

# AI for Software Engineering Research

## Pitfalls and Challenges

Willian Oizumi

# About me

**Willian Oizumi**

Senior Software Engineer @GoTo

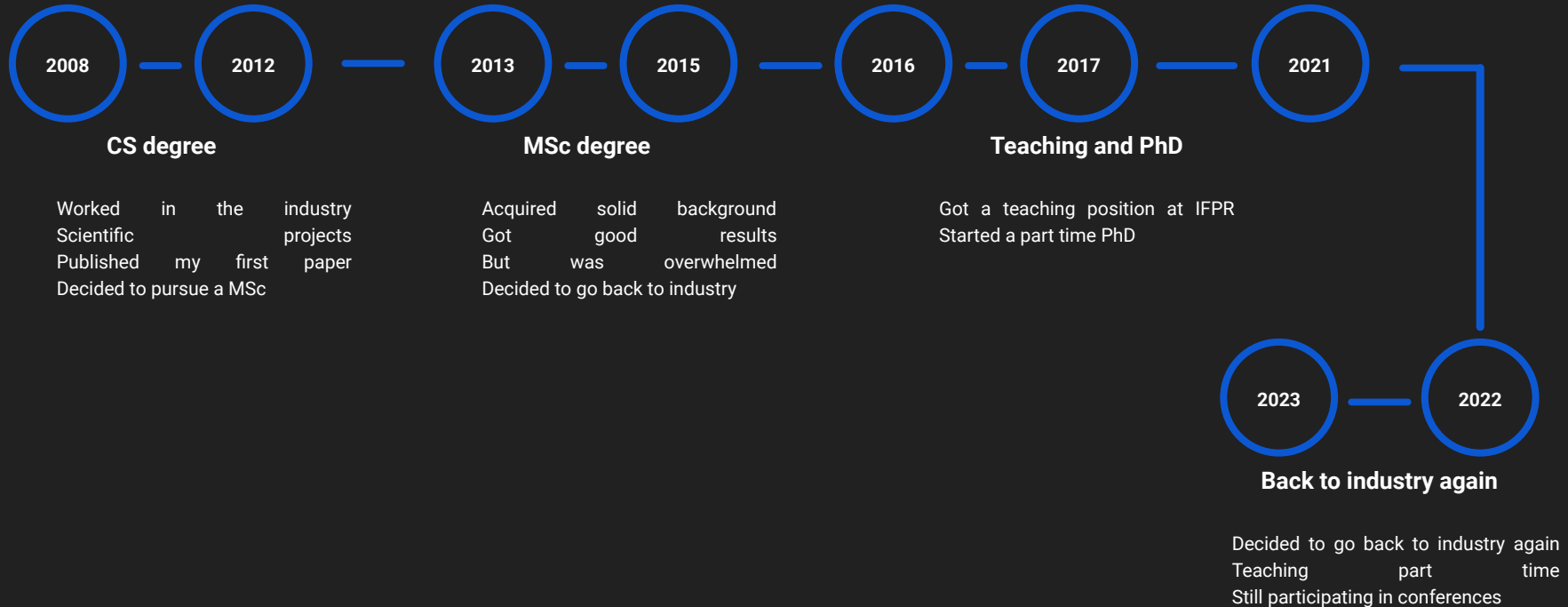
Teaching Professor @Senac-PR

PhD in Informatics - PUC-Rio (2022)



<http://wnoizumi.github.io/>

# My career so far



# My Research Experience

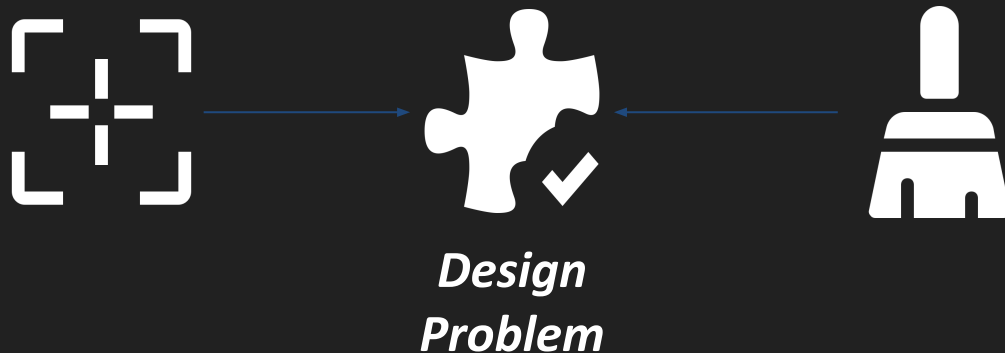
- Internal Quality Metrics
- Code Smells
- Software Architecture and Design
- Refactoring
- Search-based Software Engineering
- Software Product Lines
- Code Review
- Exploratory Study
- Controlled Experiment
- Repository Mining
- Survey
- Grounded Theory

# Our Goals Today

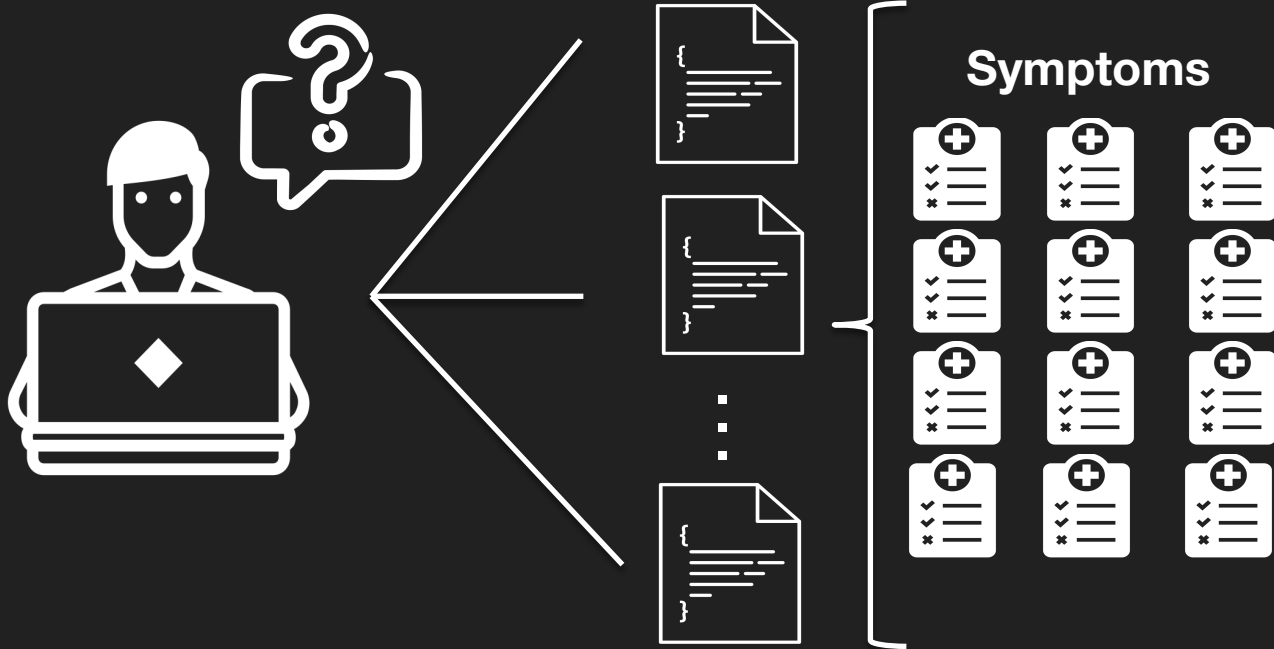
- Provide an overview of my PhD research
- Explore the pitfalls and challenges I faced during the journey
- Discuss recommendations for conducting a strong (PhD/MSc) research

# My PhD Research

Main Goal was to provide **effective support** for developers in the **identification** and **refactoring** of **design problems**



# Where and How to Refactor?



After multiples studies and years...



