Formal Description of Validity Criteria for Data and Automated Cloud-Based Validity Checking to Generate a Validated Data Set

Lucas Krauße
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https://dse.in.tum.de/



In cooperation with BMW Group

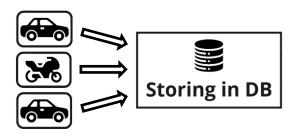


BMW processes data from customer vehicles that statistically describe driving behavior
 => FASTA data



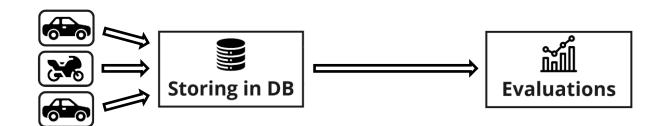


- BMW processes data from customer vehicles that statistically describe driving behavior
 => FASTA data
- Every day there are approximately 100,000 new vehicle readouts which are stored but not validated



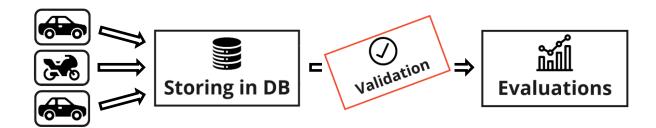


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 => FASTA data
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- Based on FASTA data, evaluations are made for a wide variety of areas





- BMW processes data from customer vehicles that statistically describe driving behavior
 => FASTA data
- Every day there are approximately 100,000 new vehicle readouts which are stored but not validated
- Based on FASTA data, evaluations are made for a wide variety of areas
- Due to measurement errors, sensor failures or faulty software these can be **incorrect**



Outline



- Motivation
- Problem Statement
- Data Validation Requirements
- Pandera (Data Validation Library)
- Conditional Language
- Validation of Multidimensional Values
- Evaluation
- Conclusion



Formal validation of FASTA data:

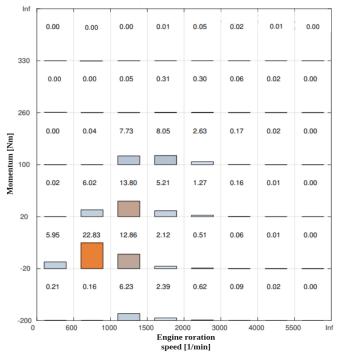


- Formal validation of FASTA data:
 - Individual columns (primitive data types)

VNR	Power	Km_Per_Day	Series_Production
MK41480	141	49.9507389	1
SR57464	190	30.7980456	0

Example individual columns

- Formal validation of FASTA data:
 - Individual columns (primitive data types)
 - Multidimensional values (e.g. Histograms)



Example: KF-PROZ multidimensional values



- Formal validation of FASTA data:
 - Individual columns (primitive data types)
 - Multidimensional values (e.g. Histograms)
 - Combination of individual attributes

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- Formal validation of FASTA data:
 - Individual columns (primitive data types)
 - Multidimensional values (e.g. Histograms)
 - Combination of individual attributes
 - Conditional rules

Operating Days	ReadoutDate	ProductionDate	Registered Date
0	736982	736961	
921	737903	736961	736982

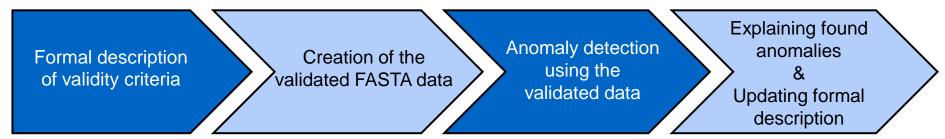
Example: Conditional rules and combination of individual attributes

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- Formal validation of FASTA data:
 - Individual columns (primitive data types)
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Pipeline:



