

Package ‘optimLanduse’

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Title Robust Land-Use Optimization

Version 2.0.0

Description Robust multi-criteria land-allocation optimization that explicitly accounts for the uncertainty of the indicators in the objective function. Solves the problem of allocating scarce land to various land-use options with regard to multiple, coequal indicators. The method aims to find the land allocation that represents the indicator composition with the best possible trade-off under uncertainty. optimLanduse includes the actual optimization procedure as described by Knoke et al. (2016) <[doi:10.1038/ncomms11877](https://doi.org/10.1038/ncomms11877)> and the post-hoc calculation of the portfolio performance as presented by Gosling et al. (2020) <[doi:10.1016/j.jenvman.2020.110248](https://doi.org/10.1016/j.jenvman.2020.110248)>.

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Imports lpSolveAPI (>= 5.5.2.0-17.7), tidyr (>= 1.1.2), dplyr (>= 1.0.0), future.apply (>= 1.9.0)

Suggests readxl, ggplot2

URL <https://github.com/Forest-Economics-Goettingen/optimLanduse/>

NeedsCompilation no

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autoSearch	<i>Optimize all possible indicator combinations</i>
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Description

This iterative search function generates a list of all possible indicator combinations. All indicator combinations are converted into a list format, where each combination corresponds to a list entry. For each of these list entries, an optimization is performed using the *initScenario* and *solveScenario* functions of the package. How these functions work in detail (incl. example code) can be seen in the help of the respective function in the package and the README from Husmann et al. (2022). The results are entirely saved into the respective list entry. In addition, each entry is appended with the currently observed land-use portfolio and the land-use portfolio when all indicators are optimized together. Out of this list, we use the Bray-Curtis measure of dissimilarity to identify the indicators driving current land-use decisions. An example and further explanation is given in the README [GitHub project page](#)

Usage

```
autoSearch(  
  coefTable,  
  landUseObs,  
  uValue = 1,  
  optimisticRule = "expectation",  
  fixDistance = 3  
)
```

Arguments

coefTable	Coefficient table in the expected <i>optimLanduse</i> format.
landUseObs	Data frame with two columns. The first column has to contain the land-use options. The second column the respective shares.
uValue	<i>u</i> Value. The uncertainty value delivered in the coefTable is multiplied with this <i>u</i> value. The value, therefore, enables scenario analyses with differing uncertainties in relation to indicator values. Higher <i>u</i> values can be interpreted as a higher risk aversion of the decision maker.

calcPerformance	<i>Attach portfolio performance and distance to target</i>
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Description

The Portfolio performances are calculated and attached to the solved `optimLanduse` object. Each performance measure describes the relative proportion to the maximum achievable (the "target") of the indicator, given the current land use distribution and the uncertainty scenario set. The lowest performing scenario of all indicators is the degree of minimal fulfillment under the worst-possible outcome. It can thus be interpreted as the guaranteed performance. At least this proportion will be achieved across all indicators.

Usage

```
calcPerformance(x)
```

Arguments

`x` An optimized `optimLanduse` object.

Details

For further information and calculation, see the supplement of Gosling et al. (2020), Formula S5 (in the supplement of the paper) and also the paragraph `optimLanduse` functions and workflow - Post-processing in Husmann et al. (2022).

Value

An optimized `optimLanduse` object with attached portfolio performance.

References

Gosling, E., Reith, E., Knoke T., Gerique, A., Paul, C. (2020): Exploring farmer perceptions of agroforestry via multi-objective optimisation: a test application in Eastern Panama. *Agroforestry Systems* **94**. doi:[10.1007/s10457020005190](https://doi.org/10.1007/s10457020005190)

Husmann, K., von Groß, V., Bödeker, K., Fuchs, J. M., Paul, C., & Knoke, T. (2022). `optimLanduse`: A package for multiobjective land-cover composition optimization under uncertainty. *Methods in Ecology and Evolution*, 00, 1– 10. <https://doi.org/10.1111/2041-210X.14000>

