

# Package ‘predfinitepop’

March 31, 2014

**Type** Package

**Title** Predictive Inference on Totals and Averages of Finite Populations Segmented in Planned and Unplanned Domains.

**Version** 1.0

**Date** 2014-03-31

**Copyright** Banco de Mexico

**Author** Juan Carlos Martinez-Ovando, Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez.

**Maintainer** Juan Carlos Martinez-Ovando <juan.martinez@banxico.org.mx>

**Description** This package computes Bayesian predictive inference on totals and averages of finite populations segmented in planned and unplanned domains. Inference is based on Bayesian nonparametric methods using species-sampling models (specifically the Dirichlet process).

**Acknowledgement** The authors appreciate advice from Pedro Gonzalez-Alegria and Claudia Velazquez-Villegas of Banco de Mexico.

**Needs Compilation** No

**License** GPL-3

## R topics documented:

conteo . . . . .	2
dataref . . . . .	3
datasample . . . . .	4
dginvgauss . . . . .	5
dinvgauss . . . . .	6
domnoplan . . . . .	6
domnoplan_composicion . . . . .	9
domnoplan_g0 . . . . .	11
domnoplan_totalcomp . . . . .	14
domplan . . . . .	15
domplan_g0 . . . . .	18
domplan_total . . . . .	21
g0_licitacion . . . . .	22
g0_simulacion . . . . .	23
ghat_simulacion . . . . .	24

gstar_simulacion	25
pesosDP	26
pginvgauss	27
pinvgauss	27
popcomp	28
poptot	29
predfinitepop	29
randDirichlet	30
rginvgauss	31
rinvgauss	31
unicstar	32

## Index 34

---

conteo	<i>Counts the number of individuals in <math>\mathcal{S}_j</math> and <math>\tilde{\mathcal{S}}_j</math>.</i>
--------	---

---

### Description

This function counts the number of individuals in  $\mathcal{S}_j$  and  $\tilde{\mathcal{S}}_j$ , with respect to the total number of individuals in a given planned domain  $\mathcal{P}_j$ .

### Usage

```
conteo(datos_j, N_j)
```

### Arguments

datos_j	Data matrix with individual measurements for $\mathcal{S}_j$
N_j	Number of individuals in $\mathcal{P}_j$ (the $j$ th planned domain) ( $N_j$ is assumed known)

### Details

"datos\_j" - This object should contain two columns labeled " $n_i$ " and "domplan".

"domplan" - Column vector with categories for the planned domains.

"n\_i" - Number of individuals in the  $i$ th group of individuals, and

"N\_j" - This object should include two columns labeled " $N_j$ " and "domplan".

"domplan" - Column vector with categories for the planned domains.

"N\_j" - Number of individuals in the  $j$ th planned domain, and

### Value

The function 'conteo' produces an output list with three elements:

M	Number of groups in " $\mathcal{S}_j$ "
CardS	Number of individuals in " $\mathcal{S}_j$ " (if " $n_i$ " = 1, for any " $j$ ", then $Cards = M = nrow(datos_j)$ )
CardNoS	Number of individuals out of sample

**Author(s)**

Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.

---

dataref

*Reference data set of a simulated population segmented in planned and unplanned domains.*

---

**Description**

This data set corresponds to a simulated population of 500 individuals grouped in two planned domains. Each planned domain of the population has been segmented, as well, into the three unplanned domains. Within the planned domain A, the segmentation is given as domain I (200), domain II (400) and domain III (200), whereas within planned domain B, the segmentation is given as domain I (780), domain II (400) and domain III (20). Individual measurements were simulated from log-normal and Weibull distributions with different parameterizations for each combination of planned and unplanned domains. See Martinez-Ovando et al. (2014) for further explanations.

**Usage**

```
data(dataref)
```

**Format**

A data frame with 500 observations on the following 3 variables.

id Identification code for simulated data.

y\_i Simulated individual outcomes.

domplan Identification for planned domains (domain A and B).

**Source**

Simulated data. See Martinez-Ovando et al. (2014) for further details.

**References**

- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

**Examples**

```
data(dataref)
```

---

datasample	<i>Data set of a simulated population segmented in planned and unplanned domains.</i>
------------	---

---

### Description

This is an object list with 20 simulated samples from the simulated population of 2,000 individuals grouped in two planned domains. Planned domain A is formed of 800 individuals, and domain B is composed of 1.2 thousand individuals. Each planned domain of the population has been segmented, as well, into the three unplanned domains.

### Usage

```
data(datasample)
```

### Format

List of length 20, with entries:

11. 'data.frame': 100 obs. of 8 variables with a sample of size 5 percent
12. 'data.frame': 200 obs. of 8 variables with a sample of size 10 percent
13. 'data.frame': 300 obs. of 8 variables with a sample of size 15 percent
14. 'data.frame': 400 obs. of 8 variables with a sample of size 20 percent
15. 'data.frame': 500 obs. of 8 variables with a sample of size 25 percent
16. 'data.frame': 600 obs. of 8 variables with a sample of size 30 percent
17. 'data.frame': 700 obs. of 8 variables with a sample of size 35 percent
18. 'data.frame': 800 obs. of 8 variables with a sample of size 40 percent
19. 'data.frame': 900 obs. of 8 variables with a sample of size 45 percent
110. 'data.frame': 1000 obs. of 8 variables with a sample of size 50 percent
111. 'data.frame': 1100 obs. of 8 variables with a sample of size 55 percent
112. 'data.frame': 1200 obs. of 8 variables with a sample of size 60 percent
113. 'data.frame': 1300 obs. of 8 variables with a sample of size 65 percent
114. 'data.frame': 1400 obs. of 8 variables with a sample of size 70 percent
115. 'data.frame': 1500 obs. of 8 variables with a sample of size 75 percent
116. 'data.frame': 1600 obs. of 8 variables with a sample of size 80 percent
117. 'data.frame': 1700 obs. of 8 variables with a sample of size 85 percent
118. 'data.frame': 1800 obs. of 8 variables with a sample of size 90 percent
119. 'data.frame': 1900 obs. of 8 variables with a sample of size 95 percent
120. 'data.frame': 2000 obs. of 8 variables with a sample of size 100 percent

