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Near-Optimal Throughput with Shallow Buffers in Reconfigurable Datacenter Networks

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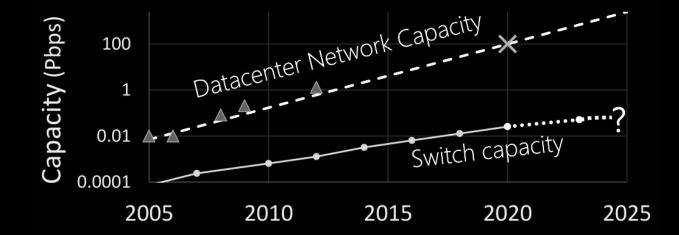




SIGMETRICS 2023



Network Demand vs Capacity Mismatch





[1] A flat datacenter network with nanosecond optical switching (SIGCOMM 2020)

A Scalable	Commodity Da	ata Center Netv	ork Architecture

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ABSTRACT

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Today's data centers may centuin tens of thousands of computers with significant aggregate bank-shift supermements. The network architecture typically consists of a tree of routing and switching denotes with propersystemic processing schemes equipdeploying the highest-end IP workshortwares, resulting topologies any ordo symport 5% of the aggregate bandwidth available at the edge of the network, while still incenting tremendous cest. Nonture design and intervention schemes application of the strength of the schemes of the strength of the schemes of the

In this paper, we show how to leverage largely commodity Binmet worksho is support the full aggraphic bandwidth of clusters consisting of ens of thousands of elements. Similar to how clusters of commodity counters have largely replaced more specialized as a second start of the second start of the second start connected commodity switches may deliver more performance at second that available from tody's high-rend solutions. Our approach requires no modifications to the end host network interface moming by vibs. In orphild and TCD and the second start of the second commodity of the orphild and TCD and the second start of the second star

Categories and Subject Descriptors

C.2.1 [Network Architecture and Design]: Network topology; C.2.2 [Network Protocols]: Routing protocols

General Terms Design, Performance, Management, Reliability

Keywords

Data center topology, equal-cost routing

1. INTRODUCTION

Growing expertise with clusters of commodity PCs have enabled a number of institutions to harness petaflops of computation power and petabytes of storage in a cost-efficient manner. Clusters consisting of tens of thousands of PCs are not unheard of in the largest

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institutions and thousand-node clusters are increasingly commor in universities, research labs, and companies. Important applications classes include scientific computing, financial analysis, data analysis and warehousing, and large-scale network services. Today, the principle bottleneck in large-scale clusters is often inter-node communication bandwidth. Many applications must exchange information with remote nodes to proceed with their local computation. For example, MapReduce [12] must perform significant data shuffling to transport the output of its map phase before proceeding with its reduce phase. Applications pupping on cluster ased file systems [18, 28, 13, 26] often require remote-node ac cess before proceeding with their I/O operations. A overy to a web search engine often requires parallel communication with every node in the cluster hosting the inverted index to return the most relevant results [7]. Even between logically distinct clusters, there re often significant communication requirements, e.g., when updating the inverted index for individual clusters performing search from the site responsible for building the index. Internet services ngly employ service oriented architectures [13], where the retrieval of a single web page can require coordination and communication with literally hundreds of individual sub-services runnine on remote nodes. Finally, the significant communication requir ments of parallel scientific applications are well known [27, 8]. There are two high-level choices for building the communicatio fabric for large-scale clusters. One ontion leverages specialized hardware and communication protocols, such as InfiniBand [2] or Myrinet [6]. While these solutions can scale to clusters of thousands of nodes with high bandwidth, they do not leverage com modity parts (and are hence more expensive) and are not natively ompatible with TCP/IP applications. The second choice lever ages commodity Ethernet switches and routers to interconnect clus er machines. This approach supports a familiar management in frastracture along with unmodified applications, operating systems and hardware. Unfortunately, aggregate cluster bandwidth scales orly with cluster size, and achieving the highest levels of band

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width iccurs non-linear cost increases with cluster size. For compatibility and cost reasons; most chied communication theory of the second second second second second second humbridin in large clusters may become versubscribed by a significant factor depending on the communication patterns. That is, too nodes consecuted in the same physical visich may be able to withinks, potentially cancers may be able to patient within the second second second second second patient second second second second second second second patients and the second second second second second second second virials at the root of the communication heratery of the bandwidth available to the ord the unmaintained heratery the second second

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VL2: A Scalable and Flexible Data Center Network

Albert Greenberg Srikanth Kandula David A. Maltz James R. Hamilton Changhoon Kim Parveen Patel

