

DeNom

A Tool to Find Problematic Nominalizations using NLP

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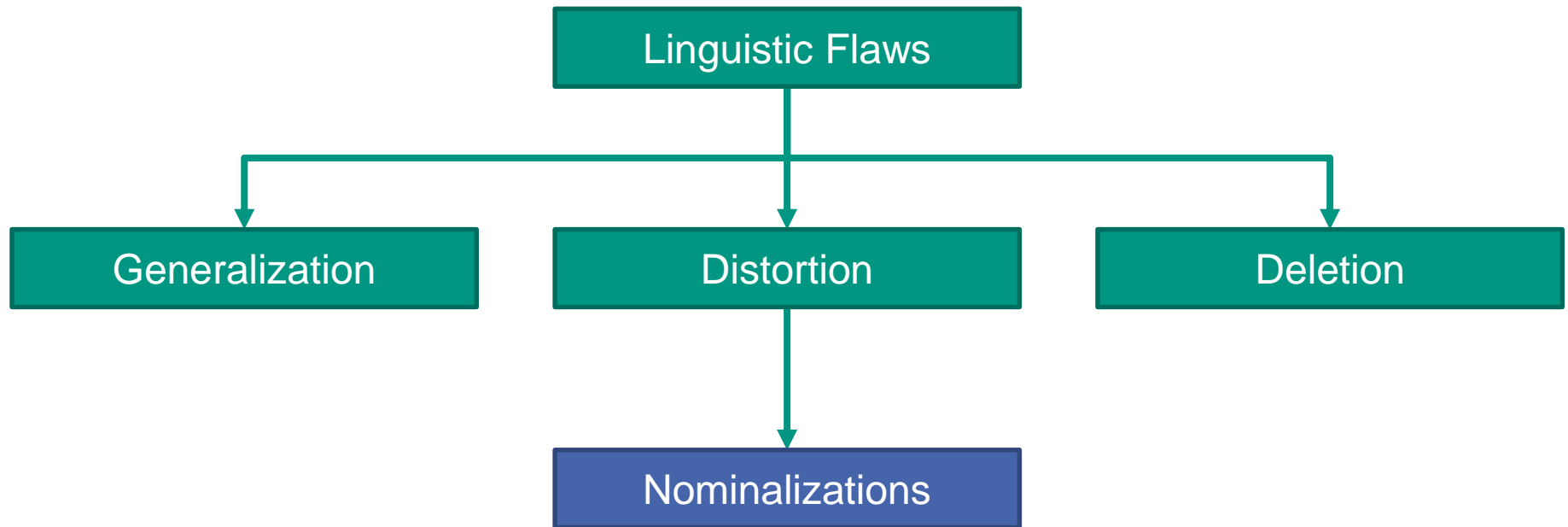


ENGINEERS

are no

poets

Linguistic Flaws in Requirements



... **return** of pallets ...

Who?

How?

To whom?

