

Finding and handling data errors

errorlocate

Edwin de Jonge and Mark van der Loo

CBS, Department of Methodology

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Error localization

Data validation and error localization answer different questions.

Data validation

Which errors are there?

Error localization

Where do I need to make changes to fix the errors?



Example

Ruleset

```
age >= 0  
age <= 120  
if (drivers_licence == TRUE) age >= 18
```

Data

age	drivers_licence
150	TRUE

Question:

Which field or fields would you change?

Error localization

Definition

Error localization is a procedure that points out fields in a data set that can be altered or imputed in such a way that all validation rules can be satisfied.

Example

Ruleset

```
if (married == TRUE ) age >= 16  
if (attends == "kindergarten") age <= 6
```

Data

age	married	attends
3	TRUE	kindergarten

Question

Which field or fields would you change?



Assignment

Use `validate` to check the data and find the variable that is incorrect.

```
if (married == TRUE ) age >= 16  
if (attends == "kindergarten") age <= 6
```

Data

age	married	attends
3	TRUE	kindergarten



Data shoot-out

```
age >= 0,  
age < 150,  
if (driver_license == TRUE) age >= 16
```

age	driver_license
10	TRUE

Question

Which field or fields would you change?



Data shoot-out

```
age >= 0,  
age < 150,  
if (driver_license == TRUE) age >= 16
```

age	driver_license
10	TRUE

Question

Which field or fields would you change?

It depends on the quality of age and driver_license. We can add more weight to age if we think that variable has better quality.



Principle of Fellegi and Holt

Find the minimal (weighted) number of fields to adjust such that all rules, including implied rules, can be satisfied.

IP Fellegi and D Holt, JASA **71** 353 17–35 (1976).

Note

This should be used as a last resort, when no further information on the location of errors is available.



Feligi Holt (FH) formalism:

But there are exceptions. . .

- In balance sheets, swapping variables (2 edits) sometimes makes more sense than adjusting one value (1 edit). (see R package:deducorrect).
- In some data, spreading a surplus or shortage on a variable over many variables is sensible. (see R package: rspa).



Implied rules?

```
turnover - total.cost == profit  
profit <= 0.6 * turnover
```

This implies (substituting profit):

```
total.cost >= 0.4 * turnover
```

We need to take into account such *essentially new* rules (edits) —unstated relations between variables that can be derived from the explicitly defined rules.



errorlocate

- R-package that implements FH.
- Is extensible (you can plug in your own detection stuff)
- provides:
 - `locate_errors`
 - `replace_errors`
 - R5 classes to add your own stuff.

errorlocate::locate_errors

```
locate_errors( data.frame( age  = 3
                           , married = TRUE
                           , attends = "kindergarten"
                           )
  , validator( if (married == TRUE) age >= 16
               , if (attends == "kindergarten") age <= 6
               )
  )
```

```
## call: x$locate(data = data, weight = weight, ...)
```

```
## located 1 error(s).
```

```
## located 0 missing value(s).
```

```
## Use 'summary', 'values', '$errors' or '$weight', to explore and retrieve
```



errorlocate::locate_errors

```
locate_errors( data.frame( age  = 3
                           , married = TRUE
                           , attends = "kindergarten"
                           )
  , validator( if (married == TRUE) age >= 16
               , if (attends == "kindergarten") age <= 6
               )
  )$errors
```

```
##           age married attends
## [1,] FALSE      TRUE  FALSE
```



Assignment (small examples)

a) Find the error in this record with `locate_errors`:

age	married	attends
26	TRUE	kindergarten

b) Find the error with `locate_errors`:

age	married	attends
15	TRUE	kindergarten

c) You have more confidence in the kindergarden variable: apply a weight of 3 to `age` in finding the errors.



Removing errors

- Detecting errors is very useful, but then what?
- Fixing philosophy is:
 - Find erroneous values.
 - Remove them (i.e. make them NA).
 - Impute them with sensible values.

Note

We could also remove erroneous records completely, but often this result in *over-deletion* and introduces a *bias*.

errorlocate::replace_errors

- Locates errors and replaces them with NA.

```
replace_errors(  
  data.frame( age      = 3  
              , married = TRUE  
              , attends = "kindergarten"  
            )  
  , validator( if (married == TRUE) age >= 16  
              , if (attends == "kindergarten") age <= 6  
            )  
)
```

```
##   age married   attends  
## 1    3      NA kindergarten
```



Assignment

Use the data set retailers from package validate.

- a) Use validate to find out which records are faulty using the rule set

```
to_pos = turnover >= 0  
or_pos = other.rev >= 0  
balance = turnover + other.rev == total.rev
```

- b) use locate_errors to find some errors.
c) use replace_errors to “fix” the data set.



```
data(retailers, package="validate")
retailers <- retailers[c("other.rev", "total.rev", "turnover")]
rules <- validator(
  to_pos = turnover >= 0
  , or_pos = other.rev >= 0
  , balance = turnover + other.rev == total.rev)
confront(retailers, rules)
```

```
## Object of class 'validation'
## Call:
##      confront(dat = retailers, x = rules)
##
## Confrontations: 3
## With fails      : 2
## Warnings        : 0
## Errors          : 0
```



```
errors <- locate_errors(retailers, rules)$errors
row_contains_error <- apply(errors, 1, any)
w <- which(row_contains_error)
errors[w,]
```

##		other.rev	total.rev	turnover
##	[1,]	TRUE	FALSE	FALSE
##	[2,]	FALSE	TRUE	FALSE
##	[3,]	NA	FALSE	TRUE
##	[4,]	TRUE	FALSE	FALSE
##	[5,]	FALSE	FALSE	TRUE



```
retailers[w,c("other.rev", "total.rev", "turnover")]
```

##	other.rev	total.rev	turnover
## 3	-33	6919	6886
## 30	1831	1831	1831
## 32	NA	107	971
## 36	98350	2747	2649
## 37	4	206	1024

```
replace_errors(retailers[w,], rules)
```

##	other.rev	total.rev	turnover
## 3	NA	6919	6886
## 30	1831	NA	1831
## 32	NA	NA	971
## 36	98350	NA	2649
## 37	4	206	NA



Internal workings:

`errorlocate:`

- translates error localization problem into a **mixed integer problem**, which is solved with `lpsolveAPI`.
- contains a small framework for implementing your own error localization algorithms.

Pipe friendly

The `replace_errors` function is pipe friendly:

```
rules <- validator(age < 150)

data_noerrors <-
  data.frame(age=160, driver_license = TRUE) %>%
  replace_errors(rules)

errors_removed(data_noerrors) # contains errors removed
```

