A Systematic Approach of Dataset Definition for a Supervised Machine Learning using NFR Framework

Matheus Marinho, Danilo Arruda, Fernando Wanderley e Anthony Lins



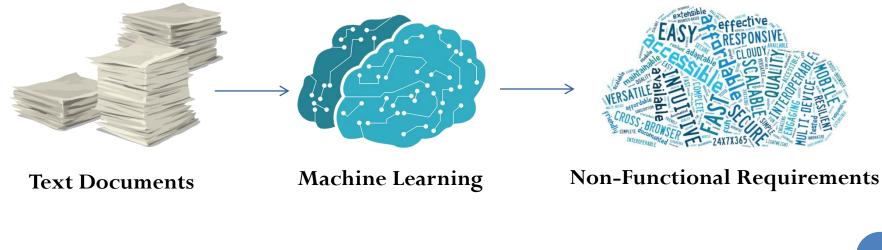
Universidade Católica de Pernambuco Centro de Ciências e Tecnologia



Introduction

Non-Functional Requirement elicitation is an important activity in the requirements engineering process, leading to determines the success or the failure of systems.

Several techniques have been proposed to identify requirements from text documents

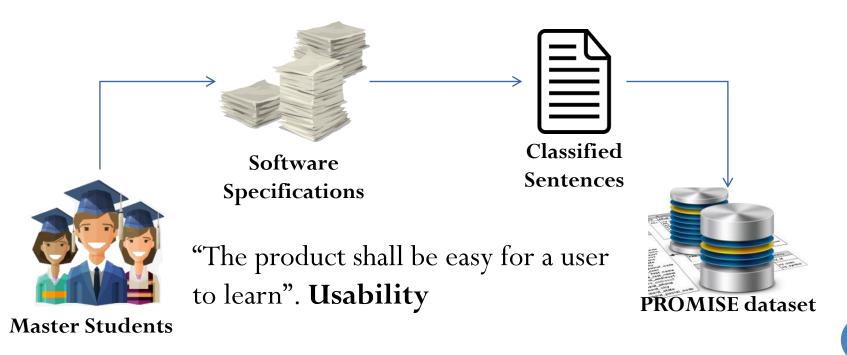


Context

An important requirements dataset: PROMISE

Was built from a research using software specifications done by master's students from DePaul University.

The PROMISE have been used as a good source of knowledge to Machine Learning training.

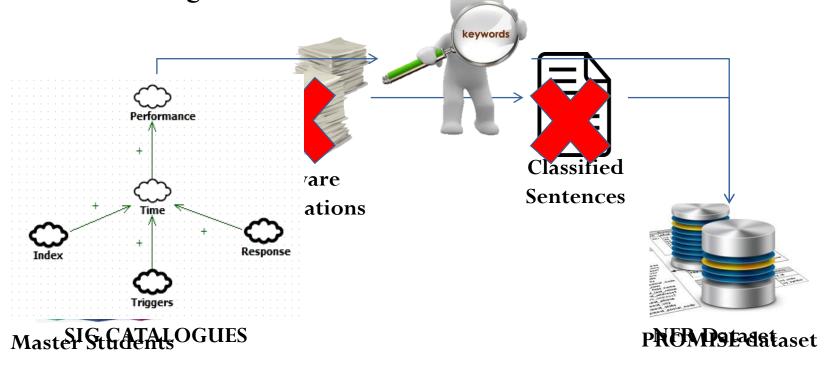


Main Goal

However, the process of generating the dataset **depends** on **people**

Our target is **reduce** this dependence

Through a **systematic process** for a dataset generation **using** keywords extracted from **SIG Catalogues**



Research Questions

To evaluate our approach, we defined the following research questions:

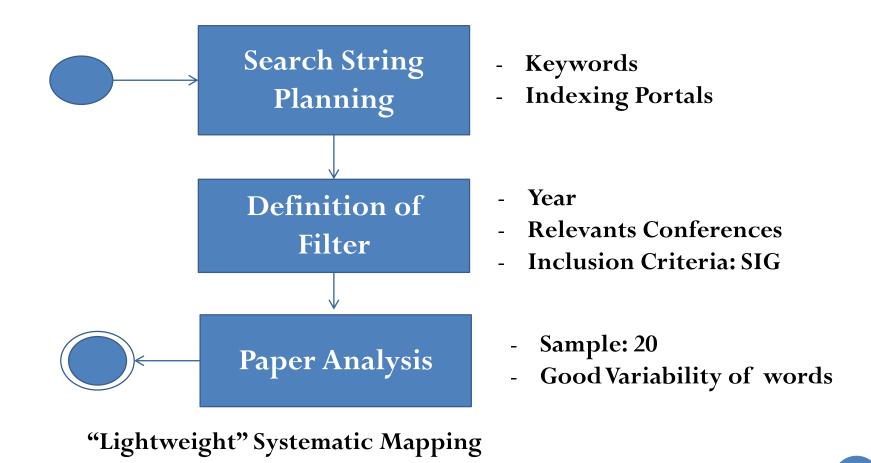
RQ1– Can we extract keywords from SIG catalogues and use these to generate a dataset to classify non-functional requirements?

