

# Package ‘GreedyExperimentalDesign’

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**Type** Package

**Title** Greedy Experimental Design Construction

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**Description** Computes experimental designs for a two-arm experiment with covariates via a number of methods.

- (0) complete randomization and randomization with forced-balance.
- (1) Greedily optimizing a balance objective function via pairwise switching. This optimization provides lower variance for the treatment effect estimator (and higher power) while preserving a design that is close to complete randomization. We return all iterations of the designs for use in a permutation test.
- (2) The second is via numerical optimization (via 'gurobi' which must be installed, see <[https://www.gurobi.com/documentation/9.1/quickstart\\_windows/r\\_ins\\_the\\_r\\_package.html](https://www.gurobi.com/documentation/9.1/quickstart_windows/r_ins_the_r_package.html)>) a la Bertsimas and Kallus.
- (3) rerandomization,
- (4) Karp's method for one covariate,
- (5) exhaustive enumeration to find the optimal solution (only for small sample sizes)
- (6) Binary pair matching using the 'nbpMatching' library
- (7) Binary pair matching plus (1) to further optimize balance
- (8) Binary pair matching plus (3) to further optimize balance
- (9) Hadamard designs

We also allow for three objective functions:  
Mahalanobis distance,  
Sum of absolute differences standardized and  
Kernel distances via the 'kernlab' library.

**License** GPL-3

**Depends** R (>= 3.2.0), rJava (>= 0.9-6), GreedyExperimentalDesignJARs (>= 1.0)

**SystemRequirements** Java (>= 7.0)

**LinkingTo** Rcpp

**Imports** Rcpp, checkmate, nbpMatching, survey, kernlab, graphics,  
grDevices, stats

**RoxygenNote** 7.1.0

**NeedsCompilation** yes

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automobile *Data concerning automobile prices.*

**Description**

The automobile data frame has 201 rows and 25 columns and concerns automobiles in the 1985 Auto Imports Database. The response variable, price, is the log selling price of the automobile. There are 7 categorical predictors and 17 continuous / integer predictors which are features of the automobiles. 41 automobiles have missing data in one or more of the feature entries. This dataset is true to the original except with a few of the predictors dropped.

**Usage**

`data(automobile)`

**Source**

K Bache and M Lichman. UCI machine learning repository, 2013. <http://archive.ics.uci.edu/ml/datasets/Automobile>

binaryMatchExperimentalDesignSearch  
*Begin a Search for Binary Matching Designs*

**Description**

This method creates an object of type `binary_experimental_design` and will find pairs. You can then use the function `resultsBinaryMatchSearch` to create randomized allocation vectors. For one column in X, we just sort to find the pairs trivially.

**Usage**

`binaryMatchExperimentalDesignSearch(X, compute_dist_matrix = NULL)`

**Arguments**

- `X` The design matrix with  $n$  rows (one for each subject) and  $p$  columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design.
- `compute_dist_matrix` The function that computes the distance matrix between every two observations in  $X$ , its only argument. The default is `NULL` signifying euclidean squared distance optimized in C++.

**Value**

An object of type `binary_experimental_design` which can be further operated upon.

**Author(s)**

Adam Kapelner

---

`binaryMatchFollowedByGreedyExperimentalDesignSearch`

*Begin a Search for Binary Matching Followed by Greedy Switch Designs*

---

**Description**

This method creates an object of type `binary_then_greedy_experimental_design` and will find optimal matched pairs which are then greedily switched in order to further minimize a balance metric. You can then use the function `resultsBinaryMatchThenGreedySearch` to obtain the randomized allocation vectors. For one column in  $X$ , the matching just sorts the values to find the pairs trivially.

**Usage**

```
binaryMatchFollowedByGreedyExperimentalDesignSearch(
  X,
  compute_dist_matrix = NULL,
  ...
)
```

**Arguments**

- `X` The design matrix with  $n$  rows (one for each subject) and  $p$  columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design.
- `compute_dist_matrix` The function that computes the distance matrix between every two observations in  $X$ , its only argument. The default is `NULL` signifying euclidean squared distance optimized in C++.
- `...` Arguments passed to `initGreedyExperimentalDesignObject`. It is recommended to set `max_designs` otherwise it will default to 10,000.