# Requirements for Refactoring Recommendation Techniques

1. Consideration of Heterogeneous Information



2. Context-Sensitive Detection

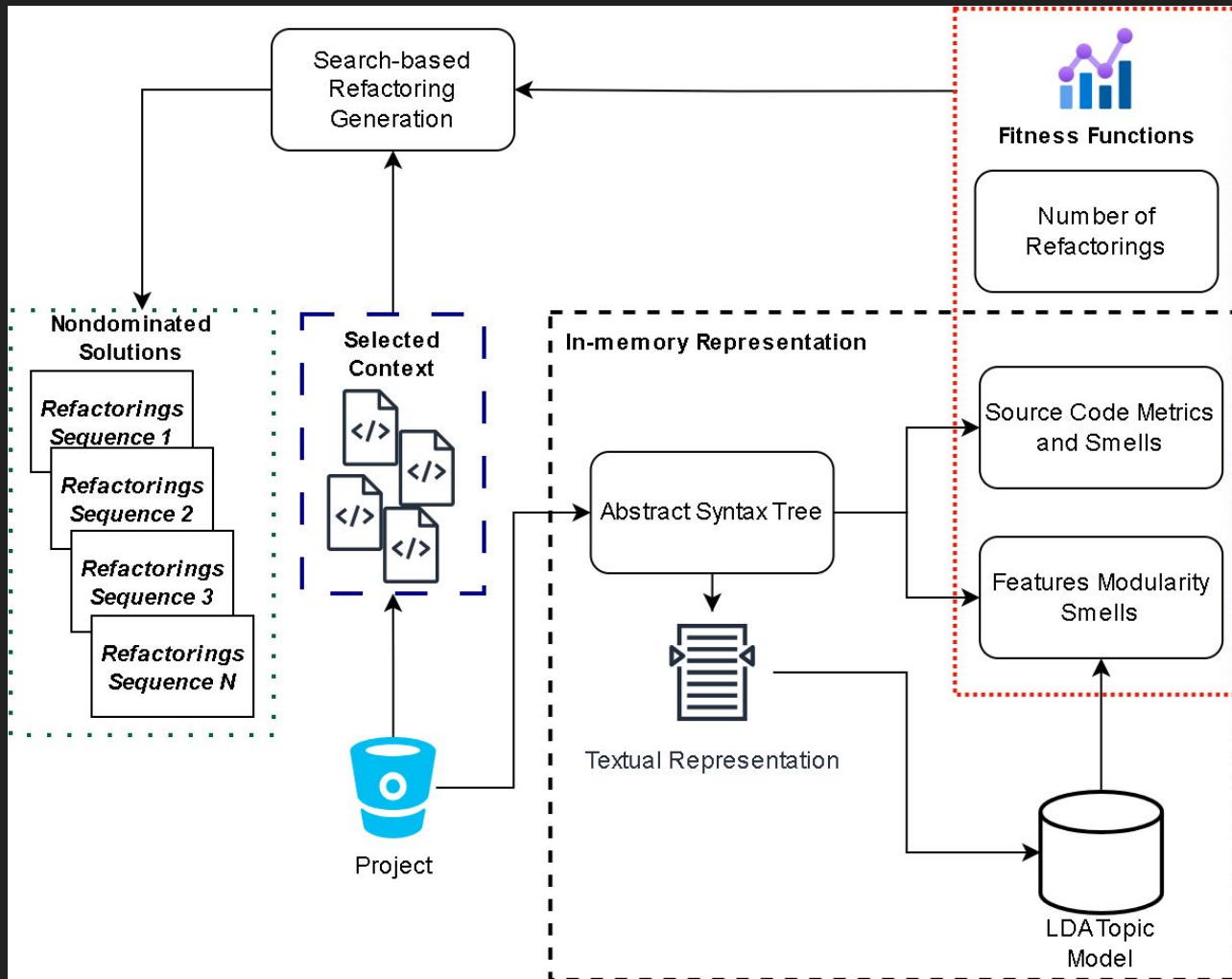


3. Feature Modularity Awareness



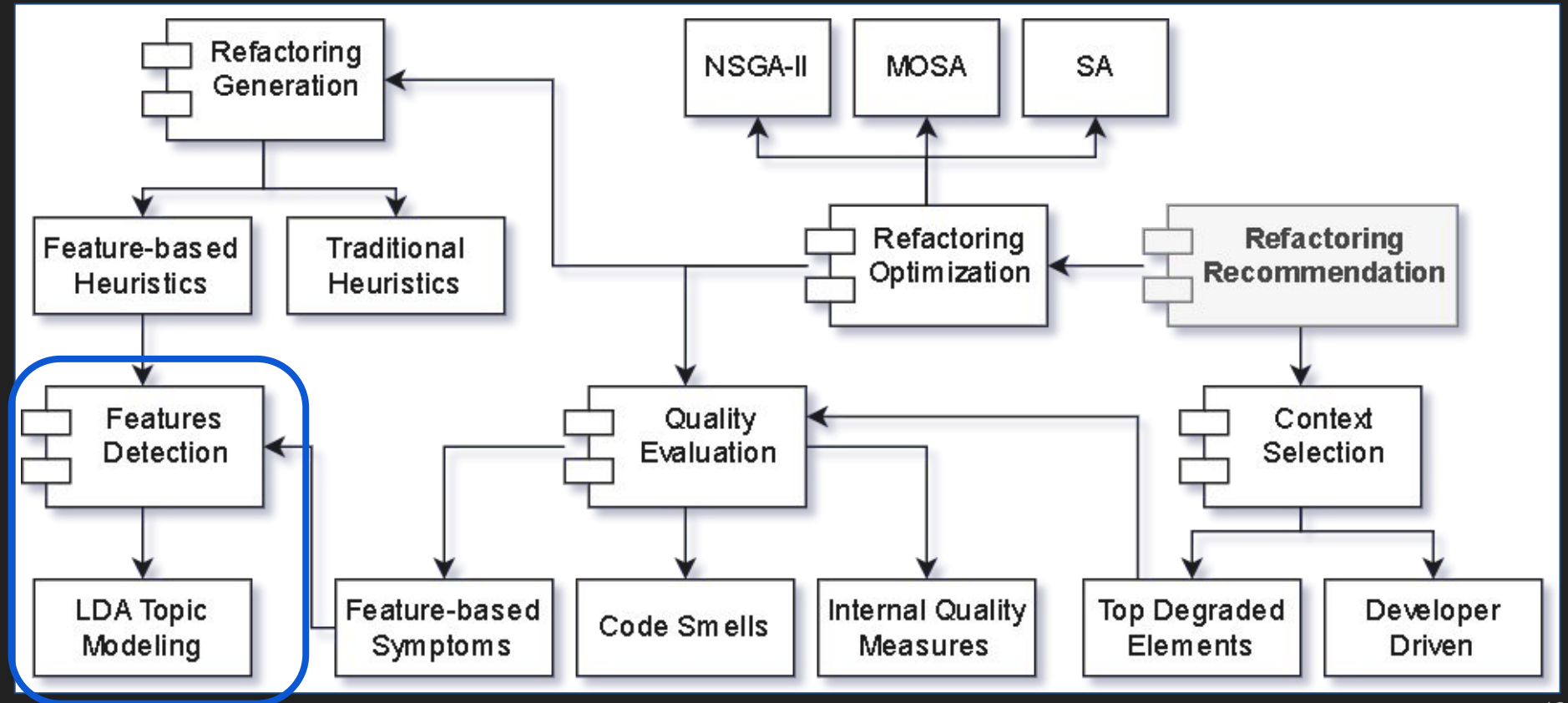
4. Effective Recommendations



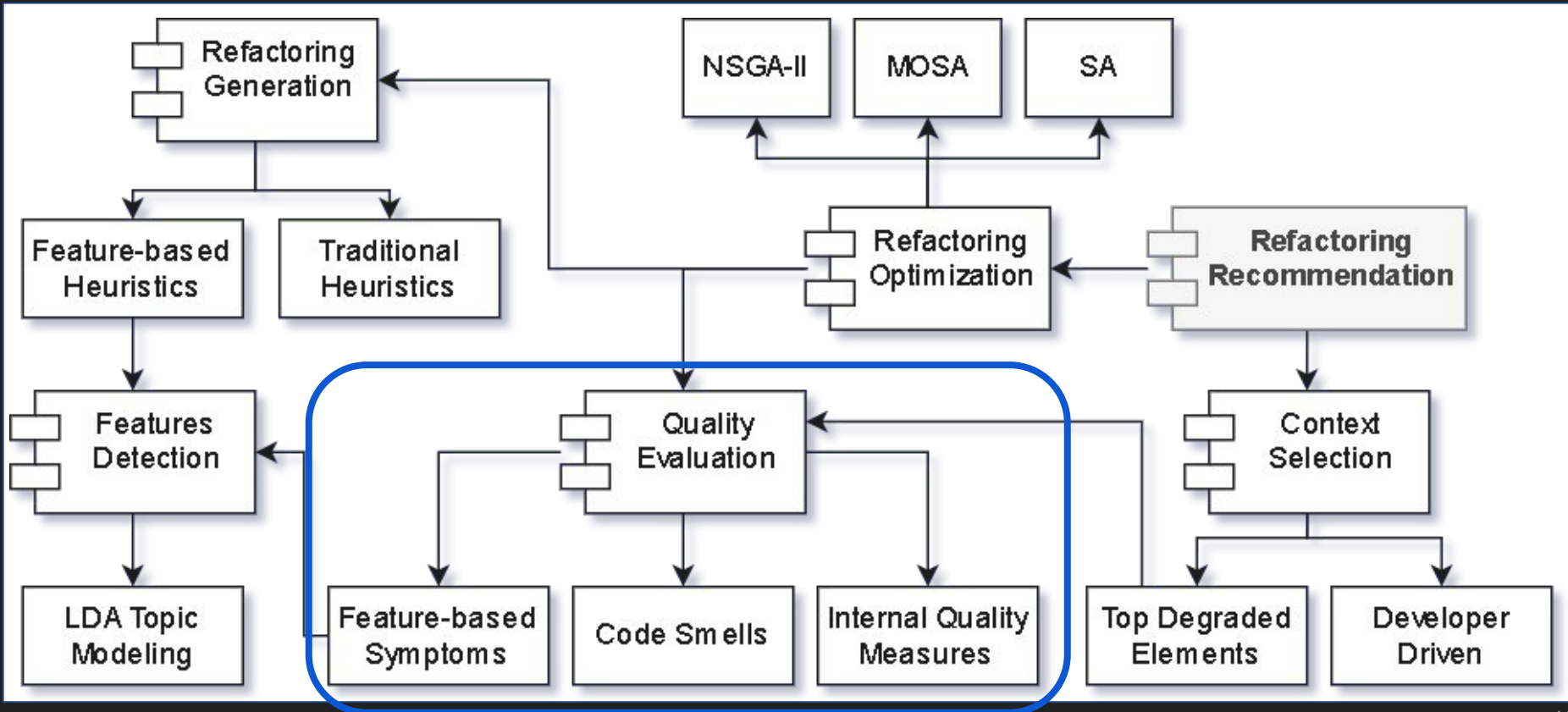




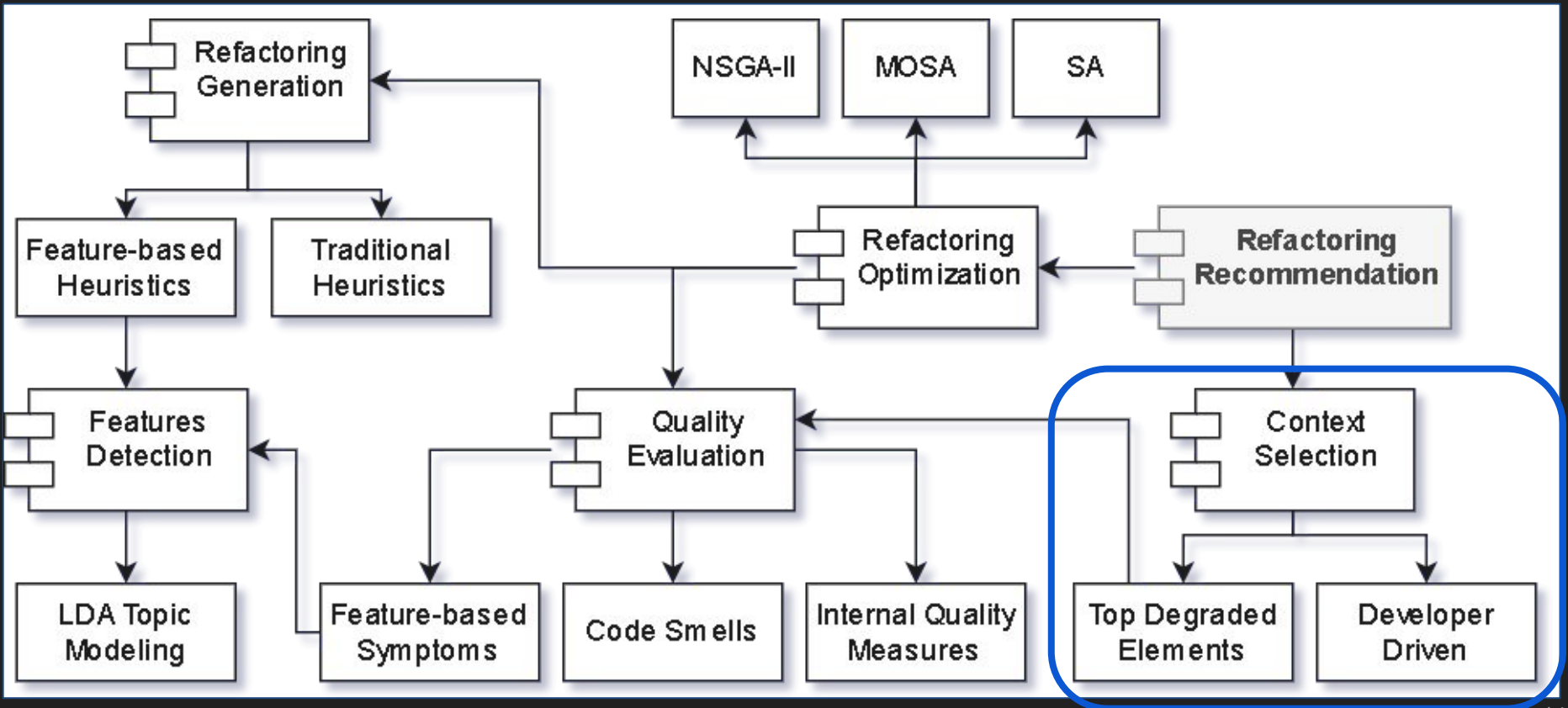
