

An Empirical Study of Web Visual Flakiness: Characterisation and Fix Strategies

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Abstract

Unpredictable behaviour in web front-ends undermines test reliability and user experience, often manifesting as inconsistent or unstable visual output. Prior work on web front-end flakiness has mainly focused on asynchronous waits, platform-specific behaviour, and logic errors, many of which stem from client-side execution, back-end dependencies, or timing variations across browsers. However, little attention has been given to pure visual flakiness—instabilities in layout, rendering, and styling that cause the same web content to appear differently across runs or environments.

In this work, we present a focused empirical study dedicated to web visual flakiness, defined as flakiness that manifests as structural or stylistic inconsistencies in web front-ends. We collected 262 cases of web visual flakiness: 144 from 31 open-source web projects and 118 from the Chromium project. Our findings show that web visual flakiness constitutes a notable portion of overall flaky cases and that visual changes are more prone to flakiness than non-visual ones, underscoring the need for focused investigation. We categorised the collected instances into two primary dimensions—structure-related (59.9%) and style-related (40.1%)—and further refined them into five concrete sub-categories by manifestation (three structure-related and two style-related). Finally, our analysis of developer fixes shows that while many address specific structural or stylistic inconsistencies, others are applied across multiple categories, revealing shared repair strategies.

Keywords:

Flaky Behaviour, Web Visuals, Web Testing, Empirical Study

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1. Introduction

As web applications grow increasingly complex, ensuring their reliability has become a critical challenge. Among many obstacles, flaky behaviours, such as test outcomes that show non-deterministic results across repeated runs, pose a constant threat to automated testing Luo et al. (2014); Parry et al. (2021); Lam et al. (2019a). This issue is particularly pronounced in the visual layer, where minor inconsistencies, such as browser-specific differences, asynchronous content loading, and ordering of DOM updates, can lead to unpredictable rendering and test failures Eck et al. (2019); Parry et al. (2022); Habchi et al. (2022); Gruber and Fraser (2022); Haben et al. (2024).

The visual layer encompasses the code and frameworks that define and render visual presentation in web rendering environments, including markup (e.g., HTML), stylesheets (e.g., CSS), and client-side logic (e.g., JavaScript). These components collectively control the structure and styling of the visual layer, making them central to both user experience and test stability. However, even subtle changes to these components can lead to visual inconsistencies that are difficult to detect and reproduce due to their non-deterministic nature (i.e., flakiness). Responsive design and dynamic content further amplify this instability across browsers and environments.

While previous research has investigated flaky tests in web and mobile environments, such as interaction-based GUI flakiness (e.g., Romano et al. (2021)) and deterministic UI rendering defects (e.g., Mahajan et al. (2016); Alameer and Halfond (2016); Althomali et al. (2021)), the focus has primarily been on functional-level non-determinism (e.g., async waits, improper test script logics, or state mismatches) or deterministic visual inconsistencies (e.g., cross-browser layout bugs). Meanwhile, studies on web flaky tests have primarily investigated non-determinism in functional or interaction-driven contexts, addressing causes such as asynchronous waits, platform-specific behaviour, or state management Parry et al. (2021); Lam et al. (2019a).

Taken together, prior studies have examined interaction-based GUI flakiness and other forms of functional non-determinism in UI behaviour, as well as deterministic visual faults, such as cross-browser layout bugs. However, they have not explicitly isolated non-deterministic visual flakiness—often manifesting as unstable layout, rendering, or styling—as a distinct empirical category. In this work, we investigate such visual flakiness as instabilities rooted in the visual layer that are inconsistently observable across executions and environments, bridging the gap between web flakiness research and

52 studies of visual correctness.

53 At the same time, web-visual testing is widely practised in industry, for
54 instance, through screenshot-based or layout-diff validation Yeh et al. (2009);
55 Choudhary et al. (2010, 2011); Stocco et al. (2018), yet its flaky behaviour
56 remains largely underexplored. Such flakiness is commonly observed in mod-
57 ern industrial web projects, highlighting the need for a dedicated empirical
58 investigation Rwemalika et al. (2019); Zou et al. (2014); Nass et al. (2023).

59 We present this study as a focused empirical investigation that system-
60 atically characterises non-deterministic visual flakiness in web applications
61 as a distinct analytical focus. To our knowledge, this study provides the
62 first systematic empirical analysis that explicitly focuses on web visual flak-
63 iness, examining its prevalence, characteristics, and developer responses. By
64 analysing 262 flaky cases from 31 open-source web projects and Chromium,
65 we identify how, where, and when visual flakiness manifests in real-world de-
66 velopment. Based on recurring patterns, we categorise them into structure-
67 related and style-related types. We also analyse how developers address
68 visual flakiness in practice, inspecting how these fixes relate to the defined
69 categories, and identifying common and cross-cutting fix strategies. In the
70 end, we address the following research questions:

- 71 • **RQ1:** What is the prevalence of flaky behaviours in web visuals?
- 72 • **RQ2:** What are the most common types of web visual flakiness, and
73 how do they manifest in practice?
- 74 • **RQ3:** What fix patterns can be employed to mitigate the flaky be-
75 haviour in web visuals?

76 Our findings, combined, highlight the importance of understanding visual
77 flakiness to improve the reliability of web testing. Below summarises the key
78 contributions:

- 79 • We conduct a focused empirical study of web visual flakiness as a dis-
80 tinct class of flaky behaviour, analysing 262 real-world cases across 31
81 open-source web projects and the Chromium codebase, and release the
82 resulting curated dataset to support future research.
- 83 • We provide a taxonomy of web visual flakiness, distinguishing two high-
84 level types, i.e., structure-related and style-related, based on recurring
85 manifestation patterns, and further refining each type into five concrete
86 subcategories.
- 87 • We investigate how developers handle web visual flakiness in practice,

88 identifying common, cross-cutting repair strategies and analysing their
89 relationships with different flakiness categories.

90 2. Methodology

91 2.1. Motivating Example

92 This study is motivated by the growing complexity of modern web front-
93 ends, where frequent modifications to DOM structure and visual styling can
94 introduce instability in the visual behaviour observed during execution. The
95 front-end of the web application is the layer where users directly interact with
96 the system, making visual instability particularly impactful. Even minor
97 inconsistencies, such as unexpected element positioning or missing visual
98 focus, can disrupt user actions and degrade user.

99 Consider the example in Figure 1a², where clicking the *Edit* menu item is
100 intended to automatically focus the text box, enabling users to begin editing
101 immediately (Figure 1b). However, as shown in Figure 1c, the text box some-
102 times fails to gain focus, requiring users to manually click it again. While
103 the underlying logic remains unchanged, this execution-time visual instabil-
104 ity, manifesting as inconsistent focus behaviour (i.e., flakiness), disrupts user
105 workflow and causes frustration, particularly in applications where users ex-
106 pect quick and fluid interactions. Such visual flakiness is often difficult to
107 detect and diagnose. Understanding the nature and recurring patterns of
108 such flakiness requires reasoning about instability in the visual layer, which
109 is inherently subtle, brittle, and sensitive to small variations Mahajan et al.
110 (2016); Pei et al. (2025).

111 2.2. Study Design

112 We conducted an empirical study to characterise flaky behaviours in web
113 visuals in real-world projects. To this end, we first collected commits and,
114 when available, their respective pull requests or issue reports related to web
115 visual flakiness from diverse web projects. We then investigated the observed
116 types of visual flakiness derived from recurring patterns and how developers
117 address these issues in practice, including common fix strategies. To support
118 this investigation, we performed a combination of manual inspection and
119 LLM-assisted annotation to categorise both visual flakiness types and their
120 corresponding fix strategies.