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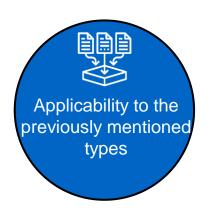
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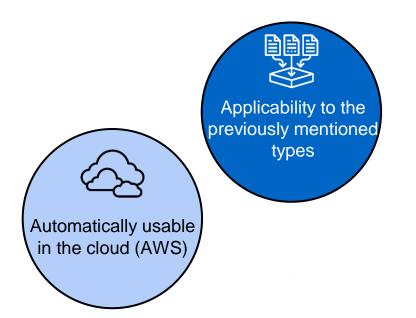
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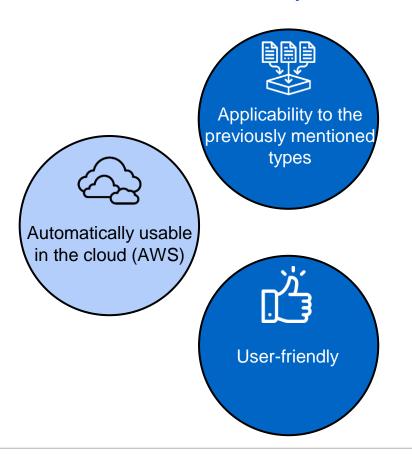
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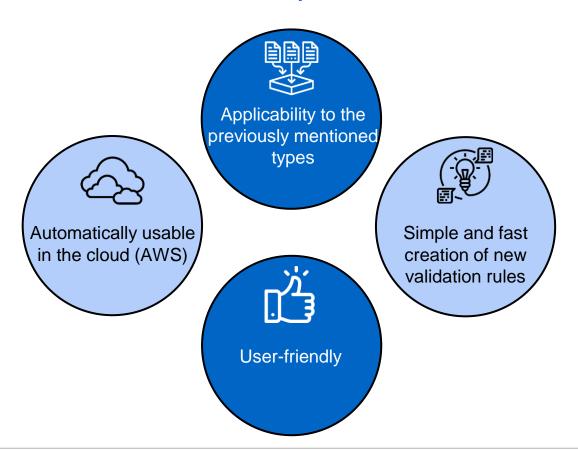


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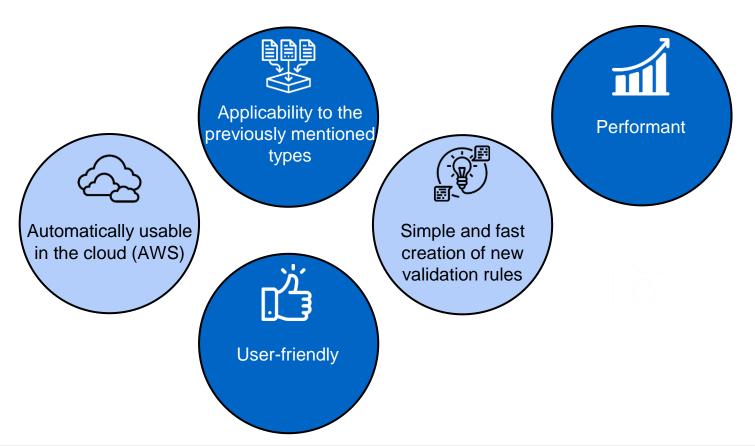