List of variables:

**id** Individual identification code

**y\_i** Group measurement

**domnoplan** Identification code for unplanned domains

**n\_i** Group size

**domplan** Identification code for planned domains

**1** Indicator variable for unplanned domain I

**2** Indicator variable for unplanned domain II

**3** Indicator variable for unplanned domain III

## Details

Within the planned domain A, the segmentation is given as domain I (200), domain II (400) and domain III (200), whereas within planned domain B, the segmentation is given as domain I (780), domain II (400) and domain III (20). Individual measurements were simulated from log-normal and Weibull distributions with different parameterizations for each combination of planned and unplanned domains. The original sample is of size 5 percent. The upcoming samples were extrated increasing the sample size in 5 percent at a time. See Martinez-Ovando et al. (2014) for further explanations.

## Source

Simulated data. See Martinez-Ovando et al. (2014) for further details.

## References

- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

## Examples

```
data(datasample)
```

---

dginvgauss

*Density of the generalized inverse Gaussian distribution*


---

## Description

Evaluates de density of the generalized inverse Gaussian distribution

## Usage

```
dginvgauss(y, m, s, f, log = FALSE)
```

## Arguments

y	Vector of responses.
m	Vector of means.
s	Vector of dispersion parameters.
f	Vector of family parameters.
log	If TRUE, log probabilities are supplied.

**Author(s)**

Jim Lindsey <jlindsey@luc.ac.be>

**References**

Extracted from package "rmutil" Version 1.0 "Utilities for Nonlinear Regression and Repeated Measurements Models"

---

dinvgauss	<i>Inverse Gaussian distribution</i>
-----------	--------------------------------------

---

**Description**

Density of the inverse Gaussian distribution

**Usage**

```
dinvgauss(y, m, s, log = FALSE)
```

**Arguments**

y	Vector of responses.
m	Vector of means.
s	Vector of dispersion parameters.
log	If TRUE, log probabilities are supplied.

**Author(s)**

Jim Lindsey <jlindsey@luc.ac.be>

**References**

Extracted from package "rmutil" Version 1.0 "Utilities for Nonlinear Regression and Repeated Measurements Models"

---

domnoplan	<i>Finite population inference on unplanned domains using reference data.</i>
-----------	---

---

**Description**

Generates Monte Carlo samples from the predictive distribution of totals of a finite population segmented in planned and unplanned domains, along with simulations of the predictive distribution for the composition of the population between the unplanned domains.

**Usage**

```
domnoplan(datos, datos_ant, domplan_N, alphaDP, colid_D, alpha_D, inter, part, nSim)
```

### Arguments

datos	$(M \times p)$ -dimensional array with positive entries for $\mathcal{S}_j$
datos_ant	$(M \times p)$ -dimensional reference array for calibration of $G_0$
domplan_N	Matrix array with counts of individuals in each planned domain
alphaDP	$J$ -dimensional array with positive entries for the parameters of the Dirichlet process for $F_j$ (with $J$ being the number of planned domains)
colid_D	$D$ -dimensional matrix array with the columns in datos that correspond to the indicator variables of the planned domains (those indicator variables represent a portion of "datos")
alpha_D	$(J \times D)$ -dimensional array with positive entries for the parameters of the multinomial-Dirichlet component for the composition across unplanned domains. Note: Each one of the $J$ rows is a vector of composition for $\mathcal{P}_j$ divided across the $D$ unplanned domains
inter	Tuning parameter for model comparison and selection (related to calibration of $G_0$ )
part	Number of partitions for predictive cross-validation (related to calibration of $G_0$ )
nSim	Number of Monte Carlo simulated replicates of the predictive distribution

### Details

- datos : Represent the data sample of the target population, unplanned domains labeled.

It should contain the following columns:

"domplan" - Categories for planned domains.

"y\_i" - Actual individual measurements (for the moment, they must be positive and continuous) for the group of observation.

"n\_i" - Number of individuals in the group (if the unit of observation in the sample are individuals, then " $n_i$ " must be equal to 1)

- datos\_ant : Represents the data reference used to calibrate  $G_{j0}$

The data must be labeled by planned domains. It should contain the following columns:

"domplan" - Categories of planned domains

"y\_i" - Positive real and individual measurements of each group in the sample (when the units of observation are the groups, " $y_i$ " should be per capita measurement)

- domplan\_N : Represents counts (or reference population) of the target population, divided by the planned domains.

Tagged data must be labeled by domains planned. It should contain the following columns:

"domplan" - Categories for planned domains.

"N\_j" - Number of individuals in each planned domain.

### Value

The function 'domnoplan' produces:

total\_domnoplan\_sim

Matrix array of dimension " $J * (3 + 2 D) * nSim$ " with predictions for the planned domains.

Column 1 - Indicator of the planned domains

Column 2 -  $T_j$  (totals of the planned domains)

Column 3 -  $N_j$  (composition of the planned domains)

Column 4 -  $(4 + D - 1) - T_{d_j}$  (totals of unplanned domains, such that  $T_j =$

$$\sum_{j=1}^J T_{d_j} (\text{across } d))$$

Column 5 -  $(4 + D)$  to  $(3 + 2 * D) - N_{d_j}$  (composition of unplanned domains,

such that  $N_j = \sum_{j=1}^J N_{d_j} (\text{across } d))$

### Author(s)

Setgio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

### References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.

- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

### See Also

[domnoplan\\_composicion](#), [domnoplan\\_totalcomp](#).

