

NLP-Based Requirements Formalization for Automatic Test Case Generation

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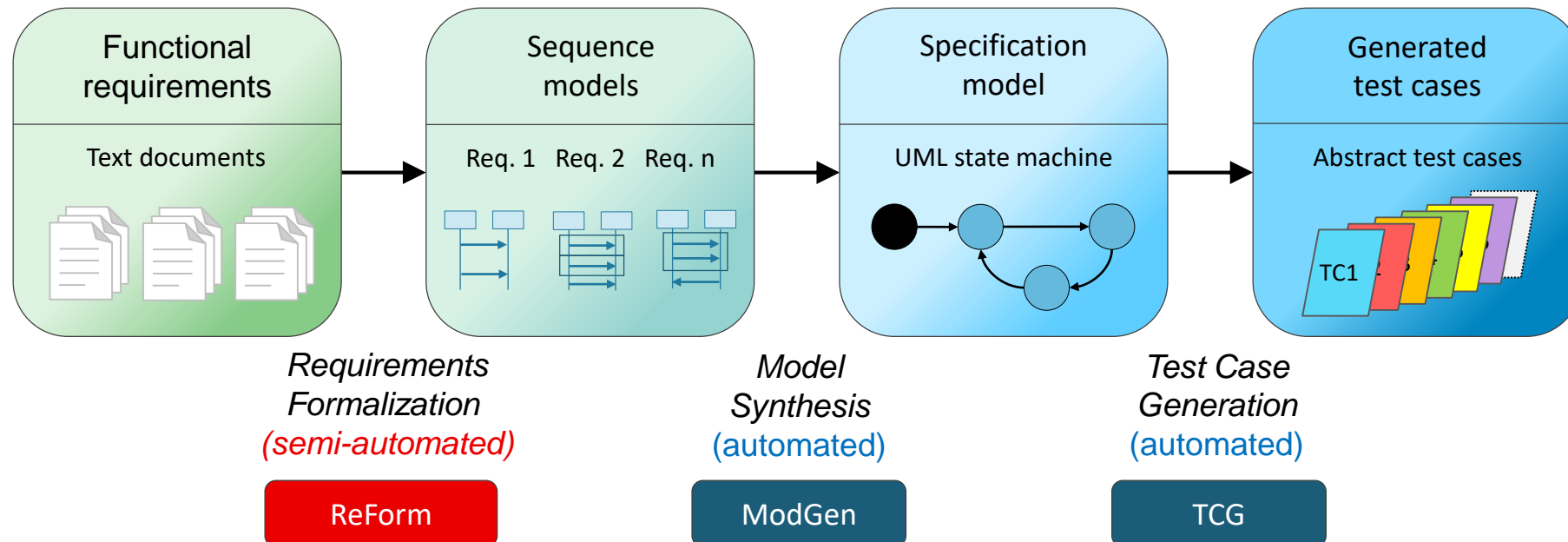
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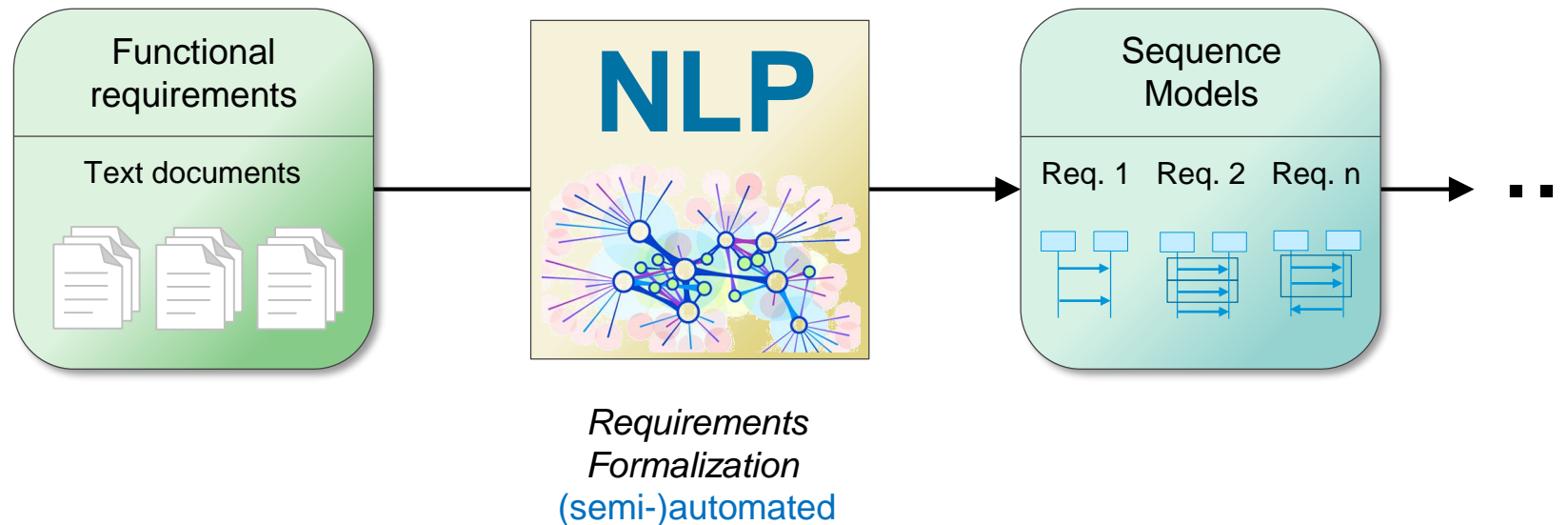
1. Introduction

