Package 'diseq'

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Title Estimation Methods for Markets in Equilibrium and Disequilibrium

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Description Provides estimation methods for markets in equilibrium and disequilibrium. Specifically, it supports the estimation of an equilibrium and four disequilibrium models with both correlated and independent shocks. It also provides post-estimation analysis tools, such as aggregation and marginal effects calculations. The estimation methods are based on full information maximum likelihood techniques given in Maddala and Nelson (1974) <doi:10.2307/1914215>. They are implemented using the analytic derivative expressions calculated in Karapanagiotis (2020) <doi:10.2139/ssrn.3525622>. The equilibrium estimation constitutes a special case of a system of simultaneous equations. The disequilibrium models, instead, replace the market clearing condition with a short side rule and allow for different specifications of price dynamics.

Language en-US

URL https://github.com/pi-kappa-devel/diseq/

BugReports https://github.com/pi-kappa-devel/diseq/issues

Depends R (>= 3.5.0)

Imports bbmle (>= 1.0.20), dplyr (>= 0.7.6), magrittr (>= 1.5), MASS (>= 7.3-50), methods, rlang (>= 0.2.1), systemfit (>= 1.1), tibble (>= 1.4.2), tidyr (>= 1.0.2), Rcpp, RcppGSL, RcppParallel

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LazyData true

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Suggests ggplot2 (>= 3.0.0), knitr (>= 1.20), numDeriv (>= 2016.8.1.1), rmarkdown (>= 1.10), testthat (>= 2.0.0)

VignetteBuilder knitr

Collate 'equation_base.R' 'system_base.R' 'equation_basic.R'
'system_basic.R' 'derivatives_basic.R'
'equation_deterministic_adjustment.R'
'system_deterministic_adjustment.R'
'derivatives_deterministic_adjustment.R'
'equation_directional.R' 'system_directional.R'
'derivatives_directional.R' 'system_fiml.R'
'derivatives_fiml.R' 'equation_stochastic_adjustment.R'
'system_stochastic_adjustment.R'
'derivatives_stochastic_adjustment.R' 'diseq.R'
'model_logger.R' 'market_model.R' 'disequilibrium_model.R'
'diseq_basic.R' 'diseq_deterministic_adjustment.R'
'diseq_directional.R' 'diseq_stochastic_adjustment.R'
'equilibrium_model.R' 'model_simulation.R'

LinkingTo Rcpp, RcppGSL

SystemRequirements C++11

NeedsCompilation yes

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R topics documented:

`diseq` *Estimation of models for markets in equilibrium and disequilibrium*

Description

The `diseq` package provides tools to estimate and analyze an equilibrium and four disequilibrium models. The equilibrium model can be estimated with either two-stage least squares or with full information maximum likelihood. The methods are asymptotically equivalent. The disequilibrium models are estimated using full information maximum likelihood. All maximum likelihood models can be estimated both with independent and correlated demand and supply shocks. The disequilibrium estimation is based on Maddala and Nelson (1974) doi: 10.2307/1914215. The package is using the expressions of the gradients of the likelihoods derived in Karapanagiotis (2020) doi: 10.2139/ssrn.3525622.

Details

Usage

The easiest way to get accustomed with the functionality of the package is to check the accompanying vignettes and the README file. These can be found in the following links:

basic_usage `vignette("basic_usage",package = "diseq")`

equilibrium_assessment `vignette("equilibrium_assessment",package = "diseq")`

Market model classes:

The model hierarchy is described in the `README` file. See the documentation of the classes for initialization details.

Equilibrium model classes:

equilibrium_model Equilibrium model that can be estimated using full information maximum likelihood or two-stage least squares

Disequilibrium model classes:

diseq_basic Disequilibrium model only with a basic short side rule

diseq_directional Disequilibrium model with directional sample separation

diseq_deterministic_adjustment Disequilibrium model with deterministic price dynamics

diseq_stochastic_adjustment Disequilibrium model with stochastic price dynamics

---

disequilibrium_model-class *Equilibrium model base class*

Description

Equilibrium model base class

---

diseq_basic-class *Basic disequilibrium model with unknown sample separation.*

Description

The basic disequilibrium model consists of three equations. Two of them are the demand and supply equations. In addition, the model replaces the market clearing condition with the short side rule. The model is estimated using full information maximum likelihood.

