

Package ‘BSCB’

July 23, 2026

Title Bayesian Simultaneous Credible Bands for Polynomial Regression

Version 1.0.2

Description Provides functions to construct two-sided Bayesian simultaneous credible bands (BSCBs) for the regression curve in univariate polynomial regression over a finite covariate interval. Six methods are implemented, including Normal-Gamma conjugate priors (with empirical Bayes, unit-information, and g-prior hyperparameter specifications), non-conjugate priors fitted via Hamiltonian Monte Carlo (HMC) using 'cmdstanr', and a non-informative independent Jeffreys prior approach. Also includes functions for computing the empirical simultaneous coverage rate (ESCR) and posterior simultaneous coverage probability (PSCP), enabling performance comparison across methods. The methodology is described in:
Yang et al. (2026). ``Bayesian simultaneous credible bands for polynomial regression" <[doi:10.48550/arXiv.2606.28015](https://doi.org/10.48550/arXiv.2606.28015)>.

License MIT + file LICENSE

Encoding UTF-8

Suggests DEoptim, testthat (>= 3.0.0), cmdstanr, knitr, rmarkdown, ggplot2

VignetteBuilder knitr

Additional_repositories <https://stan-dev.r-universe.dev>

Imports mvtnorm, MASS, OptimalDesign, instantiate, posterior

Config/testthat/edition 3

Depends R (>= 4.0)

Date 2026-07-23

URL <https://github.com/fannyyang73/BSCB>

Config/instantiate/stan-dir src/stan

BugReports <https://github.com/fannyyang73/BSCB/issues>

Config/roxygen2/version 8.0.0

NeedsCompilation yes

Author Fei Yang [aut, cre, cph] (ORCID:

<<https://orcid.org/0000-0002-3253-976X>>),

Yang Han [aut, cph] (ORCID: <<https://orcid.org/0000-0002-4143-7765>>),

Wei Liu [aut, cph] (ORCID: <<https://orcid.org/0000-0002-4719-0345>>),

Ian Hall [aut, cph] (ORCID: <<https://orcid.org/0000-0002-3033-2335>>)

Maintainer Fei Yang <fei.yang@manchester.ac.uk>

Repository CRAN

Date/Publication 2026-07-23 16:50:25 UTC

Contents

compute_bpcb_ind_jeffreys	2
compute_bscb_conjugate	5
compute_bscb_hmc	8
compute_bscb_ind_jeffreys	10
coverage_ESCR	13
coverage_PSCP	14
create_order_form	15
find_global_maximum	16
find_global_maximum_h_all	17
fn_Bayes_ECR	17
fn_Bayes_PCP	18
fn_Freq_ECR	18
fn_neg_Bayes_ECR	19
fn_neg_Bayes_PCP	19
f_L_SCB	20
f_U_SCB	20
generate_simulation_data	21
L_SCB	23
sup_T_Bayes_ESCR	23
sup_T_Bayes_PSCP	24
sup_T_Freq_ESCR	25
sup_T_simFSCB	25
U_SCB	26

Index	27
--------------	-----------

compute_bpcb_ind_jeffreys

BPCB-I-J: the Bayesian pointwise credible band using the independent Jeffreys prior

Description

Constructs a $(1 - \alpha)$ two-sided Bayesian pointwise credible band (BPCB) for polynomial regression using the independent Jeffreys prior. Unlike the simultaneous credible band, the critical constant λ is derived analytically from the marginal t-distribution as $t_{n-p-1}^{\alpha/2}$.

Usage

```
compute_bpcb_ind_jeffreys(
  X,
  Y,
  alpha = 0.05,
  a = NULL,
  b = NULL,
  L = 50000,
  AR_setting = 0,
  rho = NULL,
  theta_true = NULL,
  verbose = TRUE
)
```

Arguments

X	Numeric matrix of dimension $n \times (p + 1)$. Design matrix with intercept in the first column.
Y	Numeric vector of length n . Response variable.
alpha	Numeric. Nominal mis-coverage level; the band targets $1 - \alpha$ pointwise coverage. Default is 0.05.
a	Numeric. Left endpoint of the covariate domain $[a, b]$. Inferred from $X[, 2]$ if NULL.
b	Numeric. Right endpoint of the covariate domain $[a, b]$. Inferred from $X[, 2]$ if NULL.
L	Integer. Not used in this function (included for API consistency with other compute_bscb_* functions). Default is 500000.
AR_setting	Integer. Error covariance structure: 0 = i.i.d. errors (default); 1 = AR(1) errors.
rho	Numeric. AR(1) coefficient. Required when AR_setting = 1.
theta_true	Numeric vector of length $p + 1$. True regression coefficients. Optional; stored in the output for diagnostic use.
verbose	Logical. If TRUE (default), prints the value of the critical constant lambda.