- Rapidly changing and growing number of **requirements** in the life cycle of a device, component or system
- Increasing effort for verifying requirements and **testing** of the implementation
- Agile methods for **model-based testing** to manage test complexity and reduce test effort and cost
- **Toolchain** for requirements-based test case generation
- Time-consuming **manual step** is the creation of requirement models from textual requirements documents



1. Introduction

- Recent advances in **natural language processing (NLP)** show promising results
- We propose a **new, semi-automated technique for requirements-based model generation**



1. Introduction

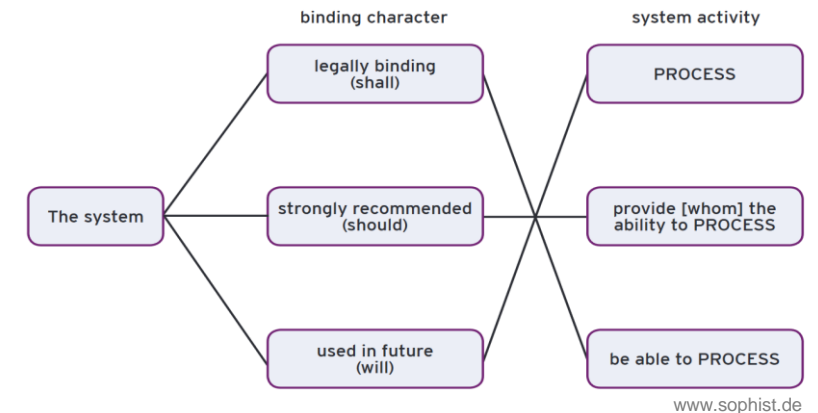
State of the art

- Several NLP approaches and tools have been investigated in recent years
- Many authors restrict their NLP approaches to a prescribed format
 - Templates
 - Set of structural rules
 - Controlled natural language
- Specific output format of generated models
 - Expert knowledge for inspection needed
 - Restricted to some requirements engineers
 - Coding practices necessary

A Comprehensive Investigation of Natural Language Processing Techniques and Tools to Generate Automated Test Cases

Imran Ahsan¹, Wasi Haider Butt², Mudassar Adeel Ahmed³ and Muhammad Waseem Anwar⁴

