Injection testing backed refactoring

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Abstract

Injection-based testing for refactoring is a pattern that minimizes the need for manual editing when altering the internal behaviour of a code base which does not have unit tests yet in place. Neither does it rely on a compilation or a linking process nor does it make assumptions on the structure of the code. Thus, it can be particularly useful for refactoring code that has been written in scripting languages, and specifically targets the research and engineering context. We describe the pattern and propose a set of functions for its application. The applicability of code injection for refactoring is highlighted via specific examples for deriving unit and integration tests. Finally, we comment on customizing the pattern and give practical advice for its implementation.

CCS Concepts

- Software and its engineering → Software development techniques; Software testing and debugging; Language features.

Keywords

refactoring, code instrumentation, legacy code, unit testing, scientific software, Matlab

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1 Introduction

We introduce the pattern Injection Testing to facilitate the refactoring of existing code, i.e., changing the internal structure of the code without affecting its external behaviour [1]. The injection testing pattern enlarges the set of code bases to which testing provisions can be added retrospectively.

The pattern allows inferring the internal behaviour of code by defining designated interception points as comments in the source code. No strong requirements are placed on the programming style of the code, which is especially important for refactoring code does not posses testing provisions, in the following called legacy code.

Techniques for refactoring are well established, see e.g., [2–9], and often rely on the existence of an accompanying test suite [10].

1.1 Existing refactoring solutions

Common interception points arise from the compiler toolchain and have been summarized under the term “seams”, “…a place where you can alter behaviour in your program without editing in that place.” [10, Chapter 4].

Pre-processing seams can be accessed via a text replacement engine that automatically changes the source code just before interpretation or compilation. Besides the need for maintaining such an engine in languages that do not feature a pre-processor, the usage of such seams necessitates a new compilation/interpretation cycle for every change that is applied to the code.

Link seams can be used in languages where the compiler/interpreter produces an intermediate representation of the code. Intercepting calls to other intermediate representations allows for the execution of arbitrary code. Similarly to the pre-processor approach, an additional linking step is required after each refactoring iteration, impeding the practical applicability of the approach.

Object seams are not compilation-related and use overloading and polymorphism for intercepting calls to class methods. This requires...
Consequently changing authors. This can lead to large code bases with insufficient testing provisions [11].

In fact, an informal survey of software papers published between 2015 and 2021 for Matlab, one of the most common programming languages in academia [12], indicates that only a small fraction of submissions includes tests, let alone automated tests using some specialized testing framework, as indicated in Table 1. This makes it hard to verify the correctness of the published results.

Table 1: Number of publications of free Matlab software whose source code is still available, in selected scientific journals between 2015 and 2021, contrasted with the number of submissions additionally including unit tested code and automatic unit testing provisions.

| Journal | Matlab submissions | unit tests | automatic unit tests |
|---------|--------------------|------------|---------------------|
| JOS      | 25                 | 9 (36%)    | 3 (12%)             |
| ACM TOMS | 19                 | 9 (47%)    | 2 (11%)             |
| JORS     | 22                 | 8 (36%)    | 2 (9%)              |
| J. Stat. Softw. | 18     | 8 (44%)    | 1 (6%)              |
| Softw. X | 78                 | 11 (14%)   | 5 (6%)              |
| Softw. Imp. | 13        | 2 (16%)    | 0 (0%)              |

1 JOS: Journal of Open Source Software, ACM TOMS: ACM Transactions on Mathematical Software, JORS: Journal of Open Research Software, J. Stat. Softw.: Journal of Statistical Software, Softw. X: Software X, Softw. Imp.: Software Impacts.

All of the above approaches pose requirements on either the programming language or the programming style in which the code has been written. Not every code meets those requirements.

2.1 Context

Given a function which needs to be refactored, in the following we refer to as system under test, that has no or insufficient testing provisions. Furthermore, this function is written in a style rendering it "un-testable".

2.2 Problem

Refactoring without having tests in place bears the risk of breaking the code. Thus, the challenge is to add tests to existing code while at the same time limiting the amount of behavioural changes to the code.

2.3 Forces

- In languages where there are no predetermined points for intercepting the behaviour of code, in the absence of sufficient testing provisions, one has to apply manual modifications to the code in order to make it testable. The manual editing of code with no sufficient testing provisions in place, however, bears the risk of unwittingly altering its behaviour. Therefore, the amount of manual edits to the code necessary for writing tests should be minimized and their potential impact should be mitigated.
- At the same time, it is desirable to be able inject at arbitrary positions in the code and to not be reliant on any particular programming style or other prerequisites for doing so.
- Also, tests should be loosely coupled to the system under test. In particular, unit tests should only use the SUT’s public interface and there should be no test code in the production code.
- Running tests should need minimal manual intervention, this means in particular they are preferably written in the same language, not need special commands for executing them, and be portable; and thus not rely on third party code.

