Package: BLPestimatoR (via r-universe)

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Type Package

Title Performs a BLP Demand Estimation

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Description Provides the estimation algorithm to perform the demand estimation described in Berry, Levinsohn and Pakes (1995) <DOI:10.2307/2171802> . The routine uses analytic gradients and offers a large number of implemented integration methods and optimization routines.

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LazyData TRUE

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`BLP_data` *Prepares data and parameters related to the BLP algorithm for estimation.*

Description

Prepares data and parameters related to the BLP algorithm for estimation.

Usage

```
BLP_data(
  model,
  market_identifier,
  product_identifier,
  par_delta,
  group_structure = NULL,
  additional_variables = NULL,
  productData,
  demographic_draws,
  integration_accuracy,
  integration_method,
  integration_draws,
  integration_weights,
  integration_seed,
  blp_inner_tol = 1e-09,
```

```
  blp_inner_maxit = 10000
)
```

Arguments

model	the model to be estimated in R's formula syntax,
market_identifier	character specifying the market identifier (variable name must be included in productData),
product_identifier	character specifying the product identifier (variable name must be included in productData),
par_delta	optional: numeric vector with values for the mean utility (variable name must be included in productData),
group_structure	optional: character specifying a group structure for clustered standard erros (variable name must be included in productData),
additional_variables	optional: character vector specifying variables you want to keep for later analysis (variable names must be included in productData)
productData	data.frame with product characteristics,
demographic_draws	optional: list with demographic draws for each market to consider observed heterogeneity (see details),
integration_accuracy	integer specifying integration accuracy,
integration_method	character specifying integration method,
integration_draws	numeric matrix of manually provided integration draws (see details),
integration_weights	numeric vector of manually provided integration weights,
integration_seed	seed for the draws of Monte Carlo based integration,
blp_inner_tol	tolerance for the contraction mapping (default: 1e-9),
blp_inner_maxit	maximum iterations for the contraction mapping (default: 10000)

Details

For any form of user provided integration draws, i.e. integration_draws (unobserved heterogeneity) or demographic_draws (observed heterogeneity), list entries must be named and contain the variable market_identifier to allow market matching. Each line in these list entries contains the draws for one market. In case of unobserved heterogeneity, list names must match the random coefficients from the model formula. The par_delta argument provides the variable name for mean utilitys. For example, in the estimation algorithm these values are used as starting guesses in the

contraction mapping. Another example is the evaluation of the GMM, which is also based on the provided mean utilitys. If you need to update `par_delta` or any other variable in the data object, use `update_BLP_data`.

Value

Returns an object of class `blp_data`.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                     x1 + x2 + x3 + x4 + x5 |
                     0+ x1 + x2 |
                     iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                      product_id = "prod_id",
                      productData = data,
                      integration_method = "MLHS" ,
                      integration_accuracy = 40,
                      integration_seed = 1)
```

---

demographicData_cereal

Draws for observed heterogeneity in Nevo's cereal example.

Description

Draws for observed heterogeneity in Nevo's cereal example.

Usage

```
demographicData_cereal
```

Format

Draws for observed heterogeneity for each demographic.

cdid market identifier,

draws_ 20 draws differing across markets.

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

`dstddelta_wrap` *Calculates derivatives of all shares with respect to all mean utilities in a given market.*

Description

Calculates derivatives of all shares with respect to all mean utilities in a given market.

Usage

```
dstddelta_wrap(blp_data, par_theta2, market, printLevel = 1)
```

Arguments

`blp_data`	data object created by the function `BLP_data`,
`par_theta2`	matrix with column and rownames providing a starting value for the optimization routine (see details),
`market`	character specifying the market in which derivatives are calculated,
`printLevel`	level of output information (default = 1)

Details

NA's in `par_theta2` entries indicate the exclusion from estimation, i.e. the coefficient is assumed to be zero. If only unobserved heterogeneity is used (no demographics), the column name of `par_theta2` must be "unobs_sd". With demographics the colnames must match the names of provided demographics (as in `demographic_draws`) and "unobs_sd". Row names of `par_theta2` must match random coefficients as specified in `model`. Constants must be named "(Intercept)".

Value

Returns a numeric matrix with derivatives. Cell in row i and col j is the derivative of share i with respect to mean utility j.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta2 <- matrix(c(0.5,2), nrow=2)
rownames(theta2) <- c("x1","x2")
colnames(theta2) <- "unobs_sd"

derivatives2 <- dstddelta_wrap(  blp_data=blp_data,
                                 par_theta2 = theta2,
                                market = 2)
```

`dstdtheta_wrap` *Calculates derivatives of all shares with respect to all non-linear parameters in a given market.*

Description

Calculates derivatives of all shares with respect to all non-linear parameters in a given market.