$$D_{nt} = X'_{d,nt}\beta_d + u_{d,nt},$$
$$S_{nt} = X'_{s,nt}\beta_s + u_{s,nt},$$
$$Q_{nt} = \min\{D_{nt}, S_{nt}\}.$$

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)
```

diseq_deterministic_adjustment-class
Disequilibrium model with deterministic price dynamics.

Description

The disequilibrium model with deterministic price adjustment consists of four equations. The two market equations, the short side rule and price evolution equation. The first two equations are stochastic. The price equation is deterministic. The sample is separated based on the sign of the price changes as in the diseq_directional model. The model is estimated using full information maximum likelihood.

$$D_{nt} = X'_{d,nt}\beta_d + P_{nt}\alpha_d + u_{d,nt},$$

$$S_{nt} = X'_{s,nt}\beta_s + P_{nt}\alpha_s + u_{s,nt},$$

$$Q_{nt} = \min\{D_{nt}, S_{nt}\},$$

$$\Delta P_{nt} = \frac{1}{\gamma}\left(D_{nt} - S_{nt}\right).$$

Examples

```
simulated_data <- simulate_model_data(
  "diseq_deterministic_adjustment", 500, 3, # model type, observed entities and time points
  -0.9, 8.9, c(0.03, -0.02), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(0.05, 0.02), # supply coefficients
  1.4 # price adjustment coefficient
)

# initialize the model
model <- new(
  "diseq_deterministic_adjustment", # model type
  c("id", "date"), "date", "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)
```

diseq_directional-class
Directional disequilibrium model with sample separation.

Description

The directional disequilibrium model consists of three equations and a separation rule. The market is described by a linear demand, a linear supply equation and the short side rule. The separation rule splits the sample into regimes of excess supply and excess demand. If a price change is positive at the time point of the observation, then the observation is classified as being in an excess demand regime. Otherwise, it is assumed that it represents an excess supply state. The model is estimated using full information maximum likelihood.

$$D_{nt} = X'_{d,nt}\beta_d + u_{d,nt},$$

$$S_{nt} = X'_{s,nt}\beta_s + u_{s,nt},$$

$$Q_{nt} = \min\{D_{nt}, S_{nt}\},$$

$$\Delta P_{nt} \geq 0 \Longrightarrow D_{nt} \geq S_{nt}.$$

Examples

```
simulated_data <- simulate_model_data(
  "diseq_directional", 500, 3, # model type, observed entities, observed time points
  -0.2, 4.3, c(0.03, 0.02), c(0.03, 0.01), # demand coefficients
  0.0, 4.0, c(0.03), c(0.05, 0.02) # supply coefficients
)

# in the directional model prices cannot be included in both demand and supply
model <- new(
  "diseq_directional", # model type
  c("id", "date"), "date", "Q", "P", # keys, time point, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)
```

---

diseq_stochastic_adjustment-class

Disequilibrium model with stochastic price dynamics.

Description

The disequilibrium model with stochastic price adjustment is described by a system of four equations. Three of of them form a stochastic linear system of market equations equations coupled with a stochastic price evolution equation. The fourth equation is the short side rule. In contrast to the deterministic counterpart, the model does not impose any separation rule on the sample. It is estimated using full information maximum likelihood.

$$D_{nt} = X'_{d,nt}\beta_d + P_{nt}\alpha_d + u_{d,nt},$$

$$S_{nt} = X'_{s,nt}\beta_s + P_{nt}\alpha_s + u_{s,nt},$$
$$Q_{nt} = \min\{D_{nt}, S_{nt}\},$$
$$\Delta P_{nt} = \frac{1}{\gamma}\left(D_{nt} - S_{nt}\right) + X'_{p,nt}\beta_p + u_{p,nt}.$$

Examples

```
simulated_data <- simulate_model_data(
  "diseq_stochastic_adjustment", 500, 3, # model type, observed entities and time points
  -0.1, 9.8, c(0.3, -0.2), c(0.6, 0.1), # demand coefficients
  0.1, 5.1, c(0.9), c(-0.5, 0.2), # supply coefficients
  1.4, 3.1, c(0.8) # price adjustment coefficient
)

