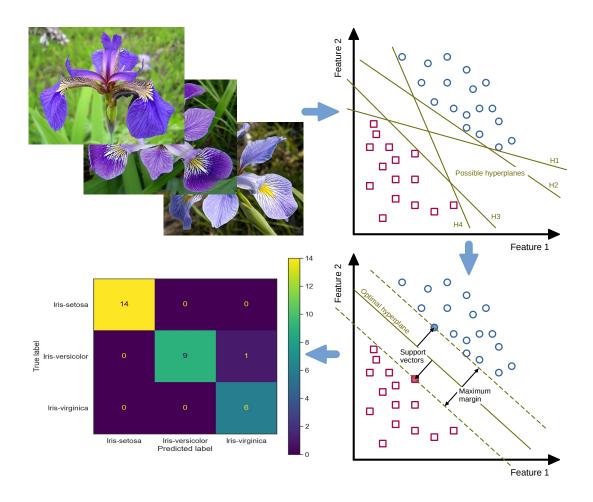
Application of the processed survey data in the analytical hierarchy process (AHP)

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This is a placeholder for the abstract that needs to be added later.



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1 Introduction

Why we use a Jupyter notebook to to publish the R program examples:

Jupyter is a new **open source** alternative to the proprietary numerical software Mathematica from Wolfram Research that is well on the way to become a standard for exchanging research results (Somers 2018; Romer 2018).

Originally Jupyter was intended as an IDE for the programming languages **Julia** and **Python**. Besides that it is also possible to install other interpreter kernels, such as the **IRkernel** for R. This can be interesting if the IDE **RStudio Desktop** is not available on the target platform used. For example, it is very difficult to install RStudio on the ARM-based embedded computer **Raspberry Pi** due to many technical dependencies. In contrast, using the R kernel in JupyterLab on the Raspberry Pi works very well and performant.

2 Global settings and dependencies

2.1 Install missing packages if not present yet

Attention: For some R packages several dependencies have to be installed first with apt install chage name>.

Dependencies for package ahpsurvey:

- R package randomNames (it depends on R 4.0, refer to https://cran.r-project.org/web/packages/randomNames/index.html)

Drawback for Raspbian buster: the dependency randomNames is not available for R v3.5.2 as it depends on R (4.0). Upgrading R in Raspbian following the instruction on https://cran.rstudio.com/bin/linux/debian/#debian-buster-stable does not work so far ...

[1] "All required packages are installed."

2.2 Load package data.table

The package data.table is used to read and manipulate tables (data.table inherits from data.frame). Install and load it:

```
[12]: library(data.table)
```

2.3 Load package ggplot2

The package ggplot2 is used to plot diagrams. Install and load it:

```
[13]: library(ggplot2)
```

2.4 Load package tidyr

The package tidyr is used to **reshape** the dataframes and provides functions like gather() or spread(). Some examples for the application can be found here: Reshaping Your Data with tidyrReshaping Your Data with tidyr.

Install and load it:

```
[14]: library(tidyr)
```

2.5 Load package dplyr

The package dplyr is necessary to manipulate dataframes using functions like select(), mutate() and left join(). Install and load it:

Hint: Setting the parameter warn.conflicts=FALSE when calling the library() function silences annoying messages about masked functions.

```
[15]: library(dplyr, warn.conflicts=FALSE)
```

2.6 Use pipes for better coding

HINT: The pipe functionality is already available by loading the library tidyr - so you don't have to load it explicitly.

What pipes like %>% are and how to use them is described here: https://statistik-dresden.de/archives/15679.

Before using pipes in R, you have to install and load the package magrittr:

```
[16]: library(magrittr, warn.conflicts=FALSE)
```

2.7 Load package ahpsurvey

The package ahpsurvey contains all the necessary mathematical and statistical methods to run the analytical hierarchy process (AHP).

```
[17]: library(ahpsurvey)
```

3 Functions for processing AHP

3.1 Set globally used input and output folders

```
[18]: str_input_path = "./output_data_manipulated"
str_output_path = "./output_data_AHP"
```

3.2 Function for reading in processed survey data from CSV files to data frames

Define a function for reading in a CSV file to a date frame.

```
[19]: func_readCSVdata_to_dataframe <- function(str_CSVfilename) {

    df_CSVdata <- fread(
        file = str_CSVfilename, encoding = "UTF-8",
        header = TRUE, sep = "\t", quote = "\""
    )

    return(df_CSVdata)
}</pre>
```

3.3 Function for generating a data frame with eigentrue values (weights)

```
[20]: func_genEigentrue_to_dataframe <- function(df_surveyData, vec_attributes) {
    list_mat_judgement <- df_surveyData %>%
        ahp.mat(vec_attributes, negconvert = TRUE)

    df_eigentrue <- ahp.indpref(list_mat_judgement, vec_attributes, method = "eigen")
    return(df_eigentrue)
}</pre>
```

