DLint: Dynamically Checking Bad Coding Practices in JavaScript

Liang Gong
Michael Pradel
Manu Sridharan
Koushik Sen

Electrical Engineering and Computer Sciences
University of California at Berkeley

http://www.eecs.berkeley.edu/Pubs/TechRpts/2015/EECS-2015-5.html

February 8, 2015
Acknowledgement

This research is supported in part by NSF Grants CCF-1423645 and CCF-1409872, by gifts from Mozilla and Samsung, a Sloan Foundation Fellowship, and by the German Federal Ministry of Education and Research (BMBF) within EC SPRIDE.
ABSTRACT

JavaScript is becoming one of the most popular languages, yet it is known for its suboptimal design. To effectively use JavaScript despite its design flaws, developers try to follow informal code quality rules that help avoiding correctness, maintainability, performance, and security problems. Lightweight static analyses, implemented in “lint-like” tools, are widely used to find violations of these rules, but are of limited use because of the language’s dynamic nature. This paper presents DLint, a dynamic analysis approach to check code quality rules in JavaScript. DLint consists of a generic framework and an extensible set of checkers that each address a particular rule. We formally describe and implement 28 checkers that address problems missed by state-of-the-art static approaches. Applying the approach in a comprehensive empirical study on over 200 popular web sites shows that static and dynamic checking complement each other. On average per web site, DLint detects 49 problems that are missed statically, including visible bugs on the web sites of IKEA, Hilton, eBay, and CNBC.

1. INTRODUCTION

JavaScript has become one of the most popular programming languages. It powers various popular web sites, applications on mobile platforms, such as Firefox OS, Tizen OS, iOS, and Android, as well as desktop platforms, such as Windows 8 and Chrome OS. Despite its great success, JavaScript is often considered a not “well-formed” language. Designed and implemented in ten days, JavaScript suffers from many unfortunate early design decisions that have prevailed with the thrive of the language to ensure backward compatibility. The suboptimal design of JavaScript causes various pitfalls that developers should avoid [17].

A popular approach to help developers avoid common pitfalls are guidelines on which language features, programming idioms, APIs, etc. to avoid, and on how to use them correctly. The developer community has learned such code quality rules over time, and documents them informally, e.g., in books [17, 28] and company-internal guidelines. Following these rules improves the overall software quality by reducing bugs, increasing performance, improving maintainability, and preventing security vulnerabilities. Since remembering and following code quality rules in JavaScript further burdens the use of an already complicated language, developers rely on automatic techniques that identify rule violations. The state-of-the-art approach for checking rules in JavaScript are lint-like checkers, such as JSLint [3], JSHint [2], ESLint [1], and Closure Linter [4]. These static checkers are widely accepted by developers and commonly used in industry.

Although static analysis is effective in finding particular kinds of problems, it is limited by the need to approximate possible runtime behavior. Most practical static checkers for JavaScript [3, 2, 1, 4] and other languages [29, 13] take a pragmatic view and favor a relatively low false positive rate over soundness. As a result, these checkers may easily miss violations of some rules and do not even attempt to check rules that require runtime information.

Figure 1 shows two examples that illustrate the limitations of static checking of code quality rules.

```javascript
// From Modernizr 2.6.2
for (i in props) { // props is an array
    prop = props[i];
    before = mStyle.style[prop];
    ...
}

(a) Violation of the rule to avoid using for-in loops on an array. Found on www.google.com/chrome.
```

```javascript
// From jQuery 2.1.0
globalEval: function(code) {
    var script, indirect = eval; // alias of eval function
    code = jQuery.trim( code );
    if (code) {
        if (code.indexOf("use strict") === 1) {
            ...
        } else {
            indirect(code); // indirect call of eval
        }
    }
}

(b) Violation of the rule to avoid eval and its variants. Found on www.repl.it.
```

Figure 1: Examples that illustrate the limitations of static checking of code quality rules.
Figure 2: Bug found by DLint on the Hilton and CNBC web sites.

violation of the rule to not iterate over an array with a for-in loop (Section 2.4.1 explains the rationale for this rule). Existing static checkers miss this violation because they cannot precisely determine whether props is an array. The code snippet is part of www.google.com/chrome, which includes it from the Modernizr library. The problem has been fixed in a newer version of the library because the code in Figure 1a misbehaves on Internet Explorer 7. The second example (Figure 1b) is a violation of the rule to avoid the notorious eval and other functions that dynamically evaluate code. The code creates an alias of eval, called indirect, and calls the eval function through this alias. We found this example on www.repl.it, which includes the code from the jQuery library. Static checkers report direct calls of eval but miss indirect calls because static call resolution is challenging in JavaScript. As noticed by others [45], aliasing eval is used on various web sites because it ensures that the code passed to eval is evaluated in the global context.

Despite the wide adoption of code quality rules and the limitations of static checking, there currently is no dynamic lint-like approach for JavaScript. This paper presents DLint, a dynamic analysis for finding violations of code quality rules in JavaScript programs. Our approach consists of a generic analysis framework and an extensible set of checkers built on top of the framework. We present 28 checkers that address common pitfalls related to inheritance, types, language usage, API usage, and uncommon values. We describe the checkers in a lightweight, declarative formalism, which yields, to the best of our knowledge, the first comprehensive description of dynamically checkable code quality rules for JavaScript. Some of the rules, e.g., Figure 1a, cannot be easily checked statically and are not addressed by existing static checkers. Other rules, e.g., Figure 1b, are addressed by existing static checkers, and DLint complements them through dynamic analysis.

