

Package ‘OSFD’

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Title Output Space-Filling Design

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Description

Methods to generate a design in the input space that sequentially fills the output space of a black-box function. The output space-filling designs are helpful in inverse design or feature-based modeling problems.

See Wang, Shangkun, Adam P. Generale, Surya R. Kalidindi, and V. Roshan Joseph. (2024), Sequential designs for filling output spaces, *Technometrics*, 66, 65–76. for details. This work is supported by U.S. National Foundation grant CMMI-1921646.

License GPL (≥ 2)

Depends Rcpp ($\geq 1.0.8$), lhs, twinning, dplyr

LinkingTo Rcpp, RcppArmadillo, twinning

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OSFD-package

*Sequential algorithms to generate designs that fill the output space.***Description**

Sequential algorithms to generate a design that produces points filling the output space. The underlying mapping f from input space to output space is assumed to be a black-box function that can be evaluated in the forward direction. Please see Wang et al. (2024) and Wang & Joseph (2025) for details.

Author(s)

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References

Wang, Shangkun, Adam P. Generale, Surya R. Kalidindi, and V. Roshan Joseph. (2024), "Sequential designs for filling output spaces", *Technometrics*, 66, 65–76.

Wang, Shangkun, and V. Roshan Joseph. (2025), "Comment: A Model-free Method for Input-Output Space-Filling Design." *Technometrics*, to appear.

ball_unif

*(Quasi) uniform points in a p-dimensional ball***Description**

ball_unif generates random or quasi-random uniform points in a p-dimensional ball.

Usage

```
ball_unif(cen, rad, n, rand = TRUE)
```

Arguments

| | |
|------|---|
| cen | a vector specifying the center of the ball. |
| rad | radius of the ball. |
| n | number of points. |
| rand | whether to generate random or quasi random points. The default value is TRUE. |

Details

ball_unif generates random uniform points or quasi uniform points by twinning algorithm in a p-dimensional ball.

Value

a matrix of the generated points.

References

Vakayil, Akhil, and V. Roshan Joseph. (2022). "Data twinning". *Statistical Analysis and Data Mining: The ASA Data Science Journal*, 15(5), 598-610.

Wang, Shangkun, Adam P. Generale, Surya R. Kalidindi, and V. Roshan Joseph. (2024), "Sequential designs for filling output spaces", *Technometrics*, 66, 65–76.

Examples

```
x = ball_unif(c(0,0), 1, 10, rand=FALSE)
plot(x, type='p')
```

IOSFD

Input-Output Space-Filling Design

Description

This function is for producing designs that explicitly balance the input and output points.

Usage

```
IOSFD(  
  D = NULL,  
  f,  
  p,  
  q,  
  lambda = 0.5,  
  n_ini = NA,  
  n,  
  scale = TRUE,  
  CAND = NULL,  
  rand_out = FALSE,  
  rand_in = FALSE  
)
```

Arguments

| | |
|---|--|
| D | a matrix of the initial design. If not specified, a random Latin hypercube design of size n_ini and dimension p will be generated as initial design. |
| f | black-box function. |
| p | input dimension. |
| q | output dimension. |

| | |
|----------|--|
| lambda | the weight for the input space. Its value should be within [0, 1]. The default value is 0.5. When lambda=0, please directly use OSFD. |
| n_ini | the size of initial design. This initial size must be specified if D is not provided. |
| n | the size of the final design. |
| scale | whether to scale the output points to 0 to 1 for each dimension. |
| CAND | the candidate points in the input space. If Null, it will be automatically generated. |
| rand_out | whether to use random uniform points or quasi random points by twinning algorithm for generating points in spheres for output space approximation. The default value is FALSE. |
| rand_in | whether to use random uniform points or quasi random points by twinning algorithm for generating points in spheres for input space candidate sets. The default value is FALSE. |

Details

IOSFD produces a design that balances the input and output points by Wang et al. (2025).

Value

| | |
|---|--|
| D | the final design points in the input space |
| Y | the output points |

References

Wang, Shangkun, and V. Roshan Joseph. (2025), "Comment: A Model-free Method for Input-Output Space-Filling Design." *Technometrics*, to appear.