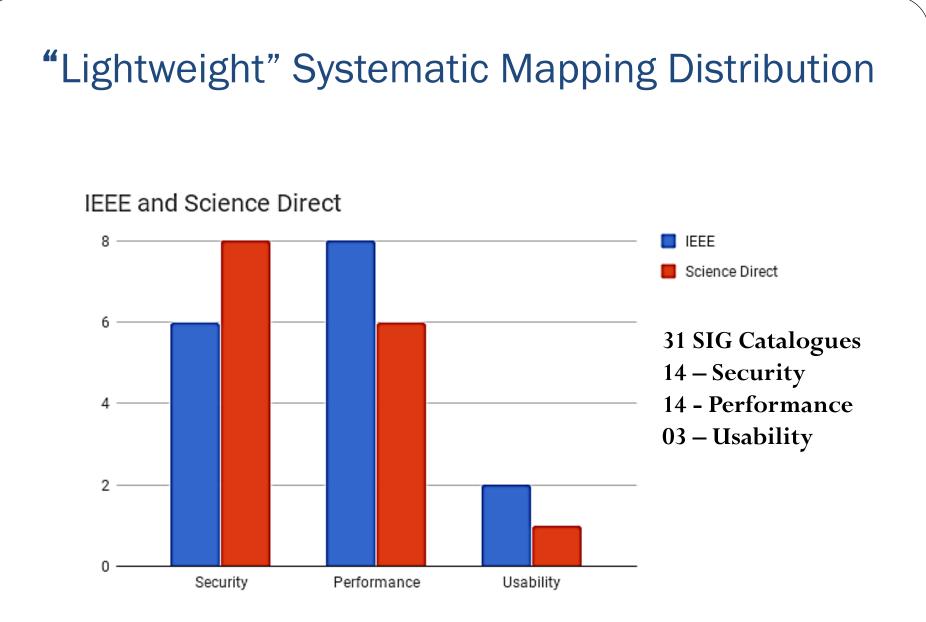
RQ2 – Can we define a systematic process for dataset generation through NFR Framework?

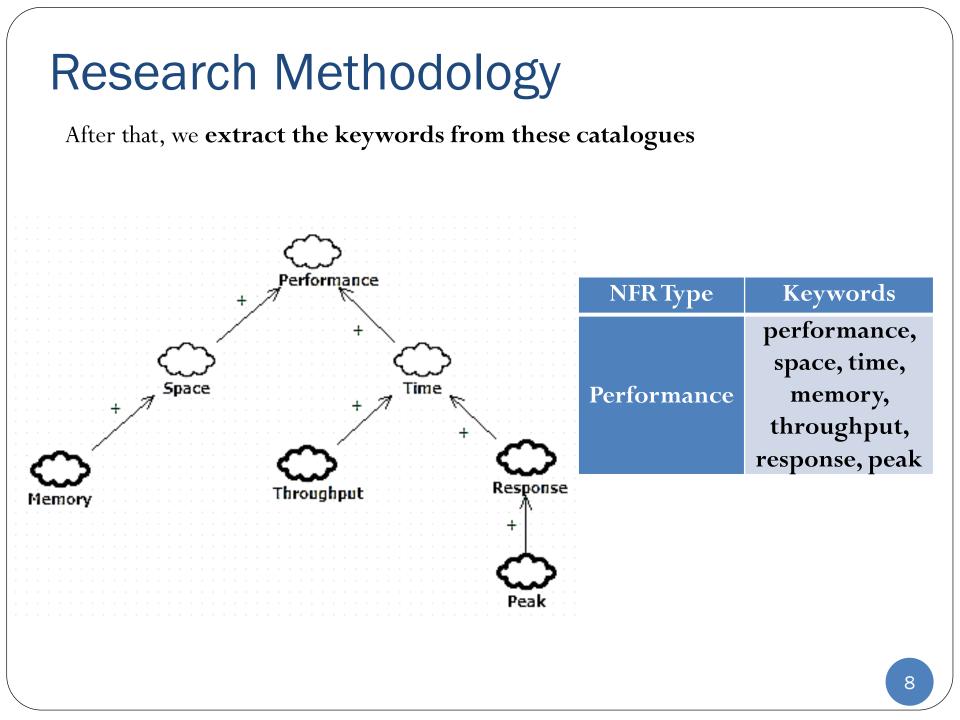
RQ3 – How well can we automatically classify non-functional requirements using this systematic process?