Having DLint and the existing static checkers raises the following research questions, which compare the effectiveness of dynamic and static analyses:

- **RQ1:** How many violations of code quality rules are detected by DLint but missed by static checkers?
- **RQ2:** How many additional code quality problems are found by dynamic variants of static checkers?

In addition, we also address this question:

- **RQ3:** How does the number of violations of code quality rules relate to the popularity of a web site?

To answer these questions, we perform an empirical study on over 200 of the world’s most popular web sites. We apply DLint and the most widely adopted existing static checker, JSHint, to these sites and compare the problems they report with each other. In total, the study involves over 4 million lines of JavaScript code and 178 million covered runtime operations. The study shows that DLint identifies 53 rule violations per web site, on average, and that 49 of these warnings are missed by static checking (RQ1). Furthermore, we find that complementing existing static checkers with dynamic variants of these checkers reveals at least 10.1% additional problems that would be missed otherwise (RQ2). We conclude from these results that static and dynamic checking complement each other, and that pursuing the DLint approach is worthwhile.

Even though this work is not primarily about bug finding (the rules we consider address a more diverse spectrum of code quality problems), we stumbled across 19 clear bugs in popular web sites when inspecting a subset of DLint’s warnings. These bugs lead to incorrectly displayed web sites and are easily noticeable by users. For example, DLint detects “undefined” hotel rates on www.hilton.com and “NaN” values on www.cnbc.com (Figure 2). The approach also successfully identifies the motivating examples in Figure 1. All examples are missed by static checking.

We envision DLint to be used as an in-house testing technique that complements static checkers in ensuring code quality. That is, our purpose is not to replace static lint-like tools but to provide an automatic approach for identifying problems missed by these tools. Our empirical results show that dynamic and static checking can each identify a unique set of violations of common code quality rules.

In summary, this paper contributes the following:

- We present the first dynamic analysis to find violations of code quality rules in JavaScript.
- We gather and formally describe 28 JavaScript code quality rules that cannot be easily checked with static analysis.
- We present an extensive empirical study on over 200 popular web sites that systematically compares the effectiveness of static and dynamic analyses in finding code quality problems. The study shows that both approaches are complementary, and it quantifies their respective benefits.
- Our implementation of DLint can be easily extended with additional checkers, providing the basis for a practical tool that fills an unoccupied spot in the JavaScript tool landscape.

2. APPROACH

This section presents DLint, a dynamic analysis to detect violations of code quality rules. DLint consists of a generic framework and a set of checkers that build upon the framework. Each checker addresses a rule and searches for violations of it. We specify when such a violation occurs in a declarative way through predicates over runtime events (Section 2.1). DLint currently contains 28 checkers that address rules related to inheritance (Section 2.2), types and type errors (Section 2.3), misuse of the JavaScript language (Section 2.4), misuse of an API (Section 2.5), and uncommon values (Section 2.6). The presented checkers reveal rule violations in various popular web sites (Section 4).
### 2.1 Rules, Events, and Runtime Patterns

The goal of this work is to detect violations of commonly accepted rules that the developer community has learned over time.

**Definition 1 (Code quality rule)**
A code quality rule is an informal description of a pattern of code or execution behavior that should be avoided or that should be used in a particular way. Following a code quality rule contributes to, e.g., increased correctness, maintainability, code readability, or performance.

We have studied 31 rules checked by JSLint [3], more than 150 rules checked by JSHint [2], and around 70 rules explained in popular guidelines [17, 28]. We find that existing static checkers may miss violations of a significant number of rules due to limitations of static analysis. Motivated by these findings, our work complements existing static checkers by providing a dynamic approach for checking code quality rules. To formally specify when a rule violation occurs, we describe violations in terms of predicates over events that happen during an execution.

**Definition 2 (Runtime event predicate)**
A runtime event predicate is a specification that matches one of the following parameterized runtime events.
- `lit(val)` matches a literal, where the literal value is `val`.
- `varRead(name, val)` matches a variable read, where the variable is called `name` and has value `val`.
- `call(base, f, args, ret,isConstr)` matches a function call, where the base object is `base`, the called function is `f`, the arguments passed to the function are `args`, the return value of the call is `ret`, and where `isConstr` specifies whether the call is a constructor call.
- `propRead(base, name, val)` matches a property read, where the base object is `base`, the property is called `name`, and the value of the property is `val`.
- `propWrite(base, name, val)` matches a property write, where the base object is `base`, the property is called `name`, and the value written to the property is `val`.
- `unOp(op, val, res)` matches a unary operation, where the operator is `op`, the input value is `val`, and the result value is `res`.
- `binOp(op, left, right, res)` matches a binary operation, where the operator is `op`, the left and right operands are `left` and `right`, respectively, and the result value is `res`.
- `cond(val)` matches a conditional, where the conditional value is `val`.
- `forIn(val)` matches a for-in loop that iterates over the object `val`.

A predicate either constrains the value of a parameter or specifies with * that the parameter can have any value. For example, `varRead('foo, *)` is a runtime predicate that matches any read of a variable called `foo`, independent of the variable’s value. The above list focuses on the events and parameters required for the runtime patterns presented in this paper. Our implementation supports additional events and parameters to enable extending DLINT with additional checkers.

