

Test Selection and Augmentation of Regression System Tests for Access Control Policy Evolution

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Abstract—As security requirements of software often change during development and maintenance, developers may modify policies according to evolving requirements. In order to increase confidence that the modification of policies is correct and does not introduce unexpected behavior, developers conduct regression testing. In order to reduce the cost of regression testing, we develop a safe regression-test-selection approach, which selects every test case that may reveal a fault in program code impacted by policy changes. Our approach develops three techniques: the first one is based on mutation analysis (that converts each rule’s decision in turn and executes and finds test cases related to each rule), the second one is based on coverage analysis (that records which rules are evaluated by executing each test case), and the third one is based on evaluated decisions of requests issued from test cases. We develop a test augmentation technique, which complements our test-selection techniques by generating additional test cases to cover not-covered-impacted-rules by existing test cases. We evaluate our approach on three real world Java programs interacting with policies. Our evaluation results show that our test selection techniques find XX test cases from an existing test cases, which covers only XX% of impacted rules on average. Our evaluation results show that our test augmentation technique generate additional test cases to cover 100% of the impacted rules.

Keywords—access control policy evolution; regression testing; test selection; test augmentation; reliability

I. INTRODUCTION

Access control is one of the privacy and security mechanisms that authorizes legitimate users to access or share critical information. Access control mechanisms are often governed by access control policies, each of which typically includes a set of rules to control which subjects can be permitted or denied to access which resources in which conditions. To facilitate specify and maintain policies in practice, system developers specify and enforce policies independently from actual functionality (i.e., business logic) of a system.

More specifically, program code, which represents the functionality, interacts with a specified policy through a security component, called policy decision point (PDP). Consider that the program code consists of methods Ms , which require decisions (e.g., permit or deny) to determine whether a given subject can have access on critical information. Typically, Ms formulate an access request that specifies that

a subject would like to have access on critical information. The Ms next submit the request to a PDP, which evaluates the request against the policy and determines whether the request should be permitted or denied. Finally, the PDP formulates and sends the decision to Ms to proceed.

In this paper, we focus on the regression testing problem in the context of policy evolution. The typical regression testing for program code interacting with a policy is as follows. Given a program code C and a policy P , the developers prepare initial system test cases, where each test case maps to rules (in the policy) exercised by the test case. Given P and its modified policy P' , the developers compare impacts of P and P' to reveal different policy behaviors, which are “dangerous” portion DP to be validated with test cases. For validating this “dangerous” portion, the developers often select only test cases (from test cases for P), which exercise DP of P' .

As security requirements of software often change during development and maintenance, system developers may modify policies according to the requirements. For example, new security requirements include new security concerns to be added into a policy. Developers may change policies without changing program code related to actual system functionality. In order to increase confidence that the modification of policies is correct and does not introduce unexpected behavior, developers periodically conduct regression testing. For policy evolution, regression testing is important because policy behavior changes may result in unexpected behaviors of program code, these behaviors can even be undesirable. Consider that the developers add a permit rule D_1 into an existing policy without any changes to the original program code. Developers validate by generating executing test cases for methods impacted by this single policy change.

For regression testing, instead of writing new system test cases, developers often write initial system test cases and reuse the test cases in practice. The naive regression testing strategy is to rerun all system test cases. However, this strategy is costly and time-consuming, especially for large-scale systems. Moreover, if the number of the initial system test cases is large, this strategy may require a significant time for developers to conduct testing. In order to reduce the cost of regression testing, developers often adopt regression

test selection, which selects and executes only test cases to expose different behaviors across different versions of the system. This approach also requires substantial costs to select and execute such system tests from the initial test cases that could reveal faults introduced by the changes. If regression-test-selection cost is smaller than rerunning all of initial system test cases, test selection helps reduce time-consuming task in validating whether the modification is correct.

In order to address the issue, we propose a safe regression-test-selection approach, which selects every test case that may reveal a fault in program code impacted by policy changes. In general, our approach automatically compares an original policy P and its modified policy P' to detect different policy behaviors. In the policy context, different policy behavior refers that, given a request, its evaluated decisions (for P and P' , respectively) are different. Our approach detects policy rules, which map such requests. These rules are dangerous portions to be validated. In our approach, we have developed three automated tools for three test-selection techniques: The first one is based on mutation analysis (that converts each rule's decision in turn and executes and finds test cases related to each rule), the second one is based on coverage analysis (that records which rules are evaluated by executing each test case), and the third one is based on evaluated decisions of requests issued from test cases. Our approach next selects only test cases related to identified rules impacted by policy changes.

