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Towards the Estimation of Quality Attributes on System Model Histories

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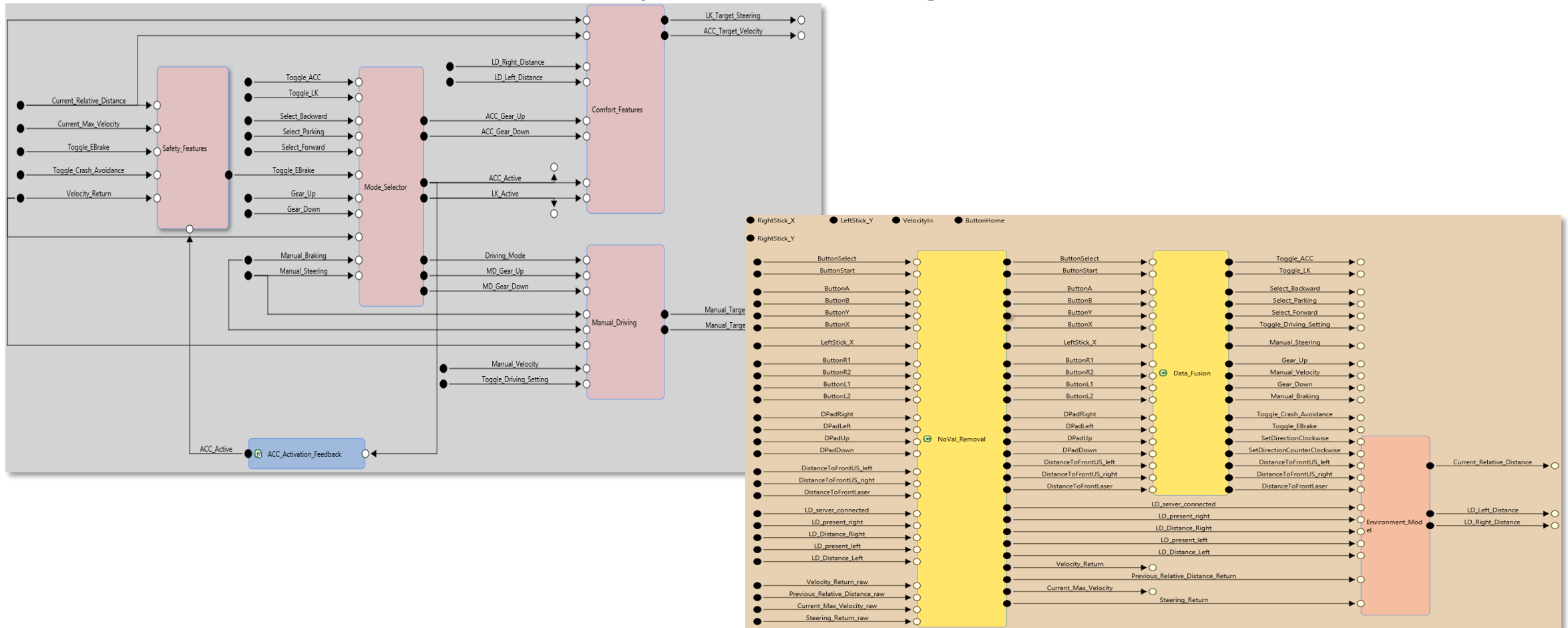