Based on runtime event predicates, DLINT allows for specifying when a program violates a code quality rule during the execution. We call such a specification a runtime pattern and distinguish between two kinds of patterns:

**Definition 3 (Single-event runtime pattern)**
A single-event runtime pattern consists of one or more event predicates over a single runtime event, where each predicate is a sufficient condition for violating a code quality rule.

For example, `varRead('foo, *)` is a single-event runtime patterns that addressed the (trivial) rule that no variable named `foo` should ever be read.

**Definition 4 (Multi-event runtime pattern)**
A multi-event runtime pattern consists of event predicates over two or more runtime events, if they occur together, are a sufficient condition for violating a code quality rule.

For example, `varRead('foo, *) ∧ varWrite('foo, *)` is a multi-event runtime pattern that addresses the (again trivial) rule to not both read and write a variable named `foo` during an execution.

Because single-event runtime patterns match as soon as a particular event occurs, they can be implemented by a stateless dynamic analysis. In contrast, multi-event runtime patterns match only if two or more related events occur. Therefore, they require a stateful dynamic analysis. Most rules we consider can be expressed as single-event runtime patterns.

The remainder of this section presents some of the code quality rules and their corresponding runtime patterns we address in DLINT. Each rule is addressed by a checker that identifies occurrences of the runtime pattern. To the best of our knowledge, we provide the first comprehensive formal description of dynamic checks that address otherwise informally specified rules. Due to limited space, we discuss only a subset of all currently implemented checkers. The full list of checkers is available on our project homepage.

**Abbreviations:** We specify runtime patterns using abbreviations: `isFct(x), isObject(x), isPrim(x), and isString(x)` are true if `x` is a function, an object, a primitive value, and a string, respectively. `isArray(x), isCSSObj(x), isFloat(x), isNumeric(x), and isRegExp(x)` are true if `x` is an array, a CSS object, a floating point value, a value that coerces into a string, respectively. `relOrEqOp` refers to a relational operator or an equality operator. Finally, `argumentProps` and `arrayProps` refer to the set of properties of the built-in `arguments` variable and the set of properties in `Array.prototype`, respectively.

### 2.2 Problems Related to Inheritance

JavaScript’s implementation of prototype-based inheritance not only offers great flexibility to developers but also provides various pitfalls that developers should avoid. To address some of these pitfalls, Table 1 shows DLINT checkers that target inheritance-related rules. The following explains two of these checkers in detail.

#### 2.2.1 Inconsistent Constructor

Each object has a constructor property that is supposed to return the constructor that has created the object. Unfortunately, JavaScript does not enforce that this property returns the constructor, and developers may (accidentally) set this property to arbitrary values. The problem is compounded by the fact that all objects inherit a constructor property from their prototype.

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Because the code does not assign the correct constructor function to Sub's prototype, accessing the constructor of an instance of Sub returns Super.

To detect such inconsistent constructors, DLint checks each read of the constructor property whether the property's value is the base object's constructor function (Checker I2 in Table 1). To access the function that has created an object, our implementation stores this function in a special property of every object created with the new keyword.

### 2.2.2 Shadowing Prototype Properties

Prototype objects can have properties, which are typically used to store data shared by all instances of a prototype. Developers of Java-like languages may think of prototype properties as static (class-level) fields. In such languages, it is forbidden to have an instance field with the same name as an existing static field. In contrast, JavaScript does not warn developers when an object property shadows a prototype property. However, shadowing is discouraged because developers may get easily confused about which property there are accessing.

To identify shadowed prototype properties, Checker I5 in Table 1 warns about property writes where the property is already defined in the base object's prototype chain. For example, the following code raises a warning:

```javascript
var s = new Sub();
```

There are two common and harmless kinds of violations of this rule in client-side JavaScript code: First, changing prototype properties of DOM objects, e.g., innerHTML. Second, overriding a super class' function, i.e., a property of the prototype object. To avoid overwhelming developers with unnecessary warnings, the checker excludes a set `shadowingAllowed` of such DOM properties and properties that refer to functions.

### 2.3 Problems Related to Types

JavaScript does not have compile time type checking and is loosely typed at runtime. As a result, various problems that would lead to type errors in other languages may remain unnoticed. Table 2 shows DLINT checkers that warn about such problems by checking type-related rules. Two of these checkers require to check for occurrences of multi-event runtime patterns. We explain three type-related checkers in the following.

#### 2.3.1 Too Many Arguments

Checker T6 warns about function calls that pass more arguments than the function expects.

```
f(x, x, x, x, x) // too many arguments
```

There are two common and harmless kinds of violations of this rule in client-side JavaScript code: First, changing prototype properties of DOM objects, e.g., innerHTML. Second, overriding a super class' function, i.e., a property of the prototype object. To avoid overwhelming developers with unnecessary warnings, the checker excludes a set `shadowingAllowed` of such DOM properties and properties that refer to functions.