²<https://github.com/angular/components/issues/18750>

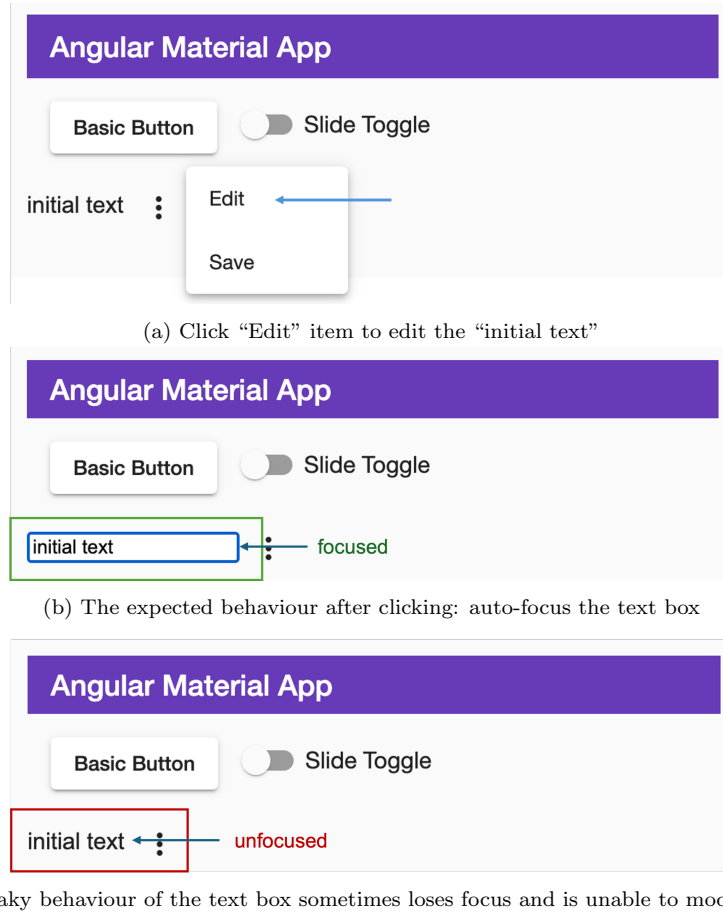


Figure 1: Example: flaky behaviour in the text box autofocus

121 2.2.1. Terminology

122 To clarify the focus of our study, we introduce two key terms that define
 123 what we consider visual-related flakiness in web projects. These definitions
 124 distinguish our scope from general categories of web or UI flakiness that
 125 include functional or interaction-driven behaviours: *visual commits* and *web*
 126 *visual flakiness*.

- 127 • **Visual commits** are commits that include changes that influence the
 128 visual behaviour or appearance of a web project, such as those related
 129 to DOM structure, styling, or rendering logic.
- 130 • **Web visual flakiness** refers to non-deterministic, intermittent incon-

131 consistencies that manifest in the visual layer of web projects. Such in-
 132 consistencies may appear, for instance, as layout shifts, rendering dif-
 133 ferences, or style-related visual inconsistencies.

134 Based on the above terminology, we define a web visual flakiness-related
 135 commit as any visual commit that exhibits flaky behaviour in the visual layer.
 136 In contrast, non-visual commits are those whose changes do not influence the
 137 rendering or presentation of web content, such as modifications unrelated to
 138 the DOM structure or styling.

139 2.2.2. Dataset Collection

140 To collect web visual flakiness cases³, we followed a systematic procedure
 141 similar to prior flakiness studies Luo et al. (2014); Romano et al. (2021);
 142 Hashemi et al. (2022). We focus on open-source web application projects
 143 on GitHub, where commit-level information is available and enables detailed
 144 analysis of code changes and their context.

145 We first identified repositories primarily using web development languages
 146 such as JavaScript and HTML and selected those with over 100 stars to
 147 ensure their representativeness. In total, we collected 31 projects comprising
 148 489,926 commits (Table 1, third column).

149 Next, we identified flaky commits by searching commit messages and,
 150 when available, associated issues or pull requests for flakiness-related key-
 151 words. We use “*flak**”, “*intermit**”, and “*unstable*” as our flakiness-related
 152 keyword set. This yielded 1,377 candidate commits related to flakiness
 153 ($\#Flaky$ in Table 1). Our primary interest lies in flakiness arising from
 154 changes whose effects manifest in the visual behaviour of web applications.
 155 Hence, in parallel with the flaky-commit search, we conducted a keyword
 156 search for visual commits using visual-related keywords, which contains “*e2e*”,
 157 “*UI*”, “*GUI*”, “*style*”, “*CSS*”, “*visual*”, “*layout*”, and “*display*”. If a commit
 158 message, issue, or pull request contained any of these visual keywords, we
 159 considered it likely to affect the visual parts of the application.⁴ In total,
 160 65,626 visual commits were identified ($\#_{commits}^{Visual}$, Table 1).

³We use the term flakiness case to refer to an instance of flaky behaviour, regardless of whether it originates from test code or application code.

⁴While some of the collected commits labelled as “*visual*” based on the keyword matching may not directly interact with visual parts, this does not affect the conclusion of our analysis in Section 3.1, as the more precise identification decreases the denominator ($\#_{commits}^{Visual}$) in the frequency analysis, further supporting the findings.

Table 1: Subjects. $\#Commits$, $\#_{Commits}^{Visual}$, $\#Flaky$, $\#FlakyVisual$ are the number of total commits, visual-related commits, all flaky-related cases, and visual-related flakiness cases

Projects	#Stars	#Commits	$\#_{Commits}^{Visual}$	#Flaky	#FlakyVisual
Chromium	18.9k	–	–	7780	118
vets-website	240	29,633	3506	235	19
ionic-framework	51k	13,970	1736	62	17
web-platform-tests	4.9k	62,667	17,486	621	15
angular-components	24.3k	12,162	2339	40	12
matrix-react-sdk	1.1k	48,466	6749	78	10
storybook	84.1k	63,182	2858	64	7
material-components	17.1k	7,453	1674	31	7
fluentui	18.4k	18,590	2768	18	6
ant-design	92.1k	28,580	4315	14	5
eui	6.1k	6,187	1127	18	5
openverse	243	11,200	621	44	5
material-ui	93.5k	25,656	3635	21	4
chakra-ui	37.6k	9,938	1865	8	4
tailwindcss	82.3k	5,568	806	7	3
bootstrap	170k	22,900	2378	5	2
vueify	39.7k	16,186	1831	5	2
quasar	25.8k	14,427	1889	5	2
svelte	78.7k	9,344	830	9	2
ghost	47.1k	39,504	4054	50	2
element	54.1k	4,612	527	4	2
blueprint	20.7k	3,581	447	4	2
balm-ui	506	3,131	129	2	2
bulma	49.2k	1,881	120	1	1
nuxt	54.3k	6,754	268	3	1
semi-design	8.4k	3,785	379	2	1
floating-ui	29.8k	2,221	246	3	1
vue.draggable	20.1k	521	22	1	1
vue-validate	10.8k	4,777	136	2	1
vue-material	9.9k	1,132	217	1	1
vite	67.7k	7,051	481	16	1
xyflow	25k	4,867	187	3	1
Open-source	240–170k	489,926	65,626	1377	144
TOTAL	-	489,926	65,626	9157	262

161 Lastly, to pinpoint web visual flakiness-related commits, we intersected
162 the sets of flaky and visual commits. Two authors manually reviewed each
163 candidate, examining the code changes⁵ and contextual information, such as
164 commit messages and, when available, related issues, pull requests, or devel-
165 oper comments, to ensure that each case genuinely involved web UI flakiness.
166 This conservative strategy—combining the intersection of strict flakiness and
167 visual keywords with manual validation—was chosen to prioritise data pre-
168 cision over exhaustive coverage to support the confidence in our findings
169 from subsequent qualitative analysis. Section 5.2 further discusses this data
170 sampling strategy. The last column (*#FlakyVisual*) in the “Open-source”
171 row of Table 1 reports 144 identified web visual flakiness cases across the 31
172 open-source projects.

173 **Chromium.** To enhance the representativeness of our study, we also in-
174 cluded Chromium, a major industrial browser project. For Chromium, we
175 adapted the keyword-based matching procedure to its issue tracker⁶, where
176 flaky behaviours are reported and managed.

