An Empirical Study of Retrofitting Legacy Unit Tests for Parameterized Unit Testing

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ABSTRACT

Owing to the significance of unit testing in the software development life cycle, unit testing has been widely adopted by software industry to ensure high-quality software. It is labor-intensive to write high-covering tests to ensure the software quality and therefore calls for automation. One such automatic testing technique to achieve high-covering tests is to write Parameterized Unit Tests (PUTs) and use them in combination with a test generation tool that accepts these PUTs. PUTs are a generalized form of conventional unit tests that accept parameters and where programmers can describe the expected behavior or specifications in a generic manner. An automatic test generation tool such as Pex from Microsoft Research accepts these PUTs and generates a set of conventional unit tests to achieve a high code coverage. We conduct an empirical study to investigate the utility of PUTs to generate highcovering conventional unit tests. In our empirical study, we use an open source C# project, called NUnit, to study the benefits of PUTs over conventional unit tests. In this paper, we present our empirical study carried out in two phases: test generalization (transforming existing manually written conventional unit tests to PUTs) and writing new PUTs (to cover the code portions under test that are not covered by the transformed PUTs). In our study, we found that test generalization increased the average block coverage by 9.68% with a maximum increase of 45.26% for one class under test. Writing new PUTs resulted in a total increase of 17.41% on an average. We also found that testing with PUTs detected 7 new defects that were not detected by the existing conventional unit tests.

1. INTRODUCTION

Unit testing is a key phase of the software development life cycle that helps detect defects at an early stage. It is essential to write high-covering tests during the unit testing phase to ensure high quality of the software under development. Conventional unit tests do not accept any parameters and include two major parts: test inputs and test oracles. In general, test inputs include the method calls that modify the object state and concrete argument values of these methods. Test oracles include the verification of method returns, some of which help verify the modified state. Often these

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conventional unit tests are written manually. However, writing all these conventional unit tests manually to achieve high coverage is often a prohibitively expensive task.

Parameterized Unit Tests (PUTs) [20] are a new advancement in unit testing. PUTs accept parameters unlike conventional unit tests, which do not accept any parameters. In particular, PUTs represent expected behavior or specifications of the code under test with symbolic values. Pex [19, 7] is a Microsoft test generation tool that accepts these PUTs and generates a set of conventional unit tests that try to achieve a high block coverage of the code under test. Pex uses an approach called dynamic symbolic execution [13, 11, 18] for generating conventional unit tests from PUTs. Initially, Pex explores the code under test with random values and collects constraints along the path executed in the code under test. Consequently. Pex explores alternate paths in the code under test by systematically flipping the captured constraints and generates concrete values (using a constraint solver) that can cover the alternate paths. For each set of concrete values that explores a new path, Pex generates a conventional unit test. Therefore, each PUT is instantiated a number of times to generate a set of conventional unit tests to achieve a high block coverage of the code under test [21]. In our empirical study, we use Pex to investigate the benefits of generalized tests (PUTs) in unit testing. The reason for using Pex is its popularity indicated by the recent public releases of Pex which recorded thousands of download counts [7]. Pex was previously used internally at Microsoft to test core components of the .NET architecture and has found serious bugs [19].

Although PUTs are more effective and can often achieve higher block coverage compared to conventional unit tests, it is a nontrivial task to write PUTs. Writing PUTs requires more effort compared to writing conventional unit tests. To address this challenge, we use suggested PUT patterns [10] and use various supporting techniques to ease the process of writing PUTs. To show the utility of PUTs over conventional unit tests in our empirical study and to ease the process of writing PUTs, we carry out a systematic procedure including two phases (1) test generalization and (2) writing additional PUTs. Test generalization is defined as the process of transforming a conventional unit test into a PUT [21]. In the test generalization phase, we take existing conventional unit tests and write PUTs to test the same or generalized behaviors as the ones tested by the conventional unit test. To transform a conventional unit test to a PUT, we first identify elements in the conventional unit test (such as arguments or the receiver object of the method under test) and promote these elements as test-method arguments. We next transform the assertions provided by the conventional unit

¹We refer to unit test generalization as test generalization throughout the paper.

test into PexAssert² assertions. We feed such transformed PUTs to Pex and record the coverage information reported by Pex. In the phase of writing new PUTs, we identify the code portions that are not covered by the transformed PUTs and write new PUTs to cover these code portions in the second phase. We define new scenarios³ and use techniques such as input-space partitioning [8] in order to write new PUTs for covering the code portions that are not covered previously.

There are two major challenges in test generalization with respect to test input values: (1) generating legal input values and (2) generating non-primitive object states. To assist a test generation tool to generate legal input values, a sufficient number of assumptions need to be specified. For example, a method under test accepting an integer can have a precondition that the argument should be greater than 1000. In our test generalization with PUTs, we add such assumptions (required by the precondition of the method under test) using PexAssume⁴ assumptions supported by Pex in a PUT. To address the issue of non-primitive object states, we write factory methods when writing PUTs. For example, if the method under test requires a non-primitive argument, Pex often cannot effectively generate a method-call sequence that can produce desirable object states for the non-primitive argument. These desirable object states are the states that help explore paths in the method under test by covering true or false branches in the method. To address this issue, we use the factory method feature provided by Pex to assist in generating desirable object states for non-primitive PUT arguments. We also use suggested test patterns [10] when writing PUTs.

In our study, we used the NUnit framework [1], which is a widely used open source unit-testing framework for C#, a counterpart of JUnit [4] for Java. We found that the block coverage achieved by transforming the conventional unit tests to PUTs has increased by 9.68% (on average) with a maximum increase of 45.26% for one class under test. We also observed a further increase of 7.74% block coverage with the additional new PUTs. Therefore, the total increase in the block coverage achieved by PUTs over conventional unit tests is 17.41%. We also found that test generalization helped cover 80 new blocks and detected 7 new defects.