<table>
<thead>
<tr>
<th>ID</th>
<th>Name</th>
<th>Code quality rule</th>
<th>Runtime event predicate(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>11</td>
<td>Enumerable-ObjProp</td>
<td>Avoid adding enumerable properties to Object. Doing so affects every for-in loop.</td>
<td>`propWrite(Object, x, x) &amp;&amp; call(Object, defineProperty, args, x, x)</td>
</tr>
<tr>
<td>12</td>
<td>Inconsistent-Constructor</td>
<td><code>x.constructor</code> should yield the function that has created <code>x</code>.</td>
<td>`propRead(base, constructor, val)</td>
</tr>
<tr>
<td>13</td>
<td>NonObject-Prototype</td>
<td>The prototype of an object must be an object.</td>
<td>`propWrite(x, name, val)</td>
</tr>
<tr>
<td>14</td>
<td>Overwrite-Prototype</td>
<td>Avoid overwriting an existing prototype, as it may break the assumptions of other code.</td>
<td>`propWrite(base, name, x)</td>
</tr>
<tr>
<td>15</td>
<td>Shadow-ProtoProp</td>
<td>Avoid shadowing a prototype property with an object property.</td>
<td>`propWrite(base, name, val)</td>
</tr>
</tbody>
</table>

#### Table 1: Inheritance-related code quality rules and runtime patterns (all are single-event patterns).

<table>
<thead>
<tr>
<th>ID</th>
<th>Name</th>
<th>Code quality rule</th>
<th>Runtime event predicate(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>FunctionVs-Prim</td>
<td>Avoid comparing a function with a primitive.</td>
<td>`binOp(relOrEqOp, left, right, s)</td>
</tr>
<tr>
<td>2</td>
<td>StringAnd-Undefined</td>
<td>Avoid concatenating a string and undefined, which leads to a string containing &quot;undefined&quot;.</td>
<td>`binOp(+, left, right, res)</td>
</tr>
<tr>
<td>3</td>
<td>ToString</td>
<td>toString must return a string.</td>
<td>`call(*, toString, s, ret, x)</td>
</tr>
<tr>
<td>4</td>
<td>Undefined-Prop</td>
<td>Avoid accessing the &quot;undefined&quot; property.</td>
<td><code>propWrite(*, &quot;undefined&quot;, s) ∧ propRead(*, &quot;undefined&quot;, s)</code></td>
</tr>
</tbody>
</table>

#### Table 2: Code quality rules and runtime patterns related to type errors.

For example, consider the following code, which mimics class-like inheritance in an incorrect way:

```javascript
function Super() {} // superclass constructor
function Sub() {} // subclass constructor
Super.call(this);
Sub.prototype = Object.create(Super.prototype);
// Sub.prototype.constructor = Sub; // should do this
```

There are two common and harmless kinds of violations of this rule in client-side JavaScript code: First, changing prototype properties of DOM objects, e.g., innerHTML. Second, overriding a super class' function, i.e., a property of the prototype object. To avoid overwhelming developers with unnecessary warnings, the checker excludes a set `shadowingAllowed` of such DOM properties and properties that refer to functions.

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#### 2.3.1 Too Many Arguments

Checker T6 warns about function calls that pass more ar-
arguments to a function than the number of arguments that the function handles. JavaScript provides two ways to access the arguments passed to a function: formal parameters and the built-in arguments variable that enables array-like access to all arguments. To handle these two ways to access arguments, the checker is implemented as a multi-event pattern. A warning is raised if and only if there is a call to f’s body, there is f’s formal arguments.length > 0. Given this, a developer would replace the loop body from above with var x = 0, array = [11, 22, 33];
for (x in array) {
  sum += array[x];
}
console.log(sum);
var fact = 42;

Unfortunately, setting a property of a primitive does not have the expected effect because the property is attached to a wrapper object that is immediately discarded afterwards. In the example, the second statement does not modify fact but a temporarily created instance of Number, and fact.isTheAnswer yields undefined after the assignment.

Developers can prevent such surprises by following the rule that setting properties of primitives should be avoided. Checker L6 checks for violations of this rule by warning about every property write event where the base value is a primitive.

### 2.4.3 Unnecessary Reference to this

The semantics of this in JavaScript differ from other languages and often confuse developers. When accessing this in the context of a function, the value depends on how the function is called. In the global context, i.e., outside of any function, this refers to the global object. Because the global object is accessible without any prefix in the global context, there is no need to refer to this, and a program that accesses this in the global context is likely to confuse the semantics of this. Checker L3 warns about accesses of this in the global context by checking whether a variable read of this yields the global object.

#### 2.5 Problems Related to API Misuse

As most APIs, JavaScript’s built-in API and the DOM API provide various opportunities for misusing the provided functionality. Motivated by commonly observed mistakes, several DLint checkers address rules related to incorrect, unsafe, or otherwise discouraged API usages (Table 4). The following explains three checkers in detail.

### 2.5.1 Truthiness of Wrapped Primitives

The built-in constructor functions Boolean, Number, and String enable developers to wrap primitive values into objects. However, because objects always coerce to true in conditionals, such wrapping may lead to surprising behavior when the truthiness of a boolean changes due to wrapping. For example, consider the following example, where the code prints “true” in the second branch, even though b is false:

```javascript
var b = false;
if (b) console.log("true");
if (new Boolean(b)) console.log("true");
```

To avoid such surprises, developers should avoid evaluating wrapped boolean in conditionals. Checker A8 warns about code where a Boolean object appears in a conditional and where the value wrapped by the object is false.

