

Package ‘bdsm’

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Title Bayesian Dynamic Systems Modeling

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Description Implements methods for building and analyzing models based on panel data as described in the paper by Moral-Benito (2013, <[doi:10.1080/07350015.2013.818003](https://doi.org/10.1080/07350015.2013.818003)>). The package provides functions to estimate dynamic panel data models and analyze the results of the estimation.

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| | |
|-------------|---------------------------------|
| bma_summary | <i>Summary of a model space</i> |
|-------------|---------------------------------|

Description

A summary of a given model space is prepared. This include things such as posterior inclusion probability (PIP), posterior mean and so on. This is the core function of the package, because it allows to make assessments and decisions about the parameters and models.

Usage

```
bma_summary(
  df,
  dep_var_col,
  timestamp_col,
  entity_col,
  model_space,
  exact_value = TRUE,
  model_prior = "uniform",
  run_parallel = FALSE
)
```

Arguments

| | |
|---------------|---|
| df | Data frame with data for the SEM analysis. |
| dep_var_col | Column with the dependent variable |
| timestamp_col | The name of the column with timestamps |
| entity_col | Column with entities (e.g. countries) |
| model_space | A matrix (with named rows) with each column corresponding to a model. Each column specifies model parameters. Compare with optimal_model_space |
| exact_value | Whether the exact value of the likelihood should be computed (TRUE) or just the proportional part (FALSE). Check SEM_likelihood for details. |
| model_prior | Which model prior to use. For now there are two options: 'uniform' and 'binomial-beta'. Default is 'uniform'. |
| run_parallel | If TRUE the optimization is run in parallel using the parApply function. If FALSE (default value) the base apply function is used. Note that using the parallel computing requires setting the default cluster. See README. |

Value

List of parameters describing analyzed models

Examples

```
library(magrittr)

data_prepared <- economic_growth[,1:7] %>%
  feature_standardization(timestamp_col = year, entity_col = country) %>%
  feature_standardization(timestamp_col = year, entity_col = country,
    cross_sectional = TRUE, scale = FALSE)

bma_result <- bma_summary(df = data_prepared, dep_var_col = gdp,
  timestamp_col = year, entity_col = country,
  model_space = economic_growth_ms)
```

| | |
|-----------------|-----------------------------|
| economic_growth | <i>Economic Growth Data</i> |
|-----------------|-----------------------------|

Description

Data used in Growth Empirics in Panel Data under Model Uncertainty and Weak Exogeneity (Moral-Benito, 2016, Journal of Applied Econometrics).

Usage

```
economic_growth
```

Format

economic_growth:

A data frame with 365 rows and 12 columns:

year Year

country Country ID

gdp Logarithm of GDP per capita (2000 US dollars at PP)

ish Ratio of real domestic investment to GDP

sed Stock of years of secondary education in the total population

pgrw Average growth rate of population

pop Population in millions of people

ipr Purchasing-power-parity numbers for investment goods

opem Exports plus imports as a share of GDP

gsh Ratio of government consumption to GDP

lnlex Logarithm of the life expectancy at birth

polity Composite index given by the democracy score minus the autocracy score

Source

<http://qed.econ.queensu.ca/jae/datasets/moral-benito001/>

economic_growth_bma_params

Example Approximate Summary of Parameters of Interest Based on Model Space

Description

A matrix representing the summary of parameters computed with `parameters_summary` based on the `economic_growth_ms` model space. TODO: describe the matrix properly after cleaning up the code of the function `parameters_summary`.

Usage

`economic_growth_bma_params`

Format

economic_growth_bma_params:

A double matrix with 5 rows and 8 columns

economic_growth_lik *Example Approximate Likelihoods Summary based on Model Space*

Description

A matrix representing the summary of likelihoods computed with `likelihoods_summary` based on the `economic_growth_ms` model space. The matrix contains likelihoods, standard deviations and robust standard deviations

Usage

`economic_growth_lik`

Format

`economic_growth_stds`:

A double matrix with 11 rows and 16 columns.

first row Likelihoods for the models

second row Almost $1/2 * BIC_k$ as in Raftery's Bayesian Model Selection in Social Research eq. 19.

rows 3-7 Standard deviations

rows 8-12 Robust standard deviations

economic_growth_ms *Example Model Space*

Description

A matrix representing the model space built using subset of regressors from the `economic_growth` dataset. The included regressors are `ish`, `sed`, `pgrw` and `pop`. Therefore the model space contains $2^4 = 16$ models (columns).

