

Package ‘sffdr’

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Type Package

Title Surrogate Functional False Discovery Rates for Genome-Wide Association Studies

Version 1.1.0

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Description Pleiotropy-informed significance analysis of genome-wide association studies with surrogate functional false discovery rates (sffDR). The sffDR framework adapts the fFDR to leverage informative data from multiple sets of GWAS summary statistics to increase power in study while accommodating for linkage disequilibrium. sffDR provides estimates of key FDR quantities in a significance analysis such as the functional local FDR and q -value, and uses these estimates to derive a functional p -value for type I error rate control and a functional local Bayes' factor for post-GWAS analyses (e.g., fine mapping and colocalization).

URL <https://github.com/ajbass/sffdr>

License LGPL

Encoding UTF-8

VignetteBuilder knitr

LazyData true

Depends R ($\geq 4.1.0$)

Imports locfit, splines, ggplot2 ($\geq 3.5.1$), patchwork ($\geq 1.3.0$),
qvalue, fastglm, withr

Suggests testthat ($\geq 3.0.0$), knitr, rmarkdown

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NeedsCompilation no

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bmi	<i>Subset of p-values from the UK Biobank</i>
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Description

A dataset containing a subset of p-values from the UK Biobank.

Format

A data frame with 10,000 rows and 4 columns:

- bmi** Body mass index
- bfp** Body fat percentage
- cholesterol** Cholesterol
- triglycerides** Triglycerides

Examples

```
# Import data
data(bmi)

# Separate main p-values and informative p-values
p <- sumstats$bmi
z <- as.matrix(sumstats[, -1])
```

fpi0est

*Estimate Functional Proportion of Null Tests***Description**

Estimates the functional proportion of null tests (π_0) using a GLM approach with a constrained binomial family. This function fits models across multiple lambda thresholds and selects the optimal estimate via MISE minimization.

Usage

```
fpi0est(
  p,
  pi0_model_obj = NULL,
  z = NULL,
  pi0_model = NULL,
  indep_snps = NULL,
  lambda = seq(0.05, 0.95, 0.05),
  constrained.p = TRUE,
  tol = 1e-09,
  maxit = 200,
  verbose = TRUE,
  ...
)
```

Arguments

p	Numeric vector of p-values.
pi0_model_obj	Optional list object returned by pi0_model . Must contain fmod (formula) and zt (rank-transformed covariates). If provided, z and pi0_model are extracted automatically. Default is NULL.
z	Optional data frame or matrix of rank-transformed covariates. Required if pi0_model_obj is not provided. Default is NULL.
pi0_model	Optional formula (as character string or formula object). Required if pi0_model_obj is not provided. Default is NULL.
indep_snps	Optional logical vector indicating independent SNPs for model fitting. Default is NULL (all SNPs used).
lambda	Numeric vector of lambda thresholds. Default is seq(0.05, 0.95, 0.05).
constrained.p	Logical; use constrained binomial family. Default is TRUE.
tol	Numeric; convergence tolerance. Default is 1e-9.
maxit	Integer; maximum iterations. Default is 200.
verbose	Logical; print progress messages. Default is TRUE.
...	Additional arguments passed to fastglm .

Details

Algorithm:

1. For each lambda threshold, fit a binomial GLM: $P(p \geq \lambda|z)$
2. Use constrained binomial family to ensure predictions in (0, 1)
3. Select optimal lambda via MISE minimization

Usage Patterns:

Pattern 1 (Recommended): Use output from `pi0_model`

```
mpi0 <- pi0_model(z)
# Clean syntax:
fpi0_out <- fpi0est(p, mpi0)
```

Pattern 2: Manually specify formula and covariates

```
fpi0_out <- fpi0est(p, z = z_matrix, pi0_model = formula_obj)
```

Value

An object of class `fpi0` (a list) containing:

fpi0 Numeric vector of functional pi0 estimates for each test.

tableLambda A data frame summarizing results for each lambda value.