This paper makes the following contributions:

- A first-of-a-kind empirical study that investigates the benefits of PUTs over conventional unit tests. In our empirical study, we show that testing with PUTs increased block coverage by 17.41% over conventional unit tests and detects 7 new defects. We also show evidence that test generalization reduces the amount of test code, thereby, helping in better test-code maintenance. The detailed results of our empirical study are available at http://ase.csc.ncsu.edu/projects/putstudy/.
- A systematic procedure of retrofitting conventional unit tests for parameterized unit testing to reduce effort in writing PUTs. Although writing PUTs requires more effort compared to writing conventional unit tests, our systematic procedure helps reduce the additional effort.
- Empirical results that show the benefits of three supporting techniques in writing effective PUTs.

Attribute	Value
#Total Files	560
#Files in NUnit.Util	72
Total LOC	53K
NUnit.Util LOC	7.2K
# Test Files of NUnit.Util	32

Table 1: Code metrics of project used in the study (NUnit)

- Empirical results that show the utility of test patterns in test generalization and writing new PUTs. We also propose two new test patterns that are useful to enable easy writing of the PUTs as observed in our testing of NUnit.
- Observations of two types of conventional unit tests that are not amenable to test generalization.

The rest of the paper is structured as follows. Section 2 describes the characteristics of the NUnit framework. Section 3 describes the test methodology followed in our study. Section 4 presents an example from the NUnit framework and explains the procedure for test generalization. Section 5 presents the categories of test patterns used in writing the PUTs. Section 6 describes how the supporting techniques are used in our test generalization. Section 7 presents the benefits of test generalization found in our study. Section 8 describes the categories of conventional unit tests that are not amenable for test generalization. Section 9 discusses the limitations of PUTs. Section 10 discusses the related work. Section 11 discusses the threats to validity. Finally, Section 12 concludes.

2. OPEN SOURCE PROJECT UNDER TEST

NUnit is a widely used open source unit-testing framework for all .NET languages, a counterpart of JUnit for Java [1]. NUnit is written in C# and uses attribute-based programming model [3] through a variety of attributes such as [TestFixture] and [Test]. The rationale behind choosing NUnit for our test generalization is the large number of manually written unit tests available with the project. These unit tests also provide information about the runtime behavior of the system. The source code of the entire project includes 560 files and about 53 KLOC. The test code includes 264 source files with 25 KLOC. This significant amount of test code makes this project a desirable subject for our empirical study. For the purpose of the study, we chose the Util package (nunit.util.dll), which is one of the core components of the framework.

The Util package includes 7.2 KLOC with 72 files and 326 methods. The numbers of test files, test methods, and test LOC account to 32, 335, and 3.4 KLOC, respectively. The reason for choosing the Util namespace⁵ for the study is two-fold (1) its significance in probably being one of the first modules being developed for the framework (2) its being an independent module reduces the amount of work required to facilitate unit testing by reducing dependent modules of the module used in the study.

3. TESTING METHODOLOGY

Our methodology of writing PUTs in our study can be classified into two major phases: (1) test generalization and (2) writing additional PUTs to achieve high coverage. The first phase involves transforming existing conventional unit tests into PUTs,

 $^{^2}$ PexAssert is a static helper class provided by Pex to evaluate assertions [2]

³Unlike the traditional definition of test scenario in system testing, we use the term scenario to refer to a set of conditions required for testing code under test.

⁴PexAssume is a static helper class provided by Pex to filter inputs [2]

⁵A namespace in C# is equivalent to a package in Java.

```
//st is of type MemorySettingsStorage and
//instantiated in the init() method of the test class
01:public void SaveAndLoadSettings() {
02: Assert.IsNull(st.GetSetting("X"));
03: Assert.IsNull(st.GetSetting("NAME"));
04: st.SaveSetting("X", 5);
05: st.SaveSetting("NAME", "Charlie");
06: Assert.AreEqual(5, st.GetSetting("X"));
07: Assert.AreEqual("Charlie", st.GetSetting("NAME"));
08:}
```

Figure 1: A conventional unit test from the Util project of the NUnit framework.

where PUTs attempt to verify the same or more generalized behavior as the ones tested by conventional unit tests. The second phase involves discovering the code portions that were not covered by the PUTs in Phase 1 and write new PUTs to achieve high coverage. We next explain the systematic procedure used in Phases 1 and 2.

Phase 1. Test Generalization. In Phase 1, for each conventional unit test, we identify possible variables that can be considered as parameters and replace constant values or variables with parameters. We next identify a test pattern to which the conventional unit test belongs. Identification of a test pattern for the PUT in advance helps to assist in writing PUTs effectively. We also define new patterns, if the conventional unit test does not fall into any of the existing test patterns. We next add necessary assumptions to make sure that the PUT tests the same or generalized expected behavior of the conventional unit test. We consequently add assertions to validate the actual behaviour of the code under test against the expected behavior. In our test generalization phase, we use factory methods supported by Pex to assist Pex in generating test inputs. When there are any non-primitive parameters, Pex may not be able to generate method-call sequences to construct desirable object states for those non-primitive parameters. These desirable object states are the states that help explore paths in the method under test by covering true or false branches in the methods. We assist Pex by providing factory methods to achieve desirable object states for non-primitive parameters.

Phase 2. Writing New PUTs. In Phase 2, our objective is to modify existing PUTs or write new PUTs to achieve higher code coverage of the classes under test. In particular, we observe the detailed coverage reports generated by Pex to identify un-covered code portions. We identify techniques to extend existing PUTs to cover those un-covered portions. In our study, we used two techniques to write additional PUTs: mock objects and input-space partitioning. We explain these supporting techniques in detail in Section 6.