Usage

```
dstdtheta_wrap(blp_data, par_theta2, market, printLevel = 1)
```

Arguments

blp_data	data object created by the function BLP_data,
par_theta2	matrix with column and rownames providing a starting value for the optimization routine (see details),
market	character specifying the market in which derivatives are calculated,
printLevel	level of output information (default = 1)

Details

NA's in par_theta2 entries indicate the exclusion from estimation, i.e. the coefficient is assumed to be zero. If only unobserved heterogeneity is used (no demographics), the column name of par_theta2 must be "unobs_sd". With demographics the colnames must match the names of provided demographics (as in demographic_draws) and "unobs_sd". Row names of par_theta2 must match random coefficients as specified in model. Constants must be named "(Intercept)".

Value

Returns a numeric matrix with derivatives. Cell in row i and col j is the derivative of share i with respect to parameter j.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                     x1 + x2 + x3 + x4 + x5 |
                     0+ x1 + x2 |
                     iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                      product_id = "prod_id",
                      productData = data,
                      integration_method = "MLHS" ,
                      integration_accuracy = 40,
                      integration_seed = 1)
```

```
theta2 <- matrix(c(0.5,2), nrow=2)
rownames(theta2) <- c("x1","x2")
colnames(theta2) <- "unobs_sd"

derivatives1 <- dstdtheta_wrap(  blp_data=blp_data,
                                  par_theta2 = theta2,
                                 market = 2)
```

`dummies_cars` *Ownership matrix in BLP's car example.*

Description

Ownership matrix in BLP's car example.

Usage

```
dummies_cars
```

Format

Dummy variables.

column i 1, if product in row j is produced by firm i, 0 otherwise

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

`estimateBLP` *Performs a BLP demand estimation.*

Description

Performs a BLP demand estimation.

Usage

```
estimateBLP(
  blp_data,
  par_theta2,
  solver_method = "BFGS",
  solver_maxit = 10000,
  solver_reltol = 1e-06,
  standardError = "heteroskedastic",
  extremumCheck = FALSE,
  printLevel = 2,
  ...
)
```

Arguments

blp_data	data object created by the function BLP_data,
par_theta2	matrix with column and rownames providing a starting value for the optimization routine (see details),
solver_method	character specifying the solver method in optim (further arguments can be passed to optim by ...)
solver_maxit	integer specifying maximum iterations for the optimization routine (default=10000),
solver_reltol	integer specifying tolerance for the optimization routine (default= 1e-6),
standardError	character specifying assumptions about the GMM residual (homoskedastic , heteroskedastic (default), or cluster)
extremumCheck	if TRUE, second derivatives are checked for the existence of minimum at the point estimate (default = FALSE),
printLevel	level of output information ranges from 0 (no GMM results) to 4 (every norm in the contraction mapping)
...	additional arguments for optim

Details

NA's in par_theta2 entries indicate the exclusion from estimation, i.e. the coefficient is assumed to be zero. If only unobserved heterogeneity is used (no demographics), the column name of par_theta2 must be "unobs_sd". With demographics the colnames must match the names of provided demographics (as in demographic_draws) and "unobs_sd". Row names of par_theta2 must match random coefficients as specified in model. Constants must be named "(Intercept)".

Value

Returns an object of class "blp_est". This object contains, among others, all estimates for preference parameters and standard errors.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
```

```
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta_guesses <- matrix(c(0.5,2), nrow=2)
rownames(theta_guesses) <- c("x1","x2")
colnames(theta_guesses) <- "unobs_sd"

blp_est <- estimateBLP(blp_data =blp_data,
                       par_theta2 = theta_guesses,
                       extremumCheck = FALSE ,
                       printLevel = 1 )
summary(blp_est)
```

`get_elasticities` *Calculates elasticities for a given variable and market.*

Description

Calculates elasticities for a given variable and market.