While test-selection techniques are useful for selecting test cases for program code impacted by policy changes, these test cases may not sufficiently cover all of rules R s of the policy impacted by the changes. We propose a test augmentation technique, which complements our test-selection techniques by generating additional test cases to cover not-covered-impacted-rules by existing test cases.

This paper makes three main contributions:

- We develop three safe test selection techniques to select only and every test cases from an existing test cases to test program code impacted by policy changes. **Is there any section that discusses the safety of the three algorithms: I am not sure that the three techniques are safe**
- We develop a test augmentation technique to generate additional new test cases to cover not-covered-impacted-rules with existing test cases selected by the preceding technique.
- TBD: Implementation and cost and benefit comparison
- **TBD: compare cost and benefits** We evaluate our approach on three real world Java programs interacting with policies. Our evaluation results show that our test selection techniques find XX test cases from an existing test cases, which covers only XX% of impacted rules on average. Our evaluation results show that our test augmentation technique generate additional test cases

to cover 100% of the impacted rules.

The rest of the paper is organized as follows.

II. BACKGROUND

A. Policy-based Software Systems: PEP-PDP separation

Policy-based software systems are regulated by access control policies that specify the different actions that subjects are allowed to perform on system resources. The policy is enforced by one or multiple Policy Enforcement Points (PEP) at the application level. A given PEP receives a subject access request and sends it to a Policy decision Point (PDP) that evaluates access control requests against policy rules. The evaluation process involves resolving all the rules in the policy. Current access control implementations follow a design strategy that separates between the PEP and the PDP, this separation is advocated mainly for two reasons:

- Specifying the policy independently from the program code enables to save time, cost and effort during the maintenance phase since developers are not required to update the application code at each policy update.
- When the policy is not hard-coded into the application, it can be specified in a standardized language and transported through many different systems platforms.

Access control policies can be specified in many languages like EPAL [1] or XACML [2]. In what follows, we consider XACML policies written in XACML language.

B. XACML Policies

This section describes the background in the field of XACML policies, policy models, and regression testing. XACML (eXtensible Access Control Markup Language) [2] is a language specification standard published by the Organization for the Advancement of Structured Information Standards (OASIS). An XACML access control specification consists of a policy set and a policy combining algorithm. XACML supports various policy models such as Role Based Access Control (RBAC) [4], [5] and Organization-Based Access Control (OrBAC) [3]. For RBAC and OrBAC, various job functions are associated to roles (e.g., staff or employee) that a user possesses. Permissions or denials to take an action on certain objects are assigned to specific roles (instead of users). OrBAC supports extra features such as role, activity (i.e., action), and view (i.e., object) hierarchy. Figure 1 is an OrBAC policy example where roles such as "Secretary" and "Borrower" are associated with subject attributes and policy decisions are made through specific roles. Let *Conditions* be the set of the environmental context such as working days or holidays, a file size, and a user's task authorization level. Let S , O , A , and C denote all the subjects, objects, actions, and *Conditions*, respectively, in an access control system. An XACML policy consists of a *policy set*, which consists of *policy sets* and *policies*. A *policy* consists of a sequence of *rules*, each of which specifies under what conditions C