4. EXAMPLE

We next explain the systematic procedure of retrofitting for PUTs to carry out unit testing in our empirical study with an example. We use the NUnit test case SaveAndLoadSettings shown in Figure 1 as an illustrative example for explaining our procedure. The objective of the unit test in our example is to verify the behavior of the class MemorySettingsStorage, which is primarily used for storage and retrieval of global values. To generalize this unit test, we first identify the concrete values used in the test case. For example, the unit test includes a concrete string value "x". We replace these concrete values with symbolic values by making them as arguments. The advantage of replacing these concrete values with symbolic values is that Pex can generate concrete values based on the constraints encountered in different paths of the method under test (MUT). Consequently, a single PUT can achieve the same test

```
00:[PexMethod]
01:public void SaveAndLoadSettingsTest1(
02: MemorySettingsStorage st, String sn, Object sv) {
03:    //Define Pex Assumptions
04:    st.SaveSetting(sn, sv);
05:    PexAssert.AreEqual(sv, st.GetSetting(sn));
06:}
```

Figure 2: A sample PUT for the conventional unit test shown in Figure 1.

```
00:[PexFactoryMethod(typeof(MSS))]
     //MSS: MemorySettingsStorage (class)
     //PAUT: PexAssumeUnderTest (Pex attribute)
01:public static MSS Create([PAUT]String[]
        sn, [PexAssumeNotNull]Object[] sv)
      PexAssume.IsTrue(sn.Length == sv.Length);
03:
04:
      PexAssume.IsTrue(sn.Length > 0);
05:
      MSS mss = new MSS();
      for (int count = 0; count < sn.Length; count++) {
07:
        PexAssume.IsTrue(sv[count] is string ||
08:
          sv[count] is int || sv[count] is bool ||
09:
          sv[count] is Enum);
10:
        mss.SaveSetting(sn[count], sv[count]);
12:
      return mss;
```

Figure 3: A factory method to assist Pex.

effectiveness (of high block coverage of the MUT) as multiple conventional unit tests with different concrete values testing the same method. In addition to generalizing the concrete values, we also generalize receiver objects.

We then analyze the conventional unit test to identify a PUT pattern [2] that the test belongs to. Identifying a pattern can help in easy generalization of the conventional unit test. For example, in the example conventional unit test, a setting is stored in the storage using the SaveSetting method and is verified with the GetSetting method. Such a conventional unit test belongs to the round-trip pattern suggested in the Pex documentation [2, 10]. If the conventional unit test does not fall into any of the pre-defined patterns, we define new patterns as shown in Section 5.2.

Figure 2 shows the skeleton of the PUT after generalizing concrete values and the receiver object. Our PUT accepts three parameters: the instance of MemorySettingsStorage, name of the setting, and its value. The SaveSetting method can be used to save either an integer value or a string value (the method accepts both types for its arguments). Therefore, the test requires two method calls shown in Statements 4 and 5 (Figure 1) to test that the method under test works as expected for both the input value types. However, we need only one method call of SaveSetting in the PUT because we accept the value type as Object, which can cover both integer and String. Indeed, the SaveSetting method also accepts bool and enum types. The generalization can automatically handle these additional types too, serving as a primary advantage of PUT as it helps reduce the test code significantly without reducing the behavior tested by the conventional unit test. We transform the assertions in the conventional unit test into PexAssert assertions to assert the same behavior. If the existing set of assertions are not sufficient, we add additional assertions to the PUT. Figure 4 shows the transformed PUT that replaces the conventional unit test.

For test generation, Pex can effectively handle primitive-type parameters such as String or integer. However, like any other test generation tool, Pex faces challenges in generating values for non-

Figure 4: Complete PUT for the conventional unit test.

primitive arguments such as st in our PUT. These non-primitive arguments often require desirable states to verify different behaviors. For example, an intention in our conventional unit test to have two SaveSetting method calls is to verify adding a new setting when there is already an existing setting in the storage. For example, consider that there is a defect in the implementation of SaveSetting that can be exposed only when there are five elements in the storage, then the desirable state for such a non-primitive argument is to have five elements already present in the storage. Therefore, to test the method under test in various scenarios, generalizing the receiver object helps in this case. However, the primary challenge in constructing desirable states for non-primitive arguments is to construct a sequence of method calls that create and mutate objects. We use a factory method supported by Pex to assist Pex in generating effective method-call sequences that can help achieve desirable object states. Figure 3 shows our factory method to assist Pex in generating effective method-call sequences. Our factory method accepts two arrays of setting names and values, and adds those entries to the storage. This factory method helps Pex to generate method-call sequences that can create desirable object states. For example, Pex can generate five names and five values as arguments to our factory method for creating a desirable object state with five elements in the storage.

Another important aspect of writing generalized PUTs is to define assumptions. For example, without any assumptions provided, Pex by default generates null values for the PUT arguments. To address the issue, we annotate a PUT method argument with a tag PexAssumeUnderTest⁶, which describes that the argument should not be null. We add further assumptions based on the behavior verified by the unit test. For example, the conventional unit test requires a assumption that the setting to be added should not already exist in the storage. We use PexAssume to add these additional assumption to the PUT such as Statement 4 (in Figure 4) in SaveAndLoadSettingsPUT1.

In a few cases, we identify that direct generalization of conventional unit tests might not achieve 100% block coverage. There could be several reasons such as the portions of the code are not covered by the behavior tested by conventional unit tests. In those cases, we identify the un-covered portions of the code and write new PUTs or modify the transformed PUTs to cover these code portions. Consider a sample code example shown in Figure 5. We highlight the un-covered portion of the code in **bold**. The reason for the un-covered code portion in this code example is that the code portion requires a delegate handler to be defined in the class. A delegate handler can be treated as a pointer to a function. These delegates can be used to encapsulate a method with a specific signature and return type. To achieve 100% coverage of the RemoveSetting method in the preceding code example, we created a trivial delegate handler and set the value to Changed. We present details on the us-

```
00:[PexMethod]
//MSS: MemorySettingsStorage (class)
//PAUT: PexAssumeUnderTest (Pex attribute)
//Changed is of type Delegate
01:public void RemoveSetting(MSS st,
        [PAUT]string settingName) {
02: st.RemoveSetting( settingName );
03: if (Changed != null)
04: Changed(this, new SettingSeventArgs(settingName));
05:}
```

Figure 5: A code sample with an un-covered portion (shown in bold) with PUTs written by transforming conventional unit tests

age of supporting techniques in our study in Section 6.

