## Package 'ikde'

## December 14, 2018

Version 0.0.1
<b>Description</b> Estimation of model marginal likelihoods for Bayesian model selection using iterative
kernel density estimation. A multitude of methods exist for performing model selection in
1 1 2 2 1 111 111 11 12 12 1 2 1 1 11 11

general and estimating marginal likelihood in specific, but none are partically well-suited to large models (such as Gaussian processes) applied to relatively limited datasets. Methods are provided to specific and construct Stan models, estimate those models, and estimate the marginal likelihood of those models.

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Imports invgamma, mvtnorm, quantreg
Suggests ggplot2 (>= 3.1.0), knitr (>= 1.20), Rdpack (>= 0.10.1),

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**Suggests** ggplot2 (>= 3.1.0), knitr (>= 1.20), Rdpack (>= 0.10.1), rmarkdown (>= 1.10)

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 ${\color{red}Needs Compilation}\ \ no$ 

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build.model

Build Stan model

## **Description**

Builds and compiles a defined Stan model

## Usage

```
build.model(ikde.model)
```

## **Arguments**

ikde.model

An object of class ikde.model

## **Details**

Builds Stan model using defined ikde.model, then compiles the model and stores DSO for fast running.

#### Value

Returns an ikde.model object with the following elements

data A list of data passed to the Stan program

transformed.data

A list describing data transformations for the Stan program to perform

parameters A list of parameters used in the Stan program

transformed.parameters

A list describing parameter transformations for the Stan program to perform

model A list describing the Stan model

stan.code Stan code for the model

stan.data Data passed to Stan for estimation

stan.dso DSO for Stan model, allows Stan to run model without recompilation

built Boolean indicating whether the model has been built

density.variable

List containing two elements: "name" of the variable on which density estimation should be performed on, and "value" indicating the value the density should be estimated

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#### **Examples**

create.restricted.models

Creates restricted models for IKDE

## **Description**

Creates set of restricted models to be used for posterior density estimation

#### Usage

```
create.restricted.models(ikde.model, eval.point)
```

## Arguments

ikde.model

An object of class ikde.model, does not necessarily have to be built

## **Details**

Posterior density can be estimated by breaking the multi-dimensional density into one-dimensional components. This method creates restricted models from which conditional densities can be estimated. Each real parameter and each entry of vector parameters are restricted one at a time, with values restricted at the specified point.

#### Value

Returns a list of built ikde.models for each restricted model

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#### **Examples**

```
data(lm.generated)
X <- lm.generated$X
y <- lm.generated$y
data <- list(N = list("int<lower=1>", nrow(X)),
             k = list("int<lower=1>", ncol(X)),
             X = list("matrix[N, k]", X),
             y = list("vector[N]", y))
parameters <- list(beta = "vector[k]",</pre>
                   sigma = "real<lower=0>")
model <- list(priors = c("beta ~ normal(0, 10)",</pre>
                         "sigma ~ inv_gamma(1, 1)"),
              likelihood = c("y ~ normal(X * beta, sigma)"))
ikde.model <- define.model(data, parameters, model)</pre>
eval.point \leftarrow list(beta = c(1, 2, 3, 4),
                   sigma = 5)
ikde.model.list <- create.restricted.models(ikde.model, eval.point)</pre>
for (restricted.ikde.model in ikde.model.list){
  cat(restricted.ikde.model$stan.code)
  cat("-----
                        ----\n")
```

define.model

Define Stan model

## **Description**

Defines Stan model, creates model code, and stores input data

## Usage

```
define.model(data, parameters, model, transformed.data = list(),
  transformed.parameters = list())
```

## **Arguments**

data A list of data passed to the Stan program. Should be of the form list(data.name

= list(data.type, data.object)).

parameters A list of parameters used in the Stan program. Should be of the form list(parameter.name

= parameter.type).

model A list describing the Stan model. Should be a list with components "priors" and

"likelihood".

transformed.data

A list describing data transformations for the Stan program to perform. Should be of the form list(variable.name = list(variable.type, variable.expression)).

transformed.parameters

A list describing parameter transformations for the Stan program to perform. Should be of the form list(variable.name = list(variable.type, variable.expression)).

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#### **Details**

Defines inputs to be used for building and eventually fitting Stan model.

