

# Package ‘BLPestimatorR’

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**Type** Package

**Title** Performs a BLP Demand Estimation

**Version** 0.3.4

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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](https://doi.org/10.2307/2171802)> . The routine uses analytic gradients and offers a large number of implemented integration methods and optimization routines.

**License** GPL-3

**LazyData** TRUE

**Depends** R (>= 4.2.0)

**Imports** Rcpp (>= 1.0.9), mvQuad, numDeriv, randtoolbox, Formula, stats, Matrix, methods

**LinkingTo** Rcpp, RcppArmadillo

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**VignetteBuilder** knitr

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BLP_data	<i>Prepares data and parameters related to the BLP algorithm for estimation.</i>
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## 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 errors (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 utilities. 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 utilities. 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_identifer="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	<i>Calculates derivatives of all shares with respect to all mean utilities in a given market.</i>
----------------	---

---

**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_identifiser="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	<i>Calculates derivatives of all shares with respect to all non-linear parameters in a given market.</i>
----------------	--

---

### 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_identifiser="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**

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

**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)

```

---

getDelta_wrap	<i>Performs a contraction mapping for a given set of non-linear parameters.</i>
---------------	---

---

## Description

Performs a contraction 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

si\_j 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_identfier="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)
```

**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**

<code>blp_data</code>	data object created by the function <code>BLP_data</code> ,
<code>par_theta2</code>	matrix with column and rownames providing the evaluation point (see details),
<code>printLevel</code>	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",
```



```

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_identifiers="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)

```

---

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)
```

---

gmm_obj_wrap	<i>Calculating the GMM objective for a given set of non-linear parameters.</i>
--------------	--

---

### 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	<i>Updates the set of linear, exogenous, random coefficient, share or mean utility variable in the data object.</i>
-----------------	---

---

**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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