3.4 Function for generating an array with consistency ratios

```
[21]: func_genCR_to_arr <- function(df_surveyData, vec_attributes) {
    arr_cr <- df_surveyData %>%
    ahp.mat(vec_attributes, negconvert = TRUE) %>%
    ahp.cr(vec_attributes, ri=0.58)

    return(arr_cr)
}
```

3.5 Function for generating a data frame with consistency ratios

3.6 Function for visualizing individual priorities and consistency ratios

```
geom_boxplot(alpha = 0, width = 0.3, color = "#808080") +
      scale_x_discrete("Attribute", label = vec_labels) +
      scale_y_continuous("Weight (dominant eigenvalue)",
                         labels = scales::percent,
                         breaks = c(seq(0,0.7,0.1))) +
      guides(color=guide_legend(title=NULL))+
      scale_color_discrete(breaks = c(0,1),
                           labels = c(paste("CR >", consistency_thres),
                                      paste("CR <", consistency_thres))) +</pre>
      labs(NULL, caption = paste("n =", nrow(df_surveyData), ",", "Mean CR =",
                               round(mean(arr_cr),3))) +
      theme_minimal() +
      ggtitle("Violins displaying priorities and consistency ratios")
  # save generated ggplot graphic to a PNG image file
  ggsave(filename = str_image_filename, width = 7, height = 7, dpi = 300)
  print(plot)
}
```

3.7 Function for generating geometric mean values from individual judgement matrices

```
[34]: func_aggpref_gmean <- function(df_surveyData, vec_attributes, arr_cr,_u
       ⇔consistency_thres=0.1, str_CRlabel) {
        df_cr <- df_surveyData %>%
          ahp.mat(vec_attributes, negconvert = TRUE) %>%
          ahp.cr(vec_attributes, ri=0.58) %>%
          data.frame() %>%
          mutate(rowid = 1:length(arr_cr), arr_cr.dum = as.factor(ifelse(arr_cr <=_u</pre>
       ⇔consistency_thres, 1, 0)))
        # rename column with consistency ratios
        colnames(df_cr)[1] <- str_CRlabel</pre>
        # combine data frame 'df_cr' with raw survey data ('df_surveyData')
        df_cr_wRaw <- cbind(df_cr, df_surveyData)</pre>
        # remove rows, where 'arr_cr.dum' == 0 (inconsistent data)
        df_cr_wRaw_cons <- df_cr_wRaw[df_cr_wRaw$arr_cr.dum != 0, ]</pre>
        # get individual judgement matrices from last 3 columns
        list_mat_judgement <- df_cr_wRaw_cons[tail(names(df_cr_wRaw_cons), 3)] %>%
          ahp.mat(vec_atts, negconvert = TRUE)
        # get geometric mean values from judgement matrices
        list_gmean_1 <- ahp.aggpref(list_mat_judgement, vec_atts, method = "eigen",_</pre>
       →aggmethod = "geometric")
        return(list_gmean_1)
```

3.8 Function for normalizing the geometric mean values

```
[25]: func_norm_gmean <- function(list_gmeans) {
         # normalization so that the sum of the geometric mean values is 1 (corresponds to_{f \sqcup}
        →100%)
        df_gmean_1 <- data.frame(list_gmeans)</pre>
        # rename column with geometric mean values (raw)
        colnames(df_gmean_1)[1] <- "gmean.raw"</pre>
        gmean_sum <- 0
        for ( val in list_gmeans ) {
          gmean_sum <- gmean_sum + val</pre>
        df_gmean_l["Sum", 1] <- gmean_sum</pre>
        for (idx in 1:length(list_gmeans)) {
          gmean norm <- list gmeans[[idx]] / gmean sum</pre>
          df_gmean_l[idx, "gmean.norm"] <- gmean_norm</pre>
        }
        gmean sum norm <- 0
        # iterate over all rows except the last, because this is the sum itself
        for ( row in 1:(nrow(df_gmean_1)-1) ) {
          gmean_sum_norm <- gmean_sum_norm + df_gmean_l[row, 2]</pre>
        df_gmean_1["Sum", 2] <- gmean_sum_norm</pre>
        return(df_gmean_1)
      }
```

4 Create data frames (tables) handling the file names of processed survey data

4.1 File table for all participants

4.2 File table for city administrations

```
[27]: df_csvInputFiles_CA <- data.table(
    file_idx = 1:4,
    keys = c("env", "soc", "eco", "crit"),
    filenames = c("rdata_CA_env_AHP_edible_Cities_2022-03-18_10-28.csv",</pre>
```