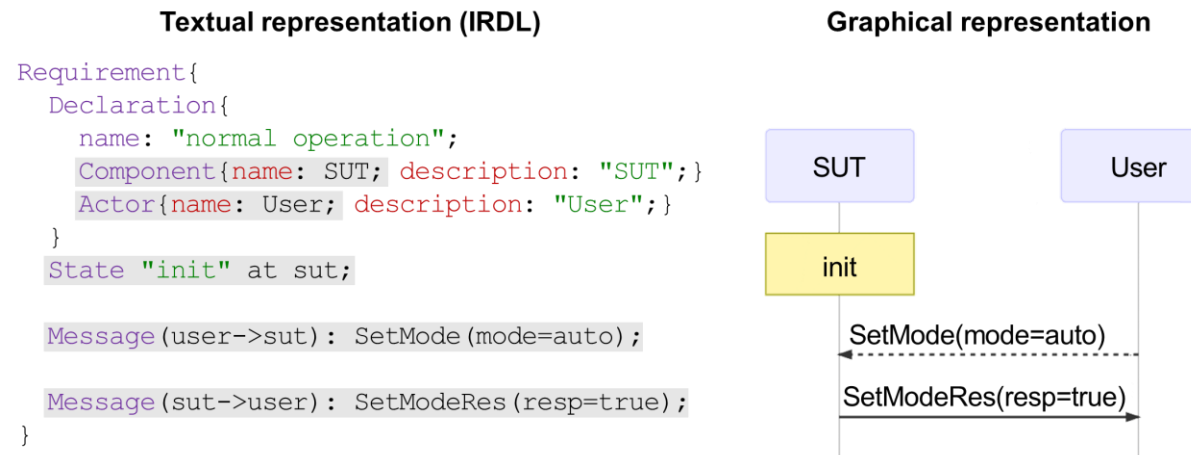
NLP-assisted software testing: A systematic mapping of the literature
Vahid Garousi^{a,*}, Sara Bauer^b, Michael Felderer^{b,c}



```
BodySense.allInstances() ->
  forAll(b|b.occupancyStatus.occupantClassFor
  AirbagControl<>OccupantClass::Error)
VoltageError.allInstances() ->
  forAll(v|v.detected = true)
BodySense.allInstances() ->
  forAll(b|b.occupancyStatus.occupantClassFor
  AirbagControl = null)
```

1. Introduction

- **Aim:** Development of a new approach and tool
 - 1) Handle an **extended range of domains and formats** of requirements
 - 2) Provide **enhanced but easily interpretable intermediate results**



- **Main contributions:**
 - 1) Development of a **rule-based approach based on NLP information**
 - 2) **Evaluation on an industrial use case using meaningful metrics**