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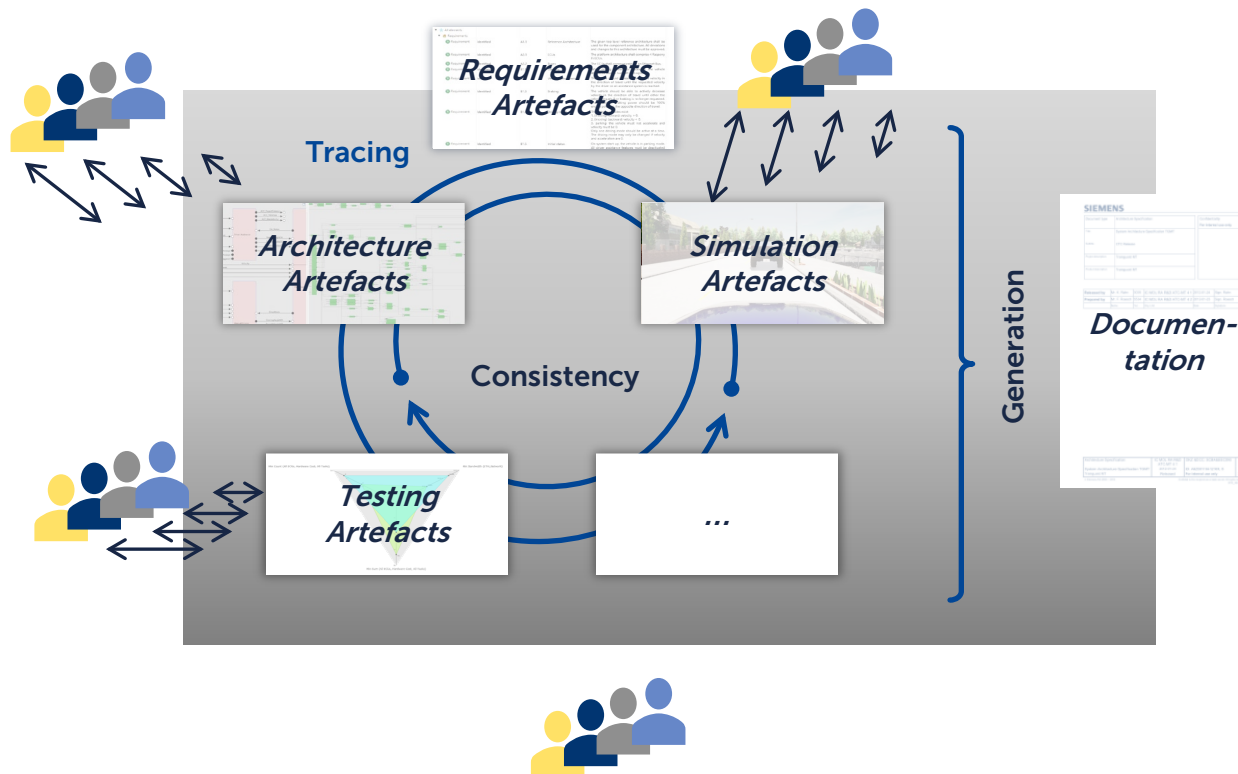
Quality Assessment in System Model Artifacts

Assessment of Model Quality in modeling artifacts



Motivation for Model Quality Assessments

The rising Complexity in Systems Engineering



Challenges:

- **High collaboration** with diverging **knowledge** of modelers
- Rising **system complexity** and **cross-domain collaboration**
- **Domain** and **project-specific** modeling **conventions** and **standards**
- **Model quality** is not **sufficiently** monitored and **communicated** in engineering lifecycle

Quality degradation over time and comprehensibility issues

Related Work & Gap

State of the Art to Maintain and Improve Model Quality

Metric-Based

Extraction of metrics and calculation of quality attributes and complexity values.

[1, 2, 3, 6, 7]

Rule-Based

Initiation of rules and syntax checks for quality assurance.

[3, 4, 7, 8]

Modeling Guidelines

Textual guidelines document modeling conventions and best practices.

[INCOSE, OMG, 4, 5, 7]

Gap: Missing cost-efficient model quality assessment approach that considers project-/domain- specific knowledge.

