

Package ‘rmp’

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Type Package

Title Rounded Mixture Package

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Description

Performs univariate probability mass function estimation via Bayesian nonparametric mixtures of rounded kernels as in Canale and Dunson (2011) <[doi:10.1198/jasa.2011.tm10552](https://doi.org/10.1198/jasa.2011.tm10552)>.

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R topics documented:

dp.post.est	1
ethylene	3
npmp	4
rmg	6

Index	10
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dp.post.est	<i>Posterior probability mass function estimation with DP prior</i>
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Description

Performs Bayesian probability mass function estimation under DP prior with Poisson base measure.

Usage

```
dp.post.est(x, y, alpha, lambda)
```

Arguments

x	Values on which to compute the pmf.
y	Vector of observed data.
alpha	DP precision parameter
lambda	Mean parameter for the Poisson base measure.

Details

Performs probability mass function estimation under the following model

$$y_i | P \sim P, i = 1, \dots, n$$

$$P \sim DP(\alpha, P_0),$$

where P_0 is Poisson with mean lambda.

Value

A vector of size length(x) containing the probability masses

Author(s)

Antonio Canale

References

Carota, C., and Parmigiani, G. (2002), "Semiparametric Regression for Count Data," *Biometrika*, **89**, 265–281.

Examples

```
data(ethylene)
y <- tapply(ethylene$impl, FUN=mean, INDEX=ethylene$id)
z <- tapply(ethylene$dose, FUN=mean, INDEX=ethylene$id)

# Estimate the pmf of the number of implants in the control group
y0 <- y[z==0]
pmf.control = dp.post.est(0:30, y0, alpha = 1)
```

ethylene

Developmental toxicity study of ethylene glycol in mice

Description

Data from the developmental toxicity study of ethylene glycol in mice conducted by the National Toxicology Program. Pregnant mice were assigned to dose groups of 0, 750, 1,500 or 3,000 mg/kg per day, with the number of implants measured for each mouse at the end of the experiment. Group sizes are 25, 24, 23 and 23, respectively.

Usage

```
data(ethylene)
```

Format

A data frame with 1192 observations on the following 7 variables.

`id` identifier for pregnant mouse
`dose` dose of ethylene glycol
`weight` weight of the fetus
`sex` sex of the fetus
`impl` number of implants in the pregnant mouse
`litsz` size of the relative litter
`mal` presence of malformation in the fetus

References

Price, C. J., Kimmel, C. A., Tyl, R. W., and Marr, M. C. (1985) "The developmental toxicity of ethylene glycol in rats and mice" *Toxicological and Applied Pharmacology* **81**, 113-127.

Examples

```
data(ethylene)
implants <- tapply(ethylene$impl, FUN=mean, INDEX=ethylene$id)

summary(implants)
m <- mean(implants)
v <- var(implants)
hist(implants, main=paste("Histogram of the number of implants (Mean = ",
round(m,2), ", Var = ", round(v,2),")"))
```

npmp

*DP mixtures of Poissons***Description**

Performs probability mass function estimation under nonparametric mixture of Poisson kernels.

Usage

```
npmp(y, k, nrep, nb, alpha=1, theta=alpha, sigma=0,
     mixing_hyperprior= FALSE, basemeasure_hyperprior = FALSE, mixing_type="DP",
     algo="slice", prior="gamma", a, b, a_a=1, b_a=1, lb=NULL, ub=NULL,
     print = 1, ndisplay = nrep/4, plot.it = FALSE, pdfwrite = FALSE, ... )
```

```
dmpoiss(y, k, nrep, nb, alpha = 1, a, b, lb = NULL, ub = NULL, print = 1,
        ndisplay = nrep/4, plot.it = FALSE, pdfwrite = FALSE, ...)
```

Arguments

y	Vector of count data
k	Truncation level for the number of cluster in the mixture. Default is length(ydis).
nrep	Number of MCMC iterations
nb	Number of burn-in iteration in the MCMC to discard
alpha	Value of the precision parameter of the Dirichlet process prior
theta	Value of the strength parameter of the Two-parameters-Poisson-Dirichlet process prior
sigma	Value of the discount parameter of the Two-parameters-Poisson-Dirichlet process prior
mixing_hyperprior	Logical. If TRUE alpha is random with gamma hyperprior
basemeasure_hyperprior	Logical. If TRUE also the parameters of the base measure are random, see details below.
mixing_type	Type of mixing distribution. Default is "DP" for Dirichlet process but also "2PD" for Two-parameters-Poisson-Dirichlet process is allowed.
algo	Type of algorithm. Current choices are: slice sampler (algo="slice") or polyurn-type sampler (algo="polya-urn").
prior	String for the base measure prior. Default is "gamma" for $\lambda \sim \text{Gamma}(a,b)$. The other choice is "normal" for $\exp(\lambda) \sim N(a,b)$
a	Shape (mean) hyperparameter for the gamma (normal) base measure
b	Scale (sd) hyperparameter for the gamma (normal) prior
a_a	Shape hyperparameter for alpha