### Examples

```
#-----
# Uses of library: "predfinitepop"
#-----

#           Read files with functions in R
rm(list=ls())

library("predfinitepop")

set.seed(12345)

#-----
# DATA
#-----
data(dataref)
data(datasample)
data(popcomp)

# Sample (with 10% of the population)

datos <- datasample[[1]]
datos_ant <- dataref
domplan_N <- popcomp[[1]]
```



```

#-----
# MODEL PARAMETERS
#-----

# a. Parameter "alpha" of the Dirichlet process
# (this is for J02 planned domains)
alphaDP <- matrix(0.03,2,1)

# b. vector with indicator variables for unplanned domains
colid_D <- t(as.matrix(c(6:8)))

# c. Matrix with parameters for unplanned domains
# (this is for J=2 planned domains)
alpha_D <- matrix(0.3,2,length(colid_D))

#-----
# CODE PARAMETERS
#-----

# A. Interval for calibration of G_0
inter <- 3

# B. Number of groups for predictive validation and model comparison
part <- 20

# C. Number of Monte Carlo replicates of the predictive distributions
nSim <- 10

#-----
# TESTING "domnoplan.R"
#-----
date.ini <- date()
domnoplan_sim <- domnoplan(datos,datos_ant,domplan_N,alphaDP,colid_D,alpha_D,inter,part,nSim)
date.fin <- date()

#
# --END --

```

---

**domnoplan\_composicion** *Generates Monte Carlo samples from the predictive distribution for the vector  $N_j$  across the  $D$  unplanned domains within a given planned domain.*

---

## Description

This function simulates Monte Carlo samples from the predictive distribution of the vector  $N_j$  across the  $D$  unplanned domains within a given planned domain  $j$ .

## Usage

```
domnoplan_composicion(datos_j,N_j,N_S_j,N_Stil_j,colid_D,alpha_D,nSim)
```

**Arguments**

<code>datos_j</code>	Data matrix with features and number of individuals in the sample $\mathcal{S}_j$
<code>N_j</code>	Number of individuals in $\mathcal{P}_j$ ( $j$ th planned domain)
<code>N_S_j</code>	Number of individuals in the sample $\mathcal{S}_j$
<code>N_Stil_j</code>	Number of individuals out the sample $\mathcal{S}_j$
<code>colid_D</code>	$D$ -dimensional matrix array with the columns in 'datos' that correspond to the indicator variables of the planned domains (those indicator variables represent a portion of 'datos')
<code>alpha_D</code>	$D$ -dimensional array with positive entieres for the parameters of the multinomial-Dirichlet component for the composition across unplanned domains (NOTE: this one makes reference to a single planned domain)
<code>nSim</code>	Number of Monte Carlo simulated replicates of the predictive distribution

**Value**

This function 'domnoplan\_composicion' produces:

<code>N_S_domnoplan</code>	Composition of the number of individuals in saample $\mathcal{S}_j$ ( $N_j^{\mathcal{S}}$ ) across the $D$ unplanned domains
<code>domnoplan_composicion_sim</code>	(1 X $D$ X nSim) matrix with Monte Carlo samples of the predictive distribution of the composition of $\tilde{\mathcal{S}}_j$

**Author(s)**

Juan Carlos Martinez-Ovando

**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

**See Also**

[domnoplan](#)

---

domnoplan_g0	<i>Finite population inference on unplanned domains using subjective <math>G_{j0}</math>.</i>
--------------	---

---

## Description

Generates Monte Carlo samples of the predictive distribution of total baseline distributions of a finite population segmented in planned and unplanned domains, using a predefined set of  $(G_{j0})_{j=1}^J$ , along with simulations of the predictive distribution for the composition of the population between the unplanned domains.

## Usage

```
domnoplan_g0(datos, domplan_N, alphaDP, colid_D, alpha_D, nSim, g0_licitacion_sal)
```

## Arguments

datos	$(M \times p)$ -dimensional array with positive entries for $\mathcal{S}_j$
domplan_N	Matrix array with counts of individuals in each planned domain
alphaDP	$J$ -dimensional array with positive entries for the parameters of the Dirichlet process for $F_j$ (with $J$ being the number of planned domains)
colid_D	$D$ -dimensional matrix array with the columns in datos that correspond to the indicator variables of the planned domains (those indicator variables represent a partion of datos)
alpha_D	$(J \times D)$ -dimensional array with positive entires for the parameters of the multinomial-Dirichlet component for the composition across unplanned domains. Note: Each one of the $J$ rows is a vector of composition for $\mathcal{P}_j$ segmented across $D$ unplanned domains
nSim	Number of Monte Carlo simulated replicates of the predictive distribution
g0_licitacion_sal	$(J \times 1)$ object list, each entry is and the object list itself associated with each $G_{j0}$ for the $J$ planned domains. The first element for arch $G_{j0}$ should be the name of the chosen distribution (see details below for alternatives), the second element should be a vector object with the parameters associated with distribution, and the third element should be its associated expectation

## Details

- datos : Represent the data sample of the target population, unplanned domains labeled.

It should contain the following columns:

"domplan" - Categories fot planned domains.

"y\_i" - Actual individual measurements/putcomes (for the moment, they must be positive) for the group of observation.

"n\_i" - Number of individuals in the group (if the unit of observation in the sample are individuals, then " $n_i$ " must be equal to 1)

- domplan\_N : Represents counts (or reference population) of the target population, divided by the planned domains.

Tagged data must be labeled by domains planned. It should contain the following columns:

"domplan" - Categories for planned domains.

"N\_j" - Number of individuals in each planned domain.

- g0\_licitacion\_sal: Chose one and only one of the distribution,

i) Gamma

ii) Weibull

iii) Lognormal

iv) Inverse-Gaussian

Parameterizations for distribution:

i) Gamma distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto x^{\alpha-1} \exp\{-x/\beta\},$$

where  $\alpha$  is the shape parameter, and  $\beta$  is the scale parameter.

ii) Weibull distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto (x/\beta)^{\alpha-1} \exp\{-(x/\beta)^\alpha\},$$

where  $\alpha$  is the shape parameter, and  $\beta$  is the scale parameter.

iii) Lognormal distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto \exp\{-(\log x - \alpha)^2 / 2\beta^2\},$$

where  $\alpha$  is the mean, and  $\beta$  is the standard deviation of the logarithm.

iv) Inverse - Gaussian distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto x^{-3/2} \exp\{-\alpha(x - \beta)^2 / 2x\beta^2\},$$

where  $\alpha$  is the shape parameter, and  $\beta$  is the mean parameter.