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1 <Policy PolicyId="Library Policy" RuleCombAlgId="Permit-overrides">
2   <Target/>
3   <Rule RuleId="1" Effect="Permit">
4     <Target>
5       <Subjects><Subject> BORROWER </Subject></Subjects>
6       <Resources><Resource> BOOK </Resource></Resources>
7       <Actions><Action> BORROWERACTIVITY </Action></Actions>
8     </Target>
9     <Condition>
10      <AttributeValue> WORKINGDAYS </AttributeValue>
11    </Condition>
12  </Rule>
13  <Rule RuleId="2" Effect="Deny">
14    <Target>
15      <Subjects><Subject> BORROWER </Subject></Subjects>
16      <Resources><Resource> BOOK </Resource></Resources>
17      <Actions><Action> BORROWERACTIVITY </Action></Actions>
18    </Target>
19    <Condition>
20      <AttributeValue> HOLIDAYS </AttributeValue>
21    </Condition>
22  </Rule>
23  <Rule RuleId="3" Effect="Permit">
24    <Target>
25      <Subjects><Subject> SECRETARY </Subject></Subjects>
26      <Resources><Resource> BOOK </Resource></Resources>
27      <Actions><Action> FIXBOOK </Action></Actions>
28    </Target>
29    <Condition>
30      <AttributeValue> MAINTENANCEDAY </AttributeValue>
31    </Condition>
32  </Rule>
33  <!-- A final, "fall-through" rule that always Denies -->
34  <Rule RuleId="FinalRule" Effect="Deny"/>
35 </policy>

```

Figure 1. An example policy specified in XACML

subject S is allowed or denied to perform action A (e.g., read) on certain object (i.e., resources) O in a system.

A rule R is of the following form:

$$R \subseteq S \times A \times O \times C \longrightarrow Dec$$

where Dec denotes a decision, which is either permit or deny. A user's request is evaluated against a policy. A request matches with a rule if the request satisfies the rule's attributes. Then the rule's decision can be returned. Formally, a request Q in the following form matches R :

$$Q \subseteq S_q \times A_q \times O_q \times C_q \text{ where } S_q \subseteq S, A_q \subseteq A, O_q \subseteq O, \text{ and } C_q \subseteq C.$$

More than one rule in a policy may be applicable to a given request. The *rule combining algorithm* is used to combine multiple rule decisions into a single decision. There are four standard rule combining algorithms: deny-overrides, permit-overrides, first applicable, and only-one-applicable. The *deny-overrides algorithm* returns Deny if any rule evaluation returns Deny or no rule is applicable. The *permit-overrides algorithm* returns Permit if any rule evaluation returns Permit. Otherwise, the algorithm returns Deny. The *first-applicable algorithm* returns whatever the evaluation of the first applicable rule returns. The *only-one-applicable algorithm* returns the decision of the only applicable rule if there is only one applicable rule, and returns error otherwise. Figure 1 shows a policy specified in XACML. The policy

consists of three XACML rules used in our real-life library access control policy subject, called *LSM* [11]. Note that we simplified XML formats to reduce space for this example. Lines 3-12 describe a rule that borrower is permitted to borroweractivity (e.g., borrowing books) book in working days. Lines 13-22 describe a rule that borrower is denied to borroweractivity book in holidays. Lines 23-32 describe a rule that secretary is permitted to fixing for book in maintenance day. As these rules are combined using the permit-overrides algorithm, the *permit-overrides algorithm* returns the Permit if a request matches at least one of rules with permit decisions. For example, if rules 2 and 3 are applicable to a request, the decision evaluated from rule 3 is given higher priority than that of rule 2.

C. Regression Testing

Software testing [12] refers to the activity of generating Tests Cases to verify the conformity of output results provided by a program to the expected output that meets its functional and non functional requirements. With the increasing complexity of software systems, this activity is gaining more and more interest in the research field and aiming to establish a certain trade-off between budget, time and quality.

Software is subject to changes occurring at the design stage or in later stages at the deployment or maintenance phases. These changes are usually supposed to meet changes in the requirements or to overcome errors that can be detected in later stages of software life cycle. Regression testing refers to the research field that is interested in retesting the system to verify that the new changes have not altered the initial system behavior. As highlighted by Rothermel et al in [13], regression testing is defined like the following: "Given a program P , a modified version P' , and a set T of test cases used previously to test P , regression analysis and testing techniques attempt to make use of a subset of T to gain sufficient confidence in the correctness of P' with respect to behaviors from P retained in P' ". The main objectives of regression testing is to reduce the costs from rerunning initial test cases that have been performed before the deployment of the software product, and to maximize the capability of selected tests to detect potential errors that are induced by changes.

1) Regression testing: Test selection and augmentation:

Regression test selection techniques involve two aspects that are usually considered conjointly:

- The first aspect is the regression test selection problem which aims at minimizing the number of tests that have to be considered Rothermel et al. [13] have provided in their work a survey on techniques that have been used in this field. The regression test selection technique is called safe when the selected tests among the initial test suite enable to detect 100% of the faults.

- The second aspect is augmentation techniques and aims at adding new tests to the initial Test Suite to cover changed parts.

2) *Regression testing in the context of software policy-based systems*: In this work, we consider regression testing in the context of policy-based software systems. We aim to verify that the changed parts of the policy have not impacted the access control mechanisms implemented in the system. portions to be validated.

The figure 1 ?? illustrates the overall scenario ADD schema

III. APPROACH

As manually selecting system tests for regression testing is tedious and error-prone, we have developed three techniques to automatically select system tests for regression in policy evolution. Among existing system tests, the objective is to select all system tests for regressing testing as follows. Our approach takes two versions of program code, which interact with v_1 (original) and v_2 (new) access control policies, respectively. Our approach takes existing system tests as an input; these tests invoke methods in program code. Our approach analyzes given program code and polices to select *only* system tests for regression testing in case of policy evolution. Among given system tests, our selected system tests invoke methods to reveal changed policy behaviors between v_1 and v_2 .

While we select all of system tests for regression, our test selection techniques cannot guarantee sufficiency of regression testing (i.e., whether tests are sufficient to reveal all of changed policy behaviors). We have developed a technique to automatically augment new system tests to satisfy sufficiency as follows. We measure sufficiency of regressing testing with our selected system tests based on rule coverage criteria []. For not-covered changed policy behaviors (i.e., rules), our approach automatically generates system tests to cover such behaviors.

Formally, C_1 denotes program code, which interacts an access control policy P_1 . C_2 denotes program code, which interacts a modified access control policy P_2 . T_0 denotes an initial test suite for C_1 . Our first step involves the regression test selection. We select $T' \subseteq T_0$ where T' is a set of test cases. T' execute on C_2 and reveal changed policy behaviors between P_1 and P_2 . We measure coverage of changed policy behaviors of P_1 and P_2 with T' . If we find not-covered policy behaviors, we create T'' , a set of new system tests for C_2 .

We next describe our proposed three test selection techniques and test augmentation technique.

A. Test Selection via Mutation Analysis

Our first proposed technique uses mutation analysis to select system tests as follows. First step is to set up rule-test

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1 <Policy PolicyId="Library Policy" RuleCombAlgId="Permit-overrides">
2   <Target/>
3   <Rule RuleId="1" Effect="Deny">
4     <Target>
5       <Subjects><Subject> BORROWER </Subject></Subjects>
6       <Resources><Resource> BOOK </Resource></Resources>
7       <Actions><Action> BORROWERACTIVITY </Action></Actions>
8     </Target>
9     <Condition>
10      <AttributeValue> WORKINGDAYS </AttributeValue>
11    </Condition>
12  </Rule>
13  ...
35 </policy>