# Overview of Main Components



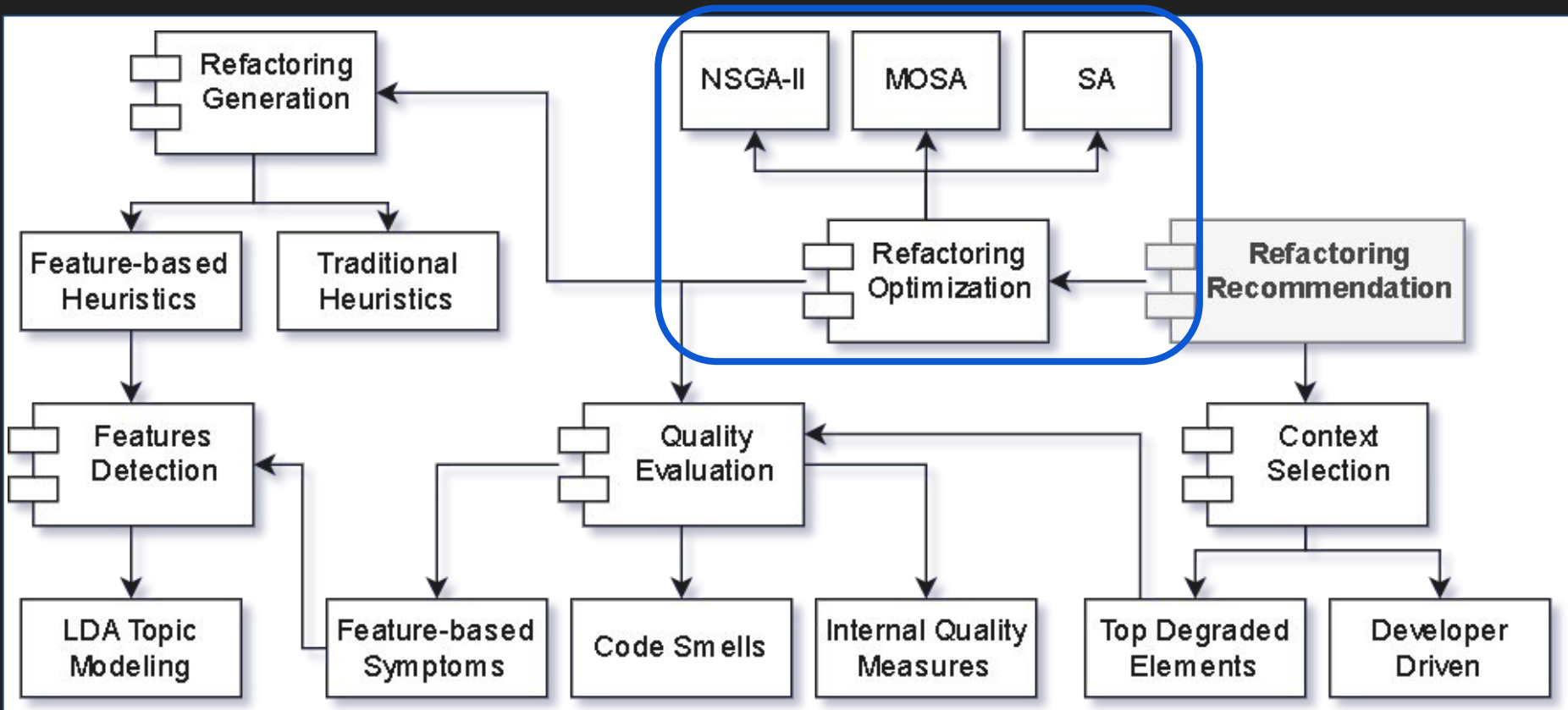
# Overview of Main Components



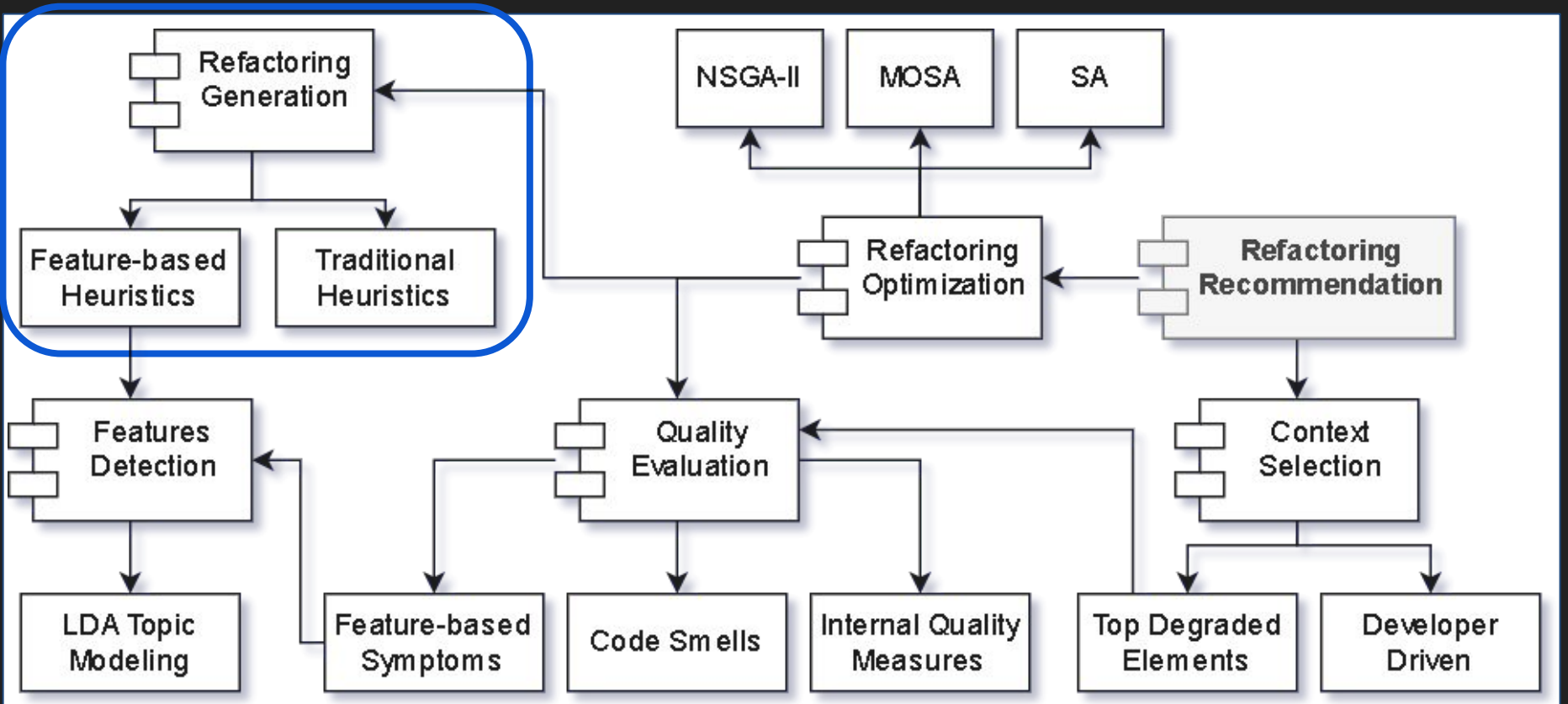
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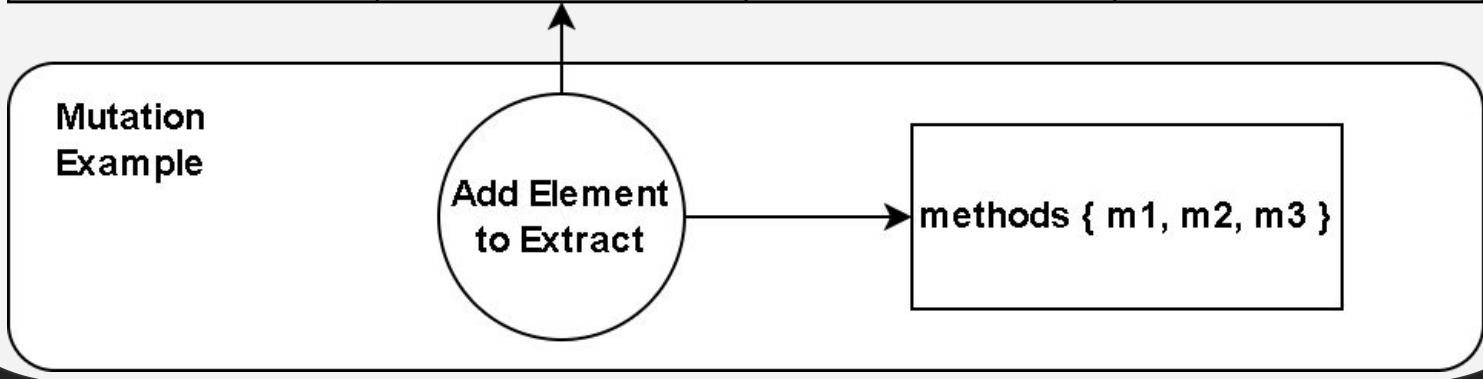


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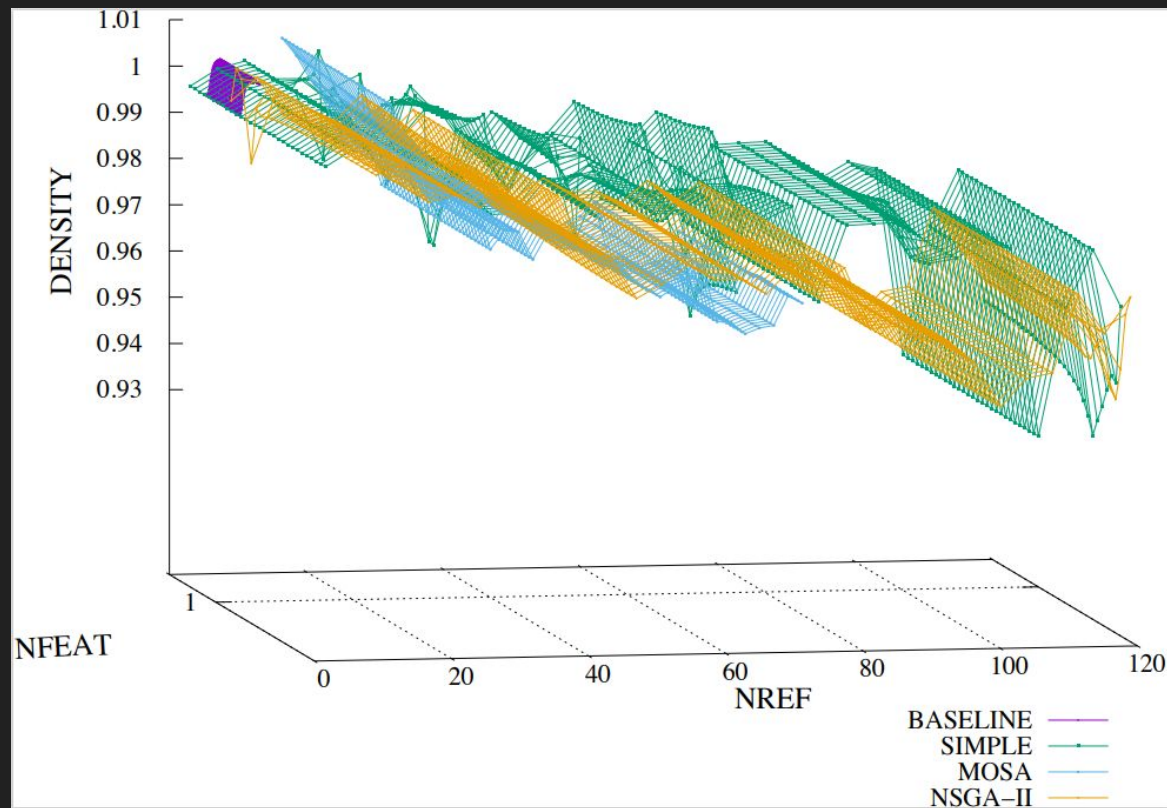