MISE The Mean Integrated Squared Error (MISE) for the chosen model.

lambda The selected optimal lambda value.

Examples

```
# Import data
data(bmi)

# Separate main p-values and conditioning p-values
p <- sumstats$bmi
z <- as.matrix(sumstats[, -1])

# Apply pi0_model to create model (uses adaptive knot selection)
fmod <- pi0_model(z)

# Estimate functional pi0
fpi0_out <- fpi0est(p, fmod)
fpi0 <- fpi0_out$fpi0

# Apply sffdr
sffdr_out <- sffdr(p, fpi0)
```

fpvalues	<i>Functional p-values</i>
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Description

Calculate functional p-values from functional local FDRs.

Usage

```
fpvalues(lfdr, p = NULL)
```

Arguments

lfdr	A vector of functional local FDRs.
p	A vector of p-values for ranking purposes. Default is NULL.

Value

A list containing:

fp	Functional p-values.
fq	Functional q-values.

kernelEstimator	<i>Kernel Density Estimation for GWAS P-values</i>
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Description

Performs kernel density estimation on p-values (univariate) or joint p-value/covariate pairs (bivariate) using local regression on the probit scale. The estimator is optimized for GWAS data with linkage disequilibrium (LD) and uses adaptive downsampling to prioritize signal-rich regions while maintaining computational efficiency.

Usage

```
kernelEstimator(
  x,
  eval.points = x,
  epsilon = 1e-15,
  epsilon.max = 1 - 1e-04,
  maxk = 5e+05,
  maxit = 200,
  target_null = 1e+05,
  trim = 0,
  nn = NULL,
  tail_threshold = -2,
  ...
)
```

Arguments

<code>x</code>	Numeric vector of p-values (for 1D density) or a 2-column matrix where the first column contains p-values and the second column contains an informative covariate/surrogate. All p-values must be in (0, 1) and the covariate/surrogate must be rank-transformed to be (0,1].
<code>eval.points</code>	Points at which to evaluate the density estimate. Defaults to <code>x</code> . For custom evaluation points, must match the dimensionality of <code>x</code> (vector or 2-column matrix).
<code>epsilon</code>	Lower bound for p-values to prevent numerical issues. P-values below this are clamped to <code>epsilon</code> . Default: $1e-15$.
<code>epsilon.max</code>	Upper bound for p-values. P-values above this are clamped to <code>epsilon.max</code> . Default: $1 - 1e-4$.
<code>maxk</code>	Maximum number of fitting points passed to <code>locfit</code> . Increase for very large datasets. Default: 500000.
<code>maxit</code>	Maximum number of iterations for local regression fitting. Default: 2000.
<code>target_null</code>	Maximum number of null SNPs to include in the weighted fit (bivariate case only). SNPs in the signal-enriched tail (defined by <code>tail_threshold</code>) are always retained; null SNPs are downsampled to this target and upweighted accordingly. Default: 100000.
<code>trim</code>	For 1D estimation, fixes the density estimate to constant values on the intervals $(0, \text{trim})$ and $(1 - \text{trim}, 1)$ to reduce boundary variance. Default: 0 (no trimming).
<code>nn</code>	Nearest-neighbor bandwidth parameter for <code>locfit</code> , expressed as a fraction of the data. If NULL (default), automatically determined based on effective sample size to span approximately 5000 neighbors (which corresponds to multiple LD blocks in GWAS). Larger values increase smoothing.
<code>tail_threshold</code>	Z-score threshold on the probit-transformed covariate scale (bivariate case only). SNPs with $z < \text{tail_threshold}$ are treated as signal-enriched and prioritized in the adaptive fit. Default: -2 (approximately 2.3% of standard normal distribution).
<code>...</code>	Additional arguments passed to <code>lp</code> for controlling local polynomial fitting (e.g., <code>deg</code> , <code>kern</code>).