<code>b_a</code>	Scale hyperparameter for alpha
<code>lb</code>	Scalar integer. Lower bound for the argument of the pmf. Default is $\max(0, \min(ydis)-10)$.
<code>ub</code>	Scalar integer. Upper bound for the argument of the pmf. Default is $\max(ydis)+10$.
<code>print</code>	Vector of integers (from 1 to 5) indicating whether to print each step of the Gibbs sampler. Specifically, 1 for current iteration, 2 for the DP cluster allocation, 3 for the posterior parameters of the mixture components, 4 for the precision of the DP, 5 for the posterior pmf.
<code>ndisplay</code>	Scalar integer. It gives the number of iterations to be displayed on screen (the function reports on the screen when every <code>ndisplay</code> iterations have been carried out)
<code>plot.it</code>	Logical, default FALSE. If TRUE a plot with empirical and estimated posterior probability mass functions is plotted.
<code>pdfwrite</code>	Logical, default FALSE. If TRUE a pdf file is written in the current working directory. Traceplots and other posterior quantities are drawn.
<code>...</code>	Additional arguments (for future implemetantions).

Details

The function `npmp` performs probability mass function estimation under nonparametric mixture of Poisson kernels, i.e.

$$\begin{aligned}
 y_i \mid \lambda_i &\sim \text{Poi}(\lambda_i), i = 1, \dots, n \\
 \lambda_i \mid G &\sim G \\
 G &\sim \Pi(P_0),
 \end{aligned}$$

where Π is a nonparametric prior (Dirichlet process or Two-parameters-Poisson-Dirichlet process) with base measure P_0 . The function `dppoiss` is a wrapper to `npmp` with `mixing_type="DP"` for back portability with version 1.0 of the package. The main part of the code is written in C language to gain computational speed. Plots and posterior summaries are in plain R code. From version 2.0 on, the blocked Gibbs sampler has been removed in place of slice sampler (Kalli et al., 2011) and poly-urn sampler. Two different base measures P_0 are implemented: `prior="gamma"` and `prior="normal"` for

$$\lambda_h \sim \text{Gamma}(a, b), \log(\lambda_h) \sim N(a, b), h = 1, \dots$$

respectively.

Value

<code>name</code>	Name of the model
<code>mixing_type</code>	Name of the mixing prior
<code>mcmc</code>	Quantities about MCMC sampling
<code>mcmc.chains</code>	MCMC chains of the parameters
<code>pmf</code>	A list containing several quantities related to the probability mass function (empirical pmf, posterior mean pmf and pointwise 95% credible intervals) computed for the values from <code>lb</code> to <code>ub</code>

parameters	A list containing the posterior mean of the cluster specific parameters (be careful of label-switching problems)
clustering	A list containing posterior quantities related to the clustering structure of the data

Author(s)

R code and porting by A. Canale, C code by A. Canale with minor contributions by N. Lunardon.

References

- Canale, A. and Dunson, D. B. (2011), "Bayesian Kernel Mixtures for Counts", *Journal of American Statistical Association*, **106**, 1528-1539.
- Kalli, M., Griffin, J., and Walker, S. (2011), "Slice sampling mixture models," *Statistics and Computing*, **21**, 93-105.

See Also

rmg

Examples

```
data(ethylene)
y <- tapply(ethylene$impl, FUN=mean, INDEX=ethylene$id)
z <- tapply(ethylene$dose, FUN=mean, INDEX=ethylene$id)

# Estimate the pmf of the number of implants in the control group
y0 <- y[z==0]
pmf.control = dpmpois(y0, k=20, nrep=11000, nb=1000, alpha=1, a=1, b=1,
lb=5, ub=24, plot.it=TRUE)
```

rmg

Nonparametric mixture of rounded Gaussians

Description

Performs Bayesian probability mass function estimation under nonparametric mixture of rounded of Gaussian kernels.

Usage

```
rmg(ydis, k=length(ydis), nrep, nb, alpha=1, theta=alpha, sigma=0,
mixing_hyperprior= FALSE, basemeasure_hyperprior = FALSE, mixing_type="DP", algo="slice",
mu0=mean(ydis), kap=var(ydis),
atau=1, btau=2, a_a=1, b_a=1, lb=NULL, ub=NULL, print = 1, ndisplay = nrep/4,
plot.it = FALSE, pdfwrite = FALSE, ...)
```

```
dpmrg(ydis, k, nrep, nb, alpha, alpha_r = FALSE, mu0 = mean(ydis), kap = var(ydis),
atau, btau, a_a = 1, b_a = 1, lb = NULL, ub = NULL,
print = 1, ndisplay = nrep/4,
plot.it = FALSE, pdfwrite = FALSE, ...)
```