Usage

`economic_growth_ms`

Format

`economic_growth_ms`:

A double matrix with 51 rows and 16 columns.

economic_growth_ms_full_proj_const

Full Model Space with Constant Projection Matrix

Description

A matrix representing the model space built using all regressors from the economic_growth dataset. Therefore the model space contains $2^9 = 512$ models (columns). The same projection matrix is used for each model.

Usage

economic_growth_ms_full_proj_const

Format

economic_growth_ms_full_proj_const:

A double matrix with 106 rows and 512 columns.

Details

TODO: to avoid NaNs when computing estimates of standard deviations, the step size in the hessian function has to be increased to $1e-2$. This is most likely caused by the fact that the likelihood values are much closer to each other after the correction for the projection matrix is introduced. Hence we have to either increase the relative tolerance of the optimization algorithm or loosen the precision when computing approximate hessian.

economic_growth_ms_full_proj_var

Full Model Space with Varying Projection Matrix

Description

A matrix representing the model space built using all regressors from the economic_growth dataset. Therefore the model space contains $2^9 = 512$ models (columns). This model space generates Posterior Inclusion Probabilities which are consistent with the results presented by Moral-Benito. The original results were approximated up to the 4th decimal place. The results obtained using this model space lead to exactly the same approximations. A different projection matrix is used for each model.

Usage

economic_growth_ms_full_proj_var

Format

economic_growth_ms_full_proj_var:
A double matrix with 106 rows and 512 columns.

exogenous_matrix *Matrix with exogenous variables for SEM representation*

Description

Create matrix which contains exogenous variables used in the Simultaneous Equations Model (SEM) representation. Currently these are: dependent variable from the lowest time stamp and regressors from the second lowest time stamp. The matrix is then used to compute likelihood for SEM analysis.

Usage

```
exogenous_matrix(df, timestamp_col, entity_col, dep_var_col)
```

Arguments

| | |
|---------------|--|
| df | Data frame with data for the SEM analysis. |
| timestamp_col | Column which determines time periods. For now only natural numbers can be used as timestamps |
| entity_col | Column which determines entities (e.g. countries, people) |
| dep_var_col | Column with dependent variable |

Value

Matrix of size $N \times k+1$ where N is the number of entities considered and k is the number of chosen regressors

Examples

```
set.seed(1)
df <- data.frame(
  entities = rep(1:4, 5),
  times = rep(seq(1960, 2000, 10), each = 4),
  dep_var = stats::rnorm(20), a = stats::rnorm(20), b = stats::rnorm(20)
)
exogenous_matrix(df, times, entities, dep_var)
```

`feature_standardization`*Perform feature standarization*

Description

This function performs **feature standarization** (also known as z-score normalization), i.e. the features are centered around the mean and scaled with standard deviation.

Usage

```
feature_standardization(  
  df,  
  timestamp_col,  
  entity_col,  
  cross_sectional = FALSE,  
  scale = TRUE  
)
```

Arguments

| | |
|------------------------------|---|
| <code>df</code> | Dataframe with data that should be prepared for LIML estimation |
| <code>timestamp_col</code> | Column with timestamps (e.g. years) |
| <code>entity_col</code> | Column with entities (e.g. countries) |
| <code>cross_sectional</code> | Whether to perform feature standardization within cross sections |
| <code>scale</code> | Whether to divide by the standard deviation TRUE or not FALSE. Default is TRUE. |

Value

A dataframe with standardized features

Examples

```
df <- data.frame(  
  year = c(2000, 2001, 2002, 2003, 2004),  
  country = c("A", "A", "B", "B", "C"),  
  gdp = c(1, 2, 3, 4, 5),  
  ish = c(2, 3, 4, 5, 6),  
  sed = c(3, 4, 5, 6, 7)  
)  
  
feature_standardization(df, year, country)
```

| | |
|---------|-----------------------|
| hessian | <i>Hessian matrix</i> |
|---------|-----------------------|

Description

Creates the hessian matrix for a given likelihood function.