4.3 File table for non-governmental organizations

4.4 File table for practitioners and experts

5 Exploit datasets of own survey with package announce for each group of participants

5.1 All participants

```
[31]: row_start = 1
      row_end = 3
      str_participants_group = "all"
      df_outputTable <- data.table()</pre>
      for ( file_idx in 1:nrow(df_csvInputFiles_all) ) {
        # create data frame from current input CSV file
        str_filename <- paste(str_input_path, df_csvInputFiles_all[file_idx, filenames],u
       ⇔sep="/")
        df_processed_survey_data <- func_readCSVdata_to_dataframe(str_filename)</pre>
        # create vectors for attributes and labels from a subset of data frame_
       → 'df_attributes_labels_all'
        vec_atts <- df_attributes_labels_all[c(row_start:row_end), attr]</pre>
        vec_labels <- df_attributes_labels_all[c(row_start:row_end), labels]</pre>
        # shift row interval for next iteration
        row_start = row_start + 3
        row_end = row_end + 3
        # generate data frame with eigentrue values (weights)
        df_eigentrue_weights <- func_genEigentrue_to_dataframe(df_processed_survey_data,u
        ⇔vec_atts)
        # generate an array with consistency ratios
        arr_CRs <- func_genCR_to_arr(df_processed_survey_data, vec_atts)</pre>
        # generate a extended data frame with consistency ratios
        consistency_thres = 0.1
        str_CRlabel <- paste("CR", df_csvInputFiles_all[file_idx, keys], sep="_")</pre>
        df_CRs <- func_genCR_to_dataframe(df_processed_survey_data, vec_atts, arr_CRs,_
       →consistency_thres, str_CRlabel)
        str_image_filename <- paste("ahp_violin", str_participants_group,_

df_csvInputFiles_PE[file_idx, keys], sep="_")

        str_image_filename <- paste(str_image_filename, ".png", sep="")</pre>
        str_image_filename <- paste(str_output_path, str_image_filename, sep="/")</pre>
        func_visuPriosCRs(df_processed_survey_data, df_CRs, arr_CRs, consistency_thres,_

-vec_atts, df_eigentrue_weights, vec_labels, str_image_filename)

        # combine data frames of eigentrue values (weights) with consistency ratios
```

Violins displaying priorities and consistency ratios

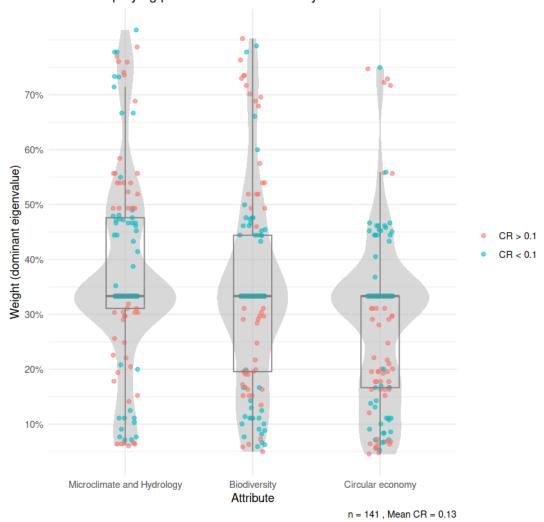


Figure 1: Violins displaying priorities and consistency ratios of all participants

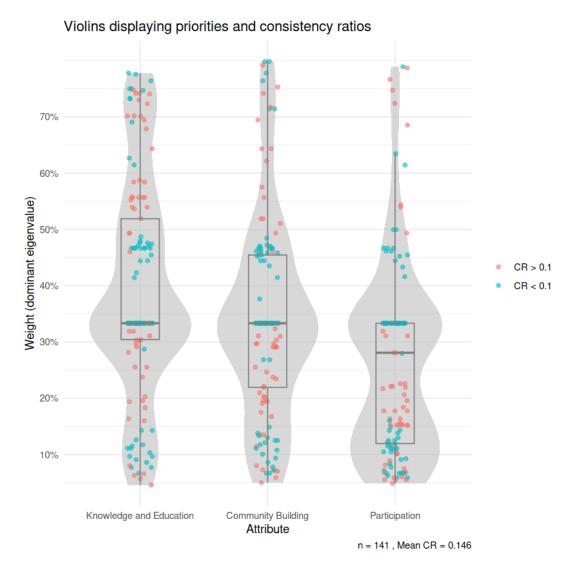


Figure 2: Violins displaying priorities and consistency ratios of all participants

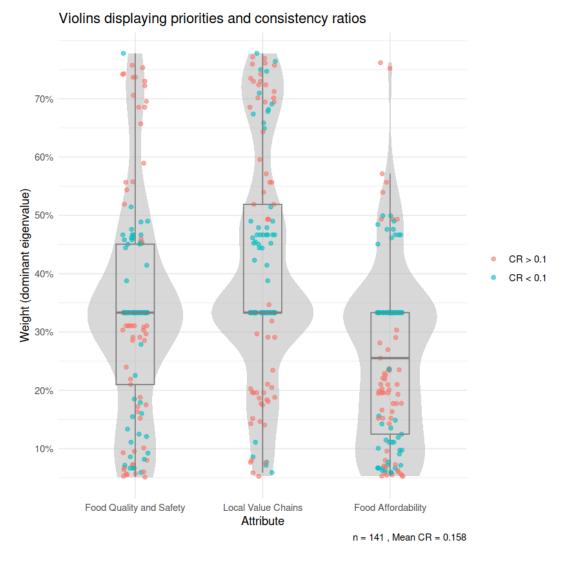


Figure 3: Violins displaying priorities and consistency ratios of all participants

