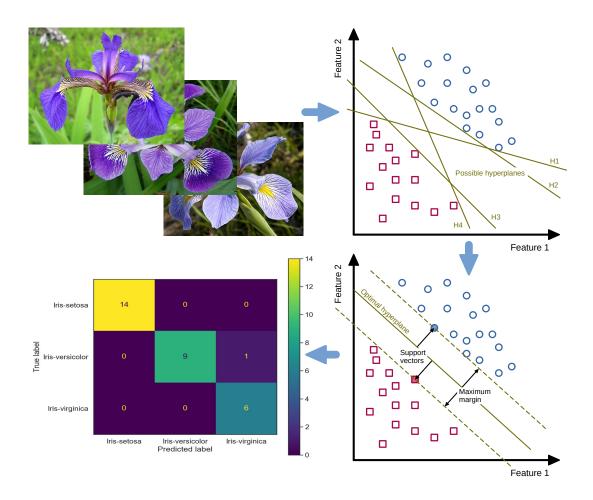
# Preparing raw CSV input data from survey for analytical hierarchy process (AHP)

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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 becoming 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

Installiere Paket nach '/home/bk/R/x86\_64-pc-linux-gnu-library/4.2'
(da 'lib' nicht spezifiziert)

#### 2.2 Load package data.table

The package data.table is used for reading and manipulating tables (data.table inherits from data.frame). Install and load it:

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

#### 2.3 Set globally used input and output folders

```
[3]: str_input_path = "./input_data_from_survey" str_output_path = "./output_data_manipulated"
```

# 2.4 Create data frame (table) handling the file names of input CSV data (raw data from survey)

# 3 Functions for manipulation of raw CSV input data of survey

#### 3.1 Function for reading in survey data from CSV files to data frame objects

Define a function for reading in a CSV file to 4 different date frames by selecting different columns.

```
df_mySurvey_2 <- fread(</pre>
  file = str_CSVfilename, encoding = "UTF-8",
  header = TRUE, sep = "\t", quote = "\"",
  # dec = ".", row.names = "CASE",
  select = c("CASE", "AS01", "AS02", "AS03",
             "RS01_01", "RS02_01", "RS03_01", "RS04_01", "RS05_01", "RS06_01")
  )
df_mySurvey_3 <- fread(</pre>
  file = str_CSVfilename, encoding = "UTF-8",
  header = TRUE, sep = "\t", quote = "\"",
  # dec = ".", row.names = "CASE",
  select = c("CASE", "AW01", "AW02", "AW03",
             "RW01_01", "RW02_01", "RW03_01", "RW04_01", "RW05_01", "RW06_01")
  )
df_mySurvey_4 <- fread(</pre>
  file = str_CSVfilename, encoding = "UTF-8",
  header = TRUE, sep = "\t", quote = "\"",
  # dec = ".", row.var = "CASE",
  select = c("CASE", "AK01", "AK02", "AK03",
             "RK01_01", "RK02_01", "RK03_01", "RK04_01", "RK05_01", "RK06_01")
  )
output <- list(df_mySurvey_1, df_mySurvey_2, df_mySurvey_3, df_mySurvey_4)</pre>
return(output)
```