5. CATEGORIZATION OF PUTS

We used suggested test patterns [10] to generalize conventional unit tests. These test patterns can help developers in writing effective PUTs. Although PUTs alleviate the problem of writing different conventional unit tests to test the code under test with different input values, writing test oracles in PUTs could be a complex task. The developers are expected to specify sufficient assertions for testing various expected behaviors of executing the code under test. To deal with this complexity, developers can use test patterns to answer important questions of "what" scenarios of the code under test need to be tested and "how" they can be asserted. In our study, we found that a few test patterns are predominantly applicable while others are helpful in a few specific cases. In addition, we found that each PUT can be categorized into more than one test pattern. In all, the patterns supported by Pex help write a wide range of PUTs for achieving high code coverage and reduce PUT writing effort for both test-driven development and testing after the application is developed. We found that it was not easy to write PUTs for a few scenarios of the code under test using these patterns. To address this issue, we proposed two new patterns that can help in writing PUTs. We first explain the categories of the existing patterns as used and not used based on whether there are PUTs that belong to the test pattern and later present our proposed new patterns.

5.1 Existing Test Patterns

We next present the classification of our PUTs into existing test patterns. Table 2 shows the 15 existing test patterns [10]. Column 1 provides the classification of each pattern in our study. We put the existing test patterns into two categories based on their usage in this study: used and not used. Column 2 provides the pattern identifier. We use these identifiers to refer to the patterns in this section. Column 3 provides the pattern name and Column 4 gives the attributes⁷ and methods supported by Pex for that pattern.

- 1. **Used:** We put a pattern under this category if that pattern was used at least once when writing PUTs. We used 9 Patterns (from Pattern 2.1 to 2.9) for writing PUTs. Figure 6 shows the distribution of pattern usage across all the 70 PUTs. The x axis of the chart shows the used patterns and the y axis shows the number of PUTs that belong to each pattern. For each test pattern, we show the number of PUTs in our study.
- 2. **Not used**: We put a pattern under this category if that pattern was not used in test generalization. We found that 5 out of

⁶PexAssumeUnderTest is a custom attribute provided by Pex.

⁷Pex provides a set of custom attributes to tag test class, test methods or parameters.

Table 2: Existing Test Patterns Supported By Pex.

Category	Pattern#	Pattern Name	Pex Supported Attributes or Methods
Used	2.1	Arrange, Act, Assert	
	2.2	Assume, Arrange, Act, Assert	
	2.3	Constructor Test	
	2.4	Roundtrip	
	2.5	Sanitized Roundtrip	
	2.6	State Relation	
	2.7	Same Observable Behavior	
	2.8	Commutative Diagram	
		Cases	PexAssert.Case()
Not used		Allowed Exception	[PexAllowedException]
	2.11	PexAllowedException	[PexExpectedGoals]
	2.12	Parameterized Stub	[PexAssumeUnderTest]
		Manual Output Review	PexStore.Value()
	2.14	Regression Tests	PexStore.ValueForValidation()
	2.15	Differential Regression Test Suite	

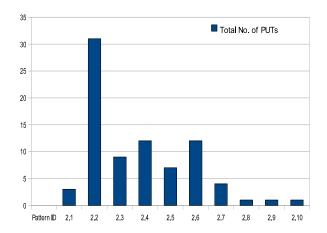


Figure 6: Distribution of test patterns for 70 PUTs. The distribution shows that in our testing, only patterns from 2.1 to 2.9 were used.

the 15 patterns were not used in writing PUTs in our study (from 2.10 to 2.15). The possible reason for not using these patterns in our test generalization is the lack of purposes for these patterns in our current study. For example, Patterns 2.14 and 2.15 are applied in the context of regression testing. However, in our current study, our focus is not on regression testing.

5.2 New Patterns

We next describe two new patterns that can be supported, and that we found useful during our test generalization for the NUnit framework.

5.2.1 Random Selection of Cached Values

Purpose: To reuse generated values by maintaining a pool and randomly picking from those values.

Motivation: We next show the motivation for such a pattern with an illustrative example. In the test generalization phase, we needed to generalize an existing unit test that requires to verify whether adding a RegisterKey to an NUnitRegistry and then clearing the NUnitRegistry work as expected. The NUnitRegistry class holds RegisterKeys as a tree structure with a main key and an unrestricted number of subkeys in a tree-structured form. Manually adding values and creating a tree structure to check both horizontal and vertical cases at the same time was possible in the existing conventional unit test. However, when the test was parameterized, we were able to write a PUT to add all RegisterKeys either to one main key or add each RegisterKey to the last added key. Due to such restriction, the unit tests generated by this PUT resulted in a reduced block coverage when compared to the conventional unit test, although there are a number of tests generated for the PUT. The rationale is that the tests generated for the PUT are expected to generate a tree structure of keys to achieve high code coverage. As the existing test patterns do not meet our current requirement, we proposed a new pattern where we maintain a pool of existing RegisterKeys and randomly select a key from this pool to add the newly created key as a subkey. Our new pattern can help construct several forms of tree structures automatically. Although we explain our motivation using the RegisterKeys test example, our pattern is general and can be applied to tests that require to reuse previously generated values.

Proposed Pattern: When the input is of the type collection or an array and the values need to be reused by the code under test, the values generated by Pex can be added to a pool using our proposed method PexStore.Pool(<name>,<value>). PexStore.Pool() method is expected to cache values (represented by value) for a particular variable (represented by <name>). Later, a value can be picked randomly from these cached values using another proposed method PexStore.Pick(<name>). PexStore.Pick() method is expected to pick a value randomly from the cached store of the variable <name>. Figure 7 shows an application of our proposed pattern.

5.2.2 Unique-value generation

Purpose: To generate unique values when a parameter is a collection of values instead of a single value.