## Value

The function 'domnoplan' produces:

total\_domnoplan\_sim

Matrix array of dimension " $J * (3 + 2D) * nSim$ " with predictions for the planned domains.

Column 1 - Indicator of the planned domains

Column 2 -  $T_j$  (totals of the planned domains)

Column 3 -  $N_j$  (composition of the planned domains)

Column 4 -  $(4 + D - 1) - T_{d_j}$  (totals of unplanned domains, such that  $T_j =$

$$\sum_{j=1}^J T_{d_j} \text{ (across } d))$$

Column 5 -  $(4 + D)$  to  $(3 + 2 * D) - N_{d_j}$  (composition of unplanned domains,

$$\text{such that } N_j = \sum_{j=1}^J N_{d_j} \text{ (across } d))$$

## Author(s)

Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

## References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

## See Also

[domnoplan\\_composicion](#), [domnoplan\\_totalcomp](#).

## Examples

```
#-----
# Uses of library:  "predfinitepop"
#-----

#           Read files with functions in R
rm(list=ls())

library("predfinitepop")

set.seed(12345)

#-----
# DATA
#-----
data(datasample)
data(popcomp)

# Sample (with 10% of the population)

datos <- datasample[[1]]
domplan_N <- popcomp[[1]]

#-----
# MODEL PARAMETERS
#-----

# a. Parameter "alpha" of the Dirichlet process
# (this is for J02 planned domains)
alphaDP <- matrix(0.03,2,1)

# b. vector with indicator variables for unplanned domains
colid_D <- t(as.matrix(c(6:8)))

# c. Matrix with parameters for unplanned domains
# (this is for J=2 planned domains)
alpha_D <- matrix(0.3,2,length(colid_D))

#-----
# CODE PARAMETERS AND INCORPORATION OF THE DISTRIBUTION
#-----
# Incorporate each element ("g0_licitacion_in_J") according to the number of domains planned
#   g0_licitacion_in_1 <- list("distribution",teta1,esp(teta1))
```

```
#
# In this case we have two domains planned
g0_licitacion_in_1 <- list("Lognormal",c(6.198,0.840),.5)
g0_licitacion_in_2 <- list("Gamma",c(6.198,0.840),.5)

g0_licitacion_sal <- list(g0_licitacion_in_1,g0_licitacion_in_2)

# B. Number of Monte Carlo replicates of the predictive distributions
nSim <- 10

#-----
# TESTING "domnoplan.R"
#-----
date.ini <- date()
domnoplan_sim <- domnoplan_g0(datos,domplan_N,alphaDP,colid_D,alpha_D,nSim,g0_licitacion_sal)
date.fin <- date()

#
# --END --
```

---

domnoplan_totalcomp	<i>Simulates Monte Carlo samples from the predictive distribution of the vector <math>T_j</math> across the <math>D</math> unplanned domains within a given planned domain.</i>
---------------------	---

---

## Description

This function simulates Monte Carlo samples from the predictive distribution of the vector  $T_j$  across the  $D$  unplanned domains in a given planned domain  $j$ .

## Usage

```
domnoplan_totalcomp(datos_j,rho,ystar,phi,g0_licitacion_sal,N_Stil_domnoplan_j_sim,
nSim,colid_D)
```

## Arguments

datos_j	Data matrix with features and number of individuals in the sample $\mathcal{S}_j$
rho	$(U \times 2)$ -dimensional vector with weights associated with the sample ties 'ystar'
ystar	Sample ties $y_i^*$ in the sample $\mathcal{S}_j$
phi	Probability weight asociated with $G_{j0}$ (the continuos part of $\hat{G}_j$ )
g0_licitacion_sal	Object list with the continuous component in $\hat{G}_j$
N_Stil_domnoplan_j_sim	$(1 \times D \times nSim)$ matrix with Monte Carlo samples of the predictive distribution of the composition of $\tilde{S}_j$
nSim	Number of Monte Carlo simulated replicates of the predictive distribution
colid_D	$D$ -dimensional matrix array with the columns in 'datos' that correspond to the indicator variables of the planned domains (those indicator variables represent a partion of 'datos')

**Value**

The function 'domnoplan\_totalcomp' produces an object list with three entries:

T\_S\_domnoplan    Composition of  $T_j$  for the planned domain  $j$  in sample  $\mathcal{S}_j$   
T\_Stil\_domnoplan\_j  
                    (1 X  $D$  X nSim)-dimensional array with simulated samples of the convolution  
                    for  $T^{\tilde{\mathcal{S}}_j}$   
domnoplan\_totalcomp\_sim  
                    (1 X  $D$  X nSim)-dimensional array with samples from the predictive distribution  
                    of the composition of  $\tilde{\mathcal{S}}_j$

**Author(s)**

Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

**See Also**

[domnoplan.](#)

---

domplan	<i>Finite population inference on unplanned domains using reference data.</i>
---------	---

---

**Description**

Generates Monte Carlo samples of the predictive distribution of totals of a finite population segmented in planned domains.

**Usage**

```
domplan(datos,datos_ant,domplan_N,alphaDP,inter,part,nSim)
```

**Arguments**

datos                    ( $M \times p$ )-dimensional array with positive entries for  $\mathcal{S}_j$   
datos\_ant                ( $M \times p$ )-dimensional reference array for calibration of  $G_0$   
domplan\_N               Matrix array with counts of individuals in each planned domain  
alphaDP                  $J$ -dimensional array with positive entries for the parameters of the Dirichlet process for  $F_j$  (with  $J$  being the number of planned domains)

inter	Tunning parameter for model comparison and selection (related to calibration of $G_0$ )
part	Number of partitions for predictive cross-validation (related to calibration of $G_0$ )
nSim	Number of Monte Carlo simulated replicates of the predictive distribution

### Details

- datos : Represents the data sample of the target population, unplanned domains labeled.