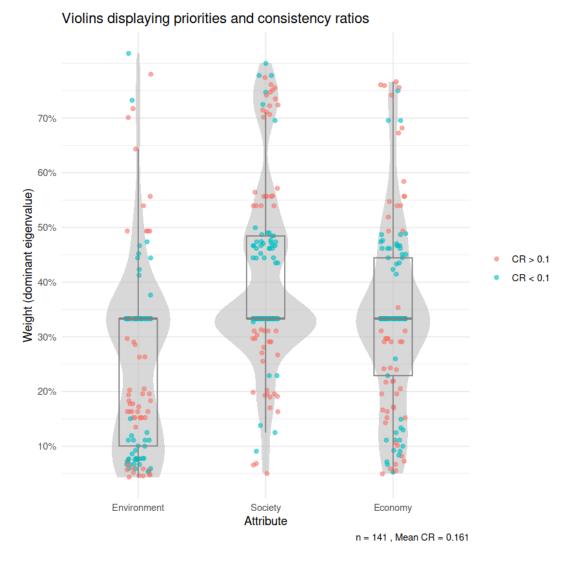


Figure 4: Violins displaying priorities and consistency ratios of all participants

5.2 Participants of city administrations

```
[50]: row_start = 1
    row_end = 3

str_participants_group = "CA"

df_outputTable <- data.table()

for ( file_idx in 1:nrow(df_csvInputFiles_CA) ) {
    # create data frame from current input CSV file
    str_filename <- paste(str_input_path, df_csvInputFiles_CA[file_idx, filenames],___
    -sep="/")
    df_processed_survey_data <- func_readCSVdata_to_dataframe(str_filename)

# create vectors for attributes and labels from a subset of data frame__
    -'df_attributes_labels_all'
    vec_atts <- df_attributes_labels_all[c(row_start:row_end), attr]
    vec_labels <- df_attributes_labels_all[c(row_start:row_end), labels]
    # shift row interval for next iteration</pre>
```

```
row_start = row_start + 3
  row_end = row_end + 3
  # generate data frame with eigentrue values (weights)
 df_eigentrue_weights <- func_genEigentrue_to_dataframe(df_processed_survey_data,u
 ⇔vec_atts)
  # generate an array with consistency ratios
  arr_CRs <- func_genCR_to_arr(df_processed_survey_data, vec_atts)</pre>
  # generate a extended data frame with consistency ratios
  consistency_thres = 0.1
  str_CRlabel <- paste("CR", df_csvInputFiles_CA[file_idx, keys], sep="_")</pre>
 df_CRs <- func_genCR_to_dataframe(df_processed_survey_data, vec_atts, arr_CRs,_
 ⇔consistency_thres, str_CRlabel)
  str_image_filename <- paste("ahp_violin", str_participants_group,_
 df_csvInputFiles_PE[file_idx, keys], sep="_")
 str_image_filename <- paste(str_image_filename, ".png", sep="")</pre>
 str_image_filename <- paste(str_output_path, str_image_filename, sep="/")
 func_visuPriosCRs(df_processed_survey_data, df_CRs, arr_CRs, consistency_thres,_
 evec_atts, df_eigentrue_weights, vec_labels, str_image_filename)
 # combine data frames of eigentrue values (weights) with consistency ratios
 df_outputTable <- cbind(df_outputTable, df_eigentrue_weights)</pre>
  # add only specific columns of 'df_CRs' (omit column 'row_id')
 df_outputTable <- cbind(df_outputTable, df_CRs[c(1, 3)])</pre>
}
# extend file name by path
str_CSVfilename_output <- paste("rdata", str_participants_group, "eigentrue_CRs", __
 ⇔sep="_")
str_CSVfilename_output <- paste(str_CSVfilename_output, ".csv", sep="")</pre>
str_CSVfilename_output <- paste(str_output_path, str_CSVfilename_output, sep="/")
# write data frame 'df_outputTable' to CSV file
write.table(df_outputTable, file = str_CSVfilename_output,
            fileEncoding = "UTF-8", row.names = FALSE,
            col.names = TRUE, sep = "\t", quote = TRUE)
```