2. Related work

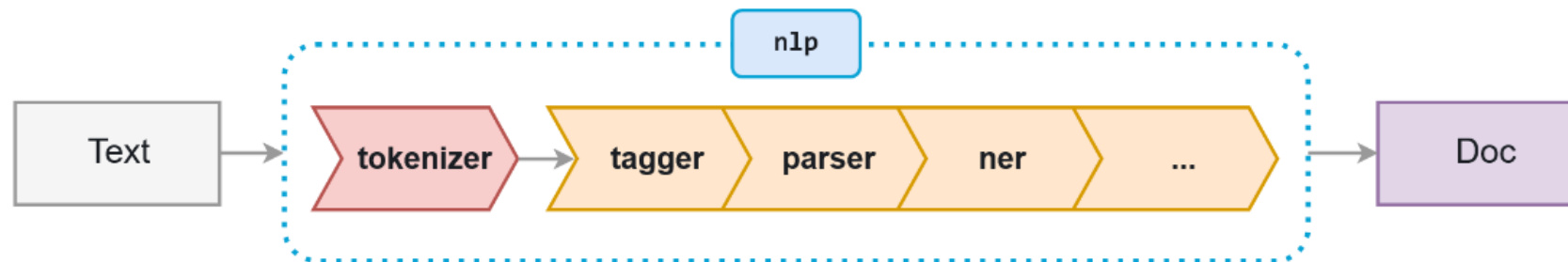
- NLP approaches restricted to a **specific domain or a prescribed format**:
 - Riebisch (2005): Creating activity diagrams from requirements following a **predefined structure**
 - Rupp (2014): SOPHIST method: formalization of structured texts, **text templates** with a defined structure
 - Mavin (2009): Small set of **structural rules** to address ambiguity, complexity and vagueness
 - Carvalho (2015): **Controlled natural language**, requirements from Data-Flow Reactive Systems (DFRS)
 - Allala (2019): Generate test cases from use cases or user stories, have to comply with a **specified format**
 - Nebut (2006): Formalization of use case descriptions by writing pre- and post-conditions in a **predefined format**
 - Goffi (2016): Complete missing test cases, provided the documentation is in a **specified template**
 - Blasi (2018): Artifacts that programmers create, belong to a **smaller subset** of specifications
- Specific output format of generated models
 - Wang (2020): Restricted Use Case Modeling (RUCM) specifications, **inspect** generated OCL constraints
 - Yue (2015): Restricted Test Case Modeling (RTCM) , **inspect** generated OCL constraints
 - Silva (2015): Test case generation using Petri Net simulation, **interpret** Colored Petri Nets
 - Fischbach (2020): Recursive dependency matching to formulate test cases from user stories

3. Methodology

■ Linguistic pre-processing

□ NLP parser: spaCy

- **Tokenization:** segment text into words, punctuations marks etc.
- **Lemmatization:** assign the base forms of words
- **Part-Of-Speech (POS) Tagging:** assign part-of-speech tags (noun, verb, etc.)
- **Dependency Parsing:** assign dependency labels (parent, children tokens)



spaCy

GET STARTED

[Installation](#)

Models & Languages

- Quickstart
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- [spaCy 101](#)
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- [New in v2.1](#)
- [New in v2.0](#)

GUIDES

- [Linguistic Features](#)
- [Rule-based Matching](#)
- [Processing Pipelines](#)
- [Vectors & Similarity](#)
- [Training Models](#)
- [Saving & Loading](#)
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- [Visualizers](#)

3. Methodology

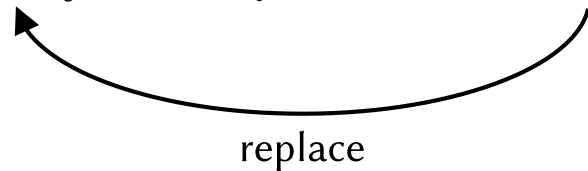
■ Linguistic pre-processing

□ Pronoun resolution

- Pronouns are identified and resolved with the farthest subject, proposed by Lappin (1994), Qiu (2004)
- Due to simplicity of the task, no need to use more sophisticated algorithms
- Resolve pronouns only if the grammatical number of the pronoun agrees with that of the antecedent
- Pleonastic pronouns (pronouns without a direct antecedent) are cited but not replaced

➤ Example:

"If the temperature of the battery is below T_{min} or it exceeds T_{max} , charging approval has to be withdrawn."



3. Methodology

■ Linguistic pre-processing

□ Decomposition

- Requirements with **multiple conditions and conjunctions** need to be mapped to individual relations
- Decomposition of complex requirements into simple clauses
 - Multiple conditions (sentences with multiple if's, while's, etc.)
 - Root conjunctions (sentences with multiple roots connected with a conjunction)
 - Noun phrase conjunctions (sentences with multiple subjects and/or objects connected with a conjunction)

➤ Example:

*"If the temperature of the battery is below T_{min} or the temperature of the battery exceeds T_{max} ,
charging approval has to be withdrawn."*