It should contain the following columns:

"domplan" - Categories fot planned domains.

"y\_i" - Actual individual measurements(for the moment, they must be positive) for the group of observation.

"n\_i" - Number of individuals in the group (if the unit of observation in the sample are individuals, then " $n_i$ " must be equal to 1)

- datos\_ant : Represents the data of reference used to calibrate  $G_0$ .

The data must be labeled by domains planned. It should contain the following columns:

"domplan" - Planned categories of domains

"y\_i" - Positive real and individual measurements of each group in the sample (when the units of observation are groups, " $y_i$ " should be a per capita measurement)

- domplan\_N : Represents counts (or reference population) of the target population, divided by the planned domains.

Tagged data must be labeled by domains planned. It should contain the following columns:

"domplan" - Categories fot planned domains.

"N\_j" - Number of individuals in each planned domain.

### Value

The function 'domplan' produces:

total\_domnoplan\_sim

Matrix array of dimension " $J * (3 + 2 D) * nSim$ " with predictions for the planned domains.

Column 1 - Indicator of the planned domains

Column 2 -  $T_j$  (totals of the planned domains)

Column 3 -  $N_j$  (composition of the planned domains)

### Author(s)

Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

### References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.

- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.



**See Also**[domplan\\_total](#)**Examples**

```

#-----
# Uses of library: "predfinitepop"
#-----

#           Read files with functions in R
rm(list=ls())

library("predfinitepop")

set.seed(12345)

#-----
# DATA
#-----
data(dataref)
data(datasample)
data(popcomp)

# Sample (with 10% of the population)

datos <- datasample[[1]]
datos_ant <- dataref
domplan_N <- popcomp[[1]]

#-----
# MODEL PARAMETERS
#-----

# a. Parameter "alpha" of the Dirichlet process
# (this is for J02 planned domains)
alphaDP <- matrix(0.03,2,1)

# b. vector with indicator variables for unplanned domains
colid_D <- t(as.matrix(c(6:8)))

# c. Matrix with parameters for unplanned domains
# (this is for J=2 planned domains)
alpha_D <- matrix(0.3,2,length(colid_D))

#-----
# CODE PARAMETERS
#-----

# A. Interval for calibration of G_0
inter <- 3

# B. Number of groups for predictive validation and model comparison
part <- 20

# C. Number of Monte Carlo replicates of the predictive distributions
nSim <- 10

```

```

#-----
# TESTING "domplan.R"
#-----
date.ini <- date()
domplan_sim <- domplan(datos,datos_ant,domplan_N,alphaDP,inter,part,nSim)
date.fin <- date()

#
# --END --

```

---

domplan_g0	<i>Finite population inference on unplanned domains using subjective <math>G_{j0}</math>.</i>
------------	---

---

### Description

Generates Monte Carlo samples of the predictive distribution of totals of a finite population segmented in planned domains, using a predefined set of baseline distributions  $(G_{j0})_{j=1}^J$ .

### Usage

```
domplan_g0(datos,domplan_N,alphaDP,nSim,g0_licitacion_sal)
```

### Arguments

datos	$(M \times p)$ -dimensional array with positive entries for $S_j$
domplan_N	Matrix array with counts of individuals in each planned domain
alphaDP	$J$ -dimensional array with positive entries for the parameters of the Dirichlet process for $F_j$ (with $J$ being the number of planned domains)
nSim	Number of Monte Carlo simulated replicates of the predictive distribution
g0_licitacion_sal	$(J \times 1)$ object list, each entry is itself an object list associated with each $G_{j0}$ for the $J$ planned domains. The first element for arch $G_{j0}$ should be the name of the chosen distribution (see details below for the admissible alternatives), the second element should be a vector object with the parameters associated with "distribution", and the third element should be its associated expectation

### Details

- datos : Represents the data sample of the target population, unplanned domains labeled.

It should contain the following columns:

"domplan" - Categories for planned domains.

"y\_i" - Actual individual measurements/outcomes (for the moment, they must be positive) for the group of observation.

"n\_i" - Number of individuals in the group (if the unit of observation in the sample are individuals, then " $n_i$ " must be equal to 1)

- domplan\_N : Represents the counts (or number of individuals) in the target population, segmented in the planned domains.

Data must be labeled by the planned domain. It should contain the following columns:

"domplan" - Categories for planned domains.

"N\_j" - Number of individuals in each planned domain.

- g0\_licitacion\_sal: Chose one and only one of the distribution,

i) Gamma

ii) Weibull

iii) Lognormal

iv) Inverse-Gaussian

Parameterizations for distribution:

i) Gamma distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto x^{\alpha-1} \exp\{-x/\beta\},$$

where  $\alpha$  is the shape parameter, and  $\beta$  is the scale parameter.

ii) Weibull distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto (x/\beta)^{\alpha-1} \exp\{-(x/\beta)^\alpha\},$$

where  $\alpha$  is the shape parameter, and  $\beta$  is the scale parameter.

iii) Lognormal distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto \exp\{-(\log x - \alpha)^2/2\beta^2\},$$

where  $\alpha$  is the mean, and  $\beta$  is the standard deviation of the logarithm.

iv) Inverse - Gaussian distribution, with parameters  $\theta = c(\alpha > 0, \beta > 0)$  and density function

$$f(x) \propto x^{-3/2} \exp\{-\alpha(x - \beta)^2/2x\beta^2\},$$

where  $\alpha$  is the shape parameter, and  $\beta$  is the mean parameter.

## Value

The function 'domplan' produces:

total\_domnoplan\_sim

Matrix array of dimension " $J * (3 + 2 D) * nSim$ " with predictions for the planned domains.