2.4 Solution

Add tests by injecting test code at runtime, defining specific entry and exit points, and thus minimize the amount of code changes; optimally by using features directly incorporated in the used programming language, so that no external tools are necessary. We will call the corresponding interception method the injection seam.

To mitigate the risk of the injected code altering the behaviour of the code base in an unintentional way, we propose the following "I-triple-A" pattern:

\[ \text{Inject} \quad \text{arrange} - \text{Act} - \text{Assert}, \]

optionally supplemented by an Inject setup stage at the very beginning and a Tear down stage at the very end.

Finally, in Sections 2.6 and 4 we discuss some points regarding the implementation of our pattern, including the presentation of the Matlab/Octave unit test framework TTEST.

3.1 Code injection

Code injection and code instrumentation refers to changing the code at hand, support object orientation and dependency injection at least to some degree.

Code for numerical computations in engineering, scientific, and educational contexts are often written incrementally and by free-hand. This can lead to large code bases.

3.2 Problem

Running tests should need minimal manual intervention, this means in particular they are preferably written in the same language, not need special commands for executing them, and be portable; and thus not rely on third party code.

4.1 Overview

In Section 2, we present our injection testing pattern and propose a set of functions for its efficient implementation.

We show how the pattern can be applied to write unit tests in Section 3.1. The code injection pattern is further exemplified in Section 3.2, where we discuss how the injection testing pattern can be used to write integration tests.

1Unit tests test well-defined, self contained parts of code, whereas integration tests test the code together with some of its dependencies with respect to specified functional requirements.

2TTEST is the Matlab unit test framework.

3Octave is a free implementation of the Matlab language.

4We use the term code injection instead of code instrumentation, to stress the point that we do not just add logging or similar functionality, but arbitrary code to an existing code base.
The refactoring of legacy code becomes easier and less likely to break existing code.

### 2.5.1 Benefits
- For most programming languages, injection testing does not pose any requirements on the coding style.
- Thus, the pattern can be applicable in situations where other approaches fail, most prominently those described in Section 1.1. This is especially important when refactoring legacy code.
- Places where interception points can be placed using the injection seam depend on the programming language, but are in general quite arbitrary; usually at each code line. This enables the testing of parts of the code.
- Using the injection seam can be faster than the pre-processor or link seam. No additional compilation/linking cycles are needed, but also code sections not relevant to the test can simply be skipped.

### 2.5.2 Drawbacks
- Injection tests are more strongly coupled to the system under test than classical unit tests. In particular, variable names, line numbers, and/or labels may have to be spelled out explicitly in the injection test. Data that previously were private to functions (like local variables) become public to the injection test; In other terminology, the injection test becomes a friend of the system under test.
- Due to the strong coupling of the injection tests with the system under test, the injection tests themselves should be refactored after the SUT is refactored successfully.
- Without sufficient support from the programming language or the development environment, injection testing may not be possible or only partly possible, e.g. whenever a correspondence the set of key functions cannot be implemented fully.

### 2.5 Consequences
The refactoring of legacy code becomes easier and less likely to break existing code.

#### 2.5.2 Drawbacks
- Code injection may degrade the system under test’s performance, as (a) the code must be supervised by some means to allow code injection, and (b) the code can not be compiled with all possible optimizations.
- Minimal manual editing of the code is still required, usually solely by adding comments, and thus, the risk of unwittingly altering its behaviour is not fully mitigated.

### 2.6 Implementation example
We will make use of the following set of key functions:
- `gotoat`: Jumps to an arbitrary line of code
- `assignat`: Assigns values to variables
- `captureat`: Stores the current state
- `returnat`: Returns from the function
- `clearat`: Removes all injected code

The correspondence between the key functions and the stages of the injection testing pattern are depicted in Figure 2.

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*Figure 2: Injection testing pattern: After an optional setup, in the `Inject Setup` stage, the code to set the desired workspace is injected. After the execution of the system under test in the `Act` stage, the results are retrieved and checked in the `Assert` stage. The final tear down concludes the pattern. The used functions are described in Section 2.6.*
The only necessary modification to the system under test may be the addition of comments to the code, indicating where to instrument the code for the injection tests. Comments are unlikely to break or alter the existing code\(^6\), particularly when some special style (e.g. `<!TEST1!>` is used, such that changes can be easily tracked and the comment’s special role is clearly indicated. If comments were accidentally removed, the accompanying test suite would fail, thus indicating the problem.

