# Abstract

We face a deluge of data that scientists need to deal with in many different domains today. While there are already good solutions to some parts of the data life cycle the applicability of the solutions to certain scientific domains often varies. Especially research domains with high degree of interdisciplinary interactions and heterogeneity in methods and data in general like ecology face problems in dealing with some valuable concepts like ontologies that potentially can be used to improve or automate some of the most common tasks in analyses like finding relevant data, cleaning and merging of datasets. We here introduce the rbefata package that connects to the open source data management platform BEFdata that has been developed and is used within the BEF-China experiment. We show the use of the package in combination with the portal using an example workflow that integrates three datasets from the BEF-China experiment representing an analysis that has been published already. We discuss the combination of the R package rbefdata and the data poral in the context of state of the art data management as well as we give an outlook on upcoming features that will bring semantical featues like smart merges based on an ontology we created.

# Introduction

With a growing awareness on the long term value of data, much effort has been put into building data management platforms, to preserve all kind of environmental and historic data, over the last years (e.g. diversity workbench, BEFdata). Many specialized solutions for different scientific disciplines appeared that provide data management plans for small scale projects or collaborations as well as for large data producing long term or remote sensing projects. An ongoing trend in that context is the development of integrative databases or data portals. They serve as nodes that collect data from smaller databases of a certain domain and they give researchers of that domain the opportunity to access a wide range of relevant data from one place. This portals in fact offer a solution to to one of the most pressing problems that we face with our valuable data today, their lost.

Another big problem with data, especially in terms of reuse of available data, is the general understanding of datasets. Usually plain datasets say nothing, to one who is not familiar with it and they are even hard to decipher by the author itself after some time has passed. It is usually hard to remember exactly what methods have been used to collect a certain columns data or what the abbreviations or headers in the dataset mean. To solve this this problem metadata frameworks have been developed and published as standards so nobody really needs to think about an own set of requirements to describe its data. The Ecological Metadata Language is only one example for that. While this theoretically solves the problem with not well described datasets it is still hard to make people use it extensively as this usually always means to learn new tools that help with the

description process (e.g morpho, data up).

While well described data helps a lot in understanding datasets and on deciding upon the relevance and applicability in a certain analysis there is still lots of manual intervention necessary after that to prepare the data for analysis (cite yourself? or xxx). It may needs to be cleaned, imputed, reshaped and merged which usually takes up to 70% of the analysis workflow, before the smart models can be applied to the data to find interesting patters (cite the workflow paper of Karin and me). This preparation steps not only are time and labour intensive but also potentially error prone, especially as the complexity of analyses grows.

Ontologies, formal representations of knowledge potentially offer a sophisticated tool to deal with that step of data preparation (cite supporting ecology as data intensive science). While they are already used in some research domains like genetics (cite xxx), other domains face more problems using it. For example in ecology, that has grown into a very collaborative, interdisciplinary and data intensive science over the last decade, to address questions on a greater temporal and spatial scale (e.g michener et al 2012). The data here is mainly provided by small scale studies spread all over the world (e.g heidorn2009 shedding light on the dark) but also through bigger long term projects like LTER (cite xxx), BEF-China (cite xxx), governmental projects and local initiatives (cite xxx). This in fact results in a wild growing, complex and heterogeneous data landscape in that we need to deal with. The application of ontologies in ecology is discussed controversially (cite xxx) which is mainly related to the heterogeneity of the research domain and it is argued that they can be a benefit, but it is hard to set up a sophisticated ontology covering all necessary terms and relation of a that complex research domain like ecology.

As there is a growing demand to use and reuse available data and to embed small heterogeneous data into a wider context in ecology we here introduce the R package rbefdata that in combination with BEFdata exactly deals with that. We showcase the functionality of the package available with version 0.3.5 creating a workflow for integration of three datasets and discuss the rbefdata package and BEFdata in the light of future developments on the integration of ontologies that will make finding data, smart merges and unit conversion possible to help researchers to deal with the upcoming challenges in dealing with data like integration of heterogeneous datasets.

# Material and Methods

## BEF-China and the BEFdata portal

The BEF-China experiment is a Biodiversity Ecosystem Functioning (BEF) experiment funded by the German science foundation (DFG, FOR 891). It is located in the subtropics of China in the provinces Jianxi and Zhejiang. The BEF-China research group (www.bef-china.de) uses two main research platforms.