Details

The function implements a multi-stage density estimation procedure:

1. **Probit transformation:** P-values are transformed to the normal scale via `qnorm` to stabilize variance and handle extreme values.
2. **Adaptive downsampling (bivariate only):** To handle large GWAS datasets efficiently, the null region (where the covariate suggests low signal) is downsampled to `target_null` SNPs, with inverse-probability weighting to preserve the density. Signal-enriched SNPs (tail) are always retained.
3. **Cascade fitting:** Multiple fitting strategies are attempted in sequence, with decreasing resolution and increasing robustness, until a valid fit is obtained.

4. **Jacobian correction:** Density estimates are transformed back to the original p-value scale using the Jacobian of the probit transformation.

The nearest-neighbor bandwidth `nn` controls smoothing and LD robustness. By targeting ~5000 neighbors (default), the estimator naturally averages over multiple LD blocks (~30-50 blocks in European ancestry populations), reducing spurious local structure while preserving true signal-covariate relationships.

Value

A data frame with columns:

- x** Evaluation points (original scale).
- fx** Estimated density at evaluation points (original scale).
- s** Evaluation points on probit scale (`qnorm(x)`).
- fs** Estimated density on probit scale.

The returned object has an attribute `"lfit"` containing the fitted `locfit` object for diagnostics.

See Also

[locfit](#), [sffdr](#)

Examples

```
## Not run:
# 1D density estimation
p <- runif(10000, 0, 1)
dens <- kernelEstimator(p)
plot(dens$x, dens$fx, type = "l")

# 2D density with informative covariate
p <- runif(10000, 0, 1)
z <- runif(10000) # rank-norm transformed covariate
x_mat <- cbind(p, z)
dens <- kernelEstimator(x_mat)

## End(Not run)
```

Description

Generates a natural spline model for the functional proportion of null tests (*fpi0*) by adaptively selecting knots based on an FDR threshold.

Usage

```
pi0_model(
  z,
  indep_snps = NULL,
  fdr_threshold = 0.25,
  min_discoveries = 2500,
  min_snps_per_knot = 100,
  n_knots = 5,
  verbose = TRUE
)
```

Arguments

<code>z</code>	Matrix/data.frame of p-values (rows=tests, cols=traits).
<code>indep_snps</code>	Logical vector indicating independent SNPs (training subset). If NULL, uses all tests. Default: NULL.
<code>fdr_threshold</code>	FDR threshold for signal definition. Default: 0.25.
<code>min_discoveries</code>	Min significant hits required to include trait. Default: 2500.
<code>min_snps_per_knot</code>	Min significant SNPs per knot interval. Default: 100. Note that this is the minimum independent SNPs required per knot interval if 'indep_snps' is provided.
<code>n_knots</code>	Target knot count. Default: 5. Automatically reduced if insufficient discoveries (via 'min_snps_per_knot') or capped by 'max_knots'.
<code>verbose</code>	Print selection details. Default: TRUE.

Details

Independent SNPs determine knot placement; all SNPs train the model to capture signal shape. For smaller datasets (<100K), consider reducing 'min_snps_per_knot' and 'min_discoveries' to improve knot placement.

Value

List containing:

<code>fmod</code>	Model formula (using <code>splines::ns</code>)
<code>zt</code>	Data frame of globally rank-transformed p-values

plot.sffdr	<i>Plotting function for sffdr object</i>
------------	---

Description

Graphical display of the sffdr object

Usage

```
## S3 method for class 'sffdr'  
plot(x, rng = c(0, 5e-08), ...)
```

Arguments

x	A sffdr object.
rng	Significance region to show. Optional.
...	Additional arguments. Currently unused.

Value

Nothing of interest.