3. Methodology

■ Syntactic entity identification

- Behavioral requirements describe a particular **action** (linguistically, verb) done by an **agent** (linguistically, subject) on the system of interest (linguistically, **object**)
 - **Action:** Main action verb in the requirement
 - Nominal action, Boolean action, simple action
 - **Subjects and Objects:** Tokens with dependencies subj and obj (and their variants like nsubj, pobj, dobj, etc.)
 - Noun chunks, compound nouns, single tokens

- Resolve **logical comparisons** (e.g. greater than, exceeds, etc.)
 - **Synonym** hyperlinks from Roget's Thesaurus
 - **Map** to corresponding equality (=), inequality (!=), inferiority (<, <=) and superiority (>, >=) symbols

- Example: *"[if the temperature of the battery is below T_{min}], [charging approval has to be withdrawn]"*
 - Subject: Battery_Temperature
 - <
 - Object: Charging_Approval
 - Subject: Charging_Approval
 - Boolean action: withdrawn

3. Methodology

■ Semantic entity identification

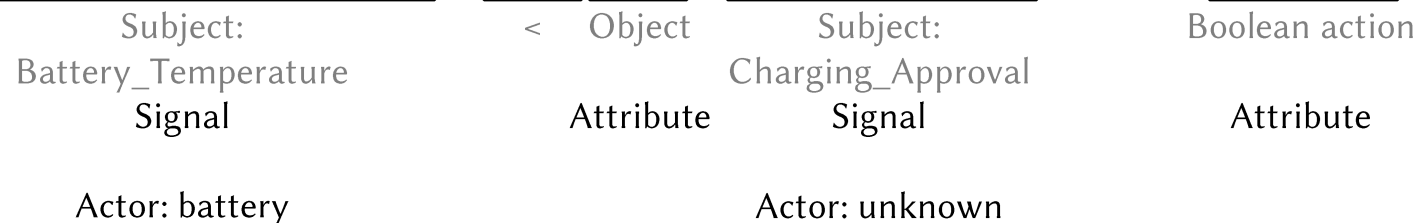
□ Mapping of syntactic to semantic entities

- **Actor or Component:** Participants involved in an interaction
- **Signal:** Interaction between different participants
- **Attributes:** Variables holding the status of an interaction
- **State:** Initial, intermediate and final states of an interaction

Syntactic entities	Semantic entities
Action	Signal
Action constraints	Attributes
Subject / Object	Actor / Component

□ More complex rules used (including action and verb types)

➤ Example: *"[if the temperature of the battery is below Tmin], [charging approval has to be withdrawn]"*



3. Methodology

■ Transformation to requirement model

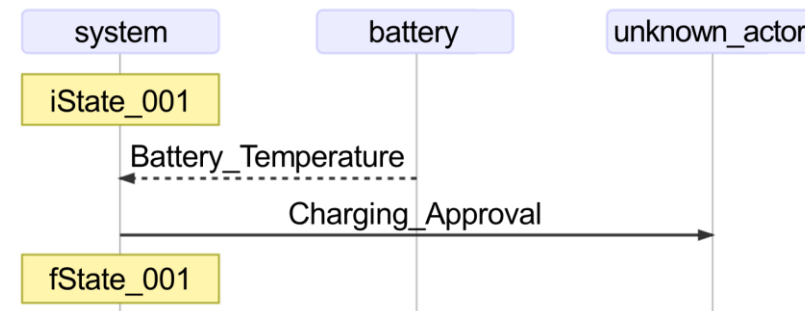
- Textual representation: ifak requirements description language (IRDL)
 - Simple text-based domain-specific language (DSL)
 - On the basis of UML sequence diagrams, a series of model elements and structures are defined
- Graphical representation: Generated sequence diagram
- Create IRDL relations for each clause and then combine them together
 - Incoming messages: SUT receives these messages provided the guard expression evaluates to be true
 - Outgoing messages: SUT sends these messages to other interaction participants

➤ Example:

Textual representation (IRDL)