Arguments

ydis	Vector of count data
k	Truncation level for the number of cluster in the mixture. Default is <code>length(ydis)</code> .
nrep	Number of MCMC iterations
nb	Number of burn-in iteration in the MCMC to discard
alpha	Value of the precision parameter of the Dirichlet process prior
theta	Value of the strength parameter of the Two-parameters-Poisson-Dirichlet process prior
sigma	Value of the discount parameter of the Two-parameters-Poisson-Dirichlet process prior
mixing_hyperprior	Logical. If TRUE alpha is random with gamma hyperprior
alpha_r	Logical. If TRUE alpha is random with gamma hyperprior
basemeasure_hyperprior	Logical. If TRUE also the parameters of the base measure are random, see details below.
mixing_type	Type of mixing distribution. Default is "DP" for Dirichlet process but also "2PD" for Two-parameters-Poisson-Dirichlet process is allowed.
algo	Type of algorithm. Current choices are: slice sampler (<code>algo="slice"</code>) or poly-urn-type sampler (<code>algo="polya-urn"</code>).
mu0	Location hyperparameter for the latent rounded Gaussian base measure
kap	Precision hyperparameter for the latent rounded Gaussian base measure
atau	Shape hyperparameter for the Gamma distribution
btau	Scale hyperparameter for the Gamma distribution
a_a	Shape hyperparameter for the Gamma distribution for alpha
b_a	Scale hyperparameter for the Gamma distribution for alpha
lb	Scalar integer. Lower bound for the argument of the pmf. Default is <code>max(0, min(ydis)-10)</code> .
ub	Scalar integer. Upper bound for the argument of the pmf. Default is <code>max(ydis)+10</code> .
print	Vector of integers (from 1 to 6) indicating whether to print each step of the Gibbs sampler. Specifically, 1 for current iteration, 2 for the data augmentation step simulating the latent continuous variables, 3 for the DP cluster allocation, 4 for the posterior parameters of the mixture components, 5 for the precision of the DP, 6 for the posterior pmf.

<code>ndisplay</code>	Scalar integer. It gives the number of iterations to be displayed on screen (the function reports on the screen when every <code>ndisplay</code> iterations have been carried out)
<code>plot.it</code>	Logical, default FALSE. If TRUE a plot with empirical and estimated posterior probability mass functions is plotted.
<code>pdfwrite</code>	Logical, default FALSE. If TRUE a pdf file is written in the current working directory. Traceplots and other posterior quantities are drawn.
<code>...</code>	Additional arguments (for future implemetantions).

Details

The `rmg` function performs Bayesian probability mass function estimation under the mixture model of Canale and Dunson (2011) with Dirichlet process or Two-parameters-Poisson-Dirichlet process as prior for the mixing measure. The model is

$$y_i \mid \mu_i, \tau_i \sim \text{RG}(\mu_i, \tau_i), i = 1, \dots, n$$

$$(\mu_i, \tau_i) \mid G \sim G$$

$$G \sim \Pi(P_0),$$

where Π is the Dirichlet process or the Two-parameters-Poisson-Dirichlet process with base measure P_0 and $\text{RG}(\mu, \tau)$ is a rounded Gaussian kernel with location μ and precision τ and tresholds $-\infty, 1, 2, \dots$. The function `dpmrg` is a wrapper to `rmg` with `mixing_type="DP"` for back portability with version 1.0 of the package. The main part of the code is written in C language to gain computational speed. Plots and posterior summaries are in plain R code. From version 2.0 on, the blocked gibbs sampler has been removed in place of slice samper (Kalli et al., 2011) and polya-urn sampler.

Value

<code>name</code>	Name of the model
<code>mixing_type</code>	Name of the mixing prior
<code>mcmc</code>	Quantities about MCMC sampling
<code>mcmc.chains</code>	MCMC chains of the parameters
<code>pmf</code>	A list containing several quantities related to the probability mass function (empirical pmf, posterior mean pmf and pointwise 95% credible intervals) computed for the values from <code>lb</code> to <code>ub</code>
<code>parameters</code>	A list containing the posterior mean of the cluster specific parameters (be careful of label-switching problems)
<code>clustering</code>	A list containing posterior quantities related to the clustering structure of the data

Author(s)

R code and porting by A. Canale, C code by A. Canale with minor contributions by N. Lunardon.

References

Canale, A. and Dunson, D. B. (2011), "Bayesian Kernel Mixtures for Counts", *Journal of American Statistical Association*, **106**, 1528-1539.

Kalli, M., Griffin, J., and Walker, S. (2011), "Slice sampling mixture models," *Statistics and Computing*, **21**, 93-105.

See Also

dpmpois

Examples

```
data(ethylene)
y <- tapply(ethylene$impl, FUN=mean, INDEX=ethylene$id)
z <- tapply(ethylene$dose, FUN=mean, INDEX=ethylene$id)

# Estimate the pmf of the number of implants in the control group
y0 <- y[z==0]
pmf.control = rmg(y0, k=20, nrep=11000, nb=1000, alpha=1, atau=1, btau=1,
lb=5, ub=24, plot.it= TRUE)
```

Index

* datasets

ethylene, 3

dp.post.est, 1

dpmposs (nmp), 4

dpmrg (rmg), 6

ethylene, 3

nmp, 4

rmg, 6