# initialize the model
model <- new(
  "diseq_stochastic_adjustment", # model type
  c("id", "date"), "date", "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", "Xp1", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)
```

---

equation_base-class *Equation base class*

Description

Equation base class

---

equation_basic-class *Basic disequilibrium model equation class*

Description

Basic disequilibrium model equation class

---

equation_deterministic_adjustment-class *Deterministic adjustment disequilibrium model equation class*

Description

Deterministic adjustment disequilibrium model equation class

`equation_directional-class`
Directional disequilibrium model equation class

Description

Directional disequilibrium model equation class

`equation_stochastic_adjustment-class`
Stochastic adjustment disequilibrium model equation class

Description

Stochastic adjustment disequilibrium model equation class

`equilibrium_model-class`
Equilibrium model estimated using full-information maximum likelihood.

Description

The equilibrium model consists of thee equations. The demand, the supply and the market clearing equations. The model can be estimated using both full information maximum likelihood and two-stage least squares.

$$D_{nt} = X'_{d,nt}\beta_d + P_{nt}\alpha_d + u_{d,nt},$$

$$S_{nt} = X'_{s,nt}\beta_s + P_{nt}\alpha_s + u_{s,nt},$$

$$Q_{nt} = D_{nt} = S_{nt}.$$

A necessary identification condition is that there is at least one control that is exclusively part of the demand and one control that is exclusively part of the supply equation. In the first stage of the two-stage least square estimation, prices are regressed on remaining controls from both the demand and supply equations. In the second stage, the demand and supply equation is estimated using the fitted prices instead of the observed.

Examples

```
simulated_data <- simulate_model_data(
    "equilibrium_model", 500, 3, # model type, observed entities and time points
    -0.9, 14.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
    0.9, 3.2, c(0.3), c(0.5, 0.02) # supply coefficients
)

# initialize the model
model <- new(
    "equilibrium_model", # model type
    c("id", "date"), "Q", "P", # keys, quantity, and price variables
    "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
    simulated_data, # data
    use_correlated_shocks = TRUE # allow shocks to be correlated
)
```

`estimate` *Model estimation.*

Description

All models are estimated using full information maximum likelihood. The `equilibrium_model` can also be estimated using two-stage least squares. The maximum likelihood estimation is based on `mle2`. If no starting values are provided, the function uses linear regression estimates as initializing values. The default optimization method is BFGS. For other alternatives see `mle2`. The implementation of the two-stage least square estimation of the `equilibrium_model` is based on `systemfit`.

Usage

```
estimate(object, ...)

## S4 method for signature 'market_model'
estimate(
  object,
  use_numerical_gradient = FALSE,
  use_numerical_hessian = TRUE,
  use_heteroscedasticity_consistent_errors = FALSE,
  cluster_errors_by = NA,
  ...
)

## S4 method for signature 'equilibrium_model'
estimate(object, method = "BFGS", ...)
```

Arguments

object	A model object.
...	Named parameter used in the model's estimation. These are passed further down to the estimation call. For the equilibrium_model model, the parameters are passed to systemfit, if the method is set to 2SLS, or to mle2 for any other method. For the rest of the models, the parameters are passed to mle2.
use_numerical_gradient	If true, the gradient is calculated numerically. By default, all the models are estimated using the analytic expressions of their likelihoods' gradients.
use_numerical_hessian	If true, the variance-covariance matrix is calculated using the numerically approximated Hessian. Calculated Hessians are only available for the basic and directional models.
use_heteroscedasticity_consistent_errors	If true, the variance-covariance matrix is calculated using heteroscedasticity adjusted (Huber-White) standard errors.
cluster_errors_by	A vector with names of variables belonging in the data of the model. If the vector is supplied, the variance-covariance matrix is calculated by grouping the score matrix based on the passed variables.
method	A string specifying the estimation method. When the passed value is among Nelder-Mead, BFGS, CG, L-BFGS-B, SANN, and Brent, the model is estimated using full information maximum likelihood based on mle2 functionality. When 2SLS is supplied, the model is estimated using two-stage least squares based on systemfit. In this case, the function returns a list containing the first and second stage estimates. The default value is BFGS.

Value

The object that holds the estimation result.

Functions

- estimate,market_model-method: Full information maximum likelihood estimation.
- estimate,equilibrium_model-method: Equilibrium model estimation.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
```

```
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate the model object (by default the maximum optimization is using BFGS)
est <- estimate(model)

# estimate the model by specifying the optimization details that are passed to the optimizer.
est <- estimate(model, control = list(reltol = 1e-4), method = "BFGS")
```

`get_aggregate_demand` *Demand aggregation.*

Description

Calculates the sample's aggregate demand at the passed set of parameters.