An experimental forest diversity gradient of 50~ha, and 27 observational plots of 30x30 m each located in the Gutianshan Nature Reserve. The observational plots were selected according to a crossed sampling design along tree species richness and stand age. The data for the workflow on carbon pools stems from 22 to 116 years consisting of 14 to 35 species (cite Bruelheide, 2010).

The BEFdata portal is an open source data management platform developed within the BEF-China project. It adheres to standards like the Ecological Metadata Language for describing datasets with metadata and is specialized in harmonizing small heterogeneous data that usually has to be dealt with in BEF. But its specialization makes it also very valuable to use in any other scientific domain that needs to deal with complex small and heterogeneous data.

The portal offers a social component where researchers can shop datasets and write a paper proposals based on the datasets in the shopping cart. In the process of creating a proposal some information like a title, a rationale, an envisaged journal and date needs to be provided. Sending in a proposal a researcher asks for access to the datasets and provides the data owners with necessary information about the paper. The data owners then can decide if and how they like to participate in the upcoming paper or if they only like to get acknowledged for providing their data (cite Karin).

### The proposal

We use an already published dataset as an example to present the functionalities and interlinkages between the BEF China dataportal and rbefdata. To test the effect of species richness on system N retention and tree sapling N uptake we conducted a 15N tracer experiment in a young tree plantation. To this end, saplings of four abundant early successional tree species have been planted in monocultures, in two- and four-species mixtures, and as individual trees. Afforestations are increasing globally to produce timber and pulp wood, but also to enhance ecosystem services such as carbon sequestration, nutrient retention, or groundwater recharge. In order to further optimise these services with regard to balanced nutrient (particularly nitrogen) cycles, it is important to know whether the use of mixtures of native tree species in afforestation projects promotes greater acquisition and retention of nitrogen compared to the currently established large-scale monocultures. Experimental design and data acquisition Four species were chosen for the experiment: Schima superba Gardn. et Champ. and Elaeocarpus decipiens Hemsley (evergreen), Quercus serrata Murray and Castanea henryi (Skan) Rehd. et Wils. (deciduous; Yu et al. 2001). The following planting schemes were established in 1-mÂ<sup>2</sup> plots. In each plot 16 saplings were planted in an array of four by four. Monocultures, two-species combinations and four-species combinations were established. The four study species provided a total of eleven species combinations four monocultures, six two-species combinations, and one four-species combination. All treatments were replicated four times, once in each of the four blocks. Pulse labelling with

15NH415NO3 (98% 15N) was performed in August and in September 2009. Leaf, fine root and soil samples have been collected in September 2010. Samples have been analysed for 15N content and leaf, fine root and soil recovery have been calculated. The sum of the three compartement recoveries is referred to as system N retention. Relative leaf, root and soil recovery was calculated as percentage of system N retention (for a detailed description of the material and methods we refer to Lang et al. 2013).

• figure shows the proposal page



Figure 1: showcase\_proposal

#### rbefdata

The rbefdata package started its development within the BEF-Cina experiment. Meanwhile it is part of the rOpenSci package porfolio (http://ropensci.org/), which is a community driven approach to wrap all science APIs and to create solutions to pull data from different repositories into R for analysis.

The package can be installed from the CRAN package repository (https://github.com/befdata/befdata) and enables access to the data, meta data structures of the platform and provides convenient methods to pull single or multiple dataset into the R environment in one step for analysis.

Additionally it offers functions that help to upload final results datasets with the script attached that has been used to derive the results from the original datasets which provides a valuable insight into data provenance and also is a stepping stone for reproducible research.

# Usecase (results)

The next step after the an accepted paper proposal is to setup the rbefdata package. This requires loading the package and setting the required package options. Having a look into the options list reveals several fields that can be filled in, like the URL to the BEFdata server, user credentials and a download folder name that is used to store downloaded freeformat files. The tematres server related URLs in the options are part of upcomming features that are non fully functional on the time of writing.

