

# Package: mig (via r-universe)

September 13, 2024

**Type** Package

**Title** Multivariate Inverse Gaussian Distribution

**Version** 1.0

**Description** Provides utilities for estimation for the multivariate inverse Gaussian distribution of Minami (2003) [doi:10.1081/STA-120025379](https://doi.org/10.1081/STA-120025379), including random vector generation and explicit estimators of the location vector and scale matrix. The package implements kernel density estimators discussed in Belzile, Desgagnes, Genest and Ouimet (2024) [doi:10.48550/arXiv.2209.04757](https://doi.org/10.48550/arXiv.2209.04757) for smoothing multivariate data on half-spaces.

**BugReports** <https://github.com/lbelzile/mig/issues>

**Imports** statmod, TruncatedNormal (>= 2.3), Rcpp (>= 1.0.12)

**Depends** R (>= 2.10)

**Suggests** numDeriv, tinytest, knitr, rmarkdown, minqa

**LinkingTo** Rcpp, RcppArmadillo

**Encoding** UTF-8

**LazyData** true

**License** MIT + file LICENSE

**VignetteBuilder** knitr

**RoxygenNote** 7.3.2

**Repository** <https://lbelzile.r-universe.dev>

**RemoteUrl** <https://github.com/lbelzile/mig>

**RemoteRef** HEAD

**RemoteSha** a42fbf87f381603f12fdf3d189c3cc23900f3d3c

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dmig	<i>Multivariate inverse Gaussian distribution</i>
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### Description

The density of the MIG model is

$$f(\mathbf{x} + \mathbf{a}) = (2\pi)^{-d/2} \boldsymbol{\beta}^\top \boldsymbol{\xi} |\boldsymbol{\Omega}|^{-1/2} (\boldsymbol{\beta}^\top \mathbf{x})^{-(1+d/2)} \exp \left\{ -\frac{(\mathbf{x} - \boldsymbol{\xi})^\top \boldsymbol{\Omega}^{-1} (\mathbf{x} - \boldsymbol{\xi})}{2\boldsymbol{\beta}^\top \mathbf{x}} \right\}$$

for points in the  $d$ -dimensional half-space  $\{\mathbf{x} \in \mathbb{R}^d : \boldsymbol{\beta}^\top (\mathbf{x} - \mathbf{a}) \geq 0\}$

### Usage

```
dmig(x, xi, Omega, beta, shift, log = FALSE)
```

```
rmig(n, xi, Omega, beta, shift, method = c("invsim", "bm"), timeinc = 0.001)
```

```
pmig(q, xi, Omega, beta, log = FALSE, method = c("sov", "mc"), B = 10000L)
```

### Arguments

x	n by d matrix of quantiles
xi	d vector of location parameters $\boldsymbol{\xi}$ , giving the expected value
Omega	d by d positive definite scale matrix $\boldsymbol{\Omega}$
beta	d vector $\boldsymbol{\beta}$ defining the half-space through $\boldsymbol{\beta}^\top \boldsymbol{\xi} > 0$
shift	d translation for the half-space $\mathbf{a}$
log	logical; if TRUE, returns log probabilities
n	number of observations
method	string; one of inverse system (invsim, default), Brownian motion (bm)
timeinc	time increment for multivariate simulation algorithm based on the hitting time of Brownian motion, default to 1e-3.
q	n by d matrix of quantiles
B	number of Monte Carlo replications for the SOV estimator

### Details

Observations are generated using the representation as the first hitting time of a hyperplane of a correlated Brownian motion.

**Value**

for dmig, the (log)-density

for rmig, an n vector if d=1 (univariate) or an n by d matrix if d > 1

an n vector of (log) probabilities

**Author(s)**

Frederic Ouimet (bm), Leo Belzile (invsim)

Leo Belzile

**Examples**

```
# Density evaluation
x <- rbind(c(1, 2), c(2,3), c(0,-1))
beta <- c(1, 0)
xi <- c(1, 1)
Omega <- matrix(c(2, -1, -1, 2), nrow = 2, ncol = 2)
dmig(x, xi = xi, Omega = Omega, beta = beta)
# Random number generation
d <- 5L
beta <- runif(d)
xi <- rexp(d)
Omega <- matrix(0.5, d, d) + diag(d)
samp <- rmig(n = 1000, beta = beta, xi = xi, Omega = Omega)
mle <- fit_mig(samp, beta = beta, method = "mle")
set.seed(1234)
d <- 2L
beta <- runif(d)
Omega <- rWishart(n = 1, df = 2*d, Sigma = matrix(0.5, d, d) + diag(d))[,,1]
xi <- rexp(d)
q <- mig::rmig(n = 10, beta = beta, Omega = Omega, xi = xi)
pmig(q, xi = xi, beta = beta, Omega = Omega)
```

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fit\_mig

*Fit multivariate inverse Gaussian distribution*


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**Description**

Fit multivariate inverse Gaussian distribution

**Usage**

```
fit_mig(x, beta, method = c("mle", "mom"), shift)
```

**Arguments**

x	n by d matrix of quantiles
beta	d vector $\beta$ defining the half-space through $\beta^\top \xi > 0$
method	string, one of mle for maximum likelihood estimation, or mom for method of moments.
shift	d translation for the half-space $a$

**Value**

a list with components:

- xi: estimate of the expectation or location vector
- Omega: estimate of the scale matrix

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geomagnetic

*Magnetic storms*

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**Description**

Absolute magnitude of 373 geomagnetic storms lasting more than 48h with absolute magnitude (dst) larger than 100 in 1957-2014.