#### Value

Returns an ikde.model object with the following elements

data A list of data passed to the Stan program

transformed.data

A list describing data transformations for the Stan program to perform

parameters A list of parameters used in the Stan program

transformed.parameters

A list describing parameter transformations for the Stan program to perform

model A list describing the Stan model

stan.code Stan code for the model

stan.data Data passed to Stan for estimation

stan.dso DSO for Stan model, allows Stan to run model without recompilation

built Boolean indicating whether the model has been built

density.variable

List containing two elements: "name" of the variable on which density estimation should be performed on, and "value" indicating the value the density should be estimated

## **Examples**

evaluate.expression

Evaluate expression from Stan program

#### **Description**

Evaluate expression from Stan program

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#### Usage

```
evaluate.expression(stan.expression, ...)
```

#### **Arguments**

```
stan.expression
```

String representing Stan expression. All variables must be passed in ... .

Any variables present in the parent environment that are needed to evaluate stan.expression

#### **Details**

First, all variables specified in ... are loaded into the environment. Then, all multipliction is replaced by

#### Value

The result of the Stan expression

## **Examples**

```
X <- matrix(1:9, nrow = 3)
b <- c(4, 5, 6)
stan.expression <- "(3 + 2) * X * (5 * b)"
# These results match:
evaluate.expression(stan.expression)
print((3 + 2) * X %*% (5 * b))
# [,1]
# [1,] 1650
# [2,] 2025
# [3,] 2400</pre>
```

evaluate.likelihood

Stan model likelihood evaluation

## Description

Evaluates likelihood of Stan model at specified evaluation point

#### Usage

```
evaluate.likelihood(ikde.model, eval.point)
```

#### **Arguments**

ikde.model An object of class ikde.model which has been builteval.point A list of parameter names and the point to evaluate priors

#### **Details**

Parses sampling statements in ikde.model\$model\$likelihood and evaluates them at the specified evaluation point.

#### Value

A real number indicating value of the log-likelihood at the evaluation point

#### **Examples**

```
data(lm.generated)
X <- lm.generated$X</pre>
y <- lm.generated$y
data <- list(N = list("int<lower=1>", nrow(X)),
              k = list("int < lower = 1 > ", ncol(X)),
              X = list("matrix[N, k]", X),
              y = list("vector[N]", y))
parameters <- list(beta = "vector[k]",</pre>
                    sigma = "real<lower=0>")
model \leftarrow list(priors = c("beta \sim normal(0, 10)",
                           "sigma ~ inv_gamma(1, 1)"),
               likelihood = c("y ~ normal(X * beta, sigma)"))
ikde.model <- define.model(data, parameters, model)</pre>
ikde.model <- build.model(ikde.model)</pre>
eval.point \leftarrow list(beta = c(1, 2, 3, 4), sigma = 5)
# These results match:
evaluate.likelihood(ikde.model, eval.point)
sum(dnorm(y, X %*% eval.point$beta, eval.point$sigma, log = TRUE))
# [1] -1054.093
```

```
evaluate.marginal.likelihood
```

Stan model marginal likelihood evaluation

#### **Description**

Evaluates marginal likelihood of Stan model at the posterior mean

#### Usage

```
evaluate.marginal.likelihood(ikde.model, burn.iter = 1000,
    sample.iter = 1000, control = NULL, refresh = NULL,
    display.output = FALSE, show.trace = FALSE)
```

#### **Arguments**

ikde.model An object of class ikde.model, does not necessarily have to be built

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#### **Details**

Uses evaluate.likelihood, evaluate.priors, and evaluate.posterior to form an estimate of marginal likelihood at the posterior mean.

#### Value

A real number indicating value of the log-marginal-likelihood at the posterior mean

#### **Examples**

evaluate.posterior

Stan model posterior evaluation

#### **Description**

Evaluates posterior of Stan model at the posterior mean

## Usage

```
evaluate.posterior(ikde.model, eval.point, burn.iter = 1000,
  sample.iter = 1000, control = NULL, refresh = NULL,
  display.output = FALSE, show.trace = FALSE)
```

#### **Arguments**

ikde.model An object of class ikde.model, does not necessarily have to be built

## **Details**

Uses list of ikde.model objects created by create.restricted.models to estimate posterior density. Each ikde.model is fit, then conditional posterior density is estimated at the specified point.