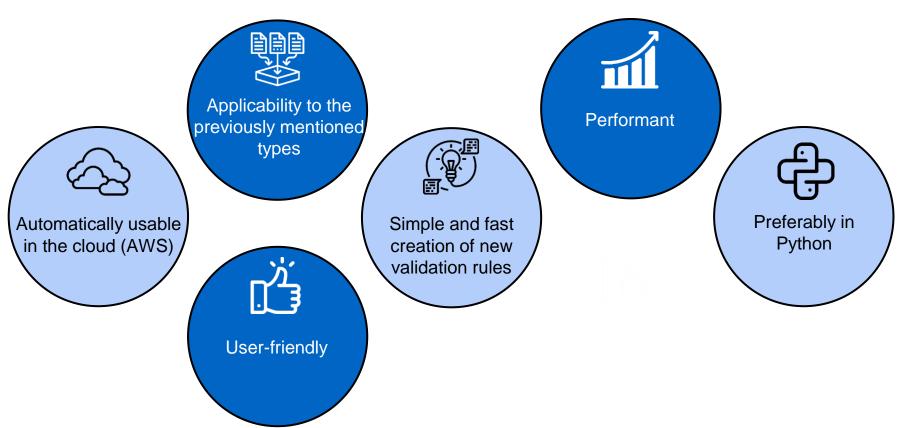




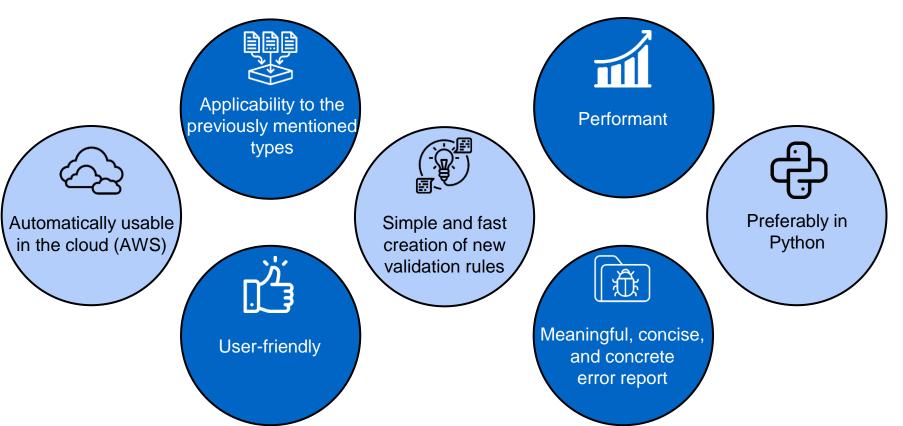












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Pandera (1)



- Definition of a Schema
- At runtime data validation on DataFrame (Excel table)
- Adding own (more complex) validation rules by an annotation
- Error report as DataFrame (Excel tables)



index	check	column	schema_context	check_number	failure_case
0	Operating Days_validation	1	DataFrameSchema		
	TMOT_temp_order	TMOT	Column	2	[63.2755 28.75814 0 0 0]
	sum_up_to_100	TMOT	Column	1	[63.2755 28.75814 0 0 0]
21	not_nullable	AvgTripLen	Column		
48	date_before_current_date	ReadoutDate	Column	1	738293
1050	in_range(0, 1000)	AvgTripLen	Column	0	1601,285714
1582	greater_than_or_equal_to(0)	OperatingDays	Column	0	-17

(1) https://pandera.readthedocs.io/en/stable/

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• Creation of more concise validation rules



- Creation of more concise validation rules
- DSL (=Domain Specific Language)

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- Creation of more concise validation rules
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- Allowed literals are **primitive data types**, **column** names, lists, null checks and pandas DataFrames/Series

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- Creation of more concise validation rules
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- Allowed literals are **primitive data types**, **column** names, lists, null checks and pandas DataFrames/Series
- Typical comparison operators, negation and in-list operator

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- Creation of more concise validation rules
- DSL (=Domain Specific Language)
- Allowed literals are primitive data types, column names, lists, null checks and pandas DataFrames/Series
- Typical comparison operators, negation and in-list operator

Addition, subtraction, (integer)
 multiplication, and (integer) division

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- Creation of more concise validation rules
- DSL (=Domain Specific Language)
- Allowed literals are primitive data types, column names, lists, null checks and pandas DataFrames/Series
- Typical comparison operators, negation and in-list operator

- Addition, subtraction, (integer)
 multiplication, and (integer) division
- Logical operators: AND and OR



- Creation of more concise validation rules
- DSL (=Domain Specific Language)
- Allowed literals are **primitive data types**, **column** names, lists, null checks and pandas DataFrames/Series
- Typical comparison operators, negation and in-list OperatingDays ReadoutDate ProductionDate RegisteredDate operator

Addition, **subtraction**, (integer) multiplication, and (integer) division

Logical operators: AND and OR

0	736982	736961	
921	737903	736961	736982

Example: Conditional rules and

combination of individual attributes

"NOT RegisteredDate.isnan AND OperatingDays == ReadoutDate - RegisteredDate OR RegisteredDate.isnan AND OperatingDays == ReadoutDate - ProductionDate"

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- Creation of more concise validation rules
- DSL (=Domain Specific Language)
- Allowed literals are primitive data types, column names, lists, null checks and pandas DataFrames/Series
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Addition, subtraction, (integer)
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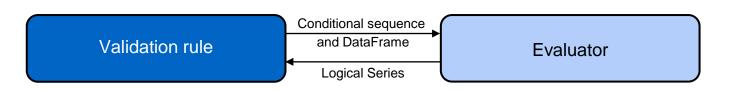
Logical operators: AND and OR

	736961	736982	0
736982	736961	737903	921

Example: Conditional rules and

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 "NOT RegisteredDate.isnan AND OperatingDays == ReadoutDate - RegisteredDate OR RegisteredDate.isnan AND OperatingDays == ReadoutDate - ProductionDate"



