

# simcadi: Similarity Indices for Categorical Distributions

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**Abstract.** In this article we introduce the command `simcadi`. It helps to calculate indicators for the similarity of categorically ordered variables and distribution, respectively. In particular, it permits the calculation of the Cosine index, and indices introduced by Finger and Kreinin (1979), Bray and Curtis (1957), Dice (1945), Sørensen (1948), Jaccard (1912), Grubel and Lloyd (1971), Ružička (1958), and Gower (1971). Moreover, it allows us to compute the development of a distribution over time. The command offers various options for an efficient handling of datasets, because it permits the calculation of benchmarks of comparison automatically, and the incorporation of complex weighting schemes.

**Note** The `simcadi` command was written for the Statistical Software Stata (©StataCorp). It is available upon request, or can be downloaded here:

[www.uni-regensburg.de/wirtschaftswissenschaften/vwl-moeller/forschung/index.html](http://www.uni-regensburg.de/wirtschaftswissenschaften/vwl-moeller/forschung/index.html)

## 1 Introduction

Measuring the relationship between two variables is essential for empirical work. If variables contain categorically ordered observations, the preferred method for measuring the relationship between variables is similarity indices and dissimilarity indices<sup>2</sup>, respectively. The most well-known indices were introduced by Finger and Kreinin (1979), Bray and Curtis (1957), Dice (1945), Sørensen (1948), Jaccard (1912), Grubel and Lloyd (1971), Ružička (1958), and Gower (1971). These indices are used extensively in economic and ecological research, and their mathematical features were studied intensively in Boyce et al. (1995), Boriah et al. (2008), Egghe (2010), and Goshtasby (2012), for example.

In this article, we introduce the `simcadi` command. It facilitates the organization of a dataset in such a way that it is easy to calculate a series of similarity indices for categorical variables and distributions. Although there is great diversity in the definition of similarity indices, the indices which are implemented in `simcadi` can all handle categorically ordered samples, and range from zero to one, with zero indicating no similarity, and one indicating maximum similarity. The indices are the result of a function of scores resulting from the comparison of the value measured for an attribute

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2. One is the logical complement of the other: a similarity index measures ‘how close’ two samples are, whereas a dissimilarity index measures the reverse.

in one sample with the value measured for the same attribute in a second sample (cf. Johnston 1976, p.11f).

Basically, `simcadi` offers three main features: First, `simcadi` can handle datasets that are organized differently (wide vs. long). That means, the comparison of two distributions is not only possible for distributions that are organized in two variables (wide), but also if two, or many distributions are captured in one single variable (long). Second, the command permits the choice of one, or more distribution(s) to which all other distributions should be compared. Alternatively, it allows the incorporation of a weighting scheme which is used to calculate a reference of comparison for each distribution. This is useful, for example, if all distributions should be compared to a weighted subset of distributions rather than to a specific distribution. To give an example, assume you want to compare the disaggregated exports of country A (B,C,...) to only those countries with which country A (B,C,...) shares a common border. Using our command, you only need the disaggregated (categorized) export flows and a contiguity matrix of all the exporting countries. Then, the command uses both to compare the export structure of country A (B,C,...) to the average export structure of all countries that share a common border with country A (B,C,...). Moreover, the command helps to implement other user-written indicators, because the `simcadi` command optionally permits the possibility of saving the variables of comparison. Third, in addition to the indices mentioned above, the command permits the indication of the change in a distribution over time, as was done by De Benedictis and Tajoli (2007b,a), for example.

Similar to metric distance measurements, these similarity indices for categorically ordered variables can also be used to quantify the difference between samples. However, the similarity indices usually do not satisfy the axiom of triangle inequality like distance measurements.<sup>3</sup> In contrast to the calculation and analysis of metric distance measurements, which is already well implemented in Stata (see Fenty 2004), we are not aware of a Stata command to calculate similarity indices.<sup>4</sup> Nevertheless, it has to be mentioned that some similarity indices can be used optionally to execute a cluster analysis in Stata (see *measure\_option*), and that the `matrix dissimilarity` command does in fact calculate some similarity measurements. However, it does so for matrices only, incorporates no options for handling the dataset or calculating distributions to work as a benchmark of comparison. Moreover, it is not specifically built to measure similarities of categorical distributions.

