Towards Improving Software Architecture Degradation Mitigation by Machine Learning

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About the presenter

- Short resume
  - 2015- : Associate Professor in Computer Science at Karlstad University
  - 2014-2015: Research Fellow at Lero – The Irish Software Research Centre
  - 2011-2014: Post-doctoral researcher at Clausthal University of Technology

- Research Interests
  - Software Architecture, in particular degradation
  - Software Evolution and Modernization

What is this about?

Software Architecture Degradation

“The continuous divergence between the intended (prescriptive) and the as-implemented (descriptive) architecture.”

Machine Learning

Thanks to machine-learning algorithms, the robot apocalypse was short-lived.

“Software learning to perform a certain class of task better over time based on previous experience.”

Motivation

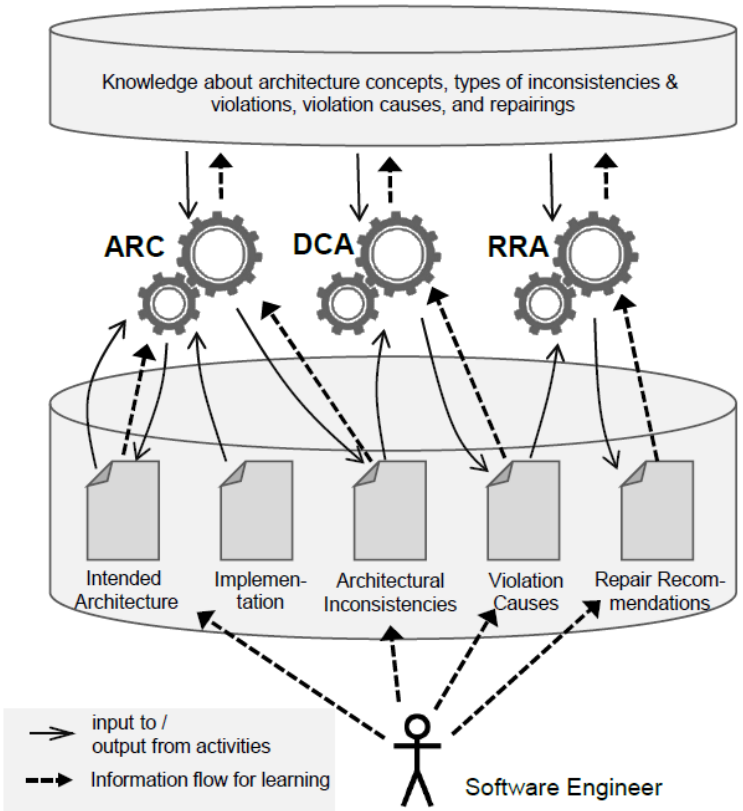
- Typical activities in architecture degradation mitigation
  - Architecture recovery
  - Consistency checking / degradation detection
  - Degradation analysis / comprehension
  - Degradation repairing

- All activities are labour-intensive and intellectually challenging

Motivating question:

How can we make use of machine learning to mitigate architecture degradation more efficiently?

Core of the approach



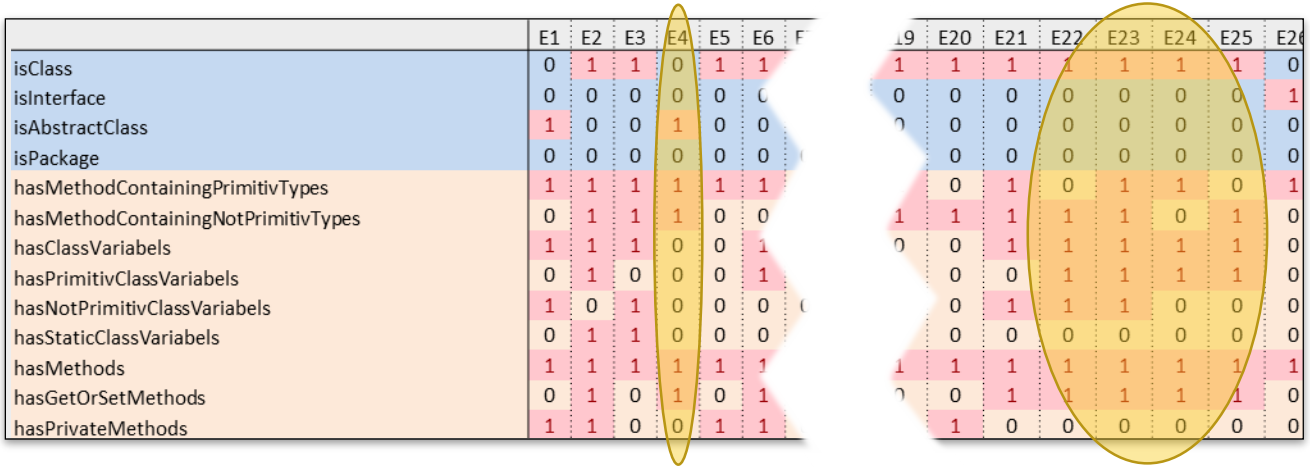
- **A**rchitecture **R**ecovery and **C**onsistency
  - Recover intended architecture from code
  - Check consistency between intended architecture and code
- **D**egradation **C**ause **A**nalysis
  - Identify the reasons for/causes of architectural inconsistencies, identify actual violations
- **R**ecommending **R**epair **A**ctions
  - Recommend refactoring of implementation or architectural adaptations to resolve degradation

