

# Package ‘errorlocate’

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**Type** Package

**Title** Locate Errors with Validation Rules

**Version** 0.4

**Description** Errors in data can be located and removed using validation rules from package 'validate'.

**License** GPL-3

**LazyData** TRUE

**URL** <https://github.com/data-cleaning/errorlocate>

**BugReports** <https://github.com/data-cleaning/errorlocate/issues>

**Depends** validate

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'dnf.R' 'errorlocalizer.R' 'errorlocate-package.r'  
'errorlocation.R' 'expr\_manip.R' 'linear.R' 'locate-errors.R'  
'mip\_lpsolve.R' 'mip\_rule.R' 'replace-errors.R' 'soft-rule.R'  
'utils.R' 'values.R'

**NeedsCompilation** no

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errorlocate-package	<i>Find errors in data given a set of validation rules.</i>
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**Description**

Find errors in data given a set of validation rules. The errorlocate helps to identify obvious errors in raw datasets.

**Details**

It works in tandem with the package [validate](#). With validate you formulate data validation rules to which the data must comply. For example:

"age cannot be negative": age >= 0

While validate can identify if a record is valid or not, it does not identify which of the variables are responsible for the invalidation. This may seem a simple task, but is actually quite tricky: a set of validation rules form a web of dependent variables: changing the value of an invalid record to repair for rule 1, may invalidate the record for rule 2.

Errorlocate provides a small framework for record based error detection and implements the Felligi Holt algorithm. This algorithm assumes there is no other information available then the values of a record and a set of validation rules. The algorithm minimizes the (weighted) number of values that need to be adjusted to remove the invalidation.

The errorlocate package translates the validation and error localization problem into a mixed integer problem and uses a mip solver to find a solution.

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## References

T. De Waal (2003) Processing of Erroneous and Unsafe Data. PhD thesis, University of Rotterdam.  
 Van der Loo, M., de Jonge, E, Data Cleaning With Applications in R  
 E. De Jonge and Van der Loo, M. (2012) Error localization as a mixed-integer program in editrules.  
 lp\_solve and Kjell Konis. (2011). lpSolveAPI: R Interface for lp\_solve version 5.5.2.0. R package version 5.5.2.0-5. <http://CRAN.R-project.org/package=lpSolveAPI>

## See Also

Useful links:

- <https://github.com/data-cleaning/errorlocate>
- Report bugs at <https://github.com/data-cleaning/errorlocate/issues>

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add_noise	<i>Add (a small amount of) noise</i>
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## Description

Utility function to add some small positive noise to weights. This is mainly done to randomly choose between solutions of equal weight. Without adding noise to weights lp solvers may return an identical solution over and over while there are multiple solutions of equal weight. The generated noise is positive to prevent that weights will be zero or negative.

## Usage

```
add_noise(x, max_delta = NULL, ...)
```

## Arguments

x	numeric vector or matrix. When x is a matrix, the function will be applied to each row of the matrix.
max_delta	when supplied noise will be drawn from $[0, \text{max\_delta}]$ otherwise see details
...	currently not used

## Details

When no max\_delta is supplied, add\_noise will use the minimum difference larger than zero divided by the length(x).

## Value

numeric vector/matrix with noise applied.

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ErrorLocalizer-class    *Base class for class locate errors based on rules and data*

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### Description

ErrorLocalizer can be used as a base class to implement a new error localization algorithm. The derived class must implement two methods: `initialize`, which is called before any error localization is done and `locate` which operates upon data. The extra parameter `...` can be used to supply algorithmic specific parameters.

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errorlocation-class    *Error location object*

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### Description

Errorlocation contains the result of a error detection. Errors can be record based or variable based.

- A record based error is restricted within one observation. `errorlocate` using the Felligi Holt algorithm assumes errors are record based.
- A variable based error is a flaw in uni- or multivariate distribution. To correct this error multiple observations or the aggregated number should be adjusted.

### Details

Current implementation assumes that errors are record based. The error locations can be retrieved using the method `values` and are a matrix of rows and columns, with the same dimensions as the `data.frame` that was checked. For errors that are purely column based, or dataset based, error-locations will return a matrix with all rows or cells set to `TRUE`. The `values` return `NA` for missing values.