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#### Value

A real number indicating value of the log-posterior at the posterior mean

#### **Examples**

```
data(lm.generated)
X <- lm.generated$X</pre>
y <- lm.generated$y
data <- list(N = list("int<lower=1>", nrow(X)),
              k = list("int<lower=1>", ncol(X)),
              X = list("matrix[N, k]", X),
              y = list("vector[N]", y))
parameters <- list(beta = "vector[k]",</pre>
                     sigma = "real<lower=0>")
model <- list(priors = c("beta ~ normal(0, 10)",</pre>
               "sigma ~ inv_gamma(1, 1)"),
likelihood = c("y ~ normal(X * beta, sigma)"))
ikde.model <- define.model(data, parameters, model)</pre>
ikde.model <- build.model(ikde.model)</pre>
stan.fit <- fit.model(ikde.model)</pre>
stan.extract <- rstan::extract(stan.fit)</pre>
eval.point <- list(beta = apply(stan.extract$beta, 2, mean),</pre>
                     sigma = mean(stan.extract$sigma))
evaluate.posterior(ikde.model, eval.point) # Only an estimation, may not exactly match presented result
# [1] -39.95366
```

evaluate.priors

Stan model prior evaluation

#### **Description**

Evaluates prior of Stan model at specified evaluation point

## Usage

```
evaluate.priors(ikde.model, eval.point)
```

## **Arguments**

ikde.model An object of class ikde.model which has been builteval.point A list of parameter names and the point to evaluate priors

#### **Details**

Parses sampling statements in ikde.model\$model\$priors and evaluates them at the specified evaluation point.

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#### Value

A real number indicating value of the log-prior at the evaluation point

#### **Examples**

```
data(lm.generated)
X <- lm.generated$X</pre>
y <- lm.generated$y
data <- list(N = list("int<lower=1>", nrow(X)),
             k = list("int<lower=1>", ncol(X)),
             X = list("matrix[N, k]", X),
             y = list("vector[N]", y))
parameters <- list(beta = "vector[k]",</pre>
                    sigma = "real<lower=0>")
model <- list(priors = c("beta ~ normal(0, 10)",</pre>
                           "sigma ~ inv_gamma(1, 1)"),
               likelihood = c("y ~ normal(X * beta, sigma)"))
ikde.model <- define.model(data, parameters, model)</pre>
ikde.model <- build.model(ikde.model)</pre>
eval.point \leftarrow list(beta = c(1, 2, 3, 4), sigma = 5)
# These results match:
evaluate.priors(ikde.model, eval.point)
sum(dnorm(eval.point$beta, 0, 10, log = TRUE),
    invgamma::dinvgamma(eval.point$sigma, 1, 1, log = TRUE))
# [1] -16.45497
```

evaluate.statement

Evaluate sampling statement from Stan program

## **Description**

Evaluate sampling statement from Stan program

## Usage

```
evaluate.statement(statement, ikde.model, eval.point)
```

#### **Arguments**

statement A string containing a sampling statement
ikde.model An object of class ikde.model, which has been built
eval.point A list of parameter names and the point to evaluate the statement

## **Details**

Parses the given sampling statement and evaluates it at the specified evaluation point. The ikde.model object and eval.point object are needed to resolve variable values in the statement.

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#### Value

A real number indicating value of the log-density of the statement at the evaluation point

#### **Examples**

```
data(lm.generated)
X <- lm.generated$X</pre>
y <- lm.generated$y
data <- list(N = list("int<lower=1>", nrow(X)),
              k = list("int<lower=1>", ncol(X)),
             X = list("matrix[N, k]", X),
             y = list("vector[N]", y))
parameters <- list(beta = "vector[k]",</pre>
                    sigma = "real<lower=0>")
model <- list(priors = c("beta ~ normal(0, 10)",</pre>
                           "sigma ~ inv_gamma(1, 1)"),
               likelihood = c("y ~ normal(X * beta, sigma)"))
ikde.model <- define.model(data, parameters, model)</pre>
ikde.model <- build.model(ikde.model)</pre>
statement <- ikde.model$model$likelihood[1]</pre>
eval.point \leftarrow list(beta = c(1, 2, 3, 4), sigma = 5)
# These results match:
evaluate.statement(statement, ikde.model, eval.point)
sum(dnorm(y, mean = X %*% eval.point$beta, sd = eval.point$sigma, log = TRUE))
# [1] -1054.093
```

fit.model

Fits Stan model

## Description

Uses a built ikde.model to draw samples from posterior distribution using Stan.