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4. Myron Hecht and Jaron Chen. 2021. Verification and Validation of SysML Models. INCOSE International Symposium 31, 1 (2021), 599–613. <https://doi.org/10.1002/j.2334-5837.2021.00857.x>
5. John Krogstie. 2012. Quality of Models. Springer London, London, 205–247. https://doi.org/10.1007/978-1-4471-2936-3_4
6. Christian F. J. Lange. 2007. Model Size Matters. In Models in Software Engineering, Thomas Kühne (Ed.). Springer Berlin Heidelberg, Berlin, Heidelberg, 211–216.
7. Parastoo Mohagheghi, Vegard Dehlen, and Tor Neple. 2009. Definitions and approaches to model quality in model-based software development – A review of literature. Information and Software Technology 51, 12 (2009), 1646–1669. <https://doi.org/10.1016/j.infsof.2009.04.004> Quality of UML Models.
8. <https://website.incquery.io/validator-for-enterprise-architect>

Problem Statement

Challenges of assessing model quality



Manual model **reviews** are **time** and **cost-intensive**.



Current automated approaches are **rule-based** and cannot capture **higher-level properties** such as comprehensibility.



High **expertise** needed for **project-specific** model **quality** evaluations.

Contribution

Work in Progress on a new Approach for Quality Estimation

- Method for creating automated data-driven model quality estimators based on expert ratings.
- Machine-learning pipeline for automated exploration of most accurate quality estimators.



Reduction of **model review effort**
by means of **automation**.



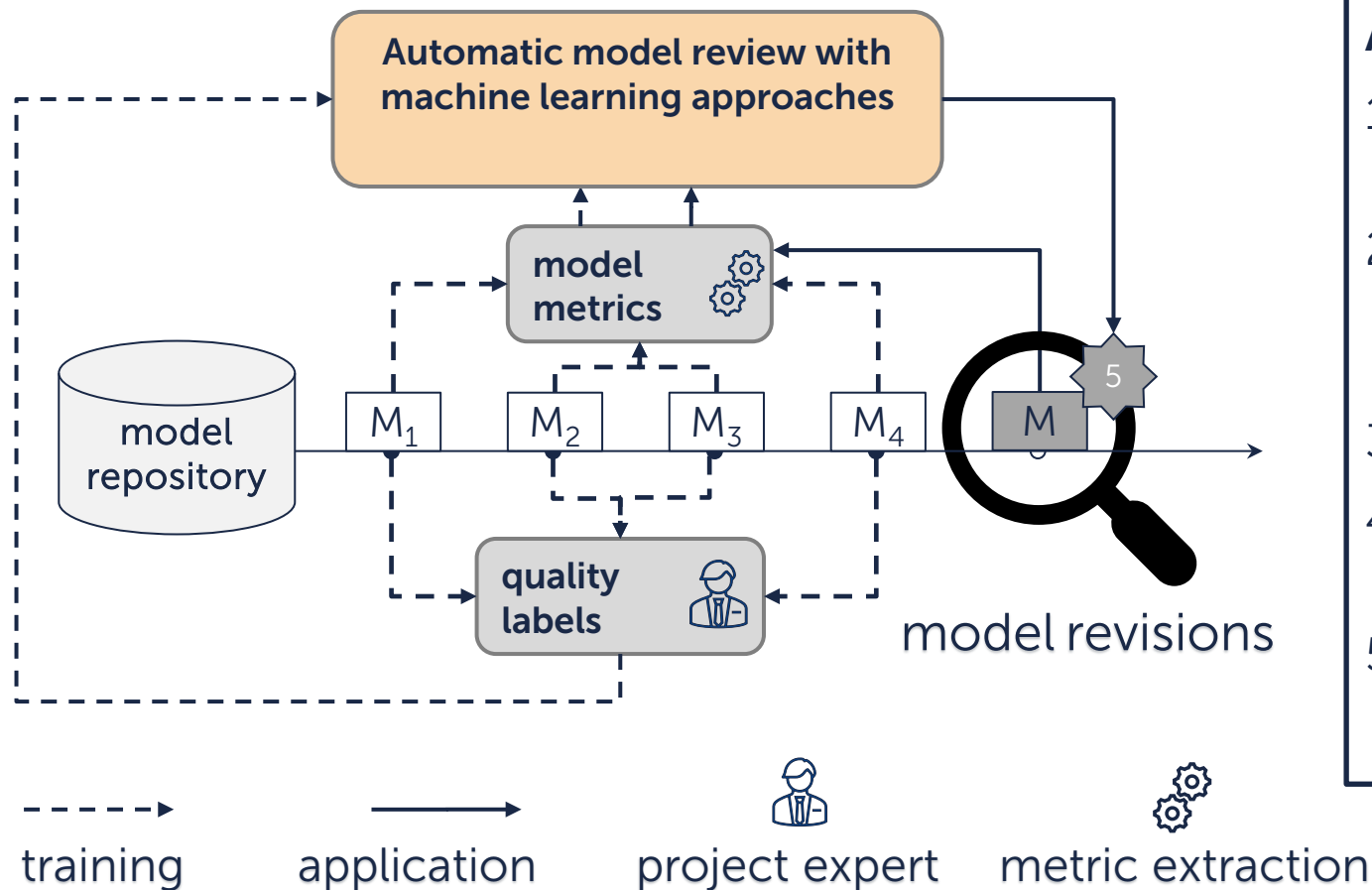
Going beyond rule-based
approaches by including
project-/domain knowledge.