**Value**

An object of type `binary_experimental_design` which can be further operated upon.

**Author(s)**

Adam Kapelner

---

`binaryMatchFollowedByRerandomizationDesignSearch`

*Begin a Search for Binary Matching Followed by Rerandomization*

---

**Description**

This method creates an object of type `binary_then_rerandomization_experimental_design` and will find optimal matched pairs which are then rerandomized in order to further minimize a balance metric. You can then use the function `resultsBinaryMatchThenRerandomizationSearch` to obtain the randomized allocation vectors. For one column in `X`, the matching just sorts the values to find the pairs trivially.

**Usage**

```
binaryMatchFollowedByRerandomizationDesignSearch(
  X,
  compute_dist_matrix = NULL,
  ...
)
```

**Arguments**

<code>X</code>	The design matrix with <code>\$n</code> rows (one for each subject) and <code>\$p</code> columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design.
<code>compute_dist_matrix</code>	The function that computes the distance matrix between every two observations in <code>X</code> , its only argument. The default is <code>NULL</code> signifying euclidean squared distance optimized in C++.
<code>...</code>	Arguments passed to <code>initGreedyExperimentalDesignObject</code> . It is recommended to set <code>max_designs</code> otherwise it will default to 10,000.

**Value**

An object of type `binary_experimental_design` which can be further operated upon.

**Author(s)**

Adam Kapelner

complete\_randomization

*Implements complete randomization (without forced balance)*

---

### **Description**

Implements complete randomization (without forced balance)

### **Usage**

```
complete_randomization(n, r, form = "one_zero")
```

### **Arguments**

n	number of observations
r	number of randomized designs you would like
form	Which form should it be in? The default is one_zero for 1/0's or pos_one_min_one for +1/-1's.

### **Value**

a matrix where each column is one of the r designs

### **Author(s)**

Adam Kapelner

---

complete\_randomization\_with\_forced\_balanced

*Implements forced balanced randomization*

---

### **Description**

Implements forced balanced randomization

### **Usage**

```
complete_randomization_with_forced_balanced(n, r, form = "one_zero")
```

### **Arguments**

n	number of observations
r	number of randomized designs you would like
form	Which form should it be in? The default is one_zero for 1/0's or pos_one_min_one for +1/-1's.

**Value**

a matrix where each column is one of the  $r$  designs

**Author(s)**

Adam Kapelner

---

compute\_gram\_matrix     *Gram Matrix Computation*

---

**Description**

Computes the Gram Matrix for a user-specified kernel using the library kernlab. Note that this function automatically standardizes the columns of the data entered.

**Usage**

```
compute_gram_matrix(X, kernel_type, params = c())
```

**Arguments**

X	The design matrix with $n$ rows (one for each subject) and $p$ columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design.
kernel_type	One of the following: "vanilla", "rbf", "poly", "tanh", "bessel", "laplace", "anova" or "spline".
params	A vector of numeric parameters. Each kernel_type has different numbers of parameters required. For more information see documentation for the kernlab library.

**Value**

The  $n \times n$  gram matrix for the given kernel on the given data.

**Author(s)**

Adam Kapelner

---

compute\_objective\_val *Computes Objective Value From Allocation Vector*

---

### Description

Returns the objective value given a design vector as well as an objective function. This is sometimes duplicated in Java. However, within Java, tricks are played to make optimization go faster so Java's objective values may not always be the same as the true objective function (e.g. logs or constants dropped).

### Usage

```
compute_objective_val(X, indic_T, objective = "abs_sum_diff", inv_cov_X = NULL)
```

### Arguments

X	The n x p design matrix
indic_T	The n-length binary allocation vector
objective	The objective function to use. Default is abs_sum_diff and the other option is mahal_dist.
inv_cov_X	Optional: the inverse sample variance covariance matrix. Use this argument if you will be doing many calculations since passing this in will cache this data.