### 2.5.2 Futile Writes of Properties

Some built-in JavaScript objects allow developers to write a particular property, but the write operation has no effect at runtime. For example, typed arrays simply ignore all out-of-bounds writes. Even though such futile writes are syntactically correct, they are likely to be unintended and, even worse, difficult to detect because JavaScript silently executes them.

Checker A4 addresses futile writes by warning about property write operations where, after the write, the base object’s property is different from the value assigned to the property. In particular, this check reveals writes where the property remains unchanged. An alternative way to check for futile writes is to explicitly search for writes to properties that are known to not have any effect. The advantage of the runtime predicate we use is to provide a generic checker that detects all futile writes without requiring an explicit list of properties.

### 2.5.3 Treating style as a String

Each DOM element has a style attribute that determines its visual appearance. The style attribute is a string in HTML, but the style property of an HTML DOM element is an object in JavaScript. For example, the JavaScript

<table>
<thead>
<tr>
<th>ID</th>
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<th>Runtime event predicate(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>A1</td>
<td>Double-Evaluation</td>
<td>Avoid eval and other ways of runtime code injection.</td>
<td>call(builtin, eval, *, *, *)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>call(builtin, Function, *, *, *)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>call(builtin, setTimeout, args, *, *)</td>
</tr>
<tr>
<td>A2</td>
<td>EmptyChar-Class</td>
<td>Avoid using an empty character class, [], in regular expressions, as it does not match anything.</td>
<td>lit(val)</td>
</tr>
<tr>
<td>A3</td>
<td>FunctionToString</td>
<td>Avoid calling Function.toString(), whose behavior is platform-dependent.</td>
<td>call(builtin, RegExp, args, *, *)</td>
</tr>
<tr>
<td>A4</td>
<td>FutileWrite</td>
<td>Writing a property should change the property’s value.</td>
<td>propWrite(base, name, val)</td>
</tr>
<tr>
<td>A5</td>
<td>MissingRadix</td>
<td>Pass a radix parameter to parseInt(), whose behavior is platform-dependent otherwise.</td>
<td>call(builtin, parseInt, args, *, *)</td>
</tr>
<tr>
<td>A6</td>
<td>SpacesIn-Regexp</td>
<td>Prefer &quot;{N}&quot; over multiple consecutive empty spaces in regular expressions for readability.</td>
<td>(lit(val) ∨ call(*, RegExp, *, *, *))</td>
</tr>
<tr>
<td>A7</td>
<td>StyleMisuse</td>
<td>CSS objects are not strings and should not be used as if they were.</td>
<td>binOp(eqOp, left, right)</td>
</tr>
<tr>
<td>A8</td>
<td>Wrapped-Primitives</td>
<td>Beware that all wrapped primitives coerce to true.</td>
<td>cond(val)</td>
</tr>
</tbody>
</table>

Table 4: Code quality rules and runtime patterns related to incorrect API usage (all are single-event patterns).
DOM object that correspond to the HTML markup `<div style="top:10px;"></div>` is a CSS object with a property named `top`. The mismatch between JavaScript and HTML types sometimes causes confusion, e.g., leading to JavaScript code that retrieves the `style` property and compares it to a string. Checker A7 identifies misuses of the `style` property by warning about comparison operations, e.g., `===` or `!==`, where one operand is a CSS object and where the other operand is a string.

### 2.6 Problems Related to Uncommon Values

The final group of checkers addresses rules related to uncommon values that often occur unintendedly (Table 5). We explain one of them in detail.

#### 2.6.1 Not a Number

The `NaN` (not a number) value may result from exceptional arithmetic operations and is often a sign of unexpected behavior. In JavaScript, `NaN` results not only from operations that produce `NaN` in other languages, such as division by zero, but also as the result of unusual type coercions. For example, applying an arithmetic operation to a non-number, such as `23 - "five"`, may yield `NaN`. Since generating `NaN` does not raise an exception or any other kind of warning in JavaScript, `NaN`-related problems can be subtle to identify and hard to diagnose.

In most programs, developers want to follow the rule that suggests to avoid occurrences of `NaN`. Checker V2 warns about violations of this rule by identifying operations that take non-`NaN` values as inputs and that produce a `NaN` value. The checker considers unary and binary operations as well as function calls.

### 3. IMPLEMENTATION

We implement DLINT as an automated analysis tool for JavaScript-based web applications and node.js applications. The system has multiple steps. First, DLINT opens a web site in Firefox, which we modify so that it intercepts all JavaScript code before executing it, including code dynamically evaluated through `eval`, `Function`, `setTimeout`, `setInterval`, and `document.write`. Second, DLINT instruments the intercepted code to add instructions that perform the checks. This part of DLINT builds upon the dynamic analysis framework Jalangi [50]. Third, while the browser loads and renders the web site, the instrumented code is executed and DLINT observes its runtime events. If an event or a sequence of events matches a runtime pattern, DLINT records this violation along with additional diagnosis information. Fourth, after completely loading the web site, DLINT automatically triggers events associated with visible DOM elements, e.g., by hovering the mouse over an element. This part of our implementation builds upon Selenium. We envision this step to be complemented by a UI-level regression test suite, by manual testing, or by a more sophisticated automatic UI testing approach [7, 15, 47]. Finally, DLINT gathers the warnings from all checkers and reports them to the developer. Our prototype implementation has around 12000 lines of JavaScript, Java and Bash code, excluding projects we build upon. The implementation is available as open-source.