**Format**

a vector of size 373

**Note**

For a detailed article presenting the derivation of the Dst index, see <http://wdc.kugi.kyoto-u.ac.jp/dstdir/dst2/onDst>

**Source**

Aki Vehtari

**References**

World Data Center for Geomagnetism, Kyoto, M. Nose, T. Iyemori, M. Sugiura, T. Kamei (2015), *Geomagnetic Dst index*, doi:10.17593/14515-74000.

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mig_kdens	<i>Multivariate inverse Gaussian kernel density estimator</i>
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**Description**

Given a matrix of new observations, compute the density of the multivariate inverse Gaussian mixture defined by assigning equal weight to each component where  $\xi$  is the location parameter.

**Usage**

```
mig_kdens(x, newdata, Omega, beta, log = FALSE)
```

**Arguments**

x	n by d matrix of quantiles
newdata	matrix of new observations at which to evaluate the kernel density
Omega	d by d positive definite scale matrix $\Omega$
beta	d vector $\beta$ defining the half-space through $\beta^\top \xi > 0$
log	logical; if TRUE, returns log probabilities

**Value**

value of the (log)-density at newdata

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mig_kdens_bandwidth	<i>Optimal scale matrix for MIG kernel density estimation</i>
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**Description**

Given an n sample from a multivariate inverse Gaussian distribution on the half-space defined by  $\{\mathbf{x} \in \mathbb{R}^d : \beta^\top \mathbf{x} > 0\}$ , the function computes the bandwidth (type="isotropic") or scale matrix that minimizes the asymptotic mean integrated squared error away from the boundary. The latter depends on the true unknown density, which is replaced using as plug-in a MIG distribution evaluated at the maximum likelihood estimator. The integral or the integrated squared error are obtained by Monte Carlo integration with N simulations

**Usage**

```
mig_kdens_bandwidth(
  x,
  beta,
  shift,
  method = c("amise", "lcv", "lscv", "rlcv"),
  type = c("isotropic", "full"),
```

```

approx = c("mig", "tnorm"),
transformation = c("none", "scaling", "spherical"),
N = 10000L,
buffer = 0.25,
pointwise = NULL,
maxiter = 2000L,
...
)

```

### Arguments

x	an n by d matrix of observations
beta	d vector defining the half-space
shift	location vector for translating the half-space. If missing, defaults to zero
method	estimation criterion, either amise for the expression that minimizes the asymptotic integrated squared error, lcv for likelihood (leave-one-out) cross-validation, lscv for least-square cross-validation or r1cv for robust cross validation of Wu (2019)
type	string indicating whether to compute an isotropic model or estimate the optimal scale matrix via optimization
approx	string; distribution to approximate the true density function $f(x)$ ; either mig for multivariate inverse Gaussian, or tnorm for truncated Gaussian.
transformation	string for optional scaling of the data before computing the bandwidth. Either standardization to unit variance scaling, spherical transformation to unit variance and zero correlation (spherical), or none (default).
N	integer number of simulations to evaluate the integrals of the MISE by Monte Carlo
buffer	double indicating the buffer from the halfspace
pointwise	if NULL, evaluates the mean integrated squared error, otherwise a d vector to evaluate the bandwidth or scale pointwise
maxiter	integer; max number of iterations in the call to optim.
...	additional parameters, currently ignored

### Value

a d by d scale matrix

### References

- Wu, X. (2019). Robust likelihood cross-validation for kernel density estimation. *Journal of Business & Economic Statistics*, 37(4), 761–770. doi:10.1080/07350015.2018.1424633
- Bowman, A.W. (1984). An alternative method of cross-validation for the smoothing of density estimates, *Biometrika*, 71(2), 353–360. doi:10.1093/biomet/71.2.353
- Rudemo, M. (1982). Empirical choice of histograms and kernel density estimators. *Scandinavian Journal of Statistics*, 9(2), 65–78. <http://www.jstor.org/stable/4615859>

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 mig\_lcv

*Likelihood cross-validation for kernel density estimation with MIG*


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**Description**

Given a data matrix over a half-space defined by beta, compute the log density using leave-one-out cross validation, taking in turn an observation as location vector and computing the density of the resulting mixture.

**Usage**

```
mig_lcv(x, beta, Omega)
```

**Arguments**

x	n by d matrix of quantiles
beta	d vector $\beta$ defining the half-space through $\beta^\top \xi > 0$
Omega	d by d positive definite scale matrix $\Omega$

**Value**

the value of the likelihood cross-validation criterion

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mig\_rlcv

*Robust likelihood cross-validation for kernel density estimation*


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**Description**

Given a data matrix over a half-space defined by beta, compute the log density using leave-one-out cross validation, taking in turn an observation as location vector and computing the density of the resulting mixture.

**Usage**

```
mig_rlcv(x, beta, Omega, xsamp, dxsamp)
```

**Arguments**

x	n by d matrix of quantiles
beta	d vector $\beta$ defining the half-space through $\beta^\top \xi > 0$
Omega	d by d positive definite scale matrix $\Omega$
xsamp	matrix of points at which to evaluate the integral
dxsamp	density of points

**Value**

the value of the likelihood cross-validation criterion

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