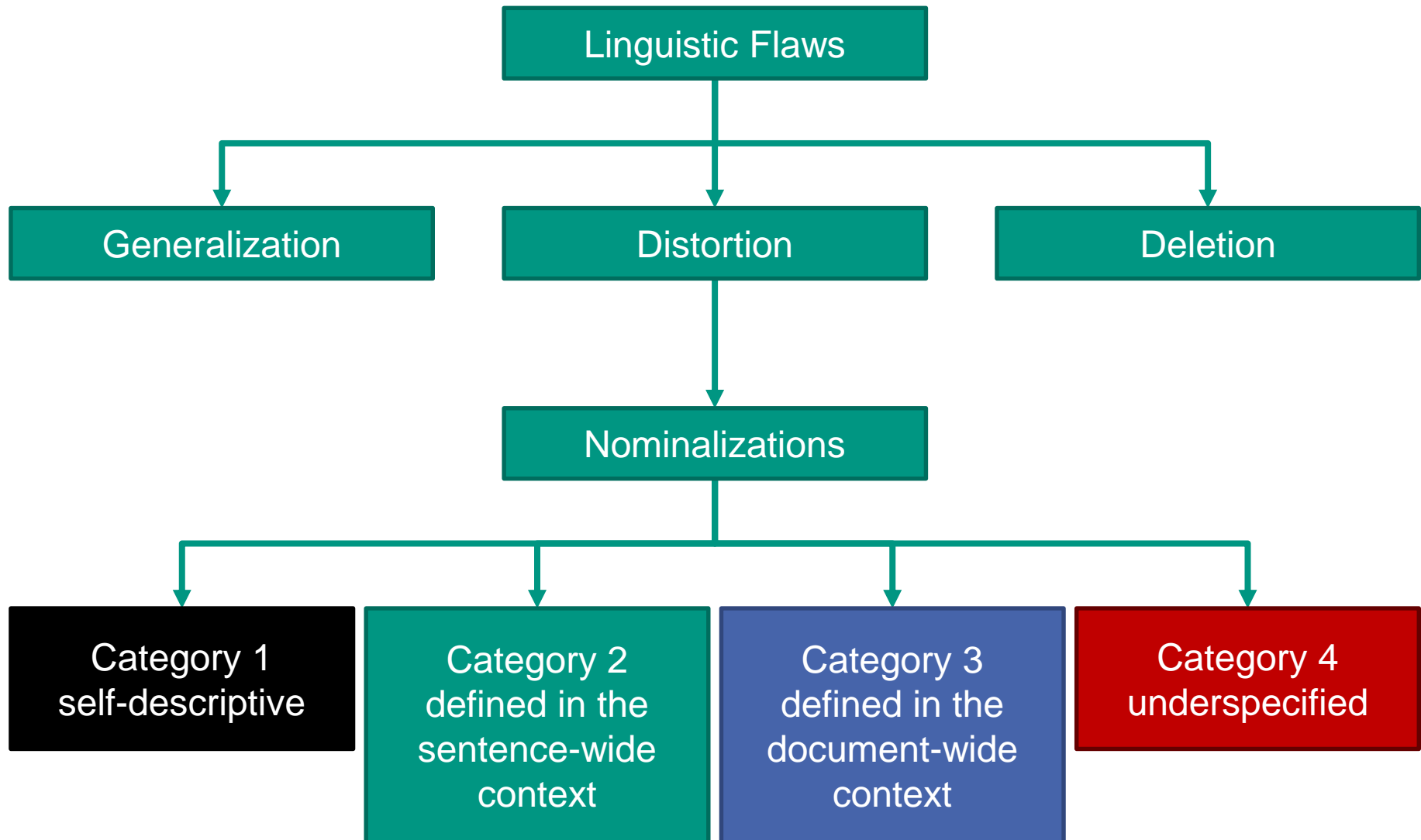
Nominalizations: Problematic yet often overlooked

- Nominalizations
 - are a problem often overlooked
 - can lead to serious problems during development
- A requirements engineer's writing rule:
Though shall not use nominalizations!
 - Ok 😊
 - Solves the problem...
- Inspection rule: Find and eliminate all nominalizations!
 - Can be identified automatically using RESI [RESI]
 - RESI is picky and produces many warnings
 - Effort too high for real-world scenarios [RESI@Automotive]

Research Question 1

ARE ALL NOMINALIZATIONS PROBLEMATIC?