A key advantage of DLINT is that the framework can easily be extended with additional dynamic checkers. Each checker registers for particular runtime events and gets called by the framework whenever these kinds of events occur. The framework dispatches events to an arbitrary number of checkers and hides the complexity of instrumentation and dispatching, making it easy to extend the set of checkers. For example, consider the execution of a `if=1.25`. DLINT instruments this statement so that the framework dispatches the following four events, in addition to executing the original code: `varRead("x", x1), propRead(x1,"g", x2), varRead("a", x3), propWrite(x3,"f", x2)`, where `x` refer to runtime values.

### 4. EVALUATION

We evaluate DLINT through an empirical study on over 200 web sites. Our main question is whether dynamically checking for violations of code quality rules is worthwhile. Section 4.2 addresses this question by comparing DLINT to a widely used static code quality checker. Section 4.3 explores the relationship between code quality and the popularity of a web site. We evaluate the performance of DLINT in Section 4.4. Section 4.5 presents examples of problems that DLINT reveals. Finally, Section 4.6 discusses threats to the validity of our conclusions.

### 4.1 Experimental Setup

The URLs we analyze come from two sources. First, we analyze the 50 most popular web sites, as ranked by Alexa. Second, to include popular web sites that are not landing pages, we search Google for trending topics mined from Facebook and include the URLs of the top ranked results. In total, the analyzed web sites contain 4 million lines of JavaScript code. Since many sites ship minified source code, where an entire script may be printed on a single line, we pass code to `js-beautify` before measuring the number of

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Table 5: Code quality rules and runtime patterns related to uncommon values (all are single-event patterns).

<table>
<thead>
<tr>
<th>ID</th>
<th>Name</th>
<th>Code quality rule</th>
<th>Runtime event predicate(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>V1</td>
<td>Float-Equality</td>
<td>Avoid checking the equality of similar floating point numbers, as it may lead to surprises due to rounding. (^2)</td>
<td>`binOp(eqOp, left, right, *)</td>
</tr>
<tr>
<td>V2</td>
<td>NaN</td>
<td>Avoid producing NaN (not a number).</td>
<td>`unOp(*, val, NaN)</td>
</tr>
<tr>
<td>V3</td>
<td>Overflow-Underflow</td>
<td>Avoid numeric overflow and underflow.</td>
<td>`binOp(*, left, right, NaN)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>`call(*)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td><code>unOp(*, val, \(\neq\) \infty)</code></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td><code>binOp(*, left, right, \(\neq\) \infty)</code></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td><code>call(builtin, *, args, \(\neq\) \infty)</code></td>
</tr>
</tbody>
</table>

\(^2\) It is a notorious fact that the expression `0.1 + 0.2 \(\neq\) 0.3` returns `false` in JavaScript.
lines of code. We fully automatically analyze each URL as described in Section 3.

To compare DLint to static checking, we analyze all code shipped by a web site with JSHint. To the best of our knowledge, JSHint is currently the most comprehensive and widely used static, lint-like checker for JavaScript. We compare the problems reported by DLint and JSHint through an AST-based analysis that compares the reported code locations and the kinds of warnings.

4.2 Dynamic versus Static Checking

DLint checks 28 rules, of which 5 have corresponding JSHint checkers. JSHint checks 150 rules, of which 9 have corresponding DLint checkers. There is no one-to-one mapping of the overlapping checkers. For example, DLint’s “DoubleEvaluation” checker corresponds to several JSHint checkers that search for calls of eval and eval-like functions. In total over all 200 web sites analyzed, DLint reports 9018 warnings from 27 checkers, and JSHint reports about 580k warnings from 91 checkers. That is, JSHint warns about significantly more code quality problems than DLint. Most of them are syntactical problems, such as “missing semicolon”, and therefore are out of the scope of a static analysis. For a fair comparison, we focus on JSHint checkers that have a corresponding DLint checker.

RQ1: How many violations of code quality rules are detected by DLint but missed by static checkers?

Figure 3a shows for each analyzed web site the percentage of warnings reported only by JSHint, by both DLint and JSHint, and only by DLint. Each horizontal line represents the distribution of warnings for a particular web site. The results show that DLint identifies warnings missed by JSHint for most web sites and that both checkers identify a common set of problems.

To better understand which DLint checkers contribute warnings that are missed by JSHint, Figure 3b shows the number of warnings reported by all DLint checkers, on average per web site. The black bars are for checkers that report problems that are completely missed by JSHint. These checkers address rules that cannot be easily checked through static analysis. The total number of DLint warnings per site ranges from 1 to 306. On average per site, DLint generates 53 warnings, of which 49 are problems that JSHint misses.

RQ2: How many additional code quality problems are found by dynamic variants of static checkers?

One of the motivations of this work is that a pragmatic static analysis may miss problems even though it searches for them. In RQ2, we focus on DLint checkers that detect rules that are also checked by JSHint and measure how many problems are missed by JSHint but revealed by DLint. Figure 4a (4b) shows the number of warnings detected by JSHint (DLint) checkers that address a rule also checked by DLint (JSHint). The figure shows that JSHint and DLint are complementary. For example, JSHint and DLint both detect 205 calls of Function, which is one form of calling the evil eval. JSHint additionally detects 359 calls that are missed by DLint because the call site is not reached at runtime. DLint additionally detects 181 calls of eval and eval-like functions, which includes calls of Function.

