Artificial Intelligence Methods for Optimization of the Software Testing Process: with Several Industrial Case Studies

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Background

Education:

- Doctor of Philosophy in Computer Science, thesis entitled "Multi-Criteria Optimization of System Integration Testing", Mälardalen University, Sweden, 2018.
- Bachelor and Master of Philosophy in Applied Mathematics

Research Area:

- Software Testing
- Optimization
- Artificial intelligence
- Natural language processing





Publication

- Sahar Tahvili and Leo Hatvani. "Artificial Intelligence Methods for Optimization of the Software Testing Process With Practical Examples and Exercises". Elsevier, July 2022. This book has been published as a peer-reviewed manuscript by Elsevier (Academic Press). The presented results in this book have not been published before. Chapter 5 of this book includes 8 industrial use cases. Furthermore, both the proposal and manuscript underwent multiple revisions before publication.
- Michael Felderer, Eduard Paul Enoiu, and Sahar Tahvili. "Artificial Intelligence Techniques in System Testing". In optimizing the Software Development Process with Artificial Intelligence. Springer, July 2023.



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 This book presents one of the first empirical studies in the field, contrasting theoretical assumptions on innovations in a real industrial environment with a large set of use cases from developed and developing testing processes at various large industries.

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 the Contrasting theoretical assumptions on innovations in a real industrial reaction of the real software testing on developed/developing Al-based solutions; serves as a guideline for conducting industrial research on the Al & software testing domain; explains all proposed solutions through real industrial case studies.

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Springer

Challenges

- For testing any software applications (such as Telecom, and safety-critical systems) a set of test cases needs to be generated.
- Test automation requires deep knowledge in the domain (Telecom, safetycritical systems), programming (Java, C#), and software testing.
- A set of test environments (testbed, test channel) are required.
- A mapping between the test scripts and test environments must be done manually and daily, which is not scalable.
- Manual scheduling of the test scripts to the test environments has required a team where they need to master:
 - The capacity, capability, and configuration for each test environment
 - The functional dependencies between test scripts
 - The duration of each test script (based on the lines of the code)





- Time and resource-consuming manual process.
- Requires deep knowledge in the domain, software testing, and test automation.
- Can not handle a large set of features and test
 environments.
- Paying no attention to the dependencies between test cases can lead to unnecessary failures between test cases by up to 40%.
- Running a test case on the wrong test environment leads directly to test failure.



Optimization Problems

- Test Environment Optimization
 - Dynamic scheduling and building test environments
- Test Execution Optimization
 - Parallel test execution
 - Faster troubleshooting processes
- Al-based Test Management System
 - Test automation
 - CI/CD Configuration Selection



A segment of the impact of artificial intelligence in software development during 2020-2021



Source: Tahvili. S and Hatvani. L, Artificial Intelligence Methods for Optimization of the Software Testing Process: With Practical Examples and Exercises, 2022

Al-aided Content Analysis: Transforming Unstructured Text into Structured Data for Efficient Research



 Collaborative, automated physical world
 Connected intelligent machines
 The internet of senses

A review of the history of text analytics



Source: Tahvili. S, Hatvani. L, Artificial Intelligence Methods for Optimization of the Software Testing Process: With Practical Examples and Exercises

Machine Learning

- NLP can be used to extract and analyze requirements from natural language documents, user feedback, and emails. It helps in understanding and prioritizing customer needs more efficiently.

Requirement Analysis SOFTWARE Evolution Design DEVELOPM CYCLF LIFE Testing Implementation

A schematic diagram of a machine learning-based approach for text mining.

- NLP can assist in test case generation by analyzing natural language requirements and converting them into test scenarios.

- **Machine Learning** can be applied to automated testing, where it learns from historical testing data to identify patterns and potential issues in the software.

- **Machine Learning** can be used for code generation, code completion, and code review. It can suggest code snippets and help identify potential coding errors.

Deep Learning

The neural system architecture of word embeddings.

Case Study 1:

- Automated Test Case Generation
- Test Data Generation
- Semantic Analysis

The spiral graph shows the dependencies between the requirements.

The clustered requirements specifications using the SBERT model. Minimizing test execution failure based on the dependencies up to 40%.

Case Study 2: Intelligence Test Management System

Intelligent Test Management System

Reads multiple types of inputs (controlled, noncontrolled natural text)

The proposed solution can create a test case when the minimal requirements are specified.

| Test Setup | | Test Procedure | | |
|-------------------------------|--------------------|--------------------|--|--|
| | | | | |
| Pass Criteria | | Pre-Post Condition | | |
| | | | | |
| Test instruction text default | Test Configuration | SW Track | | |
| Tarrad Microcerrica | Test Duration | CHS Varification | | |
| Tugges and over the | 105 Diraton | | | |
| EPIC | Test framework | Context Name | | |
| | | | | |
| | | | | |
| | Decision Support | | | |
| CI Configuration | | Automation | | |
| Pipeline 1 | | Automation | | |
| | | | | |
| | | | | |
| | | | | |

complexity of the inserted input by the end-user

Case Study 3: Dynamic scheduling and building test environments

Real-time Optimization of Testbeds for Cloudified Radio Access Networks Using Artificial Intelligence

| | Real time Optimization | on of Testbeds | | | |
|---|--------------------------|--|------------------------------|---------------------------------------|---|
| Metadata | | | | | |
| Request Number | Start Date yyyy-mm-dd | | End Date | n-dd 💼 | |
| Priority | | | | | |
| Testbeds | | | | | |
| Bin_cat Feature_0 Feature_1 Feature_2 | Feature_3 Feature_4 V | Feature_5 | | 2024 | |
| Multi_cat_feature_0 | Multi_cat_feature_1 0 | December | January February | March April | May June |
| Multi_cat_feature_2 | Multi_cat_feature_3 | $\begin{array}{c ccccccccccccccccccccccccccccccccccc$ | 2 3 4 5 6 7 8 9 Request 9 | 10 11 12 13 14 15 16 Request 41 | 17 18 19 20 21 22 23 24 25 26 Request 11 |
| const_feature_1 5 const_feature_2 3 const_feature_1 | const feature 3 4 | T2 Req T3 Req T4 Req | uest 8 | Request 3 Request 13 Request 18 | Request 5 |
| | | T_5 | Request 118 | Request 23 | Bound 2 |
| Component 1 | Component 2 | | Request / | Request 127 | Request 2 |

Some of the common artificial intelligence challenges in the industry *≸*

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• Al strategies

Architecture

- Cyber security risk
- Supporting IT systems
- Integration with Cloud

Complexity

Lack of AI explainability

- Lack of skilled workers
- Operational complexity
- High computing power
- Meta learning

Training data

- Data management
- Data governance
- Insufficient data
- Inaccurate data

Data

Ongoing fee

- Implementation cost
- Maintenance cost
- Quantifying ROI

Cost

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