Value

An object of class "bpcb_fit", a list containing:

lambda Critical constant $t_{n-p-1}^{\alpha/2}$ for the credible band.

lower_bound Function: computes the lower band at a given x.

upper_bound Function: computes the upper band at a given x.

mu_star Posterior mean of θ (GLS estimate).

dof Degrees of freedom of the marginal posterior ($n - p - 1$).

scale_mat Scale matrix Σ_0 of the marginal multivariate-t posterior distribution of θ .

cov_theta Posterior covariance matrix of θ . The posterior covariance matrix equals $\text{Cov}(\theta) = \frac{\nu}{\nu-2}\Sigma_0$, where ν is the degrees of freedom (dof).

x_range Covariate domain $[a, b]$.

theta_true True parameters (if supplied).

method Character string "independent_jeffreys".

params List of configuration parameters.

See Also

[compute_bscb_ind_jeffreys](#) for the simultaneous version, [compute_bscb_conjugate](#) for the conjugate prior version.

Examples

```
# Quadratic model with i.i.d. errors
set.seed(123)
n <- 50
x <- seq(-5, 5, length.out = n)
X <- cbind(1, x, x^2)
theta_true <- c(-6, -3, 0.25)
Y <- X %*% theta_true + rnorm(n, sd = 0.2)

fit <- compute_bpcb_ind_jeffreys(
  X      = X,
  Y      = Y,
  alpha  = 0.05,
  a      = -5,
  b      = 5,
  theta_true = theta_true,
  verbose = FALSE
)

# Critical constant (t quantile)
fit$lambda

# Evaluate the band over a grid
x_seq <- seq(-5, 5, length.out = 200)
lower_vec <- fit$lower_bound(x_seq)
upper_vec <- fit$upper_bound(x_seq)

# Plot
plot(x_seq, lower_vec, type = "l", col = "red", lty = 2, lwd = 2,
     ylim = range(c(lower_vec, upper_vec, Y)),
     xlab = "x", ylab = "y",
     main = "95% Bayesian Pointwise Credible Band (Indep. Jeffreys)")
lines(x_seq, upper_vec, col = "red", lty = 2, lwd = 2)
lines(x_seq, cbind(1, x_seq, x_seq^2) %*% theta_true,
     col = "blue", lwd = 2)
points(x, Y, pch = 16, col = "gray")
legend("topright",
     legend = c("True curve", "Data", "95% BPCB-J"),
```

```
col    = c("blue", "gray", "red"),
lty    = c(1, NA, 2),
pch    = c(NA, 16, NA))
```

compute_bscb_conjugate

BSCB-C: Bayesian Simultaneous Credible Band under the Normal-Gamma Conjugate Prior

Description

Constructs a $(1 - \alpha)$ two-sided Bayesian simultaneous credible band for polynomial regression using the normal-gamma conjugate prior. The marginal posterior of θ follows a multivariate-t distribution. The critical constant λ is estimated via Monte Carlo sampling.

Usage

```
compute_bscb_conjugate(
  X,
  Y,
  alpha = 0.05,
  a = NULL,
  b = NULL,
  L = 5e+05,
  AR_setting = 0,
  rho = NULL,
  hyperparameter = c("empirical", "unit_info", "g_prior"),
  optimize_type = c("P", "G", "D"),
  theta_true = NULL,
  verbose = TRUE
)
```

Arguments

X	Numeric matrix of dimension $n \times (p + 1)$. Design matrix with intercept in the first column.
Y	Numeric vector of length n . Response variable.
alpha	Numeric. Nominal mis-coverage level; the band targets $1 - \alpha$ simultaneous coverage. Default is 0.05.
a	Numeric. Left endpoint of the covariate domain $[a, b]$. Inferred from $X[, 2]$ if NULL.
b	Numeric. Right endpoint of the covariate domain $[a, b]$. Inferred from $X[, 2]$ if NULL.
L	Integer. Number of Monte Carlo draws for computing the critical constant λ . Default is 500000.

AR_setting	Integer. Error covariance structure: 0 = i.i.d. errors (default); 1 = AR(1) errors.
rho	Numeric. AR(1) coefficient. Required when AR_setting = 1.
hyperparameter	Character. Hyperparameter specification for the Normal-Gamma prior: "empirical" = empirical Bayes; "unit_info" = unit-information prior; "g_prior" = Zellner's g-prior.
optimize_type	Character. Method for computing $\sup_{x \in [a, b]} T(x)$: "P" = polyroot analytical method (recommended); "G" = global optimisation; "D" = differential evolution (DEoptim).
theta_true	Numeric vector of length $p + 1$. True regression coefficients. Optional; stored in the output for diagnostic use.
verbose	Logical. If TRUE (default), prints progress messages including the value of the critical constant lambda.