Linguistic Flaws in Requirements

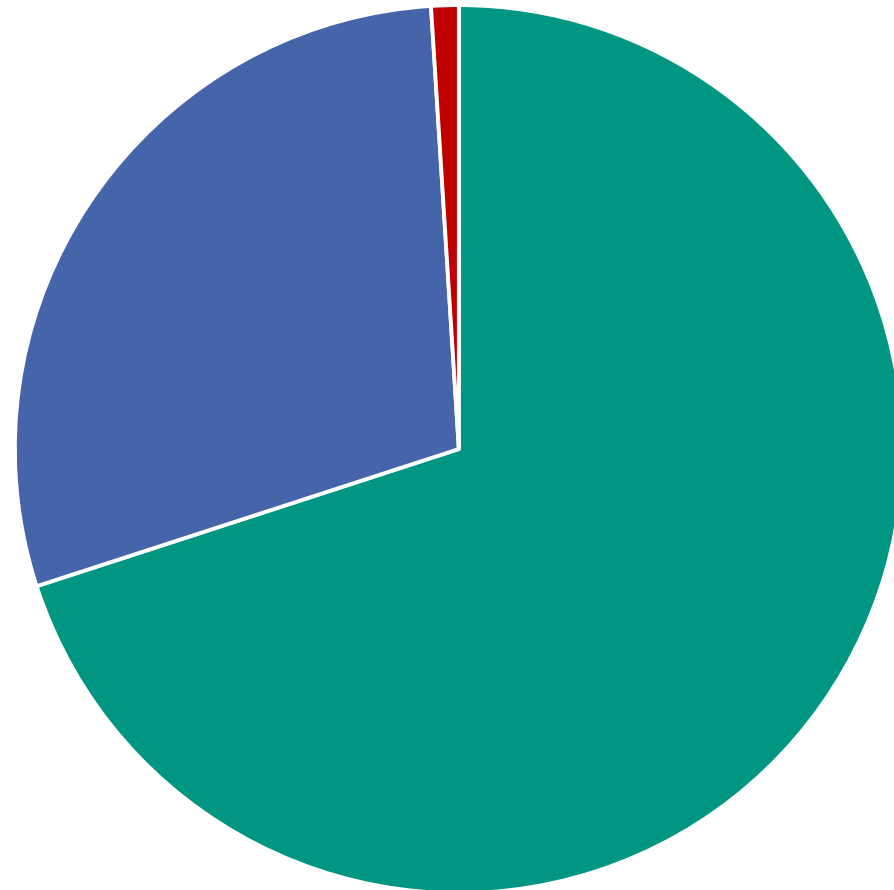


Results of the (Manual) Preparatory Study

Nominalizations

5 specifications
>40,000 words
356 nominalizations in total

0 % Category 1 (!)	■
70 % Category 2	■
29 % Category 3	■
1 % Category 4	■

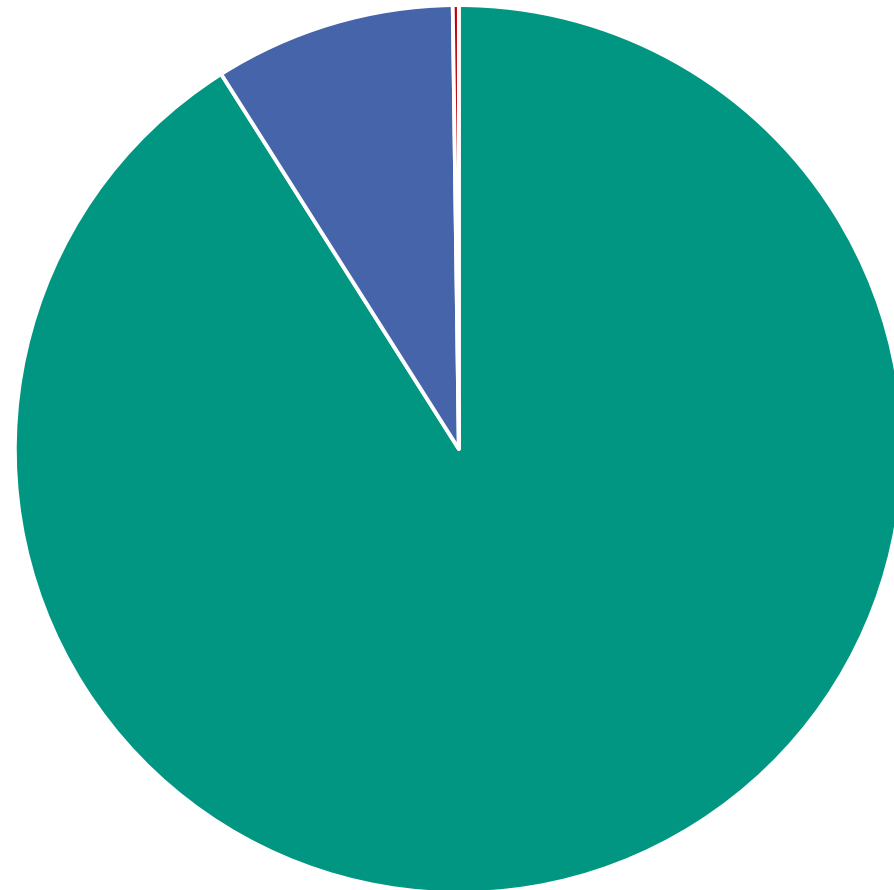


