method for class 'mmrm_tmb'
AIC(object, corrected = FALSE, ..., k = 2)

## S3 method for class 'mmrm_tmb'
BIC(object, ...)

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

## S3 method for class 'mmrm_tmb'
residuals(object, type = c("response", "pearson", "normalized"), ...)

## S3 method for class 'mmrm_tmb'
simulate(
  object,
  nsim = 1,
  seed = NULL,
  newdata,
  ...,
  method = c("conditional", "marginal")
)

```

### Arguments

object	(mmrm_tmb) the fitted MMRM object.
complete	(flag) whether to include potential non-estimable coefficients.



...	mostly not used; Exception is <code>model.matrix()</code> passing ... to the default method.
<code>newdata</code>	( <code>data.frame</code> ) optional new data, otherwise data from object is used.
<code>se.fit</code>	(flag) indicator if standard errors are required.
<code>interval</code>	(string) type of interval calculation. Can be abbreviated.
<code>level</code>	(number) tolerance/confidence level.
<code>nsim</code>	(count) number of simulations to use.
<code>formula</code>	( <code>mrmr_tmb</code> ) same as object.
<code>data</code>	( <code>data.frame</code> ) object in which to construct the frame.
<code>include</code>	(character) names of variable types to include. Must be NULL or one or more of <code>c("subject_var", "visit_var", "group_var", "response_var")</code> .
<code>full</code>	(flag) indicator whether to return full model frame (deprecated).
<code>na.action</code>	(string) na action.
<code>x</code>	( <code>mrmr_tmb</code> ) same as object.
<code>sigma</code>	cannot be used (this parameter does not exist in MMRM).
<code>corrected</code>	(flag) whether corrected AIC should be calculated.
<code>k</code>	(number) the penalty per parameter to be used; default <code>k = 2</code> is the classical AIC.
<code>type</code>	(string) unscaled (response), pearson or normalized. Default is response, and this is the only type available for use with models with a spatial covariance structure.
<code>seed</code>	unused argument from <code>stats::simulate()</code> .
<code>method</code>	(string) simulation method to use. If "conditional", simulated values are sampled given the estimated covariance matrix of object. If "marginal", the variance of the estimated covariance matrix is taken into account.

### Details

`include` argument controls the variables the returned model frame will include. Possible options are "response\_var", "subject\_var", "visit\_var" and "group\_var", representing the response variable, subject variable, visit variable or group variable. character values in new data will always be factorized according to the data in the fit to avoid mismatched in levels or issues in `model.matrix`.

**Value**

Depends on the method, see Functions.

**Functions**

- `coef(mmrm_tmb)`: obtains the estimated coefficients.
- `fitted(mmrm_tmb)`: obtains the fitted values.
- `predict(mmrm_tmb)`: predict conditional means for new data; optionally with standard errors and confidence or prediction intervals. Returns a vector of predictions if `se.fit == FALSE` and `interval == "none"`; otherwise it returns a `data.frame` with multiple columns and one row per input data row.
- `model.frame(mmrm_tmb)`: obtains the model frame.
- `model.matrix(mmrm_tmb)`: obtains the model matrix.
- `terms(mmrm_tmb)`: obtains the terms object.
- `logLik(mmrm_tmb)`: obtains the attained log likelihood value.
- `formula(mmrm_tmb)`: obtains the used formula.
- `vcov(mmrm_tmb)`: obtains the variance-covariance matrix estimate for the coefficients.
- `VarCorr(mmrm_tmb)`: obtains the variance-covariance matrix estimate for the residuals.
- `deviance(mmrm_tmb)`: obtains the deviance, which is defined here as twice the negative log likelihood, which can either be integrated over the coefficients for REML fits or the usual one for ML fits.
- `AIC(mmrm_tmb)`: obtains the Akaike Information Criterion, where the degrees of freedom are the number of variance parameters (`n_theta`). If corrected, then this is multiplied with  $m / (m - n\_theta - 1)$  where  $m$  is the number of observations minus the number of coefficients, or  $n\_theta + 2$  if it is smaller than that (Hurvich and Tsai 1989; Burnham and Anderson 1998).
- `BIC(mmrm_tmb)`: obtains the Bayesian Information Criterion, which is using the natural logarithm of the number of subjects for the penalty parameter  $k$ .
- `print(mmrm_tmb)`: prints the object.
- `residuals(mmrm_tmb)`: to obtain residuals - either unscaled ('response'), 'pearson' or 'normalized'.
- `simulate(mmrm_tmb)`: simulate responses from a fitted model according to the simulation method, returning a `data.frame` of dimension  $[n, m]$  where  $n$  is the number of rows in `newdata`, and  $m$  is the number `nsim` of simulated responses.

**References**

- Hurvich CM, Tsai C (1989). "Regression and time series model selection in small samples." *Biometrika*, **76**(2), 297–307. doi:10.2307/2336663.
- Burnham KP, Anderson DR (1998). "Practical use of the information-theoretic approach." In *Model selection and inference*, 75–117. Springer. doi:10.1007/9781475729177\_3.
- Gałeczki A, Burzykowski T (2013). "Linear mixed-effects model." In *Linear mixed-effects models using R*, 245–273. Springer.

**See Also**

[mmrm\\_methods](#), [mmrm\\_tidiars](#) for additional methods.

**Examples**