Research Methodology

To reach our goal we **first** propose a way to **find SIG catalogues**

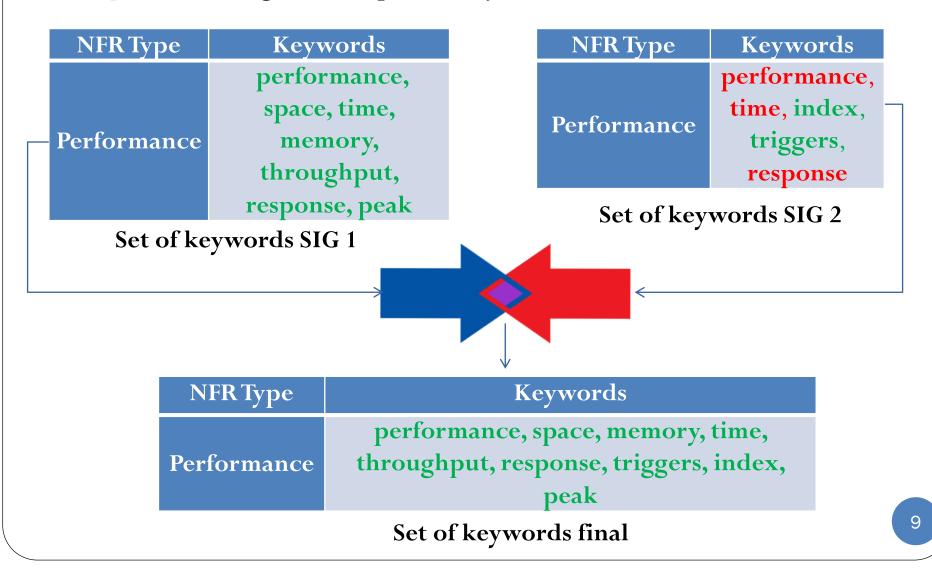






Research Methodology

And perform a merge of the repeated keywords.

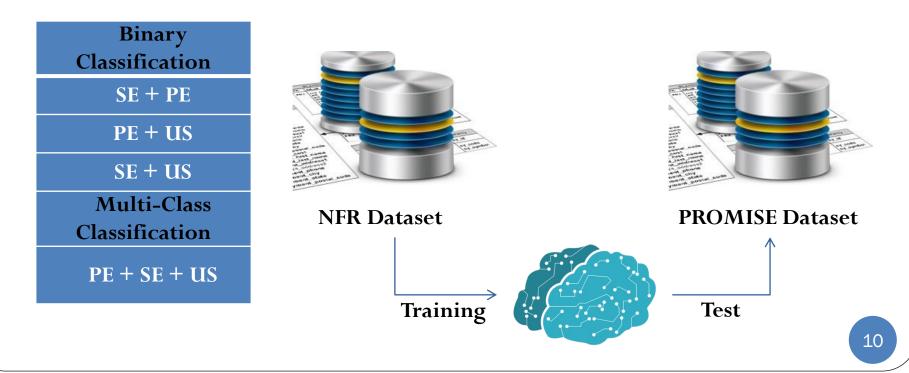


Research Methodology

For this research we select the requirements of **security**, **performance** and **usability**

Then we defined **three binary classifiers** and **one multi-classifier** to evaluate the classification accuracy of our dataset.

Classifying the nonfunctional requirements contained in the PROMISE.



RQ1: Can we extract keywords from SIG catalogues and use these to generate a dataset to classify non-functional requirements?

We can verify, it is possible to extract keywords from SIG catalogues.

Comparing the keywords found in our dataset with the words contained in PROMISE, it was possible to observe that 15% of the keywords in our dataset are contained in PROMISE.

| NFR Type | Total of Keywords | | |
|-------------|-------------------|--|--|
| Performance | 25 keywords | | |
| Security | 24 keywords | | |
| Usability | 28 keywords. | | |
| Total | 77 keywords | | |

| NFR Type | Keywords | | |
|-------------|---|--|--|
| Performance | performance, space, time , throughput, response , | | |
| | memory, consumption, fast , index, triggers, storage, low, | | |
| | run, runtime, perform, execute, mean, peak, compress, | | |
| | dynamic, offset, reduce, fixing, early, processing | | |
| Security | security, confidentiality, integrity, availability, accuracy, | | |
| | completeness, secure, access, registration, authorization, | | |
| | identification, authentication, validation, transaction, | | |
| | user , password , control, encryption, key, spoofing, | | |
| | attack, policy, logging, permission. | | |
| Usability | Usefulness, uniformity, simplicity, operability, | | |
| | intuitiveness, adaptability, comprehensibility, | | |
| | friendliness, performability , accessibility, | | |
| | configurability, understandability, integration, | | |
| | management, usability, compliance, cognition, | | |
| | applicability, language, support, tutorial, training, help, | | |
| | flexibility, easy , use, graphic, timeliness. | | |

To verify the viability of our preliminary dataset, we performing tests with the 4 classifiers.