Results of the (Half-Automated) Preparatory Study

Nominalizations

6 specifications
>33,000 words
499 nominalizations detected

0 % Category 1 (!) ■
83 % Category 2 ■
8 % Category 3 ■
0.2 % Category 4 ■
+ some false positives



Research Question 2

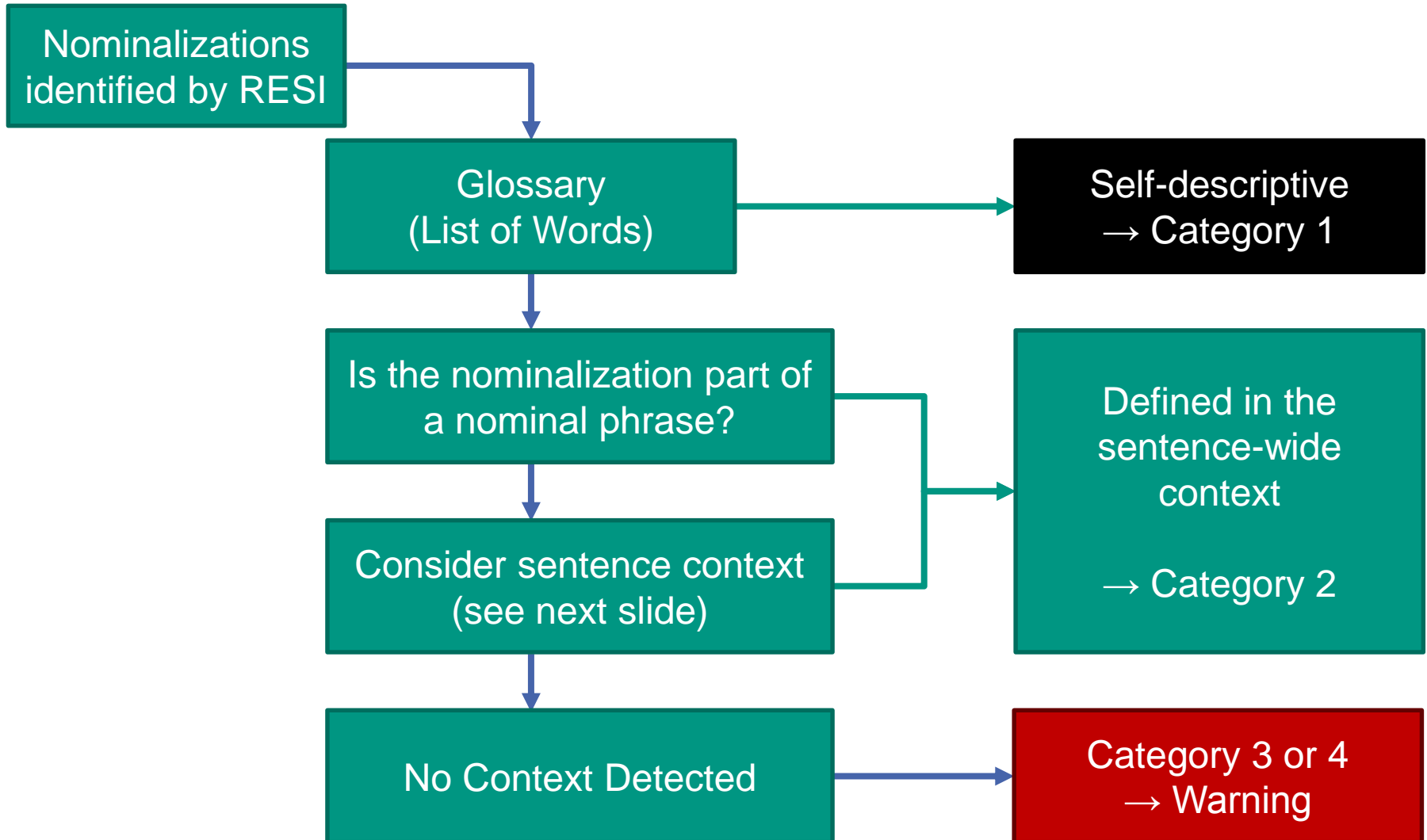
CAN WE IDENTIFY *ACCEPTABLE* NOMINALIZATIONS AUTOMATICALLY?

