

# Package ‘DiSCos’

January 20, 2025

**Title** Distributional Synthetic Controls Estimation

**Version** 0.1.1

**Description** The method of synthetic controls is a widely-adopted tool for evaluating causal effects of policy changes in settings with observational data. In many settings where it is applicable, researchers want to identify causal effects of policy changes on a treated unit at an aggregate level while having access to data at a finer granularity. This package implements a simple extension of the synthetic controls estimator, developed in Gunsilius (2023) <[doi:10.3982/ECTA18260](https://doi.org/10.3982/ECTA18260)>, that takes advantage of this additional structure and provides nonparametric estimates of the heterogeneity within the aggregate unit. The idea is to replicate the quantile function associated with the treated unit by a weighted average of quantile functions of the control units. The package contains tools for aggregating and plotting the resulting distributional estimates, as well as for carrying out inference on them.

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**BugReports** <https://github.com/Davidvandijcke/DiSCos/issues>

**URL** <http://www.davidvandijcke.com/DiSCos/>,  
<https://github.com/Davidvandijcke/DiSCos>

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DiSCo *Distributional Synthetic Controls*

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## Description

This function implements the distributional synthetic controls (DiSCo) method from Gunsilius (2023). as well as the alternative mixture of distributions approach.

## Usage

```
DiSCo(
  df,
  id_col.target,
  t0,
  M = 1000,
  G = 1000,
  num.cores = 1,
  permutation = FALSE,
  q_min = 0,
  q_max = 1,
  CI = FALSE,
  boots = 500,
  replace = TRUE,
  uniform = FALSE,
  cl = 0.95,
  graph = FALSE,
  qmethod = NULL,
  qtype = 7,
```

```

seed = NULL,
simplex = FALSE,
mixture = FALSE,
grid.cat = NULL
)

```

## Arguments

<code>df</code>	Data frame or data table containing the distributional data for the target and control units. The data table should contain the following columns: <ul style="list-style-type: none"> <li>• <code>y_col</code> A numeric vector containing the outcome variable for each unit. Units can be individuals, states, etc., but they should be nested within a larger unit (e.g. individuals or counties within a state)</li> <li>• <code>id_col</code> A numeric vector containing the aggregate IDs of the units. This could be, for example, the state if the units are counties or individuals</li> <li>• <code>time_col</code> A vector containing the time period of the observation for each unit. This should be a monotonically increasing integer.</li> </ul>
<code>id_col.target</code>	Variable indicating the name of the target unit, as specified in the <code>id_col</code> column of the data table. This variable can be any type, as long as it is the same type as the <code>id_col</code> column of the data table.
<code>t0</code>	Integer indicating period of treatment.
<code>M</code>	Integer indicating the number of control quantiles to use in the DiSCo method. Default is 1000.
<code>G</code>	Integer indicating the number of grid points for the grid on which the estimated functions are evaluated. Default is 1000.
<code>num.cores</code>	Integer, number of cores to use for parallel computation. Default is 1. If the permutation or CI arguments are set to TRUE, this can be slow and it is recommended to set this to 4 or more, if possible. If you get an error in "all cores" or similar, try setting <code>num.cores=1</code> to see the precise error value.
<code>permutation</code>	Logical, indicating whether to use the permutation method for computing the optimal weights. Default is FALSE.
<code>q_min</code>	Numeric, minimum quantile to use. Set this together with <code>q_max</code> to restrict the range of quantiles used to construct the synthetic control. Default is 0 (all quantiles). Currently NOT implemented for the <code>mixture</code> approach.
<code>q_max</code>	Numeric, maximum quantile to use. Set this together with <code>q_min</code> to restrict the range of quantiles used to construct the synthetic control. Default is 1 (all quantiles). Currently NOT implemented for the <code>mixture</code> approach.
<code>CI</code>	Logical, indicating whether to compute confidence intervals for the counterfactual quantiles. Default is FALSE. The confidence intervals are computed using the bootstrap procedure described in Van Dijcke et al. (2024).
<code>boots</code>	Integer, number of bootstrap samples to use for computing confidence intervals. Default is 500.
<code>replace</code>	Logical, indicating whether to sample with replacement when computing the bootstrap samples. Default is TRUE.