Solution Representation			
ClassA	ClassB	ClassC	ClassD
MM {m1->ClassD}	EC [	MM {m1->ClassE}	No Refactoring
MM {m2->ClassF}	fields {f1, f2, f3}	EC [	
MF {f1->ClassD}	methods {m1, m2}	fields {f1, f2}	
MF {f2->ClassF}	]	methods {m2, m3} ]	
0	1	2	3



# Example of Solution Space for the okhttp Project



# Quality Impact of Solutions

Project	# Smell Difference				# Features Difference				LCOM Difference			
	BL	SP	MOSA	NSGA-II	BL	SP	MOSA	NSGA-II	BL	SP	MOSA	NSGA-II
Fresco	-6	<b>-23</b>	-16	<b>-23</b>	-1,5	<b>-7</b>	-4	<b>-7</b>	0,0468	-0,2043	-0,0723	<b>-0,2212</b>
RxJava	-15	-35	-13	<b>-37</b>	-2	-10	-5,5	<b>-11</b>	-0,0544	0,0151	0,0415	<b>-0,0747</b>
Jenkins	-24	-27	-21	<b>-29</b>	-4	-5	-5	<b>-6</b>	-0,0261	-0,0923	-0,0921	<b>-0,1378</b>
Spring Security	-5	<b>-19</b>	-17	<b>-19</b>	0	<b>-5</b>	<b>-5</b>	-4	0,0007	-0,0913	-0,0469	<b>-0,1145</b>
Dubbo	-11	<b>-24</b>	-11	-23	-1	<b>-2</b>	<b>-2</b>	-1	0,1247	-0,1452	-0,0475	<b>-0,1972</b>
OkHttp	-6,5	-26	-20	<b>-27</b>	1	-3	<b>-4</b>	<b>-4</b>	-0,0741	-0,2193	-0,1549	<b>-0,2541</b>
All	-10,5	<b>-26</b>	-16	<b>-26</b>	-2	-5	-4	<b>-6</b>	-0,0202	-0,114	-0,066	<b>-0,151</b>

# Quality Impact of Solutions

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**NSGA-II** outperformed other strategies  
in most projects

# Thesis Contributions

01

## Requirements and Guidelines

- Empirically identified requirements
- Definition and evaluation of key requirements

02

## Complete Approach for Dealing with Design Problems

- Robust evidence on the identification and removal of design problems
- Single solution addressing all key requirements

03

## Search-based Refactoring

- Alternative and effective solution representation
- Heuristics-based initial population
- Domain-specific mutation operators

04

## Open Science

- Open source tools (Organic and OrganicRef)
- Replication package for all studies

What went wrong?  
(what could be better?)

#1

Too much indecision at the beginning

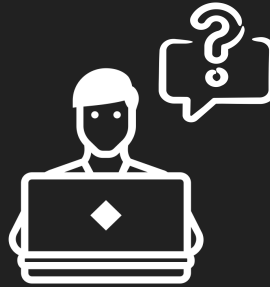
Technical Debt

Code Smells

IHC

Source Code  
Metrics

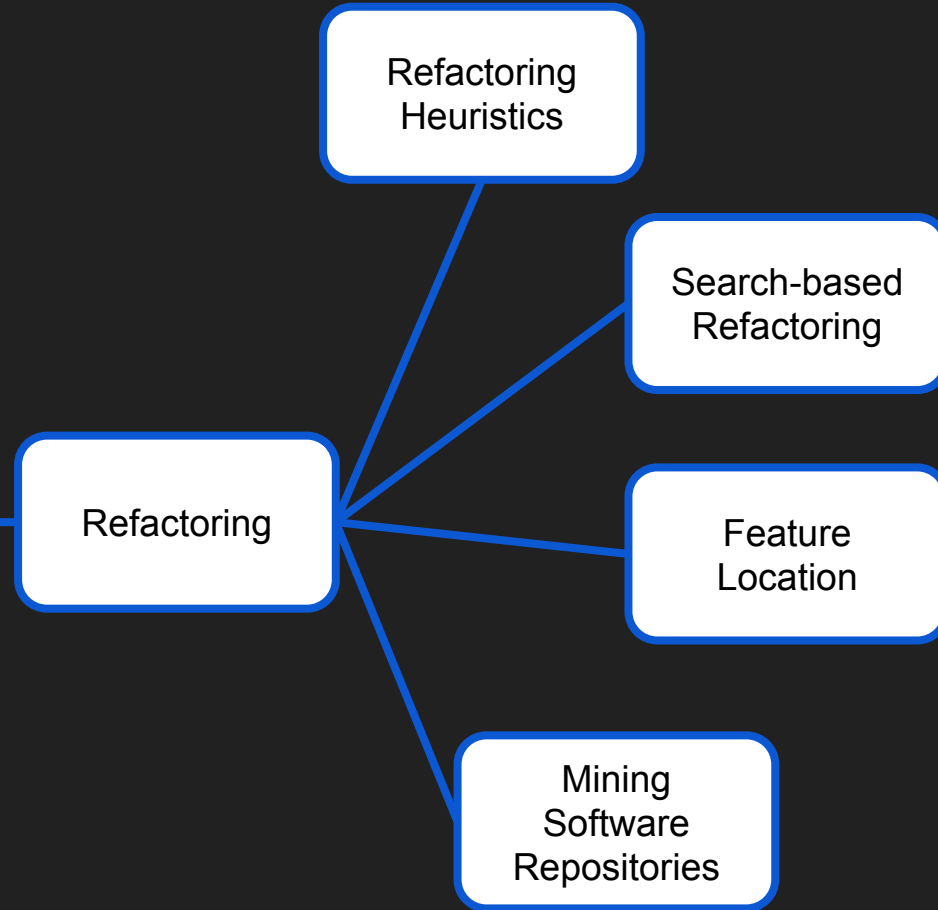
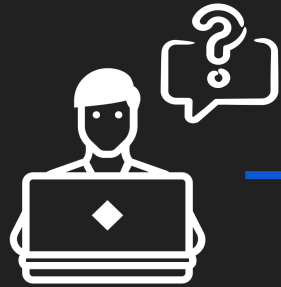
Refactoring



