

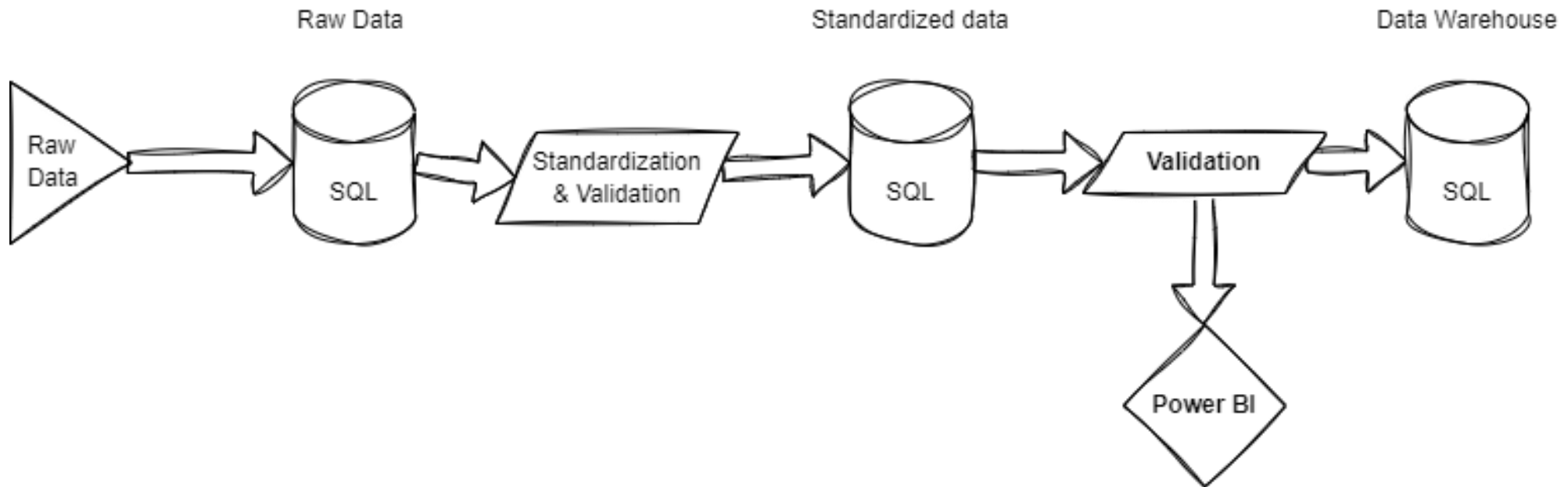
Automating Data Validation on SQL Server Using R and the Machine Learning package

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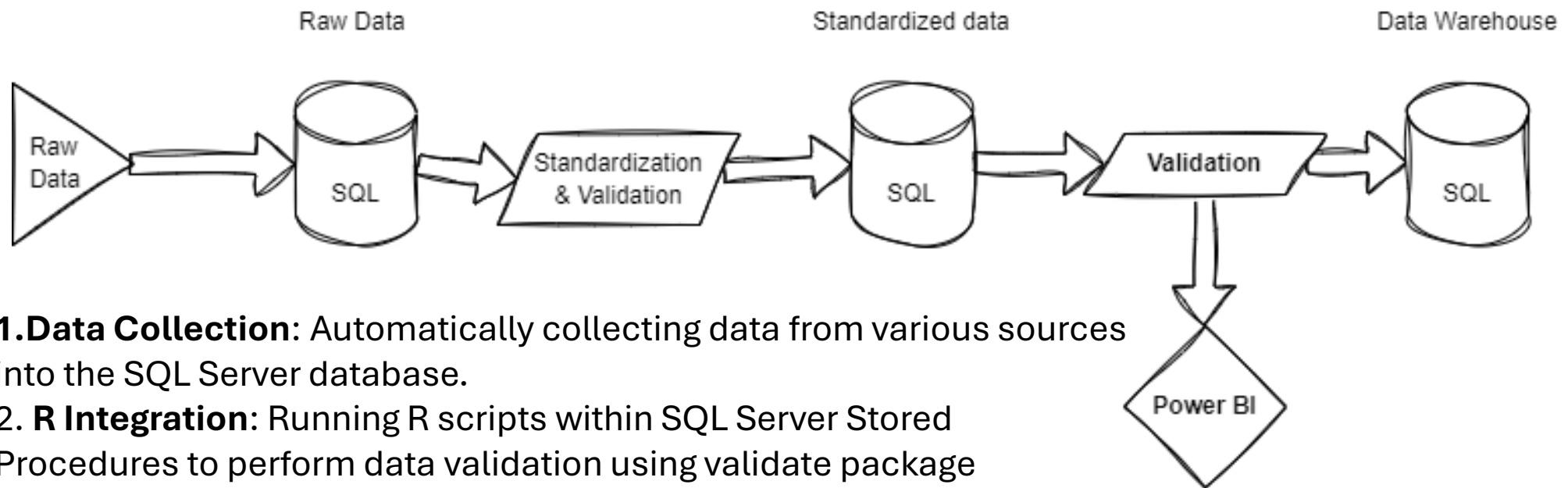
Automated data validation within dataflow on SQL Server