<code>uniform</code>	Logical, indicating whether to construct uniform bootstrap confidence intervals. Default is FALSE. If FALSE, the confidence intervals are pointwise.
<code>cl</code>	Numeric, confidence level for the (two-sided) confidence intervals.
<code>graph</code>	Logical, indicating whether to plot the permutation graph as in Figure 3 of the paper. Default is FALSE.
<code>qmethod</code>	Character, indicating the method to use for computing the quantiles of the target distribution. The default is NULL, which uses the <code>quantile</code> function from the stats package. Other options are "qkden" (based on smoothed kernel density function) and "extreme" (based on parametric extreme value distributions). Both are substantially slower than the default method but may be useful for fat-tailed distributions with few data points at the upper quantiles. Alternatively, one could use the <code>q_max</code> option to restrict the range of quantiles used.
<code>qtype</code>	Integer, indicating the type of quantile to compute when using <code>quantile</code> in the <code>qmethod</code> argument. The default 7. See the documentation for the <code>quantile</code> function for more information.
<code>seed</code>	Integer, seed for the random number generator. This needs to be set explicitly in the function call, since it will invoke <code>RNGkind</code> which will set the seed for each core when using parallel processes. Default is NULL, which does not set a seed.
<code>simplex</code>	Logical, indicating whether to use to constrain the optimal weights to the unit simplex. Default is FALSE, which only constrains the weights to sum up to 1 but allows them to be negative.
<code>mixture</code>	Logical, indicating whether to use the mixture of distributions approach instead. See Section 4.3. in Gunsilius (2023). This approach minimizes the distance between the CDFs instead of the quantile functions, and is preferred for categorical variables. When working with such variables, one should also provide a list of support points in the <code>grid.cat</code> parameter. When that is provided, this parameter is automatically set to TRUE. Default is FALSE.
<code>grid.cat</code>	List, containing the discrete support points for a discrete grid to be used with the mixture of distributions approach. This is useful for constructing synthetic distributions for categorical variables. Default is NULL, which uses a continuous grid based on the other parameters.

### Details

This function is called for every time period in the `DiSCo` function. It implements the `DiSCo` method for a single time period, as well as the mixture of distributions approach. The corresponding results for each time period can be accessed in the `results.periods` list of the output of the `DiSCo` function. The `DiSCo` function returns the average weight for each unit across all periods, calculated as a uniform mean, as well as the counterfactual target distribution produced as the weighted average of the control distributions for each period, using these averaged weights.

### Value

A list containing the following elements:

- `results.periods` A list containing, for each time period, the elements described in the return argument of `DiSCo_iter`, as well as the following additional elements:

- DiSco
  - \* `quantile` The counterfactual quantiles for the target unit.
  - \* `weights` The optimal weights for the target unit.
  - \* `cdf` The counterfactual CDF for the target unit.
- `weights` A numeric vector containing the synthetic control weights for the control units, averaged over time. When `mixture` is TRUE, these are the weights for the mixture of distributions, otherwise they are the weights for the quantile-based approach.
- `CI` A list containing the confidence intervals for the counterfactual quantiles and CDFs, if `CI` is TRUE. Each element contains two named subelements called `upper`, `lower`, `se` which are the upper and lower confidence bands and the standard error of the estimate, respectively. They are  $G \times T$  matrices where  $G$  is the specified number of grid points and  $T$  is the number of time periods. The elements are:
  - `cdf` The bootstrapped CDF
  - `quantile` The bootstrapped quantile
  - `quantile_diff` The bootstrapped quantile difference
  - `cdf_diff` The bootstrapped CDF difference
  - `bootmat` A list containing the raw bootstrapped samples for the counterfactual quantiles and CDFs, if `CI` is TRUE. These are not meant to be accessed directly, but are used by DiSCoTEA to compute aggregated standard errors. Advanced users may wish to access these directly for further analysis. The element names should be self-explanatory. #'
  - `control_ids` A list containing the control unit IDs used for each time period, which can be used to identify the weights associated with each control as the returned weights have the same order as the control IDs.
  - `perm` A `permut` object containing the results of the permutation method, if `permutation` is TRUE. Call `summary` on this object to print the overall results of the permutation test. #'
  - `evgrid` A numeric vector containing the grid points on which the quantiles were evaluated.
  - `params` A list containing the parameters used in the function call.