Reducing expert knowledge
required to **assess model quality**.

Solution Approach

Quality Estimator Concept



Approach Steps:

1. Setup **and extract model metrics** from model version history.
2. Let **experts** decide on a **company/project specific quality catalogue**.
3. Rate **model elements** manually.
4. Store **metric** and **quality attribute** history.
5. Train and **explore estimators semi-automated** and **apply** them.

Approach: Metrics for Component-based Models

Step 1: Extract **metrics** for each revision in model **version history**

Name	Description	T.
#Contained Elements	Number of direct children of the element under review (EUR).	Element M. [11]
#Total Elements	Sum of elements in the EUR's sub-tree including itself.	
#Total Leaf Elements	Sum of leaf elements in the EUR's sub-tree.	
Nesting Level	The hierarchy level of the overall system model tree where the EUR is located.	
#Total Commentable Elements	Sum of elements that can have a comment in the EUR's sub-tree.	
#Total Commented Elements	Sum of elements that contain a comment in the EUR's sub-tree.	
#Ports	Number of input and output ports of the EUR.	Port Metrics
#Total Ports	Sum of ports in the EUR's sub-tree.	
#Total Input Ports	Sum of input ports in the EUR's sub-tree.	
#Total Output	Sum of output ports in the EUR's sub-tree.	
#Channels	Number of channels that are contained by the EUR.	Struct. M.
Clustering Coefficient	Graph-theoretic measure depicting how the EUR's direct children cluster together [17].	
Density	Ratio of #Channels in the EUR and the maximum number of channels possible [31].	
Surface Coverage	Percentage of covered area in a EUR's diagram view (within its min. bounding box) [26].	Diag.
Deviation of Channel lengths	Standard deviation of the channel lengths in a diagram view of the EUR [31].	