177 Specifically, we searched issue reports for flakiness-related keywords and
178 filtered them with visual-related terms, using the same keyword sets as for
179 the open-source projects. Although Chromium maintains a mirrored GitHub
180 repository, the correspondence between issue reports and individual commits
181 is often incomplete, indirect, or missing. As a result, commit-level infor-
182 mation (e.g., code diffs and commit statistics) is not consistently available
183 for Chromium and, thus, commit-level analyses are not performed for this
184 dataset. The collection process yielded 118 web visual flakiness cases from
185 Chromium (the second row in Table 1), resulting in a total of 262 web visual
186 flakiness cases across all studied subjects (the last row in Table 1).

187 The two primary data sources employed in this investigation exhibit a no-
188 table structural distinction. The open-source web projects dataset is derived
189 from GitHub commits, which enables the identification of web visual flakiness
190 and corresponding fixes through commit messages and code diffs. In contrast,

⁵We mainly considered changes to files written in HTML, JavaScript, TypeScript, CSS, or SCSS, as these languages commonly define and influence the visual structure and style of web applications.

⁶<https://issues.chromium.org/issues>. Chromium issues hosted on this platform include both current records and historical issues migrated from the legacy Monorail system, which we analyse in this study.

the Chromium dataset is derived from issue tracker records, where flaky behaviours are reported and discussed, often without a reliable correspondence to individual commits. Consequently, analyses that rely on commit-level granularity—such as LLM-assisted categorisation (Section 2.2.4) and flakiness frequency per commit (RQ1)—are only conducted for the open-source projects. For Chromium, the same taxonomic framework and validation procedure are applied to manually analyse issue descriptions, comments, and linked patches, ensuring methodological consistency despite differences in data availability.

2.2.3. Web Visual Flakiness Taxonomy

To better understand and analyse web visual flakiness, we manually inspected the collected instances in our dataset to search for recurring patterns. This analysis revealed two broad families of issues: some involved unstable DOM structures, where elements were not reliably present or positioned (e.g., elements missing, misaligned, or late), while others involved changes in appearance due to inconsistencies in style interpretation or resource loading (e.g., CSS conflicts or delayed fonts). Based on these recurring patterns, we derive a taxonomy of web visual flakiness consisting of two high-level families: **structure-related** and **style-related**.

- **Structure-related flakiness:** occurs when visual inconsistencies arise due to variations in how elements are rendered, positioned, or updated. These inconsistencies may stem from changes to the DOM structure (e.g., layout shifts due to dynamic element insertion or removal), asynchronous or delayed updates to DOM elements, or failures in loading content that result in missing or incomplete visual components; this type of flakiness occurs regardless of whether the DOM structure is explicitly modified, as even stable structures can exhibit inconsistent rendering under dynamic or environment-dependent conditions. Hence, structure-related flakiness involves all inconsistencies that relate to the state of the DOM structure, including element presence, positioning, and readiness, where positioning refers to whether an element is correctly realised in the layout rather than how its visual style is rendered.
- **Style-related flakiness:** occurs when visual inconsistencies arise due to variations in how style properties (e.g., CSS rules, font rendering, or visual appearance attributes) are applied, interpreted, or available across different environments, browsers, or rendering engines at run-

time. These inconsistencies affect the visual appearance of elements without involving changes to the DOM structure or the spatial presence and placement of elements. Style-related flakiness primarily affects visual properties such as colour, size, and font.

In summary, structure-related flakiness concerns which visual elements are present and where they appear on the page, while style-related flakiness concerns how visual elements look. These two represent high-level categories in our taxonomy. Each category includes multiple concrete subcategories that capture distinct recurring manifestations observed in our dataset, which can help developers diagnose and address flaky behaviours. For *structure-related web visual flakiness*, we identify three primary subcategories:

- **DOM Structure Issues:** flakiness where DOM elements (i.e., structural components) are not reliably or fully available in the DOM. This includes elements being delayed, missing, removed, or incompletely rendered due to asynchronous behaviour or race conditions, as well as explicit structural changes (e.g., adding, removing, or reordering elements). These are existential problems independent of event execution – *Whether and how elements are rendered or present*.
- **DOM Layout Instability:** flakiness where the spatial arrangement of elements is inconsistent, leading to overlapping, misaligned, or shifted components. Such effects may be triggered by structural changes such as added, removed, or repositioned nodes. The key distinction from DOM Structure Issues is that the inconsistency appears in layout position rather than in element presence or rendering – *How structural elements are spatially positioned*.
- **Timing-sensitive Visual Triggers:** flakiness that occurs when visual behaviours or event handling are triggered too early or too late relative to when structural elements become ready in the DOM. Examples include situations where expected responses, such as focus, visibility, or transitions, fail because the element was inserted, updated, or removed at the wrong time relative to the event execution. In other words, the failure arises from a synchronisation problem between event execution and element readiness, and not about the presence or position within the DOM. – *When those elements become available relative to an event or trigger*.

For *style-related web visual flakiness*, we identified two primary subcate-

263 gories:

- 264 • **Style Interpretation Issues:** flakiness, where the visual appearance
265 of elements is inconsistent because style rules are interpreted or ap-
266 plied differently at runtime. Examples include variations across envi-
267 ronments (e.g., differences in CSS properties, such as flexbox and font),
268 as well as inconsistencies within the same environment (e.g., cascade or-
269 der, specificity conflicts, or dynamic overrides via inline style injection)
270 – *How styles are interpreted and applied.*
- 271 • **Style Resource Latency:** flakiness where the visual appearance changes
272 inconsistently due to delays in loading style-dependent resources. Ex-
273 amples include elements appearing in a fallback or unstyled form due
274 to yet-to-be-loaded style resources, such as fonts or stylesheets – *When*
275 *style-dependent resources are applied.*

276 2.2.4. LLM-assisted categorisation

277 To support consistent and scalable annotation, we employed a large lan-
278 guage model (gpt-4o-mini) to assist in labelling verified web visual flakiness
279 instances from open-source web projects. This LLM-assisted procedure was
280 applied across multiple analysis dimensions defined in this study, including
281 the categorisation of visual flakiness types based on the taxonomy introduced
282 in Section 2.2.3, as well as the categorisation of corresponding developer fix
283 strategies analysed in the later research question (RQ3).

284 For each instance, a structured description including commit metadata
285 and messages, code diffs, and, if available, pull request details or linked issue
286 reports, was processed using a predefined prompt to generate a preliminary
287 label. After preliminary labels were generated through LLM, two authors
288 manually reviewed all label assignments. The LLM-generated labels served
289 as an initial reference during the manual review process, helping structure the
290 inspection and support consistent interpretation of the category definitions
291 across the dataset.

292 The LLM’s confidence scores associated with each label assignment were
293 used to prioritise manual review effort. For label assignments with confi-
294 dence scores of 0.7 or higher, two authors performed a confirmatory check to
295 verify that the suggested category—whether a visual flakiness subcategory
296 or a fix strategy—aligned with the observed evidence in the code diffs and
297 commit messages associated with resolving the reported flaky behaviour. If

Prompt Template

// Definitions (provided to the model)

Flakiness Type Definitions. Descriptions of web visual flakiness types (three structure-related and two style-related), e.g.,

1. DOM structure issues: ...
2. ...

Fix Strategy Definitions. Descriptions of fix strategies, e.g.,

1. Manage element changes: ...
2. ...

// Task

You are categorising the visual flakiness type and the corresponding fix strategy of a web visual flakiness instance.

// Supporting Evidence

Consider code diffs and issue or pull request descriptions together with commit metadata (e.g., author, date, statistics), commit messages, PR state/labels/-commits/files, and issue labels/state as supporting evidence.

// Instructions

Decide the most likely single flakiness type and single fix strategy from the allowed lists below. Use EXACT strings from the allowed sets. If the evidence is sparse, select the closest plausible category and assign a lower confidence.

Allowed Flakiness Types: {...}

Allowed Fix Strategies: {...}

// Output Format

Return STRICT JSON with keys:

flakiness_type (string),
fix_strategy (string),
confidence (float 0–1),
rationale (string),
evidence (string).