### Author(s)

Adam Kapelner

---

compute\_randomization\_metrics  
*Computes Randomization Metrics (explained in paper) about a design algorithm*

---

### Description

Computes Randomization Metrics (explained in paper) about a design algorithm

### Usage

```
compute_randomization_metrics(designs)
```

### Arguments

designs	A matrix where each column is one design.
---------	---



**Value**

A list of resulting data: the probability estimates for each pair in the design of randomness where estimates close to ~0.5 represent random assignment, then the entropy metric the distance metric, the maximum eigenvalue of the allocation var-cov matrix (operator norm) and the squared Frobenius norm (the sum of the squared eigenvalues)

**Author(s)**

Adam Kapelner

---

generate\_stdzied\_design\_matrix

*Generates a design matrix with standardized predictors.*

---

**Description**

This function is useful for debugging.

**Usage**

```
generate_stdzied_design_matrix(n = 50, p = 1, covariate_gen = rnorm, ...)
```

**Arguments**

n	Number of rows in the design matrix
p	Number of columns in the design matrix
covariate_gen	The function to use to draw the covariate realizations (assumed to be iid). This defaults to rnorm for $N(0,1)$ draws.
...	Optional arguments to be passed to the covariate_dist function.

**Value**

The design matrix

**Author(s)**

Adam Kapelner

---

 GreedyExperimentalDesign

*Greedy Experimental Design Search*


---

**Description**

A tool to find a priori experimental designs with good balance greedily

**Author(s)**

Adam Kapelner <kapelner@qc.cuny.edu>

**References**

Kapelner, A

---

greedy\_orthogonalization\_curation

*Curate More Orthogonal Vectors Greedily*


---

**Description**

This function takes a set of allocation vectors and pares them down one-by-one by eliminating the vector that can result in the largest reduction in  $\text{Avg}[|r_{ij}|]$ . It is recommended to begin with a set of unmirrored vectors for speed. Then add the mirrors later for whichever subset you wish.

**Usage**

```
greedy_orthogonalization_curation(W, Rmin = 2, verbose = FALSE)
```

**Arguments**

W	A matrix in $\{-1, 1\}^R \times n$ which have R allocation vectors for an experiment of sample size n.
Rmin	The minimum number of vectors to consider in a design. The default is the true bottom, two.
verbose	Default is FALSE but if not, it will print out a message for each iteration.

**Value**

A list with two elements: (1) `avg_abs_r_ij_by_R` which is a data frame with  $R - Rmin + 1$  rows and columns R and average absolute  $r_{ij}$  and (2) `Wsorted` which provides the collection of vectors in sorted by best average absolute  $r_{ij}$  in row order from best to worst.

**Author(s)**

Adam Kapelner

---

 greedy\_orthogonalization\_curation2

*Curate More Orthogonal Vectors Greedily*


---

### Description

This function takes a set of allocation vectors and pares them down one-by-one by eliminating the vector that can result in the largest reduction in  $\text{Avg}[|r_{ij}|]$ . It is recommended to begin with a set of unmirrored vectors for speed. Then add the mirrors later for whichever subset you wish.

### Usage

```
greedy_orthogonalization_curation2(W, R0 = 100, verbose = FALSE)
```

### Arguments

W	A matrix in $\{-1, 1\}^R \times n$ which have R allocation vectors for an experiment of sample size n.
R0	The minimum number of vectors to consider in a design. The default is the true bottom, two.
verbose	Default is FALSE but if not, it will print out a message for each iteration.

### Value

A list with two elements: (1) `avg_abs_r_ij_by_R` which is a data frame with  $R - R_{\min} + 1$  rows and columns R and average absolute  $r_{ij}$  and (2) `Wsorted` which provides the collection of vectors in sorted by best average absolute  $r_{ij}$  in row order from best to worst.

### Author(s)

Adam Kapelner

---

 hadamardExperimentalDesign

*Create a Hadamard Design*


---

### Description

This method returns unique designs according to a Hadamard matrix

### Usage

```
hadamardExperimentalDesign(X, strict = TRUE, form = "zero_one")
```

**Arguments**

X	The design matrix with $n$ rows (one for each subject) and $p$ columns (one for each measurement on the subject). The measurements aren't used to compute the Hadamard designs, only the number of rows.
strict	Hadamard matrices are not available for all $n$ .
form	Which form should it be in? The default is one_zero for 1/0's or pos_one_min_one for +1/-1's.

**Value**

An matrix of dimension  $R \times n$  where  $R$  is the number of Hadamard allocations.