Usage

```
get_elasticities(
  blp_data,
  share_info,
  theta_lin,
  variable,
  products,
  market,
  printLevel = 1
)
```

Arguments

`blp_data`	data object created by the function `BLP_data`,
`share_info`	object with individual and aggregated choice probabilities created by the function `getShareInfo`,
`theta_lin`	linear parameter of the variable for which elasticities are calculated for,
`variable`	character specifying a variable for which elasticities are calculated for,

products	optional: character vector of specific products,
market	character specifying the market in which elasticities are calculated
printLevel	level of output information (default = 1)

Value

Returns a matrix with elasticities. Value in row j and col i for a variable x, gives the effect of a change in product i's characteristic x on the share of product j.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                             Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                             Xexo = c("x1", "x2", "x3", "x4", "x5"),
                             Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                             true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                                    Xlin.true.price = -0.2,
                                                    Xrandom.true = rep(2,K),
                                                    instrument.effects = rep(2,10),
                                                    instrument.Xexo.effects = rep(1,5)),
                             price.endogeneity = list( mean.xi = -2,
                                                       mean.eita = 0,
                                                      cov = cbind( c(1,0.7), c(0.7,1))),
                             printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta_guesses <- matrix(c(0.5,2), nrow=2)
rownames(theta_guesses) <- c("x1","x2")
colnames(theta_guesses) <- "unobs_sd"

shareObj <- getShareInfo(  blp_data=blp_data,
                           par_theta2 = theta_guesses,
                           printLevel = 1)


get_elasticities(blp_data=blp_data,
                 share_info = shareObj ,
                 theta_lin = 1,
                 variable = "price",
```

```
products = c("4","20"),
market = 1)
```

`getDelta_wrap` *Performs a contration mapping for a given set of non-linear parameters.*

Description

Performs a contration mapping for a given set of non-linear parameters.

Usage

```
getDelta_wrap(blp_data, par_theta2, printLevel = 1)
```

Arguments

`blp_data`	data object created by the function `BLP_data`,
`par_theta2`	matrix with column and rownames providing a starting value for the optimization routine (see details),
`printLevel`	level of output information (default = 1)

Details

NA's in `par_theta2` entries indicate the exclusion from estimation, i.e. the coefficient is assumed to be zero. If only unobserved heterogeneity is used (no demographics), the column name of `par_theta2` must be "unobs_sd". With demographics the colnames must match the names of provided demographics (as in `demographic_draws`) and "unobs_sd". Row names of `par_theta2` must match random coefficients as specified in `model`. Constants must be named "(Intercept)".

Starting guesses for the contraction mapping are provided with `BLP_data`.

Value

Returns an object of class "blp_cm" with results from the contraction mapping.

`delta` resulting vector of mean utilities after the contraction mapping

`counter` inner iterations needed to convergence

`sij` market share integral evaluations for each product (in rows) for the final mean utility

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta_guesses <- matrix(c(0.5,2), nrow=2)
rownames(theta_guesses) <- c("x1","x2")
colnames(theta_guesses) <- "unobs_sd"

delta_eval <- getDelta_wrap(  blp_data=blp_data,
                              par_theta2 = theta_guesses,
                              printLevel = 4)
```

`getJacobian_wrap` *Calculating the Jacobian for a given set of non-linear parameters and mean utilities.*

Description

Calculating the Jacobian for a given set of non-linear parameters and mean utilities.

Usage

```
getJacobian_wrap(blp_data, par_theta2, printLevel = 1)
```

Arguments

blp_data	data object created by the function BLP_data,
par_theta2	matrix with column and rownames providing the evaluation point (see details),
printLevel	level of output information (default = 1)

Details

NA's in par_theta2 entries indicate the exclusion from estimation, i.e. the coefficient is assumed to be zero. If only unobserved heterogeneity is used (no demographics), the column name of par_theta2 must be "unobs_sd". With demographics the colnames must match the names of provided demographics (as in demographic_draws) and "unobs_sd". Row names of par_theta2 must match random coefficients as specified in model. Constants must be named "(Intercept)".

Value

Returns a matrix with the jacobian (products in rows, parameters in columns).

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta_guesses <- matrix(c(0.5,2), nrow=2)
rownames(theta_guesses) <- c("x1","x2")
colnames(theta_guesses) <- "unobs_sd"
```

```
jacobian <- getJacobian_wrap(blp_data=blp_data,
                             par_theta2 = theta_guesses,
                             printLevel = 2)
head(jacobian)
```

`getShareInfo` *Calculates information related to predicted shares for a given set of non-linear parameters and data.*

Description

Calculates information related to predicted shares for a given set of non-linear parameters and data.

Usage

```
getShareInfo(blp_data, par_theta2, printLevel = 1)
```

Arguments

`blp_data`	data object created by the function `BLP_data` (provides, among others, mean utilitys and integration draws),
`par_theta2`	matrix with column and rownames providing the evaluation point (see details),
`printLevel`	level of output information (default = 1)

Value

Returns a list with information related to predicted shares.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )

model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )
```

```
blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta_guesses <- matrix(c(0.5,2), nrow=2)
rownames(theta_guesses) <- c("x1","x2")
colnames(theta_guesses) <- "unobs_sd"

shares <- getShareInfo(  blp_data=blp_data,
                           par_theta2 = theta_guesses,
                           printLevel = 4)
```

`gmm_obj_wrap` *Calculating the GMM objective for a given set of non-linear parameters.*

Description

Calculating the GMM objective for a given set of non-linear parameters.