Use only allowed values.

Figure 2: System prompt used for LLM-assisted categorisation. The system prompt provides type definitions of visual flakiness and fix strategy. It restricts label(type) selection to a predefined set (*Allowed*), and specifies the required structured output format. Statements prefixed with "//" are comments added for readability and are not part of the prompt text itself.

Prompt Template

// Instance-specific context (user prompt)

Context: {A structured description of the flakiness instance, including commit metadata, commit messages, code diffs, and, when available, titles and descriptions from related pull requests or linked issue reports.}

Figure 3: User prompt for flakiness instance-specific context. The user prompt provides structured contextual information for each flakiness instance, including commit metadata, code diffs, and, when available, related issue or pull request descriptions.

the confidence score was below 0.7, the authors jointly re-reviewed all relevant artefacts (e.g., code diffs, commit messages, and related pull requests), performing a more in-depth manual inspection and resolving any ambiguities through discussion until consensus was reached.

The final categorisation of each instance, across both visual flakiness types and fix strategies, was determined based on this combined LLM-assisted and manual review process. No additional visual flakiness subcategories emerged in the final categorisation results beyond those derived from the initial pattern analysis described in Section 2.2.3 during the categorisation; the resulting fix strategy assignments are used solely to support the analysis in RQ3 (Section 3.3), where the fix strategy categories are introduced and discussed.

Figures 2 and 3 illustrate the prompts used for the LLM-assisted categorisation. The system prompt provides natural-language definitions of visual flakiness types (sub-categories) and fix strategies, specifies the categorisation task and the types of supporting evidence to consider, lists the allowed labels for each dimension, and defines the required structured output format and labelling rules. The user prompt provides instance-specific context for each case, such as commit metadata, code diffs, and, when available, related issue or pull request descriptions.

As discussed in Section 2.2.2, our dataset also includes web visual flakiness cases from Chromium, for which commit-level artefacts required for LLM-assisted categorisation (e.g., code diffs and commit statistics) are not consistently available. Therefore, the LLM-assisted categorisation procedure above is not applied to the Chromium cases. Instead, Chromium cases are manually labelled by two authors through collaborative review. Each case is assigned to one of the existing web visual flakiness subcategories defined in our taxonomy and derived from the open-source web projects, based on issue

325 descriptions and discussion comments recorded in the issue tracker and, when
326 explicitly available, linked patches or code review references. The same two
327 authors jointly review each case and resolve disagreements through discussion
328 until consensus is reached. No new flakiness subcategories were introduced
329 beyond those identified from the open-source web projects.

330 2.2.5. Research Questions

331 This study aims to deepen the understanding of flaky behaviours in web
332 visuals by exploring the following research questions:

333 **RQ1:** *What is the prevalence of flaky behaviours in web visuals?*

334 To assess the prevalence and impact of visual-related flakiness in web
335 applications and their tests, we conduct three investigations from different
336 perspectives. First, we examine the prevalence of visual commits to establish
337 a baseline for how frequently visual components are modified in our target
338 projects. This provides initial context for evaluating the role of visual changes
339 in flaky behaviour. Next, we analyse the ratio of flaky commits to total com-
340 mits within visual changes and within non-visual ones (e.g., backend or API).
341 This step assesses whether visual commits are more likely to introduce flaky
342 behaviour than other types of code changes. Lastly, we calculate the propor-
343 tion of visual flaky commits among all flaky commits across projects. While
344 the previous step compares the relative likelihoods of flakiness between visual
345 flakiness and non-visual flakiness, this step evaluates the overall contribution
346 of visual flakiness to the unreliability of web projects. Note that Chromium
347 is excluded from this analysis, as its flakiness is tracked through issues, not
348 commits (Section 3.1).

349 **RQ2:** *What are the most common types of web visual flakiness, and how do
350 they manifest in practice?*

351 To answer this RQ, we analyse the distribution of visual flakiness across
352 two primary categories—structure-related and style-related—and their re-
353 spective concrete subtypes (three structure and two style). This two-level
354 categorisation allows hierarchical reasoning about the sources of visual insta-
355 bility, enabling both an overall understanding and a fine-grained examination
356 of individual patterns. In the following analysis, we further illustrate repre-
357 sentative cases for each subtype to show how these flaky behaviours manifest
358 in real-world projects, together with a quantitative analysis of their distri-
359 bution and trends.

360 **RQ3:** *What fix patterns can be employed to mitigate the flaky behaviour in*
 361 *web visuals?*

362 Flaky behaviours degrade test stability, complicate debugging, and thereby
 363 diminish user experience. While prior work has proposed general solutions
 364 for flakiness, web visual flakiness may require more targeted strategies. By
 365 analysing how developers address web visual flakiness through the changes
 366 applied to address it, we aim to provide practical insights into fix strategies
 367 for different flakiness types, helping developers mitigate them effectively.

368 3. Results

369 3.1. RQ1: Prevalence of Web Visual Flakiness

370 Figure 4 presents the prevalence of visual commits across the 31 open-
 371 source projects, measured as the ratio of visual commits to total commits
 372 in each project. Chromium is excluded from this analysis due to the un-
 373 availability of reliable commit-level data (Section 2.2.2). On average, visual
 374 commits constitute 11.7% of all commits in these studied projects. This
 375 finding suggests that changes to visual components account for a substantial
 376 portion of ongoing web development activities and may therefore contribute
 377 significantly to overall project instability, warranting dedicated inspection.

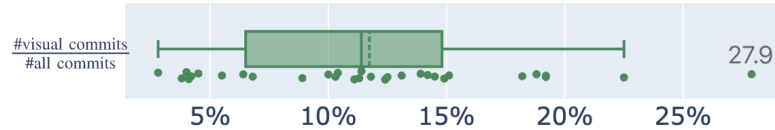
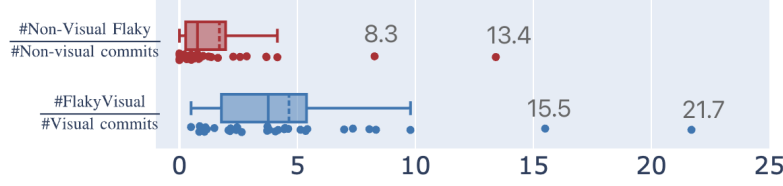


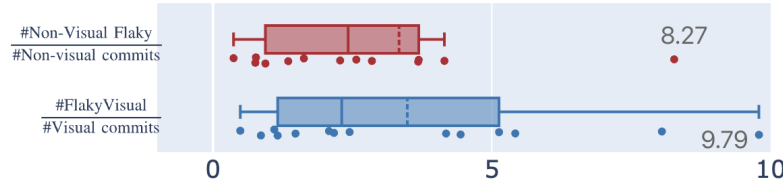
Figure 4: # visual commits to # all commits per project (%). The mean is 11.7%, and the median is 11.4%.

378 Next, we compare the frequency of flaky commits—i.e., the likelihood of
 379 commits exhibiting flakiness—between visual and non-visual commits. Fig-
 380 ure 5 presents the distribution of flaky commit ratios for each project (each
 381 dot represents one project) within the visual and the non-visual commits.

382 In this study, we employed a keyword-based search to collect relevant
 383 commits and issue reports, following prior flakiness studies Romano et al.
 384 (2021). Despite combining multiple metadata sources (commit messages, is-
 385 sue reports, and pull requests), this approach inevitably misses some cases
 386 due to incomplete or inconsistent textual descriptions—for instance, when



(a) $\#Flaky > 0$ (no threshold). For $flakyfreq_{visual}$, mean = 4.6, median = 3.7; for $flakyfreq_{non-visual}$, mean=1.7, median=0.8 (‰)



(b) $\#Flaky > 10$. For $flakyfreq_{visual}$, mean = 3.5, median = 2.3; for $flakyfreq_{non-visual}$, mean = 3.3, median = 2.4 (‰)

Figure 5: Flaky frequency. $flakyfreq_{visual} = \frac{\#FlakyVisual}{\#Visual\ commits}$ vs. $flakyfreq_{non-visual} = \frac{\#Non-Visual\ Flaky}{\#Non-visual\ commits}$. (‰) denotes commits per thousand.