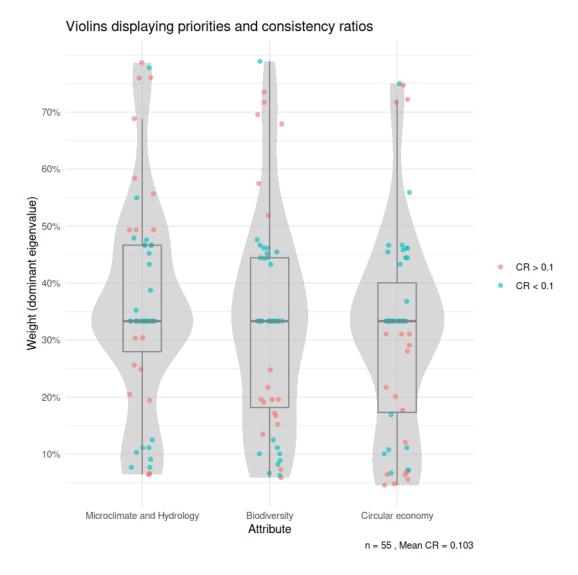


Figure 5: Violins displaying priorities and consistency ratios of participants of city administrations

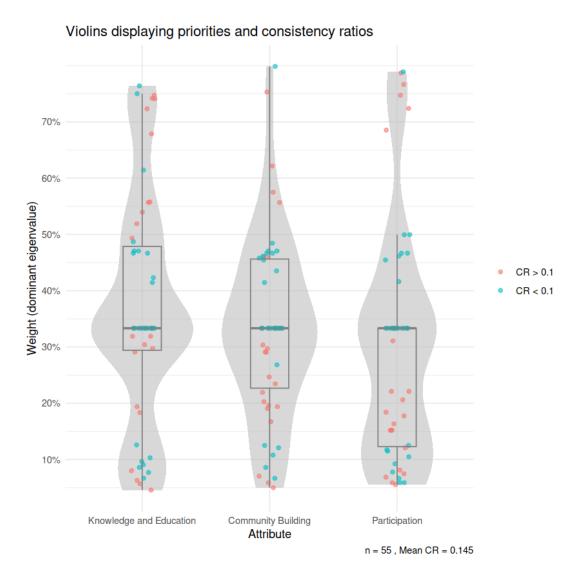


Figure 6: Violins displaying priorities and consistency ratios of participants of city administrations

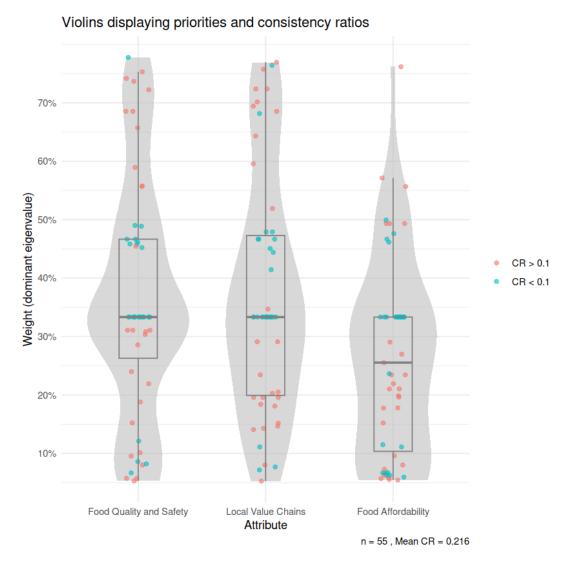


Figure 7: Violins displaying priorities and consistency ratios of participants of city administrations

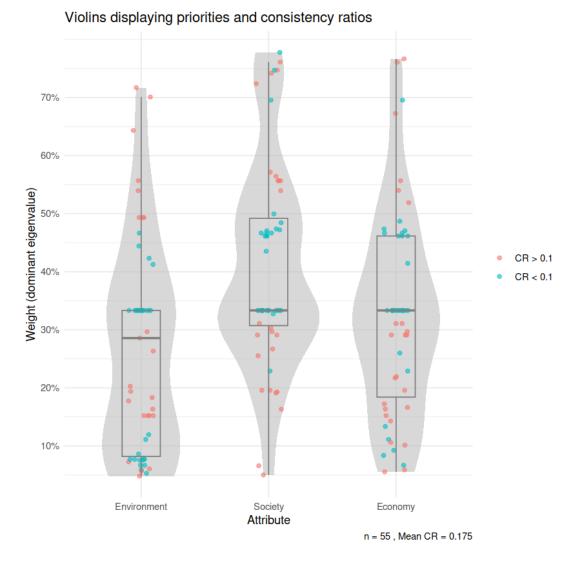


Figure 8: Violins displaying priorities and consistency ratios of participants of city administrations