Usage

```
get_aggregate_demand(object, parameters)

## S4 method for signature 'market_model'
get_aggregate_demand(object, parameters)
```

Arguments

`object`	A model object.
`parameters`	A vector of model's parameters.

Value

The sum of the demanded quantities evaluated at the given parameters.

See Also

get_demanded_quantities

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)
```

```
# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate the model object
est <- estimate(model)

# get estimated aggregate demand
get_aggregate_demand(model, est@coef)
```

`get_aggregate_supply` *Supply aggregation.*

Description

Calculates the sample's aggregate supply at the passed set of parameters.

Usage

```
get_aggregate_supply(object, parameters)

## S4 method for signature 'market_model'
get_aggregate_supply(object, parameters)
```

Arguments

`object`	A model object.
`parameters`	A vector of model's parameters.

Value

The sum of the supplied quantities evaluated at the given parameters.

See Also

get_supplied_quantities

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate the model object
est <- estimate(model)

# get estimated aggregate supply
get_aggregate_supply(model, est@coef)
```

---

get_demanded_quantities

Demanded quantities.

Description

Calculates the demanded quantity for each observation.

Usage

```
get_demanded_quantities(object, parameters)

## S4 method for signature 'market_model'
get_demanded_quantities(object, parameters)
```

Arguments

`object`	A model object.
`parameters`	A vector of model's parameters.

Value

A vector with the demanded quantities evaluated at the given parameter vector.

See Also

get_aggregate_demand

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate the model object
est <- estimate(model)

# get estimated demanded quantities
demq <- get_demanded_quantities(model, est@coef)
```

---

get_demand_descriptives

Demand descriptive statistics

Description

Calculates and returns basic descriptive statistics for the model's demand data. Factor variables are excluded from the calculations.

Usage

```
get_demand_descriptives(object)

## S4 method for signature 'market_model'
get_demand_descriptives(object)
```

Arguments

object	A model object.

Value

A data `tibble` containing descriptive statistics.

`get_marginal_effect_at_mean` *Marginal effects at the mean.*

Description

Returns the estimated marginal effects evaluated at the mean of a variable.

Usage

```
get_marginal_effect_at_mean(object, estimation, variable)

## S4 method for signature 'disequilibrium_model'
get_marginal_effect_at_mean(object, estimation, variable)
```

Arguments

`object`	A disequilibrium model object.
`estimation`	A model estimation object (i.e. a `mle2` object).
`variable`	Variable name for which the effect is calculated.

Value

The marginal effect of the passed variable evaluated at the estimated mean.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate a model object
est <- estimate(model)
```

```
# get the marginal effects at the mean of variable "X1"
get_marginal_effect_at_mean(model, est, "X1")
```

`get_mean_marginal_effect` *Mean marginal effects*

Description

Returns the average estimated marginal effect a variable.

Usage

```
get_mean_marginal_effect(object, estimation, variable)

## S4 method for signature 'disequilibrium_model'
get_mean_marginal_effect(object, estimation, variable)
```

Arguments

`object`	A disequilibrium model object.
`estimation`	A model estimation object (i.e. a `mle2` object).
`variable`	Variable name for which the effect is calculated.

Value

The mean of the estimated marginal effects of the passed variable.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate a model object
```

```
est <- estimate(model)

# get the mean marginal effects of variable "X1"
get_mean_marginal_effect(model, est, "X1")
```

`get_model_description` *Model description.*

Description

A unique identifying string for the model.

Usage

```
get_model_description(object)

## S4 method for signature 'market_model'
get_model_description(object)
```

Arguments

`object`	A model object.

Value

A string representation of the model.

`get_normalized_shortages` *Normalized shortages.*

Description

Returns the shortages normalized by the variance of the difference of the shocks at a given point.

Usage

```
get_normalized_shortages(object, parameters)

## S4 method for signature 'disequilibrium_model'
get_normalized_shortages(object, parameters)
```

Arguments

`object`	A disequilibrium model object.
`parameters`	A vector of parameters at which the shortages are evaluated.

Value

A vector with the estimated normalized shortages.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate a model object
est <- estimate(model)

# get estimated normalized shortages
nshort <- get_normalized_shortages(model, est@coef)
```

---

get_number_of_observations

Number of observations.