387 developers use ambiguous language or omit flakiness-(or visual) related key-
 388 words. Consequently, projects often contained too few identified flaky com-
 389 mits (Table 1), which has also been observed in prior flakiness studies Ro-
 390 mano et al. (2021); Hashemi et al. (2022). As this could bias the comparison,
 391 we excluded projects with fewer than ten identified flaky commits from the
 392 thresholded analysis to ensure reliability, resulting in 14 out of 31 projects.
 393 Accordingly, Figure 3 presents two groups of boxplots: one that includes all
 394 projects (Figure 5b) and another that applies the threshold (Figure 5a).

395 Overall, visual commits exhibit a higher frequency of flakiness than non-
 396 visual commits. The frequencies shown are normalised per thousand commits
 397 (‰). In both analyses, visual commits are consistently more likely to involve
 398 flaky behaviour. The median flakiness frequency for visual commits is 3.7‰
 399 (2.3‰ under the threshold), compared to 0.8‰ (2.4‰ under the thresh-
 400 old) for non-visual commits. The corresponding mean frequencies are 4.6‰
 401 (3.5‰) and 1.7‰ (3.3‰), respectively. These results indicate that the vi-
 402 sual layers of web projects are often more prone to flaky behaviour than
 403 non-visual parts.

404 Having established that visual commits are both prevalent and more likely
 405 to exhibit flaky behaviour than non-visual ones, we lastly examined the over-
 406 all contribution of visual flakiness to all flaky cases across projects. Figure 6

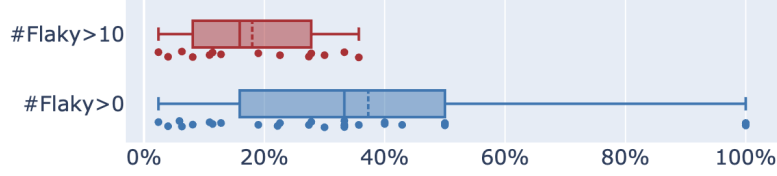


Figure 6: $\#FlakyVisual$ to $\#Flaky$ per project (%). $\#Flaky > 0$ and $\#Flaky > 10$ denote including only the projects with more than 0 (no threshold) and more than 10 flakiness instances. For $\#Flaky > 0$, mean = 37.2%, median = 33.3%; for $\#Flaky > 10$, mean = 17.9% and median=15.9%.

407 shows the distribution of the proportion of visual flaky commits among all
 408 flaky commits, with and without thresholding (i.e., including only projects
 409 with at least ten flaky commits). As shown in the lower boxplot of Figure 4,
 410 visual flakiness accounts for approximately 37.2% of all flaky cases, while in
 411 the thresholded subset (upper boxplot), 17.9% of all flaky commits are visu-
 412 ally related across the remaining 14 projects. Combined with the previously
 413 established prevalence and relatively high likelihood of visual commits being
 414 flaky, these observations highlight the need for a deeper investigation of the
 415 underlying characteristics of web visual flakiness.

Answer to RQ1. Visual commits, which account for an average of 11.7% of total commits, are often more prone to flakiness than non-visual ones. With and without the threshold of at least 10 flaky commits per project to be considered, visual flaky commits constitute a notable portion of all flaky cases (37.2% and 17.9%). These results combined underscore the inherently unstable nature of the visual layer of web projects and motivate a detailed examination of its underlying characteristics.

417 3.2. *RQ2: Common Types of Web Visual Flakiness*

418 Table 2 summarises the distribution of web visual flakiness in our dataset.
 419 Among the 262 analysed cases, 59.9% (157) are structure-related and 40.1%
 420 (105) are style-related. Similar trends appear in both Chromium (47.5%
 421 vs. 52.5%) and the 31 open-source web projects (70.1% vs. 29.9%). In
 422 the open-source projects, although style-related flakiness forms a smaller
 423 portion than in Chromium, it still accounts for nearly one-third of all visual
 424 flakiness, showing its practical importance. Overall, these results indicate
 425 that both structure- and style-related flakiness are major sources of web
 426 visual instability, with neither overwhelmingly dominating the other.

Table 2: Distributions of Web Visual Flakiness. The *Webs* refer to the open-source web projects. The values within the parentheses denote the percentage of flakiness instances in each type. The last column *Total* shows the combined statistics of *Webs* and *Chromium*.

Type	Subtype of Flaky Behaviour	# Flaky Cases (%)		
		Webs	Chromium	Total
Structure -related	DOM Structure Issues	34 (23.6%)	25 (21.2%)	59 (22.5%)
	DOM Layout Instability	17 (11.8%)	19 (16.1%)	36 (13.7%)
	Timing-sensitive Visual Triggers	50 (34.7%)	12 (10.2%)	62 (23.7%)
	All structured-related	101 (70.1%)	56 (47.5%)	157 (59.9%)
Style -related	Style Interpretation Issues	35 (24.3%)	45 (38.1%)	80 (30.5%)
	Style Resource Latency	8 (5.6%)	17 (14.4%)	25 (9.5%)
	All style-related	43 (29.9%)	62 (52.5%)	105 (40.1%)
All	-	144 (100%)	118 (100%)	262 (100%)

427 The following section presents the detailed distribution of structure- and
428 style-related flakiness and provides real-world examples illustrating how each
429 type of web visual flakiness manifests in practice.

430 3.2.1. *Structure-related Flakiness*

431 Within structure-related flakiness (the last column in Table 2), the dis-
432 tribution is largely consistent across Chromium and the Web projects, except
433 for Timing-sensitive Visual Triggers, which show the most notable variation—
434 34.7% in the Web set but only 10.2% in Chromium. This gap may reflect
435 the differences in execution environments. Web applications often rely on
436 asynchronous updates or client-side scripts that are more directly affected
437 by subtle variations in timing, whereas the rendering pipeline of Chromium
438 is more centralised and internally synchronised. In contrast, DOM Structural
439 Issues and DOM Layout Instability exhibit similar proportions across both
440 sets, indicating that these types of structural flakiness (i.e., the presence, ap-
441 pearance, and position of elements) occur at comparable rates in Chromium
442 and general web projects. The following examples illustrate how each type
443 of structure-related flakiness manifests in practice.

444 **DOM Structure Issues (22.5%)** DOM structure issues refer to cases
445 where the expected DOM hierarchy or elements are inconsistently constructed
446 across executions. This can happen due to structural disruptions (e.g., el-
447 ement insertions or removals) or timing-related factors (e.g., asynchronous

448 rendering or race conditions) that affect *whether* they are properly built.

```
1 <div class="mat-mdc-snack-bar-label" #label>
2   /* fix: added aria-hidden, aria-live */
3 -   <ng-template cdkPortal></ng-template>
4 +   <div aria-hidden="true">
5 +     <ng-template cdkPortal></ng-template>
449 6 +   </div>
7 +   <div [attr.aria-live]="_live"></div>
8 </div>
```

Listing 1: Displaying and using elements dynamically

450 Listing 1⁷ presents an example where rendering flakiness occurred in dy-
451 namically inserted content (i.e., a snack bar). The issue arose because the
452 DOM structure lacked a properly defined `aria-live` region, causing unsta-
453 ble update behaviour during content rendering. Lines 4-7 add a `div` with
454 `aria-hidden="true"` (Lines 4-6) and another `div` marked with `aria-live`
455 (Line 7), ensuring a stable and complete DOM structure that prevents pre-
456 mature/inconsistent updates during dynamic rendering.