In the following sections we introduce the calculation of the indices, explain the syntax, and offer a series of examples to show the flexibility of the command.

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3. According to this axiom, the distance between object *a* and *b* is shorter than the sum of the distances via another point, *c*.

4. The R-Project community offers the `vegdist` command. This is part of the `vegan` package (see <https://github.com/vegandevs/vegan>) and enables some dissimilarity indices to be computed. However, the options for handling the dataset to calculate an individual benchmark of comparison are limited.

## 2 The `simcadi` command

`simcadi` implements seven indices all of which are defined between zero and one. We overview the formulae of calculation in section 1, where  $s_{Atk} = x_{Atk}/X_{At}$  is the value shares of distribution  $A$  in category  $k$  at time  $t$ , with  $X_{At} = \sum_k x_{Atk}$ . The value shares of distribution  $B$ , to which distribution  $A$  is compared, are denoted with  $s_{Btk}$ , and  $M_A$  is the number of columns (excluding missing values).

Please note that the naming conventions vary. The Finger-Kreinin Index, for example, was reinvented several times and is also known under various names as found in works by researchers like Czekanowski (1909), Dice (1945), and Sørensen (1948). Another example is the quantitative version of the Jaccard index, which is sometimes named after Ružička (1958). To give a further example, the Bray-Curtis Index which is probably the most well known index, is sometimes also called the Finger-Kreinin Index, or vice versa, because when using shares—as `simcadi` does by default—both indicators are algebraically identical.

In a nutshell, all measurements consider the overlap of two distributions in some way, whereby the Finger-Kreinin Index (FKI) does so the most directly. The main advantage of the Finger-Kreinin Index is that it is “not influenced by the relative sizes or scales” (Finger and Kreinin 1979, p.906) of the category, and is invariant with respect to proportional sub-classifications, which is shown with an axiomatic approach in Sun and Ng (2000). It should be considered that some indices may give misleading values if data entries are not strictly positive.

The Change in Distribution Index differs from the indices mentioned previously, because it does not compare two distributions but uses one of the indices mentioned previously to calculate the change of one distribution over time. Thus, it is a function of the distribution in two points in time:  $t$  and  $t'$ . We exemplify its calculation with the Finger-Kreinin Index in section 1. However, any other similarity index can be used for calculation. Consequently, it is also defined between zero and one, with zero indicating that the distribution did not change, and one indicating that the distribution in  $t'$  has nothing in common with the distribution in  $t$ .

### 2.1 Syntax

There are three ways to use the command: The first permits the comparison of two variables (*variable\_1 variable\_2*).

```
simcadi variable_1 variable_2 [if][in], class(varname) [options]
```

The second can be used to compare a number of distributions (specified with `id()`) that are captured in one single variable (*variable\_1*) to the distributions specified in `wcountry()`.

```
simcadi variable_1 [if][in], class(varname) id(varname) wcountry(name)
```