Approach: Quality Attributes for Component-based Models

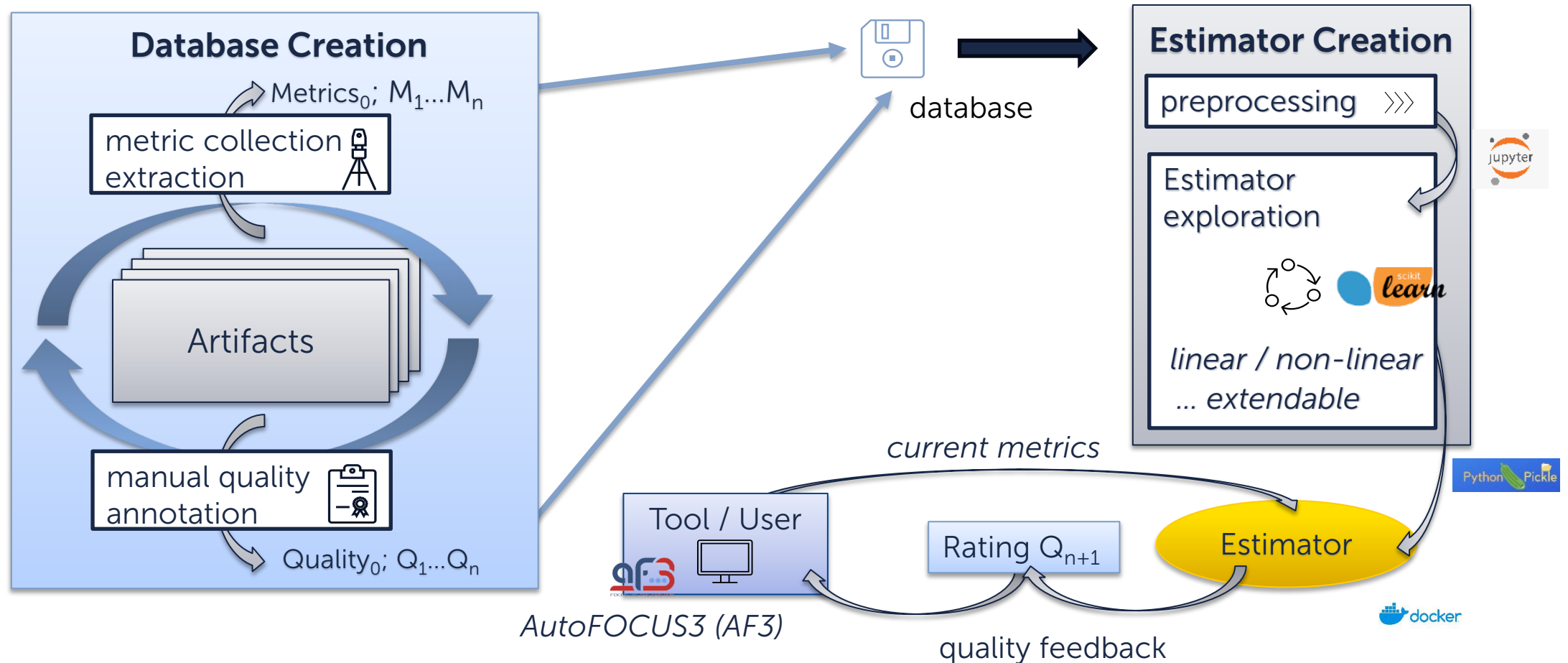
Step 2: Define **attributes** and (manually) **rate** model **elements**

Example: Quality Attributes from Case Study (focus: comprehensibility)

Name	Description	Aspects
Graphical Element Representation	Does the visualization of the element under review represents its impact and role for the system under review?	size, presentation, position, icon, ...
Graphical Data Flow	Does the graphical representation convey an intuitive understanding of the data flow?	which elements consume / produce / transform data?
Element Naming	Do the element names fit in the context of the element under review? Do they convey an intuitive understanding of the element's functionality?	name of element under review in relation to name of its constituent elements
Interface Representation	Are the port and channel names as well as the data types comprehensible in the context of the system under review?	port naming, datatype names and definitions
Abstraction Level	Is the decomposition of the element under review comprehensible? Are its constituent elements at the expected granularity level?	Aspects: assess if the children of the element under review are on a similar abstraction level

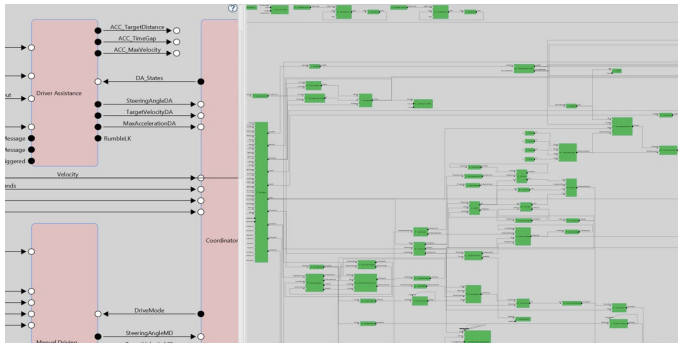
Approach: Quality Estimator Architecture