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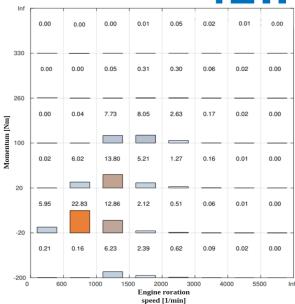
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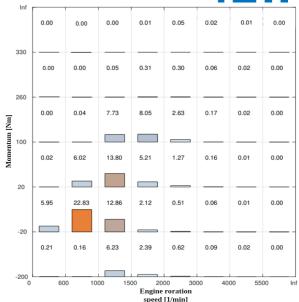
• A Total of 48 values per vehicle evaluation



Example: KF-PROZ multidimensional values

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- A Total of 48 values per vehicle evaluation
- Engine variant specific



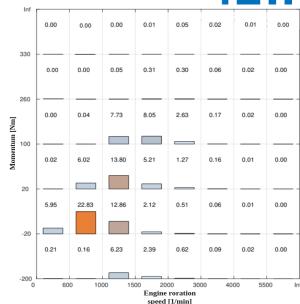
Example: KF-PROZ multidimensional values

- A Total of 48 values per vehicle evaluation
- Engine variant specific
- Machine Learning classification approach

Machine Learning

Model

(Black Box)

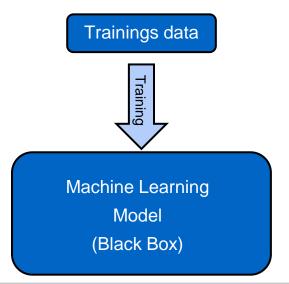


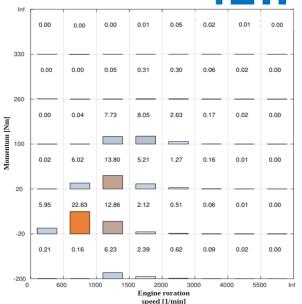
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- A Total of 48 values per vehicle evaluation
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• Trained with small portion of the FastaData (24,000 vehicle

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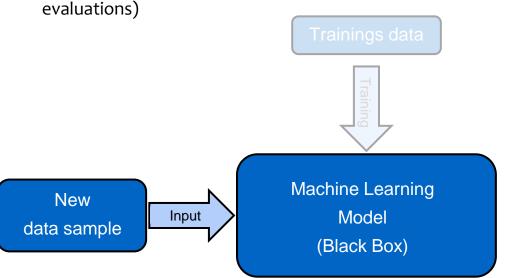


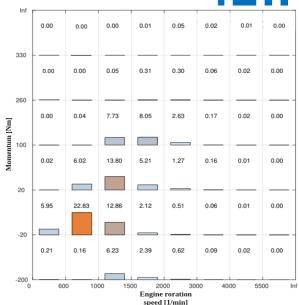


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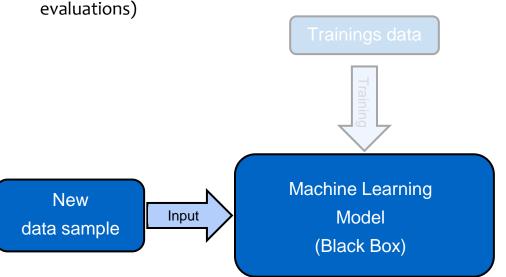


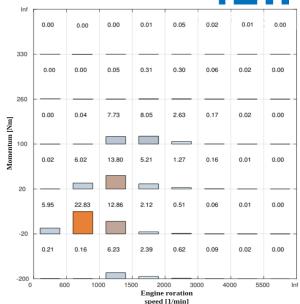


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Example: KF-PROZ multidimensional values

Motorvariante A

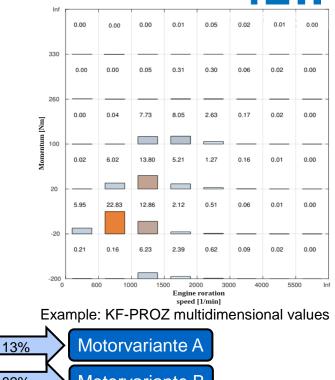
Motorvariante B

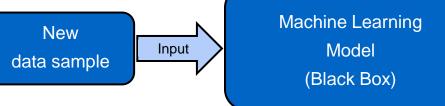
Motorvariante C

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- A Total of 48 values per vehicle evaluation
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Trained with small portion of the FastaData (24,000 vehicle evaluations)