**Author(s)**

Adam Kapelner

---

initGreedyExperimentalDesignObject

*Begin A Greedy Pair Switching Search*

---

**Description**

This method creates an object of type `greedy_experimental_design` and will immediately initiate a search through  $1_T$  space for forced balance designs.

**Usage**

```
initGreedyExperimentalDesignObject(
  X = NULL,
  max_designs = 10000,
  objective = "mahal_dist",
  Kgram = NULL,
  wait = FALSE,
  start = TRUE,
  max_iters = Inf,
  semigreedy = FALSE,
  diagnostics = FALSE,
  num_cores = 1
)
```

**Arguments**

X	The design matrix with $n$ rows (one for each subject) and $p$ columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design. This parameter must be specified unless you choose objective type "kernel" in which case, the Kgram parameter must be specified.
---	--

max_designs	The maximum number of designs to be returned. Default is 10,000. Make this large so you can search however long you wish as the search can be stopped at any time by using the <code>stopSearch</code> method
objective	The objective function to use when searching design space. This is a string with valid values "mahal_dist" (the default), "abs_sum_diff" or "kernel".
Kgram	If the objective = kernel, this argument is required to be an $n \times n$ matrix whose entries are the evaluation of the kernel function between subject $i$ and subject $j$ . Default is NULL.
wait	Should the R terminal hang until all max_designs vectors are found? The default is FALSE.
start	Should we start searching immediately (default is TRUE).
max_iters	Should we impose a maximum number of greedy switches? The default is Inf which a flag for "no limit."
semigreedy	Should we use a fully greedy approach or the quicker semi-greedy approach? The default is FALSE corresponding to the fully greedy approach.
diagnostics	Returns diagnostic information about the iterations including (a) the initial starting vectors, the switches at every iteration and information about the objective function at every iteration (default is FALSE due to speed concerns).
num_cores	The number of CPU cores you wish to use during the search. The default is 1.

**Value**

An object of type `greedy_experimental_design_search` which can be further operated upon

**Author(s)**

Adam Kapelner

**Examples**

```
## Not run:
library(MASS)
data(Boston)
#pretend the Boston data was an experiment setting
#first pull out the covariates
X = Boston[, 1 : 13]
#begin the greedy design search
ged = initGreedyExperimentalDesignObject(X,
max_designs = 1000, num_cores = 3, objective = "abs_sum_diff")
#wait
ged

## End(Not run)
```

---

```
initKarpExperimentalDesignObject  
    Begin Karp Search
```

---

### Description

This method creates an object of type `karp_experimental_design` and will immediately initiate a search through  $\$1\_T\$$  space. Note that the Karp search only works for one covariate (i.e.  $\$p=1\$$ ) and the objective "`abs_sum_diff`".

### Usage

```
initKarpExperimentalDesignObject(  
  X,  
  wait = FALSE,  
  balanced = TRUE,  
  start = TRUE  
)
```

### Arguments

<code>X</code>	The design matrix with $\$n\$$ rows (one for each subject) and $\$p\$$ columns (one for each measurement on the subject). This is the design matrix you wish to search for a more karp design.
<code>wait</code>	Should the R terminal hang until all <code>max_designs</code> vectors are found? The default is FALSE.
<code>balanced</code>	Should the final vector be balanced? Default and recommended is TRUE.
<code>start</code>	Should we start searching immediately (default is TRUE).

### Value

An object of type `karp_experimental_design_search` which can be further operated upon

### Author(s)

Adam Kapelner

---

```
initOptimalExperimentalDesignObject
```

*Begin a Search for the Optimal Solution*

---

### Description

This method creates an object of type `optimal_experimental_design` and will immediately initiate a search through  $S_{1-T}$  space. Since this search takes exponential time, for most machines, this method is futile beyond 28 samples. You've been warned!