Usage

```
gmm_obj_wrap(blp_data, par_theta2, printLevel = 2)
```

Arguments

`blp_data`	data object created by the function `BLP_data`,
`par_theta2`	matrix with column and rownames providing a starting value for the optimization routine (see details),
`printLevel`	level of output information ranges from 1 (no GMM results) to 4 (every norm in the contraction mapping)

Details

NA's in `par_theta2` entries indicate the exclusion from estimation, i.e. the coefficient is assumed to be zero. If only unobserved heterogeneity is used (no demographics), the column name of `par_theta2` must be "unobs_sd". With demographics the colnames must match the names of provided demographics (as in `demographic_draws`) and "unobs_sd". Row names of `par_theta2` must match random coefficients as specified in `model`. Constants must be named "(Intercept)".

Value

Returns a list with results from the GMM evaluation.

`local_min` GMM point evaluation

`gradient` GMM derivative with respect to non-linear parameters

`delta` result of the contraction mapping

`xi` residuals of GMM evaluation

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                    x1 + x2 + x3 + x4 + x5 |
                    0+ x1 + x2 |
                    iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                     product_id = "prod_id",
                     productData = data,
                     integration_method = "MLHS" ,
                     integration_accuracy = 40,
                     integration_seed = 1)

theta_guesses <- matrix(c(0.5,2), nrow=2)
rownames(theta_guesses) <- c("x1","x2")
colnames(theta_guesses) <- "unobs_sd"

gmm <- gmm_obj_wrap(  blp_data=blp_data,
                      par_theta2 = theta_guesses,
                      printLevel = 2)
gmm$local_min
```

originalDraws_cereal *Draws for unobserved heterogeneity in Nevo's cereal example.*

Description

Draws for unobserved heterogeneity in Nevo's cereal example.

Usage

```
originalDraws_cereal
```

Format

Each list entry contains draws (unobserved heterogeneity) for a random coefficient.

cdid market identifier,

draws_ 20 draws differing across markets.

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

productData_cars *Product data of BLP's car example.*

Description

Product data of BLP's car example.

Usage

```
productData_cars
```

Format

A data frame with product data of 2217 cars in 20 markets.

share car market share,

price car price,

hpwt horsepower-weight ratio,

air 1, if car has air conditioning, 0 otherwise,

mpg market identifier,

space length times width of the car,

const constant,

id uniquely identifies a car,

cdid uniquely identifies the market of a product,

firmid uniquely identifies the firm of a product (corresponds to column number in the ownership matrix).

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

`productData_cereal` *Product data of Nevo's cereal example.*

Description

Product data of Nevo's cereal example.

Usage

```
productData_cereal
```

Format

A data frame with product data of 24 cereals in each of 94 markets.

share cereals market share,

price cereals price,

const constant,

sugar cereals sugar,

mushy cereals mushy,

cdid market identifier,

product_id uniquely identifies a product in a market,

productdummy uniquely identifies a product in a market,

IV1 1. instrument,

IV2 2. instrument,

IV3 3. instrument,

IV4 4. instrument,

IV5 5. instrument,

IV6 6. instrument,

IV7 7. instrument,

IV8 8. instrument,

IV9 9. instrument,

IV10 10. instrument,

IV11 11. instrument,

IV12 12. instrument,

IV13 13. instrument,

IV14 14. instrument,

IV15 15. instrument,

IV16 16. instrument,

IV17 17. instrument,

IV18 18. instrument,

IV19 19. instrument,

IV20 20. instrument

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

`simulate_BLP_dataset` *This function creates a simulated BLP dataset.*

Description

This function creates a simulated BLP dataset.

Usage