The most essential setting for the workflow we present here is the user credentials. These are used to authenticate the user against the portal to ensure the access to the data has been granted and to log the data access. Setting the URL is not required here as it defaults to the BEF-China project instance of the BEFdata portal that we retrieve data from. If one has set up an own instance of the BEFdata portal, this URL needs to be changed so the package communicates with the right server (see box below).

```
require(rbefdata)
# options list
bef.options()
## [1] "http://china.befdata.biow.uni-leipzig.de"
##
## $tematres url
## [1] "http://tematres.befdata.biow.uni-leipzig.de/vocab/index.php"
##
## $tematres service url
## [1] "http://tematres.befdata.biow.uni-leipzig.de/vocab/services.php"
## $download_dir
## [1] "downloads"
##
## $user_credentials
## [1] ""
# querry single options
bef.options("url")
```

```
## [1] "http://china.befdata.biow.uni-leipzig.de"

# set credentials example
bef.options(user_credentials = "aölkjspoiul12")

# set URL example
bef.options(url = "http://my.own.befdata.instance.com")
```

After the setup of the rbefdata package we can start right away using data from the proposal. A (proposal) download function is used for that which draws all associated datasets of the proposal into the R environment in one single step (see blow). The functin reuires the ID of the proposal that can be found in the URL of the proposal on the BEFdata portal (see blow).

```
# the id is 90
http://befdataproduction.biow.uni-leipzig.de/paperproposals/90
# proposal id is
datasets = bef.get.datasets_for_proposal(id = 90)
extract_one_dataset = datasets[[1]]
• Inspect datasets
```

• after download (attributes())

• in general bef.get.metadata\_for(dataset = id)

As the BEFdata portal provides metadata via EML standard, we also make use of that within rbefdata. Each dataset on download is associated with the metadata provided by its authors. This information can be extracted using the built-in R function attributes() (see code box below). As this requires granted access rights to the dataset, there is also a function especially to only draw the metadata which is always free for download.

```
# extract title of one dataset
attributes(datasets[[1]])$title

## [1] "Competition of saplings for N -Pilot- system 15N retention"

# for all dataset in the proposal
titles = sapply(datasets, function(x) attributes(x)$title)
titles

## [1] "Competition of saplings for N -Pilot- system 15N retention"
## [2] "Plottreatment and -location within the blocks of the Pilot-Experiment"
```

```
# other metadata available
names(attributes(datasets[[1]]))
    [1] "names"
                                    "class"
    [3] "row.names"
                                    "title"
##
   [5] "abstract"
                                   "publicationDate"
   [7] "language"
                                    "creators"
   [9] "authors"
                                    "intellectualRights"
##
## [11] "distribution"
                                    "keywords"
## [13] "generalTaxonomicCoverage" "samplingDescription"
## [15] "spatial_coverage"
                                    "temporal_coverage"
                                    "columns"
## [17] "related_material"
# only get metadtata by dataset id
bef.portal.get.metadata(dataset = 335)$title
## [1] "Competition of saplings for N -Pilot- system 15N retention"
```

- both datasets contain a plot\_id that can be used for merge
  - extract into separate datasets
  - We need to merge them two by two.
  - to analyse the dataset of system N retention we do need information about species diversity of the plots and the information about which plot is placed in which block from the design dataset. Furthermore we need values of the initial basal diameter from the dataset Nrecov

```
[1] "block"
                                     "x"
##
    [3] "y"
##
                                     "plot_id"
   [5] "control_ID"
                                     "block_community_code"
    [7] "community_number"
                                     "species_mixture"
##
##
   [9] "species_diversity"
                                     "species_pool"
## [11] "species_code"
                                     "research_group_colour"
## [13] "control"
                                     "closed_canopy"
## [15] "density"
                                     "Natives"
## [17] "depth"
                                     "harvest"
## [19] "fungicide"
                                     "inoculation"
                                     "native"
## [21] "pesticide"
## [23] "genetic_diverstiy"
                                     "seed addition"
                                     "plot_treatment_connected"
## [25] "fertilizer"
## [27] "sp1"
                                     "sp2"
## [29] "sp3"
                                     "sp4"
## [31] "sp5"
                                     "sp7"
## [33] "sp8"
                                     "sp11"
## [35] "sp_connected"
```