Similar approaches are in use for tracing function calls in Matlab\(^15\), and for implementing Mutation testing in Java\(^16\).

Only some programming languages, e.g. Java, support code injection directly. When there is no language support, it is often possible to use the debugger for implementing code injection. The relevant features of the debugger are:

- Running a program step by step
- (Conditionally) stopping the program at a certain point
- Inspecting the current state
- Jumping to a certain point

Often, these features can be accessed from within the programming language, e.g. in interpreted languages like Matlab, Python, or R. Also often the debugger can be controlled using third party libraries, e.g. `libgdb` for `gdb` (discontinued 1993)\(^17\), `lldb` for the LLVM toolchain\(^18\), Windows Debugger (`WinDbg`) for Windows\(^19\).

Failing that, one can usually still write macros in some scripting language for the debugger, e.g. `GDBMl` for `gdb`\(^20\).

3 EXAMPLES

In this section we give two examples on using our injection testing pattern for deriving a unit test and an integration test. Apart from the functions proposed in Section 2.6, we use the following tools:

- A method to easily save snapshots of some state
- A unit test framework for automated unit tests
- An assertion framework to easily write assertions

3.1 Example 1: Unit test

We demonstrate the injection testing pattern on the dummy legacy function `foo` given in Listing 1\(^7\), using Matlab style code. The task is to refactor parts of the body of `foo` out into a sub function while touching the code as little as possible. We only present the relevant lines of code.

Write a succeeding unit test. The function `foo` is not suited for classical unit tests, since the lines of code to be tested are not accessible via standard means.

```
function foo( a1, a2 )
    % lots of code
    sum = 0
    for i = 1:a1 % loop over i = 1 to a1 (both included)
        sum = sum + i; end
    % lots of code
```

Listing 1: Function foo

In the first step we add the labels `<FOO:1>` and `<FOO:2>` which indicate where to instrument the code with injection tests, see Listing 2.

```
function foo( a1, a2 )
    % lots of code
    % <FOO:1> % comment for testing
    sum = 0
    for i = 1:a1
        sum = sum + i; end
    % <FOO:2> % comment for testing
    % lots of code
```

Listing 2: Function foo augmented

In this form, the code is ready for injection testing and we can write a unit test suite. An example unit test suite is given in Listing 3.

```
% inject setup
gotoat( 'foo', 'goto', '<FOO:1>' )
captureat( 'foo', 'at', '<FOO:2>' )
returnat( 'foo', 'at', '<FOO:2>' )

% test 15
% inject arrange
assignat( 'foo', 'at', '<FOO:1>', 'a1', 15 )
% act
foo()
% assert
X = captureat() % obtain values
EXPECT_EQ( X.sum, 120 ) % compare result

% test 0
% inject arrange
assignat( 'foo', 'at', '<FOO:1>', 'a1', 0 )
% act
foo()
% assert
X = captureat()
EXPECT_EQ( X.f002, 0 )

% tear down
clearat( 'foo' )
```

Listing 3: Unit test suite for foo

Listing 4.2: Function executed and stored for later retrieval. The function `captureat` injects code such that the value of the variable `sum` at the line with comment `<FOO:2>` is stored for later retrieval. The function `returnat` injects code so that the function `foo` returns to the caller site whenever the control flow reaches line `<FOO:2>`.

In the `% test 1` part we arrange the data to be injected into the function `foo`; at the line with comment `<FOO:1>` the variable `a1` will be assigned the value 15. Afterwards we execute the system under test by calling it and retrieve the stored data by `captureat`. In the `% assert` section we check whether the retrieved value of `sum` equals our expected value 120. The second test `% test 2` follows the same pattern.

In the `% tear down` part, after the injection tests have finished, we clean up using `clearat` which removes all injected code from the function `foo`.

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\(^6\) An example where comments could break code are old BASIC dialects which used hard coded line numbers to specify the control flow.