Value

An object of class "bscb_fit", a list containing:

lambda Critical constant for the credible band.

lower_bound Function: computes the lower band at a given x.

upper_bound Function: computes the upper band at a given x.

mu_star Posterior mean of θ .

dof Degrees of freedom of the marginal posterior.

scale_mat Scale matrix Σ_0 of the marginal multivariate-t posterior distribution of θ .

cov_theta Posterior covariance matrix of θ . The posterior covariance matrix equals $\text{Cov}(\theta) = \frac{\nu}{\nu-2} \Sigma_0$, where ν is the degrees of freedom (dof).

x_range Covariate domain $[a, b]$.

lambda_samples Monte Carlo samples used to compute λ .

theta_true True parameters (if supplied).

method Character string "conjugate".

params List of configuration parameters.

See Also

[compute_bscb_ind_jeffreys](#) for the independent Jeffreys prior version, [compute_bpceb_ind_jeffreys](#) for the pointwise band.

Examples

```
# Example 1: Simple quadratic model with i.i.d. errors

set.seed(123)
n <- 50
p <- 2
x <- seq(-5, 5, length.out = n)
X <- cbind(1, x, x^2)
```

```

theta_true <- c(-6, -3, 0.25)
e_sd <- 0.2

# Generate response
epsilon <- rnorm(n, mean = 0, sd = e_sd)
Y <- X %*% theta_true + epsilon

# Notably, this is a quick example. In theory, L should set to L=500,000 or larger;
fit <- compute_bscb_conjugate(X=X, Y=Y, alpha = 0.05, a = -5, b = 5, L = 5000,
                             AR_setting = 0, # 0: iid error; 1: autoregressive error
                             rho = NULL,
                             hyperparameter = "empirical",
                             optimize_type = "P",
                             theta_true = theta_true,
                             verbose = FALSE)

# View results
print(fit$lambda)          # Critical value
print(fit$mu_star)         # Posterior mean of theta
print(fit$cov_theta)       # Posterior covariance matrix

# Compute BSCB-C at a specific point
x_new <- 0.5
lower <- fit$lower_bound(x_new)
upper <- fit$upper_bound(x_new)
cat("At x =", x_new, ": [", lower, ", ", upper, "]\n")

# Vectorized computation for plotting
x_seq <- seq(-5, 5, length.out = 1000)
lower_vec <- fit$lower_bound(x_seq)
upper_vec <- fit$upper_bound(x_seq)
y_true <- cbind(1, x_seq, x_seq^2) %*% theta_true

# Visualization
plot(x_seq, lower_vec, type = "l", col = "red", lty = 2, lwd = 2,
     ylim = range(c(lower_vec, upper_vec, Y)),
     xlab = "x", ylab = "y",
     main = "95% Bayesian Simultaneous Credible Band (Conjugate Prior)")
lines(x_seq, upper_vec, col = "red", lty = 2, lwd = 2)
lines(x_seq, y_true, col = "blue", lwd = 2)
points(x, Y, pch = 16, col = "gray")
legend("topright",
      legend = c("True curve", "Data", "95% BSCB-C"),
      col = c("blue", "gray", "red"),
      lty = c(1, NA, 2),
      pch = c(NA, 16, NA),
      lwd = 2)

```

compute_bscb_hmc

Compute BSCB via Hamiltonian Monte Carlo

Description

Constructs a Bayesian Simultaneous Credible Band (BSCB) for polynomial regression under non-conjugate priors using Hamiltonian Monte Carlo (HMC) via Stan. Supports two prior specifications: Normal-Normal and Normal-half-Cauchy. The critical constant λ is estimated by Monte Carlo, and the Posterior Simultaneous Coverage Probability (PSCP) is also returned.

Usage

```
compute_bscb_hmc(
  Y,
  X,
  V = diag(nrow(X)),
  a,
  b,
  theta_true = NULL,
  alpha = 0.05,
  prior_type = c("normal_half_cauchy", "normal_normal"),
  normal_theta_sd = 10,
  normal_sigma_sd = 5,
  cauchy_scale = 2,
  iter_sampling = 4000,
  iter_warmup = 4000,
  chains = 4,
  thin_number = 1,
  adapt_delta = 0.95,
  max_treedepth = 15,
  AR_setting = 0,
  rho = 0,
  optimize_type = c("P", "G", "D"),
  L = 5e+05,
  draw_num = 10000
)
```

Arguments

Y	Numeric vector of responses of length n.
X	Design matrix of dimension $n \times (p+1)$, including an intercept column. The second column must contain the raw covariate values.
V	Error covariance matrix of dimension $n \times n$. Use <code>diag(n)</code> for i.i.d. errors (default).
a	Left endpoint of the covariate domain $[a, b]$.
b	Right endpoint of the covariate domain $[a, b]$.