Step 4 & 5: Prototype Implementation

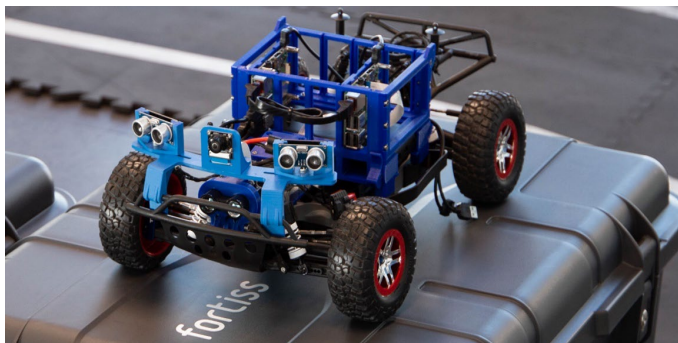


Experiment: Platform and Data

AutoFOCUS3 system models describing assisted driving functions



<https://af3.fortiss.org/>



<https://git.fortiss.org/ff1/>



Experiment Data and Context

- Software architecture models describing decomposition, data-flow and system behavior.
- 103 model versions from 14-week practical course at TU Munich
- 1837 elements rated for experiment.
- Quality ratings from 1 (bad) to 3 (good).
- Training data 80%; Test data 20%

Experiment: Data Processing and Training



Pipeline for Automatic Exploration of different Estimation Models

Linear Regression

Correlation Analysis with Pearson, Spearman & Kendall



None of our extracted metrics highly correlates with one of our annotated comprehensibility ratings.

Support Vector Machines

Grid Search with varying C and Gamma Values and Radial Basis Function kernel.



Best accuracy at reaching 0.79 on test data for the graphical element representation rating.

Random Forests

Exploration of parameter constellations resulting in high accuracy on the test data.



Best results on the test data with average accuracy of 0.94 on the test data for all attributes.

