Memoization of Browser Computations for a Faster Mobile Web

Ayush Goel Matthew Furlong Hyun Jong Lee

1 Introduction

Page Load Time (PLT) of a website is a key performance metric that significantly impacts user-experience, as pointed out by many recent studies from both academia and industry [2, 3]. User-experience in web browsing is directly correlated to companies' revenues: Amazon shows that reducing 100ms in PLT results in 1 percent revenue increase and Shopzilla reports that improving PLT from 6 to 1.2 seconds increased their revenue by 12 percent [8].

There have been many prior work in improving the PLT of mobile devices, ranging from offloading computation and network tasks to proxies [13, 18, 22] to re-prioritising requests at client side by letting the client itself discover all resources on a page [4, 12]. It is worthwhile to note that existing works attempt to surrogate web-tasks of mobile devices from resource-rich server environment [15].

The end-to-end PLT for many webpages is far from ideal: an order of tens of seconds on mobile devices [21] and on the order of seconds for stationary desktops. Many existing solutions often require server-side modification, which strongly discourages content providers to use these new solutions. However, a client side solution would be agnostic of the content provider/server that is used to render the web pages and can therefore optimize page load times for all web pages alike. Most of the existing work on client side optimiztions focuses on efficient ways to optimise web cache [23]. Prior work shows that computation latency is the driving factor behind slow page load time for mobile devices, as compared to newtork latency. [20].

In this work, we propose a novel PLT optimization technique that caches output from previous code execution of a webpage (e.g., Javascript, inline HTML, css) on mobile devices to reduce user-perceived PLT at the small cost of increased storage requirements. Prior techniques have gone as far as caching the compiled code, either on the client side or on the server side, to save on the compilation time when the web content remains unchanged [23]. We take this a step further, and cache the output of the execution of all the code on a web page. (Note the use the word code, to distinguish it from other components of a webpage which include layout and data). Recent work has shown that most of the webpages remain unchanged over a large period of time. For content-rich pages, the amount of updates vary across Web pages. In the best (worst) case, 20% (75%) of the HTML page is changed over a month. Most changes are made to data (e.g., links to images, titles) while little change is made to the layout and code [23]. This implies that most of the code output could be reused, essentially eliminating code execution time from the critical path of a web page load. This would bring down the entire page load time to the time taken in rendering and painting the layout. Caching the computation as a technique to optimize the execution time has already been explored at a data center level [7] and it has shown tremendous improvement with more than 35% of jobs benefiting from caching. We are trying to apply a similar technique on the mobile client's browser.

2 Motivation

Major web browsers like Chrome, Firefox, and Safari have recenlty invested a lot of resources, time and energy into improving web performance on mobile devices, specifically by targeting the network usage. However, the network now comprises less than 30% [11] of the total critical path for an average page load on a mobile device. This includes caching almost 95% of the resources that are fetched from the server [20], dns presolution, dns caching, tcp reconnect etc. Chrome released a paper last year showing how improved caching algorithms, despite having significant improvements on the desktop, don't have the same proportionate imrpovements on mobile devices. This is primarily attributed to the fact that computation comprises more than 65% of the critical path during a page load. This illustrates the need to further optimize the compuatation time.

During the Chrome dev summit this year, their team announced the latest improvements they have made in their browser to improve the page load time. Interstingly, most of their work focuses on improving the compilation and parsing time by introducing compile and parser cache. Recent studies [16] still report that the median page load time for a mobile website is about 14 seconds. Research [16] shows that a user will only wait for 3 seconds before abandoning a web site if it shows no response at all. A lot of prior work [11] has been done to compare the page load times on mobile vs desktop, and