### Usage

```
initOptimalExperimentalDesignObject(
  X = NULL,
  objective = "mahal_dist",
  Kgram = NULL,
  wait = FALSE,
  start = TRUE,
  num_cores = 1
)
```

### Arguments

<code>X</code>	The design matrix with $n$ rows (one for each subject) and $p$ columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design.
<code>objective</code>	The objective function to use when searching design space. This is a string with valid values "mahal_dist" (the default), "abs_sum_diff" or "kernel".
<code>Kgram</code>	If the <code>objective = kernel</code> , this argument is required to be an $n \times n$ matrix whose entries are the evaluation of the kernel function between subject $i$ and subject $j$ . Default is <code>NULL</code> .
<code>wait</code>	Should the R terminal hang until all <code>max_designs</code> vectors are found? The default is <code>FALSE</code> .
<code>start</code>	Should we start searching immediately (default is <code>TRUE</code> ).
<code>num_cores</code>	The number of CPU cores you wish to use during the search. The default is 1.

### Value

An object of type `optimal_experimental_design_search` which can be further operated upon

### Author(s)

Adam Kapelner

---

```
initRerandomizationExperimentalDesignObject
  Begin a Rerandomization Search
```

---

### Description

This method creates an object of type `rerandomization_experimental_design` and will immediately initiate a search through  $S_1$  space for forced-balance designs.

### Usage

```
initRerandomizationExperimentalDesignObject(
  X = NULL,
  obj_val_cutoff_to_include,
  max_designs = 1000,
  objective = "mahal_dist",
  Kgram = NULL,
  wait = FALSE,
  start = TRUE,
  num_cores = 1
)
```

### Arguments

<code>X</code>	The design matrix with $n$ rows (one for each subject) and $p$ columns (one for each measurement on the subject). This is the design matrix you wish to search for a more optimal design.
<code>obj_val_cutoff_to_include</code>	Only allocation vectors with objective values lower than this threshold will be returned. If the cutoff is infinity, you are doing BCRD and you should use the <code>complete_randomization_with_forced_balanced</code> function instead.
<code>max_designs</code>	The maximum number of designs to be returned. Default is 10,000. Make this large so you can search however long you wish as the search can be stopped at any time by using the <code>stopSearch</code> method
<code>objective</code>	The objective function to use when searching design space. This is a string with valid values "mahal_dist" (the default), "abs_sum_diff" or "kernel".
<code>Kgram</code>	If the <code>objective = kernel</code> , this argument is required to be an $n \times n$ matrix whose entries are the evaluation of the kernel function between subject $i$ and subject $j$ . Default is <code>NULL</code> .
<code>wait</code>	Should the R terminal hang until all <code>max_designs</code> vectors are found? The default is <code>FALSE</code> .
<code>start</code>	Should we start searching immediately (default is <code>TRUE</code> ).
<code>num_cores</code>	The number of CPU cores you wish to use during the search. The default is 1.



**Value**

An object of type rerandomization\_experimental\_design\_search which can be further operated upon.

**Author(s)**

Adam Kapelner

---

plot.greedy\_experimental\_design\_search  
*Plots a summary of a greedy\_experimental\_design\_search object*

---

**Description**

Plots a summary of a greedy\_experimental\_design\_search object

**Usage**

```
## S3 method for class 'greedy_experimental_design_search'  
plot(x, ...)
```

**Arguments**

x                    The greedy\_experimental\_design\_search object to be summarized in the plot  
...                   Other parameters to pass to the default plot function

**Value**

An array of order statistics from [plot\\_obj\\_val\\_order\\_statistic](#) as a list element

**Author(s)**

Adam Kapelner

---

plot\_obj\_val\_by\_iter *Plots the objective value by iteration*

---

**Description**

Plots the objective value by iteration

**Usage**

```
plot_obj_val_by_iter(res, runs = NULL)
```

**Arguments**

res	Results from a greedy search object
runs	A vector of run indices you would like to see plotted (default is to plot the first up to 9)

**Author(s)**

Adam Kapelner

---

plot\_obj\_val\_order\_statistic  
*Plots an order statistic of the object value as a function of number of searches*

---

**Description**

Plots an order statistic of the object value as a function of number of searches

**Usage**

```
plot_obj_val_order_statistic(  
  obj,  
  order_stat = 1,  
  skip_every = 5,  
  type = "o",  
  ...  
)
```

**Arguments**

<code>obj</code>	The <code>greedy_experimental_design_search</code> object whose search history is to be visualized
<code>order_stat</code>	The order statistic that you wish to plot. The default is 1 for the minimum.
<code>skip_every</code>	Plot every <code>nth</code> point. This makes the plot generate much more quickly. The default is 5.
<code>type</code>	The type parameter for plot.
<code>...</code>	Other arguments to be passed to the plot function.