Fully automated data flow in SQL Stored Procedures



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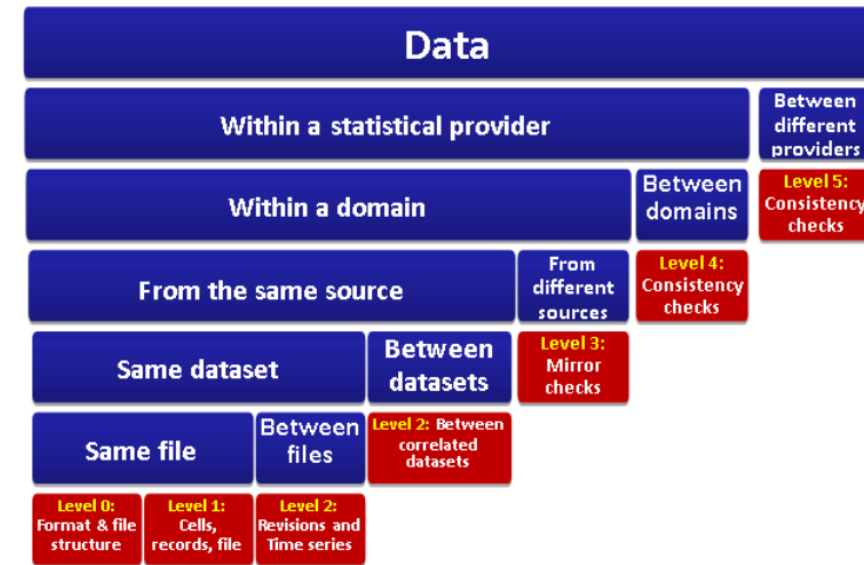


- 1.Data Collection:** Automatically collecting data from various sources into the SQL Server database.
- 2. R Integration:** Running R scripts within SQL Server Stored Procedures to perform data validation using validate package
- 3. Scheduling:** Setting up SQL Server Agent jobs to automate the execution of Stored Procedures at defined intervals.

Developement of the process:

1 Validation rules in [rules]

- R code directly saved in SQL Server table for every rule
- Validation rules according to validation classes, e.g. Cellular, within data set, against other files, data sets,....
- In collaboration with the (end)users



[validation].[classes]

class_ID	name	description	level_ID
1	Data delivery	NULL	1
2	Number of columns	NULL	1
3	Column data type	NULL	1
4	Variable length	NULL	1
5	Variable range	NULL	2

[validation].[rules]

rule_ID	rule_name	rule_description	rule_class	r_code	table_reference	reference_column	valid_from	valid_to	active	comment
16	empty_entry_contract_ID	No NULL in column contract_ID	6	!is.na(contract_ID)	contracts	contract_ID	6/1/2024	12/31/9999	1	NULL
32	squaremeter_size	Squaremeter size in between [0,500]	5	in_range(squaremeter, 0, 500)	housing	squaremeter	7/23/2024	9/13/2024	0	NULL



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Developement of the process:

2 Validation in R

- Development of the R code in R Studio
- Collection of data and validation rules from SQL tables
- Select which data to validate by collection_IDs

`[validation].[errors]`

- Validation

error_ID	contract_ID	rule_ID	collection_date	collection_ID
22599	614	21	26:16.0	37
22648	23463	31	26:16.0	37

- Extraction of results into excel / SQL table

Developement of the process:

2 Validation in R

```
1  validate <- function(datafile){
2
3    name_datafile <- deparse(substitute(datafile))
4    # Validation for this datafile
5    rules_for_datafile <- validator(.data = rules[rules$table_reference == name_datafile,])
6    # run validation of these rules
7    validation_result <- confront(datafile, rules_for_datafile, key="contract_ID")
8    # validation results put together in table
9    df_validation_result <- as.data.frame(validation_result)
10   overall_validation_result <- summary(validation_result)[,1:5]
11   # contract_IDs that did not stand validation and the rules they failed in
12   broken_rules <- df_validation_result %>%
13     filter(value != TRUE) %>%
14     mutate(name = as.integer(substr(name,2,nchar(name)))) %>%
15     distinct(name, contract_ID)
```