Table 1: Formulae for Calculating Similarity Indices

Indicator	Description
Finger-Kreinin Index (FKI)	$FKI_{At} = \sum_k \min(s_{Atk}, s_{Btk})$
The Bray-Curtis Index (BCI)	$BCI_{At} = 1 - \frac{\sum_k  s_{Atk} - s_{Btk} }{\sum_k (s_{Atk} + s_{Btk})}$
Gower Index (GOW)	$GOW_{it} = 1/M_A \cdot \sum_k \left( \frac{ s_{Atk} - s_{Btk} }{(\max(s_{At}) - \min(s_{At}))} \right)$
Ružička Index (RUZ)	$RUZ_{At} = \frac{\sum_k \min(s_{Atk}, s_{Btk})}{\sum_k \max(s_{Atk}, s_{Btk})}$
Jaccard Index (JAC)	$JAC_{At} = \frac{\sum_k (s_{Atk} \cdot s_{Btk})}{(\sum_k s_{Atk}^2 + \sum_k s_{Btk}^2 - \sum_k s_{Atk} \cdot s_{Btk})}$
Grubel-Lloyd Index (GLI)	$GLI_{At} = \frac{1}{k} \sum_k \frac{(s_{Atk} + s_{Btk*}) -  s_{Atk} - s_{Btk} }{(s_{Atk} + s_{Btk})}$
Cosine Index (COS)	$COS_{At} = \frac{\sum_k (s_{Atk} \cdot s_{Btk})}{\sqrt{\sum_k s_{Atk}^2} + \sqrt{\sum_k s_{Btk}^2}}$
Change in Distribution (CID)	$CID_{At,t'}^{\text{Similarity Index}}(s_{Atk}, s_{A,t',k})$
Example:	$CID_{At,t'}^{\text{FKI}} = 1 - \sum_k \min(s_{Atk}, s_{A,t=t-t',k})$

[options]

The third should be used to incorporate an external weighting scheme set by **using filename**. **simcadi** uses this scheme to calculate a benchmark distribution of comparison for each distribution.

```
simcadi variable_1 using filename [if][in], class(varname) id(varname)
[options]
```

## 2.2 Options

**class(varname)** specifies the variable containing the categories of the distributions (e.g. the trade classification).

**id(varname)** specifies the variable that identifies the distributions (e.g. the countries).

**wcountry(name)** sets the distribution with which each distribution should be compared. Note: if more than one distribution is set, the unweighted average of the distributions is calculated to work as the distribution of comparison.

**wvarname(varname)** specifies the name of the variable that contains the weights (only applicable if a weighting matrix is assigned).

`varpartner(varname)` specifies the name of the second variable in the weighting scheme file (e.g. the trading partner). Only applicable only if a weighting matrix is assigned.

`cid(#)` calculates the Change in Distribution index (for each indicator chosen). `#` sets a period to which each `'id(varname)'` is compared.

`realvalues` use the real values instead of the shares. Note: Applying real values follows that some indices (Gower) are defined between zero and one. (The default uses shares.)

`time(#)` sets the time period (e.g. 2012).

`timevar(varname)` specifies the name of the time variable.

`savecomp(filename)` stores the file with the distributions/variables that are compared (Note: The user can use this file to calculate further indices.)

`saveresult(filename)` stores the calculated indices under the name filename.

`detail` offers a detailed description of the calculation process.

`braycurtis` calculates the Bray-Curtis Index (only allowed if the variable of interest contains positive values only).

`cosine` calculates the Cosine Index.

`finger` calculates the Finger-Kreinin Index.

`gower` calculates the Gower Index.

`grubel` calculates the Grubel-Lloyd Index.

`jaccard` calculates the Jaccard Index.

`ruzicka` calculates the Ruzicka Index.

### 3 Examples...

#### ...how `simcadi` can handle 'wide' data

Assume you have consumption data for three people (Adam, Brittany, Charlie) who consume three goods (beer, bread, water) at two different periods (1, 2) in the wide format. The following Stata output shows how to compare the consumption profile of Adam with that of Charlie:

```
. use example_wide, clear
. list
```

time	good	Adam	Brittany	Charlie
------	------	------	----------	---------

1.	1	beer	18	9	0		
2.	1	bread	18	15	10		
3.	1	water	18	13	10		
4.	2	beer	19	7	1		
5.	2	bread	18	11	12		
6.	2	water	17	16	12		

  

```
. simcadi Charlie Adam, class(good) timevar(time) realvalues
Calculation of the indices...
Merge the results...
Results were not saved
. list
```