**Value**

An array of order statistics as a list element

**Author(s)**

Adam Kapelner

---

`print.binary_experimental_design`

*Prints a summary of a `binary_experimental_design` object*

---

**Description**

Prints a summary of a `binary_experimental_design` object

**Usage**

```
## S3 method for class 'binary_experimental_design'  
print(x, ...)
```

**Arguments**

<code>x</code>	The <code>binary_experimental_design</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

```
print.binary_then_greedy_experimental_design
    Prints a summary of a binary_then_greedy_experimental_design
    object
```

---

**Description**

Prints a summary of a `binary_then_greedy_experimental_design` object

**Usage**

```
## S3 method for class 'binary_then_greedy_experimental_design'
print(x, ...)
```

**Arguments**

<code>x</code>	The <code>binary_then_greedy_experimental_design</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

```
print.binary_then_rerandomization_experimental_design
    Prints a summary of a binary_then_rerandomization_experimental_design
    object
```

---

**Description**

Prints a summary of a `binary_then_rerandomization_experimental_design` object

**Usage**

```
## S3 method for class 'binary_then_rerandomization_experimental_design'
print(x, ...)
```

**Arguments**

<code>x</code>	The <code>binary_then_rerandomization_experimental_design</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

```
print.greedy_experimental_design_search
    Prints a summary of a greedy_experimental_design_search ob-
    ject
```

---

**Description**

Prints a summary of a `greedy_experimental_design_search` object

**Usage**

```
## S3 method for class 'greedy_experimental_design_search'
print(x, ...)
```

**Arguments**

x	The <code>greedy_experimental_design_search</code> object to be summarized in the console
...	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

```
print.karp_experimental_design_search
    Prints a summary of a karp_experimental_design_search object
```

---

**Description**

Prints a summary of a `karp_experimental_design_search` object

**Usage**

```
## S3 method for class 'karp_experimental_design_search'
print(x, ...)
```

**Arguments**

x	The <code>karp_experimental_design_search</code> object to be summarized in the console
...	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

```
print.optimal_experimental_design_search
    Prints a summary of a optimal_experimental_design_search ob-
    ject
```

---

**Description**

Prints a summary of a `optimal_experimental_design_search` object

**Usage**

```
## S3 method for class 'optimal_experimental_design_search'
print(x, ...)
```

**Arguments**

<code>x</code>	The <code>optimal_experimental_design_search</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

```
print.rerandomization_experimental_design_search
    Prints a summary of a rerandomization_experimental_design_search
    object
```

---

**Description**

Prints a summary of a `rerandomization_experimental_design_search` object

**Usage**

```
## S3 method for class 'rerandomization_experimental_design_search'
print(x, ...)
```

**Arguments**

<code>x</code>	The <code>rerandomization_experimental_design_search</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default print function

**Author(s)**

Adam Kapelner

---

`resultsBinaryMatchSearch`*Returns unique allocation vectors that are binary matched*

---

**Description**

Returns unique allocation vectors that are binary matched

**Usage**

```
resultsBinaryMatchSearch(  
  obj,  
  num_vectors = 1000,  
  objective = NULL,  
  form = "zero_one"  
)
```

**Arguments**

<code>obj</code>	The <code>binary_experimental_design</code> object where the pairs are computed.
<code>num_vectors</code>	How many random allocation vectors you wish to return. The default is 1000.
<code>objective</code>	Should we compute all the objective values for each allocation? Default is NULL for "no". If non-null, it needs to either be "mahal_dist" or "abs_sum_diff".
<code>form</code>	Which form should it be in? The default is <code>one_zero</code> for 1/0's or <code>pos_one_min_one</code> for +1/-1's.