theta_true	Numeric vector of true regression coefficients of length $p+1$. Used to evaluate ESCR in simulation studies. Set to NULL (default) when the true coefficients are unknown.
alpha	Nominal miscoverage rate. The credible band targets $1 - \alpha$ simultaneous coverage. Default is 0.05.
prior_type	Character string specifying the prior on (θ, σ) . Either "normal_half_cauchy" (default, recommended) or "normal_normal".
normal_theta_sd	Prior standard deviation for each component of θ under the Normal-Normal prior. Default is 10.
normal_sigma_sd	Prior standard deviation for σ under the Normal-Normal prior. Default is 5.
cauchy_scale	Scale parameter of the half-Cauchy prior on σ under the Normal-half-Cauchy prior. Default is 2.
iter_sampling	Number of post-warmup HMC draws per chain. Default is 4000.
iter_warmup	Number of warmup draws per chain. Default is 4000.
chains	Number of Markov chains. Default is 4.
thin_number	Positive integer. Thinning interval for posterior draws. A value of k retains every k -th draw from each chain. Default is 1 (no thinning).
adapt_delta	Target acceptance probability for the NUTS sampler. Default is 0.95.
max_treedepth	Maximum tree depth for the NUTS sampler. Default is 15.
AR_setting	Integer. 0 for i.i.d. errors (default), 1 for AR(1) errors.
rho	AR(1) autocorrelation coefficient. Only used when <code>AR_setting == 1</code> . Default is 0.
optimize_type	Character. Method for computing $\sup_{x \in [a, b]} T(x)$: "P" = polyroot analytical method (recommended, default); "G" = global optimisation (grid-based); "D" = differential evolution (DEoptim).
L	Number of Monte Carlo draws used to estimate the critical constant λ . Default is 500000.
draw_num	Number of Monte Carlo draws used to estimate the PSCP. Default is 10000.

Value

An object of class "bscb_fit", which is a list with the following components:

lambda	Estimated critical constant at level $1 - \alpha$.
lower_bound	Lower credible band evaluated on a fine grid over $[a, b]$.
upper_bound	Upper credible band evaluated on a fine grid over $[a, b]$.
theta_true	True regression coefficients (if supplied).
order_form	Polynomial order form used internally.
mu_star	Posterior mean of θ (length $p+1$).
cov_theta	Posterior covariance matrix of θ .

theta_mat	Matrix of posterior draws, (chains * floor(iter_sampling / thin_number)) x (p+1).
x_range	Numeric vector c(a, b).
call	The matched call.
method	Character string "HMC".
n	Sample size.
p	Polynomial degree.
alpha	Nominal miscoverage rate.
data	List containing the design matrix X and response vector Y.
lambda_samples	Numeric vector of length L containing the Monte Carlo supremum draws used to derive lambda.
params	List of additional settings: AR_setting, rho, prior_type, normal_theta_sd, normal_sigma_sd, cauchy_scale, iter_sampling, iter_warmup, chains, thin_number, L, draw_num, optimize_type.

See Also

[compute_bscb_conjugate](#), [compute_bscb_ind_jeffreys](#)

Examples

```
set.seed(42)
n <- 20; p <- 2
x_seq <- seq(-5, 5, length.out = n)
X <- cbind(1, x_seq, x_seq^2)
theta_true <- c(-6, -3, 0.25)
Y <- as.numeric(X %*% theta_true + rnorm(n, sd = 0.2))
fit <- compute_bscb_hmc(
  Y = Y, X = X, V = diag(n),
  a = -5, b = 5,
  theta_true = theta_true,
  prior_type = "normal_half_cauchy",
  L = 1000, draw_num = 500 # small values for illustration only
)
fit$lambda
fit$params$prior_type
```

compute_bscb_ind_jeffreys

BSCB-J: Bayesian Simultaneous Credible Band under the Independent Jeffreys Prior

Description

Constructs a $(1 - \alpha)$ two-sided Bayesian simultaneous credible band for polynomial regression using the independent Jeffreys prior. The marginal posterior of θ follows a multivariate-t distribution with degrees of freedom $n - p - 1$.

Usage

```
compute_bscb_ind_jeffreys(
  X,
  Y,
  alpha = 0.05,
  a = NULL,
  b = NULL,
  L = 5e+05,
  AR_setting = 0,
  rho = NULL,
  optimize_type = c("P", "G", "D"),
  theta_true = NULL,
  verbose = TRUE
)
```

Arguments

X	Numeric matrix of dimension $n \times (p + 1)$. Design matrix with intercept in the first column.
Y	Numeric vector of length n . Response variable.
alpha	Numeric. Nominal mis-coverage level; the band targets $1 - \alpha$ simultaneous coverage. Default is 0.05.
a	Numeric. Left endpoint of the covariate domain $[a, b]$. Inferred from $X[, 2]$ if NULL.
b	Numeric. Right endpoint of the covariate domain $[a, b]$. Inferred from $X[, 2]$ if NULL.
L	Integer. Number of Monte Carlo draws for computing the critical constant λ . Default is 500000.
AR_setting	Integer. Error covariance structure: 0 = i.i.d. errors (default); 1 = AR(1) errors.
rho	Numeric. AR(1) coefficient. Required when <code>AR_setting = 1</code> .
optimize_type	Character. Method for computing $\sup_{x \in [a, b]} T(x)$: "P" = polyroot analytical method (recommended); "G" = global optimisation; "D" = differential evolution (DEoptim).
theta_true	Numeric vector of length $p + 1$. True regression coefficients. Optional; stored in the output for diagnostic use.
verbose	Logical. If TRUE (default), prints progress messages.