Considering all checkers shown in Figure 4, DLint reports 10.1% additional warnings that are missed by JSHint even though JSHint checks for the rule. Manual inspection of these problems shows that they are due to code that is hard or impossible to analyze for a pragmatic static checker, e.g., code that assigns Function to another variable and calls it through this alias.

We conclude from the results for RQ1 and RQ2 that dynamically checking for violations of code quality rules is worthwhile. DLint complements existing static checkers by revealing problems that are missed statically and by finding violations of rules that cannot be easily checked through static analysis.

Coverage: Dynamic analysis is inherently limited to the parts of a program that are covered during an execution. To understand the impact of this limitation, we compute coverage as the number of basic operations that are executed at least once divided by the total number of basic operations. For example, if \( (b) \ a=1 \) consists of three basic operations: read \( b \), test truthiness of \( b \), and write \( a \). If during the execution, \( b \) was always falsy, the coverage rate would be \( 2/3 = 66.7\% \). Figure 3c shows the coverage achieved during the executions that DLint analyzes. Each line represents the coverage rate of one web site. Overall, coverage is relatively low for most web sites, suggesting that, given a richer test suite, DLint could reveal additional rule violations.

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9 JSHint checks more code quality rules than JSLint. ESLint is a re-implementation of JSLint to support pluggable checkers.
rupted user interfaces and displaying incorrect data, while stumbling across 19 examples of serious problems, such as cor-

4.5 Examples of Detected Problems

We address this question by analyzing web sites of different popularity ranks and by computing the overall number of warnings produced by DLint and JSHint. More specifically, we gather a random sample of web sites from the Alexa top 10, 100, . . . , 10000 web sites and apply both checkers to each set of samples. Figure 5 shows the number of reported warnings. The left figure shows the relation between the number of JSHint warnings reported per LOC (vertical axis) and the ranking of the web site (horizontal axis). The right figure shows the relation between the number of DLint warnings per operation executed (vertical axis) and the ranking of the web site (horizontal axis). The correlation is 0.6 and 0.45 for DLint and JSHint, respectively, suggesting that popular web sites tend to have fewer violations of code quality rules.

4.4 Performance of the Analysis

Figure 3d shows the overall time required to apply DLint to a web site. The time is composed of the time to intercept and instrument JavaScript code, the time to execute code and render the page, and the time to summarize and report warnings. Most of the time is spent executing the instrumentation code, which naturally is higher than executing non-instrumented code [50]. Given that DLint is a fully automatic testing approach, we consider the performance of our implementation to be acceptable.

4.3 Code Quality versus Web Site Popularity

RQ3: How does the number of violations of code quality rules relate to the popularity of a web site?

We address this question by analyzing web sites of different popularity ranks and by computing the overall number of warnings produced by DLint and JSHint. More specifically, we gather a random sample of web sites from the Alexa top 10, 100, . . . , 10000 web sites and apply both checkers to each set of samples. Figure 5 shows the number of reported warnings. The left figure shows the relation between the number of JSHint warnings reported per LOC (vertical axis) and the ranking of the web site (horizontal axis). The right figure shows the relation between the number of DLint warnings per operation executed (vertical axis) and the ranking of the web site (horizontal axis). The correlation is 0.6 and 0.45 for DLint and JSHint, respectively, suggesting that popular web sites tend to have fewer violations of code quality rules.

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4.5 Examples of Detected Problems

The main goal of checking code quality rules is not to detect serious bugs but to help developers to avoid code that potentially causes problems. Nevertheless, we were happy to stumble across 19 examples of serious problems, such as corrupted user interfaces and displaying incorrect data, while inspecting warnings reported by DLint. Due to limited space, this section only reports some illustrative examples of such problems, all of which are missed by JSHint. Additional examples are given in the introduction.

Not a Number. Checker V2 reveals several occurrences of NaN that are visible to visitors of a site. For example, Figure 6 shows some NaN bugs detected by DLint on the web sites of IKEA (1) and eBay (2), where articles cost the incredibly cheap “$NaN”, and an official website of a basketball team10 (3), where a basketball player had “NaN” steals and turnovers. DLint reports the code location where a NaN originates, making it relatively easy to understand the root cause. For example, the IKEA web site loads an XML file containing the data to display, parses the data, and dynamically inserts the results into the DOM. Unfortunately, some data items are missing and the JavaScript code initializes the corresponding variables with undefined, which are involved in an arithmetic operation that finally yields NaN.

Overflows and Underflows. Figure 6(4-5) shows two bugs related to arithmetic overflow and underflow detected by DLint on the sites of Tackle Warehouse and CCNEX. They are due to arithmetic operations that yield an infinite value that propagates to the DOM.

Futile Write. DLint warns about the following code snippet on Twitch, a popular video streaming web site:

```javascript
window.onbeforeunload=
    "Twitch.player.getPlayer().pauseVideo();"

window.onunload="Twitch.player.getPlayer().pauseVideo();"
```

The code attempts to pause the video stream when the page is about to be closed. Unfortunately, these two event

10http://www.uconnhuskies.com/
handlers are still null after executing the code. Although it is not uncommon to use a string as a parameter indicating the code to be executed, e.g., for setTimeout and setInterval, developers must assign a function object as an event handler of a DOM element. Writing a non-function into an event handler property is simply ignored by DOM implementations.

