

# RECOMMENDING WHEN DESIGN TECHNICAL DEBT SHOULD BE SELF-ADMITTED

Computer Engineering



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# Concepts and Definitions

## *Technical Debt (TD)*

### Definition

TDs are not quite right code which we postpone making it right [1].  
They are temporary and suboptimal solutions or workarounds.

### Reasons

Quickly fix an issue

Early conception stages

Lack of comprehension, skill or experience

# Concepts and Definitions

## *Technical Debt (TD)*

### Types

Design	Object-Oriented design principles
Requirement	Functional and non functional requirement trade-offs
Code	Quality of source code
Test	Quality of testing activities
Documentation	Misleading Information
Defect	Known and possible defects

### Impacts

Slower conception and execution  
Lower maintainability and quality  
Higher production cost

# Concepts and Definitions

## *Self-Admitted Technical Debt (SATD)*

### Definition

SATDs are technical debts which are consciously introduced and explicitly documented by developers.

### Characteristics

Diffusion	31% of files contain SATD, 51 instances per system
Evolution	Persistent and increase with time
Actors	Experienced developers introduce them

# Elements of the Problematic

## *Self-Admitted Technical Debt (SATD)*

### Issues

#### Weak Performance Results

#### Ineffective Strategies

- Code Refactoring
- Continuous Tracking
- Proactiveness

#### Flawed or Incomplete Approaches

- Pattern Matching
- Machine Learning with Natural Language Processing

## Objectives

### *Main Research Question*

How can we identify and detect **technical debts** in a software project using **source code or features** and known **self-admitted technical debts** with a **machine learning** approach?

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#### TEDIOUS

Metrics and Warnings  
Machine Learners

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#### CNN

Source Code  
Neural Network

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# Objectives

## *Application Objectives*

### Recommendation System

To encourage self-admitting technical debts

To ease tracking

To fix issues faster

### Complement to existing smell detectors

Or use as an alternative

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Tracking

Managing

Improving

# Objectives

## *Design Objectives*

### Defining and Extracting Features

**TEDIOUS** : Metrics and Warnings

**CNN** : Source Code

### Identifying SATDs

### Preprocessing Features

**TEDIOUS** : Multi-Collinearity, Feature Selection, Normalization, Rebalancing

**CNN** : Tokenization, Word Embedding

### Building and Applying Machine Learners

**TEDIOUS** : Machine Learning Models

**CNN** : Neural Network

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## 1. TEDIOUS

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# Study Definition

## Studied Projects

Project	Number of				Nb of Com € M	Nb of SATD		SATD M %
	F	C	M	Com		€ M	€ M	
Ant	1,113	1,575	11,052	20,325	13,359	1	57	0.5%
ArgoUML	1,922	2,579	14,346	64,393	17,722	203	425	2%
Columba	1,549	1,884	7,035	33,415	10,305	8	418	5%
Hibernate	2,129	2,529	17,405	15,901	9,073	21	377	2%
jEdit	394	889	4,785	15,468	10,894	6	77	2%
jFreeChart	1,017	1,091	10,343	22,827	15,412	4	1,881	18%
jMeter	1,048	1,328	8,680	19,721	12,672	95	424	5%
jRuby	970	2,063	14,163	10,599	7,809	16	275	2%
Squirrel	2,325	4,123	16,648	25,216	15,574	35	173	1%

F : Files

C : Classes

M : Methods

Com : Comments



# Study Definition

## *Research Questions*

RQ1

How does TEDIIOUS work for recommending SATD within project?

RQ2

How does TEDIIOUS work for recommending SATD within project?

RQ3

How would a method-level smell detector compare with TEDIIOUS?