• However, this synthesis dataset now includes many columns that we do not need. Thus they will be excluded in a synthesis dataset we create.

```
syndata = merge(Nretention, design)
# overview about the content of the synthesis dataset
names(syndata)
##
    [1] "plot_id"
                                     "recov_plot"
##
    [3] "perleaf_plot"
                                     "perroot_plot"
##
                                    "persoil_plot"
   [5] "perbio_plot"
   [7] "gbd T0.mm."
                                    "block"
   [9] "x"
                                    "v"
##
                                     "block_community_code"
## [11] "control_ID"
## [13] "community_number"
                                    "species_mixture"
## [15] "species_diversity"
                                     "species_pool"
## [17] "species_code"
                                     "research_group_colour"
## [19] "control"
                                     "closed_canopy"
## [21] "density"
                                    "Natives"
## [23] "depth"
                                    "harvest"
## [25] "fungicide"
                                     "inoculation"
## [27] "pesticide"
                                     "native"
## [29] "genetic_diverstiy"
                                    "seed addition"
## [31] "fertilizer"
                                     "plot_treatment_connected"
                                     "sp2"
## [33] "sp1"
```

# the synthesis dataset

```
## [35] "sp3"
                                  "sp4"
## [37] "sp5"
                                  "sp7"
## [39] "sp8"
                                  "sp11"
## [41] "sp_connected"
syndata = syndata[-c(9:14, 16:41)]
names(syndata)
## [1] "plot_id"
                          "recov_plot"
                                              "perleaf_plot"
## [4] "perroot_plot"
                          "perbio_plot"
                                              "persoil_plot"
## [7] "gbd_T0.mm."
                          "block"
                                              "species_diversity"
### check for data properties (keep?)
str(syndata)
## 'data.frame': 42 obs. of 9 variables:
## $ plot_id
                    : Factor w/ 42 levels "pilot1D01", "pilot1D02", ...: 1 2 3 4 5 6 7 8 9 :
## $ recov_plot
                    : num 15.64 13.03 20.29 9.6 8.35 ...
## $ perleaf_plot : num 2.92 5.52 1.06 4.19 15.1 ...
## $ perroot_plot : num 1.001 1.369 0.516 1.086 5.098 ...
## $ perbio_plot
                     : num 3.92 6.89 1.57 2.03 17.96 ...
## $ persoil_plot
                     : num 96.1 93.1 98.4 98 82 ...
## $ gbd_T0.mm.
                     : num NA 3.69 2.88 6 7 ...
## $ block : Factor w/ 8 levels "1","2","3","4",..: 1 1 1 1 1 1 1 1 2 ...
## $ species_diversity: int 1 1 1 2 2 2 2 2 4 1 ...
### we want to use 'species_diversity' as a factor
syndata$species_diversity = as.factor(syndata$species_diversity)
attach(syndata)
### check response variables for normality and transform
# qqnorm(recov_plot^.5) qqline(recov_plot^.5,lty=2)
syndata$recov_plot_t = syndata$recov_plot^0.5
# qqnorm(perleaf_plot^.5) qqline(perleaf_plot^.5,lty=2)
syndata$perleaf_plot_t = syndata$perleaf_plot^0.5
### All response variables have been square root transformed
syndata$perroot_plot_t = syndata$perroot_plot^0.5
syndata$perbio_plot_t = syndata$perbio_plot^0.5
```

```
syndata$persoil_plot_t = syndata$persoil_plot^0.5
### round response varibles
syndata$recov_plot_t = round(syndata$recov_plot_t, digits = 2)
syndata$perleaf_plot_t = round(syndata$perleaf_plot_t, digits = 2)
syndata$perroot_plot_t = round(syndata$perroot_plot_t, digits = 2)
syndata$persoil_plot_t = round(syndata$persoil_plot_t, digits = 2)
attach(syndata)
## The following objects are masked from syndata (position 3):
##
##
       block, gbd_T0.mm., perbio_plot, perleaf_plot, perroot_plot,
##
       persoil_plot, plot_id, recov_plot, species_diversity
  • we want to analyse our data by linear mixed effects models. Since our
     plots are clumbed in space, we use block as random factor we will use the
    we will use the packages (nlme) for model analysis and (multcomp) for
     post-hoc comparisons
## Loading required package: mvtnorm Loading required package: survival
## Loading required package: splines Loading required package: MASS Loading
## required package: nnet
#### Model one Overall recovery/N retention
model1 = lme(recov_plot_t ~ gbd_T0.mm. + species_diversity, syndata, random = ~1 |
    block, na.action = na.omit, method = "REML")
anova(model1)
##
                     numDF denDF F-value p-value
## (Intercept)
                         1
                              34
                                    871.4 < .0001
## gbd_T0.mm.
                         1
                               34
                                      7.5 0.0098
## species_diversity
                         2
                              34
                                      2.9 0.0708
summary(glht(model1, linfct = mcp(species_diversity = "Tukey")))
##
##
     Simultaneous Tests for General Linear Hypotheses
## Multiple Comparisons of Means: Tukey Contrasts
##
##
## Fit: lme.formula(fixed = recov_plot_t ~ gbd_T0.mm. + species_diversity,
       data = syndata, random = ~1 | block, method = "REML", na.action = na.omit)
##
```

```
##
## Linear Hypotheses:
             Estimate Std. Error z value Pr(>|z|)
## 2 - 1 == 0
                                  -1.51
               -0.378
                           0.251
                                            0.279
## 4 - 1 == 0
                0.479
                           0.420
                                    1.14
                                            0.480
## 4 - 2 == 0
                0.858
                           0.398
                                    2.15
                                            0.076 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Adjusted p values reported -- single-step method)