Description

Returns the number of observations that are used by an initialized model. The number of used observations may differ from the numbers of observations of the data set that was passed to the model's initialization.

Usage

```
get_number_of_observations(object)

## S4 method for signature 'market_model'
get_number_of_observations(object)
```

Arguments

object	A model object.

Value

The number of used observations.

`get_prefixed_const_variable` *Constant coefficient variable name.*

Description

The constant coefficient name is constructed by concatenating the equation prefix with `CONST`.

Usage

```
get_prefixed_const_variable(object)

## S4 method for signature 'equation_base'
get_prefixed_const_variable(object)
```

Arguments

`object`	An equation object.

Value

The constant coefficient name.

`get_prefixed_control_variables` *Control variable names.*

Description

The controls of the equation are the independent variables without the price variable. Their names are constructed by concatenating the equation prefix with the name of the price column.

Usage

```
get_prefixed_control_variables(object)
```

Arguments

`object`	An equation object.

Value

A vector with the control variable names.

`get_prefixed_independent_variables` *Independent variable names.*

Description

The names of the independent variables are constructed by concatenating the equation prefix with the column names of the data `tibble`.

Usage

```
get_prefixed_independent_variables(object)
```

Arguments

`object`	An equation object.

Value

A vector with the independent variable names.

`get_prefixed_price_variable` *Price coefficient variable name.*

Description

The price coefficient name is constructed by concatenating the equation prefix with the name of the price column.

Usage

```
get_prefixed_price_variable(object)

## S4 method for signature 'equation_base'
get_prefixed_independent_variables(object)

## S4 method for signature 'equation_base'
get_prefixed_price_variable(object)

## S4 method for signature 'equation_base'
get_prefixed_control_variables(object)
```

Arguments

`object`	An equation object.

Value

The price coefficient variable name.

`get_prefixed_variance_variable` *Variance variable name.*

Description

The variance variables is constructed by concatenating the equation prefix with "VARIANCE".

Usage

```
get_prefixed_variance_variable(object)

## S4 method for signature 'equation_base'
get_prefixed_variance_variable(object)
```

Arguments

`object`	An equation object.

Value

The variable name for the variance of the shock of the equation.

`get_relative_shortages` *Relative shortages.*

Description

Returns the shortages normalized by the supplied quantity at a given point.

Usage

```
get_relative_shortages(object, parameters)

## S4 method for signature 'disequilibrium_model'
get_relative_shortages(object, parameters)
```

Arguments

`object`	A disequilibrium model object.
`parameters`	A vector of parameters at which the shortages are evaluated.

Value

A vector with the estimated normalized shortages.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate a model object
est <- estimate(model)

# get estimated relative shortages
rshort <- get_relative_shortages(model, est@coef)
```

---

get_shortage_probabilities

Shortage probabilities.

Description

Returns the shortage probabilities, i.e. the probabilities of an observation coming from an excess demand regime, at the given point.

Usage

```
get_shortage_probabilities(object, parameters)

## S4 method for signature 'disequilibrium_model'
get_shortage_probabilities(object, parameters)
```

Arguments

object	A disequilibrium model object.
parameters	A vector of parameters at which the shortage probabilities are evaluated.

Value

A vector with the estimated shortage probabilities.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate a model object
est <- estimate(model)

# get the estimated shortage probabilities
probs <- get_shortage_probabilities(model, est@coef)
```

---

get_supplied_quantities *Supplied quantities.*

Description

Calculates the supplied quantity for each observation.

Usage

```
get_supplied_quantities(object, parameters)

## S4 method for signature 'market_model'
get_supplied_quantities(object, parameters)
```

Arguments

object	A model object.
parameters	A vector of model's parameters.

Value

A vector with the supplied quantities evaluated at the given parameter vector.