```
simulate_BLP_dataset(
  nmkt,
  nbrn,
  Xlin,
  Xexo,
  Xrandom,
  instruments,
  true.parameters = list(),
 price.endogeneity = list(mean.xi = -2, mean.eita = 0, cov = cbind(c(1, 0.7), c(0.7,
    1))),
  printlevel = 1,
  seed
)
```

Arguments

nmkt	number of markets
nbrn	number of products
Xlin	character vector specifying the set of linear variables
Xexo	character vector specifying the set of exogenous variables (subset of Xlin)
Xrandom	character vector specifying the set of random coefficients (subset of Xlin)
instruments	character vector specifying the set of instrumental variables
true.parameters	list with parameters of the DGP
	Xlin.true.except.price "true" linear coefficients in utility function except price
	Xlin.true.price "true" linear price coefficient in utility function
	Xrandom.true "true" set of random coefficients
	instrument.effects "true" coefficients of instrumental variables to explain endogenous price
	instrument.Xexo.effects "true" coefficients of exogenous variables to explain endogenous price
price.endogeneity	list with arguments of the multivariate normal distribution
	mean.xi controls for the mean of the error term in the utility function
	mean.eita controls for the mean of the error term in the price function
	cov controls for the covariance of xi and eita
printlevel	0 (no output) 1 (summary of generated data)
seed	seed for the random number generator

Details

The dataset is balanced, so every market has the same amount of products. Only unobserved heterogeneity can be considered. Variables that enter the equation as a Random Coefficient or exogenously must be included in the set of linear variables. The parameter.list argument specifies the "true" effect on the individual utility for each component. Prices are generated endogenous as a function of exogenous variables and instruments, where the respective effect sizes are specified in instrument.effects and instrument.Xexo.effects. Error terms xi and eita are drawn from a multivariate normal distribution, whose parameters can be set in price.endogeneity. Market shares are generated by MLHS integration rule with 10000 nodes.

Value

Returns a simulated BLP dataset.

Examples

```
K<-2 #number of random coefficients
```

theta_guesses_cereal *Parameter starting guesses for Nevo's cereal example.*

Description

Parameter starting guesses for Nevo's cereal example.

Usage

```
theta_guesses_cereal
```

Format

A matrix with 4 random coefficients (rows) and columns for 4 demographics and one unobserved heterogeneity column (5 cols in total).

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

update_BLP_data *Updates the set of linear, exogenous, random coefficient, share or mean utility variable in the data object.*

Description

Updates the set of linear, exogenous, random coefficient, share or mean utility variable in the data object.

Usage

```
update_BLP_data(data_update, blp_data)
```

Arguments

data_update	data.frame with variables to update (must contain the market_identifier and product_identifier variables as in blp_data),
blp_data	data object created by the function BLP_data

Value

Returns an object of class blp_data.

Examples

```
K<-2 #number of random coefficients
data <- simulate_BLP_dataset(nmkt = 25, nbrn = 20,
                        Xlin = c("price", "x1", "x2", "x3", "x4", "x5"),
                        Xexo = c("x1", "x2", "x3", "x4", "x5"),
                        Xrandom = paste0("x",1:K),instruments = paste0("iv",1:10),
                        true.parameters = list(Xlin.true.except.price = rep(0.2,5),
                                               Xlin.true.price = -0.2,
                                               Xrandom.true = rep(2,K),
                                               instrument.effects = rep(2,10),
                                               instrument.Xexo.effects = rep(1,5)),
                        price.endogeneity = list( mean.xi = -2,
                                                  mean.eita = 0,
                                                  cov = cbind( c(1,0.7), c(0.7,1))),
                        printlevel = 0, seed = 234234 )


model <- as.formula("shares ~  price + x1 + x2 + x3 + x4 + x5 |
                     x1 + x2 + x3 + x4 + x5 |
                     0+ x1 + x2 |
                     iv1 + iv2 + iv3 + iv4 + iv5 + iv6 + iv7 + iv8 +iv9 +iv10" )

blp_data <- BLP_data(model = model, market_identifier="cdid",
                      product_id = "prod_id",
                      productData = data,
                      integration_method = "MLHS" ,
                      integration_accuracy = 40,
                      integration_seed = 1)

new_data <- data.frame(price = seq(1,10,length.out=500),
                         x1 =  seq(2,10,length.out=500),
                         cdid = sort(rep(1:25,20)),
                         prod_id = rep(1:20,25) )
blp_data_example_updated <-update_BLP_data(blp_data = blp_data,
                                           data_update = new_data)
```

`w_guesses_cereal` *Mean utility starting guesses for Nevo's cereal example.*

Description

Mean utility starting guesses for Nevo's cereal example.

Usage

```
w_guesses_cereal
```

Format

A numeric vector of 2256 values.

Source

https://dataverse.harvard.edu/file.xhtml?persistentId=doi:10.7910/DVN/26803/SOF9FW&version=1.0

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