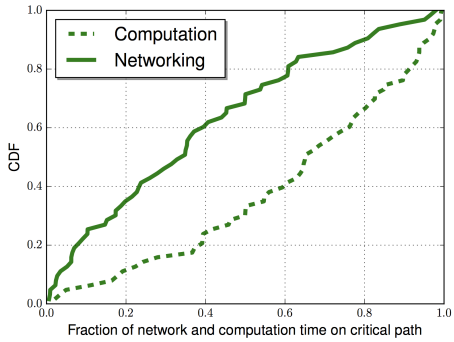

Figure 1: **Runtime information on mobile devices**

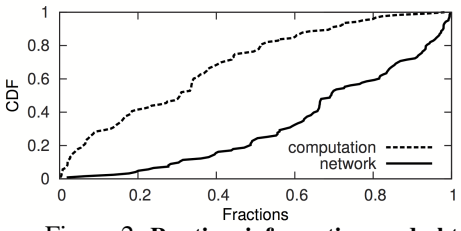

Figure 2: **Runtime information on desktops**

recent results from 2016 claim that despite the increasing compute resources in mobile devices, the computation time on mobile is significanlty higher than their desktop counterparts. Our experiments on the most popular news and sports websites on the latest mobile hardware and the latest Chrome verison reveal that despite these recent efforts, scripting still takes significantly more time ,as compared to the other components of the total computation time. With our understanding of computation being the current bottleneck for high page load times, we conduct a set of experiments to evaluate the reasons for this exceedingly high computation time. In order to do this, We break down computation into four categories: scripting, loading, paitnint, and rendering. and observe that scripting essentially takes more than 70% of the total computation time which is more than all the other categories combined (Figure 3). This makes it all the more important to do an in-depth analysis of the computation time to clearly understand where exactly this time is being spent. Using results from this study, we would eventually design our caching framework for the javascript execution output.

3 Design

We propose a new technique to improve the page load time by reducing the javscript time, specifically the execution and script evaluation time. In order to do this, we intend to build a new caching framework for the modern web browsers, specifically Google Chrome. We choose to focus on Chrome because it accounts for about 50% of the market share in terms of browser usage. Our caching framework will store the javascript execution result. This can mean a lot of things due to the dynamic nature javascript. Most of the times it is simply the return value of the javascript functions. At other times it can be a modified DOM structure or just some intermediate result which will be further processed by other javascript functions, later down the execution timeline. We will talk about the exact definition of javascript execution output in section 3.1 The expiry of this javascript exectution cache is supposed to be same as the expiry of the javascript source cache. This essentially is derived from the x-cache header field in the response, which determines how long the javascript source would reside in the browser cache. There are a lot of caveats to this approach, in the sense coming up with an optimal data structure to hold the javascipt execution cache, handling non determinism of the javascript code, and in our work we will explore all of these. The biggest challenge when developing a new caching framework is modifying the current browser's code in order to evaluate the efficacy of our caching framework. Since a lot of browsers already implement caching at the javascript runtime level, such as compiler and parser caches, a lot of this architecure can be borrowed for the execution cache as well. Another potential challenge will be the memory overhead. Most of the popular websites which spend about 70% of their time on javascript execution are running 1000s of javascript functions. Saving the output of all of these functions will add extra memory overhead to current browsers.

3.1 Javascript Execution Output

We have come up with a definition of the javascript execution output, which we will use in order to build our caching framework. Currently we are working on two different granularities to capture the javascript execution output. The key idea behind capturing a javascript's execution output is to capture the global changes made to the environment as a result of the javascript execution. This eliminates the need to capture any local computation, intermediate results computed and any output in general which doesn't affect the global state of the browser. We define the global state of the browser to be represented by the window object, which is an instance of the open window/tab of the browser. This window object is supported by all the major browsers like Chrome, Firefox, Safari. If the execution of any javascript doesn't modify the window object in any way, then essentially that javascript has no affect to the global browser window state and therefore has null impact. Such a javascript can be done away with, as long as any other global object which has used results from this javascript's local computation was cached. This is the key idea behind our caching framework.

The modification to the global window object can be either to the current properties of the window object, which were modified by the javacript, or in form of creation of new properties. Any global variable and function that is defined by the javascript becomes a new property of the window object.