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Abstract

To be agile and cost effective, data centers should allow dynamic resource allocation across large server pools. In particular, the data center network should enable any server to be assigned to any ser vice. To meet these goals, we present VLz, a practical network archi tecture that scales to support huge data centers with uniform high capacity between servers, performance isolation between services and Ethernet layer-2 semantics. VL2 uses (1) flat addressing to allow service instances to be placed anywhere in the network, (2) Valiant Load Balancing to spread traffic uniformly across network paths, and (3) end-system based address resolution to scale to large server pools, without introducing complexity to the network control plane /L2's design is driven by detailed measurements of traffic and faul data from a large operational cloud service provider. VL2's imple mentation leverages proven network technologies, already available t low cost in high-speed hardware impleme able and reliable network architecture. As a result, VL2 networks can be deployed today, and we have built a working prototype. We evaluate the merits of the VL2 design using measurement, analysis, and experiments. Our VL2 prototype shuffles 2.7 TB of data among 75 servers in 395 seconds - sustaining a rate that is 94% of the max-

Categories and Subject Descriptors: C.2.1 [Computer-Communi cation Network]: Network Architecture and Design General Terms: Design, Performance, Reliability Kerworks: Data center network, commodifization

1. INTRODUCTION

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Permission to make digital or hard copies of all or part of this work for provad or classroom set is granted without fee provided that copies are no made of estimated for petitor commercial advantage and that copies regulation to post an average of the set of the set of the set of the regulation to post an server as to reduze the to blues, requires program termission and/set of the set or eduze the set of the set of the set SIGCOMM '09. August 17-21, 2009, Barcelona, Spain Covering 12099 ACM '096-160558 '04040086 ...510.00. Agility promises improved risk management and cost savings. Without agility, each service must pre-allocate enough servers to meet difficult to predict demand spiks, or risk failure at the brink of success. With agility, the data center operator can meet the fluctuating demands of individual services from a large shared server pool, resulting in higher server utilization and lower costs.

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Unfortunately, the designs for today's data center network prevent agility in several ways. First, existing architectures do not provide enough capacity between the servers they interconnect. onventional architectures rely on tree-like network configurations built from high-cost hardware. Due to the cost of the equipment the capacity between different branches of the tree is typically overubscribed by factors of 1:5 or more, with paths through the highes levels of the tree oversubscribed by factors of 1-80 to 1-140. This lim its communication between servers to the point that it fragments the erver pool - congestion and computation hot-spots are prevalen even when spare capacity is available elsewhere. Second, while data enters host multiple services, the network does little to prevent traffic flood in one service from affecting the other services around it — when one service experiences a traffic flood, it is common for all hose sharing the same network sub-tree to suffer collateral damage Third, the routing design in conventional networks achieves scale by assigning servers topologically significant IP addresses and dividing servers among VLANs. Such fragmentation of the address space limits the utility of virtual machines, which cannot migrate out of their original VLAN while keeping the same IP address. Further the fragmentation of address space creates an enormous configura ion burden when servers must be reassigned among services, and the human involvement typically required in these reconfigurations

limits the speed of deployment: To overcome three limitations in today's design and achieve agality, we arrange for the network to implement a familiar and concrete model gue each service the likelon that all the servers assigned to a, and only those servers, are constructed by a single this likelone even the time of each service varies from a server to to no.ose. Realizing this vision concretely translates into building a network that metrics the following three objectives:

 Uniform high capacity: The maximum rate of a server-to-server traffic flow should be limited only by the available capacity on the network-interface cards of the sending and receiving servers, and assigning servers to a service should be independent of network topology.

 Performance isolation: Traffic of one service should not be affected by the traffic of any other service, just as if each service was connected by a separate physical switch.

 Layer-2 semantics: Just as if the servers were on a LAN—where any IP address can be connected to any port of an Ethernet switch due to flat addressing—data-center management software should be able to easily assign any server to any service and configure

Jupiter Rising: A Decade of Clos Topologies and Centralized Control in Google's Datacenter Network

Arjun Singh, Joon Ong, Amit Agarwal, Glen Anderson, Ashby Armistead, Roy Bannon, Seb Boving, Gaurav Desai, Bob Felderman, Paulie Germano, Anand Kanagala, Jeff Provost, Jason Simmons, Elichi Tanda, Jim Wanderer, Urs Hölzle, Stephen Stuart, and Amin Vahdat Google, Inc. jupiter-siccomm@apoole.com