Precision is the fraction of instances retrieved that are relevant.

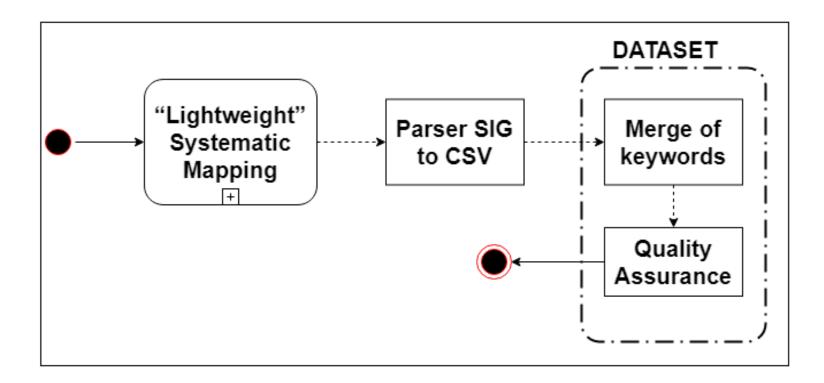
The mean precision of the binary classifiers was 81.50%.

The mean precision of the multi-class classifier was 70.66%.

The best precision of the 4 classifiers was obtained when we classified the usability requirements.

| Binary Classification | | | | | | |
|----------------------------|-----|-----------|--|--|--|--|
| Dataset | NFR | Precision | | | | |
| SE + PE | SE | 0.69 | | | | |
| | PE | 0.79 | | | | |
| PE + US | PE | 0.75 | | | | |
| | US | 0.96 | | | | |
| SE + US | SE | 0.75 | | | | |
| | US | 0.95 | | | | |
| Multi-Class Classification | | | | | | |
| PE + SE + US | PE | 0.57 | | | | |
| | SE | 0.66 | | | | |
| | US | 0.89 | | | | |

RQ2: Can we define a systematic process for dataset generation through NFR Framework? Observing what was done in the research methodology it was possible to define the systematic process composed of 4 steps



In this **activity** the classification **accuracy** of the generated dataset is **verified**.

Accuracy is the fraction of the total sample that is correctly identified.

If the accuracy is **not** acceptable, new **searches for SIG catalogues** are done or **insertion of new keywords** through **taxonomy** tools.

We also use the taxonomy tools for **balancing** the **dataset**



17

RQ3: How well can we automatically classify non-functional requirements using this systematic process?

At the end of the systematic process it was possible to generate this dataset

We obtained an **precision** average of **88.00%** on the **four** proposed classifiers.

The mean **precision** of the <u>binary classifiers</u> was **89.50%**. **+8%**

The <u>multi-classifier</u> obtained an mean **precision** of **83.66%**. +13%

| NFR Type | Total of Keywords | | |
|-------------|-------------------|--|--|
| Performance | 29 keywords | | |
| Security | 29 keywords | | |
| Usability | 29 keywords. | | |
| Total | 87 keywords | | |

Recall is the **fraction** of **instances relevant** that are was retrieved

| Configuration | | | Metrics | | | | |
|----------------------------|-----|-----------|---------|--------|--|--|--|
| Binary Classification | | | | | | | |
| Dataset | NFR | Precision | Recall | F1 | | | |
| SE + PE | SE | 0.98 | 0.85 | 0.91 | | | |
| | PE | 0.89 | 0.98 | (0.93) | | | |
| PE + US | PE | 0.85 | 0.87 | 0.86 | | | |
| | US | (0.89) | 0.87 | 0.88 | | | |
| SE + US | SE | 0.78 | 0.98 | 0.87 | | | |
| | US | (0.98) | 0.71 | 0.83 | | | |
| Multi-Class Classification | | | | | | | |
| PE + SE + US | PE | 0.86 | 0.83 | (0.85) | | | |
| | SE | 0.72 | 0.94 | 0.82 | | | |
| | US | (0.93) | (0.68) | 0.78 | | | |

Future Works

- Make a more rigorous systematic mapping
- Apply our dataset in real projects
- Expand the number of non-functional requirements category in our dataset
- Compare and use other supervised machine learning algorithms

Conclusion

- Use of keywords extracted from SIG Catalogues for dataset generation.
- The definition of a systematic process for generating a dataset, allowing other researchers to generate their own datasets
- The use of a dictionary of synonyms, bringing a taxonomic view.
- Confirm the benefits that machine learning can bring to requirements engineering.