457 **DOM Layout Instability (13.7%)** DOM layout instability concerns flak-
458 iness affecting *how* existing DOM elements are spatially positioned, rather
459 than whether they are present. Such flakiness manifests as inconsistent spa-
460 tial arrangements, leading to overlaps, misalignment, or displacement of ele-
461 ments.

```
1 /* fix: narrow viewport width flaky errors */
2 - <div #table style="margin: 16px">
3 + <div #table style="margin: 16px; max-width:
4   90vw; max-height: 90vh;">
5   <mat-table [dataSource]="dataSource">
6     <ng-container matColumnDef="before">
7     </ng-container>
8   </mat-table>
9 </div>
```

Listing 2: Container size makes the page structure unstable

⁷<https://github.com/angular/components/commit/cf8e8eee>

Listing 2⁸ shows a case of DOM layout instability due to uncontrolled expansion of a container `<div>`. The absence of maximum width and height constraints allowed the table to grow unpredictably, causing structural shifts such as overflow or misalignment. The fix directly modifies the DOM by adding bounding attributes, which constrain spatial behaviour and stabilise rendering. Though expressed via style-like properties, the impact is structural, preventing disordered layouts rather than adjusting visual appearance.

Timing-sensitive Visual Triggers (23.7%) Timing-sensitive visual triggers refer to flakiness that manifests due to variation in *when* the visual state or DOM becomes observable or stable, rather than *whether* it appears at all. Such flakiness arise from timing mismatches between test actions and the completion of UI rendering or state updates.

```
1   <input #textInput type="text" [value]="text"
2     (focus)="onAcceptEdit()"
3     (blur)="onAcceptEdit()"
4   /input>
5   export class TextEditorComponent {
6     @ViewChild("textInput")
7     ngAfterViewInit() {
475    /* fix: add delay to ensure the state */
9   +   window.setTimeout(() => {
10      this.textInput.nativeElement.focus();
11   +   }, 5);
12  }}
```

Listing 3: Unexpected element focus behaviour

In Listing 3⁹, DOM Structure flakiness occurs due to timing mismatches in when the input field becomes available for interaction. The input field is expected to receive focus when triggered, as indicated by the focus operation ("onAcceptEdit") (Line 2). However, since rendering is not always synchronised with when elements become fully available, focus may be applied too early. This can cause the input field to fail to receive focus, preventing immediate text input intermittently.

⁸<https://github.com/angular/components/commit/d2b499da>

⁹<https://github.com/angular/components/commit/4ae63b39>

483 3.2.2. *Style-related Flakiness*

484 For style-related flakiness, Style Interpretation Issues are the most preva-
485 lent type in both Chromium (38.1%) and the Web projects (24.3%) (Ta-
486 ble 2). Style Resource Latency appears less frequently, though the magni-
487 tude of difference varies: it accounts for 14.4% of style-related flakiness in
488 Chromium but only 5.6% in the Web projects. This difference likely stems
489 from Chromium’s heavy reliance on diverse and externally hosted style re-
490 sources, which increases the likelihood of delayed or inconsistent loading,
491 potentially leading to more visual instability.

492 **Style Interpretation Issues (30.5%)** This type of flakiness often arises
493 when CSS properties or style interactions are handled inconsistently across
494 rendering environments. Even minor differences in CSS or interactions be-
495 tween style rules—such as differences in the evaluation of properties like
496 display or text-indent between platforms or browsers —can result in visually
497 different outcomes across executions. Such inconsistencies make it challeng-
498 ing for developers to ensure a stable and uniform visual presentation in web
499 applications.

```
1  <style>
2      input::-webkit-input-placeholder, textarea
3          ::-webkit-input-placeholder {
4          color: #aaaaaa;
5          text-indent: 0;}
6      /* fix: add rule improve compatibility */
500 6 +  textarea::-webkit-input-placeholder {
7 +      color: #aaaaaa;
8 +      text-indent: -3px;}
9  </style>
```

Listing 4: *text-indent* properties cause flakiness behaviour

501 Listing 4¹⁰ (i.e., style codes) presents an example of style-related flaki-
502 ness caused by CSS attribute compatibility issues. The placeholder text and
503 cursor become misaligned when the text area is focused due to variations
504 in how browsers interpret the text-indent property for placeholders. This
505 inconsistency arises from differences in CSS rule enforcement across render-
506 ing engines. To improve compatibility, the fix explicitly defines *text-indent*

¹⁰<https://github.com/ionic-team/ionic-framework/commit/76ca4757>

507 for `textarea::-webkit-input-placeholder`, ensuring more consistent be-
508 haviour across browsers.

509 **Style Resource Latency (9.5%)** Style Resource Latency occurs when es-
510 sential assets — such as fonts, images, or external style sheets — are not
511 fully loaded before the visual rendering is complete. Such delays can lead to
512 temporary or inconsistent appearance, as the web page may be incorrectly
513 styled or displayed without its intended formatting.

```
1  main_resource.Write(HTML(  
2    <!doctype html>  
3    <head>  
4      /* fix:  cache aware font Loading */  
5    <link rel="preload" as="font" type="font/  
514    woff2"  
6      href="https://example.com/font.woff  
    ">))
```

Listing 5: Font loading contributes to flakiness

515 Listing 5¹¹ illustrates a case where flakiness arises due to delayed font
516 loading (Line 6). If custom fonts take too long to load or fail to load,
517 browsers fall back to default fonts, which may differ in size or spacing, causing
518 misaligned or invisible elements. This mismatch between the expected and
519 the actual leads to visual inconsistencies. Developers fixed this issue using
520 preload hints to ensure timely font availability and prevent such flakiness.

521 Overall, Chromium shows a more even distribution of structure- and style-
522 related flakiness compared to the Web projects. We suspect that the key con-
523 tributor is the smaller proportion of Timing-sensitive Visual Triggers flaki-
524 ness, possibly reflecting Chromium’s more synchronised event scheduling. At
525 the same time, its complex, multi-threaded rendering pipeline may introduce
526 additional nondeterminism during layout computation and resource loading,
527 contributing to the slightly higher proportions of DOM Layout Instability
528 and Style Resource Latency observed. DOM construction remains largely
529 deterministic within a single engine build. Hence, the modest increase in
530 Style Interpretation Issues may instead reflect the broader redistribution of
531 flaky cases from the decrease in timing-sensitive flakiness, rather than inher-
532 ent nondeterminism in the style engine itself.

¹¹<https://issues.chromium.org/issues/40114104>

533 **Flaky behaviours across different front-end frameworks.**

534 Modern web development increasingly relies on front-end frameworks. In
535 this empirical study, we additionally examined whether the choice of frame-
536 work influences the distribution of visual flakiness types. Among 31 web
537 projects, 87 of 144 cases (60.6%) involved frameworks, such as React, Vue,
538 and Angular¹². As shown in Figure 7, similar patterns emerge across React,
539 Vue, and Angular, suggesting that the underlying framework does not have
540 much influence on the manifestation of web visual flakiness.

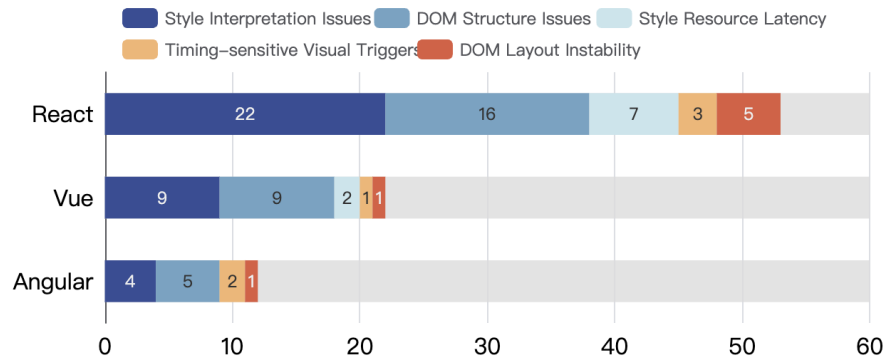


Figure 7: The distribution of flakiness across UI frameworks.

¹²The remaining 39.4% of projects do not use front-end frameworks; some were developed before frameworks became widely adopted, while others are lightweight applications or static websites that avoid the extra complexity that frameworks introduce.