Motivation: In our test generalization phase, for four PUTs, we were unable to achieve the same test effectiveness (i.e., block coverage) by the generated conventional unit tests as that of the existing unit tests. We found the reason to be a required object state: the required object state for the four PUTs to achieve high block cover-

```
00: [PexMethod()]
01:public void TestClearRoutinesPUT([PAUT]String[] key) {
02: PexAssume.IsTrue(key.Length > 1);
    for (int j = 0; j < \text{key.Length}; j++) {
03:
        PexAssume.IsNotNull(key[j]);
04:
05:
    NUnitRegistry.TestMode = true;
06:
07:
    using (RegistryKey mainKey =
        NUnitRegistry.CurrentUser) {
08:
09:
        //enabling appending values to a list
10:
        PexStore.Pool("keys",mainKey);
        for (int j = 0; j < key.Length; <math>j++) {
11:
12:
            RegisterKey parentKey =
13:
               PexStore.Pick("keys") as RegisterKey;
14:
            RegistryKey subKey =
               mainKey.CreateSubKey(key[k - 1]);
15:
16:
            PexStore.Pool("keys", subKey);
17:
18:
        NUnitRegistry.ClearTestKeys();
19:
        PexAssert.IsTrue(mainKey.SubKeyCount == 0);
20:
21:}
```

Figure 7: A new test pattern that helps reuse previously generated values using a cache. The PexStore.Pool() method can be used to append the value mainKey to the cache of variable keys and PexStore.Pick() method can be used to pick a random value from the cache of the variable keys.

age can be created only when a unique set of elements is available. For such PUTs, we had to write additional code to ensure that Pex generates only unique values. For example, in the following PUT, we include a for loop with a PexAssume (as shown below) on each element of the collection to make sure that Pex generates a set of unique values.

```
//PAUT: PexAssumeUnderTest
01:public void SubstorageSettingsPUT1(
     [PAUT]String subName, [PAUT]String[] name)
02:{
03:
        PexAssume.IsNotNull(value);
04:
        PexAssume.IsTrue(name.Length == value.Length);
       /*assist Pex to generate unique values by
 **iterating over each array element and adding
 **the assumption it is not equal to any other
 **arrav element*/
        for (int i = 0; i < name.Length; i++) {
05:
           for (int j = 0; j < name.Length; j++) {
06:
07:
             PexAssume.IsFalse(name[i].Equals(name[j]));
08:
        }
09:
10:
```

Proposed Pattern: When the input is of the type collection or an array, we propose a new pattern using a custom attribute called PexGenerateUnique that can inform Pex to generate unique values. For example, applying the proposed pattern to the preceding example can result in the following easy-to-write code. We also assume that PexGenerateUnique subsumes the properties of PexAssumeUnderTest.

For the four PUTs that required generation of a set of unique values, we could apply the proposed pattern as shown in the preceding code example and achieve the test effectiveness.

6. SUPPORTING TECHNIQUES USED IN WRITING PUTS

We next detail on the supporting techniques used in writing PUTs. In writing PUTs, we use the factory method and mock object supporting techniques. We used an additional technique of input-space partitioning to achieve higher code coverage in our Phase 2. We next describe these techniques used in our study.

6.1 Factory Methods

A commonly used technique in test generalization to assist effective test generation is writing factory methods. In our study of writing PUTs and executing them with Pex, we observed that one of the primary reasons for not achieving high block coverage is due to lack of method-call sequences for achieving desirable object states. Although Pex includes a heuristic demand-driven strategy for generating method-call sequences, we found that Pex's strategy can generate method-call sequences effectively in certain limited scenarios where the constructors either accept primitive arguments or explicitly state the actual type of the argument. To address this issue, we use the factory-method feature provided by Pex. These factory methods allow developers to write method-call sequences that can help Pex in achieving desirable object states. Figure 3 shows an example factory method that we used in our study. Our factory method accepts two arrays of setting names and values, and adds those entries to the storage. This factory method helps generate different object states for MemorySettingStorage. For example, using our factory method, Pex can generate an object of MemorySettingStorage with five elements in the storage.

In our test generalization, we constructed five factory methods for assisting Pex in generating desirable object states. We observed that these factory methods are quite helpful in achieving high block coverage.

6.2 Mock Objects

Mock objects help test features in isolation by replacing functionalities with mock objects. In Phase 2, we added 21 new PUTs and found two methods where we needed to use mock objects. Two existing test methods SaveEmptyConfigs and SaveNormalProject involved testing saving of an NUnitProject. When a new project is created (i.e., an instance of NUnitProject is created), an xml file is saved in the project directory for saving the project configurations. These test methods add configurations on this project file and assert if the file is saved in the right format and contains the added configuration information. SaveEmptyConfigs test method tests an empty project, with default debug and release configurations and the latter tests a project saved with multiple configurations. In Phase 1, we were not able to generalize these two tests as they needed interaction with an xml file from a specific location (the specific location refers to the directory location where the project is saved; when a project is saved, a new directory is created as the project directory). The xml file was expected to already exist physically for the test to execute so that the code under test can access the file and add configuration information to the file, and test if the project is correctly saved with the configurations. The existing tests save the projects and assert by reading the xml file using a stream reader and check against an expected string (which is constructed based on the configurations that are added). Generalization of these conventional unit tests is not straightforward as every conventional unit test generated by Pex requires a physical file in the expected file location. The code under test has high dependency on external factors, i.e., file reading and writing from a specific location.

We handled this situation of not being able to generalize due to

```
. . . . . . . . .
    public MockXmlTextWriter(string filename,
      Encoding encoding)
03:
        this.fileName = filename;
04:
05:
    public void WriteAttributeString
     (string attributeName, string value)
06:
        xmlString = xmlString + " " + attributeName
07:
            + "=" + "" + value + "";
08:
    public void WriteEndElement()
10:
        xmlString = xmlString + " />";
11:
12:
13:
    public void Close()
14:
        xmlString = xmlString.Replace("/> />", "/>" +
15:
            System.Environment.NewLine + "</
               + startString + ">");
16:
        CreatedProjects.currentProject = xmlString;
17: }
```

Figure 8: Sample code from the MockXmlTextWriter mock object. In the WriteAttributeString method, we append the argument string to a global string xmlString and this global string represents the content written to the xml file.

a high dependency on external environments by mocking the required xml file writer in Phase 2. When the code under test saves a project, it opens the xml file using XmlTextWriter. In order to avoid the complexity of generating a "real" file at a "real" file location (on creating a project) and writing into the file, we mocked the expected behaviour of XmlTextWriter as MockXmlFileWriter. This mock object, unlike the real object, appends the text to a string, preserving the output of the actual behaviour of XmlTextWriter. The mock object behavior results in the form of a string while the actual object would result in a file. Nevertheless, as suggested by the Pex tutorial [2], we did not mock every method of the actual class, but mocked only the methods used by the code under test. Figure 8 shows a code snippet from the mocked object.