Working with RESI and DeNom

- When working with RESI, the user
 - must import the specification from a text file and preprocess it
 - connect to an ontology
 - choose the inspection rule(s)
(nominalizations, ambiguous words, complete process words, determiners)
 - carry out the inspection

- DeNom automates most of the repetitive work
 - Import of specifications from DOORS
 - Preprocessing
 - Connecting and configuring the ontology

DeNom's Workflow



How do we do it?

Category 2: Defined in the intra-sentence context

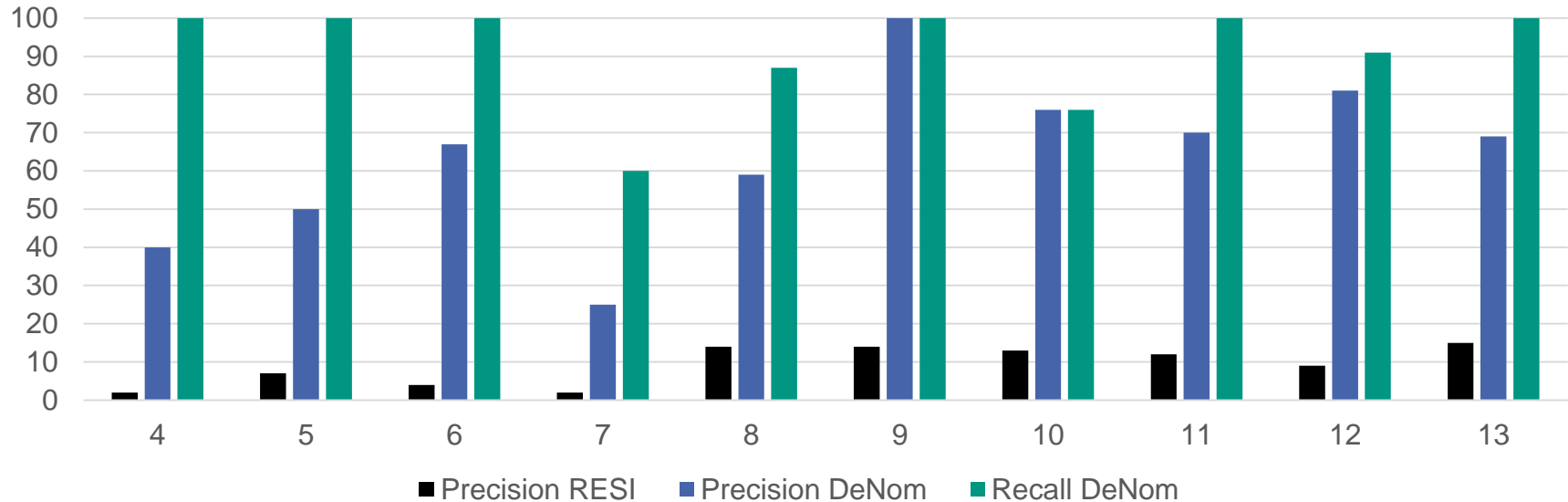
- DeNom uses RESI's completeness test for process words (i.e. verbs)
 - Transform the nominalization to the corresponding verb
 - Ask the ontology which arguments are needed by the verb
 - Check if the arguments are given in the sentence

■ ... return of pallets...