1.	time	finger-d	jaccar-d	braycu-d	gowerR-d	cosine-d	grubel-d
	1	20	.4433497	.5405405	1.133333	.8164966	.4761905
	ruzick-d			compare2			
	.3703704			Charlie with Adam			

  

2.	time	finger-d	jaccar-d	braycu-d	gowerR-d	cosine-d	grubel-d
	2	25	.532767	.6329114	.8787879	.8274392	.5758621
	ruzick-d			compare2			
	.462963			Charlie with Adam			

If you are only interested in the development of the distribution from period one to period two, you can use the `cid` option together with `time`:

```
. simcadi Charlie Adam, class(good) time(2) timevar(time) cid(1) finger cosine
Calculation of the indices...
Merge the results...
Results were not saved
. list
```

	time	CIDfin-d	finger-d	CIDcos-d	cosine-d	compare2
1.	2	.04	.6881481	.0017316	.8274392	Charlie with Adam

### ...how simcadi can handle 'long' data

`simcadi` is especially built to handle datasets in the long format. The Stata excerpt below shows how to measure the similarity of Adam's consumption with that of Brittany and Charlie using the `wcountry` option if all the consumption data are captured within a single variable.

```
. list
```

--

	name	time	good	consume
1.	Adam	1	beer	18
2.	Adam	1	bread	18
3.	Adam	1	water	18
4.	Brittany	1	beer	9
5.	Brittany	1	bread	15
6.	Brittany	1	water	13
7.	Charlie	1	beer	0
8.	Charlie	1	bread	10
9.	Charlie	1	water	10
10.	Adam	2	beer	19
11.	Adam	2	bread	18
12.	Adam	2	water	17
13.	Brittany	2	beer	7
14.	Brittany	2	bread	11
15.	Brittany	2	water	16
16.	Charlie	2	beer	1
17.	Charlie	2	bread	12
18.	Charlie	2	water	12

```
. simcadi consume , class(good) id(name) wcountry(Adam) ///
>       time(1) timevar(time) savecomp(filename, replace) ///
>       finger gower
```

A: No obvious errors.

Q: Did the user specify a weight dataset?

A: NO. The exports should be compared to the equally weighted average of the following countries (Adam).

Automatic calculation of the weighting scheme...  
...DONE

file filename.dta saved

Calculation of the indices...

Warning: The Gower Index cannot be calculated properly. Please check your data and the prerequisites of the Gower Index.

Merge the results...  
...DONE

Results were not saved

. list

	name	time	finger-d	gowerg-d
1.	Adam	1	1	.
2.	Brittany	1	.9099099	.3703704
3.	Charlie	1	.6666667	.4444444

. use filename, clear

. list

	name	time	good	comp2	comp1
1.	Adam	1	beer	.3333333	.3333333

2.	Adam	1	bread	.3333333	.3333333
3.	Adam	1	water	.3333333	.3333333
4.	Brittany	1	beer	.3333333	.2432432
5.	Brittany	1	bread	.3333333	.4054054
6.	Brittany	1	water	.3333333	.3513514
7.	Charlie	1	beer	.3333333	0
8.	Charlie	1	bread	.3333333	.5
9.	Charlie	1	water	.3333333	.5

A warning note shows that the Gower Index cannot be calculated properly. Here, it is easy to show what went wrong by considering the formula of the Gower Index, and by looking at the content of the file ‘filename’, which was saved by the option `savecomp()`: The denominator of the Gower Index is zero for the comparison of Adam with Adam, which is undefined. We want to highlight that the saved file can also be used to calculate a user-specified index. Moreover, the `wcountry` option also allows for more than one distribution to be included. `wcountry(Adam Brittany Charlie)`, for example, would yield a benchmark of comparison that contains the average shares of consumption for all three persons. The next section explains how to use more sophisticated weighting schemes.