See Also

get_aggregate_supply

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate the model object
est <- estimate(model)

# get estimated supplied quantities
supq <- get_supplied_quantities(model, est@coef)
```

---

get_supply_descriptives

Supply descriptive statistics

Description

Calculates and returns basic descriptive statistics for the model's supply data. Factor variables are excluded from the calculations.

Usage

```
get_supply_descriptives(object)

## S4 method for signature 'market_model'
get_supply_descriptives(object)
```

Arguments

object	A model object.

Value

A data tibble containing descriptive statistics.

---

has_shortage *Checks if an observation is in a shortage stage.*

Description

Returns TRUE for the indices at which the shortages of the market are non-negative, i.e. the market is in an excess demand state. Returns FALSE for the remaining indices. The evaluation of the shortages is performed using the passed parameter vector.

Usage

```
has_shortage(object, parameters)

## S4 method for signature 'disequilibrium_model'
has_shortage(object, parameters)
```

Arguments

object	A disequilibrium model object.
parameters	A vector of parameters at which the shortage probabilities are evaluated.

Value

A vector of Boolean values indicating observations with shortages.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
```

```
)

# estimate a model object
est <- estimate(model)

# get the indices of  estimated shortages
has_short <- has_shortage(model, est@coef)
```

---

initialize_market_model *Model initialization*

Description

Model initialization

Usage

```
## S4 method for signature 'diseq_basic'
initialize(
  .Object,
  key_columns,
  quantity_column,
  price_column,
  demand_specification,
  supply_specification,
  data,
  use_correlated_shocks = TRUE,
  verbose = 0
)

## S4 method for signature 'diseq_deterministic_adjustment'
initialize(
  .Object,
  key_columns,
  time_column,
  quantity_column,
  price_column,
  demand_specification,
  supply_specification,
  data,
  use_correlated_shocks = TRUE,
  verbose = 0
)

## S4 method for signature 'diseq_directional'
initialize(
```

```
  .Object,
  key_columns,
  time_column,
  quantity_column,
  price_column,
  demand_specification,
  supply_specification,
  data,
  use_correlated_shocks = TRUE,
  verbose = 0
)

## S4 method for signature 'diseq_stochastic_adjustment'
initialize(
  .Object,
  key_columns,
  time_column,
  quantity_column,
  price_column,
  demand_specification,
  supply_specification,
  price_specification,
  data,
  use_correlated_shocks = TRUE,
  verbose = 0
)

## S4 method for signature 'equilibrium_model'
initialize(
  .Object,
  key_columns,
  quantity_column,
  price_column,
  demand_specification,
  supply_specification,
  data,
  use_correlated_shocks = TRUE,
  verbose = 0
)
```

Arguments

.Object	The object to be Constructed.
key_columns	Key columns of the data set.
quantity_column	The quantity variable of the data set.
price_column	The price variable of the data set.

demand_specification
: A formula representation of the right hand side of the demand equation.

supply_specification
: A formula representation of the right hand side of the supply equation.

data
: The data set.

use_correlated_shocks
: Should the model be estimated using correlated shocks?

verbose
: Verbosity level.

time_column
: The time column of the data set.

price_specification
: A formula representation of the price equation.

Details

Common initialization

Variable construction

The constructor prepares the model's variables using the passed specifications. The specification strings are expected to follow the syntax of `formula`. The construction of the model's data uses the variables that are extracted by these specification. The demand variables are extracted by a formula that uses the `quantity_column` on the left hand side and the `demand_specification` on the right hand side of the formula. The supply variables are constructed by the and the `quantity_column` and the `supply_specification`. In the case of the `diseq_stochastic_adjustment` model, the price dynamics' variables are extracted using the `quantity_column` and the `price_specification`

Data preparation

1. If the passed data set contains rows with NA values, they are dropped. If the verbosity level allows warnings, a warning is emitted reporting how many rows were dropped.

2. After dropping the rows, factor levels may be invalidated. If needed the constructor readjusts the factor variables by removing the unobserved levels. Factor indicators and interaction terms are automatically created.

3. The primary column is constructed by pasting the values of the key_columns.

4. In the case of the `diseq_directional`, `diseq_deterministic_adjustment`, and the `diseq_stochastic_adjustment` models, a column with lagged prices is constructed. Since lagged prices are unavailable for the observation of the first time points, these observations are dropped. If the verbosity level allows the emission of information messages, the constructor prints the number of dropped observations.

5. In the case of the `diseq_directional`, and the `diseq_stochastic_adjustment` models, a column with price differences is created.

Value

The initialized model.