## To adjust for the unbalanced experimental design, Anova Type II (package
## "car" (Fox & Weisberg 2011)) was used to test for main effects
model1c = Anova(model1, type = "II")
model1c
## Analysis of Deviance Table (Type II tests)
##
## Response: recov_plot_t
##
                    Chisq Df Pr(>Chisq)
## gbd T0.mm.
                     7.42 1
                                 0.0065 **
## species_diversity 5.73 2
                                 0.0570 .
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# vizual check plot(model1,resid(.)~fitted(.))
# plot(model1,recov_plot_t~fitted(.))
#### Model2 percentage leaf recovery of plot recovery
model2 = lme(perleaf_plot_t ~ species_diversity, syndata, random = ~1 | block,
   method = "REML")
anova(model2)
##
                    numDF denDF F-value p-value
                             36 273.07 <.0001
## (Intercept)
                        1
                                   6.55 0.0037
## species_diversity
                        2
                             36
summary(glht(model2, linfct = mcp(species_diversity = "Tukey")))
##
##
     Simultaneous Tests for General Linear Hypotheses
```

```
##
## Multiple Comparisons of Means: Tukey Contrasts
##
##
## Fit: lme.formula(fixed = perleaf_plot_t ~ species_diversity, data = syndata,
      random = ~1 | block, method = "REML")
##
##
## Linear Hypotheses:
             Estimate Std. Error z value Pr(>|z|)
## 2 - 1 == 0
              1.052
                           0.293
                                    3.59
                                          <0.001 ***
## 4 - 1 == 0
              0.865
                           0.497
                                    1.74
                                             0.18
## 4 - 2 == 0 -0.187
                                   -0.39
                                             0.92
                           0.479
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Adjusted p values reported -- single-step method)
Anova(model2, type = "II")
## Analysis of Deviance Table (Type II tests)
##
## Response: perleaf_plot_t
                    Chisq Df Pr(>Chisq)
## species_diversity 13.1 2
                                 0.0014 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# vizual check plot(model2,resid(.)~fitted(.))
# plot(model2,perleaf_plot_t~fitted(.))
### Model3 percentage root recovery of overall recovery
model3 = lme(perroot_plot_t ~ species_diversity, syndata, random = ~1 | block,
   method = "REML")
anova(model3)
##
                    numDF denDF F-value p-value
## (Intercept)
                        1
                             36
                                 373.7 <.0001
                                    7.1 0.0024
## species_diversity
                        2
                             36
summary(glht(model3, linfct = mcp(species_diversity = "Tukey")))
##
```

```
##
     Simultaneous Tests for General Linear Hypotheses
##
## Multiple Comparisons of Means: Tukey Contrasts
##
##
## Fit: lme.formula(fixed = perroot_plot_t ~ species_diversity, data = syndata,
      random = ~1 | block, method = "REML")
##
## Linear Hypotheses:
             Estimate Std. Error z value Pr(>|z|)
## 2 - 1 == 0
                0.600
                           0.170
                                    3.53
                                            0.001 **
## 4 - 1 == 0
                0.731
                           0.289
                                            0.029 *
                                    2.53
## 4 - 2 == 0
                           0.278
                0.130
                                    0.47
                                            0.883
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Adjusted p values reported -- single-step method)
Anova(model3, type = "II")
## Analysis of Deviance Table (Type II tests)
##
## Response: perroot_plot_t
                    Chisq Df Pr(>Chisq)
## species_diversity 14.3 2
                                0.00079 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# vizual check plot(model3,resid(.)~fitted(.))
# plot(model3,perleaf_plot_t~fitted(.))
### Model 4 percentage soil recovery of overall recovery
model4 = lme(persoil_plot_t ~ species_diversity, syndata, random = ~1 | block,
   method = "REML")
anova(model4)
                    numDF denDF F-value p-value
## (Intercept)
                        1
                             36
                                  26200 < .0001
## species_diversity
                        2
                             36
                                      4 0.0274
summary(glht(model4, linfct = mcp(species_diversity = "Tukey")))
```

```
##
##
    Simultaneous Tests for General Linear Hypotheses
##
## Multiple Comparisons of Means: Tukey Contrasts
##
##
## Fit: lme.formula(fixed = persoil_plot_t ~ species_diversity, data = syndata,
      random = ~1 | block, method = "REML")
##
##
## Linear Hypotheses:
             Estimate Std. Error z value Pr(>|z|)
              -0.294
## 2 - 1 == 0
                           0.127 - 2.32
                                            0.050
## 4 - 1 == 0
              -0.501
                                   -2.33
                                            0.049 *
                           0.215
## 4 - 2 == 0
              -0.207
                           0.207 -1.00
                                            0.570
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Adjusted p values reported -- single-step method)
Anova(model4, type = "II")
## Analysis of Deviance Table (Type II tests)
## Response: persoil_plot_t
                    Chisq Df Pr(>Chisq)
## species_diversity 7.96 2
                                  0.019 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# vizual check plot(model4,resid(.)~fitted(.))
# plot(model4,perleaf_plot_t~fitted(.))
```