5.3 Participants of non-governmental organizations

```
row_start = row_start + 3
  row_end = row_end + 3
  # generate data frame with eigentrue values (weights)
 df_eigentrue_weights <- func_genEigentrue_to_dataframe(df_processed_survey_data,u
 ⇔vec_atts)
  # generate an array with consistency ratios
  arr_CRs <- func_genCR_to_arr(df_processed_survey_data, vec_atts)</pre>
  # generate a extended data frame with consistency ratios
  consistency_thres = 0.1
  str_CRlabel <- paste("CR", df_csvInputFiles_NGO[file_idx, keys], sep="_")</pre>
 df_CRs <- func_genCR_to_dataframe(df_processed_survey_data, vec_atts, arr_CRs,_
 ⇔consistency_thres, str_CRlabel)
  str_image_filename <- paste("ahp_violin", str_participants_group,_
 df_csvInputFiles_PE[file_idx, keys], sep="_")
 str_image_filename <- paste(str_image_filename, ".png", sep="")</pre>
 str_image_filename <- paste(str_output_path, str_image_filename, sep="/")
 func_visuPriosCRs(df_processed_survey_data, df_CRs, arr_CRs, consistency_thres,_
 evec_atts, df_eigentrue_weights, vec_labels, str_image_filename)
 # combine data frames of eigentrue values (weights) with consistency ratios
 df_outputTable <- cbind(df_outputTable, df_eigentrue_weights)</pre>
  # add only specific columns of 'df_CRs' (omit column 'row_id')
 df_outputTable <- cbind(df_outputTable, df_CRs[c(1, 3)])</pre>
}
# extend file name by path
str_CSVfilename_output <- paste("rdata", str_participants_group, "eigentrue_CRs", __
 ⇔sep="_")
str_CSVfilename_output <- paste(str_CSVfilename_output, ".csv", sep="")</pre>
str_CSVfilename_output <- paste(str_output_path, str_CSVfilename_output, sep="/")
# write data frame 'df_outputTable' to CSV file
write.table(df_outputTable, file = str_CSVfilename_output,
            fileEncoding = "UTF-8", row.names = FALSE,
            col.names = TRUE, sep = "\t", quote = TRUE)
```

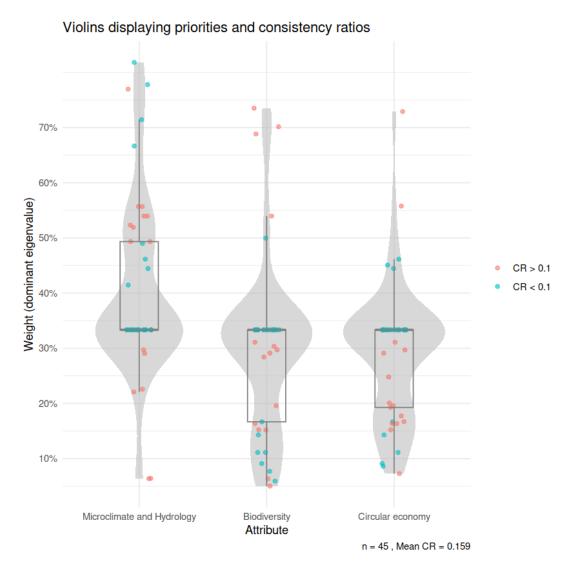


Figure 9: Violins displaying priorities and consistency ratios of participants of non-governmental organizations

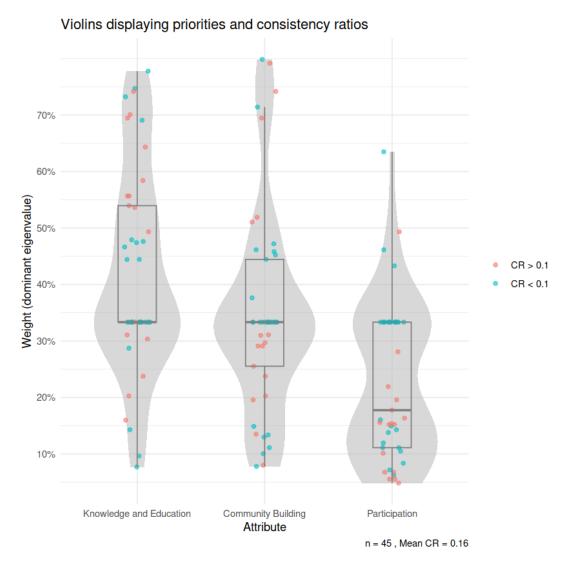


Figure 10: Violins displaying priorities and consistency ratios of participants of non-governmental organizations

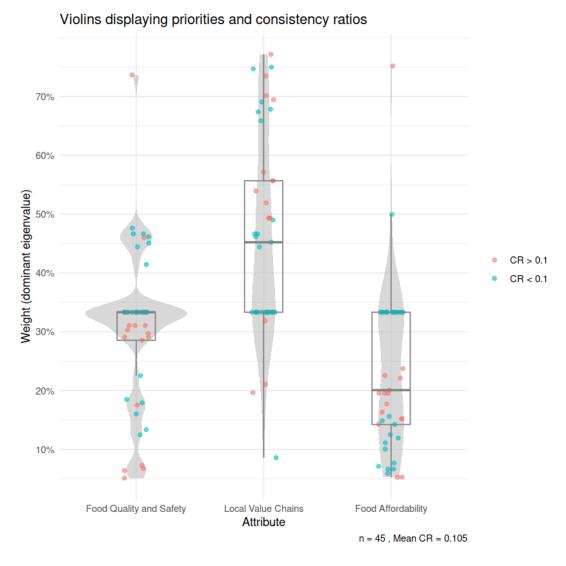


Figure 11: Violins displaying priorities and consistency ratios of participants of non-governmental organizations