## References

Gunsilius FF (2023). “Distributional synthetic controls.” *Econometrica*, **91**(3), 1105–1117.

Van Dijcke D, Gunsilius F, Wright AL (2024). “Return to Office and the Tenure Distribution.” Working Paper 2024-56, University of Chicago, Becker Friedman Institute for Economics.()

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DiSCoT

*Store aggregated treatment effects*

---

## Description

S3 object holding aggregated treatment effects

**Usage**

```
DiSCoT(  
  agg,  
  treats,  
  ses,  
  grid,  
  ci_lower,  
  ci_upper,  
  t0,  
  call,  
  cl,  
  N,  
  J,  
  agg_df,  
  perm,  
  plot  
)
```

**Arguments**

agg	aggregation method
treats	list of treatment effects
ses	list of standard errors
grid	grid
ci_lower	list of lower confidence intervals
ci_upper	list of upper confidence intervals
t0	start time
call	call
cl	confidence level
N	number of observations
J	number of treated units
agg_df	dataframe of aggregated treatment effects and their confidence intervals
perm	list of per mutation results
plot	a ggplot object containing the plot for the aggregated treatment effects using the agg parameter

**Value**

S3 object of class DiSCoT with associated summary and print methods

DiSCoTEA

*Aggregate treatment effects from DiSCo function.***Description**

Function to aggregate treatment effects from the output of the DiSCo function, plot the distribution of the aggregation statistic over time, and report summary tables.

**Usage**

```
DiSCoTEA(
  disco,
  agg = "quantileDiff",
  graph = TRUE,
  t_plot = NULL,
  savePlots = FALSE,
  xlim = NULL,
  ylim = NULL,
  samples = c(0.25, 0.5, 0.75)
)
```

**Arguments**

<code>disco</code>	Output of the DiSCo function.
<code>agg</code>	String indicating the aggregation statistic to be used. Options include <ul style="list-style-type: none"> <li>• <code>quantileDiff</code> Difference in quantiles between the target and the weighted average of the controls.</li> <li>• <code>quantile</code> Plots both the observed and the counterfactual quantile functions. No summary statistics will be produced.</li> <li>• <code>cdfDiff</code> Difference in CDFs between the target and the weighted average of the controls.</li> <li>• <code>cdf</code> Plots both the observed and the counterfactual CDFs. No summary statistics will be produced.</li> </ul>
<code>graph</code>	Boolean indicating whether to plot graphs (default is TRUE).
<code>t_plot</code>	Optional vector of time periods ( <code>t_col</code> values in the original dataframe) to be plotted (default is NULL, which plots all time periods).
<code>savePlots</code>	Boolean indicating whether to save the plots to the current working directory (default is FALSE). The plot names will be <code>[agg]_[start_year]_[end_year].pdf</code> .
<code>xlim</code>	Optional vector of length 2 indicating the x-axis limits of the plot. Useful for zooming in on relevant parts of the distribution for fat-tailed distributions.
<code>ylim</code>	Optional vector of length 2 indicating the y-axis limits of the plot.
<code>samples</code>	Numeric vector indicating the range of quantiles of the aggregation statistic ( <code>agg</code> ) to be summarized in the summary property of the S3 class returned by the function (default is <code>c(0.25, 0.5, 0.75)</code> ). For example, if <code>samples = c(0.25, 0.5,</code>

0.75), the summary table will include the average effect for the 0-25th, 25-50th, 50-75th and 75-100th quantiles of the distribution of the aggregation statistic over time.