Usage

```
hessian(lik, theta, ...)
```

Arguments

| | |
|-------|--|
| lik | function |
| theta | kx1 matrix |
| ... | other parameters passed to lik function. |

Value

Hessian kxk matrix where k is the number of parameters included in the theta matrix

Examples

```
lik <- function(theta) {  
  return(theta[1]^2 + theta[2]^2)  
}  
  
hessian(lik, c(1, 1))
```

| | |
|------------------------|--------------------------------------|
| initialize_model_space | <i>Initialize model space matrix</i> |
|------------------------|--------------------------------------|

Description

This function builds a representation of the model space, by creating a dataframe where each column represents values of the parameters for a given model. Real value means that the parameter is included in the model. A parameter not present in the model is marked as NA.

Usage

```
initialize_model_space(  
  df,  
  timestamp_col,  
  entity_col,  
  dep_var_col,  
  init_value = 1  
)
```

Arguments

| | |
|---------------|--|
| df | Data frame with data for the SEM analysis. |
| timestamp_col | Column which determines time periods. For now only natural numbers can be used as timestamps |
| entity_col | Column which determines entities (e.g. countries, people) |
| dep_var_col | Column with dependent variable |
| init_value | Initial value for parameters present in the model. Default is 1. |

Details

Currently the set of features is assumed to be all columns which remain after excluding `timestamp_col`, `entity_col` and `dep_var_col`.

A power set of all possible exclusions of linear dependence on the given feature is created, i.e. if there are 4 features we end up with 2^4 possible models (for each model we independently decide whether to include or not a feature).

Value

matrix of model parameters

Examples

```
library(magrittr)  
  
data_prepared <- economic_growth[,1:7] %>%  
  feature_standardization(timestamp_col = year, entity_col = country) %>%  
  feature_standardization(timestamp_col = year, entity_col = country,  
    cross_sectional = TRUE, scale = FALSE)  
  
initialize_model_space(data_prepared, year, country, gdp)
```

| | |
|-----------------|--|
| join_lagged_col | <i>Dataframe with no lagged column</i> |
|-----------------|--|

Description

This function allows to turn data in the format with lagged values for a chosen column (i.e. there are two columns with the same quantity, but one column is lagged in time) into the format with just one column

Usage

```
join_lagged_col(  
  df,  
  col,  
  col_lagged,  
  timestamp_col,  
  entity_col,  
  timestep = NULL  
)
```

Arguments

| | |
|---------------|---|
| df | Dataframe with data with a column with lagged values |
| col | Column with quantity not lagged |
| col_lagged | Column with the same quantity as col, but the values are lagged in time |
| timestamp_col | Column with timestamps (e.g. years) |
| entity_col | Column with entities (e.g. countries) |
| timestep | Difference between timestamps (e.g. 10) |

Value

A dataframe with two columns merged, i.e. just one column with the desired quantity is left.

Examples

```
df <- data.frame(  
  year = c(2000, 2001, 2002, 2003, 2004),  
  country = c("A", "A", "B", "B", "C"),  
  gdp = c(1, 2, 3, 4, 5),  
  gdp_lagged = c(NA, 1, 2, 3, 4)  
)  
  
join_lagged_col(df, gdp, gdp_lagged, year, country, 1)
```

likelihoods_summary *Approximate standard deviations for the models*

Description

Approximate standard deviations are computed for the models in the given model space. Two versions are computed.