# Discussion

- General discussion
- high need to effective use/reuse data
- relevant data needs to be simply detectable
- BEFdata and rbefdata
- in combination provide a solution to
  - data storage

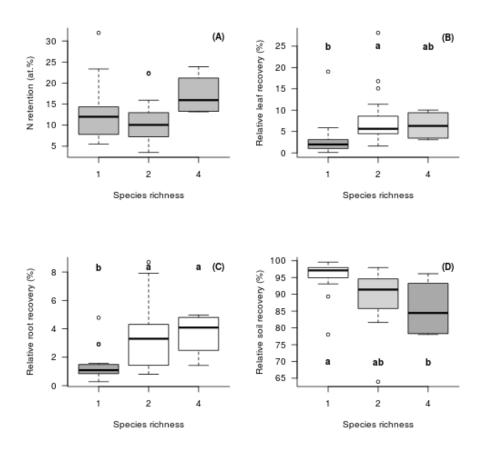


Figure 2: plot of chunk final\_plot

- describing data with metadata
- collaboration and data sharing
- simply pull data into analysis software and push data back
- data provenance by attaching R scripts to uploads
- will provide solution with next versions
  - easier finding relevant data
  - smart merges (including unit conversions)

As there is a growing demand to effectively reuse available data this puts much pressure on the development of solutions that help researchers not only to find but also to integrate heterogeneous small data into a wider context of in different analyses (cite xxx, data intensive science, long tail). The combination of BEFdata and the rbefdata package provides a solutions to a one part of the data life cycle and especially introduces a solution to deal with high heterogeneous data.

We recently stared to develop an ontology using a tematres server containing knowledge extracted from portals that deal with data management for ecological research. The tematres server offers an API so all the contained terms can be accessed by the upcoming version of rbefdata

The formalization developed is and will be based on the knowledge used in biodiversity research. Thus we will here discuss the software combination BEFdata and rbefdata in the light of the upcoming features and in general context state of the art data management today. In one of the next versions to be rolled out the BEFdata portal will get a semantical annotation feature. This will give admins and data admins the ability to tag each column of datasets with a general term that best describes the content. So the field will contain top terms of the ontology. The tagging will be reflected in the API and can thus be simply queried to use the information within the R package. Using the knowledge about the content of a column in the R package will enable us to do support smart merges that work

tematres (homepage) into BEFdata and the rbefdata package so they play well together semantically.