## Details

This function takes in the output of the `DiSCo_per` function and computes aggregate treatment effect using a user-specified aggregation statistic. The default is the differences between the counterfactual and the observed quantile functions (`quantileDiff`). If `graph` is set to `TRUE`, the function will plot the distribution of the aggregation statistic over time. The S3 class returned by the function has a `summary` property that will print a selection of aggregated effects (specified by the `samples` parameter) for the chosen `agg` method, by post-treatment year (see examples below). This summary call will only print effects if the `agg` parameter requested a distribution difference (`quantileDiff` or `cdfDiff`). The other aggregations are meant to be inspected visually. If the permutation parameter was set to `TRUE` in the original `DiSCo` call, the summary table will include the results of the permutation test. If the original `DiSCo` call was restricted to a range of quantiles smaller than  $[0, 1]$  (i.e. `q_min > 0` or `q_max < 1`), the `samples` parameter is ignored and only the aggregated differences for the quantile range specified in the original call are returned.

## Value

A `DiSCoT` object, which is an S3 class that stores a list of treatment effects, their standard errors, the corresponding confidence intervals (if specified), and a dataframe with treatment effects aggregated according to the `agg` input. The S3 class also has a `summary` property that will print a selection of aggregated effects (specified by the `samples` parameter) for the chosen `agg` method, by post-treatment year, as well as the permutation test results, if specified.

---

DiSCo\_iter

*Estimate DiSCo in a single period*

---

## Description

This function implements the DiSCo method for a single time period, as well as the mixture of distributions approach. Its return values contain valuable period-specific estimation outputs.

## Usage

```
DiSCo_iter(
  yy,
  df,
  evgrid,
  id_col.target,
  M,
  G,
  T0,
  qmethod = NULL,
  qtype = 7,
```



```

    q_min = 0,
    q_max = 1,
    simplex = FALSE,
    controls.id,
    grid.cat,
    mixture
)

```

## Arguments

yy	Integer indicating the current year being processed.
df	Data frame or data table containing the distributional data for the target and control units. The data table should contain the following columns: <ul style="list-style-type: none"> <li>• <code>y_col</code> A numeric vector containing the outcome variable for each unit. Units can be individuals, states, etc., but they should be nested within a larger unit (e.g. individuals or counties within a state)</li> <li>• <code>id_col</code> A numeric vector containing the aggregate IDs of the units. This could be, for example, the state if the units are counties or individuals</li> <li>• <code>time_col</code> A vector containing the time period of the observation for each unit. This should be a monotonically increasing integer.</li> </ul>
evgrid	A vector of grid points on which to evaluate the quantile functions.
id_col.target	Variable indicating the name of the target unit, as specified in the <code>id_col</code> column of the data table. This variable can be any type, as long as it is the same type as the <code>id_col</code> column of the data table.
M	Integer indicating the number of control quantiles to use in the DiSCo method. Default is 1000.
G	Integer indicating the number of grid points for the grid on which the estimated functions are evaluated. Default is 1000.
T0	Integer indicating the last pre-treatment period starting from 1.
qmethod	Character, indicating the method to use for computing the quantiles of the target distribution. The default is <code>NULL</code> , which uses the <code>quantile</code> function from the <code>stats</code> package. Other options are <code>"qkden"</code> (based on smoothed kernel density function) and <code>"extreme"</code> (based on parametric extreme value distributions). Both are substantially slower than the default method but may be useful for fat-tailed distributions with few data points at the upper quantiles. Alternatively, one could use the <code>q_max</code> option to restrict the range of quantiles used.
qtype	Integer, indicating the type of quantile to compute when using <code>quantile</code> in the <code>qmethod</code> argument. The default 7. See the documentation for the <code>quantile</code> function for more information.
q_min	Numeric, minimum quantile to use. Set this together with <code>q_max</code> to restrict the range of quantiles used to construct the synthetic control. Default is 0 (all quantiles). Currently NOT implemented for the <code>mixture</code> approach.
q_max	Numeric, maximum quantile to use. Set this together with <code>q_min</code> to restrict the range of quantiles used to construct the synthetic control. Default is 1 (all quantiles). Currently NOT implemented for the <code>mixture</code> approach.