```

Figure 2. An example mutant policy by changing $R1$'s rule decision (i.e., effect)

correlation. Given a policy P , we creates its rule-decision-change (RDC) mutant policy Pr_i by changing decision (e.g., Permit to Deny) of rule r_i in turn. An example mutant policy is shown in Figure 2. The mutant policy mutates Rule 1's decision in the original example policy in Figure 1; original Rule 1's decision Permit is changed to Deny. The technique finds affected tests for this rule decision change. The technique executes system tests T on program code for P and Pr_i , respectively. To detect changed policy behaviors, the technique monitors responses of evaluated requests formulated from system tests ST . When the technique find system tests, which evaluate different policy decisions against P and P' , respectively, our technique correlates rule r_i and ST . We continue our technique until we find each rule's correlated system test cases in turn.

Second step is to select system tests for regression on P and its modified policy P_m . Our technique conducts change impact analysis of P and P_m to find which rules' decision are changed. We select system tests correlated with such decision changed rules.

While the technique can quickly select system tests in the second step, the technique requires rule-test setup (in the first step), which could be costly in terms of execution times. Given n rules in P , we execute T for $2 \times n$ times. As the first step is applied for only existing rules R in P , our technique requires addition of rule-test correlation for newly added rules R_n where $R_n \notin R$ in P_m . In addition, if a new system test is introduced, we execute this test for $2 \times n$ times. However, an important benefit is that we are enabled to conduct rule-test set-up once before encountering policy changes in terms of correlated rules.