Software  
Architecture

Internal Quality  
Attributes

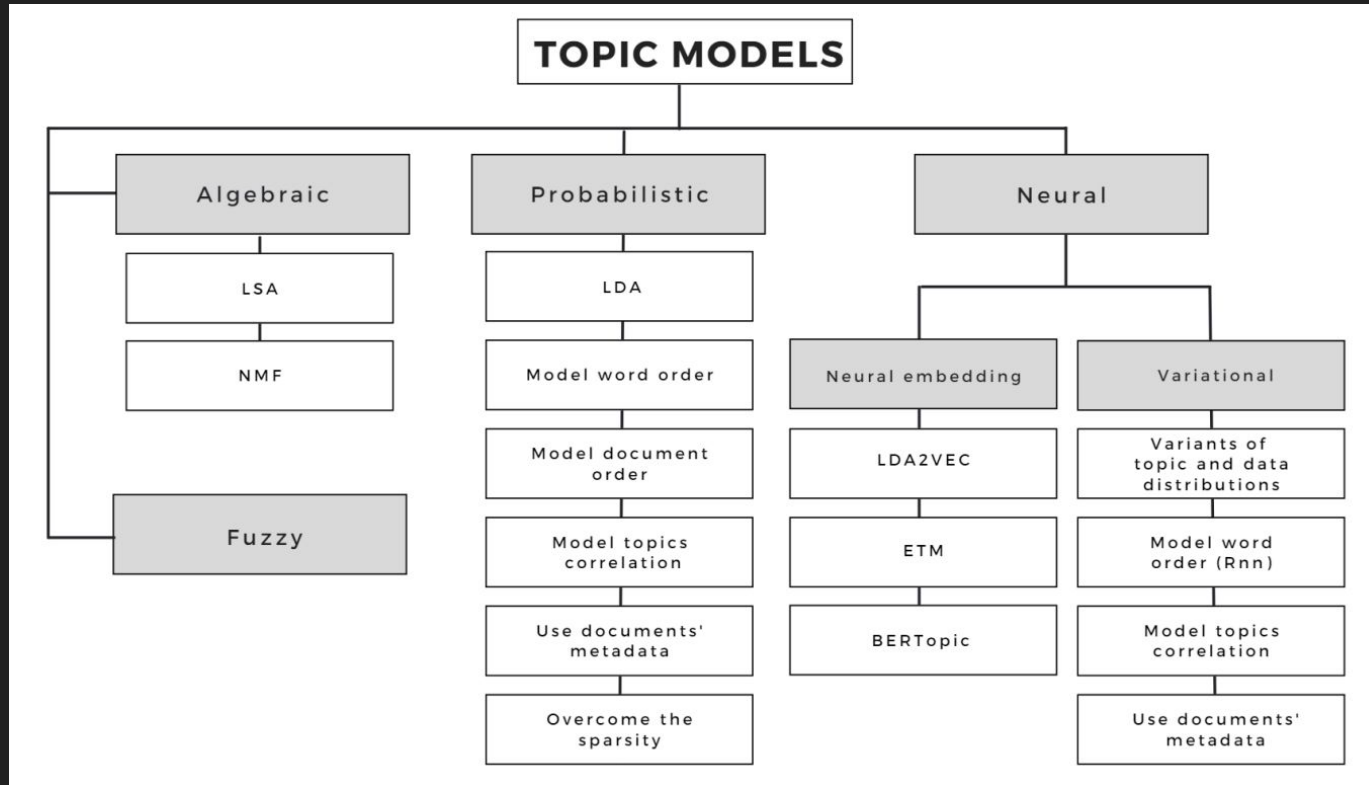




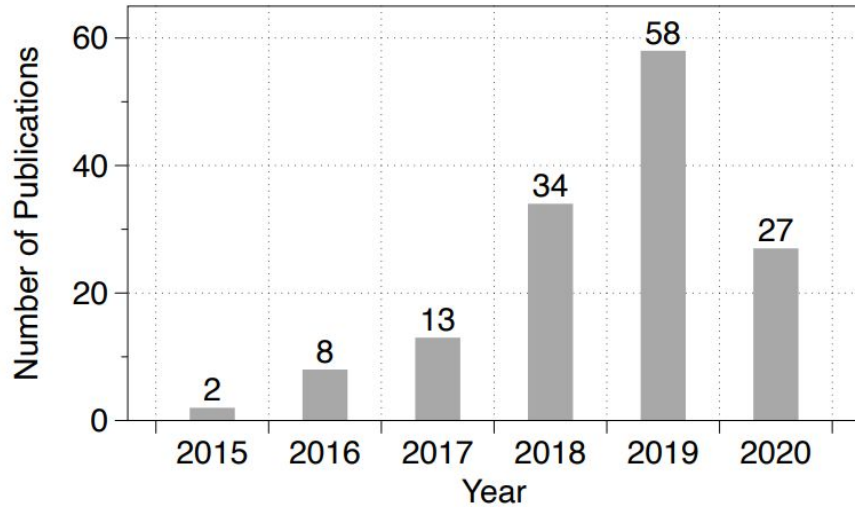
#2

Lack of attention to the fast-evolving  
literature

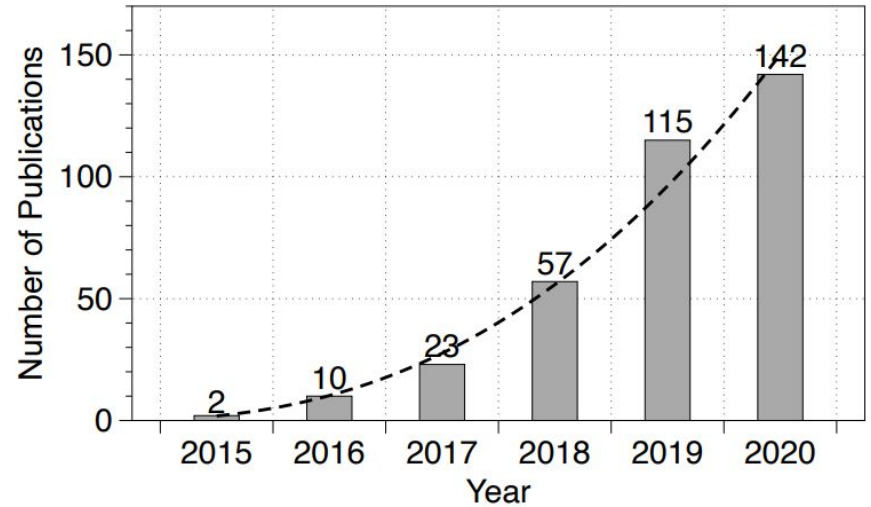
# Example: State-of-the-art for Topic Modeling



# Another Example: Deep Learning for Software Engineering



(a) Number of publications per year.



(b) Cumulative number of publications per year.

Yang, Yanming, et al. "A survey on deep learning for software engineering." ACM Computing Surveys (CSUR) 54.10s (2022): 1-73.

# Number of Papers per Topic - ICSE

Quick search on dblp, only looking at the sessions and papers titles:

Year	* Learning	NN/DL	LLM / Pre-trained	SBSE
2016	3	0	0	1
2017	4	1	0	3
2018	1	3	0	9
2019	4	11	0	1
2020	18	10	0	0
2021	16	12	1	0
2022	39	22	8	0
2023	33	17	11	0

Is SBSE dead?

# Applications of Search-based Software Testing to Trustworthy Artificial Intelligence

**Track** [SSBSE 2022 Keynotes](#)

**When** **Thu 17 Nov 2022 09:00 - 10:30** at **ERC SR 9 - Plenary + Keynote 1** Chair(s): [Mike Papadakis](#)

**Abstract** Increasingly, many systems, including critical ones, rely on machine learning (ML) components to achieve autonomy or adaptiveness. Such components, having no specifications or source code, impact the way we develop but also verify such systems. This talk will report on experiences and lessons learned in applying search-based solutions to test and analyse such ML-enabled systems. Indeed, our results have shown that metaheuristic search plays a key role in enabling the effective test automation of ML models and the systems they are integrated into. Though other techniques are also required to achieve scalability and enable safety analysis, for example, the black-box nature of ML components naturally lends itself to search-based solutions.

**Bio** <http://www.lbriand.info>

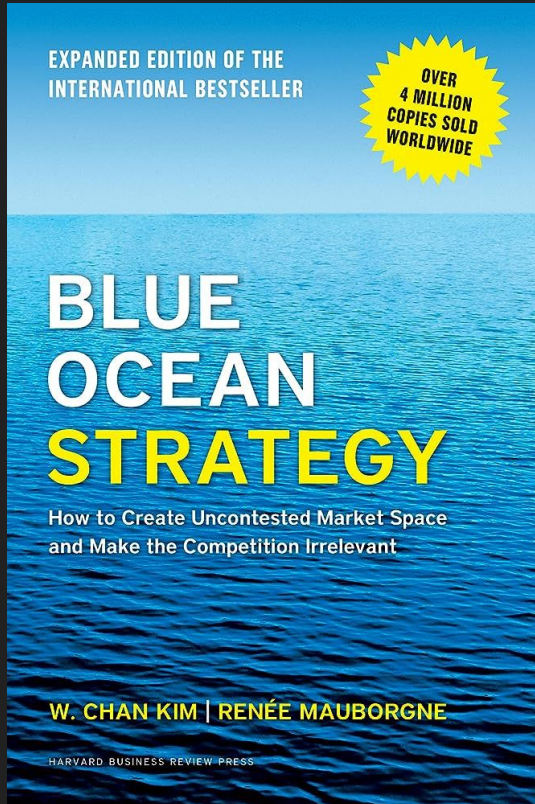


Lionel Briand

University of Luxembourg; University of Ottawa

<https://conf.researchr.org/home/ssbse-2022>

# Red/Blue Ocean Strategy

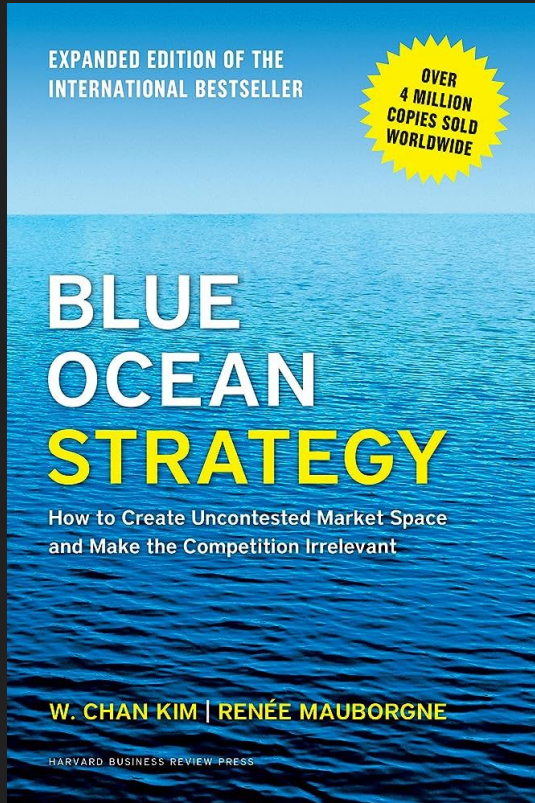


RED OCEAN STRATEGY	VS	BLUE OCEAN STRATEGY
Compete in <b>existing</b> market space		Create <b>uncontested</b> market space
<b>Beat</b> the competition		Make the competition <b>irrelevant</b>
Exploit <b>existing</b> demand		Create and capture <b>new</b> demand
<b>Make</b> the value-cost trade-off		<b>Break</b> the value-cost trade-off
Align the whole system of a firm's activities with its strategic choice of <b>differentiation <u>or</u> low cost</b>		Align the whole system of a firm's activities with its strategic choice of <b>differentiation <u>and</u> low cost</b>



# Red/Blue Ocean Strategy

Most research is in the **red ocean**



RED OCEAN STRATEGY	VS	BLUE OCEAN STRATEGY
Compete in <b>existing</b> market space		Create <b>uncontested</b> market space
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Align the whole system of a firm's activities with its strategic choice of <b>differentiation or low cost</b>		Align the whole system of a firm's activities with its strategic choice of <b>differentiation and low cost</b>

#3

Too much effort spent in contributions  
unrelated to my thesis

#4

Too much focus on technical aspects rather  
than on the scientific ones

## Lots of effort on...

- Writing beautiful code
- Creating a tool that is easy to use
- Testing different frameworks and libraries to use in the tool
- Collaborating in other students' studies

## Rather than...

- Being up to date with the literature
- Finding out which datasets could be used in my research
- Validating and adapting the proposed technique
- Designing a sound and robust evaluation study
- Doing preliminary studies and publishing first author papers

# Empirical Standards for Software Engineering Research

Empirical Standards					About	Tools	Standards	Supplements	FAQ	People
General Standard	Action Research	Benchmarking	Case Study	Case Survey						
Data Science	Engineering Research	Experiments	Grounded Theory	Longitudinal						
Meta Science	Mixed Methods	Optimization Studies	Qualitative Surveys	Quantitative Simulation						
Questionnaire Surveys	Replication	Repository Mining	Systematic Reviews							

Available at: <https://sigsoft.org/EmpiricalStandards/docs/>

# Get to Know What is Available

Datasets

Replication  
Packages  
(Zenodo/Figshare)

MSR Mining  
Challenge

Create and publish  
your own dataset

(Pretrained)  
Models

Tools / Source  
Code

# Get to Know What is Available

Datasets

(Pretrained)  
Models

Tools / Source  
Code

Hugging Face

GitHub

Replication  
Packages  
(Zenodo/Figshare)

# Get to Know What is Available

Datasets

(Pretrained)  
Models

Tools / Source  
Code

GitHub

From your or  
others Labs

Replication  
Packages  
(Zenodo/Figshare)



#5

The chosen scope was larger than it should  
have been

# Revisiting the Requirements

1. Consideration of Heterogeneous Information
2. Context-Sensitive Detection
3. Feature Modularity Awareness
4. Effective Recommendations