## Usage

```
fit.model(ikde.model, burn.iter = 1000, sample.iter = 1000,
  chains = 1, control = NULL, refresh = NULL,
  display.output = FALSE)
```

#### **Arguments**

ikde.model

An object of class ikde.model which has been built

## Details

Takes a built ikde.model object, which contains model DSO, and fits the model using rstan::stan.

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#### Value

An object of S4 class stanfit. See rstan::stan for more details.

## **Examples**

```
data(lm.generated)
X <- lm.generated$X</pre>
y <- lm.generated$y
data <- list(N = list("int<lower=1>", nrow(X)),
              k = list("int<lower=1>", ncol(X)),
X = list("matrix[N, k]", X),
              y = list("vector[N]", y))
parameters <- list(beta = "vector[k]",</pre>
                     sigma = "real<lower=0>")
model \leftarrow list(priors = c("beta \sim normal(0, 10)",
                            "sigma ~ inv_gamma(1, 1)"),
               likelihood = c("y ~ normal(X * beta, sigma)"))
ikde.model <- define.model(data, parameters, model)</pre>
ikde.model <- build.model(ikde.model)</pre>
stan.fit <- fit.model(ikde.model)</pre>
stan.extract <- extract(stan.fit)</pre>
print(apply(stan.extract$beta, 2, mean)) # Only an estimation, may not exactly match presented result
# [1] 3.236087 1.629510 4.496279 1.211404
```

lm.generated

Randomly generated multivariate linear model data

## Description

A dataset for estimation of linear models

## Usage

lm.generated

## **Format**

A list with two components:

- **X** Matrix of independent variables
- y Vector of dependent variable observations

## **Details**

Generated with the following code:

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```
set.seed(100)

N <- 100
k <- 4
sd <- 10

X <- cbind(1, matrix(runif(N * (k - 1), -10, 10), ncol = k - 1))
beta <- runif(k, -5, 5)
y <- X
y <- c(y)</pre>
```

prostatic.nodes

Prostatic nodal development data

## Description

A dataset replicated from Chib (1995) indicating presence of prostatic nodal development among patients prostate cancer

## Usage

```
prostatic.nodes
```

#### **Format**

A data frame with 53 observations of 7 variables:

Case Patient identifier

- y Binary outcome indicating nodal development
- X.1 Explanatory variable
- X.2 Explanatory variable
- X.3 Binary explanatory variable
- X.4 Binary explanatory variable
- X.5 Binary explanatory variable

#### **Details**

These data were replicated from Chib (1995)

#### References

Chib S (1995). "Marginal likelihood from the Gibbs output." *Journal of the American Statistical Association*, **90**(432), 1313–1321.

stan.dist.to.r.dist Mapping between Stan and R distribution functions

## Description

Mapping between Stan and R distribution functions

## Usage

```
stan.dist.to.r.dist
```

#### **Format**

An object of class list of length 12.

## **Details**

A list of Stan distributions, associated R distribution functions, and arguments to those functions.

```
stan.operator.to.r.operator
```

Mapping between Stan and R operators

## Description

Mapping between Stan and R operators

## Usage

```
stan.operator.to.r.operator
```

## **Format**

An object of class list of length 3.

#### **Details**

A list of Stan operators (regex) and associated R operators.

%stan\*%

%stan\*%

Function to replicate multiplication in Stan

## **Description**

Function to replicate multiplication in Stan

## Usage

```
x %stan*% y
```

## **Arguments**

```
x First term in producty Second term in product
```

## **Details**

Accepts arguments x and y. If either is a singleton, returns the value of x\*y (in R notation). If both arguments are matrices or vectors, returns x

## Value

Returns an object of the same type as the base

## **Examples**

```
X <- matrix(1:9, nrow = 3)
b <- c(4, 5, 6)

(3 + 2) * X %stan*% (5 * b)
#      [,1]
# [1,] 1650
# [2,] 2025
# [3,] 2400</pre>
```

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