

Putting DNS in Context

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"I was drivin' through the misty rain Searchin' for a mystery train Boppin' through the wild blue Tryin' to make a connection with you"





Previous Work

inter-related such that the Web browsers comtion strategies for 'ficantly impact vlain why a - loaded

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Predominant one. Despite being one of the oldest Internet systems DNS still operates with semi-obscure interactions among its systems holders: domain owners. network onerators. onerating systems and

DNS still operates with semi-obscure interactions among its stake holders: domain owners, network operators, operating systems and app developers. The goal of this work is to holistically understand

holders: domain owners, network operators, operating systems, and app developers. The goal of this work is to holistically systems, and the dvnamics of DNS in mobile traffic along with the role of each of

app developers. The goal of this work is to holistically understand the dynamics of DNS in mobile traffic along with the role of each of ite etakeholdere Wence two complementary (anonymized) datasets: the dynamics of DNS in mobile traffic along with the role of each of the state of the section of

its stakeholders. We use two complementary (anonymized) datasets traffic logs provided by a European mobile network operatod datasets with 19M customers. and traffic logs from 5.000 users of Lumen. a

traffic logs provided by a European mobile network operator (MNO) with 19M customers, and traffic logs from 5,000 users of Lunen, a traffic monitoring and for Android. We complement such nassive

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of total network flows. The time to live (TTL) values for such due tends to change on a longer time-scale. Further. denending on the

mains are mostly short (< 1min), despite domain-to-IPs mapping tends to change on a longer time-scale. Further, depending on the one-rators recursive resolver architecture, end-user devices receive

tends to change on a longer time-scale. Further, depending on the operators recursive resolver architecture, end-user devices roceive even smaller TTL values leading to subontimal effectiveness of

operators recursive resolver architecture, end-user devices receive even smaller TTL values leading to suboptimal effectiveness of the on-device DNS cache Desnite a number of on-device and in

even smaller TTL values leading to suboptimal effectiveness of network ontimizations available to minimize DNS overhead which

the on-device DNS cache. Despite a number of on-device and in network optimizations available to minimize DNS overhead and in wa find correction to 10% of nage load time (PLT) on average

network optimizations available to minimize DNS overhead, which we find corresponding to 10% of page load time (PLT) on average, we have not found wide evidence of their adoption in the wild.

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Dissecting DNS Stakeholders in Mobile Networks

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According to recent estimates, mobile traffic is expected to have a seven fold increase by 2021 [9]. At such growing rate, improving network architectures and understanding quality of experience (OoE)

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In the end-user QoE. Many previous studies investigated QoE in mobile networks consing on inefficiencies of the access technology and networks

Many previous studies investigated QoE in mobile networks focusing on inefficiencies of the access technology and networks nath 15 7 17 19 22 25 311 and in-nath nroxies [1 14 27]. DNS has

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been considered as part of general traffic performance studies [7, 17, 19], and (small scale) active experiments [23]. Less attention [7, 17, instead viven to DNS in mobile networks and most immortantly. 19], and (small scale) active experiments [23]. Less attention was instead given to DNS in mobile networks and, most importantly,

Is impact on users QoE. Despite recent evolutions such as 4G, mobile networks last mile is till a charod arroece medium cuffering from high latenry Henre com-Despite recent evolutions such as 4G, mobile networks last mile is still a shared access medium suffering from high latency. Hence com munications should be optimized (or avoided) when possible. Con still a shared access medium suffering from high latency. Hence com munications should be optimized (or avoided) when possible com-sidering DNS, on-device DNS caches and app-specific optimizations munications should be optimized (or avoided) when possible. Con-sidering DNS, on-device DNS caches and app-specific optimizations such as nre-fetching [15] are adopted to minimize DNS lookups. sidering DNS, on-device DNS caches and app-specific optimizations such as pre-fetching [15] are adopted to minimize DNS lookups. However: DNS stakeholders-domain owners, mobile network on. such as pre-fetching [15] are adopted to minimize DNS lookups. However, DNS stakeholders-domain owners, mobile network op-

Mario Almeida, Alessandro Finamore * Diego Perino * Narseo Vallina-Rodriguez * Matteo Varvello*

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sources fetched by HTTP with JavaScript and CSS eval-

ABSTRACT

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Demystifying Page Load Performance with WProf

Web Page load time is a key performance metric that pany techniques aim to reduce Unfortunately the com-

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Plexity of modern web Pages makes it difficult to iden-all the performance bottlenecks. We present WProf, a bottlenecks. We present we detailed tify performance bottlenecks. The produce a detailed lightweight in-browser profiler that produce a detailed tify performance bottlenecks. We present WProf, a lightweight in browser profiler that produces a detailed dependency graph of the activities that make up a new

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dependency graph of the activities that make up a page load. WProf is based on a model we developed to cap-ture the constraints between natural load page pare

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ture the constraints between network load, page pars-ing, JavaScript/CSS evaluation, and rendering activity in nonder browcere we combine word renorte with

ing, JavaScript/CSS evaluation, and rendering activity in popular browsers. We combine WProf reports with oritical path analysis to study the page load time of 350

In popular browsers. We combine WProt reports with critical path analysis to study the page load time of 350 Web page under a variety of cattinge including the use

critical pain analysis to study the page load time of 320 Web pages under a variety of settings including the use of end-hoet caching

Web pages under a variety of settings including the use of end-host caching, SPDY instead of HTTP, and the mod nagesneed certier extension We find that connected

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HIOU pagespeeu server extension. We have a much as 37 tion is a significant factor that makes up as much as 37

uon is a significant factor that makes up as much as of the critical path, and that synchronous JavaScript

or the critical paul, and that synchronous Javascrift a significant role in page load time by blocking

Xiao Sophia Wang, Aruna Balasubramanian, Arvind Krishnamurthy, and David Wetherall University of Washington

Dataset From Case Connection Zone







Methodology: Blocking



DNS Record Origin

 We'll now use our two primitives—pairing and blocking—to catagorize application transactions based on where the corresponding DNS information comes from

No DNS



Prefetched



Prefetched



Local Cache



Local Cache



Shared Resolver Cache



Auth Resolution Required



DNS Record Origin

	% Connections	
No DNS	7.2	
Local Cache	42.9	57.9% of conns. have no direct DNS costs
Prefetched	7.8	
Shared Resolver Cache	26.3	42.1% of conns.
Auth Resolution Required	15.7	don't have needed DNS information
		on hand

Absolute DNS Cost



Relative DNS Cost





CDF

DNS Cost

- We say application connections requiring a DNS lookup experience "significant DNS cost" when both the following criteria are met:
 - DNS lookup duration > 20msec
 - DNS percentage contribution > 1%

DNS Cost



64% of transactions show an *insignificant* DNS cost using both criteria 27% of transactions show an *insignificant* DNS cost using one criteria

<9% of transactions show an significant DNS cost using both criteria

Also In The Paper ...

- More methodological details
 - including ethical considerations in appendix

- Analysis as a function of resolver
 - ISP vs. Google vs. OpenDNS vs. Cloudflare

Brief analysis of how to improve DNS further



Questions? Comments?



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