Answer to RQ2. The analysis shows that structure-related and style-related types together account for nearly equal portions of web visual flakiness. Among the structure-related types, Timing-sensitive visual triggers are the most frequent, whereas Style Interpretation Issues are the most common among style-related ones, followed by Style Resource Latency. Overall, both Chromium and the Web projects exhibit similar distributions of visual flakiness types. However, Chromium exhibits a lower proportion of Timing-sensitive visual triggers and a comparatively higher share of style-related flakiness. The smaller ratio of timing-sensitive triggers is likely attributed to more synchronised and centralised rendering and event pipeline of Chromium, reducing timing variability across executions. In contrast, the relatively higher ratio of style-related flakiness might partly result from fewer structure-related cases overall, rather than an inherent increase in style-level instability.

541

542 3.3. *RQ3: Common Flakiness Fix Patterns of Developers*

543 Based on the categorisation process described in Section 2.2.4, we identi-
544 fied seven categories of recurring fix strategies developers use to address web
545 visual flakiness.

- 546 • **Manage element changes:** insert, remove, or update DOM elements
547 to prevent inconsistencies in element presence or structure.
- 548 • **Delay:** introduce a conservative timing buffer to handle variability
549 in dynamic rendering or state transitions, ensuring that subsequent
550 operations are not processed prematurely.
- 551 • **Explicit size position:** define layout dimensions (e.g., width, height,
552 margins) to avoid instability caused by flexible or implicit sizing.
- 553 • **Use grid/flexbox:** employ structured layout systems such as CSS
554 Grid or Flexbox to improve spatial consistency and reduce layout mis-
555 alignments.
- 556 • **CSS compatibility:** modify CSS rules to ensure consistent behaviour
557 across different browsers and rendering engines by applying baseline
558 styles that override browser-specific defaults to standardise the initial
559 rendering behaviour across environments
- 560 • **Element-event coordination:** align target elements and triggering
561 events in timing and state to prevent missed or inconsistent interac-
562 tions.

- **Preload & Cache:** ensure key resources (e.g., fonts, images) are available when needed by preloading them and leveraging caching to prevent delays or rendering failures.

We further analysed how these fix strategies are distributed across different types of web visual flakiness to understand which strategies developers often employ to handle structure- and style-related flakiness issues, as shown in Table 3. Overall, developers tend to employ fix strategies that closely correspond to how each type of web visual flakiness manifests. For instance, in structure-related flakiness, such as DOM Structure issues, developers most frequently applied Element–event coordination, used in 23 of 34 cases, to stabilise interactions between elements and events. In Timing-sensitive Visual Triggers, Delay and Element–Event Coordination were the two dominant strategies, both addressing asynchronous rendering and event ordering issues. For DOM Layout Instability, developers often constrain layout flexibility to prevent unintended spatial shifts by adopting the Explicit size position strategy. For the Style Interpretation, nearly all fixes (33 of 35) were about CSS Compatibility, addressing inconsistent rendering semantics through CSS adjustments. Lastly, for Style Resource Latency, timing- and loading-related strategies, such as Delay and Preload & Cache, were most frequently applied to ensure stable resource availability and rendering.

At the same time, we observed that specific fix strategies are not strictly confined to a single type of web visual flakiness. For instance, CSS Compatibility was applied across all five categories of flakiness, and Delay was utilised in DOM Layout Instability, Timing-sensitive Visual Triggers, Style Interpretation Issues, and Style Resource Latency. Element-event coordination, the most adopted strategy in DOM Structure Issues, was also the most common strategy among all fixes for structure-related flakiness, appearing in 38 of 144 cases (26.4%). We conjecture that this broad applicability stems from the general role these strategies play in stabilising cross-cutting sources of non-determinism. For instance, modifying CSS rules for CSS Compatibility could mitigate rendering inconsistencies that manifest across structure and style components. Similarly, the Delay strategy reduces premature or mismatched interactions by introducing timing buffers, while Element–event coordination can ensure the synchronisation of elements and their triggering events, thus generally preventing state mismatches in timing-sensitive scenarios.

Developer priorities for different types.

Table 3: Repair strategies for 144 cases of web projects’ visual flakiness. Overall, developers typically apply fix strategies aligned with specific visual flakiness types (e.g., *Element–Event Coordination* for *DOM Structure Issues*, and *Delays* for *Timing-sensitive Visual Triggers*), while some strategies—such as *CSS compatibility*—are used across both structure- and style-related flakiness.

Type	Cause	Fix method	# Flaky Cases (%) Webs
Structure -related	DOM Structure Issues	CSS compatibility	3 (2.1%)
		Manage element changes	8 (5.6%)
		Element-event coordination	23 (16.0%)
	DOM Layout Instability	CSS compatibility	2 (1.4%)
		Delay	2 (1.4%)
		Manage element changes	1 (0.7%)
		Element-event coordination	1 (0.7%)
		Explicit size position	5 (3.5%)
		Use grid/flexbox	6 (4.2%)
	Timing-sensitive Visual Triggers	CSS compatibility	3 (2.1%)
		Delay	26 (18.1%)
		Element–event coordination	14 (9.7%)
		Manage element changes	7 (4.9%)
Style -related	Style Interpretation Issues	CSS compatibility	33 (22.9%)
		Delay	1 (0.7%)
		Manage element changes	1 (0.7%)
	Style Resource Latency	CSS compatibility	1 (0.7%)
		Delay	2 (1.4%)
		Preload & Cache	4 (2.8%)
		Manage element changes	1 (0.7%)

599 We further examined how developers prioritise addressing different types
600 of web visual flakiness using the Chromium dataset, where each issue is as-
601 signed an explicit priority level. Figure 8 shows the distribution of these
602 priorities, P1 (high), P2 (medium), and P3 (low), across structure-related
603 and style-related categories of web visual flakiness. Specifically, the first
604 and third subfigures in Figure 5 illustrate priority distributions within style-
605 related and structure-related flakiness, respectively. Overall, while a substan-

606 tial proportion of both structure-related (44.6%) and style-related (59.7%)
607 issues were assigned medium priority (P2), the structure-related issues were
608 more frequently labelled as low priority (P3) than the style-related (41.1%
609 vs. 24.2%). This implies that developers may allocate relatively more at-
610 tention to style-related flakiness in Chromium. A likely explanation is that
611 style-related flakiness often produces visually broader or more noticeable dis-
612 ruptions, such as broken layouts or inconsistent rendering, compared to the
613 structure-related ones that may remain localised to specific elements. In fact,
614 our previous analysis on the distribution of web visual flakiness shows that
615 style-related flakiness represents a larger overall share than in the open-source
616 web projects.

617 At a finer granularity, DOM Layout Instability exhibited the highest pro-
618 portion of high-priority (P1) assignments (21.2%), followed by Style Resource
619 Latency (17.6%) and Style Interpretation Issues (15.6%). These results fur-
620 ther support the above conjectures, suggesting that developers often priori-
621 tise unstable behaviours with broad visual impact, especially those that affect
622 page layout or rendering consistency. Timing-Sensitive Visual Triggers were
623 mostly given medium priority (83.3%), indicating a medium level of urgency.
624 DOM Structure Issues had the highest proportion of low-priority (P3) as-
625 signments (56%), suggesting that element-specific inconsistencies are often
626 viewed as less critical to the overall user experience. This further explains
627 the higher proportion of low priority in structure-related flakiness compared
628 to style-related flakiness.

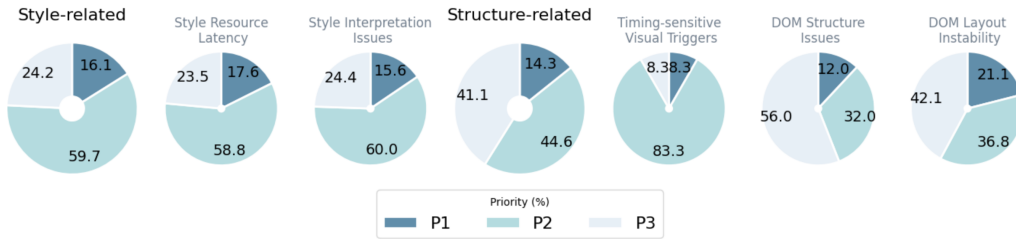


Figure 8: The distribution of developers addressing priorities (P1 (high), P2 (medium), and P3 (low)) under different causes, the proportions are in percentages (%).