ML in Architecture Recovery & Consistency

- Architecture Recovery often understood as: clustering implementation elements into architectural components
- Clustering a typical task in unsupervised learning

- There are more complex architectural concepts like patterns and guidelines
- Architectural decisions affect more than the system decomposition
- Result of recovery is most often not the intended but the implemented architecture

Main idea: express architectural concepts in terms of code-based features



	E1	E2	E3	E4	E5	E6	F
isClass	0	1	1	0	1	1	
isInterface	0	0	0	0	0	0	
isAbstractClass	1	0	0	1	0	0	
isPackage	0	0	0	0	0	0	
hasMethodContainingPrimitivTypes	1	1	1	1	1	1	
hasMethodContainingNotPrimitivTypes	0	1	1	1	0	0	
hasClassVariabels	1	1	1	0	0	1	
hasPrimitivClassVariabels	0	1	0	0	0	1	
hasNotPrimitivClassVariabels	1	0	1	0	0	0	
hasStaticClassVariabels	0	1	1	0	0	0	
hasMethods	1	1	1	1	1	1	
hasGetOrSetMethods	0	1	0	1	0	1	
hasPrivateMethods	1	1	0	0	1	1	

	E20	E21	E22	E23	E24	E25	E26
isClass	1	1	1	1	1	1	0
isInterface	0	0	0	0	0	0	1
isAbstractClass	0	0	0	0	0	0	0
isPackage	0	0	0	0	0	0	0
hasMethodContainingPrimitivTypes	0	1	0	1	1	0	1
hasMethodContainingNotPrimitivTypes	1	1	1	1	0	1	0
hasClassVariabels	0	1	1	1	1	1	0
hasPrimitivClassVariabels	0	0	1	1	1	1	0
hasNotPrimitivClassVariabels	0	1	1	1	0	0	0
hasStaticClassVariabels	0	0	0	0	0	0	0
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hasGetOrSetMethods	0	1	1	1	1	1	0
hasPrivateMethods	1	0	0	0	0	0	0

ML in Degradation Cause Analysis

- Idea: express causes as structural patterns + metrics

Violation Cause	Architecturally Misplaced Method
Description	A method *f* refers to a class/interface *D*, accesses one of its fields, or calls one of its methods, causing a violation, but *f* neither belongs to its containing class *C* nor to the surrounding module *M*.
Symptom	
Symptom	*modsim(f) = 0*
Symptom	*cohesion(f,C) = 0*
Symptom	*modsim(C) = 1*

S. Herold, M. English, J. Buckley, S. Counsell and M. Ó. Cinnéide, "Detection of violation causes in reflexion models," *2015 IEEE 22nd International Conference on Software Analysis, Evolution, and Reengineering (SANER).*

- Challenges
  - Interpretation of metric values for violations instances – symptoms express “ideal” case
  - Overlap with other violation causes
- Use of ML techniques
  - Classification of “safe” violation cause instances
  - Discovery of novel violation causes

ML in Recommending Repair Actions

- Refactoring recommender

[A, cannot, declare, B]		
B b; S	⟹ replace([B], [B']), if B' ∈ super(B) ∧ typecheck([B' b; S]) ∧ B' ∉ M_B	D1
B b; S	⟹ replace([B], [B']), if B' ∈ sub(B) ∧ typecheck([B' b; S]) ∧ B' ∉ M_B	D2
B b = exp; S	⟹ propagate([exp], b, [S]), if can(A, access, B)	D3
g (B b) { S }	⟹ remove([B b]), if typecheck([g(){ S }])	D4
catch (B b) { S }	⟹ replace([B], [B']), if B' ∈ super(B) ∧ typecheck([catch(B' b){ S }]) ∧ B' ∉ M_B	D5
[A, cannot, access, B]		
b.f	⟹ replace([b.f], [D; c.g]), if g = delegate(f) ∧ {D,c} = gen_decl(g) ∧ type(c) ∉ M_B	D6

Terra, R., Valente, M. T., Czarnecki, K., and Bigonha, R. S. (2015), A recommendation system for repairing violations detected by static architecture conformance checking, *Softw. Pract. Exper.*, 45, 315– 342,

- Fixed, predefined
  - Set of rules
  - Priorities of rules
  - Action parts

- Machine Learning could support
  - Adapt priorities by considering additional features and observing acceptance/rejection of recommendations
  - Observe additional manual actions after accepting recommendations and refine action parts / learn new rules

Conclusion

- We believe there is huge potential for ML in software architecture maintenance and evolution
- Access to data is expected to be a huge challenge
- Validated intended architectures and degradation analysis by experts needed for training

Thanks for your attention!

Any questions?

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