Column 1 - Indicator of the planned domains

Column 2 -  $T_j$  (totals of the planned domains)

Column 3 -  $N_j$  (composition of the planned domains)

## Author(s)

Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

## References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

**See Also**

[domplan\\_total](#)

**Examples**

```
#-----
# Uses of library: "predfinitepop"
#-----

#           Read files with functions in R
rm(list=ls())

#library("predfinitepop")

set.seed(12345)

#-----
# DATA
#-----
data(datasample)
data(popcomp)

# Sample (with 10% of the population)

datos <- datasample[[1]]
domplan_N <- popcomp[[1]]

#-----
# MODEL PARAMETERS
#-----

# a. Parameter "alpha" of the Dirichlet process
# (this is for J02 planned domains)
alphaDP <- matrix(0.03,2,1)

# b. vector with indicator variables for unplanned domains
colid_D <- t(as.matrix(c(6:8)))

# c. Matrix with parameters for unplanned domains
# (this is for J=2 planned domains)
alpha_D <- matrix(0.3,2,length(colid_D))

#-----
# CODE PARAMETERS AND INCORPORATION OF THE DISTRIBUTION
#-----
# Incorporate each element ("g0_licitacion_in_J") according to the number of domains planned
#   g0_licitacion_in_1 <- list("distribution",teta1,esp(teta1))
#
# In this case we have two domains planned
g0_licitacion_in_1 <- list("Lognormal",c(6.198,0.840),.5)
g0_licitacion_in_2 <- list("Gamma",c(6.198,0.840),.5)

g0_licitacion_sal <- list(g0_licitacion_in_1,g0_licitacion_in_2)

# B. Number of Monte Carlo replicates of the predictive distributions
nSim <- 10
```

```
#-----
# TESTING "domplan.R"
#-----
date.ini <- date()
domplan_sim <- domplan_g0(datos,domplan_N,alphaDP,nSim,g0_licitacion_sal)
date.fin <- date()

#
# --END --
```

---

domplan_total	<i>Monte Carlo simulation of total for a given planned domain.</i>
---------------	--

---

### Description

Simulates Monte Carlo samples of size ' $nSim$ ' of the final distribution for the total for the given planned domain  $\mathcal{P}_j$ .

### Usage

```
domplan_total(datos_j,rho,ystar,phi,g0_licitacion_sal,N_Stil_j,nSim)
```

### Arguments

datos_j	Data matrix with features and number of individuals in the sample ' $\mathcal{S}_j$ '
rho	$(U \times 2)$ -dimensional vector with weights associated with the sample ties ' $ystar$ '
ystar	Sample ties $y_i^*$ , in the sample $\mathcal{S}_j$
phi	Probability weight asociated with ' $G_{j0}$ ' (the continuos part of ' $\hat{G}_j$ ')
g0_licitacion_sal	Object list with the continuous component in ' $\hat{G}_j$ '
N_Stil_j	Composition of individuals out of ' $\mathcal{S}_j$ '
nSim	Number of Monte carlosimulated replicates of the predictive distribution

### Value

The function 'domplan\_total' produces and object list with two entries:

T_S_j	Composition of ' $T_j$ ' of the given planned domains in the sample ' $\mathcal{S}_j$ '
domplan_total_sim	$(1 * nSim)$ matrix with samples from the predictive distribution of the total ' $T_{Stil_j}$ '

### Author(s)

Juan Carlos Martinez-Ovando

## References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

## See Also

[domnoplan](#)

---

g0_licitacion	<i>Prior elicitation of <math>G_{j0}</math> using reference data.</i>
---------------	---

---

## Description

This function computes the prior elicitation of  $G_{j0}$  using reference data for the planned domain  $j$ .  $G_{j0}$  is used by the SSM as the baseline function. The distribution is elicited using a predictive cross-validation procedure for model comparison and selection among the following alternatives: Gamma, Weibull, Lognormal and Inverse-Gaussian. The function also computes the expectation of the chose distribution.

## Usage

```
g0_licitacion(datos_ant, inter, part)
```

## Arguments

datos_ant	Reference data for calibration of $G_{j0}$
inter	Tunning parameter for model comparison and selection
part	Number of groups for predictive validation. The larger the number of groups the fewer observations are to be removed for calibration

## Details

datos\_ant: Data collected in a previous time or from another source for each planned domain. It must contain two columns, labeled by 'domplan' and 'y\_i' 'domplan' - Labels of planned domains 'y\_i' - individual outcome (for the sake of implementation, this must be positive)

## Value

The function 'g0\_licitacion' produces an object list with three elements:

distribution	List arrangement: String object for 'distribution' i) Gamma ii) Weibull iii) Log-normal iv) Inverse Gaussian
theta	Vector of parameters associated with 'distribution'
mu	Expected value of 'distribution'

**Author(s)**

Adriana Roldan-Rodriguez

**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

---

g0_simulacion	<i>Computes Monte Carlo samples from a given <math>G_{j0}</math>.</i>
---------------	---

---

**Description**

This function computes Monte Carlo samples from a given continuous  $G_{j0}$ .

**Usage**

```
g0_simulacion(g0_licitacion_sal,nSim)
```

**Arguments**

`g0_licitacion_sal`  
Object list with three elements (produced with 'g0\_licitacion'):

- a) String object for 'distribution'
  - i) Gamma
  - ii) Weibull
  - iii) Log-normal
  - iv) Inverse-Gaussian
- b) theta - Vector of parameters associated with 'distribution'
- c) mu - Expected value of 'distribution'

`nSim`                      Number of Monte Carlo simulations

**Value**

This function 'g0\_simulacion' produces:

`g0_simulacion_sal`  
(nSim X 1) dimensional array with simulated data

**Author(s)**

Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando

## References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

## See Also

[ghat\\_simulacion](#)

---

ghat_simulacion	<i>Simulates Monte Carlo samples from <math>\hat{G}_j</math>.</i>
-----------------	---

---

## Description

This function simulates Monte Carlo samples from  $\hat{G}_j$ .