\(^7\) The `:` operator generates a regularly spaced vector of numbers, and is usually used to define the bounds of a for loop.
Refactor. Having our unit tests in place we can safely refactor the function foo, as given in Listing 4.

```matlab
function foo( a1, a2 )
% lots of code
% <FOO:1>
sum = sum0( a1 )
% <FOO:1>
% lots of code
end

function x = sum0( x )
x = x * ( x + 1 ) / 2
end
```

Listing 4: Refactored foo

When we are sure that the refactoring did not change the behaviour, one should refactor the code and the unit tests once more, see Listings 5 and 6.

```matlab
function foo( a1, a2 )
% lots of code
sum = sum0( a1 )
% lots of code
end

function x = sum0( x )
x = x * ( x + 1 ) / 2
end
```

Listing 5: Twice refactored foo

Listing 6: Refactored unit tests for foo

3.2 Example 2: Integration test

Another example for using the injection test pattern is the gradual refactoring of the function bar, whose functionality is not apparent to the programmer, see Listing 7.

Write a succeeding unit test.

```matlab
function bar( a )
% 1000 lines of code
% <BAR:0>
% another 1000 lines of code
% <BAR:1>
end
```

Listing 7: Function bar

Using injection testing, we store the full state of the program at various locations when run in its initial form, i.e. before refactoring. After refactoring we compare the saved state with the new state. If they coincide, we can assume that the behaviour of the system under test did not change. An exemplary unit test suite is given in Listing 8.

What happens in Listing 8. The only substantial difference to our unit test suite for foo is the use of the helper function `CACHE`. This is just a thin convenience wrapper for storing data to disk, used as follows: When a file with name equal to its first argument does not exist, it stores the value of the second argument to disk. Otherwise, it discards the second argument and loads the stored data from disk.

```matlab
% inject setup
captureat( 'bar', 'at', '<BAR:0>' )
captureat( 'bar', 'at', '<BAR:1>' )

% test 5
% arrange
a = 5
% act
X = captureat() % obtain values
% assert
EXPECT_EQ( CACHE('BAR0_5', X.BAR0), X.BAR0 )
EXPECT_EQ( CACHE('BAR1_5', X.BAR1), X.BAR1 )

% test 7
a = 7
% test continuos as above

% tear down
clearat( 'bar' )
```

Listing 8: Integration tests for bar

This time we inject code such that the whole workspace of the function bar is captured whenever the control flow reaches the lines with comments <BAR:0> and <BAR:1>. The call `X = captureat()` in the % act section then retrieves the stored data and stores it in X. The assert section now compares the two stored states with the snapshot taken from before refactoring.

4 IMPLEMENTATION IN MATLAB

We implemented the set of key functions proposed in Section 2.6 using Matlab’s debugger and conditional breakpoints. Matlab’s conditional breakpoints evaluate a string at run-time. If the result is `truthy`, the code run is stopped at that location; but when it is `falsy`, the code run continues normally. Using conditional breakpoints for code injection in Matlab has some restrictions:

1. Conditional breakpoints only accept valid Matlab commands, but not anonymous functions.
2. The injected code must return a `falsy` value in order to avoid the debugger stopping its execution.
3. If the injected code throws an error, it is caught automatically by Matlab and the program run stops, i.e. the debugger starts.
4. Injected code is always executed before the code at the injected line. Code cannot be injected between statements.

To execute anonymous functions, we store them in a persistent variable in some function, and generate a string which then executes that anonymous function. To ensure that the return value of the injected code is `false`, it gets wrapped in a function returning `false` and evaluated by `evalin( '"caller', ___ )`. Errors thrown by the injected code are caught inside the function which evaluates the string or anonymous function.

4.1 Example implementation of evalat

To illustrate the execution of anonymous functions in Matlab, and thus execution of arbitrary code, a minimum implementation of a

\[ A \text{ truthy value is a value which implicitly evaluates to true, for example in an if condition, e.g. true, 1 or an array with only non-zero values. Contrary, a falsy value implicitly evaluates to false, e.g. false, 0, or an array with at least one zero.} \]

\[ B \text{The function evalin( '"caller', cmd ) executes a command cmd in the callers workspace, and in particular has access to the callers workspace. Note that evalin cannot be used recursively.} \]
function evalat is given in Listing 9. The key functions captureat, assignat both can be derived from this one.

Note, to avoid parsing the inputs, the interface of evalat is different from the interface of the ...at functions in the listings above.

function ret = evalat( fun, lne, h );
    persistent cache;
    if( nargin==0 );
        ret = cache;
    return; end;
    cache = h;
    h = [ 'returnfalse( 'assign(''handle'',evalat()) ) || ...' returnfalse( handle() ); ' ];
dbstop( 'in',fun, 'at',num2str(lne), 'if',h );

function ret = returnfalse( varargin );
    ret = false;

function ret = assign( name, value );
    try; ret = evalin( 'caller', [name '']; );
    catch; ret = []; end;
    assignin( 'caller', name, value );

Listing 9: evalat

What happens in Listing 9. The function evalat accepts three arguments, Fun is the function where code shall be injected, lne is the line number where code shall be injected, h is an anonymous function to be executed at the specified position.