The two granuarities at which we'll capture the javascript execution output are:

1. Javascript file

2. Javascript function

The first point above is a higher granularity capture. To cater to this, we capture the entire window object state before the page load begins, and after the page has been loaded. Then we do a basic diff analysis of these two window states. This difference is essentially the global effects of that specific javascript. This helps us understand how much caching of the javascript execution output would ultimately impact the page loading time. To the best of our understanding, if 95% properties of the global window object remain unchanged upon execution of a particular javascript, then we have a theoretical upper bound of 95% that we can achieve as far as reduction in javascript execution time is concerned, by employing our caching framework. We conducted a couple of experiments to understand how much change is made to the window object when the page is loaded after 3s, 3 hours and 3 days. To our suprise, the change in the window object is extremely insignificant encouraging us to expect high impacts of our caching framework.

However javascript is rarely executed/evaluated as a file as a whole. Therefore to better understand the computation effects of a javascript, we evaluate the execution output at the second granularity, at a function level. This will give us a much better understanding of the nature of the javascript execution, as javascript code is usually invoked one function at a time. The approach here, is to create a call graph of the javascript execution, during a page load. Each node in the graph represents a function that was actually invoked (note that is this a dynamic call graph, built during the actual page load). Each node contains a signature for the function that was executed. The signature contains the function name, arguments, any global variables delcared, initialized or computed and the return value. We use this signature to compare graphs across two different loads to establish how much of the execution can be cached.

Caching of javascript execution is to be eventually implemented at a function level. The results from the above two studies help us in defining the upper bounds of the benefits of our approach. We are yet to implement the actual caching framework and evaluate the actual benefits of the approach.

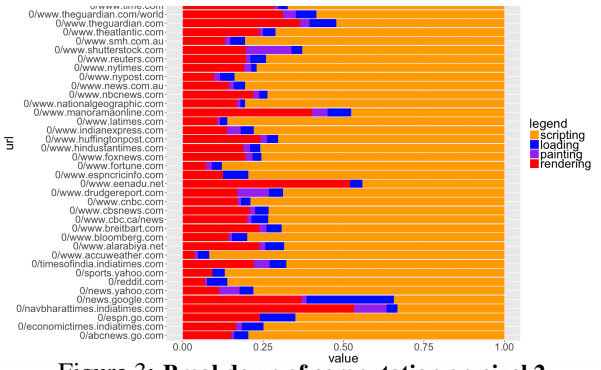

Figure 3: **Breakdown of computation on pixel 2**

4 Progress

As a preliminary step in this direction, we first established a corpus of the top 75 news and sports websites on which we would run our experiments in order to cater to the most popular and compute intensive websites. These websites are taken from the Alexa top website list. We ran all our experiments on Google Pixel 2. We used Chrome v 61 to run all of our experiments on the mobile device. We leveraged chrome developer tools in order to capture the runtime traces, both for network and compute. We then analyzed these runtime traces to draw insight into the critical path of the website, the total time being spent on the compute vs the total time being spent on the network, and most importantly the finer level breakdown of the computation time to understand the true nature of computation on mobile devices.

We categorised computation time into four categorires, scripting, loading, rendering and painting. Scripting is the total time being spent on compiling, evaluating and executing javascript. Loading consists of parsing the html and css, which happens immediately after the payload for the network requests are received by the browser. Loading takes these payload objects and parses them before converting them to a DOM tree. Once the DOM tree is built, the rendering engine converts this DOM tree into a render tree, which contains the exact coordinates and the shape of each of the DOM node. This process comrpises the rendering time of the web page. Painting time is the time taken to process the render tree, and convert each pixel into a bitmap. Figure 3 shows the computation break down for these first level of categories for the Google Pixel 2. We further break down this time into the finer level events which are returned by Google Chrome's trace and then group them by their event name.

These results are extremely coherent with our design of implementing a javsvcript chrome caching mechanism and show the possibility of a large improvement in the total page load due to the high percentage share of execution time as seen in Figure 4.