## Usage

```
ghat_simulacion(rho, ystar, phi, g0_licitacion_sal, N_Stil_j)
```

## Arguments

rho	( $U \times 2$ )-dimensional vector with weights associated with 'ystar'
ystar	Sample ties $y_i^*$ in $\mathcal{S}_j$
phi	Probability weight asociated with $G_{j0}$ (the continuos part of $\hat{G}_j$ )
g0_licitacion_sal	Object list with three elements (produced with 'g0_licitacion'): <ul style="list-style-type: none"> <li>a) String object for 'distribution'               <ul style="list-style-type: none"> <li>i) Gamma</li> <li>ii) Weibull</li> <li>iii) Log-normal</li> <li>iv) Inverse-Gaussian</li> </ul> </li> <li>b) theta - Parameters associated with 'distribution'</li> <li>c) mu - Expected value of 'distribution'</li> </ul>
N_Stil_j	Number of samples to simulate (number of individuals in $\tilde{\mathcal{S}}_j$ ) out of the sample

## Value

'ghat\_simulacion' produces:

```
ghat_simulacion_sal
(N_Stil_j X 1) dimensional vector with simulated data
```



**Author(s)**

Juan Carlos Martinez-Ovando

**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

**See Also**

[g0\\_simulacion](#), [gstar\\_simulacion](#).

---

<code>gstar_simulacion</code>	<i>Generates Monte Carlo samples from the discrete component of <math>\hat{G}_j</math></i>
-------------------------------	--

---

**Description**

This function generates Monte Carlo samples from the discrete component of  $\hat{G}_j$  for the Dirichlet process prior in the planned domain  $j$ .

**Usage**

```
gstar_simulacion(rho, ystar, nSim)
```

**Arguments**

<code>rho</code>	$(U \times 2)$ -dimensional vector with weights associated with 'ystar'
<code>ystar</code>	Sample ties $y_i^*$ in $\mathcal{S}_j$
<code>nSim</code>	Number of Monte Carlo simulations

**Value**

The function 'gstar\_simulacion' produces:

```
gstar_simulacion_sal
```

(nSim X 1) dimensional vector with simulated data

**Author(s)**

Juan Carlos Martinez-Ovando

## References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

## See Also

[ghat\\_simulacion](#)

---

pesosDP	<i>Computes the weights associated with the predictive distribution <math>\hat{G}_j</math> for a given planned domain <math>j</math> under the Dirichlet processes prior.</i>
---------	---

---

## Description

This function computes the weights associated with the predictive distribution  $\hat{G}_j$  for a given planned domain  $j$ , using the sample ties  $(y_j^*)$ , under the Dirichlet processes prior.

## Usage

```
pesosDP(unicos_sal, alphaDP)
```

## Arguments

unicos_sal	( $U \times 2$ ) matrix with sample ties and asociated frequencies (produced with 'unicos')
alphaDP	Scale parameter of the Dirichlet process

## Details

The matrix with sample ties 'unicos\_sal' must include a column named 'm\_k' for the frequencies of the sample ties.

## Value

The function 'pesosDP' produces an object list with two elements:

rho	( $U \times 2$ ) dimensional vector with weights associated with 'ystar' (the sample ties)
phi	Probability weight asociated with $G_{j0}$ (the continuos part of $\hat{G}_j$ )

## Author(s)

Sergio I. Olivares-Guzman, Juan Carlos Martinez-Ovando

## References

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

---

pginvgauss

*Generalized inverse Gaussian distribution*

---

## Description

Probability of the generalized inverse Gaussian distribution.

## Usage

pginvgauss(q, m, s, f)

## Arguments

q	Vector of quantiles.
m	Vector of means.
s	Vector of dispersion parameters.
f	Vector of family parameters.

## Author(s)

Jim Lindsey <jlindsey@luc.ac.be>

## References

Extracted from package "rmutil" Version 1.0 "Utilities for Nonlinear Regression and Repeated Measurements Models"

---

pinvgauss

*Inverse Gaussian distribution*

---

## Description

Probability of the inverse Gaussian distribution.

## Usage

pinvgauss(q, m, s)

**Arguments**

q	Vector of quantiles.
m	Vector of means.
s	Vector of dispersion parameters.

**Author(s)**

Jim Lindsey <jlindsey@luc.ac.be>

**References**

Extracted from package "rmutil" Version 1.0 "Utilities for Nonlinear Regression and Repeated Measurements Models"

---

popcomp	<i>Composition of the simulated population segmented in planned and unplanned domains.</i>
---------	--

---

**Description**

This object corresponds to a list with the composition of the simulated population across planned and unplanned domains. See Martinez-Ovando et al. (2014) for further details.

**Usage**

```
data(popcomp)
```

**Format**

The format is: List of 3 \$ : num [1:2] 800 1200 \$ : num [1:3] 980 800 220

**Source**

Simulated data. See Martinez-Ovando et al. (2014) for further details.

**References**

- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

**Examples**

```
data(popcomp)
```

---

poptot	<i>Totals of the simulated population segmented in planned and unplanned domains.</i>
--------	---

---

### Description

This object corresponds to a list with the totals of the simulated population across planned and unplanned domains. See Martinez-Ovando et al. (2014) for further details.

### Usage

```
data(poptot)
```

### Format

The format is:

[[1 ]] Composition of the population across planned domains (working format).

[[2 ]] Composition of the population across planned domains.

[[3 ]] Composition of the population across unplanned domains .

### Source

Simulated data. See Martinez-Ovando et al. (2014) for further details.

### References

- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

### Examples

```
data(poptot)
```

---

predfinitepop	<i>Predictive Inference on Totals and Averages of Finite Populations Segmented in Planned and Unplanned Domains.</i>
---------------	--

---

### Description

This package computes Bayesian predictive inference on totals and averages of finite populations segmented in planned and unplanned domains. Inference is based on Bayesian nonparametric methods using species-sampling models (specifically the Dirichlet process).

**Details**

Package: predfinitepop  
 Type: Statistical  
 Version: v-1  
 Date: 2014-03-31

**Author(s)**

Juan Carlos Martinez-Ovando, Banco de Mexico, <juan.martinez@banxico.org.mx>  
 Sergio I. Olivares-Guzman, Banco de Mexico, <solivares@banxico.org.mx>  
 Aadriana Roldan-Rodriguez, Banco de Mexico, <aroldan@banxico.org.mx>

**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

---

randDirichlet	<i>Simulates samples from the Dirichlet distribution with (px1) vector parameter <math>\alpha</math>.</i>
---------------	---

---

**Description**

This function simulates samples from the Dirichlet distribution with (px1) vector parameter  $\alpha$ .