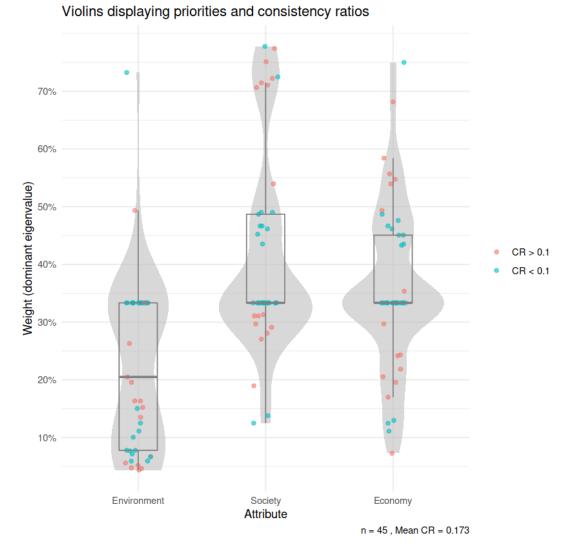


Figure 12: Violins displaying priorities and consistency ratios of participants of non-governmental organizations

5.4 Participants of practitioners and experts

```
# shift row interval for next iteration
  row_start = row_start + 3
 row_end = row_end + 3
  # generate data frame with eigentrue values (weights)
 df_eigentrue_weights <- func_genEigentrue_to_dataframe(df_processed_survey_data,__
 ⇔vec_atts)
  # generate an array with consistency ratios
  arr_CRs <- func_genCR_to_arr(df_processed_survey_data, vec_atts)</pre>
  # generate a extended data frame with consistency ratios
  consistency_thres = 0.1
  str_CRlabel <- paste("CR", df_csvInputFiles_PE[file_idx, keys], sep="_")</pre>
  df_CRs <- func_genCR_to_dataframe(df_processed_survey_data, vec_atts, arr_CRs,_
 ⇔consistency_thres, str_CRlabel)
  str_image_filename <- paste("ahp_violin", str_participants_group,_
 df_csvInputFiles_PE[file_idx, keys], sep="_")
 str_image_filename <- paste(str_image_filename, ".png", sep="")</pre>
 str_image_filename <- paste(str_output_path, str_image_filename, sep="/")</pre>
 func_visuPriosCRs(df_processed_survey_data, df_CRs, arr_CRs, consistency_thres, u
 evec_atts, df_eigentrue_weights, vec_labels, str_image_filename)
 # img <- image_graph(width = 800, height = 800, res = 24)
  # img <- image_read(str_image_filename)</pre>
  # print(img)
 # combine data frames of eigentrue values (weights) with consistency ratios
 df_outputTable <- cbind(df_outputTable, df_eigentrue_weights)</pre>
 # add only specific columns of 'df_CRs' (omit column 'row_id')
 df_outputTable <- cbind(df_outputTable, df_CRs[c(1, 3)])</pre>
# extend file name by path
str_CSVfilename_output <- paste("rdata", str_participants_group, "eigentrue_CRs", u
 ⇔sep="_")
str_CSVfilename_output <- paste(str_CSVfilename_output, ".csv", sep="")</pre>
str_CSVfilename_output <- paste(str_output_path, str_CSVfilename_output, sep="/")
# write data frame 'df_outputTable' to CSV file
write.table(df_outputTable, file = str_CSVfilename_output,
            fileEncoding = "UTF-8", row.names = FALSE,
            col.names = TRUE, sep = "\t", quote = TRUE)
```

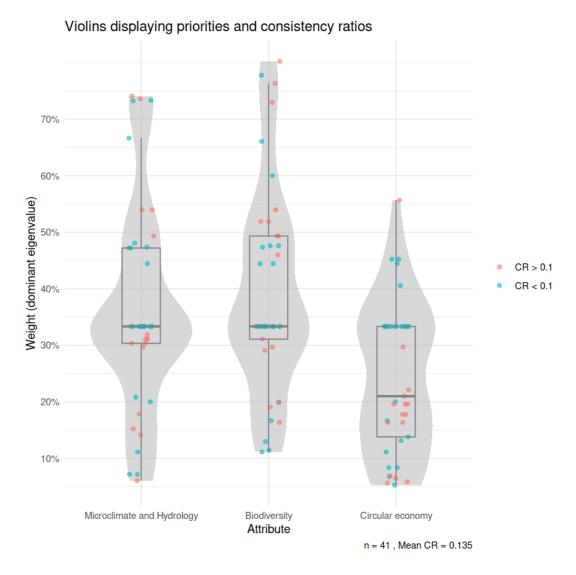


Figure 13: Violins displaying priorities and consistency ratios of participants of practitioners and experts