Recently in their 2017 dev summit, the Chrome team

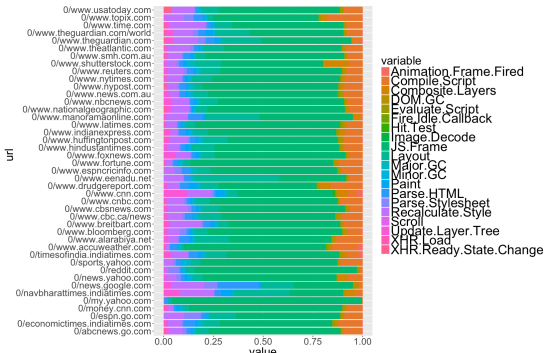

Figure 4: **Breakdown of computation into finer events on pixel 2**

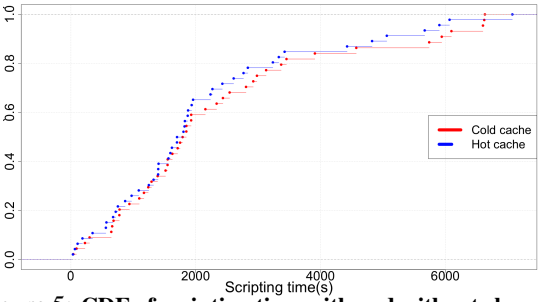

Figure 5: **CDF of scripting time with and without chrome's optimisations**

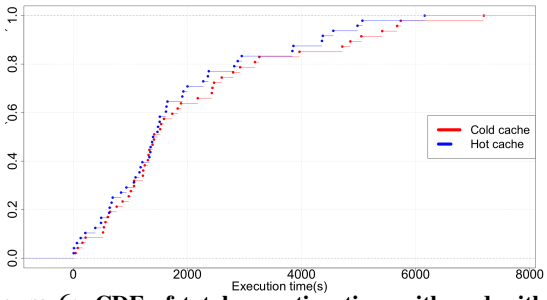

Figure 6: **CDF of total execution time with and without chrome's optimisations**

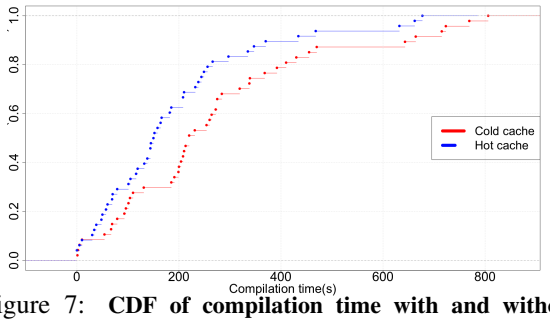

Figure 7: **CDF of compilation time with and without chrome's optimisations**

discussed the various optimisation techniques they have employed in the latest Chrome browser to improve the total page load time. We did a comparison of the total page load time with and without Chrome's optimizations to see the improvements. In order to do this, we captured the trace from Alexa's top 75 news and sports website once with a fresh cache, ie cold cache, and then subsequently with a hot cache which contains all of Chrome's optimizations, including its compiler and parser cache. As we can see from Figure 7, there has been a significant reduction in the overall compilation time, with about 100ms reduction in median compile time. This is primarily due to the introduction of compiler and parser cache. The line corresponding to Cold cache refers to the fresh load of all the websites, whereas the line corresponding to the Hot cache refers to the subsequent load which makes use of Chrome's caching framework. This is also reflected partially in the overall scripting time as shown in Figure 5. Note that scripting time is the sum of compilation, execution and other minor javascript events in the execution pipeline like garbage collection. However the interesting thing to note, is that despite all these optimizations, we observe almost neglible improvement in the median execution time of the javascript, as shown in Figure 6. This serves as a huge motivation for the vast potential in improving the overall page load time, by optimizing the javascript execution time.

After establishing the impact of a javascript execution caching framework, we conducted experiments to understand the nature of the javascript computation. We have explained the results from these experiments in section 3.1. We observed that most of the properties of the global window object remain unchanged. For a time difference of three seconds, we observe only a 2.5% of properties to be have changed. This is kind of expected as not a lot is suppose to change within three seconds between two web page loads. However interestingly, even for a gap of three hours between two loads, only 3.5% of the properties change and for a gap for three days, only 4.5% properties changed. These are the 95% percentile numbers, and therefore further motivate us to expect extremely high gains from a javascript caching framework.