Author(s)

Andrew J. Bass

See Also

[sffdr](#)

Examples

```
# import data  
data(bmi)  
  
# Separate main p-values and conditioning p-values  
p <- sumstats$bmi  
z <- as.matrix(sumstats[, -1])  
  
# Apply pi0_model to create model  
fmod <- pi0_model(z)  
  
# Estimate functional pi0  
fpi0_out <- fpi0est(p, z = fmod$zt, pi0_model = fmod$fmod)  
fpi0 <- fpi0_out$fpi0  
  
# Apply sffdr  
sffdr_out <- sffdr(p, fpi0)
```

```
# Plot significance results
plot(sffdr_out, rng = c(0, 5e-4))
```

sffdr

Surrogate Functional False Discovery Rate Analysis

Description

Estimate functional p-values, q-values, and local false discovery rates (lfdr) for GWAS data leveraging summary statistics from related traits. Functional p-values map from the functional q-value (FDR-based measure) to a p-value for type I error rate control, accounting for pleiotropy that impacts the prior probability of association.

Usage

```
sffdr(
  p.value,
  fpi0,
  surrogate = NULL,
  epsilon = .Machine$double.xmin,
  nn = NULL,
  fp_ties = TRUE,
  seed = 2026,
  verbose = TRUE,
  ...
)
```

Arguments

p.value	Numeric vector of p-values to analyze.
fpi0	Numeric vector of functional pi0 estimates, obtained from fpi0est . Values must be in [0, 1].
surrogate	Optional numeric vector (same length as p.value) used as a surrogate variable for compression. If NULL (default), uses fpi0 as the surrogate.
epsilon	Numeric; lower bound for p-value clamping during density estimation. Default is .Machine\$double.xmin.
nn	Numeric; nearest-neighbor bandwidth for kernelEstimator . If NULL (default), automatically selected as ~5000 neighbors.
fp_ties	Logical; whether to break ties in functional p-values using the original p-value ordering. Default is TRUE.
seed	Integer; random seed for reproducibility of rank tie-breaking. Default is 2026.
verbose	Logical; print progress messages. Default is TRUE.
...	Additional arguments passed to kernelEstimator .

Details

The surrogate functional FDR (sfFDR) methodology extends the functional FDR framework to leverage multiple informative variables (e.g., functional annotations, GWAS summary statistics) for increased power while controlling false discovery rates.

Workflow:

1. Estimate functional π_0 (proportion of nulls) using `fpi0est`
2. Call `sffdr()` with p-values and estimated functional π_0
3. Use returned functional p-values/q-values/local FDRs for significance testing

Surrogate Variable: If not specified, the estimated functional π_0 is used as the surrogate variable.

Value

An S3 object of class "sffdr" containing:

<code>call</code>	The function call.
<code>pvalues</code>	Original p-values.
<code>fpvalues</code>	Functional p-values.
<code>fqvalues</code>	Functional q-values.
<code>flfdr</code>	Functional local false discovery rates.
<code>fpi0</code>	Functional π_0 estimates.
<code>fx</code>	Joint density estimates at observed (p-value, surrogate) pairs.

Author(s)

Andrew J. Bass

See Also

`fpi0est`, `plot.sffdr`, `kernelEstimator`

Examples

```
# Import data
data(bmi)

# Separate main p-values and conditioning p-values
p <- sumstats$bmi
z <- as.matrix(sumstats[, -1])

# Apply pi0_model to create model
# (note: use indep_snps argument to specify independent SNPs for training)
fmod <- pi0_model(z)

# Estimate functional pi0
# (note: use indep_snps argument to specify independent SNPs for training)
fpi0_out <- fpi0est(p, fmod)
```

```
fpi0 <- fpi0_out$fpi0  
  
# Apply sffdr  
sffdr_out <- sffdr(p, fpi0)  
  
# Plot significance results  
plot(sffdr_out)  
  
# Extract functional quantities  
fp <- sffdr_out$fpvalues  
fq <- sffdr_out$fqvalues  
flfdr <- sffdr_out$flfdr
```

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