### Acknowledgements

Thanks to all the data owners of the proposal for providing access to the datasets.

### Literature

This will be done externally as markdown has no good way to deal with references and stuff. We can collect here a list of links to references. I will then collect them with e.g zotero to export a format we can hand them in for publication.

# Appendix

• maybe will not be used extensively but we will see

### **Figures**

• vizualization plugin (keywords)

One can visualize the keywords associated with the dataset of a BEFdata portal using the vitalization functionality. This gives a short overview about the contents the portal data is dealing with.

```
## Warning: BEF conference 2011 could not be fit on page. It will not be
## plotted. Warning: replanted species could not be fit on page. It will not
## be plotted. Warning: Gram-negative bacteria could not be fit on page. It
## will not be plotted. Warning: Gram-positive bacteria could not be fit on
## page. It will not be plotted. Warning: shannon diversity could not be fit
## on page. It will not be plotted. Warning: tree performance could not be
## fit on page. It will not be plotted. Warning: wood perforation plates
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## wavelength 214nm could not be fit on page. It will not be plotted.
## Warning: cadmium at wavelength 228nm could not be fit on page. It will not
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## will not be plotted. Warning: experimental design could not be fit on
## page. It will not be plotted. Warning: gene diversity could not be fit on
## page. It will not be plotted. Warning: geomorphology could not be fit on
## page. It will not be plotted. Warning: leaf physical resistance could not
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## diversity could not be fit on page. It will not be plotted. Warning:
## response variable could not be fit on page. It will not be plotted.
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## plotted. Warning: standard deviation could not be fit on page. It will not
## be plotted. Warning: tree identifier could not be fit on page. It will not
## be plotted. Warning: wood bending could not be fit on page. It will not be
## plotted. Warning: wood shearing could not be fit on page. It will not be
## plotted. Warning: wood shrinkage could not be fit on page. It will not be
## plotted. Warning: wood stretching could not be fit on page. It will not be
## plotted. Warning: aboveground biomass could not be fit on page. It will
## not be plotted. Warning: aeromorphic organic layer could not be fit on
## page. It will not be plotted. Warning: basal area increment could not be
## fit on page. It will not be plotted. Warning: branch water potential could
## not be fit on page. It will not be plotted. Warning: cavity nesting
## hymenoptera could not be fit on page. It will not be plotted. Warning:
## coarse root density could not be fit on page. It will not be plotted.
```

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## Warning: community similarity could not be fit on page. It will not be
## plotted. Warning: community weighted mean trait could not be fit on page.
## It will not be plotted. Warning: crown projection area could not be fit on
## page. It will not be plotted. Warning: ecosystem functioning could not be
## fit on page. It will not be plotted. Warning: experimental treatment could
## not be fit on page. It will not be plotted. Warning: forest canopy could
## not be fit on page. It will not be plotted. Warning: functional trait
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## could not be fit on page. It will not be plotted. Warning: phylogenetic
## distinctness could not be fit on page. It will not be plotted. Warning:
## phytophagous insects could not be fit on page. It will not be plotted.
## Warning: position could not be fit on page. It will not be plotted.
## Warning: rainfall simulator could not be fit on page. It will not be
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## will not be plotted. Warning: Weibull distribution could not be fit on
## page. It will not be plotted. Warning: wood ground tissue could not be fit
## on page. It will not be plotted. Warning: wood mechanics could not be fit
## on page. It will not be plotted. Warning: wood porosity could not be fit
## on page. It will not be plotted.
```

### Tables

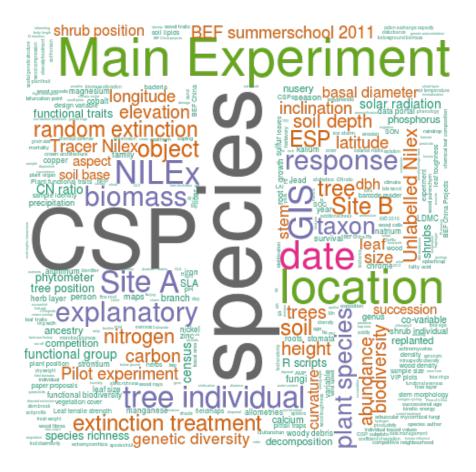


Figure 3: plot of chunk vizalize\_keywords