- Verb frame for return:
(and

(objectGiven :ACTION :OBJECT) ← what?
(isa :ACTION ReturningSomething)
(giver :ACTION :SUBJECT) ← who?
(givee :ACTION :OBLIQUE-OBJECT))) ← to whom?

Evaluation



- 10 specifications, >59,000 words
- 1,136 nominalizations
 - only 84 of them are problematic
 - DeNom shows 129 warnings
- Precision of RESI on average: 8% ($F_1=15\%$)
- Precision of DeNom on average: 65% (with a recall of 88%, $F_1=75\%$)

Conclusions and Future Work

- Improved user-experience
 - Improved performance (precision: 8% → 65%)
 - Less manual work
- Improve DeNom
 - Consider document wide context
 - Only one of RESI's four automatic inspections considered
 - Further reduce user interaction
- Large scale evaluation
 - In situ evaluation @Daimler
 - Integrate DeNom into Daimler's requirements inspection tool

References

- [Rupp] C. Rupp, *Requirements-Engineering und -Management. Professionelle, iterative Anforderungsanalyse für die Praxis*, 5., aktualisierte und erw. Aufl. Hanser Fachbuchverlag, 2009.
- [RESI] S. J. Körner and T. Brumm, “Natural Language Specification Improvement with Ontologies,” *International Journal of Semantic Computing (IJSC)*, vol. 03, no. 04, pp. 445–470, 2010.
- [PVMyth] J. Krisch and F. Houdek, “The Myth of Bad Passive Voice and Weak Words -- An Empirical Investigation in the Automotive Industry,” in *RE'15:23rd IEEE International Requirements Engineering Conference*, Ottawa, Ontario, Canada, 2015.
- [RESI@Automotive] S. J. Körner, M. Landhäußer, and W. F. Tichy, “Transferring Research Into the Real World: How to Improve RE with AI in the Automotive Industry,” in *1st International Workshop on Artificial Intelligence for Requirements Engineering*, 2014.