Examples

```
# test function: inverse-radius function (Wang et.al 2023)
inverse_r = function(x){
  epsilon = 0.1
  y1 = 1 / (x[1]^2 + x[2]^2 + epsilon^2) ^ (1/2)
  if (x[2]==0){
    y2 = 0
  }else if (x[1]==0) {
    y2 = pi / 2}else{
    y2 = atan(x[2] / x[1])
  }
  return (c(y1=y1, y2=y2))
}

set.seed(2022)
p = 2
q = 2
f = inverse_r
n_ini = 10
n = 50
```

```
iosfd = IOSFD(f=f, p=p, q=q, n_ini=n_ini, n=n)
D = iosfd$D
Y = iosfd$Y
```

| | |
|--------|-------------------------|
| mMdist | <i>Minimax distance</i> |
|--------|-------------------------|

Description

mMdist computes the minimax distance of a design in a specified region. A large uniform sample from the specified region is needed to compute the minimax distance.

Usage

```
mMdist(X, X_space)
```

Arguments

| | |
|---------|--|
| X | a matrix specifying the design. |
| X_space | a large sample of uniform points in the space of interest. |

Details

mMdist approximates the minimax distance of a set of points X by the large sample X_space in the space of interest.

Value

the minimax distance.

References

Johnson, Mark E., Leslie M. Moore, and Donald Ylvisaker. (1990), "Minimax and Maximin Distance Designs", *Journal of Statistical Planning and Inference*, 26, 131–148.

Wang, Shangkun, Adam P. Generale, Surya R. Kalidindi, and V. Roshan Joseph. (2024), "Sequential designs for filling output spaces", *Technometrics*, 66, 65–76.

Examples

```
# the minimax distance of a random Latin hypercube design
D = randomLHS(5, 2)
mMdist(D, replicate(2, runif(1e5)))
```

Description

This function is for producing designs that fill the output space.

Usage

```
OSFD(
  D = NULL,
  f,
  p,
  q,
  n_ini = NA,
  n,
  scale = TRUE,
  method = "EI",
  CAND = NULL,
  rand_out = FALSE,
  rand_in = FALSE
)
```

Arguments

| | |
|----------|--|
| D | a matrix of the initial design. If not specified, a random Latin hypercube design of size n_ini and dimension p will be generated as initial design. |
| f | black-box function. |
| p | input dimension. |
| q | output dimension. |
| n_ini | the size of initial design. This initial size must be specified if D is not provided. |
| n | the size of the final design. |
| scale | whether to scale the output points to 0 to 1 for each dimension. |
| method | two choices: 'EI' or 'Greedy'; the default is 'EI'. |
| CAND | the candidate points in the input space. If Null, it will be automatically generated. |
| rand_out | whether to use random uniform points or quasi random points by twinning algorithm for generating points in spheres for output space approximation. The default value is FALSE. |
| rand_in | whether to use random uniform points or quasi random points by twinning algorithm for generating points in spheres for input space candidate sets. The default value is FALSE. |

Details

OSFD produces a design that fills the output space using the sequential algorithm by Wang et al. (2024).

Value

D the final design points in the input space
Y the output points

References

Wang, Shangkun, Adam P. Generale, Surya R. Kalidindi, and V. Roshan Joseph. (2024), "Sequential designs for filling output spaces", *Technometrics*, 66, 65–76.

Examples

```
# test function: inverse-radius function (Wang et.al 2024)
inverse_r = function(x){
  epsilon = 0.1
  y1 = 1 / (x[1]^2 + x[2]^2 + epsilon^2) ^ (1/2)
  if (x[2]==0){
    y2 = 0
  }else if (x[1]==0) {
    y2 = pi / 2}
  }else{
    y2 = atan(x[2] / x[1])
  }
  return (c(y1=y1, y2=y2))
}

set.seed(2022)
p = 2
q = 2
f = inverse_r
n_ini = 10
n = 50
osfd = OSFD(f=f, p=p, q=q, n_ini=n_ini, n=n)
D = osfd$D
Y = osfd$Y
```

spanfill

Generate points to approximate the space spanned by the existing points

Description

spanfill generates points to approximate a space based on existing points. These approximate points can be used to find local fill distance in the space or be used as candidate points in active learning.

Usage

```
spanfill(X, bound = FALSE)
```

Arguments

| | |
|-------|--|
| X | a matrix specifying the existing points |
| bound | a binary variable indicating whether to bound the generated points to 0 to 1 in each dimension. If bound=TRUE, all the generated points will be projected to the unit hypercube. The default value is FALSE. |

Details

spanfill generates points to approximate the space spanned by the existing points. Details can be found in Wang et al. (2024).

Value

a matrix of the generated points to approximate the space.

References

Wang, Shangkun, Adam P. Generale, Surya R. Kalidindi, and V. Roshan Joseph. (2024). "Sequential designs for filling output spaces", *Technometrics*, 66, 65–76.

Examples

```
X = matrix(runif(20), ncol=2)
spanfill_points = spanfill(X)
plot(spanfill_points, type='p')
```


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