**Author(s)**

Adam Kapelner

---

`resultsBinaryMatchThenGreedySearch`*Returns unique allocation vectors that are binary matched*

---

**Description**

Returns unique allocation vectors that are binary matched

**Usage**

```
resultsBinaryMatchThenGreedySearch(  
  obj,  
  num_vectors = NULL,  
  compute_obj_vals = FALSE,  
  form = "zero_one"  
)
```

**Arguments**

obj	The binary_then_greedy_experimental_design object where the pairs are computed.
num_vectors	How many random allocation vectors you wish to return. The default is NULL indicating you want all of them.
compute_obj_vals	Should we compute all the objective values for each allocation? Default is FALSE.
form	Which form should it be in? The default is one_zero for 1/0's or pos_one_min_one for +1/-1's.

**Author(s)**

Adam Kapelner

---

resultsBinaryMatchThenRerandomizationSearch

*Returns unique allocation vectors that are binary matched*

---

**Description**

Returns unique allocation vectors that are binary matched

**Usage**

```
resultsBinaryMatchThenRerandomizationSearch(
  obj,
  num_vectors = NULL,
  compute_obj_vals = FALSE,
  form = "zero_one"
)
```

**Arguments**

obj	The binary_then_greedy_experimental_design object where the pairs are computed.
num_vectors	How many random allocation vectors you wish to return. The default is NULL indicating you want all of them.
compute_obj_vals	Should we compute all the objective values for each allocation? Default is FALSE.
form	Which form should it be in? The default is one_zero for 1/0's or pos_one_min_one for +1/-1's.

**Author(s)**

Adam Kapelner



---

resultsGreedySearch    *Returns the results (thus far) of the greedy design search*

---

### Description

Returns the results (thus far) of the greedy design search

### Usage

```
resultsGreedySearch(obj, max_vectors = 9, form = "one_zero")
```

### Arguments

obj	The greedy_experimental_design object that is currently running the search
max_vectors	The number of design vectors you wish to return. NULL returns all of them. This is not recommended as returning over 1,000 vectors is time-intensive. The default is 9.
form	Which form should it be in? The default is one_zero for 1/0's or pos_one_min_one for +1/-1's.

### Author(s)

Adam Kapelner

### Examples

```
## Not run:
library(MASS)
data(Boston)
#pretend the Boston data was an experiment setting
#first pull out the covariates
X = Boston[, 1 : 13]
#begin the greedy design search
ged = initGreedyExperimentalDesignObject(X,
max_designs = 1000, num_cores = 2, objective = "abs_sum_diff")
#wait
res = resultsGreedySearch(ged, max_vectors = 2)
design = res$ending_indicTs[, 1] #ordered already by best-->worst
design
#what is the balance on this vector?
res$obj_vals[1]
#compute balance explicitly in R to double check
compute_objective_val(X, design) #same as above
#how far have we come?
ged
#we can cut it here
stopSearch(ged)

## End(Not run)
```

---

resultsKarpSearch      *Returns the results (thus far) of the karp design search*

---

**Description**

Returns the results (thus far) of the karp design search

**Usage**

resultsKarpSearch(obj)

**Arguments**

obj                      The karp\_experimental\_design object that is currently running the search

**Author(s)**

Adam Kapelner

---

resultsOptimalSearch      *Returns the results (thus far) of the optimal design search*

---

**Description**

Returns the results (thus far) of the optimal design search

**Usage**

resultsOptimalSearch(obj, num\_vectors = 1, form = "one\_zero")

**Arguments**

obj                      The optimal\_experimental\_design object that is currently running the search

num\_vectors            How many allocation vectors you wish to return. The default is 1 meaning the best vector. If Inf, it means all vectors.

form                    Which form should it be in? The default is one\_zero for 1/0's or pos\_one\_min\_one for +1/-1's.

**Author(s)**

Adam Kapelner

---

`resultsRerandomizationSearch`*Returns the results (thus far) of the rerandomization design search*

---

**Description**

Returns the results (thus far) of the rerandomization design search

**Usage**

```
resultsRerandomizationSearch(  
  obj,  
  include_assignments = FALSE,  
  form = "one_zero"  
)
```

**Arguments**

<code>obj</code>	The <code>rerandomization_experimental_design</code> object that is currently running the search
<code>include_assignments</code>	Do we include the assignments (takes time) and default is FALSE.
<code>form</code>	Which form should the assignments be in? The default is <code>one_zero</code> for 1/0's or <code>pos_one_min_one</code> for +1/-1's.