By using the mock object technique, we were able to achieve generalization of both the existing tests of SaveEmptyConfigs and SaveNormalProject, resulting in a block coverage of 61.64% and 67.92%, respectively. Recall that we were not able to generalize the existing conventional unit tests when we did not use the mock object.

6.3 Input-Space Partitioning

Input-space partitioning [8] helps partition the input space into disjoint blocks, where the union of all the blocks should result in the complete input space. We explain how we performed the input-space partitioning in our study to achieve a higher coverage of the method SaveSetting of the class SettingsGroup. The SaveSetting method accepts an argument of type Object. The method accepts several types such as int, string, bool, and enum, and a different path of the code is covered for each type. Therefore, in order to achieve high code coverage, a PUT to test this method should be designed to generate conventional unit tests that take different types of the argument. For simplicity, we explain how we dealt with integers and strings only. We defined two partitions where the first partition includes integers and the second

partition includes strings. We wrote separate PUTs for covering these partitions. We repeated the same procedure for other input types. Consequently, Pex achieved high coverage as it was able to generate different input types and cover several program paths.

7. BENEFITS OF PUTS

We next present the benefits of retrofitting conventional unit tests for PUTs. In test generalization, we transformed 57 conventional unit tests of 10 test classes resulting in 49 PUTs. In order to achieve higher block coverage, we wrote 21 new PUTs for 6 of the 10 classes under study. Table 3 shows the results of writing PUTs. Column "Test Class" shows the names of the test class and Column "Test Methods" shows the statistics of existing conventional test methods, the test methods that are amenable to test generalization, the percentage of amenable conventional unit tests, the number of transformed PUTs, and the number of new PUTs that were written. The five sub-columns represented by "Conventional", "#Amenable", "%", "#PUT", and "#New PUT" give cumulative figures for all test methods in the corresponding class. "#PUT" shows the number of PUTs written in Phase 1 and "#New PUT" shows the number of new PUTs written in Phase 2 to achieve higher coverage. Section 8 provides details of the test methods that are not amenable to test generalization with illustrative examples. Column "% Coverage" shows the block coverage reported by Pex on executing tests. These two sub-columns show the average of all the test methods in each class. Column "Avg. New Blocks" shows the average number of new blocks covered by PUTs. As PUTs can verify more general behavior, we found that PUTs achieve a high block coverage and cover new blocks that are not covered by conventional unit tests. Column "#Defects" shows the number of defects that were detected by the PUTs and were not detected by the existing conventional unit tests. This column again shows a cumulative value of all the test methods in a test class.

The benefits discussed here primarily reflect the results of the test generalization phase as we compare the benefits of PUTs over the existing conventional unit tests. Section 9 discusses the amount of effort we took in writing the new PUTs in comparison to generalizing the existing conventional unit tests. We found three major benefits of retrofitting conventional unit tests for PUTs: higher code coverage, detection of new defects, and reduced test code. We observed that generalization increases code coverage and detected new defects that were not detected by the existing conventional unit tests. We also identified that a single PUT often helps to replace multiple conventional unit tests, thereby reducing the amount of test code (as shown by Column "#Convention" and Column "#PUT" in Table 3). Sections 7.1, 7.2, and 7.3 explain these details.

7.1 Coverage

As generalized tests often help cover more scenarios, we found that test generalization helped to have an effective increase in the block coverage as shown in Table 3. For example, test generalization of the RegistrySettingsStorage class shows an increase in the coverage of 45.24%. In addition, the tests generated for the PUTs in the test generalization achieved coverage of new blocks that are not covered by the existing conventional unit tests. In order to achieve more code coverage, we wrote 21 new PUTs for 6 classes and obtained an average increase of 35% code coverage (considering only those classes for which we wrote new PUTs).

7.2 Defects

After test generalization we found 7 new defects that were not detected by the existing conventional unit tests. We next explain

Table 3: Benefits of Retrofitting PUTs in Unit Test	Table 3:	3: Benefits of	of Retrofitting	PUTs in	Unit Testin	g
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Test Class	Test Methods				% Coverage			Avg. New	#Defects	
								Blocks		
	#Conventional	#Amenable	%	#PUT	#New PUT	Conventional	PUT	With		
								New PUT		
NUnitProjectSave.cs	3	1	33.33	1	2	35.71	40.98	57.91	10	0
NUnitRegistryTests.cs	5	5	100.00	4		58.97	72.80	100.00	0	1
TestAgentTests.cs	2	2	100.00	2	1	100.00	100.00	NA	0	0
RegistrySettings-										
StorageTests.cs	6	5	83.33	6	4	45.34	90.60	100.00	0	1
MemorySettings-										
StorageTests.cs	6	4	66.67	4		100.00	100.00	NA	2	0
PathUtilTests.cs	7	3	42.86	6		85.00	85.00	NA	0	3
RecentFilesTests.cs	22	21	95.45	5	6	59.94	76.08	84.17	14	0
ServerUtilityTests.cs	2	2	100.00	3	1	90.32	90.32	95.16	0	2
SettingsGroupTests.cs	5	5	100.00	6	7	66.94	90.95	94.13	2	0
ProcessRunnerTests.cs	0	NA	NA	NA	NA	NA	NA	NA	NA	NA

a defect detected by our test generalization. The NUnitRegistry class stores the RegistryKeys in a tree-structured hierarchy. For building the key hierarchy, a default key is taken as a main key and the given keys are added as sub-keys to the main key or to the other sub-keys. During testing, adding a key hierarchy and on checking for the count or on clearing the keys, we found abnormal behavior for two tests. The PUT was written to take three test inputs. For one of the tests generated by Pex, the test inputs were t, t, and t, and the other test case took the test inputs as 0, 0, and 0. For the first test case, when the three inputs were added to a main key (two as sub-keys and the other as a sub-key to one of the added subkeys), the count check for the keys passed, i.e., PexAssert (2, mainKey.SubKeyCount) passed. The same assertion failed for the second case (with test inputs "\0") with an assertion failure message, "expected 2, got 1". This defect shows that the failure was possibly due to missing check on invalid characters.