Usage

```
likelihoods_summary(
  df,
  dep_var_col,
  timestamp_col,
  entity_col,
  model_space,
  exact_value = TRUE,
  model_prior = "uniform",
  run_parallel = FALSE
)
```

Arguments

| | |
|---------------|---|
| df | Data frame with data for the SEM analysis. |
| dep_var_col | Column with the dependent variable |
| timestamp_col | The name of the column with timestamps |
| entity_col | Column with entities (e.g. countries) |
| model_space | A matrix (with named rows) with each column corresponding to a model. Each column specifies model parameters. Compare with optimal_model_space |
| exact_value | Whether the exact value of the likelihood should be computed (TRUE) or just the proportional part (FALSE). Check SEM_likelihood for details. |
| model_prior | Which model prior to use. For now there are two options: 'uniform' and 'binomial-beta'. Default is 'uniform'. |
| run_parallel | If TRUE the optimization is run in parallel using the parApply function. If FALSE (default value) the base apply function is used. Note that using the parallel computing requires setting the default cluster. See README. |

Value

Matrix with columns describing likelihood and standard deviations for each model. The first row is the likelihood for the model (computed using the parameters in the provided model space). The second row is almost $1/2 * BIC_k$ as in Raftery's Bayesian Model Selection in Social Research eq. 19 (see TODO in the code below). The third row is model posterior probability. Then there are rows with standard deviations for each parameter. After that we have rows with robust standard deviation (not sure yet what exactly "robust" means).

Examples

```

data_centered_scaled <-
  feature_standardization(df = bdsm::economic_growth[,1:7],
                          timestamp_col = year, entity_col = country)
data_cross_sectional_standardized <-
  feature_standardization(df = data_centered_scaled, timestamp_col = year,
                          entity_col = country, cross_sectional = TRUE,
                          scale = FALSE)

likelihoods_summary(df = data_cross_sectional_standardized,
                    dep_var_col = gdp, timestamp_col = year,
                    entity_col = country, model_space = economic_growth_ms)

```

| | |
|------------------|---------------------------------------|
| matrices_from_df | <i>List of matrices for SEM model</i> |
|------------------|---------------------------------------|

Description

List of matrices for SEM model

Usage

```

matrices_from_df(
  df,
  timestamp_col,
  entity_col,
  dep_var_col,
  lin_related_regressors = NULL,
  which_matrices = c("Y1", "Y2", "Z", "cur_Y2", "cur_Z", "res_maker_matrix")
)

```

Arguments

| | |
|------------------------|--|
| df | Dataframe with data for the likelihood computations. |
| timestamp_col | Column which determines time stamps. For now only natural numbers can be used. |
| entity_col | Column which determines entities (e.g. countries, people) |
| dep_var_col | Column with dependent variable |
| lin_related_regressors | Vector of strings of column names. Which subset of regressors is in non trivial linear relation with the dependent variable (dep_var_col). In other words regressors with non-zero beta parameters. |
| which_matrices | character vector with names of matrices which should be computed. Possible matrices are "Y1", "Y2", "Z", "cur_Y2", "cur_Z", "res_maker_matrix". Default is c("Y1", "Y2", "Z", "cur_Y2", "cur_Z", "res_maker_matrix") in which case all possible matrices are generated |

Value

Named list with matrices as its elements

Examples

```
matrices_from_df(economic_growth, year, country, gdp, c("pop", "sed"),
                 c("Y1", "Y2"))
```

`optimal_model_space` *Finds MLE parameters for each model in the given model space*

Description

Given a dataset and an initial value for parameters, initializes a model space with parameters equal to initial value for each model. Then for each model performs a numerical optimization and finds parameters which maximize the likelihood.