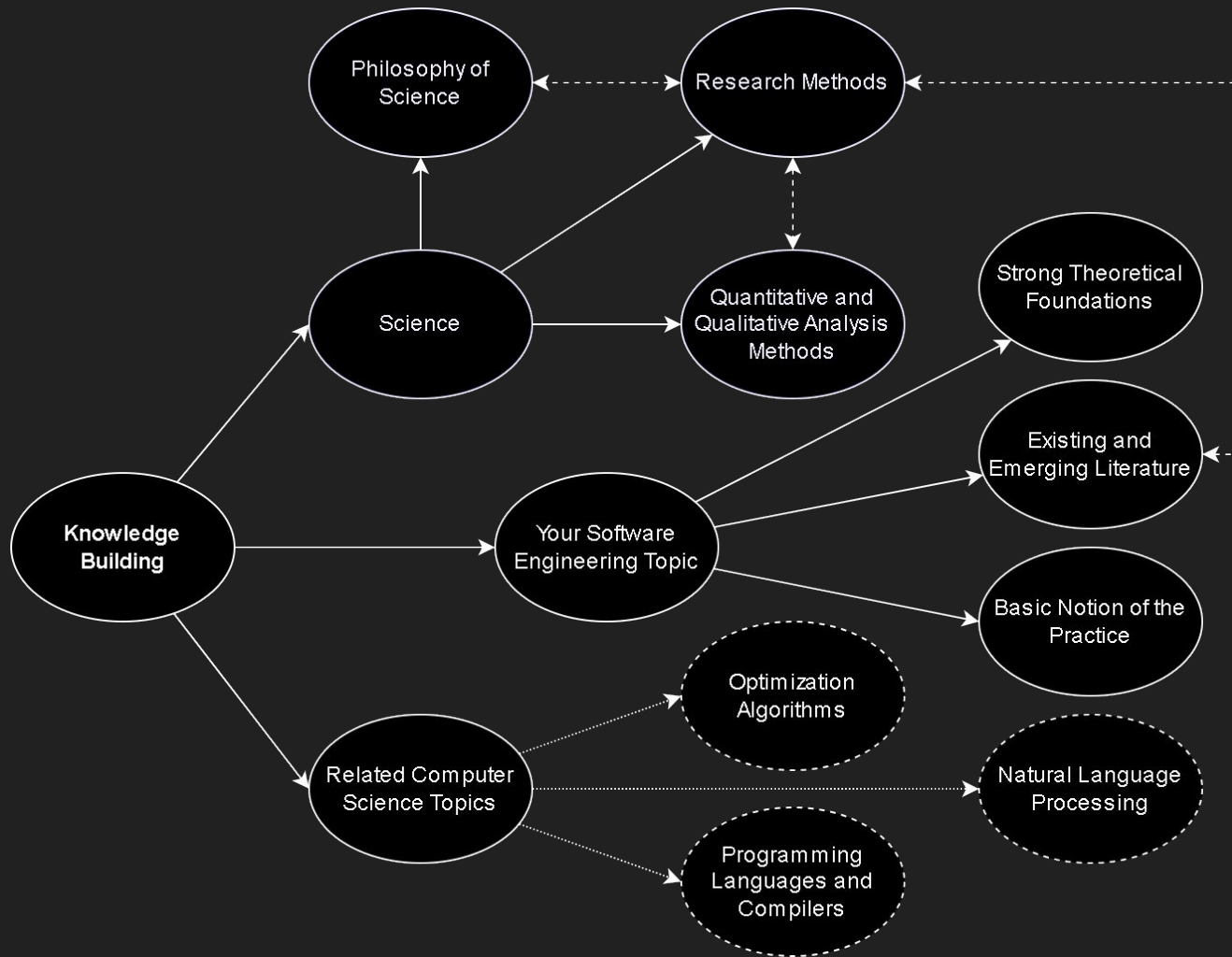
# Revisiting the Requirements

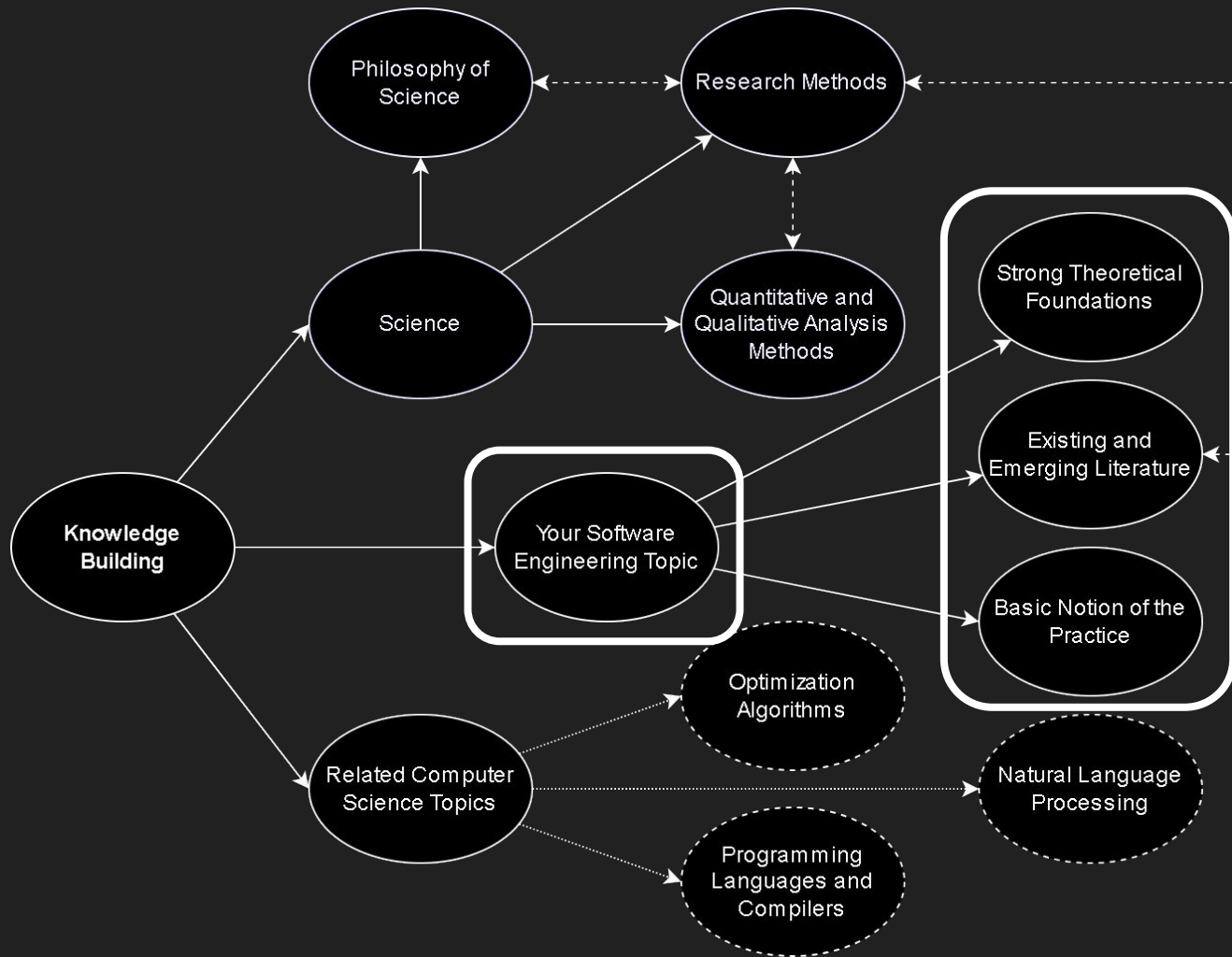
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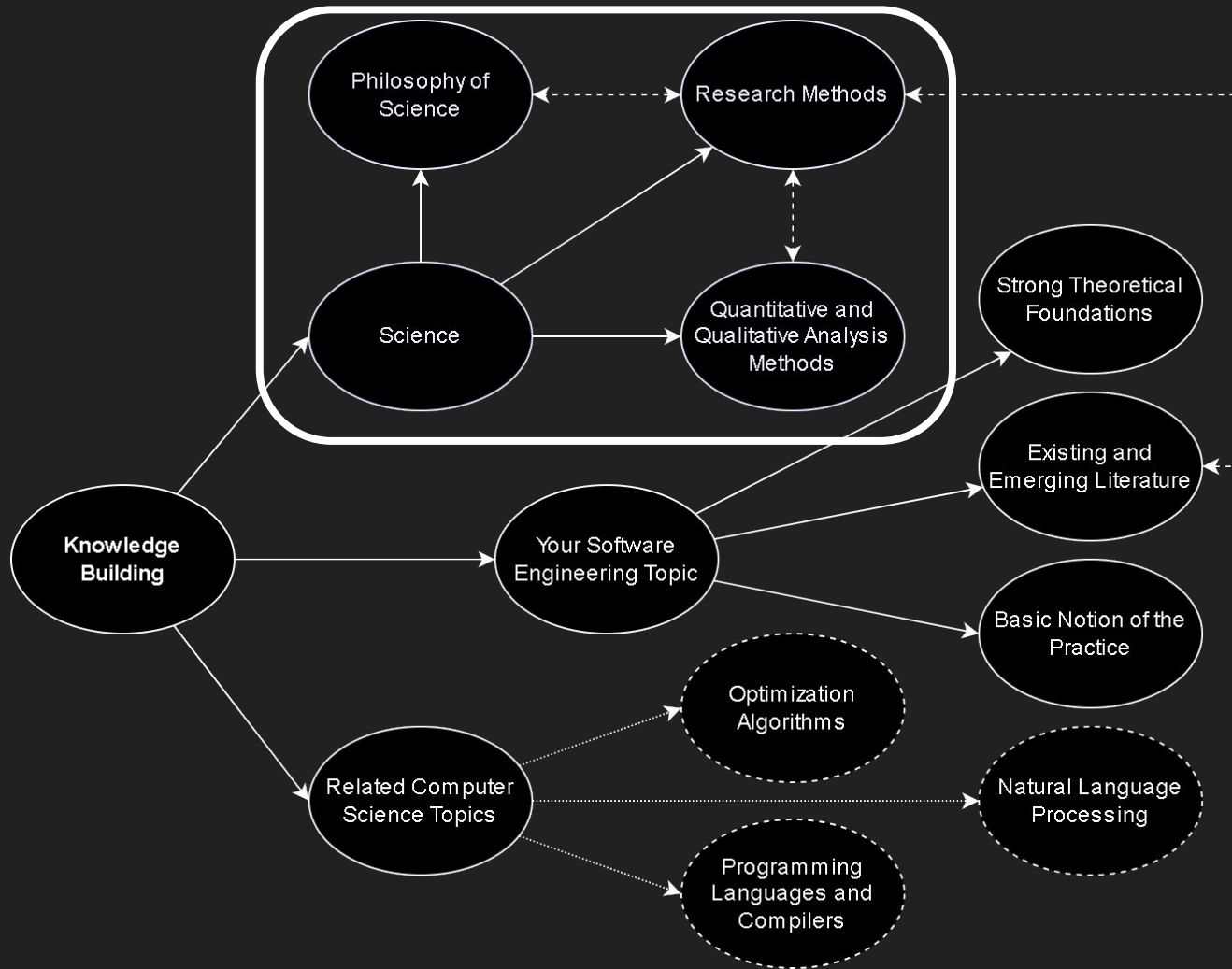
Each requirement alone could be  
deeply explored in a thesis