```

formula <- FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID)
object <- fit_mmrmm(formula, fev_data, weights = rep(1, nrow(fev_data)))
# Estimated coefficients:
coef(object)
# Fitted values:
fitted(object)
predict(object, newdata = fev_data)
# Model frame:
model.frame(object)
model.frame(object, include = "subject_var")
# Model matrix:
model.matrix(object)
# terms:
terms(object)
terms(object, include = "subject_var")
# Log likelihood given the estimated parameters:
logLik(object)
# Formula which was used:
formula(object)
# Variance-covariance matrix estimate for coefficients:
vcov(object)
# Variance-covariance matrix estimate for residuals:
VarCorr(object)
# REML criterion (twice the negative log likelihood):
deviance(object)
# AIC:
AIC(object)
AIC(object, corrected = TRUE)
# BIC:
BIC(object)
# residuals:
residuals(object, type = "response")
residuals(object, type = "pearson")
residuals(object, type = "normalized")

```

---

```
print.cov_struct
```

*Print a Covariance Structure Object*

---

**Description**

Print a Covariance Structure Object

**Usage**

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

**Arguments**

x (cov\_struct)  
a covariance structure object.

... Additional arguments unused.

**Value**

x invisibly.

---

refit\_multiple\_optimizers

*Refitting MMRM with Multiple Optimizers*

---

**Description**

**[Experimental]**

**Usage**

```
refit_multiple_optimizers(fit, ..., control = mrmr_control(...))
```

**Arguments**

fit (mrmr\_fit)  
original model fit from `fit_single_optimizer()`.

... Additional arguments passed to `mrmr_control()`.

control (mrmr\_control)  
object.

**Value**

The best (in terms of log likelihood) fit which converged.

**Note**

For Windows, no parallel computations are currently implemented.

**Examples**

```
fit <- fit_single_optimizer(
  formula = FEV1 ~ RACE + SEX + ARMCD * AVISIT + us(AVISIT | USUBJID),
  data = fev_data,
  weights = rep(1, nrow(fev_data)),
  optimizer = "nlminb"
)
best_fit <- refit_multiple_optimizers(fit)
```

---

std_start	<i>Standard Starting Value</i>
-----------	--------------------------------

---

**Description**

Obtain standard start values.

**Usage**

```
std_start(cov_type, n_visits, n_groups, ...)
```

**Arguments**

cov_type	(string) name of the covariance structure.
n_visits	(int) number of visits.
n_groups	(int) number of groups.
...	not used.

**Details**

std\_start will try to provide variance parameter from identity matrix. However, for ar1 and ar1h the corresponding values are not ideal because the  $\rho$  is usually a positive number thus using 0 as starting value can lead to incorrect optimization result, and we use 0.5 as the initial value of  $\rho$ .

**Value**

A numeric vector of starting values.

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