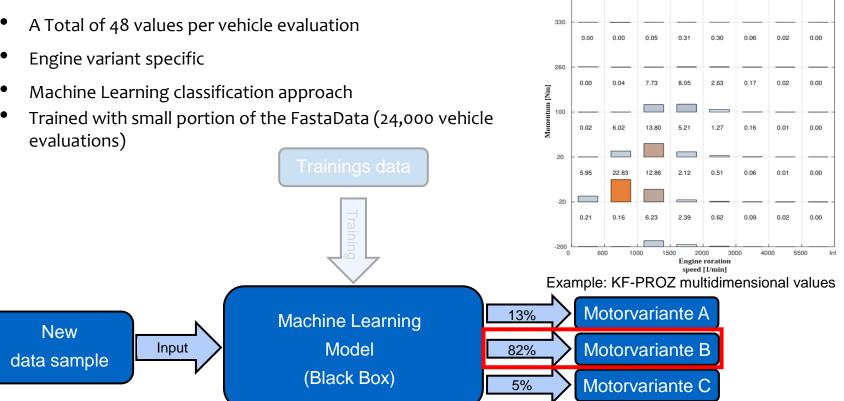


Motorvariante B 82% Motorvariante C 5%

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0.01

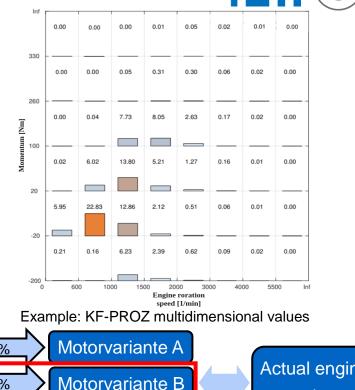
0.05

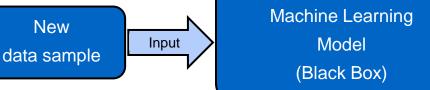


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- A Total of 48 values per vehicle evaluation
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13% Motorvariante B 82% Motorvariante C 5%

Actual engine variant

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Evaluation (1)



• Setup:

- Intel Xeon Gold 6240 CPU (2.60 GHz, only 4 of the 16 cores used due to virtualization)
- 132 GB DRAM
- Taking the average of multiple benchmarks

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- Setup:
 - Intel Xeon Gold 6240 CPU (2.60 GHz, only 4 of the 16 cores used due to virtualization)
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 - Taking the average of multiple benchmarks
- ML Model Multidimensional Values:
 - ~ 97% accuracy

Evaluation (1)



- Setup:
 - Intel Xeon Gold 6240 CPU (2.60 GHz, only 4 of the 16 cores used due to virtualization)
 - 132 GB DRAM
 - Taking the average of multiple benchmarks
- ML Model Multidimensional Values:
 - ~ 97% accuracy
- Conditional Language vs. Direct implementation:

Implementation	10,000 Records	50,000 Records	100,000 Records	1,000,000 Records
Direct	206.7 ms	975.6 ms	1954.6 ms	19179.3 ms
Conditional Language	225.3 ms	996.6 ms	1992.2 ms	19313.8 ms
Difference Δ of Direct	18.6 ms	21.0 ms	37.6 ms	134.5 ms
and Conditional Lan-				
guage Implementation				

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Evaluation (2) – Pandera Validation



- Pandera validation:
 - In total 86 Validation rules
 - Combination of rules:
 - directly implemented and
 - formulated with the conditional language

Function	10,000 Records	50,000 Records	100,000 Records	1,000,000 Records
Schema validation	2.441 sec	9.296 sec	17.500 sec	173.362 sec

• Out of 10,000 random FASTA records 5.5% had at least one validation rule violation

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Conclusion



FASTA data validation improves the overall data quality with a short runtime

Multidimensional values validation rules are realizable by classification ML Models

The created Conditional Language is universally applicable on logic tasks

All processes can be ported to the Amazon Web Services (AWS) Cloud

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Sources



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• Icons from <u>Flaticon.com</u>. Artists: Becris, Freepik, Kiranshastry, Payungkead, Pauseo8, ultimatearm