Usage

```
optimal_model_space(
  df,
  timestamp_col,
  entity_col,
  dep_var_col,
  init_value,
  exact_value = TRUE,
  run_parallel = FALSE,
  control = list(trace = 2, maxit = 10000, fnscale = -1, REPORT = 100, scale = 0.05)
)
```

Arguments

| | |
|----------------------------|--|
| <code>df</code> | Data frame with data for the SEM analysis. |
| <code>timestamp_col</code> | The name of the column with time stamps |
| <code>entity_col</code> | Column with entities (e.g. countries) |
| <code>dep_var_col</code> | Column with the dependent variable |
| <code>init_value</code> | The value with which the model space will be initialized. This will be the starting point for the numerical optimization. |
| <code>exact_value</code> | Whether the exact value of the likelihood should be computed (TRUE) or just the proportional part (FALSE). Check SEM_likelihood for details. |
| <code>run_parallel</code> | If TRUE the optimization is run in parallel using the parApply function. If FALSE (default value) the base apply function is used. Note that using the parallel computing requires setting the default cluster. See README. |
| <code>control</code> | a list of control parameters for the optimization which are passed to optim . Default is <code>list(trace = 2, maxit = 10000, fnscale = -1, REPORT = 100, scale = 0.05)</code> , but note that <code>scale</code> is used only for adjusting the <code>parscale</code> element added later in the function code. |

Value

List of parameters describing analyzed models

Examples

```
library(magrittr)

data_prepared <- economic_growth[,1:7] %>%
  feature_standardization(timestamp_col = year, entity_col = country) %>%
  feature_standardization(timestamp_col = year, entity_col = country,
    cross_sectional = TRUE, scale = FALSE)

model_space <- optimal_model_space(df = data_prepared, dep_var_col = gdp,
  timestamp_col = year, entity_col = country,
  init_value = 0.5)
```

parameters_summary *BMA summary for parameters of interest*

Description

TODO This is just the code previously present in the morel-benito.R script wrapped as a function (to get rid of the script). Well written code and docs are still needed

Usage

```
parameters_summary(
  regressors,
  bet,
  pvarh,
  pvarr,
  fy,
  fyt,
  ppmsize,
  cout,
  nts,
  pts,
  variables_n
)
```

Arguments

| | |
|------------|------|
| regressors | TODO |
| bet | TODO |
| pvarh | TODO |

| | |
|-------------|------------------|
| pvarr | TODO |
| fy | TODO |
| fyt | TODO |
| ppmsize | TODO |
| cout | TODO |
| nts | TODO (negatives) |
| pts | TODO (positives) |
| variables_n | TODO |

Value

TODO dataframe with results

regressor_names_from_params_vector

Helper function to extract names from a vector defining a model

Description

For now it is assumed that we can only exclude linear relationships between regressors and the dependent variable.

Usage

```
regressor_names_from_params_vector(params)
```

Arguments

params a vector with parameters describing the model

Details

The vector needs to have named rows, i.e. it is assumed it comes from a model space (see [initialize_model_space](#) for details).

Value

Names of regressors which are assumed to be linearly connected with dependent variable within the model described by the params vector.

Examples

```
params <- c(alpha = 1, beta_gdp = 1, beta_gdp_lagged = 1, phi_0 = 1, err_var = 1)
regressor_names_from_params_vector(params)
```

residual_maker_matrix *Residual Maker Matrix*

Description

Create residual maker matrix from a given matrix m . See article about [projection matrix](#) on the Wikipedia.

Usage

```
residual_maker_matrix(m)
```

Arguments

m Matrix

Value

$M \times M$ matrix where M is the number of rows in the m matrix.

Examples

```
residual_maker_matrix(matrix(c(1,2,3,4), nrow = 2))
```

SEM_B_matrix *Coefficients matrix for SEM representation*

Description

Create coefficients matrix for Simultaneous Equations Model (SEM) representation.

Usage

```
SEM_B_matrix(alpha, periods_n, beta = c())
```

Arguments

α numeric
 periods_n integer
 β numeric vector. Default is $c()$ for no regressors case.

Value

List with two matrices B_{11} and B_{12}

Examples

```
SEM_B_matrix(3, 4, 4:6)
```

SEM_C_matrix *Coefficients matrix for initial conditions*

Description

Create matrix for Simultaneous Equations Model (SEM) representation with coefficients placed next to initial values of regressors, dependent variable and country-specific time-invariant variables.