B. Test Selection via System Test Execution

Our previous technique finds correlation of all of existing rules N in a given policy. To reduce such correlation setup efforts, we develop a technique to correlate only rules, which can be evaluated by system tests. Our intuition is that system tests may interact only small number of rules in a policy instead of a whole rules in the policy. Therefore, we require correlation of system tests for only such small number of rules. First step is to set up rule-test correlation. The technique executes system tests T on

Algorithm 1: Test Selection via Mutation Analysis Algorithm

```

TestSelection1( $P, P_m, T$ ):  $T'$ 
Input: XACML Policy  $P$ , modified policy  $P_m$ , Initial Sytem Test Cases  $T$ 
Output:  $T' \subseteq T$  where  $T'$  is the subset of  $T$  selected for use in regression testing  $P_m$ 
 $T' = \emptyset$ 
/*Rule-test set-up phase*/
for each rule  $r_i$  in Policy  $P$  do
   $T_{r_i} = \emptyset$  where  $T_{r_i} \subseteq T$  are the tests correlated to  $r_i$ 
  /*We mutate the policy  $P$  by creating a rule-decision change (RDC) on  $r_i$  to get  $P_{r_i}$ */
   $P_{r_i} \xleftarrow{RDC(r_i)} P$ 
  Execute  $T$  with  $P_{r_i}$ 
  for each  $t$  in  $T$  do
    Let  $E(t)$  be the test execution result,
     $E(t) = \text{Success}, \text{Failure}$ 
    if  $E(t) \leftarrow \text{Failure}$  then
       $T_{r_i} \leftarrow T_{r_i} \cup t$ 
    end if
  end for
  Map( $r_i, T_{r_i}$ )
end for
/*Test selection phase*/
 $\{r_m\}_{i=1..m} \leftarrow \text{diff}(P, P_m)$ 
for Each rule  $r_i$  in  $\{r_m\}_{i=1..m}$  do
   $T' \leftarrow T' \cup T_{r_i}$ 
end for
return  $T'$ 

```

program code for P . The technique monitors which rules in a policy are evaluated with requests formulated from system tests ST . Our technique correlates rule r_i and ST . We continue our technique until we find correlated system test cases with rules in a policy. Second step is the same with the previous technique. This step is to select system tests for regression on P and its modified policy P_m . Our technique conducts change impact analysis of P and P_m to find which rules' decision are changed. We select system tests correlated with such decision changed rules. Both the two techniques are white-box testing, where access control policies are available. An important benefit of this technique is to reduce cost in terms of mutation analysis and execution times. This technique does not require generating mutants by changing rule's decision in turn. Moreover, the technique can significantly reduce execution time. While the technique can quickly select system selects in the second step, the technique requires rule-test setup (in the first step), which could be costly in terms of execution times. Consider that requests rs are formulated from system tests interact only n_1 rules ($n_1 \leq n$) in a policy. We execute T only once. Our

Algorithm 2: Test Selection via System Test Execution

```

TestSelection2( $P, P_m, T$ ):  $T'$ 
Input: XACML Policy  $P$ , modified policy  $P_m$ , Initial Sytem Test Cases  $T$ 
Output:  $T' \subseteq T$  where  $T'$  is the subset of  $T$  selected for use in regression testing  $P_m$ 
 $T' = \emptyset$ 
/*Rule-test set-up phase*/
for Each test case  $t$  in  $T$  do
  Execute  $t$  with  $P$ 
   $MAP_1 = \text{Map}(t, \{r_p\}_{i=1..p})$  where  $\{r_p\}_{i=1..p}$  are the rules in  $P$  that are triggered by  $t$ 
end for
Generate a mapping rule/test from  $MAP_1$ 
 $MAP_2 = \text{Map}(r_i, T_{r_i}) \leftarrow MAP_1$ 
/*Test selection phase*/
 $\{r_m\}_{i=1..m} \leftarrow \text{diff}(P, P_m)$ 
for Each rule  $r_i$  in  $\{r_m\}_{i=1..m}$  do
   $T' \leftarrow T' \cup T_{r_i}$ 
end for
return  $T'$ 

```

technique requires addition of rule-test correlation for newly added rules R_n where $R_n \notin R$ in P_m as the same with the previous technique.