**Author(s)**

Adam Kapelner

---

`searchTimeElapsed`*Returns the amount of time elapsed*

---

**Description**

Returns the amount of time elapsed

**Usage**

```
searchTimeElapsed(obj)
```

**Arguments**

<code>obj</code>	The <code>experimental_design</code> object that is currently running the search
------------------	--

**Author(s)**

Adam Kapelner

---

`standardize_data_matrix`*Standardizes the columns of a data matrix.*

---

**Description**

Standardizes the columns of a data matrix.

**Usage**

```
standardize_data_matrix(X)
```

**Arguments**

X                    The n x p design matrix

**Value**

The n x p design matrix with columns standardized

**Author(s)**

Adam Kapelner

---

`startSearch`*Starts the parallelized greedy design search.*

---

**Description**

Once begun, this function cannot be run again.

**Usage**

```
startSearch(obj)
```

**Arguments**

obj                    The experimental\_design object that will be running the search

**Author(s)**

Adam Kapelner

---

stopSearch	<i>Stops the parallelized greedy design search.</i>
------------	---

---

**Description**

Once stopped, it cannot be restarted.

**Usage**

```
stopSearch(obj)
```

**Arguments**

obj	The experimental_design object that is currently running the search
-----	---

**Author(s)**

Adam Kapelner

---

summary.binary_experimental_design	<i>Prints a summary of a binary_experimental_design object</i>
------------------------------------	--

---

**Description**

Prints a summary of a binary\_experimental\_design object

**Usage**

```
## S3 method for class 'binary_experimental_design'  
summary(object, ...)
```

**Arguments**

object	The binary_experimental_design object to be summarized in the console
...	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner

```
summary.binary_then_greedy_experimental_design
```

```
Prints a summary of a binary_then_greedy_experimental_design
object
```

---

**Description**

Prints a summary of a binary\_then\_greedy\_experimental\_design object

**Usage**

```
## S3 method for class 'binary_then_greedy_experimental_design'
summary(object, ...)
```

**Arguments**

object	The binary_then_greedy_experimental_design object to be summarized in the console
...	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner

---

```
summary.binary_then_rerandomization_experimental_design
```

```
Prints a summary of a binary_then_rerandomization_experimental_design
object
```

---

**Description**

Prints a summary of a binary\_then\_rerandomization\_experimental\_design object

**Usage**

```
## S3 method for class 'binary_then_rerandomization_experimental_design'
summary(object, ...)
```

**Arguments**

object	The binary_then_rerandomization_experimental_design object to be summarized in the console
...	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner

---

```
summary.greedy_experimental_design_search
```

```
Prints a summary of a greedy_experimental_design_search object
```

---

**Description**

Prints a summary of a greedy\_experimental\_design\_search object

**Usage**

```
## S3 method for class 'greedy_experimental_design_search'  
summary(object, ...)
```

**Arguments**

object	The greedy_experimental_design_search object to be summarized in the console
...	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner

---

```
summary.karp_experimental_design_search
```

```
Prints a summary of a karp_experimental_design_search object
```

---

**Description**

Prints a summary of a karp\_experimental\_design\_search object

**Usage**

```
## S3 method for class 'karp_experimental_design_search'  
summary(object, ...)
```

**Arguments**

object	The karp_experimental_design_search object to be summarized in the console
...	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner

---

```
summary.optimal_experimental_design_search
```

```
Prints a summary of a optimal_experimental_design_search ob-
ject
```

---

**Description**

Prints a summary of a `optimal_experimental_design_search` object

**Usage**

```
## S3 method for class 'optimal_experimental_design_search'
summary(object, ...)
```

**Arguments**

<code>object</code>	The <code>optimal_experimental_design_search</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner

---

```
summary.rerandomization_experimental_design_search
```

```
Prints a summary of a rerandomization_experimental_design_search
object
```

---

**Description**

Prints a summary of a `rerandomization_experimental_design_search` object

**Usage**

```
## S3 method for class 'rerandomization_experimental_design_search'
summary(object, ...)
```

**Arguments**

<code>object</code>	The <code>rerandomization_experimental_design_search</code> object to be summarized in the console
<code>...</code>	Other parameters to pass to the default summary function

**Author(s)**

Adam Kapelner



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