Functions

- `initialize,diseq_basic-method`: Basic disequilibrium model base constructor
- `initialize,diseq_deterministic_adjustment-method`: Disequilibrium model with deterministic price adjustment constructor

- `initialize,diseq_directional-method`: Directional disequilibrium model base constructor
- `initialize,diseq_stochastic_adjustment-method`: Disequilibrium model with stochastic price adjustment constructor
- `initialize,equilibrium_model-method`: Full information maximum likelihood model constructor

---

`market_model-class` *Model base class*

Description

Model base class

Slots

`logger` Logger object.

`key_columns` Vector of column names that uniquely identify data records. For panel data this vector should contain an entity and a time point identifier.

`time_column` Column name for the time point data.

`explanatory_columns` Vector of explanatory column names for all model's equations.

`data_columns` Vector of model's data column names. This is the union of the quantity, price and explanatory columns.

`columns` Vector of primary key and data column names for all model's equations.

`model_tibble` Model data `tibble`.

`model_type_string` Model type string description.

`system` Model's system of equations.

`maximize_log_likelihood`
Maximize the log-likelihood.

Description

Maximizes the log-likelihood using the GSL implementation of the BFGS algorithm. This function is primarily intended for advanced usage. The `estimate` functionality is a fast, analysis-oriented alternative. If the GSL is not available, the function returns a trivial result list with status set equal to -1. If the C++17 execution policies are available, the implementation of the optimization is parallelized.

Usage

```
maximize_log_likelihood(
  object,
  start,
  step,
  objective_tolerance,
  gradient_tolerance
)

## S4 method for signature 'equilibrium_model'
maximize_log_likelihood(
  object,
  start,
  step,
  objective_tolerance,
  gradient_tolerance
)
```

Arguments

`object`	A model object.
`start`	Initializing vector.
`step`	Optimization step.
`objective_tolerance`	Objective optimization tolerance.
`gradient_tolerance`	Gradient optimization tolerance.

Value

A list with the optimization output.

See Also

estimate

Examples

```
simulated_data <- simulate_model_data(
    "equilibrium_model", 500, 3, # model type, observed entities, observed time points
    -0.9, 14.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
    0.9, 3.2, c(0.03), c(0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
    "equilibrium_model", # model type
    c("id", "date"), "Q", "P", # keys, quantity, and price variables
```

```
    "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
    simulated_data, # data
    use_correlated_shocks = TRUE # allow shocks to be correlated
)

# maximize the model's log-likelihood
mll <- maximize_log_likelihood(
    model,
    start = NULL, step = 1e-5,
    objective_tolerance = 1e-4, gradient_tolerance = 1e-3
)
```

`minus_log_likelihood` *Minus log-likelihood.*

Description

Returns the opposite of the log-likelihood. The likelihood functions are based on Maddala and Nelson (1974) doi: 10.2307/1914215. The likelihoods expressions that the function uses are derived in Karapanagiotis (2020) doi: 10.2139/ssrn.3525622. The function calculates the model's log likelihood by evaluating the log likelihood of each observation in the sample and summing the evaluation results.

Usage

```
minus_log_likelihood(object, parameters)

## S4 method for signature 'diseq_basic'
minus_log_likelihood(object, parameters)

## S4 method for signature 'diseq_deterministic_adjustment'
minus_log_likelihood(object, parameters)

## S4 method for signature 'diseq_directional'
minus_log_likelihood(object, parameters)

## S4 method for signature 'diseq_stochastic_adjustment'
minus_log_likelihood(object, parameters)

## S4 method for signature 'equilibrium_model'
minus_log_likelihood(object, parameters)
```

Arguments

`object`	A model object.
`parameters`	A vector of parameters at which the function is to be evaluated.

Value

The opposite of the sum of the likelihoods evaluated for each observation.

`model_logger-class` *Logger class*

Description

Logger class

Slots

`verbosity` Controls the intensity of output messages. Errors are always printed. Other than this, a value of

1 prints warnings,

2 prints basic information,

3 prints verbose information and,

4 prints debug information.

`scores` *Likelihood scores.*

Description

It calculates the gradient of the likelihood at the given parameter point for each observation in the sample. It, therefore, returns an n x k matrix, where n denotes the number of observations in the sample and k the number of estimated parameters. There order of the parameters is the same as the one that is used in the summary of the results.