Value

An object of class "bscb_fit", a list containing:

lambda Critical constant for the credible band.

lower_bound Function: computes the lower band at a given x .

upper_bound Function: computes the upper band at a given x .

mu_star Posterior mean of θ (GLS estimate).

dof Degrees of freedom of the marginal posterior ($n - p - 1$).

scale_mat Scale matrix Σ_0 of the marginal multivariate-t posterior distribution of θ .

cov_theta Posterior covariance matrix of θ . The posterior covariance matrix equals $\text{Cov}(\theta) = \frac{\nu}{\nu-2} \Sigma_0$, where ν is the degrees of freedom (dof).

x_range Covariate domain $[a, b]$.

lambda_samples Monte Carlo samples used to compute λ .

theta_true True parameters (if supplied).

method Character string "independent_jeffreys".

params List of configuration parameters.

Examples

```
# Quadratic model with i.i.d. errors
# This is for a quick demonstration;
# For actual use, please set L = 500,000.
set.seed(123)
n <- 50
x <- seq(-5, 5, length.out = n)
X <- cbind(1, x, x^2)
theta_true <- c(-6, -3, 0.25)
Y <- X %*% theta_true + rnorm(n, sd = 0.2)

fit <- compute_bscb_ind_jeffreys(
  X      = X,
  Y      = Y,
  alpha  = 0.05,
  a      = -5,
  b      = 5,
  L      = 5000,
  theta_true = theta_true,
  verbose = FALSE
)

# Critical constant
fit$lambda

# Evaluate the band over a grid
x_seq    <- seq(-5, 5, length.out = 200)
lower_vec <- fit$lower_bound(x_seq)
upper_vec <- fit$upper_bound(x_seq)
```

```
# Full example with recommended L
fit_full <- compute_bscb_ind_jeffreys(
  X = X, Y = Y, alpha = 0.05, a = -5, b = 5,
  L = 50000, theta_true = theta_true
)
```

coverage_ESCR	<i>Compute the coverage of BSCB</i>
---------------	-------------------------------------

Description

Compute the coverage of BSCB

Usage

```
coverage_ESCR(fit, optimize_type = c("P", "G", "D"), verbose = FALSE)
```

Arguments

fit	A BSCB fit object containing lambda, mu_star, cov_theta, theta_true, x_range, order_form
optimize_type	Character. Method for computing $\sup_{x \in [a, b]} T(x)$: "P" = polyroot function (recommended); "G" = global optimisation; "D" = Doptimize function from package DEoptim.
verbose	Logical. If TRUE (default), prints the value of the critical constant lambda.

Value

Integer: 1 if covered, 0 if not covered

Examples

[illegible]

```

        verbose = FALSE)

# Check the empirical simultaneous coverage rate (ESCR)
is_covered <- coverage_ESCR(fit, optimize_type = "P", verbose = TRUE)
cat("Coverage indicator:", is_covered, "\n")

```

coverage_PSCP

*Compute the Posterior Simultaneous Coverage Probability (PSCP)***Description**

Estimates the posterior simultaneous coverage probability (PSCP) of a constructed BSCB by Monte Carlo integration over the posterior distribution of θ . For each posterior draw $\hat{\theta}$, the supremum $\sup_{x \in [a, b]} T(x)$ is computed and compared against the critical constant λ . The PSCP is the proportion of draws for which $\sup T(x) \leq \lambda$.

Usage

```

coverage_PSCP(
  fit,
  draw_num = 10000,
  optimize_type = c("P", "G", "D"),
  verbose = FALSE
)

```

Arguments

fit	An object of class "bscb_fit" returned by compute_bscb_conjugate or compute_bscb_ind_jeffreys . Must contain lambda, mu_star, cov_theta, dof, x_range, and order_form.
draw_num	Integer. Number of Monte Carlo draws for estimating PSCP. Default is 10000.
optimize_type	Character. Method for computing $\sup_{x \in [a, b]} T(x)$: "P" = polyroot analytical method (recommended); "G" = global optimisation; "D" = differential evolution (DEoptim).
verbose	Logical. If TRUE, prints the estimated PSCP value. Default is FALSE.

Value

Numeric. Estimated posterior simultaneous coverage probability, a value in $[0, 1]$.

See Also

[coverage_ESCR](#), [compute_bscb_conjugate](#), [compute_bscb_ind_jeffreys](#)

Examples

```
# This is for a quick demonstration;
# For actual use, please set L = 500000 and draw_num = 10000.
set.seed(123)
n <- 50
x <- seq(-5, 5, length.out = n)
X <- cbind(1, x, x^2)
theta_true <- c(-6, -3, 0.25)
Y <- X %*% theta_true + rnorm(n, sd = 0.2)

fit <- compute_bscb_conjugate(
  X      = X,
  Y      = Y,
  alpha  = 0.05,
  a      = -5,
  b      = 5,
  L      = 1000,
  theta_true = theta_true,
  verbose = FALSE
)

coverage_PSCP(fit, draw_num = 500, optimize_type = "P", verbose = TRUE)

# Full example with recommended draw_num
coverage_PSCP(fit, draw_num = 10000, optimize_type = "P")
```

create_order_form	Create a polynomial basis vector function
-------------------	---

Description

Returns a function that maps a scalar x to the polynomial basis vector $(1, x, x^2, \dots, x^p)$.