Upon calling with three arguments, the anonymous function h is stored in the persistent variable cache. A persistent variable retains its values between function calls. Then, the function dbstop adds in the function fun at the specified position lne a conditional breakpoint, which will execute the code listed in Listing 10. We put those lines of code in its own listing for better readability.

Listing 10: Code of conditional breakpoint in evalat

Now, when the function fun is called and the program flow reaches line lne the function evalat is called without arguments. Thus, evalat returns the value of the persistent variable cache, this is exactly the anonymous function h we want to execute. The anonymous function h is passed to the function assign, which creates a variable with name 'handle' in the workspace of the function fun. All of this code is wrapped inside a call to returnfalse, which ensures that always false is returned.

A conditional breakpoint only stops when the injected code returns true. Since the first part of the injected code returns false the second part is evaluated. Now the just assigned variable handle, which is our anonymous function h, is executed. The result of the anonymous function h is passed again to returnfalse which again ensures that false is returned. Thus, the debugger does not stop the program.

4.2 Example implementation of returnat

A more involved, but shorter example shows how to programmatically return early from a function. The idea is to provoke an error at a user defined position in a function and catch the exception. As already noted, simply throwing an error in some injected code would not work, since, whenever injected code throws, the debugger stops the program. Instead, we have to make sure that an error is thrown after the injected code was executed. This we achieve by some tough means; we clear the function’s workspace, see Listing 12.

Listing 12: returnat

What happens in Listing 12. The function returnat accepts two arguments, Fun is the function which shall be executed, lne is the line number at which we want to return. When called with two arguments, dbstop adds a conditional breakpoint in the function fun at the specified position lne, which executes a call to returnat without arguments. Afterwards the function fun is called. When program flow reaches the specified location lne, returnat is called without arguments. Thus, the workspace of Fun is cleared by executing evalin('caller', 'clear'). Finally false is returned, so that the debugger does not stop the program. The next time a variable is accessed in the function fun, an error is thrown, therefore the fun returns. This error is caught in the catch block in the function returnat. In our example implementation we display the caught error, but any other code is equally possible. A usage example is given in Listing 13.

Listing 13: Usage example of returnat

4.3 TTEST

Functionality for implementing the code injection pattern, specifically the key functions assignat, captureat, evalat, returnat, are contained in the unit test framework TTEST for Matlab and

The function nargin returns the number of passed arguments. Note that, calling a function without parentheses is equivalent to calling it without arguments, i.e. nargin is the same as nargin( ).

Persistent variables retain their value between calls to the function. They get automatically initialized with [].

If the function’s arguments are called varargin, then the variable varargin contains all passed arguments wrapped inside a cell array.

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If the function’s arguments are called varargin, then the variable varargin contains all passed arguments wrapped inside a cell array.

In the listing, the lines starting with "w" are supposed to be entered in the Matlab command window. The other lines present the expected result.

Anonymous function are declared by an "@" character, followed by the list of arguments in parentheses.

This example is tested with Matlab 2020a; with other versions it may not work.
Code injection backed refactoring

Octave. TTEST has been written specifically with testing code in a scientific context in mind. It supports the testing of scripts, local and sub functions, has utilities for caching results for integration tests, and adds support for injection testing and partly for design by contract [21].

TTEST is published under a permissive open source license and available at gitlab.com/tommsch/TTEST. The full documentation of TTEST, together with a comparison of Matlab unit test frameworks, can be found in [22].

The following projects use TTEST (list non exhaustive): • Auditory Modelling Toolbox (Ver. 1.1) [23] • toolboxes [24] • Large Time Frequency Analysis Toolbox (Ver. >2.4) [25]

5 CONCLUSION

We presented a pattern for the injection-based refactoring as a means for handling otherwise not testable code, along with a set of functions suitable for its implementation. We gave examples on its usage and practical advice for their implementation in scripting languages via making use of the debugger’s functionalities. We provide a free implementation of the pattern’s key functionality in our TTEST unit testing framework.

Further work comprises enhancements of the usability and customizability of the pattern by improving on the underlying functionality in the toolbox, e.g. by implementing a gotoat function, allowing for the direct execution of arbitrary sections in the code.

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