C. Test Selection via Play Back

To reduce such correlation setup efforts in the previous techniques, we develop a technique, which does not require correlation setup. Our approach executes system tests T on program code for P . Our approach records all requests issued to policy decision point (PDP) for each system test case. For test selection, our technique evaluates all issued requests against P and P_m , respectively. Our technique selects requests (with corresponding system test cases) to evaluate different decisions for two different policy versions

Our previous two techniques require correlation rule-test setup. The techniques analyze two versions of policies statically or dynamically under test to find which rules' are changed. However, these technique require correlation setup. For this approach, we execute all of system test cases for only once. For additional system tests, we execute the tests only once. An important benefit is that, we don't require that access control policies are available as the preceding two approaches. For security reasons, developers may not want to reveal their access control policies.

D. Test Augmentation

TBD

Algorithm 3: Test Selection via Play Back

TestSelection3(P, P_m, T): T'

Input: XACML Policy P , modified policy P_m , Initial Sytem Test Cases T

Output: $T' \subseteq T$ where T' is the subset of T selected for use in regression testing P_m

$T' = \emptyset$

$R_{T'} = \emptyset$ where $R_{T'}$ are the requests corresponding to T'

Execute system requests R

for each request Re in R **do**

if $decision(Re/P) \neq decision(Re/P_m)$ **then**

$R_{T'} \leftarrow R_{T'} \cup Re$

end if

end for

$T' \leftarrow R_{T'}$

return T'

IV. IMPLEMENTATION

V. EXPERIMENT

We carried out our evaluation on a desktop PC, running Windows 7 with Intel Core i5, 2410 Mhz processor, and 4 GB of RAM. We have implemented test selection techniques and mutant generation in Java. We generate mutants by changing rules in access control policies To simulate regression in access control policies. We used three types of mutants injected in the policies; the first one is RMR (Rule Removal), RA (Rule Addition), and CRE (Change Rule Effect) mutants. [ToDo: citation and explain more] In our mutants, we only change rules in access control policies, we do not simulate regression in test code in subjects. To measure the efficiency of our three techniques, we conducted evaluation as follows. We compared elapsed time to analyze test-rule correlation analysis, change impact analysis, and test selection by each technique. For the first two techniques, we require test-rule correlation analysis and change impact analysis, which should be done before actual test selection. We compare also selected tests by our three techniques to compare that all of these techniques return the same tests for regression. The objective of this evaluation is to investigate how our approach impacts performance for subjects and safety to select all tests for regression.

A. Subjects

The subjects include three real-life Java programs each which interact with access control policies [?]. We next describe our three subjects.

- Library Management System (LMS) provides web services to manage books in a public library.
- Virtual Meeting System (VMS) provides web conference services. VMS allows users to organize online meetings in a distributed platform.

- Auction Sale Management System (ASMS) allows users to buy or sell items on line. A seller initiates an auction by submitting a description of an item she wants to sell with its expected minimum price. Users then participate in bidding process by bidding the item. For the bidding on the item, users have enough money in her account before bidding.

Table I shows information in our subjects. [ToDo: explain more] Table II shows information in our subjects. [ToDo: explain more] Our subjects are equipped with Sun PDP [?], which is a popularly used PDP to evaluate requests. Policies in LMS, VMS, and ASMS contain a total of 720, 945, and 1760 rules, respectively.

B. Objectives and Measures

In the evaluation, we intend to answer the following research questions:

- RQ1: How many of test cases in a test suite (from an existing test suite) selected by our test selection techniques? This question helps to show that our techniques can reuse ZZ% of a test suite. We also compare our approach with random test select technique to show how effectively our approach selects test suite to cover changed policy behaviors.
- RQ2: Do our protest selections are safe? This question helps to show that our techniques select all tests, which shows only and all test cases to reveal changed behaviors in access control policies.
- RQ3: What are elapsed time for our techniques to conduct test selection by given subjects? This question helps to compare performance of our techniques by measuring efficiency with regards to elapsed time.
- RQ4: How higher we can achieve additional policy coverage ratio by our test augmentation technique? This question helps to show that our technique can generate/augment test suite to cover 100% of changed policy behaviors. We also compare our approach with random test generation technique to show how effectively our approach augment test suite to cover not-coverted changed policy behavriors.

C. Metrics

We use following 4 metrics in our evaluation.