Usage

```
create_order_form(p)
```

Arguments

p Non-negative integer. Polynomial degree.

Value

A function $f(x)$ that returns $(1, x, \dots, x^p)$ as a numeric vector (for scalar x) or matrix (for vector x).

Examples

```
f <- create_order_form(p = 2)
f(3) # returns c(1, 3, 9)
f(c(1, 2, 3)) # returns a matrix
```

find_global_maximum	<i>Find the global maximum of $T(x)$ via grid search and local optimisation</i>
---------------------	--

Description

Fallback method using a coarse grid search combined with uniroot and optimize. For most cases, find_global_maximum_h_all (Liu's analytic method) is preferred.

Usage

```
find_global_maximum(
  fn,
  a,
  b,
  order_form,
  theta,
  mu_star,
  cov_mat,
  tol = 1e-06,
  n_grid = 100
)
```

Arguments

fn	Function. The objective function $T(x)$ to maximise.
a	Numeric. Left endpoint of the search interval.
b	Numeric. Right endpoint of the search interval.
order_form	Function. Polynomial basis function from create_order_form.
theta	Numeric vector. Posterior draw of theta.
mu_star	Numeric vector. Posterior mean of theta.
cov_mat	Numeric matrix. Covariance matrix.
tol	Numeric. Numerical tolerance. Default 1e-6.
n_grid	Integer. Number of grid points. Default 100.

Value

A list with components maximum, x_max, and all_candidates.

find_global_maximum_h_all

Find the global maximum of $T(x)$ analytically via polyroot (Liu's method)

Description

Computes $\sup_{x \in [a, b]} T(x)$ by finding the stationary points of $h(x) = T(x)^2 = N(x)/D(x)$ via polyroot, where $N(x)$ and $D(x)$ are polynomials. This is the recommended method.

Usage

```
find_global_maximum_h_all(a, b, d, cov_mat)
```

Arguments

a	Numeric. Left endpoint of the search interval.
b	Numeric. Right endpoint of the search interval.
d	Numeric vector of length $p + 1$. Direction vector $(\theta - \mu^*)$.
cov_mat	Numeric matrix of dimension $(p + 1) \times (p + 1)$. Covariance matrix.

Value

A list with components maximum, x_max, and all_candidates.

fn_Bayes_ECR

T(x) for computing ESCR: uses true parameter theta_true

Description

$T(x)$ for computing ESCR: uses true parameter theta_true

Usage

```
fn_Bayes_ECR(x, theta_true, mu_star, cov_theta)
```

Arguments

x	Numeric. Evaluation point.
theta_true	Numeric vector. True regression coefficients.
mu_star	Numeric vector. Posterior mean of theta.
cov_theta	Numeric matrix. Posterior covariance of theta.

Value

Numeric scalar. Value of $T(x)$.

fn_Bayes_PCP	<i>T(x) for computing lambda (PSCP): uses posterior draw theta_hat</i>
--------------	--

Description

T(x) for computing lambda (PSCP): uses posterior draw theta_hat

Usage

```
fn_Bayes_PCP(x, theta_hat, mu_star, cov_theta)
```

Arguments

x	Numeric. Evaluation point.
theta_hat	Numeric vector. Posterior draw of theta.
mu_star	Numeric vector. Posterior mean of theta.
cov_theta	Numeric matrix. Posterior covariance of theta.

Value

Numeric scalar. Value of T(x).

fn_Freq_ECR	<i>T(x) to compute ESCR for frequentist methods</i>
-------------	---

Description

T(x) to compute ESCR for frequentist methods

Usage

```
fn_Freq_ECR(x, theta_true, lm_theta_hat, S, inv)
```

Arguments

x	Numeric. Evaluation point.
theta_true	Numeric vector. True regression coefficients.
lm_theta_hat	Numeric vector. OLS estimate of theta.
S	Numeric. Residual standard error.
inv	Numeric matrix. Inverse of $X^T X$.

Value

Numeric scalar. Value of T(x).

fn_neg_Bayes_ECR	<i>T(x) for computing ESCR for DEoptim: uses true parameter theta_true</i>
------------------	--

Description

T(x) for computing ESCR for DEoptim: uses true parameter theta_true

Usage

```
fn_neg_Bayes_ECR(x, theta_true, mu_star, cov_theta)
```

Arguments

x	Numeric. Evaluation point.
theta_true	Numeric vector. True regression coefficients.
mu_star	Numeric vector. Posterior mean of theta.
cov_theta	Numeric matrix. Posterior covariance of theta.