Development of the process:

3 Writing validation results into excel (Dataprovider)

```
17 # write validation results in excel file for data provider response
18 overall_result_with_name <- overall_validation_result %>%
19   left_join(as.data.frame(rules_for_datafile)) %>%
20   mutate(name = description) %>%
21   select(name, items, passes, fails, nNA)
22 broken_rules_with_name <- broken_rules %>%
23   mutate(name = paste0("X", name)) %>%
24   left_join(as.data.frame(rules_for_datafile)) %>%
25   mutate(name = description) %>%
26   select(name, contract_ID)
27 write.xlsx(overall_result_with_name,
28           file = wd_file,
29           sheetName = paste0(name_datafile, "-all rules"),
30           col.names = TRUE, row.names = TRUE, append = TRUE)
31 write.xlsx(broken_rules_with_name,
32           file = wd_file,
33           sheetName = paste0(name_datafile, "-failures"),
34           col.names = TRUE, row.names = TRUE, append = TRUE)
35
```



Developement of the process:

4 Writing validation results into SQL tables

```
36 # writing errors in SQL [validation].[errors]
37 if(nrow(broken_rules) > 0){
38   errors <- as.data.frame(cbind(contract_ID      = broken_rules["contract_ID"],
39   |                           rule_ID           = broken_rules["name"],
40   |                           collection_date    = as_datetime(format(Sys.time(), '%Y-%m-%d %H:%M:%S')),
41   |                           collection_ID     = collection_ID)) %>%
42   |   rename(rule_ID = name)
43   dbWriteTable(con,
44   |             DBI::Id(schema = "validation", table = "errors"),
45   |             errors,
46   |             append = TRUE)
47 }
48 }
49 }
```

[validation].[error]

error_ID	contract_ID	rule_ID	collection_date	collection_ID
22599	614	21	26:16.0	37
22648	23463	31	26:16.0	37



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Transfer to SQL Stored Procedure, basic example

```
CREATE PROCEDURE [dbo].[CalculateValues]
AS
BEGIN
    EXEC sp_execute_external_script
        @language = N'R',
        @script = N' #library()
                    a <- InputDataSet$a
                    b <- InputDataSet$b
                    c <- a / b
                    d <- a * b
                    OutputDataSet <- data.frame(a, b, c, d)',
        @input_data_1 = N'SELECT * FROM [dbo].[example];',
        @output_data_1_name = N'OutputDataSet'
    WITH RESULT SETS ((a FLOAT, b FLOAT, c FLOAT, d FLOAT));
END;

EXEC [dbo].[CalculateValues]
```

[dbo].[example]

	a	b
1	2	4
2	3	6

OutputDataSet

	a	b	c	d
1	2	4	0.5	8
2	3	6	0.5	18



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Obstacles

- R version control
(SQL Server 2022 Machine Learning package : R 4.2.0)
- R package management
- Language issues (Icelandic letters in the data, utf-8) on SQL Server 2019 (R 3.5.2)
- Loading in more than one SQL table per Stored Procedure – in progress
- Resource management

Pros/Cons of automation directly on SQL Server using R code within ML package

Pro

- Resssource **efficient**: Processing ressources and time
- Data is processed where it is stored
- User/Machine independent automation
- Perfect addition to fully automized data flow process
- Standardized way of data validation (reusable)
- Comparability between process qualities
- Continous data quality insurance
- Implementing of additional validation checks
- Improved accuracy in detecting data anomalies and inconsistencies
- Successful **scheduling** of validation tasks
- Potential for **Machine Learning / AI** implementation

Con

- Need for Machine Learning package on SQL Server
- Recommended from SQL Microsoft Server 2022 onwards (otherwise language issues)
- Learning threshold to implement R code into Stored Procedures
- Only one SQL table can be read in in per Stored Procedure at a time
- Installation of R packages (IT support) within ML package

THANK YOU

Do you have any questions?

Used R packages

- **Validate[1]: for validation** cran.r-project.org/web/packages/validate/
 - declare rules
 - apply them on dataset
- **xlsx[2]: production of validation files to return to data providers, only if applicable** cran.r-project.org/web/packages/xlsx/
- **DBI [3] / odbc[4] for read in of SQL tables / writing results into SQL tables , only if not run on SQL Server** cran.r-project.org/web/packages/DBI/ / cran.r-project.org/web/packages/odbc/