#### ...how `simcadi` can handle user-specified weighting schemes

Sometimes the user wishes to specify a more complex weighting scheme. To give an example, trade economist are often interested in the export composition of countries, but do not wish to compare the exports of all countries to a single export composition, rather than comparing the exports of each country to only those countries that are on the same continent, or to countries that share a common border. What is even more complex, but quite common in spatial science, is that economists wish to weight the countries in the comparison differently with the inverse distance between countries, for example. In the following, we explain how this can be done using `simcadi`.

Assume you want to compare Brittany with all the others, but the men should only be compared to other men. Additionally, for men, you want to put more weight on other men’s distributions. This means that you do not want to compare each distribution to a single distribution, but aim to compare each distribution with a specific distribution. The following Stata excerpt shows how this can be done. We add the `detail` option to give a deeper insight into what the command is doing: First, the command checks whether or not obvious errors exist. These could be incorrectly named variables, or weighting data that were named or specified incorrectly, for example. Second, `simcadi` computes a distribution of comparison for each person as follows:

$$s_{i*tk} = \sum_{j:j \neq i} (w_{ij} \cdot s_{jtk}), \quad (1)$$

where  $w_{ij}$  denotes the weight between subjects  $i$  and  $j$ . Our command automatically ensures that the weights add up to one for each subject,  $\sum_j w_{ij} = 1 \forall i$ . Depending on



the weighting matrix, the resulting benchmark may be different for each subject. For more details, please see the following Stata Output excerpt:

```
. use weightdata,clear
. list
```

	name	name2	weight
1.	Adam	Adam	1
2.	Adam	Brittany	0
3.	Adam	Charlie	2
4.	Brittany	Adam	1
5.	Brittany	Brittany	1
6.	Brittany	Charlie	1
7.	Charlie	Adam	2
8.	Charlie	Brittany	0
9.	Charlie	Charlie	1

```
. use example_consume, clear
. simcadi consume using weightdata, class(good) id(name) time(1) timevar(time) ///
> varpartner(name2) detail finger grubel savecomp(compfile, replace)
```

---

```
Q: Any obvious errors?
Value variable exists          ---> consume
Time variable exists          ---> time
id variable exists            ---> name
Weight variable (in weight dataset) exists ---> weight
Categorization exists         ---> good
Variable name exists in the weight dataset
Variable name2 exists in the weight dataset
A: No obvious errors.
```

---

```
Q: Which index should be calculated?
A: The Grubel-Lloyd Index.
A: The Finger-Kreinin Index.
A: The indicator(s) are calculated using shares.
3 = # of distinct values of name
```

---

```
Q: Did the user specify a weight dataset?
A: YES ---> weightdata
```

---

```
Q: Do the weights add up to one for each name (in the raw weight data)
A: NO, we need to adjust the weight...
...weights successfully adjusted
```

---

```
Calculation of the s_i*kt...
3 = # of distinct values of good
27 = # of observations in the bilateral dataset
Merge the weighting scheme with trade data...
```

---

```
Q: Are all countries in the weighting scheme matched successfully?
Note: If this is not the case, the weights do not add up to one for each name
A: YES
```

---

```
file compfile.dta saved
Calculation of the indices...
```

```

Calculation of the Grubel-Lloyd index...
...DONE
Calculation of the Finger-Kreinin index...
...DONE
Merge the results...
...DONE

```

---

```
Results were not saved
```

```
. list
```

	name	time	finger-d	grubel-d
1.	Adam	1	.7777778	.7380952
2.	Brittany	1	.9489489	.9384114
3.	Charlie	1	.7777778	.5833333

```
. use compfile, clear
```

```
. list
```

	name	time	good	comp2	comp1
1.	Adam	1	beer	.1111111	.3333333
2.	Adam	1	bread	.4444445	.3333333
3.	Adam	1	water	.4444445	.3333333
4.	Brittany	1	beer	.1921922	.2432432
5.	Brittany	1	bread	.4129129	.4054054
6.	Brittany	1	water	.3948949	.3513514
7.	Charlie	1	beer	.2222222	0
8.	Charlie	1	bread	.3888889	.5
9.	Charlie	1	water	.3888889	.5

## 4 References

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