**Style Misuse.** DLINT found the following code on Craigslist:

```javascript
if (document.body.style === "width:100\%") { ... }
```

The developer tries to compare style with a string, but style is an object that coerces into a string that is meaningless to compare with, e.g., CSS2Properties in Firefox.

**Properties of Primitives.** Besides web sites, we also apply DLINT to the Sunspider and Octane benchmark suites. Due to space limitations, detailed results are omitted. The following is a problem that Checker L6 detects in Octane’s GameBoy Emulator benchmark:

```javascript
var decode64 = "";
if (dataLength > 3 & dataLength % 4 == 0) {
    while (index < dataLength) {
        decode64 += String.fromCharCode(...);
    }
    if (sixbits[3] >= 0x40) {
        sixbits[3] = 0x00;
    }
    if (sixbits[3] >= 0x40) {
        sixbits[3] = 0x00;
    }
    decode64.length -= 1; // writing a string’s property
}
```

Line 7 tries to remove the last character of decode64. Unfortunately, this statement has no side effect because the string primitive is coerced to a String object before accessing length, leaving the string primitive unchanged.

### 4.6 Threats to Validity

The validity of the conclusions drawn from our results are subject to several threats. First, both DLINT and JSHint include a limited set of checkers, which may or may not be representative for dynamic and static analyses that check code quality rules in JavaScript. Second, since DLINT and JSHint use different reporting formats, our approach for matching the warnings from both approaches may miss warnings that refer to the same problem and may incorrectly consider reports as equivalent. We carefully inspect and revise the matching algorithm to avoid such mistakes.

## 5. RELATED WORK

Several dynamic analyses for JavaScript have been proposed recently, e.g., TypeDevil [42] detects type inconsistencies in JavaScript, EventRacer [44] identifies event-related data races, and Mesbah et al. check for cross-browser issues [39]. JITProf is a profiler that identifies performance bottlenecks caused by code that JIT engines cannot effectively optimize [24]. Each of these approaches addresses a particular kind of programming error, whereas DLINT is a generic framework for checking code quality rules. Other dynamic program analyses for JavaScript include determinacy analysis [49], information flow analyses [16, 9], library-aware static analysis [34], and symbolic execution approaches [48]. BEAR [23] allows for inferring user interaction requirements from runtime traces. In contrast to these approaches, our work aims at finding violations of coding practices.

Static analyses beyond the checkers discussed elsewhere in this paper include analyses to detect potential type errors [54, 32, 27, 25] and a sophisticated points-to analysis [53]. Feldthaus et al. propose a refactoring framework for specifying and enforcing JavaScript practices [20]. In contrast to these approaches, our work explores how dynamic analysis can complement existing static analyses. Several authors propose approaches for semi-automatically [37, 22] or automatically [14] repairing JavaScript applications. Automatically repairing violations of code quality rules detected by DLINT is an interesting direction for future work.

Artzi et al. propose a UI-level test generation framework for web applications [8]. EventBreak [41] uses a performance-guided test generation algorithm to identify unresponsiveness of web applications. These and other UI-level test generation approaches [38, 35, 19, 7, 18, 55, 15] may be combined with DLINT to extend the set of analyzed executions.

The complementary nature of static and dynamic analysis can be exploited by combining both approaches [52]. Feldthaus et al. combine dynamic and static analysis of JavaScript to check if TypeScript [40] interfaces match a library’s implementation [21]. DSD-Crasher [?] analyses static code and runtime behaviour to find bugs. Other approaches combine both approaches to find security vulnerabilities in web sites [26] and to debug and repair faults in Java programs [51].

FindBugs [30] is a static checker of code quality rules in Java, which is similar in spirit to the checkers we compare with and which also has been widely used in industry [12, 11]. PQL [36] is a “program query language” designed to express code quality rules that programmers should follow. It focuses on rules that can be expressed with a subset of the runtime events supported by DLINT. For example, PQL cannot express queries over unary and binary operations.

Several empirical studies to understand the abilities of bug detection tools have been performed. Rutar et al. compare static bug finding tools for Java [46]. Rahman et al. compare static bug checkers and defect prediction approaches [43]. Ayewah et al. [10] study and discuss the warnings reported by FindBugs [30]. Our work differs from these studies by directly comparing static and dynamic analyses for JavaScript.

## 6. CONCLUSION

We described DLINT, a dynamic analysis approach that consists of an extensible framework and 28 checkers that address problems related to inheritance, types, language misuse, API misuse, and uncommon values. Our work contributes the first formal description of these otherwise informally documented rules and the first dynamic checker for rule violations. We apply DLINT in a comprehensive empirical study on over 200 of the world’s most popular web sites and show that dynamic checking complements state-of-the-art static checkers. Static checking misses at least 10.1% of the problems it is intended to find. Furthermore, DLINT addresses problems that are hard or impossible to reveal statically, leading to 49 problems that are missed statically by found by DLINT, on average per web site. Since our approach scales well to real-world web sites and is easily extensible, it provides a first step in filling an currently unoccupied spot in the JavaScript tool landscape.

### Acknowledgments

This research is supported in part by NSF Grants CCF-1423645 and CCF-1409872, by gifts from Mozilla and Samsung, a Sloan Foundation Fellowship, and by the German Federal Ministry of Education and Research (BMBF) within EC SPRIDE.
7. REFERENCES