Usage

```
scores(object, parameters)

## S4 method for signature 'diseq_basic'
scores(object, parameters)

## S4 method for signature 'diseq_deterministic_adjustment'
scores(object, parameters)

## S4 method for signature 'diseq_directional'
scores(object, parameters)

## S4 method for signature 'diseq_stochastic_adjustment'
scores(object, parameters)

## S4 method for signature 'equilibrium_model'
scores(object, parameters)
```

Arguments

object	A model object.
parameters	A vector with model parameters.

Value

The score matrix.

Examples

```
simulated_data <- simulate_model_data(
  "diseq_basic", 500, 3, # model type, observed entities, observed time points
  -0.9, 8.9, c(0.3, -0.2), c(-0.03, -0.01), # demand coefficients
  0.9, 4.2, c(0.03), c(-0.05, 0.02) # supply coefficients
)

# initialize the model
model <- new(
  "diseq_basic", # model type
  c("id", "date"), "Q", "P", # keys, quantity, and price variables
  "P + Xd1 + Xd2 + X1 + X2", "P + Xs1 + X1 + X2", # equation specifications
  simulated_data, # data
  use_correlated_shocks = TRUE # allow shocks to be correlated
)

# estimate the model object (by default the maximum optimization is using BFGS)
est <- estimate(model)

# Calculate the score matrix
scores <- scores(model, est@coef)
```

---

simulate_model_data *Simulate model data.*

Description

Returns a data `tibble` with simulated data from a generating process that matches the passed model string. By default, the simulated observations of the controls are drawn from a normal distribution.

Usage

```
simulate_model_data(
  model_string,
  nobs,
  tobs,
  alpha_d,
```

```
  beta_d0,
  beta_d,
  eta_d,
  alpha_s,
  beta_s0,
  beta_s,
  eta_s,
  gamma,
  beta_p0,
  beta_p,
  sigma_d = 1,
  sigma_s = 1,
  sigma_p = 1,
  rho_ds = 0,
  rho_dp = 0,
  rho_sp = 0,
  seed = NA_integer_,
  price_generator = function(nobs) stats::rnorm(n = nobs, mean = 2.5, sd = 0.5),
  control_generator = function(nobs) stats::rnorm(n = nobs, mean = 2.5, sd = 0.5)
)

## S4 method for signature 'ANY'
simulate_model_data(
  model_string,
  nobs,
  tobs,
  alpha_d,
  beta_d0,
  beta_d,
  eta_d,
  alpha_s,
  beta_s0,
  beta_s,
  eta_s,
  gamma,
  beta_p0,
  beta_p,
  sigma_d = 1,
  sigma_s = 1,
  sigma_p = 1,
  rho_ds = 0,
  rho_dp = 0,
  rho_sp = 0,
  seed = NA_integer_,
  price_generator = function(nobs) stats::rnorm(n = nobs, mean = 2.5, sd = 0.5),
  control_generator = function(nobs) stats::rnorm(n = nobs, mean = 2.5, sd = 0.5)
)
```

Arguments

model_string	Model type. It should be among equilibrium_model, diseq_basic, diseq_directional, diseq_deterministic_adjustment, and diseq_stochastic_adjustment.
nobs	Number of simulated entities.
tobs	Number of simulated dates.
alpha_d	Price coefficient of demand.
beta_d0	Constant coefficient of demand.
beta_d	Coefficients of exclusive demand controls.
eta_d	Demand coefficients of common controls.
alpha_s	Price coefficient of supply.
beta_s0	Constant coefficient of supply.
beta_s	Coefficients of exclusive supply controls.
eta_s	Supply coefficients of common controls.
gamma	Price equation's stability factor.
beta_p0	Price equation's constant coefficient.
beta_p	Price equation's control coefficients.
sigma_d	Demand shock's standard deviation.
sigma_s	Supply shock's standard deviation.
sigma_p	Price equation shock's standard deviation.
rho_ds	Demand and supply shocks' correlation coefficient.
rho_dp	Demand and price shocks' correlation coefficient.
rho_sp	Supply and price shocks' correlation coefficient.
seed	Pseudo random number generator seed.
price_generator	Pseudo random number generator callback for prices.
control_generator	Pseudo random number generator callback for non-price controls.

Value

The simulated data.

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