Examples

```
require(ggplot2)
require(readxl)

dat <- read_xlsx(exampleData("exampleGosling.xlsx"))
init <- initScenario(dat, uValue = 2,
                     optimisticRule = "expectation",
                     fixDistance = 3)
```

```

result <- solveScenario(x = init)
performance <- calcPerformance(result)

# Visualize the distance

ggplot(performance$scenarioTable,
       aes(x = indicator,
           y = performance,
           color = indicator)) +
geom_point() +
geom_hline(yintercept =
           min(performance$scenarioTable$performance),
           linetype = "dashed", color = "red") +
ylim(0, 1)

```

dataPreparation

*Transform data into the expected format***Description**

The input data must suit the specific expected optimLanduse format prior to initialization and optimization. This function provides the possibility to easily transform data from the commonly used form of the exemplary data [exampleData](#) into the expected format. Please consider that the application of this function is not mandatory and in most cases not required. Best practice is to transform your data yourself into the expected format. Detailed information about the expected format and possible data processing can be found on the [GitHub project page](#). Note that incomplete rows, which include NA-values will be deleted and an error message will be thrown.

Usage

```
dataPreparation(dat, uncertainty = "SE", expVAL = "mean")
```

Arguments

dat	Data frame or tibble in the format of the exampleData . Please refer to the GitHub project page for more details.
uncertainty	Indicates the column name of the uncertainty measure. Typical is "SE" for standard error or "SD" for standard deviation.
expVAL	Indicates the column name of the expected value.

Value

A formatted coefficients table with land-use options and indicator values ready for initialization via [initScenario](#).

References

Gosling, E., Reith, E., Knoke, T. et al. Exploring farmer perceptions of agroforestry via multi-objective optimisation: a test application in Eastern Panama. *Agroforest Syst* 94, 2003–2020 (2020). <https://doi.org/10.1007/s10457-020-00519-0>

Examples

```
require(readxl)
dat <- read_xlsx(exampleData("exampleGosling_dataPrep.xlsx"), col_names = TRUE)
dat <- dataPreparation(dat, uncertainty = "sd", expVAL = "mean")
```

exampleData

Exemplary data in the required format

Description

optimLanduse comes bundled with exemplary data for land-use optimization. The files can also be found on your computer in the package folder `./extdata`. These examples provide some quick applications of the package for demonstration and an example of the expected data structure of the data. Consider also the [GitHub project page](#) for exemplary applications of the package.

Usage

```
exampleData(fileName = "exampleGosling.xlsx")
```

Arguments

fileName	Name of the example file. See 'details' section for further explanation of all provided examples.
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Details

exampleGosling.xlsx contains the freely available data from Gosling et al. (2020). *exampleEmpty.xlsx* contains a template for your data.

Value

The path to the example file on your computer.

References

Gosling, E., Reith, E., Knoke, T. et al. Exploring farmer perceptions of agroforestry via multi-objective optimisation: a test application in Eastern Panama. *Agroforest Syst* 94, 2003–2020 (2020). <https://doi.org/10.1007/s10457-020-00519-0>

Examples

```
require(readxl)
path <- exampleData()
read_xlsx(path, col_names = FALSE)
path <- exampleData("exampleGosling.xlsx")
read_xlsx(path, col_names = FALSE)
```

initScenario	<i>Initialize the robust optimization</i>
--------------	---

Description

The function initializes an *optimLanduse* S3 object on the basis of a coefficients table. Please note that the coefficients table must follow the expected *optimLanduse* format. The expected format is explained in the example on the [GitHub project page](#) and in the publication in *Methods in Ecology and Evolution* (Husmann et al. ,2022)

Usage

```
initScenario(
  coefTable,
  uValue = 1,
  optimisticRule = "expectation",
  fixDistance = 3
)
```

Arguments

coefTable	Coefficient table in the expected <i>optimLanduse</i> format.
uValue	<i>u</i> Value. The uncertainty value delivered in the coefTable is multiplied with this <i>u</i> value. The value, therefore, enables scenario analyses with differing uncertainties in relation to indicator values. Higher <i>u</i> values can be interpreted as a higher risk aversion of the decision maker.
optimisticRule	Either <i>expectation</i> or <i>uncertaintyAdjustedExpectation</i> . The rule indicates whether the optimistic outcomes of an indicator are directly reflected by their expectations or if the indicator is calculated as expectation + uncertainty when "more is better" or expectation - uncertainty respectively when "less is better". An optimization based on <i>expectation</i> considers only downside risks.
fixDistance	This optional numeric value allows to define distinct uncertainty levels for the calculation of the uncertainty space and the averaged distances of a certain land-cover composition (see Equation 9 in Husmann et al. (2020)). Passing NA disables fixDistance. In this case, the uncertainty space is defined by uValue.