Currently, we are working on capturing the javascript execution at the second granularity, ie at the function level, which we covered in section 3.1. In order to do this, we have built a web proxy which sits between the client browser and the news and sports websites. Everytime a request is made by the client, the proxy intercepts the request, and injects instrumentation code in the javascript files and the inline script tags inside the html files. This instrumented code is read by the javascript debugger when the page is actually loaded, and the debugger then builds a callgraph, with each node representing a function that was invoked. Once a graph is built, we will use some a kind of a graph diffing algorithm, to quantify how much of the call graph was modified, across the two loads.

5 Related Work

Work on improving web performance has been ongoing for more than two decades now. Prior work has focused on various components of the overall page load time from remodifying the source code of the webpage itself, to optimizing the network component of the overall execution time. More recently, some work has been in improving the computatio time latency.

Erman et all [6] has shown that unlike desktop browsers, optimizations such as SPDY/HTTP2 does not improve PLT on mobile browsers. They show that this is because of the negative interactions between the cellular state machine and the transport protocol. Similary, Qian et al [14] show that caching does not provide page load improvements for mobile browsers. A recent paper from Google in 2016 [20] showed how there is very little improvement to the overall page load time despite significant improvements in the cache hit rate.

Much of the research on explicitly improving mobile browser performance has seen mixed results. FLywheel [1] is Google's compression proxy that compresses web content to significanlty reduce the use of expensive cellular data. The authors note that whie Flywheel succeeds in reducing the data usage, its effect on page load time is more mixed; it helps the performance of certain pages and hurts the performance of others. Flexiweb [17] is built over Google's compression proxy to ensure that the proxy does not hurt page load times. However, FlexiWeb is not designed to explicitly improve page load performance. Wang et al [21] show that speculative loading in one of only client only approaches that can improve mobile browser performance. However, speculative loading requires knowledge of what objects are likely to be requested by the user.

Other research works have looked at metrics that are orthogonal to the page load time metric. Parcel [19] uses a proxy approach to divide the page load process between the mobile device and the proxy. Because Parcel is a network appraoch, the evaluations are primarily focused on the reduction of network latency. Klotski [5] focuses on increasing the number of objects rendered in the first 5 seconds to improve the user quality of experience.

Other client side improvements reduce energy usage and computational delays using parallel browsers [9, 10] and improved hardware [24] By improving the parallelization for necessary page load tasks, such as rendering, these systems reduce energy usage and have a positive impact on page load times.

While there have been several recent efforts on improving mobile browser performance they have not been uniformly successful due to their various limitations.

6 Future work

Most of the project up till now, was meant to serve as a motivation for building a caching framework for the javascript execution output. All our experiments have given positive results on the expected benefits of a caching framework. We also have established an upper bound on the posible decrease in the page load time in the best case scenario.

The actual implementation of the caching framework will be our next step. This is more like an engineering effort, which will determine the efficacy of our idea and will actually evaluate it. In order to do this, we will have to modify the production level source code of the Chrome browser, as that is the browser we are focusing on for our research (the reason for which is mentioned in the introduction).

Our understanding of the current caching framework used for compiler and parser cache can be used as a reference of our javascript caching framework.

7 Contribution

Our work had a fair distribution among the three co authors. This being the main reseach project of Ayush Goel, he was primarily responsible for setting up the testbed for the experiments and conducting them. All the code for the different anaylsis done for the page loads, like network analysis, trace analysis and capturing the window object state and running a simple diff algorithm is written by Ayush Goel. He has studied the current optimisations in place done by Chrome, and evaluated the improvement in the loading time with these optimisations enabled.

Matthew Furlong, the second co author has contributed in finding the javascript libraries we currently use to capture the network and timeline trace for each web page load, on top of which most of our other analysis code is written. He has contributed in analysing data from our experiments by building CDFs of the results and comparing them with each other.

Hyunjong Lee, the third co author has contributed in writing the mid semester and the final reports, generating figures and writing the content.

Most importantly, Matthew and Hyunjong were constantly involved in the critical discussion of the key ideas, and helped in estabilishing the next steps for our research project.

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