Answer to RQ3. While developers tend to choose fix strategies that align closely with the specific types of web visual flakiness, some fix strategies appear across multiple types, reflecting their broad applicability. Additional analysis of issue priorities shows that developers tend to prioritise flakiness with broader visual impact, such as layout or style-related problems, over those with narrower or isolated effects.

629

630 4. Discussion

631 4.1. Positioning web visual flakiness against general UI issues and previous 632 Web flakiness studies

633 Unlike general web visual issues discussed in prior studies Althomali et al.
634 (2021); Alameer and Halfond (2016); Mahajan et al. (2016); Romano et al.
635 (2021), which are typically deterministic or consistently reproducible, web
636 visual flakiness is inherently intermittent and non-deterministic. It often
637 stems from asynchronous rendering, transient DOM states, or timing in-
638 consistencies, and manifests across the structure and style dimensions that
639 we categorised in this study. Such characteristics of these issues complicate
640 their detection and resolution, as they depend on subtle runtime interactions
641 rather than static or deterministic failures. Prior studies on Web flakiness
642 have primarily examined broad or functional causes (e.g., network latency,
643 environment configuration) or deterministic rendering faults Romano et al.
644 (2021). In contrast, our work isolates and empirically characterises flaky be-
645 haviours that specifically originate in the visual layer, i.e., those directly tied
646 to the presentation logic of web structures and styles, thereby addressing a
647 class of issues that has often been overlooked in prior work.

648 5. Threats to Validity

649 5.1. Internal validity

650 While our categorisations of flaky behaviours were performed through it-
651 erative discussion and cross-validation among the two authors, along with
652 LLM assistance, the process is inherently affected by subjective judgment.
653 Although we ensured consistency through shared criteria and multiple passes
654 over the data, potential bias or misclassification may remain. Flakiness
655 in web visuals can stem from subtle timing issues, rendering inconsisten-
656 cies, style misinterpretations, etc. While we applied multiple-round manual

657 checks, some edge cases may still have been misplaced. In addition, our cat-
658 egorisation may not capture all manifestations in diverse contexts. Future
659 work could refine and expand these criteria for a more precise.

660 5.2. *Construct validity*

661 Construct validity threats primarily stem from the use of keyword-based
662 filtering of commit messages, issue reports, and pull requests to identify vi-
663 sual commits and web visual flakiness. Keyword matching may introduce
664 misclassification, including false negatives when visual or flaky behaviour is
665 only implicitly described, as well as false positives when keywords do not
666 correspond to actual visual or flaky behaviour.

667 To mitigate this, we adopted a conservative sampling strategy focused on
668 precision. Specifically, we focused only on the interaction of flaky and visual
669 keywords, manually validating each candidate case. Two authors jointly in-
670 spected code changes and artifacts to confirm the presence of visual flakiness,
671 and thereby further eliminated false positives, increasing confidence in the
672 final dataset.

673 As this conservative choice likely results in false negatives, the reported
674 prevalence should be interpreted as a lower bound. However, since our cate-
675 gorisation and fix-pattern analyses focus on recurring behavioural character-
676 istics observed across consistently validated cases, rather than on complete
677 enumeration, this limitation is less likely to affect the qualitative findings.

678 5.3. *External validity*

679 A potential limitation of our study is the generalisability of the results
680 beyond the specific projects and test suites analysed. Although our find-
681 ings are rooted in web projects and Chromium, they may also be relevant to
682 other contexts, such as mobile applications and hybrid frameworks. Mobile
683 browsers, which exhibit a variety in hardware setups and screen sizes, are
684 also prone to similar structural and style inconsistencies, especially when it
685 comes to responsive development. Despite differences in technical implemen-
686 tation, manifestations of flakiness, like layout shifts and inconsistent style
687 application, remain conceptually aligned. This suggests that our categorisa-
688 tions and observed fix patterns may help inform broader efforts to address
689 visual flakiness across diverse platforms.

6. Related work

Flaky tests have been the subject of extensive research in the software testing community, particularly in programming languages, such as Java, Python, and JavaScript Luo et al. (2014); Lam et al. (2020a,b); Gruber et al. (2021); Hashemi et al. (2022); Leong et al. (2019). Numerous studies have examined the prevalence, causes, and mitigation strategies for flaky tests in code-based systems Lam et al. (2020a); Pinto et al. (2020); Dutta et al. (2020). For example, Luo et al. conducted a pioneering study on flaky tests, analysing large-scale systems to identify the underlying causes of test flakiness Luo et al. (2014). While these findings are valuable for traditional software, they do not directly address the challenges posed by style-related issues, where visual rendering discrepancies could also lead to flakiness.

For research on flaky tests in web environments, Hashemi et al. analysed flaky behaviour in JavaScript frameworks, identifying browser-related issues as significant contributors to flakiness testing Hashemi et al. (2022). Romano et al. analysed flaky UI tests from 62 web and Android projects and identified four categories of causes for flaky UI tests Romano et al. (2021). Moran et al. focused on identifying the root cause of flakiness in web applications by analysing the execution of tests under diverse combinations of environmental factors that may contribute to flakiness Morán et al. (2020). Pei et al. investigated the async await flakiness in web tests, focusing on the DOM and time issues Pei et al. (2024). However, these studies rarely isolate style-related code as a distinct source of flaky behaviour, although CSS is highly susceptible to the inconsistencies that lead to flakiness. Our study aims to fill this gap by specifically investigating flaky behaviours in CSS, providing a more nuanced understanding of the problem in web development.

Besides empirical studies on flaky tests, there has been considerable work on mitigation strategies Ziftci and Cavalcanti (2020); Camara et al. (2021); Haben et al. (2021); Rahman and Shi (2024); Fallahzadeh and Rigby (2022). Bell et al. leveraged code evolution and code coverage to determine whether new test failures between two commits are due to flaky tests Bell et al. (2018). There have also been several tools that have been targeted to detect or repair certain types of flaky tests, such as iDFlakies Lam et al. (2019b), iFixFlakies Shi et al. (2019), FLASH Dutta et al. (2020), Flakify Fatima et al. (2022), Flakeflagger Alshammari et al. (2021), and so on Dutta et al. (2021); Cordy et al. (2022); Cai et al. (2024).

While there is a growing of research on flaky tests in software systems,

727 the focus on web visual-related flakiness is minimal. Our research builds on
728 previous studies of flaky tests by extending the analysis to the web visuals.
729 By focusing specifically on flaky behaviour in style and visual layers, this
730 study aims to bridge the gap in current literature and provide actionable
731 insights for developers dealing with flaky behaviours.

732 7. Conclusion

733 This paper presents a focused empirical study of web visual flakiness
734 as a distinct and explicitly characterised class of flakiness, based on 262
735 cases collected from 31 open-source projects (144 cases) and the Chromium
736 project (118 cases), revealing its recurring presence across diverse mod-
737 ern web projects. We found that visual commits, comprising 11.4% of all
738 commits, exhibit a higher frequency of flakiness than non-visual commits.
739 We categorised web visual flakiness into two primary dimensions—structure-
740 related and style-related—, which we further refined into three and two con-
741 crete subtypes, respectively. Through an inductive analysis of developers’
742 fixes, we found seven recurring fix strategies, which address common insta-
743 bility patterns, such as element-event coordination, timing, layout stabili-
744 sation, and style consistency. Although some are closely associated with
745 specific flakiness types, others span across multiple types, showing the inter-
746 twined nature of structure and style aspects in rendering. Together, these
747 findings highlight the need to treat web visual flakiness as a distinct subject
748 in web development and testing. To facilitate further research and repro-
749 ducibility, we make our dataset and replication package publicly available at:
750 <https://doi.org/10.5281/zenodo.17324581>

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