- Policy coverage information for changed policy behaviors.
- Number of test cases reused by the test selection technique.
- Elapsed time of test-rule correlation, change impact analysis, and test selection.
- Number of test cases generated by the test augmentation technique.

[ToDo: explain more]

Table I
SUBJECTS

	LOC	# of Test Methods	# of Security Test Methods	# of Covered Rules	# of Not-covered Rules	% of Covered Rules
LMS	3749	46	29	42	0	100
VMS	3734	52	10	13	106	12
ASMS	7836	93	91	109	21	83

Table II
ACCESS CONTROL POLICIES IN OUR SUBJECTS

	# Subjects	# Actions	# Resources	# Conditions	# Explicit Rules	# Implicit Rules	# Total Rules
LMS	6	10	3	4	42	678	720
VMS	7	15	3	3	106	839	945
ASMS	8	11	5	4	129	1631	1760

VI. RELATED WORK

Our previous work developed policy testing approaches for policy structural coverage [10], request generation [8], and mutation testing [9]. Our previous work [6] also proposed a generic model-based conformance checking approach for access control policies written in XACML. These pieces of work do not rely on properties for generating test requests to detect a fault in a policy. Our previous work [7] developed an approach for measuring the quality of policy properties in policy verification. Given user user-specified properties, the quality of properties are measured based on fault-detection capability. While these approach focus on test request generation, in this paper, our technique targets at test selection (among existing system tests) for policy evolution.

Related Work here!!

VII. CONCLUSION

Conclusion here!!

REFERENCES

- [1] IBM, Enterprise Privacy Authorization Language (EPAL), Version 1.2 . <http://www.w3.org/Submission/2003/SUBM-EPAL-20031110>, 2003.
- [2] OASIS eXtensible Access Control Markup Language (XACML). <http://www.oasis-open.org/committees/xacml/>, 2005.
- [3] A. Abou El Kalam, R. El Baida, P. Balbiani, S. Benferhat, F. Cuppens, Y. Deswarte, A. Miège, C. Saurel, and G. Trouessin. Organization Based Access Control. In *Proc. 4th IEEE International Workshop on Policies for Distributed Systems and Networks (Policy 203)*, June 2003.
- [4] Annie Anderson. XACML profile for role based access control (RBAC). OASIS Committee Draft 01, 2004.
- [5] David F. Ferraiolo, Ravi Sandhu, Serban Gavrila, D. Richard Kuhn, and Ramaswamy Chandramouli. Proposed NIST standard for role-based access control. *ACM Trans. Inf. Syst. Secur.*, 4(3):224–274, 2001.
- [6] Vincent C. Hu, Evan Martin, JeeHyun Hwang, and Tao Xie. Conformance checking of access control policies specified in XACML. In *Proc. 1st IEEE International Workshop on Security in Software Engineering (IWSSE)*, 2007.
- [7] Evan Martin, JeeHyun Hwang, Tao Xie, and Vincent Hu. Assessing quality of policy properties in verification of access control policies. In *Proc. Annual Computer Security Applications Conference (ACSAC)*, pages 163–172, 2008.
- [8] Evan Martin and Tao Xie. Automated test generation for access control policies via change-impact analysis. In *Proc. 3rd International Workshop on Software Engineering for Secure Systems (SESS)*, pages 5–11, 2007.
- [9] Evan Martin and Tao Xie. A fault model and mutation testing of access control policies. In *Proc. 16th International Conference on World Wide Web (WWW)*, pages 667–676, 2007.
- [10] Evan Martin, Tao Xie, and Ting Yu. Defining and measuring policy coverage in testing access control policies. In *Proc. 8th International Conference on Information and Communications Security (ICICS)*, pages 139–158, 2006.
- [11] Tejeddine Mouelhi, Yves Le Traon, and Benoit Baudry. Transforming and selecting functional test cases for security policy testing. In *Proc. 2nd International Conference on Software Testing, Verification, and Validation (ICST 2009)*, 2009.
- [12] Glenford J. Myers. *Art of Software Testing*. John Wiley & Sons, Inc., New York, NY, USA, 1979.
- [13] Gregg Rothermel and Mary Jean Harrold. Analyzing regression test selection techniques. *IEEE Trans. Softw. Eng.*, 22:529–551, August 1996.