Usage

```
SEM_C_matrix(alpha, phi_0, periods_n, beta = c(), phi_1 = c())
```

Arguments

| | |
|-----------|--|
| alpha | numeric |
| phi_0 | numeric |
| periods_n | numeric |
| beta | numeric vector. Default is c() for no regressors case. |
| phi_1 | numeric vector. Default is c() for no regressors case. |

Value

matrix

Examples

```
alpha <- 9
phi_0 <- 19
beta <- 11:15
phi_1 <- 21:25
periods_n <- 4
SEM_C_matrix(alpha, phi_0, periods_n, beta, phi_1)
```

SEM_dep_var_matrix *Matrix with dependent variable data for SEM representation*

Description

Create matrix which contains dependent variable data used in the Simultaneous Equations Model (SEM) representation on the left hand side of the equations. The matrix contains the data for time periods greater than or equal to the second lowest time stamp. The matrix is then used to compute likelihood for SEM analysis.

Usage

```
SEM_dep_var_matrix(df, timestamp_col, entity_col, dep_var_col)
```

Arguments

| | |
|---------------|--|
| df | Data frame with data for the SEM analysis. |
| timestamp_col | Column which determines time periods. For now only natural numbers can be used as timestamps |
| entity_col | Column which determines entities (e.g. countries, people) |
| dep_var_col | Column with dependent variable |

Value

Matrix of size $N \times T$ where N is the number of entities considered and T is the number of periods greater than or equal to the second lowest time stamp.

Examples

```
set.seed(1)
df <- data.frame(
  entities = rep(1:4, 5),
  times = rep(seq(1960, 2000, 10), each = 4),
  dep_var = stats::rnorm(20), a = stats::rnorm(20), b = stats::rnorm(20)
)
SEM_dep_var_matrix(df, times, entities, dep_var)
```

| | |
|----------------|-------------------------------------|
| SEM_likelihood | <i>Likelihood for the SEM model</i> |
|----------------|-------------------------------------|

Description

Likelihood for the SEM model

Usage

```
SEM_likelihood(
  params,
  data,
  timestamp_col,
  entity_col,
  dep_var_col,
  lin_related_regressors = NULL,
  per_entity = FALSE,
  exact_value = TRUE
)
```

Arguments

| | |
|------------------------|--|
| params | Parameters describing the model. Can be either a vector or a list with named parameters. See 'Details' |
| data | Data for the likelihood computations. Can be either a list of matrices or a dataframe. If the dataframe, additional parameters are required to build the matrices within the function. |
| timestamp_col | Column which determines time stamps. For now only natural numbers can be used. |
| entity_col | Column which determines entities (e.g. countries, people) |
| dep_var_col | Column with dependent variable |
| lin_related_regressors | Which subset of columns should be used as regressors for the current model. In other words regressors are the total set of regressors and lin_related_regressors are the ones for which linear relation is not set to zero for a given model. |
| per_entity | Whether to compute overall likelihood or a vector of likelihoods with per entity value |
| exact_value | Whether the exact value of the likelihood should be computed (TRUE) or just the proportional part (FALSE). Currently TRUE adds: 1. a normalization constant coming from Gaussian distribution, 2. a term disappearing during likelihood simplification in Likelihood-based Estimation of Dynamic Panels with Predetermined Regressors by Moral-Benito (see Appendix A.1). The latter happens when transitioning from equation (47) to equation (48), in step 2: the term $\text{trace}(\text{HG}_{22})$ is dropped, because it can be assumed to be constant from Moral-Benito perspective. To get the exact value of the likelihood we have to take this term into account. |

Details

The params argument is a list that should contain the following components:

alpha scalar value which determines linear dependence on lagged dependent variable

phi_0 scalar value which determines linear dependence on the value of dependent variable at the lowest time stamp

err_var scalar value which determines classical error component (Σ_{11} matrix, σ_{ϵ}^2)

dep_vars double vector of length equal to the number of time stamps (i.e. time stamps greater than or equal to the second lowest time stamp)

beta double vector which determines the linear dependence on regressors different than the lagged dependent variable; The vector should have length equal to the number of regressors.