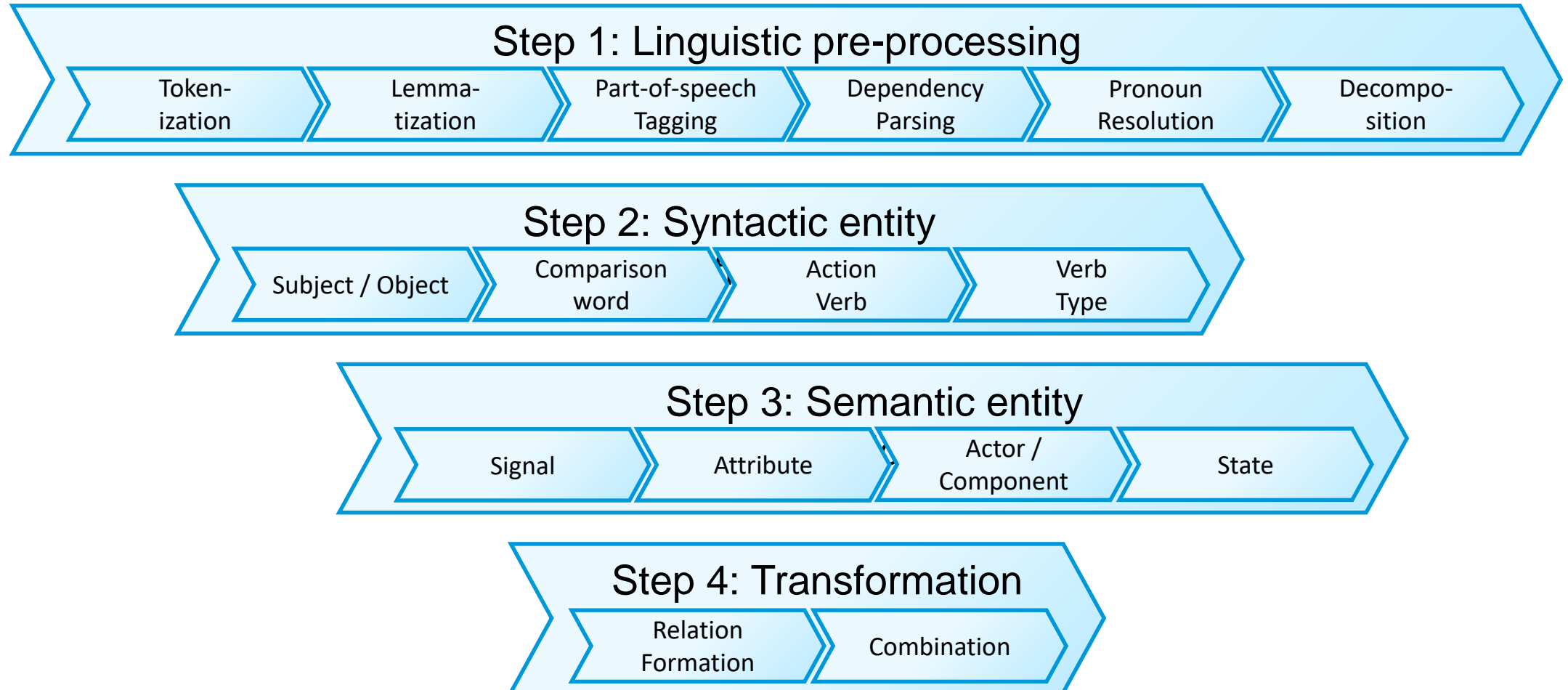
```
State iState_001 at system;  
Check(battery->system) :Battery_Temperature  
  [msg.value < Tmin || msg.value > Tmax];  
Message(system->unknown_actor) :  
  Charging_Approval(false);  
State fState_001 at system;
```

Graphical representation



3. Methodology

- Steps in our NLP approach:



4. Application

- Industrial use case from the e-mobility domain defined by AKKA
- Charging approval system of an electric vehicle with a charging station
- Aims to provide a typical **basic scenario** and development workflow in software development for an automotive electronic control unit (ECU)
- Defining **requirements**, using model-based software development and deploying the functionality on an ECU
- Function “charging approval” implements a simple function, decides upon specific input signals, if the charging process of the battery is allowed or not
- For example, charging approval is given or withdrawn depending on the
 - battery temperature, voltage or state of charge
 - the requested current is adjusted according to the battery temperature
 - error behavior is handled for certain conditions



5. Results

■ Evaluation metrics

- R set of textual requirements
- X_r set of expected artifacts (semantic entities) for a requirement $r \in R$
- Y_r set of generated artifacts (semantic entities) for a requirement $r \in R$
- $X = \bigcup_{r \in R} X_r$ set of expected artifacts in all requirements
- $Y = \bigcup_{r \in R} Y_r$ set of generated artifacts in all requirements

□ Definition of **metrics** to measure the performance of the method:

- 1) **Completeness:** For an individual requirement, this metric denotes the number of expected artifacts $x \in X_r$ for which a corresponding (not necessarily identical) generated artifact $y \in Y_r$ exists, in relation to the total number of expected artifacts $|X_r|$.
- 2) **Correctness:** For an individual requirement, this metric denotes the number of generated artifacts $y \in Y_r$ for which a corresponding, semantically identical (up to naming conventions) expected artifact $x \in X_r$ exists, in relation to the total number of generated artifacts $|Y_r|$.
- 3) **Consistency:** This metric denotes the number of generated artifacts $y \in Y$ for which a corresponding expected artifact $x \in X$ exists and is used identically in all requirements $r \in R$, in relation to the total number of generated artifacts $|Y|$.

5. Results

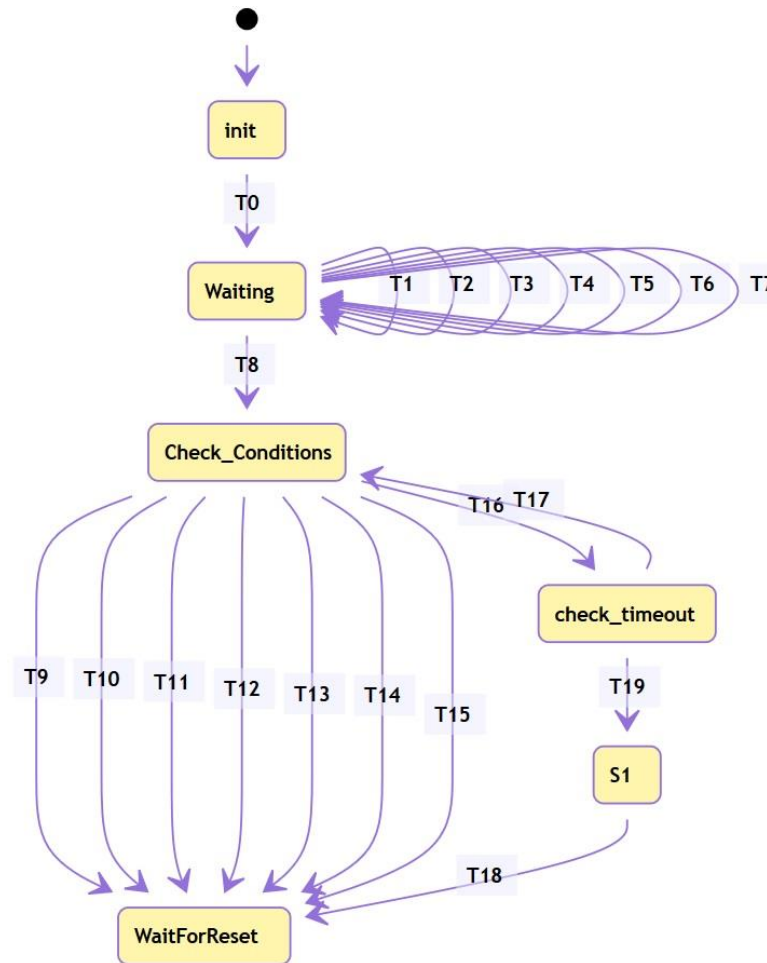
- **Evaluation** of the algorithm on the charging approval system
 - Charging approval SUT is described by **14 separate requirement statements**
 - Domain knowledge: **Predefined list** of signals, attributes, etc. or integrated by direct **user interaction** from an expert with knowledge about the system