Details

Separating the initialization from the optimization is to save computation time in batch analysis. The separated function calls allow the user to perform multiple optimizations from one initialized object. This could save time in the scenario or sensitivity analysis.

A detailed description of the input parameters can be found in Husmann et al. (2022).

Value

An initialized *optimLanduse* S3 object ready for optimization.

References

Husmann, K., von Groß, V., Bödeker, K., Fuchs, J. M., Paul, C., & Knoke, T. (2022). *optimLanduse*: A package for multiobjective land-cover composition optimization under uncertainty. *Methods in Ecology and Evolution*, 00, 1– 10. <https://doi.org/10.1111/2041-210X.14000>

Examples

```
require(readxl)
dat <- read_xlsx(exampleData("exampleGosling.xlsx"))

init <- initScenario(dat,
                     uValue = 2,
                     optimisticRule = "expectation",
                     fixDistance = 3)
```

solveScenario	<i>Perform the optimization</i>
---------------	---------------------------------

Description

The function solves the optimization framework specified by the initialized *optimLanduse* object.

Usage

```
solveScenario(
  x,
  digitsPrecision = 4,
  lowerBound = 0,
  upperBound = 1,
  landUseRestriction = NA,
  paretoY = NA,
  paretoX = NA,
  paretoMaxDistance = NA
)
```

Arguments

<code>x</code>	The initialized <i>optimLanduse</i> object. See initScenario for the initialization.
<code>digitsPrecision</code>	Precision of the loss value. <code>digitsPrecision</code> is the possibility to influence the calculation time.
<code>lowerBound</code>	Optional lower bounds for the land-use options. Must be 0 or a vector in the dimension of the land-use options.
<code>upperBound</code>	Optional upper bounds for the land-use options. Must be 1 or a vector in the dimension of the land-use options.

landUseRestriction	Optional named numeric vector imposing an upper-bound restriction on the shares of individual land-use options. The names must be a subset of the land-use options of the initialized optimLanduse object; each value sets the maximum allowed share for that option. Passing NA (the default) disables the restriction.
paretoY	Optional character vector naming the indicator(s) used as the objective (the Y dimension) when computing a point of the Pareto frontier. Must be supplied together with paretoX. Passing NA (the default) disables the Pareto formulation and performs the standard optimization.
paretoX	Optional character vector naming the indicator(s) used as the constraint (the X dimension) of the Pareto optimization. Must be supplied together with paretoY. Passing NA disables the Pareto formulation.
paretoMaxDistance	Optional numeric value giving the minimum level that the paretoX indicator(s) must attain, i.e. the threshold along the X dimension that constrains the Pareto optimization. Passing NA disables the Pareto formulation.

Details

The methodological background and the formulation of the optimization framework are described in Knoke et al. (2016) and in Husmann et al. (2022)

Value

A solved landUse portfolio ready for export or further data processing.

References

- Knoke, T., Paul, C., Hildebrandt, P. et al. (2016): Compositional diversity of rehabilitated tropical lands supports multiple ecosystem services and buffers uncertainties. *Nat Commun* **7**, 11877. [doi:10.1038/ncomms11877](https://doi.org/10.1038/ncomms11877)
- Husmann, K., von Groß, V., Bödeker, K., Fuchs, J. M., Paul, C., & Knoke, T. (2022). optimLanduse: A package for multiobjective land-cover composition optimization under uncertainty. *Methods in Ecology and Evolution*, 00, 1– 10. <https://doi.org/10.1111/2041-210X.14000>

Examples

```
require(readxl)
dat <- read_xlsx(exampleData("exampleGosling.xlsx"))

init <- initScenario(dat,
                     uValue = 2,
                     optimisticRule = "expectation",
                     fixDistance = 3)

result <- solveScenario(x = init)
```

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