Value

Numeric scalar. Value of T(x).

fn_neg_Bayes_PCP	<i>Negative T(x) for minimisation-based optimisers (e.g. DEoptim)</i>
------------------	---

Description

Negative T(x) for minimisation-based optimisers (e.g. DEoptim)

Usage

```
fn_neg_Bayes_PCP(x, theta_hat, mu_star, cov_theta)
```

Arguments

x	Numeric. Evaluation point.
theta_hat	Numeric vector. Posterior draw of theta.
mu_star	Numeric vector. Posterior mean of theta.
cov_theta	Numeric matrix. Posterior covariance of theta.

Value

Numeric scalar. Negative value of T(x).

f_{L_SCB}	<i>Vertical distance from true curve to lower band boundary</i>
--------------	---

Description

Vertical distance from true curve to lower band boundary

Usage

```
f_L_SCB(x, cov_theta, mu_star, lambda_best_optim, theta_true)
```

Arguments

x	Numeric. Evaluation point.
cov_theta	Numeric matrix. Posterior covariance of theta.
mu_star	Numeric vector. Posterior mean of theta.
lambda_best_optim	Numeric. Critical constant lambda.
theta_true	Numeric vector. True regression coefficients.

Value

Numeric. Positive if true curve is above lower bound.

f_{U_SCB}	<i>Vertical distance from true curve to upper band boundary</i>
--------------	---

Description

Vertical distance from true curve to upper band boundary

Usage

```
f_U_SCB(x, cov_theta, mu_star, lambda_best_optim, theta_true)
```

Arguments

x	Numeric. Evaluation point.
cov_theta	Numeric matrix. Posterior covariance of theta.
mu_star	Numeric vector. Posterior mean of theta.
lambda_best_optim	Numeric. Critical constant lambda.
theta_true	Numeric vector. True regression coefficients.

Value

Numeric. Positive if true curve is below upper bound.

generate_simulation_data

Generate simulation datasets for polynomial regression

Description

Generates a fixed design matrix X and a list of response vectors Y for use in simulation studies of Bayesian simultaneous credible bands. The design can be either equally-spaced (ES) or D-optimal (DO).

Usage

```
generate_simulation_data(
  p,
  n,
  e_sd,
  theta_true,
  a = -5,
  b = 5,
  replication = 2,
  design_index = 2,
  center_index = 1,
  n_ES_x = n,
  n_DO_init_x = 3e+05,
  AR_index = 0,
  rho = 0.1,
  batch_index = 1,
  seed = NULL
)
```

Arguments

<code>p</code>	Integer. Polynomial degree. Must be 1, 2, or 3.
<code>n</code>	Integer. Sample size.
<code>e_sd</code>	Numeric. Error standard deviation (sigma in the paper).
<code>theta_true</code>	Numeric vector of length $p + 1$. True regression coefficients.
<code>a</code>	Numeric. Left endpoint of the covariate domain $[a, b]$.
<code>b</code>	Numeric. Right endpoint of the covariate domain $[a, b]$.
<code>replication</code>	Integer. Number of simulation replications.
<code>design_index</code>	Integer. Design type: 1 = equally-spaced (ES); 2 = D-optimal (DO).
<code>center_index</code>	Integer. Centering of covariates: 1 = mean-centred (default); 0 = uncentred; 2 = standardised.
<code>n_ES_x</code>	Integer. Number of equally-spaced design points. Only used when <code>design_index</code> = 1.

<code>n_D0_init_x</code>	Integer. Candidate pool size for D-optimal search. A large value (e.g. 300000) ensures that 6 support points are selected. Only used when <code>design_index = 2</code> .
<code>AR_index</code>	Integer. Error structure: 0 = i.i.d. (default); 1 = AR(1).
<code>rho</code>	Numeric. AR(1) coefficient. Only used when <code>AR_index = 1</code> .
<code>batch_index</code>	Integer. Batch index used as part of the random seed (<code>set.seed(1000 * batch_index + i)</code> for replication <code>i</code>).
<code>seed</code>	Integer or NULL. Base random seed used internally for the D-optimal design search (<code>OptimalDesign::od_KL</code>). If NULL (default), no seed is set and results may vary between runs. When seed is supplied, results are fully reproducible because the internal search is stopped after a fixed number of restarts (<code>rest.max</code>), rather than after a fixed time budget.

Value

A list containing:

X Design matrix of dimension $n \times (p + 1)$.

Y.list List of replication response vectors, each of length `n`.

optimal_x Vector of selected support points.

optimal_weights Vector of observation counts at each support point.