# Mind Maps to Guide your Journey

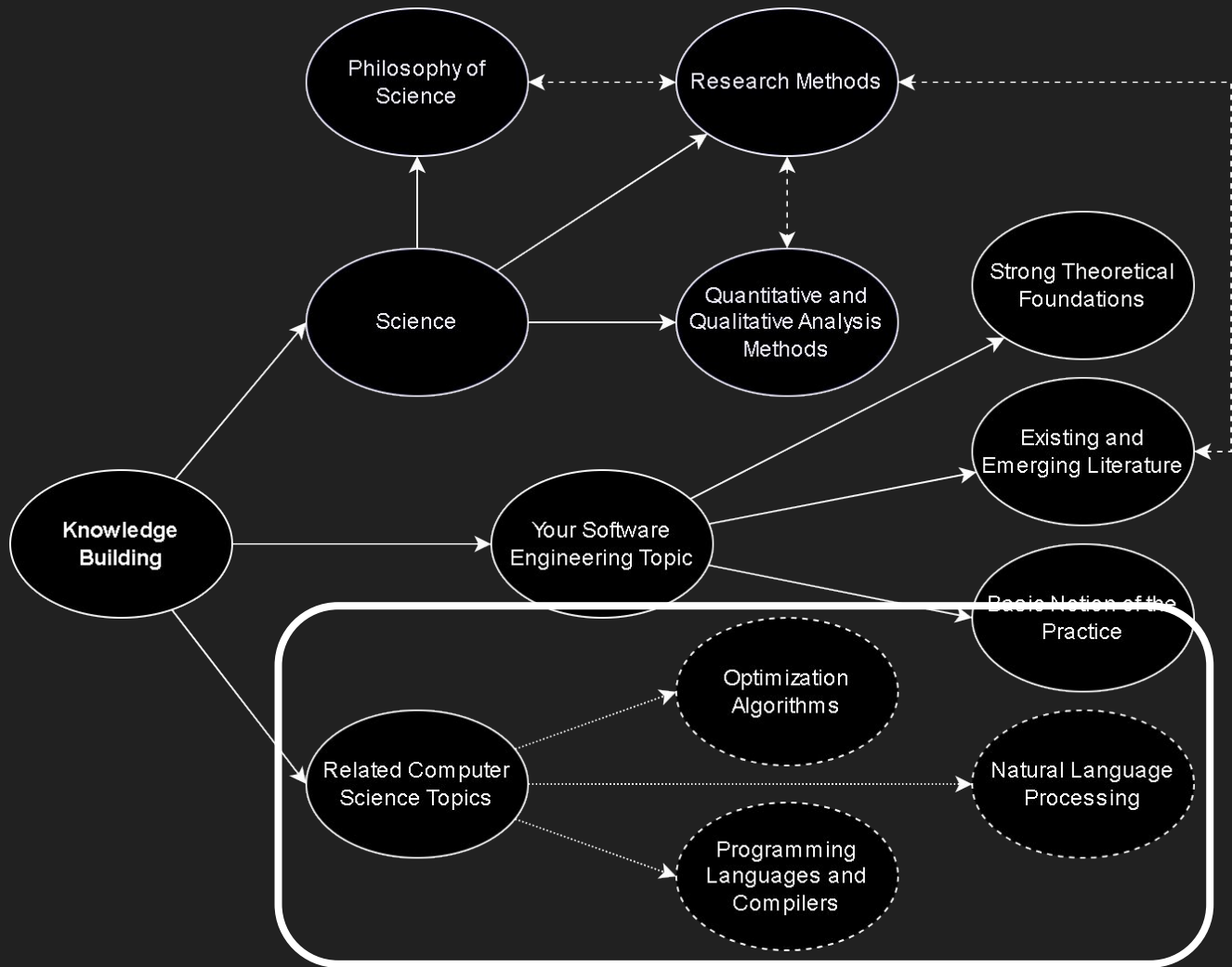
# Mind Map 1: Knowledge Building



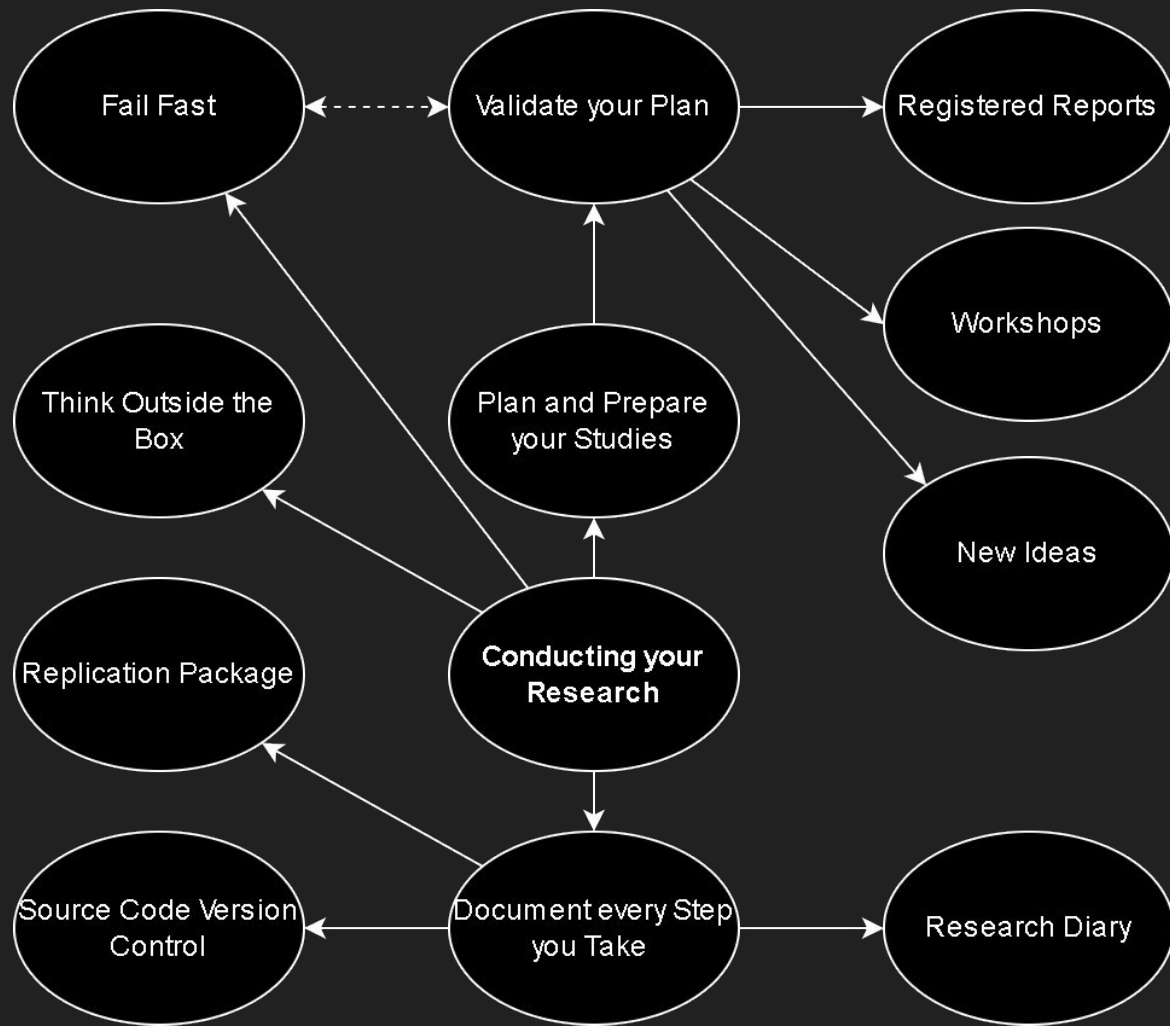


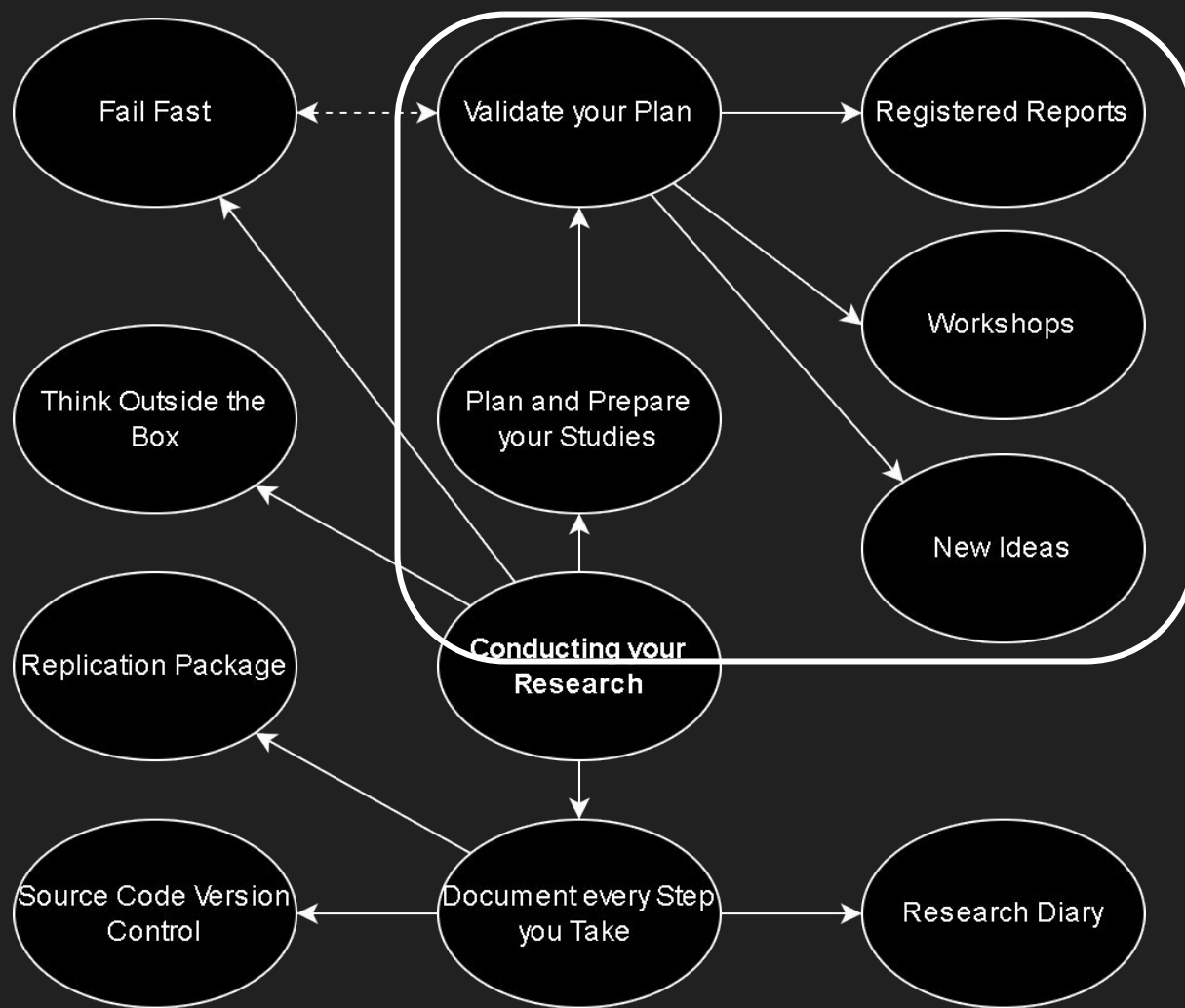


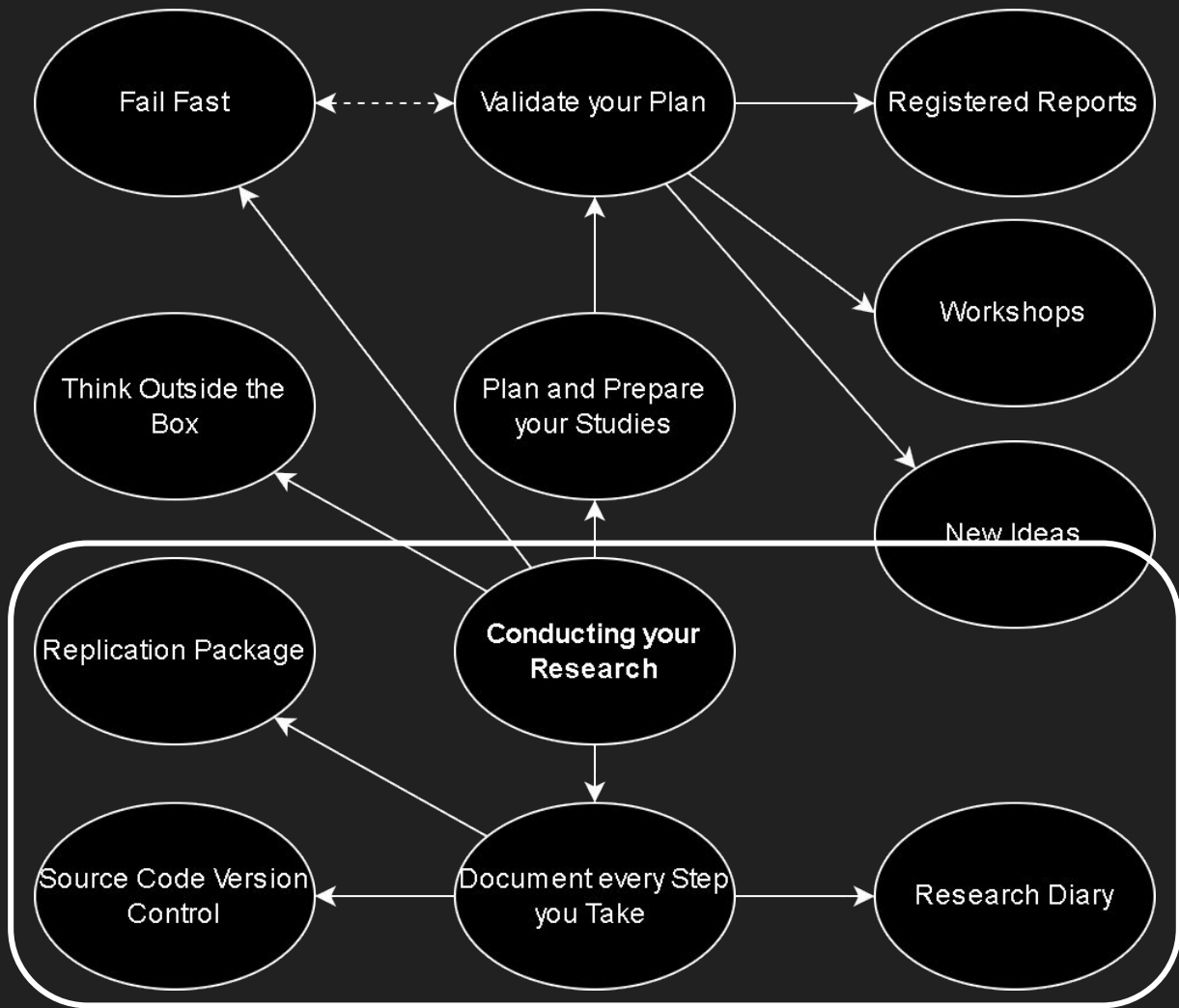


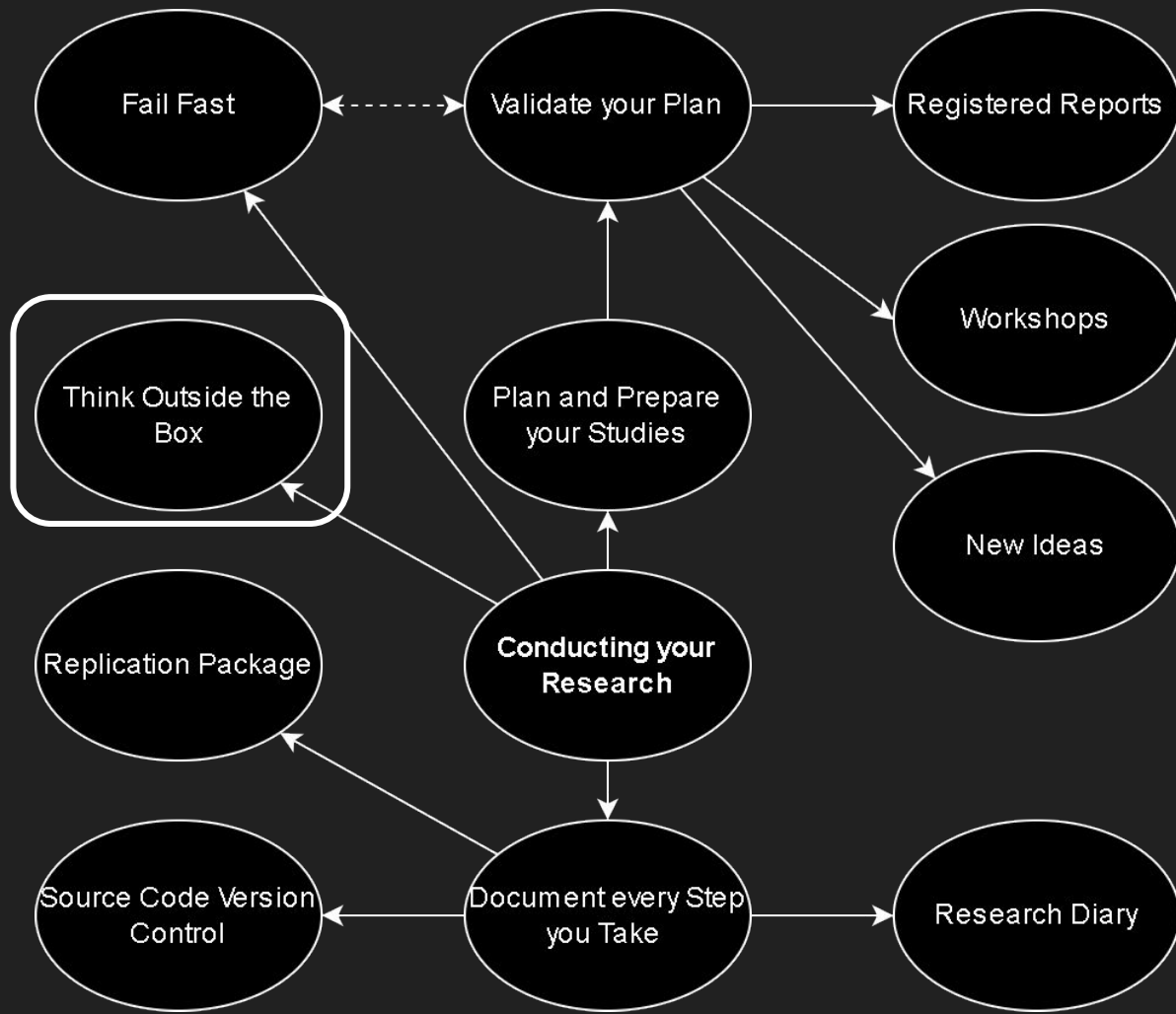


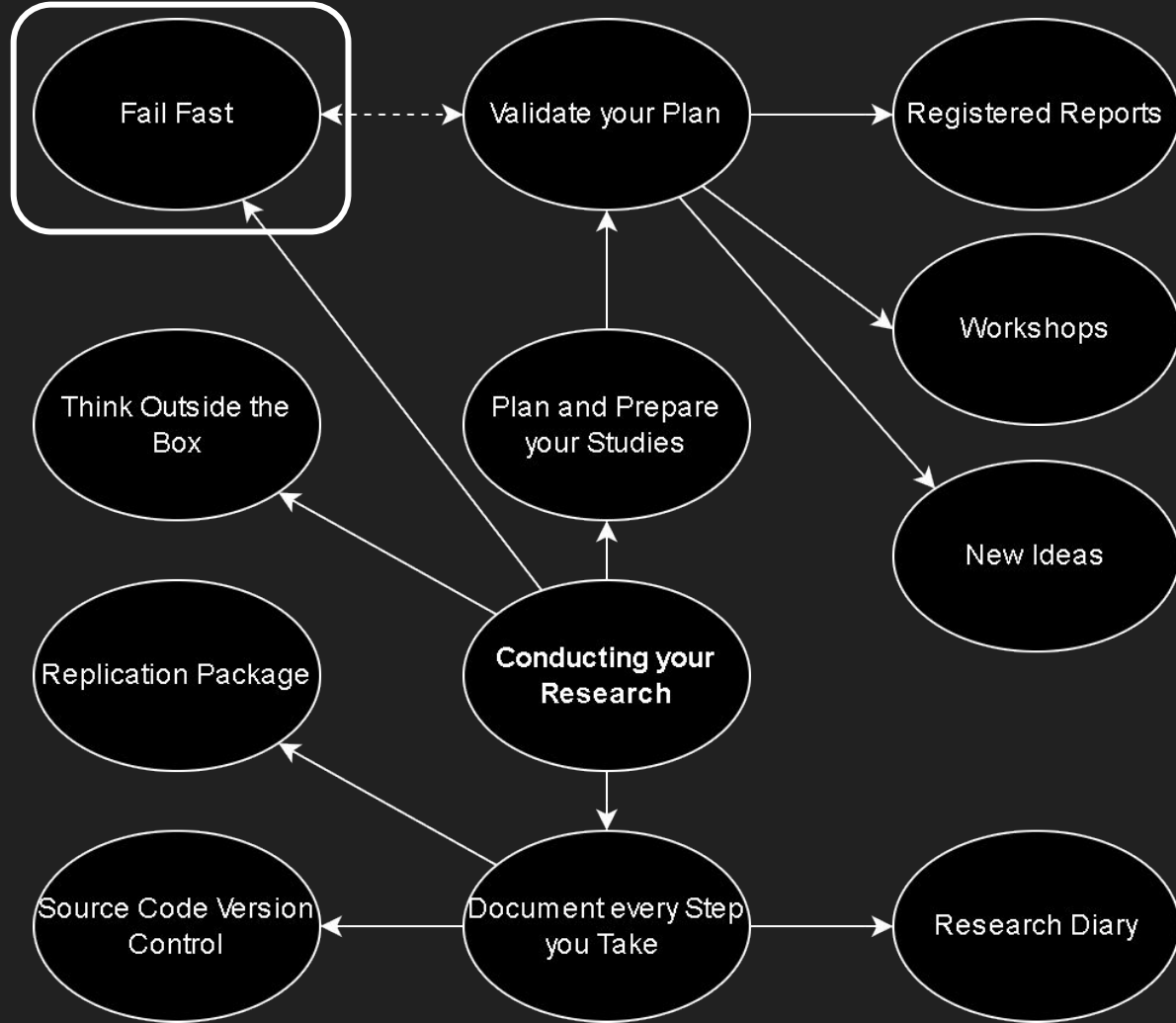
# Mind Map 2: Conducting your research











## In a Nutshell

- Select your research topic as soon as possible
- Avoid too large scopes
- Put effort on building the required knowledge
- Try to fill the gaps with collaborations
- Keep tracking the literature and adapt if needed



# AI for Software Engineering

## Pitfalls and Challenges

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