### Fields

- `$errors`: matrix indicating which values are erroneous (`TRUE`), missing (`NA`) or valid (`FALSE`)
- `$weight`: The total weight per record. A weight of 0 means no errors were detected.

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errors_removed	<i>Get location of removed errors from a 'cleaned' data set</i>
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**Description**

errors\_removed retrieves the errors detected by [replace\\_errors](#)

**Usage**

```
errors_removed(x, ...)
```

**Arguments**

- x                    data.frame that was checked for errors
- ...                  not used

**Value**

[errorlocation-class](#) object

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expect_values	<i>expect values</i>
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**Description**

expect values

**Usage**

```
expect_values(values, weights, ...)
```

**Arguments**

- values              named list of values.
- weights             named numeric of equal length as values.
- ...                  not used

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FHLocalizer-class	<i>Feligi-Holt Errorlocalizer</i>
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### Description

Implementation of the Feligi-Holt algorithm using the ErrorLocalizer base class. Given a set of validation rules and a dataset the Feligi-Holt algorithm finds for each record the smallest (weighted) combination of variables that are erroneous (if any).

### Note

Most users do not need this class and can use [locate\\_errors](#).

errorlocalizer implements feligi holt using a MIP-solver. For problems in which coefficients of the validation rules or the data are too different, you should consider scaling the data.

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is_categorical	<i>Check if rules are categorical</i>
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### Description

Check if rules are categorical

### Usage

```
is_categorical(x, ...)
```

### Arguments

x	validator or expression object
...	not used

### Value

logical indicating which rules are purely categorical/logical

### Examples

```
v <- validator( A %in% c("a1", "a2")
               , B %in% c("b1", "b2")
               , if (A == "a1") B == "b1"
               , y > x
               )

is_categorical(v)
```

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is_conditional	<i>Check if rules are conditional rules</i>
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**Description**

Check if rules are conditional rules

**Usage**

```
is_conditional(rules, ...)
```

**Arguments**

rules	validator object containing validation rules
...	not used

**Value**

logical indicating which rules are conditional

**Examples**

```
v <- validator( A %in% c("a1", "a2")
               , B %in% c("b1", "b2")
               , if (A == "a1") x > 1 # conditional
               , if (y > 0) x >= 0 # conditional
               , if (A == "a1") B == "b1" # categorical
               )

is_conditional(v)
```

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is_linear	<i>Check which rules are linear rules.</i>
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**Description**

Check which rules are linear rules.

**Usage**

```
is_linear(x, ...)
```

**Arguments**

x	<a href="#">validator</a> object containing data validation rules
...	not used

**Value**

logical indicating which rules are (purely) linear.

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locate_errors	<i>Locate errors in data</i>
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**Description**

Locate erroneous fields in rows of data using validation rules or a specific errorlocalizer object. This method returns found errors, according to the specified method *x*. If these errors are to be removed automatically use method [replace\\_errors](#).

**Usage**

```
locate_errors(data, x, ..., timeout = 60)

## S4 method for signature 'data.frame,validator'
locate_errors(data, x, weight = NULL, ref = NULL, ..., timeout = 60)

## S4 method for signature 'data.frame,ErrorLocalizer'
locate_errors(data, x, weight = NULL, ref = NULL, ..., timeout = 60)
```

**Arguments**

<i>data</i>	data to be checked
<i>x</i>	validation rules or errorlocalizer object to be used for finding possible errors.
<i>...</i>	optional parameter to be used by a specific method
<i>timeout</i>	maximum number of seconds that the localizer should use per record.
<i>weight</i>	numeric optional weight vector to be used in the error localization.
<i>ref</i>	data.frame optional reference data to be used in the rules checking

**Value**

[errorlocation-class](#) object describing the errors found.