Examples

```
# Example 1: quadratic model, D-optimal design
sim_data <- generate_simulation_data(
  p      = 2,
  n      = 20,
  e_sd   = 0.2,
  theta_true = c(-6, -3, 0.25),
  a      = -5,
  b      = 5,
  replication = 1,
  design_index = 2,
  center_index = 1
)

X      <- sim_data$X
Y.list <- sim_data$Y.list
```

```
# Example 2: cubic model, equally-spaced design
sim_data2 <- generate_simulation_data(
  p      = 3,
  n      = 20,
  e_sd   = 0.2,
  theta_true = c(1, 2, -1, 0.5),
  a      = -5,
```

```

    b          = 5,
    replication = 1,
    design_index = 1,
    center_index = 1
  )

```

L_SCB	<i>Evaluate lower band at x</i>
-------	---------------------------------

Description

Evaluate lower band at x

Usage

```
L_SCB(x, cov_theta, mu_star, lambda_best_optim)
```

Arguments

x	Numeric. Evaluation point.
cov_theta	Numeric matrix. Posterior covariance of theta.
mu_star	Numeric vector. Posterior mean of theta.
lambda_best_optim	Numeric. Critical constant lambda.

Value

Numeric. Lower band value at x.

sup_T_Bayes_ESCR	<i>To compute ESCR for BSCB and BPCB For BSCB use cov_mat = cov_theta; for BPCB use cov_mat = scale_mat.</i>
------------------	--

Description

To compute ESCR for BSCB and BPCB For BSCB use cov_mat = cov_theta; for BPCB use cov_mat = scale_mat.

Usage

```
sup_T_Bayes_ESCR(a, b, theta_true, mu_star, cov_mat)
```

Arguments

a	Numeric. Left endpoint.
b	Numeric. Right endpoint.
theta_true	Numeric vector. True regression coefficients.
mu_star	Numeric vector. Posterior mean of theta.
cov_mat	Numeric matrix. Covariance matrix.

Value

List with maximum and x_max.

sup_T_Bayes_PSCP	<i>To compute the critical constant for BSCB and BPCB; To compute PSCP for BSCB and BPCB</i>
------------------	--

Description

To compute the critical constant for BSCB and BPCB; To compute PSCP for BSCB and BPCB

Usage

sup_T_Bayes_PSCP(a, b, theta_hat, mu_star, cov_mat)

Arguments

a	Numeric. Left endpoint.
b	Numeric. Right endpoint.
theta_hat	Numeric vector. Posterior draw of theta.
mu_star	Numeric vector. Posterior mean of theta.
cov_mat	Numeric matrix. Covariance matrix.

Value

List with maximum and x_max.

sup_T_Freq_ESCR	<i>To compute the ESCR for FSCB and FPCB</i>
-----------------	--

Description

To compute the ESCR for FSCB and FPCB

Usage

```
sup_T_Freq_ESCR(a, b, theta_true, lm_theta_hat, cov_mat)
```

Arguments

a	Numeric. Left endpoint.
b	Numeric. Right endpoint.
theta_true	Numeric vector. True regression coefficients.
lm_theta_hat	Numeric vector. OLS estimate of theta.
cov_mat	Numeric matrix. Scaled covariance matrix ($S^2 \times (X^T X)^{-1}$).

Value

List with maximum and x_max.

sup_T_simFSCB	<i>To compute the critical constant for simFSCB</i>
---------------	---

Description

To compute the critical constant for simFSCB

Usage

```
sup_T_simFSCB(a, b, W_sample, cov_mat)
```

Arguments

a	Numeric. Left endpoint.
b	Numeric. Right endpoint.
W_sample	Numeric vector. Simulated draw.
cov_mat	Numeric matrix. Inverse of $X^T X$.

Value

List with maximum and x_max.

U_SCB	<i>Evaluate upper band at x</i>
-------	---------------------------------

Description

Evaluate upper band at x

Usage

U_SCB(x, cov_theta, mu_star, lambda_best_optim)

Arguments

- x Numeric. Evaluation point.
- cov_theta Numeric matrix. Posterior covariance of theta.
- mu_star Numeric vector. Posterior mean of theta.
- lambda_best_optim Numeric. Critical constant lambda.

Value

Numeric. Upper band value at x.

Index

compute_bpcb_ind_jeffreys, [2](#), [6](#)
compute_bscb_conjugate, [4](#), [5](#), [10](#), [14](#)
compute_bscb_hmc, [8](#)
compute_bscb_ind_jeffreys, [4](#), [6](#), [10](#), [10](#),
[14](#)
coverage_ESCR, [13](#), [14](#)
coverage_PSCP, [14](#)
create_order_form, [15](#)

f_L_SCB, [20](#)
f_U_SCB, [20](#)
find_global_maximum, [16](#)
find_global_maximum_h_all, [17](#)
fn_Bayes_ECR, [17](#)
fn_Bayes_PCP, [18](#)
fn_Freq_ECR, [18](#)
fn_neg_Bayes_ECR, [19](#)
fn_neg_Bayes_PCP, [19](#)

generate_simulation_data, [21](#)

L_SCB, [23](#)

sup_T_Bayes_ESCR, [23](#)
sup_T_Bayes_PSCP, [24](#)
sup_T_Freq_ESCR, [25](#)
sup_T_simFSCB, [25](#)

U_SCB, [26](#)