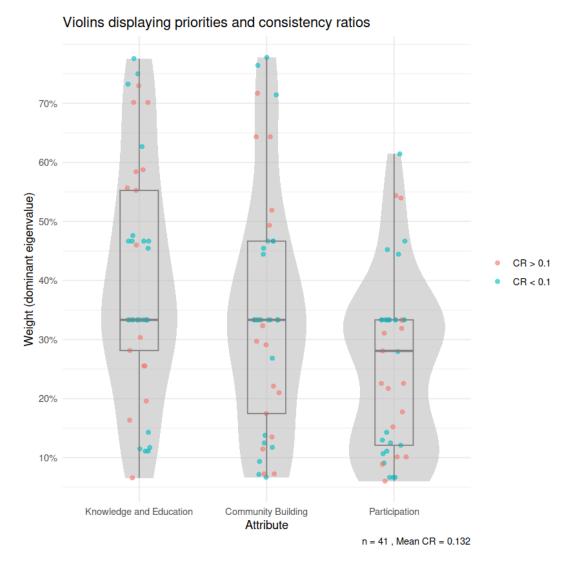


Figure 14: Violins displaying priorities and consistency ratios of participants of practitioners and experts

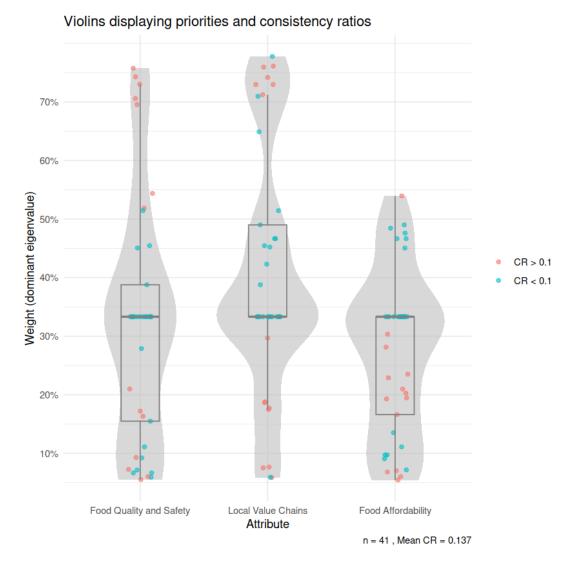


Figure 15: Violins displaying priorities and consistency ratios of participants of practitioners and experts

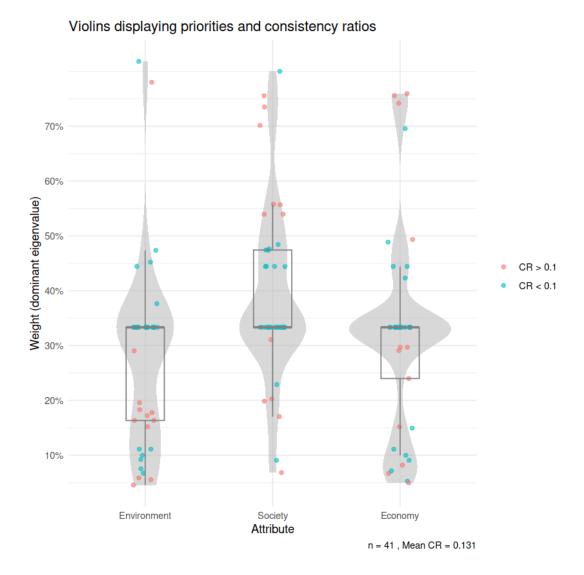


Figure 16: Violins displaying priorities and consistency ratios of participants of practitioners and experts

5.5 Calculation of aggregated preference weights for consistent datasets

```
[36]: list_gmean <- func_aggpref_gmean(df_processed_survey_data, vec_atts, arr_CRs,u
consistency_thres=0.1, str_CRlabel)

df_gmean <- func_norm_gmean(list_gmean)

df_gmean</pre>
```

		gmean.raw	gmean.norm
A data.frame: 4×2		<dbl></dbl>	<dbl $>$
	Envi	0.3739039	0.4179807
	Soci	0.3023657	0.3380094
	Econ	0.2182787	0.2440099
	Sum	0.8945482	1.0000000

6 Summary and outlook

[]:

7 References

Online references

Romer, Paul (Apr. 13, 2018). Jupyter, Mathematica, and the Future of the Research Paper. English. URL: https://paulromer.net/jupyter-mathematica-and-the-future-of-the-research-paper/ (visited on 09/08/2022) (cit. on p. 2).

Somers, James (Apr. 5, 2018). The Scientific Paper Is Obsolete. English. The Atlantic. URL: https://www.theatlantic.com/science/archive/2018/04/the-scientific-paper-is-obsolete/556676/ (visited on 09/08/2022) (cit. on p. 2).