<code>simplex</code>	Logical, indicating whether to use to constrain the optimal weights to the unit simplex. Default is FALSE, which only constrains the weights to sum up to 1 but allows them to be negative.
<code>controls.id</code>	List of strings specifying the column names for the control units' identifiers.
<code>grid.cat</code>	List, containing the discrete support points for a discrete grid to be used with the mixture of distributions approach. This is useful for constructing synthetic distributions for categorical variables. Default is NULL, which uses a continuous grid based on the other parameters.
<code>mixture</code>	Logical, indicating whether to use the mixture of distributions approach instead. See Section 4.3. in Gunsilius (2023). This approach minimizes the distance between the CDFs instead of the quantile functions, and is preferred for categorical variables. When working with such variables, one should also provide a list of support points in the <code>grid.cat</code> parameter. When that is provided, this parameter is automatically set to TRUE. Default is FALSE.

### Details

This function is part of the DiSCo method, called for each time period. It calculates the optimal weights for the DiSCo method and the mixture of distributions approach for a single time period. The function processes data `f` or both the target and control units, computes the quantile functions, and evaluates these on a specified grid. The function is designed to be used within the broader context of the DiSCo function, which aggregates results across multiple time periods.

### Value

A list with the following elements:

- `DiSCo_weights` Weights calculated using the DiSCo method.
- `mixture`
  - `weights` Optimal weights for the mixture approach.
  - `distance` Value of the objective function for the mixture approach.
  - `mean` Weighted mixture of the controls' CDFs.
- `target`
  - `cdf` Empirical CDF of the target. Only computed when `mixture=TRUE`.
  - `grid` Grid on which the quantile and CDF functions were evaluated.
  - `data` Original data for the target unit.
  - `quantiles` Quantiles for the target unit, evaluated on the specified grid.
- `controls`
  - `data` Original data for the control units.
  - `cdf` Empirical CDFs of the control units. Only computed when `mixture=TRUE`.
  - `quantiles` Quantiles for the control units, evaluated on the specified grid. .
- `controls.q` Quantiles for the control units, evaluated on the specified grid.

---

dube	<i>Data from (Dube 2019)</i>
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**Description**

As used in the empirical application of Gunsilius (2023).

**Usage**

dube

**Format**

dube:

A data frame with 652,870 rows and 3 columns:

**id\_col** State FIPS

**time\_col** Year

**y\_col** `adj0contpov` variable in Dube (2019). Captures the distribution of equalized family income from wages and salary, defined as multiples of the federal poverty threshold. ...

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ex_gmm	<i>ex_gmm</i>
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---

**Description**

Example data for DiSCo command. Returns simulated target and control that are mixtures of Gaussian distributions.

**Usage**

`ex_gmm(Ts = 2, num.con = 30, numdraws = 1000)`

**Arguments**

Ts	an integer indicating the number of time periods
num.con	an integer indicating the number of control units
numdraws	an integer indicating the number of draws

**Value**

target	a vector.
control	a matrix.

---

permut	<i>permut</i>
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**Description**

Object to hold results of permutation test

**Usage**

```
permut(distp, distt, p_overall, J_1, q_min, q_max, plot)
```

**Arguments**

distp	List of squared Wasserstein distances between the control units
distt	List of squared Wasserstein distances between the target unit and the control units
p_overall	Overall p-value
J_1	Number of control units
q_min	Minimum quantile
q_max	Maximum quantile
plot	ggplot object containing plot of squared Wasserstein distances over time for all permutations.

**Value**

A list of class `permut`, with the same elements as the input arguments.

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summary.DiSCoT	<i>summary.DiSCoT</i>
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---

**Description**

Summary of DiSCoT object

**Usage**

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

**Arguments**

object	DiSCoT object
...	Additional arguments

**Value**

summary of DiSCoT object

---

<code>summary.permut</code>	<i>summary.permut</i>
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**Description**

Summarize permutation test results

**Usage**

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

**Arguments**

<code>object</code>	Object of class <code>permut</code>
<code>...</code>	Additional arguments

**Value**

Prints permutation test results

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