Figure 4

An overview of DeNom's processing steps

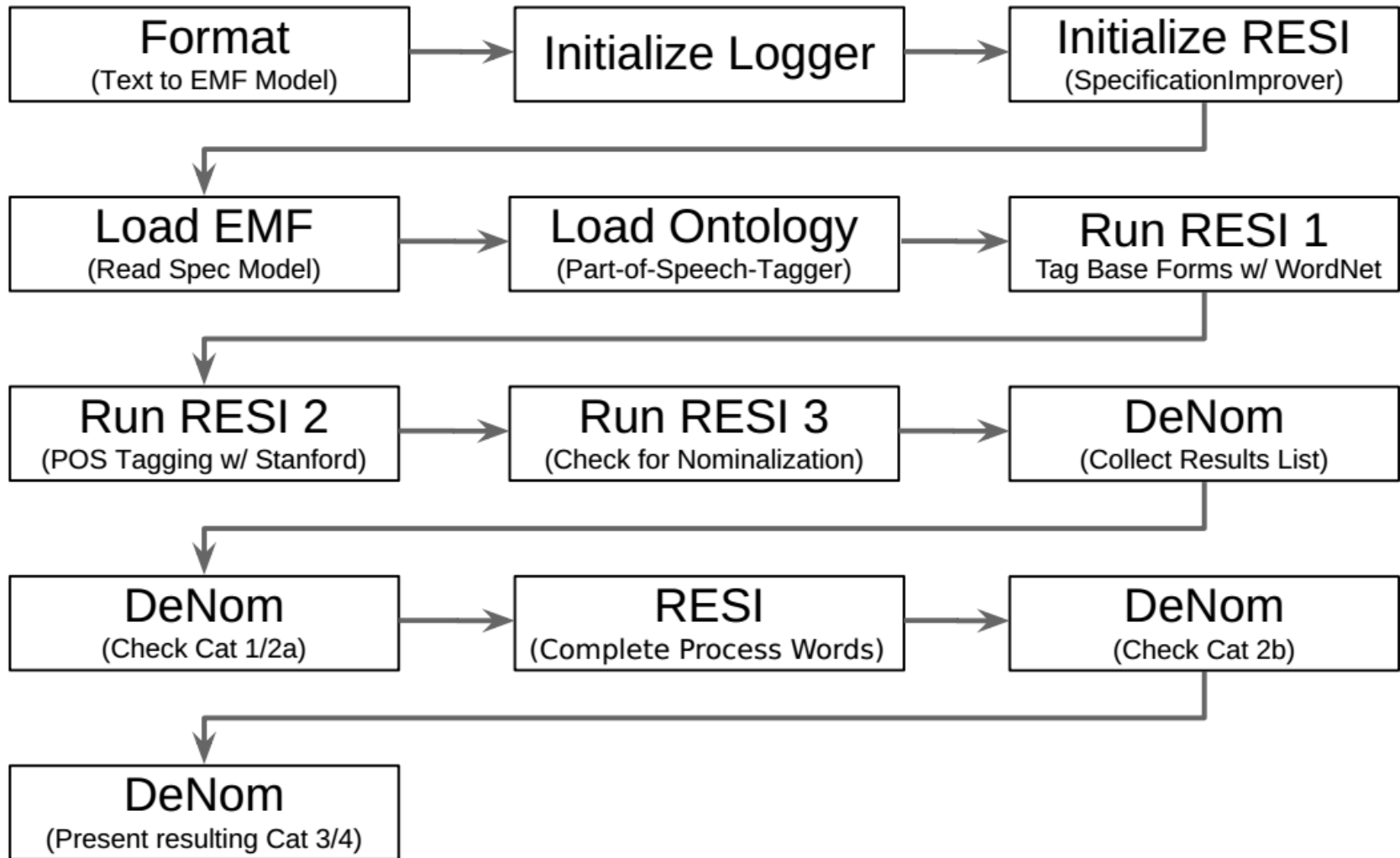


Table 1
Results of the Manual Preparatory Study

	Words	Nom.	Cat. 1	Cat. 2	Cat. 3	Cat. 4
SRS 1	9,942	85	0.0%	70.6%	29.4%	0.0%
SRS 2	23,104	158	0.0%	59.5%	36.7%	2.5%
SRS 3	2,129	21	0.0%	81.0%	14.3%	0.0%
SRS 4	3,687	62	0.0%	95.2%	4.8%	0.0%
SRS 5	1,598	30	0.0%	56.7%	43.3%	0.0%
Sum	40,460	356	0	247	102	4
			0.0%	69.4%	28.7%	1.1%

Table 2
RESI's Results in the Preparatory Study

	Words	Nom.	Cat. 1	Cat. 2	Cat. 3	Cat. 4
SRS 3	2,158	25	0.0%	88.0%	8.0%	0.0%
SRS 4	3,687	56	0.0%	83.9%	1.8%	0.0%
SRS 5	1,598	15	0.0%	73.3%	6.7%	6.7%
SRS 6	6,069	116	0.0%	91.4%	5.2%	0.0%
SRS 7	12,581	133	0.0%	84.2%	5.3%	0.0%
SRS 8	7,403	154	0.0%	74.7%	16.2%	0.0%
Sum	33,496	499	0	413	42	1
			0.0%	82.8%	8.4%	0.2%

Table 3

Results of the Evaluation for DeNom (D) and RESI (R). We calculated DeNom's recall with respect to RESI's results.

Doc.	# Words	Nom.		probl. Nom.	Precision		Recall D
		R	D		R	D	
SRS 4	3,687	81	5	2	2%	40%	100%
SRS 5	1,598	15	2	1	7%	50%	100%
SRS 6	6,069	108	6	4	4%	67%	100%
SRS 7	12,580	246	12	3	2%	25%	60%
SRS 8	7,403	167	34	20	14%	59%	87%
SRS 9	2,923	28	4	4	14%	100%	100%
SRS 10	8,098	130	17	13	13%	76%	76%
SRS 11	2,590	57	10	7	12%	70%	100%
SRS 12	10,444	243	26	21	9%	81%	91%
SRS 13	4,094	61	13	9	15%	69%	100%
Sum	59,486	1,136	129	84	8%	65%	88%