**Examples**

```
rules <- validator( profit + cost == turnover
                    , cost >= 0.6 * turnover # cost should be at least 60% of turnover
                    , turnover >= 0 # can not be negative.
                    )
data <- data.frame(profit=755, cost=125, turnover=200)
le <- locate_errors(data, rules)

print(le)
summary(le)
```



```

v_categorical <- validator( A %in% c("a1", "a2")
                           , B %in% c("b1", "b2")
                           , if (A == "a1") B == "b1"
                           )

data <- data.frame(A = c("a1", "a2"), B = c("b2", "b2"))
locate_errors(data, v_categorical)$errors

v_logical <- validator( A %in% c(TRUE, FALSE)
                       , B %in% c(TRUE, FALSE)
                       , if (A == TRUE) B == TRUE
                       )

data <- data.frame(A = TRUE, B = FALSE)
locate_errors(data, v_logical, weight=c(2,1))$errors

# try a condinational rule
v <- validator( married %in% c(TRUE, FALSE), if (married==TRUE) age >= 17 )
data <- data.frame( married = TRUE, age = 16)
locate_errors(data, v, weight=c(married=1, age=2))$errors

```

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MipRules-class

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*Create a mip object from a validator object*


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## Description

Create a mip object from [validator](#) object. This is a utility class that translates a validator object into a mixed integer problem that can be solved. Most users should use [locate\\_errors](#) which will handle all translation and execution automatically. This class is provided so users can implement or derive an alternative solution.

## Methods

The MipRules class contains the following methods:

- `$execute` calls the mip solver to execute the rules.
- `$to_lp`: transforms the object into a `lp_solve` object
- `$is_infeasible` Checks if the current system of mixed integer rules is feasible.
- `$set_values`: set values and weights for variables (determines the objective function).

## Examples

```

rules <- validator(x > 1)
mr <- miprules(rules)
mr$to_lp()
mr$set_values(c(x=0), weights=c(x=1))
mr$execute()

```

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replace_errors	<i>Replace erroneous fields with NA or a suggested value</i>
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## Description

Find erroneous fields using [locate\\_errors](#) and replace these fields automatically with NA or a suggestion that is provided by the error detection algorithm.

## Usage

```
replace_errors(data, x, ref = NULL, ..., value = c("NA", "suggestion"))

## S4 method for signature 'data.frame,validator'
replace_errors(data, x, ref = NULL, ..., value = c("NA", "suggestion"))

## S4 method for signature 'data.frame,ErrorLocalizer'
replace_errors(data, x, ref = NULL, ..., value = c("NA", "suggestion"))

## S4 method for signature 'data.frame,errorlocation'
replace_errors(data, x, ref = NULL, ..., value = c("NA", "suggestion"))
```

## Arguments

data	data to be checked
x	<a href="#">validator</a> object
ref	optional reference data set
...	these parameters are handed over to <a href="#">locate_errors</a>
value	NA

## Details

Note that you can also use the result of [locate\\_errors](#) with `replace_errors`. When the procedure takes a long time and `locate_errors` was called previously this is the preferred way, because otherwise `locate_errors` will be executed again. The errors that were removed from the `data.frame` can be retrieved with the function [errors\\_removed](#). For more control over error localization see [locate\\_errors](#).

## Value

data with erroneous values removed.

## Note

In general it is better to replace the erroneous fields with NA and apply a proper imputation methods. Suggested values from the error localization method may introduce an unwanted bias.

**See Also**[errorlocation-class](#)**Examples**

```

rules <- validator( profit + cost == turnover
                    , cost - 0.6*turnover >= 0
                    , cost >= 0
                    , turnover >= 0
                    )
data <- data.frame(profit=755, cost=125, turnover=200)

data_no_error <- replace_errors(data,rules)

# faulty data was replaced with NA
data_no_error

errors_removed(data_no_error)

# a bit more control, you can supply the result of locate_errors
# to replace_errors, which is a good thing, otherwise replace_errors will call
# locate_errors internally.
error_locations <- locate_errors(data, rules)
replace_errors(data, error_locations)

```

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translate_mip_lp	<i>translate linear rules into an lp problem</i>
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**Description**

translate linear rules into an lp problem

**Usage**

```
translate_mip_lp(rules, objective = NULL, eps = 0.001, ...)
```

**Arguments**

rules	mip rules
objective	function
eps	accuracy for equality/inequality
...	additional <a href="#">lp.control</a> parameters that are set for the mip problem

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