7.3 Test Code

Test generalization also helped reduce the test code as shown in Column "#PUT" of Table 3. Often, test generalization either helps reduce the amount of code in a single test method or helps combine multiple test methods into a single PUT. Figure 9 shows an example PUT of the pattern type *Cases* that combined five conventional unit tests. Each conventional unit test verifies one case in the corresponding PUT. In addition, the PUT achieved higher coverage compared to the five conventional unit tests as the MaxValue is now accepted as an argument and the concrete values are generated from the argument based on captured constraints.

8. CONVENTIONAL UNIT TESTS NOT AMENABLE TO TEST GENERALIZATION

In our study, we found that 12.75% of conventional unit tests are not amenable to test generalization. There are two common cases under which we found that conventional unit tests are not amenable to test generalization.

Default checks. We found that there are multiple conventional unit tests that verify default values, which are often static values. We suspect that the purpose of these tests could be to make sure that the developers do not change the static values accidentally. We suspect that these tests are not amenable for test generalization as values verified by these tests are constants.

```
[PexMethod]
public void CountOverOrAtMaxPUT1(int MaxValue) {
  recentFiles.MaxFiles = MaxValue;
  PexAssert
  .Case(MaxValue < MIN)
   .Implies(() => MIN == recentFiles.MaxFiles)
  .Case(MaxValue == MIN)
  .Implies(() => MaxValue == recentFiles.MaxFiles)
  .Case(MaxValue > MIN && MaxValue < MAX)
  .Implies(() => MaxValue == recentFiles.MaxFiles)
  .Case(MaxValue == MAX)
  .Implies(() => MaxValue == recentFiles.MaxFiles)
  .Case(MaxValue > MAX)
  .Implies(() => MaxValue == recentFiles.MaxFiles)
  .Case(MaxValue > MAX)
  .Implies(() => MAX == recentFiles.MaxFiles);
}
```

Figure 9: Single PUT constructed from five conventional tests.

Missing test oracles. We found that test generalization may cause the loss of test oracles in some cases. These cases can often occur when generalizing conventional unit tests that belong to the PUT patterns *Roundtrip* and *Commutative diagram*. We explain this issue using the illustrative example shown below.

The Canonicalize method in PathUtils accepts a string parameter and uses a complex procedure to transform the input string into a standard form. It is easy to identify the expected output for concrete strings such as "C:/folder1/.../folder2/.../file.tmp". However, when the conventional unit test is generalized with a parameter for the input string, it is challenging to identify the expected output. Developers need to implement a logic to assert the behavior of the method under test. The amount of effort required in such cases could be higher than the effort required to write the implementation of the actual method under test and therefore such conventional unit tests are not amenable to test generalization.

9. DISCUSSION

One of the major limitations of PUTs is that PUTs requires more effort from developers than writing conventional unit tests requires.

Although PUTs reduce the complexity of writing multiple conventional unit tests with various concrete test inputs, developers need additional expertise in writing such PUTs as PUTs are more generic compared to conventional unit tests. For example, writing PUTs requires developers to prescribe a test oracle that can deal with the generality of test inputs. We show that to reduce the complexity of writing PUTs, we adopted a methodology of writing PUTs in two phases. In Phase 1, we generalized the existing conventional unit tests to PUTs using suggested test patterns. In Phase 2, we used supporting techniques to write more PUTs to assist Pex in generating high-covering tests.

In our study, we observed that we took longer time to write PUTs in Phase 2 compared to the time we took in Phase 1. In Phase 1, we transformed the existing conventional unit tests to PUTs and the existing unit tests assist in writing PUTs as shown in Section 4 with an example. In Phase 2, we discovered the un-covered code portions and wrote more PUTs to assist Pex to generate tests to cover the un-covered code portions. Based on our experience, we believe that to enjoy the test effectiveness achieved by writing PUTs and to reduce the cost involved in writing PUTs, a practical solution could be retrofitting PUTs by transforming these conventional unit tests to PUTs. We expect that writing a single conventional unit test to test a method under test and then transforming it to a PUT can help developers in writing the PUTs. Such a single conventional unit test can act as an example unit test representing the intention of "what" needs to be tested. We expect that this methodology can both ease the process of writing PUT and still achieve high test effectiveness in unit testing.

As shown in our study, although the usage of the suggested test patterns and the supporting techniques can reduce effort of developers in writing PUTs and allow the test generation tool to generate high-covering tests, the complexity lies in being able to use them. In general, developers might consider it a tougher job to learn the supporting techniques and use them to write PUTs than writing multiple possible conventional unit tests. Nevertheless, our study shows that PUTs are more effective than conventional unit tests in detecting defects and also in achieving high code coverage. Therefore, we believe that the choice of writing PUTs is a trade-off between cost and benefit.

10. RELATED WORK

Pex [19, 20, 21] accepts PUTs and uses symbolic execution to generate test inputs. Similarly, other existing tools such as Parasoft Jtest [5] and CodeProAnalytiX [6] adopt the design-by-contract approach [15] and allow developers to specify method preconditions, postconditions, and class invariants for the unit under test and carry out symbolic execution or random testing to generate test inputs. More recently, Saff et al. [17] propose theory-based testing and generalize six Java applications to show that the proposed theory-based testing is more effective compared to traditional example-based testing. A theory is a partial specification of a program behavior and is a generic form of test methods where assertions should hold for all inputs that satisfy the assumptions specified in the test methods. A theory is similar to a PUT and Saff et al.'s approach uses these defined theories and applies the constraint solving mechanism based on path coverage to generate test inputs similar to Pex. The results of their study show a 25% of test methods amenable to test generalization compared to the 80.18% of test methods amenable to test generalization achieved in Phase 1 of our empirical study (calculated by the average of the Column "% amenable" in Table 3). Furthermore, their study does not provide the methodology of test generalization or show empirical evidence

of benefits of test generalization shown in our study.

There are existing approaches [16, 9, 12] that can automatically generate required method-call sequences that achieve different object states. However, in practice, each approach has its own limitations. For example, Pacheco et al.'s approach [16] generates method-call sequences randomly by incorporating feedback from already generated method-call sequences. However, such a random approach can still face challenges in generating desirable method-call sequences, as often there is little chance of generating required sequences at random. In our test generalization, we manually write factory methods to assist Pex in generating desirable object states for non-primitive data types.