**Usage**

```
randDirichlet(alpha,n)
```

**Arguments**

alpha	$p$ -dimensional vector with positive entries
n	Number of simulated replicates

**Value**

This function produces:

randDir	$(n \times p)$ matrix with 'n' simulated replicates
---------	---

**Author(s)**

Juan Carlos Martinez-Ovando

**References**

- "Non-Uniform Random Variate Generation", Luc Devroy (1986), Berlin: Springer-Verlag.

---

 rginvgauss

*Generalized inverse Gaussian distribution*


---

**Description**

Simulation from the generalized inverse Gaussian distribution.

**Usage**

```
rginvgauss(n = 1, m, s, f)
```

**Arguments**

n	Number of values to generate.
m	Vector of means.
s	Vector of dispersion parameters.
f	Vector of family parameters.

**Author(s)**

Jim Lindsey <jlindsey@luc.ac.be>

**References**

Extracted from package "rmutil" Version 1.0 "Utilities for Nonlinear Regression and Repeated Measurements Models"

---

 rinvgauss

*Inverse Gaussian distribution*


---

**Description**

Simulation from the inverse Gaussian distribution.

**Usage**

```
rinvgauss(n = 1, m, s)
```

**Arguments**

n	Number of values to generate.
m	Vector of means.
s	Vector of dispersion parameters.

**Author(s)**

Jim Lindsey <jlindsey@luc.ac.be>

**References**

Extracted from package "rmutil" Version 1.0 "Utilities for Nonlinear Regression and Repeated Measurements Models"

---

unicstar	<i>Identifies the sample ties in <math>\mathcal{S}_j</math> and computes their associated sample frequencies.</i>
----------	---

---

**Description**

This function identifies the sample ties ( $y_k^*$ ) in  $\mathcal{S}_j$  and computes their associated sample frequencies.

**Usage**

```
unicstar(datos_j)
```

**Arguments**

datos_j	Data matrix with individual measurements in " $\mathcal{S}_j$ "
---------	---

**Details**

The column with individual measurements in "datos\_j" should be named "y\_i"

**Value**

The function 'unicstar' produces an object list with two elements:

U	Number of sample ties in " $\mathcal{S}_j$ "
unicstar	Data matrix with sample ties ' $y_i$ ' and asociated frequencies

**Author(s)**

Sergio I. Olivares-Guzman, Adriana Roldan-Rodriguez, Juan Carlos Martinez-Ovando



**References**

- "A Bayesian nonparametric framework to inference on totals of finite populations," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2013. Contributions of Young Researchers to Bayesian Statistics (eds. E. Lazarone & F. Ieva), Chapter 15 Springer.
- "Predictive inference on finite populations segmented in planned and unplanned domains," Martinez-Ovando, J. C., Olivares-Guzman, S. I., Roldan-Rodriguez, A., 2014. Submitted. Also available as Banco de Mexico Working Paper 2014-04.

# Index

- \*Topic **Calibration**
  - g0\_licitacion, [22](#)
- \*Topic **Compositional analysis**
  - domnoplan, [6](#)
  - domnoplan\_g0, [11](#)
- \*Topic **Convolutions**
  - domnoplan, [6](#)
  - domplan, [15](#)
  - domplan\_g0, [18](#)
- \*Topic **Domains planned**
  - domnoplan, [6](#)
  - domnoplan\_g0, [11](#)
  - domplan, [15](#)
- \*Topic **Domains unplanned**
  - domnoplan, [6](#)
  - domnoplan\_g0, [11](#)
- \*Topic **Model comparison and selection**
  - g0\_licitacion, [22](#)
- \*Topic **Monte Carlo simulation**
  - g0\_simulacion, [23](#)
  - ghat\_simulacion, [24](#)
  - gstar\_simulacion, [25](#)
- \*Topic **Planned domains**
  - domplan\_g0, [18](#)
- \*Topic **Population Total**
  - domplan\_g0, [18](#)
- \*Topic **Prediction**
  - g0\_simulacion, [23](#)
  - ghat\_simulacion, [24](#)
  - gstar\_simulacion, [25](#)
- \*Topic **Prior elicitation**
  - g0\_licitacion, [22](#)
- \*Topic **Total Population**
  - domnoplan, [6](#)
  - domnoplan\_g0, [11](#)
  - domplan, [15](#)
- \*Topic **convolutions**
  - domnoplan\_g0, [11](#)
- \*Topic **datasets**
  - dataref, [3](#)
  - datasample, [4](#)
  - popcomp, [28](#)
  - poptot, [29](#)
- conteo, [2](#)
- dataref, [3](#)
- datasample, [4](#)
- dginvgauss, [5](#)
- dinvgauss, [6](#)
- domnoplan, [6](#), [10](#), [22](#)
- domnoplan., [15](#)
- domnoplan\_composicion, [9](#)
- domnoplan\_g0, [11](#)
- domnoplan\_totalcomp, [14](#)
- domnoplan\_totalcomp., [8](#), [13](#)
- domplan, [15](#)
- domplan\_g0, [18](#)
- domplan\_total, [17](#), [20](#), [21](#)
- g0\_licitacion, [22](#)
- g0\_simulacion, [23](#)
- ghat\_simulacion, [24](#), [24](#), [26](#)
- gstar\_simulacion, [25](#)
- gstar\_simulacion., [25](#)
- pesosDP, [26](#)
- pginvgauss, [27](#)
- pinvgauss, [27](#)
- popcomp, [28](#)
- poptot, [29](#)
- predfinitepop, [29](#)
- randDirichlet, [30](#)
- rginvgauss, [31](#)
- rinvgauss, [31](#)
- unicstar, [32](#)