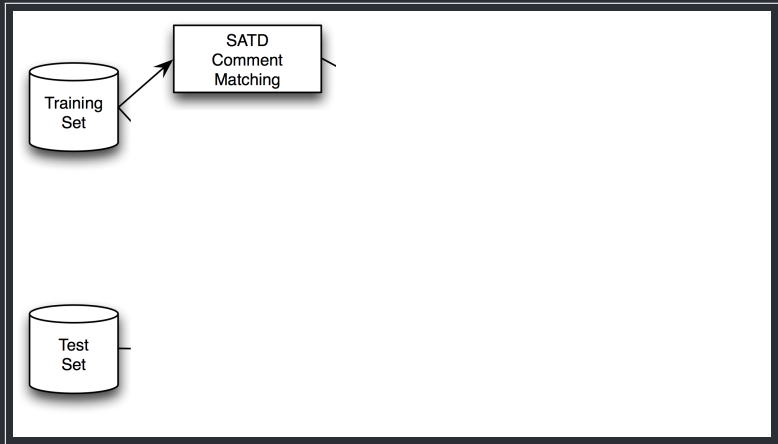
# The Approach

## *Train and Test Sets*



# The Approach

## *SATD Matching*



# The Approach

## *SATD Matching*

### Identification of Self-Admitted Technical Debt

Method-Level

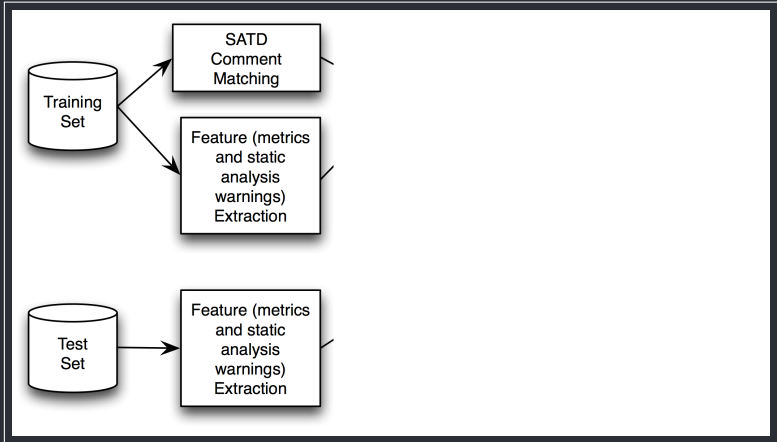
Pattern Matching

Tagging



# The Approach

## *Features Extraction*



# The Approach

## Features Extraction

```
<comment type="block" format="javadoc" pos:line="106" pos:column="5">/**
 * Get account using the email address to identify it.
 *
 * @param address
 *         email address
 * @return account item
 */</comment>
<function><type><specifier pos:line="113" pos:column="5">public</specifier> <name pos:line="113" pos:column="12">AccountItem</name></type> <name pos:
line="113" pos:column="24">getAccount</name><parameter_list pos:line="113" pos:column="34">(<param-decl><type><name pos:line="113" pos:column="35">
String</name></type> <name pos:line="113" pos:column="42">address</name></decl></param>)</parameter_list> <block pos:line="113" pos:column="51">{
```

## Source Code Metrics

XML Representation

Remove SATD Methods

Compute source code and readability metrics

# The Approach

## *Features Extraction*

### List of Metrics

LOC

Nb of Statements

Nb of Comments

McCabe Cyclomatic Complexity

Nb of Expressions Nb of Passed Param

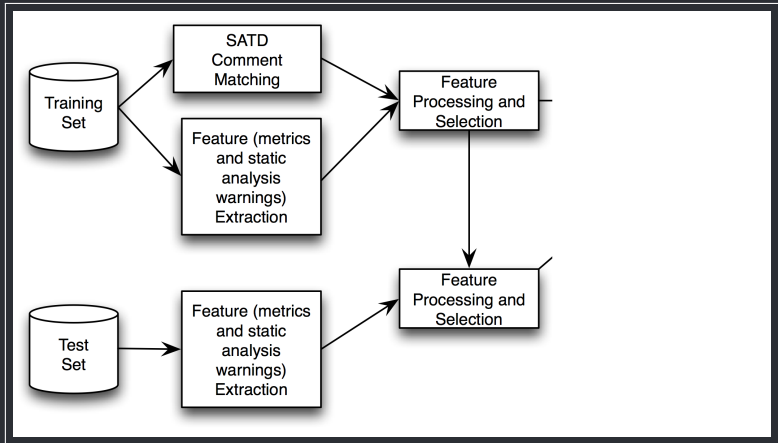
Nb of Identifiers

Nb of Call Sites

Nb of Declarations

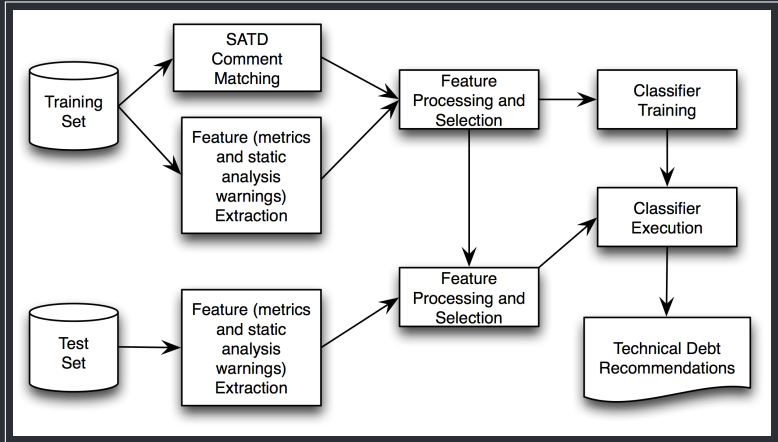
# The Approach

## Feature Preprocessing



# The Approach

## Machine Learners



Method-level Image du process (separer en differents process et expliquer) - Training and test sets - SATD matching - Features - Preprocessing - Building and applying machine learners (training and testing)

Analysis Method Within project performance Cross project performance TEDIOUS vs method level Qualitative analysis

Inspirations (Kim, Dos Santos) Description

Method-level Image du process (separer en differents process et expliquer) - Training and test sets - SATD matching - Source code - Preprocessing/Word Embeddings - Building and applying machine learners (training and testing)

Research questions (main and 3 sub-questions) 9 OOS + Java More method-related than class-related Unbalance

Source code comments only Source code with comments Source code without comments Source code partially with comments

Construct Internal Conclusion Reliability External

Number of metrics/warning Number of LOC Process inherent errors 21/22

Default configurations (no real optimizations) Generalization of results Only in Java Performance evaluation can be better for CNN  
TEDIOUS approach CNN approach Dataset TEDIOUS results CNN  
results Applicability scenarios - Recommendation system -  
Complement to smell detectors Improvements - Optimization -  
Pattern matching process - More examples to train - More positive  
examples - More metrics/warnings and source code - Other TD types  
- Extend to other languages and domains

# Bibliography

$T_{\text{E}}\text{X}$ ,  $\text{\LaTeX}$ , and Beamer

- [1] W Cunningham. *The wycash portfolio management system*. dans Addendum to the Proceedings on Object-oriented Programming Systems, Languages, and Applications (Addendum), série OOPSLA '92, New York, NY, USA: ACM, 1992, pp 29–30, DOI: 10.1145/157709.157715. En ligne: <http://doi.acm.org/10.1145/157709.157715>.