phi_1 double vector which determines the linear dependence on initial values of regressors different than the lagged dependent variable; The vector should have length equal to the number of regressors.

phis double vector which together with psis determines upper right and bottom left part of the covariance matrix; The vector should have length equal to the number of regressors times number of time stamps minus 1, i.e. $\text{regressors}_n * (\text{periods}_n - 1)$

psis double vector which together with phis determines upper right and bottom left part of the covariance matrix; The vector should have length equal to the number of regressors times number of

time stamps minus 1 times number of time stamps divided by 2, i.e. $\text{regressors_n} * (\text{periods_n} - 1) * \text{periods_n} / 2$

Value

The value of the likelihood for SEM model (or a part of interest of the likelihood)

Examples

```
set.seed(1)
df <- data.frame(
  entities = rep(1:4, 5),
  times = rep(seq(1960, 2000, 10), each = 4),
  dep_var = stats::rnorm(20), a = stats::rnorm(20), b = stats::rnorm(20)
)
df <-
  feature_standardization(df, timestamp_col = times, entity_col = entities)
SEM_likelihood(0.5, df, times, entities, dep_var)
```

SEM_psi_matrix

Matrix with psi parameters for SEM representation

Description

Matrix with psi parameters for SEM representation

Usage

```
SEM_psi_matrix(psis, timestamps_n, features_n)
```

Arguments

`psis` double vector with psi parameter values
`timestamps_n` number of time stamps (e.g. years)
`features_n` number of features (e.g. population size, investment rate)

Value

A matrix with `timestamps_n` rows and $(\text{timestamps_n} - 1) * \text{feature_n}$ columns. Psis are filled in row by row in a block manner, i.e. blocks of size `feature_n` are placed next to each other

Examples

```
SEM_psi_matrix(1:30, 4, 5)
```

SEM_regressors_matrix *Matrix with regressors data for SEM representation*

Description

Create matrix which contains regressors data used in the Simultaneous Equations Model (SEM) representation on the left hand side of the equations. The matrix contains regressors data for time periods greater than or equal to the second lowest time stamp. The matrix is then used to compute likelihood for SEM analysis.

Usage

```
SEM_regressors_matrix(df, timestamp_col, entity_col, dep_var_col)
```

Arguments

| | |
|---------------|--|
| df | Data frame with data for the SEM analysis. |
| timestamp_col | Column which determines time periods. For now only natural numbers can be used as timestamps |
| entity_col | Column which determines entities (e.g. countries, people) |
| dep_var_col | Column with dependent variable |

Value

Matrix of size $N \times (T-1) \times k$ where N is the number of entities considered, T is the number of periods greater than or equal to the second lowest time stamp and k is the number of chosen regressors. If there are no regressors returns NULL.

Examples

```
set.seed(1)
df <- data.frame(
  entities = rep(1:4, 5),
  times = rep(seq(1960, 2000, 10), each = 4),
  dep_var = stats::rnorm(20), a = stats::rnorm(20), b = stats::rnorm(20)
)
SEM_regressors_matrix(df, times, entities, dep_var)
```

| | |
|------------------|---|
| SEM_sigma_matrix | <i>Covariance matrix for SEM representation</i> |
|------------------|---|

Description

Create covariance matrix for Simultaneous Equations Model (SEM) representation. Only the part necessary to compute concentrated likelihood function is computed (cf. Appendix in the Moral-Benito paper)

Usage

```
SEM_sigma_matrix(err_var, dep_vars, phis = c(), psis = c())
```

Arguments

| | |
|----------|----------------|
| err_var | numeric |
| dep_vars | numeric vector |
| phis | numeric vector |
| psis | numeric vector |

Value

List with two matrices Sigma11 and Sigma12

Examples

```
err_var <- 1
dep_vars <- c(2, 2, 2, 2)
phis <- c(10, 10, 20, 20, 30, 30)
psis <- c(101, 102, 103, 104, 105, 106, 107, 108, 109, 110, 111, 112)
SEM_sigma_matrix(err_var, dep_vars, phis, psis)
```

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