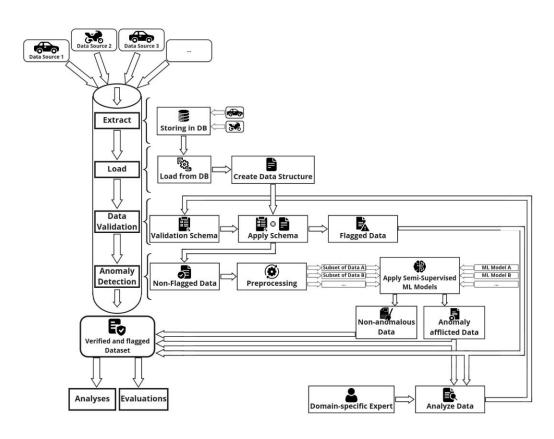


Backup

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Backup – Overview

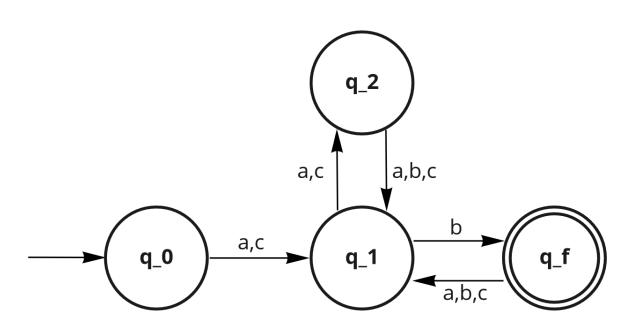




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Backup – DSL Automata explained

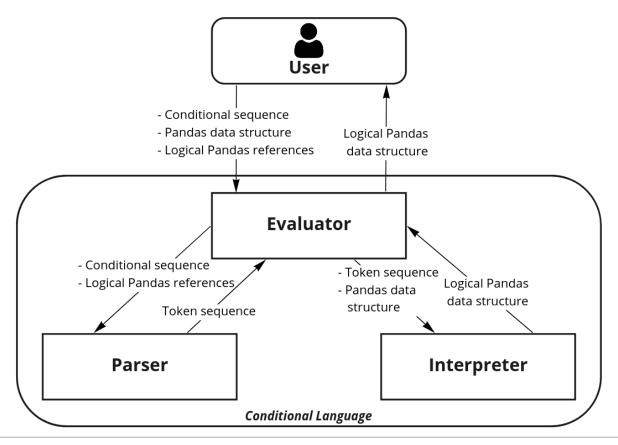




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Backup – DSL Explanation of the Evaluator Abstraction ()

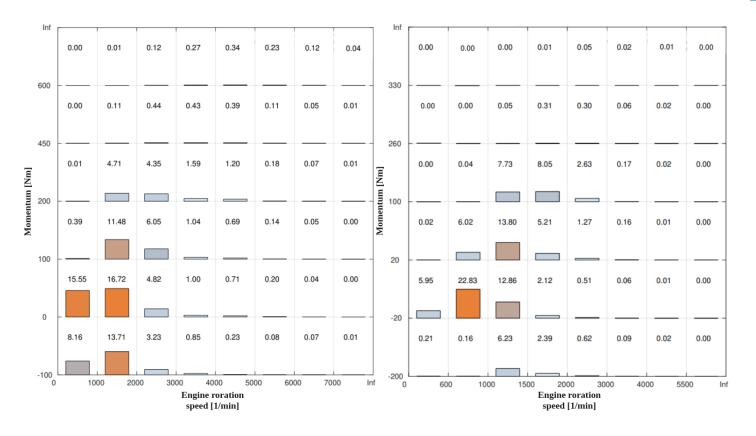




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Backup – Multidimensional Values KF-Proz Comparison (**)





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Backup – Evaluation(s)



- Multidimensional (trained) classification ML Model:
 - (Input) 200 ReLu 200 ReLu (Output)

- Found violations of validation rules of 10,000 random FASTA data records:
 - ~ 85 % NULLs
 - ~ 5 % not summed to 100
 - ~ 3 % Incorrect engine variant distribution (KF-Proz)
 - ~ 2 % each:
 - Presented Operating Days conditional validation rule
 - Date before current year
 - Not in specified range
 - Pattern match violation
 - ~ 1 % smaller than o