In our previous work [14], we presented an empirical study to analyze the use of parameterized model in unit testing with PUTs. We showed that using a mock object can ease the process of unit testing and identified challenges faced in testing code when there are multiple APIs that need to be mocked. In our current study, we also use mock objects in our testing with PUTs. However, our previous study showed the benefits of mock objects in unit testing, while our current study shows the use of mock objects to help achieve test generalization. In another of our previous work around PUTs [22], we propose mutation analysis to help developers in identifying likely locations in PUTs that can be improved to make more general PUTs. In contrast, our current study suggests a methodology of retrofitting conventional unit tests for parameterized unit testing.

11. THREATS TO VALIDITY

The threats to external validity primarily include the degree to which the subject programs, faults, and conventional unit tests are representative of true practice. The number of classes used in our empirical study is relatively small, although the defects detected during our study are real defects. These threats could be reduced by conducting more studies with wider types of subjects in our future work. The threats to internal validity are due to manual process involved in writing PUTs. Our study results can be biased based on our experience and knowledge of the subject programs. These threats can be reduced by conducting more case studies with more subject programs and other human subjects. The results in our study can also vary based on other factors such as the effectiveness of our written PUTs and test generation capability of Pex. One threat to construct validity is that our study uses the block coverage reports generated by Pex to reflect the quality of tests.

12. CONCLUSION

We conducted an empirical study to investigate the utility of PUTs in unit testing. We first generalize the existing conventional tests by transforming conventional unit tests into PUTs, and then write new PUTs to increase code coverage. In Phase 1 of our study, we generalized 57 conventional unit tests in the NUnit framework to write 49 PUTs. We identified benefits of test generalization such as increase in the block coverage by 9.68% (on average) with a maximum increase of 45.26% for one class under test and detection of 7 new defects. In Phase 2 of our study, we wrote 21 new PUTs and achieved an increase in the block coverage of 17.41% over the conventional unit tests. We also identified types of conventional tests that are not amenable for test generalization and proposed new PUT patterns. We present details on the utility of suggested test patterns and supporting techniques that help in writing PUTs. We further discuss the limitations of retrofitting unit tests for PUTs in terms of effort required in writing these PUTs.

13. REFERENCES

- [1] NUnit, 2002. http://nunit.com/index.php.
- [2] Pex Documentation, 2006. http://research.microsoft.com/Pex/documentation.aspx.
- [3] Test-Driven Development in .NET, The NUnit Testing Framework, 2006. http://www.developerfusion.com/article/5240/testdriven-development-in-net/2/.
- [4] JUnit for Test Driven Development, 2008. http://www.junit.org/.
- [5] Parasoft Jtest, 2008. http://www.parasoft.com/ jsp/products/home.jsp?product=Jtest.
- [6] CodePro AnalytiX, 2009. http://www.eclipse-plugins.info/eclipse/plugin_details.jsp?id=943.
- [7] Pex Automated White box Testing for .NET, 2009. http://research.microsoft.com/Pex/.
- [8] P. Ammann and J. Offutt. *Introduction to Software Testing*. Cambridge University Press, 2008.
- [9] C. Csallner and Y. Smaragdakis. JCrasher: an automatic robustness tester for Java. *Software Practice and Experience*, 34(11), 2004.
- [10] J. de Halleux and N. Tillmann. Parameterized Test Patterns For Effective Testing with Pex, 2008. http://research.microsoft.com/en-us/ projects/pex/pexpatterns.pdf.
- [11] P. Godefroid, N. Klarlund, and K. Sen. DART: Directed Automated Random Testing. In *Proceedings of the 2005* ACM SIGPLAN Conference on Programming Language Design and Implementation (PLDI), pages 213–223, 2005.
- [12] S. Khurshid, C. S. Pasareanu, and W. Visser. Generalized symbolic execution for model checking and testing. In Proceedings of International Conference on Tools and Algorithms for the Construction and Analysis of Systems (TACAS), pages 553–568, 2003.
- [13] J. C. King. Symbolic Execution and Program Testing. Communications of the ACM, 19(7):385–394, 1976.

- [14] M. R. Marri, T. Xie, N. Tillmann, J. de Halleux, and W. Schulte. An empirical study of testing file-system-dependent software with mock objects. In Proceedings of the 2009 International Workshop on Automation of Software Test (AST), Business and Industry Case Studies (to appear), 2009.
- [15] B. Meyer. Object-Oriented Software Construction. Prentice Hall PTR, 2000.
- [16] C. Pacheco, S. K. Lahiri, M. D. Ernst, and T. Ball. Feedback-directed random test generation. In *Proceedings of the 2007 International Conference on Software Engineering (ICSE)*, pages 75–84, 2007.
- [17] D. Saff, M. Boshernitsan, and M. D. Ernst. Theories in Practice: Easy-to-write Specifications that Catch Bugs,. Technical report.
- [18] K. Sen, D. Marinov, and G. Agha. CUTE: a concolic Unit Testing Engine for C. In *Proceedings of the 2005 joint* meetings of the European Software Engineering Conference and the ACM SIGSOFT Symposium on the Foundations of Software Engineering (ESEC/FSE), pages 263–272, 2005.
- [19] N. Tillmann and J. de Halleux. Pex White Box Test Generation for .NET. In *Proceedings of the 2008 International Conference on Tests and Proofs (TAP)*, pages 134–153, 2008.
- [20] N. Tillmann and W. Schulte. Parameterized Unit Tests. In Proceedings of the 2005 joint meetings of the European Software Engineering Conference and the ACM SIGSOFT Symposium on the Foundations of Software Engineering (ESEC/FSE), pages 253–262, 2005.
- [21] N. Tillmann and W. Schulte. Unit Tests Reloaded: Parameterized Unit Testing with Symbolic Execution. *IEEE Software*, 23(4):38–47, 2006.
- [22] T. Xie, N. Tillmann, P. de Halleux, and W. Schulte. Mutation analysis of parameterized unit tests. In *Proceedings of the* 2009 International Workshop on Mutation Analysis (Mutation) (to appear), 2009.