#### 3.2 Function for manipulation of the read in data and store in new data frame

```
[6]: func_scrambleData <- function(df_inputData, vec_colnames_search_1,_
      ovec_colnames_search_2, vec_colnames_out) {
       # Generate new data frame ...
       df_outputData <- data.frame(matrix(ncol = 3, nrow = 0))</pre>
       # ... and name the columns
       colnames(df_outputData) <- vec_colnames_out</pre>
       # Generate 1. column
       for ( row_idx in 1:nrow(df_inputData) ) {
         # filter column names by vector element
         if (df_inputData[row_idx, colnames(df_inputData) %in% vec_colnames_search_1[1],_
      ⇔with=FALSE] == 1) {
           int_tmp_val <- as.integer(df_inputData[row_idx, colnames(df_inputData) %in%_

¬vec_colnames_search_2[1], with=FALSE])
           int_tmp_val <- int_tmp_val * -1 - 1</pre>
           df_outputData[row_idx, vec_colnames_out[1]] <- int_tmp_val</pre>
         else if (df_inputData[row_idx, colnames(df_inputData) %in%_
      →vec_colnames_search_1[1], with=FALSE] == -1) {
           df_outputData[row_idx, vec_colnames_out[1]] <- 1</pre>
         }
```

```
else if (df_inputData[row_idx, colnames(df_inputData) %in%_
→vec_colnames_search_1[1], with=FALSE] == 2) {
    int_tmp_val <- as.integer(df_inputData[row_idx, colnames(df_inputData) %in%u
⇔vec_colnames_search_2[2], with=FALSE])
    int_tmp_val <- int_tmp_val + 1</pre>
    df_outputData[row_idx, vec_colnames_out[1]] <- int_tmp_val</pre>
  }
# Generate 2. column
for ( row_idx in 1:nrow(df_inputData) ) {
  # filter column names by vector element
  if (df_inputData[row_idx, colnames(df_inputData) %in% vec_colnames_search_1[2],_
⇔with=FALSE] == 1) {
    int_tmp_val <- as.integer(df_inputData[row_idx, colnames(df_inputData) %in%_
→vec_colnames_search_2[3], with=FALSE])
    int_tmp_val <- int_tmp_val * -1 - 1</pre>
    df_outputData[row_idx, vec_colnames_out[2]] <- int_tmp_val</pre>
  else if (df_inputData[row_idx, colnames(df_inputData) %in%_

    vec_colnames_search_1[2], with=FALSE] == -1) {
    df_outputData[row_idx, vec_colnames_out[2]] <- 1</pre>
  }
  else if (df_inputData[row_idx, colnames(df_inputData) %in%_
ovec_colnames_search_1[2], with=FALSE] == 2) {
    int_tmp_val <- as.integer(df_inputData[row_idx, colnames(df_inputData) %in%_

→vec_colnames_search_2[4], with=FALSE])
    int_tmp_val <- int_tmp_val + 1</pre>
    df_outputData[row_idx, vec_colnames_out[2]] <- int_tmp_val</pre>
  }
}
# Generate 3. column
for ( row_idx in 1:nrow(df_inputData) ) {
  # filter column names by vector element
  if (df_inputData[row_idx, colnames(df_inputData) %in% vec_colnames_search_1[3],_
⇔with=FALSE] == 1) {
    int_tmp_val <- as.integer(df_inputData[row_idx, colnames(df_inputData) %in%_
→vec_colnames_search_2[5], with=FALSE])
    int_tmp_val <- int_tmp_val * -1 - 1</pre>
    df_outputData[row_idx, vec_colnames_out[3]] <- int_tmp_val</pre>
  }
  else if (df_inputData[row_idx, colnames(df_inputData) %in%_
→vec_colnames_search_1[3], with=FALSE] == -1) {
    df_outputData[row_idx, vec_colnames_out[3]] <- 1</pre>
  else if (df_inputData[row_idx, colnames(df_inputData) %in%_
ovec_colnames_search_1[3], with=FALSE] == 2) {
    int_tmp_val <- as.integer(df_inputData[row_idx, colnames(df_inputData) %in%_
→vec_colnames_search_2[6], with=FALSE])
    int_tmp_val <- int_tmp_val + 1</pre>
```

```
df_outputData[row_idx, vec_colnames_out[3]] <- int_tmp_val
}

# return scrambled data frame
return(df_outputData)
}</pre>
```

#### 3.3 Function for writing resulting data frame to CSV file

```
[7]: func_writeDataframe_to_CSVfile <- function(str_path, str_CSVfilename, df_dataframe, u
      ⇔str filenameExtension) {
       # Split file name on second underscore, found here:
      # https://stackoverflow.com/questions/32398427/
      \neg r-split-a-character-string-on-the-second-underscore/32398489#32398489
      list_str_split <- strsplit(sub('(^[^_]+_[^_]+)_(.*)$', '\\1 \\2',_
      str_CSVfilename), ' ')
       # extend the file name prefix and glue together with old suffix
       str_CSVfilename_extended <- paste(list_str_split[[1]][1], str_filenameExtension,_
      ⇔list_str_split[[1]][2], sep="_")
       # extend file name by path
       str_CSVfilename_extended <- paste(str_path, str_CSVfilename_extended, sep="/")
       write.table(df_dataframe, file = str_CSVfilename_extended,
                   fileEncoding = "UTF-8", row.names = FALSE,
                   col.names = TRUE, sep = "\t", quote = TRUE)
     }
```

## 4 Manipulate the data and store in new CSV files for each criteria

#### 4.1 Environmental sub-criteria

Walk over all input CSV files, manipulate the data, and write the results to output CSV files:

#### 4.2 Social sub-criteria

Walk over all input CSV files, manipulate the data, and write the results to output CSV files:

#### 4.3 Economic sub-criteria

Walk over all input CSV files, manipulate the data, and write the results to output CSV files:

#### 4.4 Criteria (main criteria)

Walk over all input CSV files, manipulate the data, and write the results to output CSV files:

```
[11]: vec_colnames_search_1 <- c('AK01', 'AK02', 'AK03')
vec_colnames_search_2 <- c('RK01_01', 'RK02_01', 'RK03_01', 'RK04_01', 'RK05_01',

$\times' \text{RK06_01'} \\
vec_colnames_out <- c('Envi_Soci', 'Envi_Econ', 'Soci_Econ')}
```

### 5 Summary and outlook

[]:

#### 6 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).