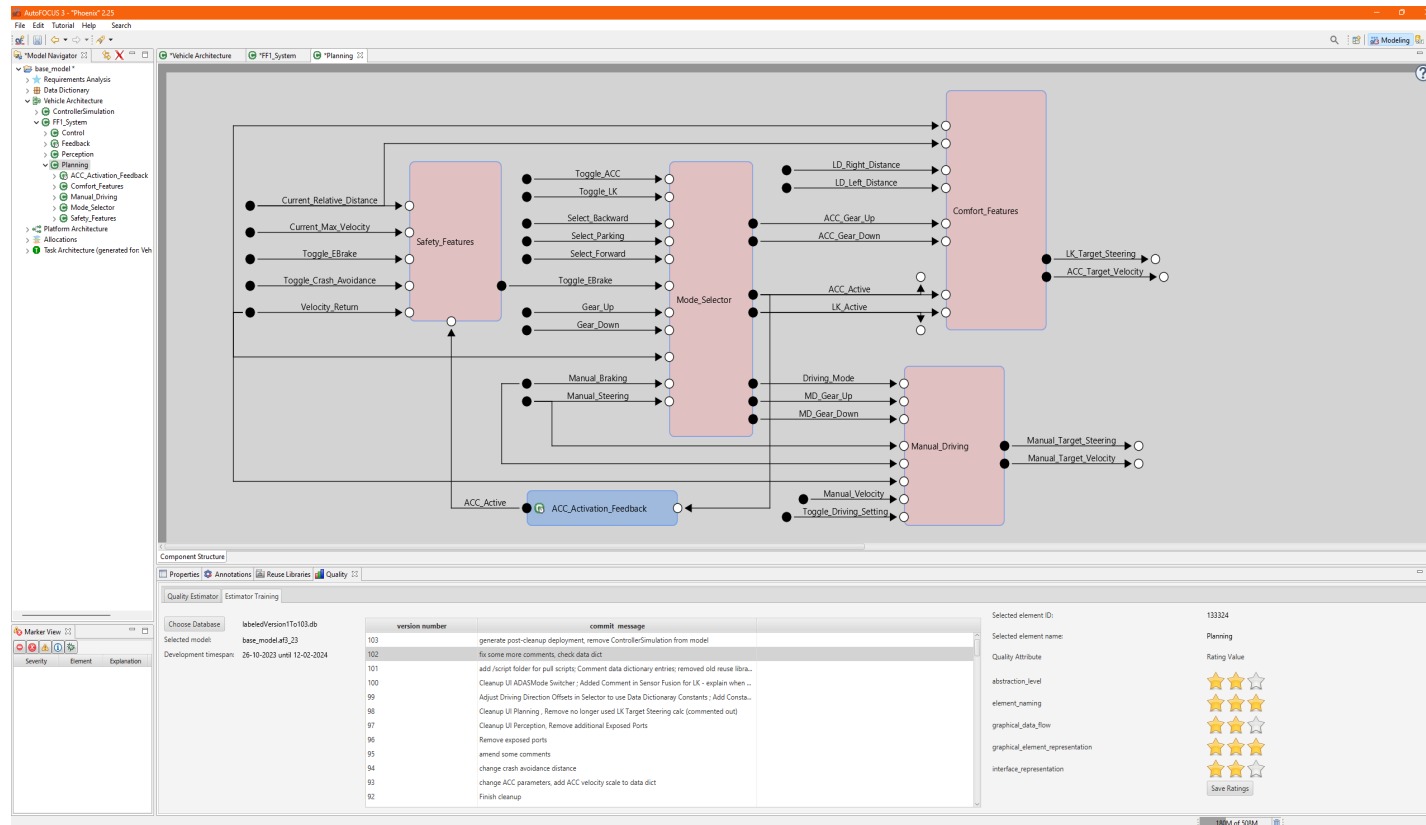
Experiment: Random Forest Results

Exploration of 3472 different Predictor Sets per Quality Attribute

Predictors	Ratings	Accuracy
Nesting Level, #Channels, Surface Coverage, Deviation of Channel Lengths	Abstraction Level	0.94
Nesting Level, #Ports, #Total Ports, Density, Surface Coverage	Element Naming	0.92
Nesting Level, #Contained Elements, #Total Elements, Deviation of Channel Lengths	Graphical Data Flow	0.94
Clustering Coefficient, #Channels, #Total Elements, Deviation of Channel Lengths	Graphical Element Representation	0.95
Nesting Level, #Total Ports, #Total Leaf Elements, Surface Coverage	Interface Representation	0.94

Prototype: Quality Estimator Tooling

AutoFOCUS 3 Quality Plugin



Prototype Functionality

- **Automated Extraction** of **model metrics** from model version history into SQLite databases.
- **Navigation module** to checkout commits and **rate** selected model elements manually.
- **Load** and **store estimators**
- **Docker service** running with API waiting for model metrics to estimate **quality attributes**.

Future Work on Model Quality Estimation

Approaches to reduce Limitations and Threats to Validity

Threats to Validity and Limitations

Data Quality

- Bias in labeling due to one person labeling the data along with low granularity in the assessment
- Initial selection of metrics originating from AF3 metamodel
- Distribution of ratings on existing model data.

Approach

- Industrial applicability

Future Work and Extensions

Data Quality

- Study with 4x the data labeled by multiple people with more sophisticated quality ratings with higher granularity
- Metric selection also developed for industry case study
- Researching ways to improve training data quality label distribution.

Approach

- Case study with industrial partner

Thank you for your Attention!

Key take aways, benefits, limits and challenges



Domain-/Project-
specific **quality**
assessment

Live quality
feedback

Saving costs by
detecting quality
degradation early

Analysis of quality
development in
model histories



Scarcity, quality
and **constraints** of
ground truth data

Causality between
metrics and
quality attributes

Adaption of
tooling to **industry**
standards

Approach **Industry**
Applicability

Contact



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Feel free to reach out to me for further
discussions.