	without domain knowledge		with domain knowledge	
	macro avg.	micro avg.	macro avg.	micro avg.
Completeness	78.2%	79.8%	81.4%	84.1%
Correctness	74.9%	78.8%	78.3%	82.1%
Consistency	94.1%		96.4%	

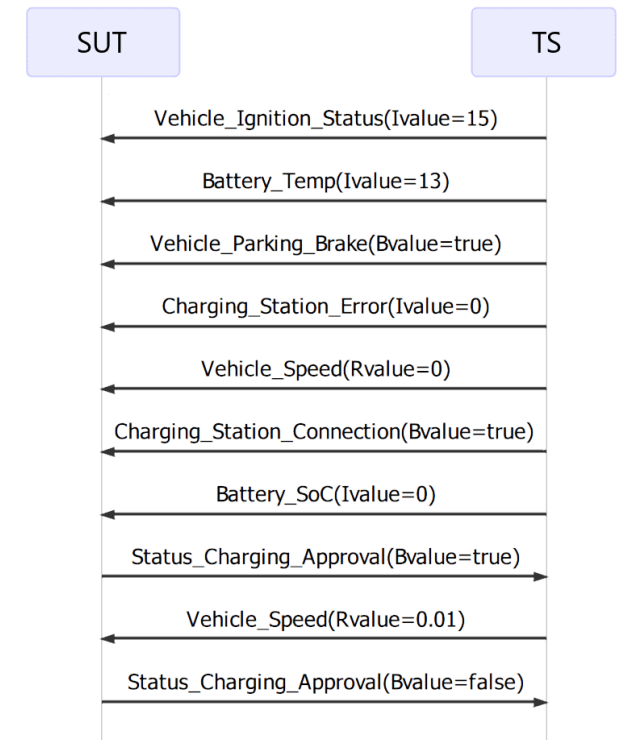
- Method shows **good results**, most of signals and other artifacts were detected **correctly and completely**
- Having a list of artifact declarations in advance produces even more accurate predictions
- Should **save a lot of time**, manual creation of sequence diagrams takes a lot of time by reading documentations and having discussions, to create the logical structure and to add all the details

5. Results

- **Model synthesis and test generation**
 - Requirement models as input for model synthesis using ifak's prototypical tool ModGen
 - UML state machine
 - 6 states
 - 20 transitions with appropriate signals, guards and actions
 - ifak's prototypical tool TCG with coverage criteria "all-paths"
 - 73 test cases



UML state machine of charging approval



Exemplary test case

6. Conclusion

- **NLP-based method** for machine-aided model generation from textual requirements
 - Cover a **wide range** of requirements formulations without being restricted to a specific domain or format
 - Generated requirement models are given in a **user-friendly, comprehensible** representation
- **Evaluated** our approach on the industrial use case of a battery charging approval system
 - Produce **complete, correct and consistent** artifacts to a high degree
 - **Reduce the human effort** of creating test cases from textual requirements
- **Outlook:**
 - **Refine** the rule-based approach further, reducing the need for manual modifications
 - Identify **semantic entities** by
 - Training of a Named Entity Recognition (NER) algorithm: Intensive labelling work
 - Semantic Role Labeling (SRL): Pre-trained models available
 - Consider **further application domains**, such as in rail, industrial communication and automotive

Acknowledgments

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Thank you for your attention!

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