ABSTRACT

We present our approach for overcoming the cost, operational complexity, and limited scale endemic to datacenter networks a decade ago. Three themes unify the five generations of datacenter networks detailed in this paper. First, multi-stage Clos topologies built from commodity switch silicon can support cost-effective de ployment of building-scale networks. Second, much of the general, but complex, decentralized network routing and management protocols supporting arbitrary deployment scenarios were overkill for single-operator. pre-planned datacenter networks. We built a centralized control mechanism based on a global configuration pushed to all datacenter switches. Third, mod lar hardware design coupled with simple, robust software allowed our design to also support inter-cluster and wide-area networks. Our datacenter networks run at dozens of sites across the planet, scaling in capacity by 100x over ten years to more than 1Pbps of bisection handwidth

CCS Concepts

Keywords Datacenter Networks; Clos topology; Merchant Silicon; Centralized control and management

Networks → Data center networks:

1 INTRODUCTION

Datacenter networks are critical to delivering web services, modern storage infrastructure, and are a key en-

Permission to make digital to head organs of gauser all of this work for generating or classroom one is granted without for gravity are not studie of distribuild for profit or commercial absoluting and that organs have the based of the studies of the studies of the studies of the studies in work much be based. For all due to exact the convertibution's SIGCOMM '15 August 17-21, 2015, Londone, Datterd Kingdoon 0.2015 CorptyPitchWork by disc researchardwise's ACM ISBN 97-1-450-5542-51508. abler for chould computing. Bandwidth demands in the disaccenter are doubling every 15-15 models (Figure []), even faster than the wish area literater. A number of reting the start of the start of the start of the start of the intervent start of the start of the start of the start intervent start with a start or start of the start can deliver higher spatialer results by accessing more data the the critical path of noise/start or greats. Finally, constellations of correstiont approximations often share subtional results and the start of the start of the start of the start index generations. We start, and start in the start of the start index generation, we be start, and start in the start of the start of the start index generation.

Ten years ago, we found the cost and operational omplexity associated with traditional datacenter net work architectures to be prohibitive. Maximum network scale was limited by the cost and capacity of the highest end switches available at any point in time 24 These switches were engineering marvels, typically re cycled from products targeting wide area deployments WAN switches were differentiated with hardware sur port/offload for a range of protocols (e.g., IP multicast) or by pushing the envelope of chip memory (e.g., Internet-scale routing tables, off chip DRAM for deep buffers, etc.). Network control and management pro tocols targeted autonomous individual switches rather than pre-configured and largely static datacenter fabrics. Most of these features were not useful for datacen ters, increased cost, complexity, delayed time to market and made network management more difficult.

Datacentre twitches were also built as complex chassis targeting the highest levels of availability. In a WMN Internet deployment, losing a single switch/router can have substantial impact on applications. Because WAN links are so expensive, it makes sense to invest in high availability. However, more platient to trade cost switches operating in a multi-weaked WMN environment with adultary end hosts require support for many potrocols to ensure interoperativity. In single-operator data





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Jellyfish: Networking Data Centers, Randomly Ankit Single': Chi-Yao Hong', Lucian Papel, P. Brigheng Goff ¹ University of Clulifornia, Berkeley ² University of California, Berkeley		ffective Low-Diameter Network Topology Torsten Hoeffer TITV Zarch bereffictud	Xpander: Towards Optimal-Performance Dat Asaf Valadarsky [®] Gal Stabat ^{ef} asaf valadarsky@mail.huji.ac.il gal.shabat@mail.huji.ac.il Michael Schapiram@huji.ac.il	tacenters Michael Dinitz ¹ mdinitz®es.jhu.edu
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MB is supported by the 2013 Google European Doctoral Fellowship in Parallel ¹Numbers for random topologies are updated from values obtained using the Backing installation in the Jacob and a scheduled with endotient formation.

ation datacenters, e.g., low-diameter networks such as Slim spect to the size of S. We present a formal definition of Fly 9 or random networks like Jellyfish 48 have an im4

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A consequence of our work is to Abstract Unfortunately, in our experience, this lack of a clear under The data center network is increasingly a cost reliabilshow that entirely network-level failure recovery can be standing of lifecycle management complexity often results ity and performance bottleneck for cloud computing. Al-Most recent datacenter topology designs have focused on practical and nearly instantaneous in a data center setting. in costly mistakes in the design of datacenters that are disthough multi-tree topologies can provide scalable bandperformance properties such as latency and throughput. In Addressing this need for network-layer recovery. Fatcovered during deployment and therefore cannot be rectified. width and traditional routing algorithms can provide eventhis paper, we explore a new dimension, life cycle manage-Tree architectures have proposed using a centralized man-A Cost Effec Our paper is a first step towards useful characterizations of tual fault tolerance, we argue that recovery speed can be ment complexity, which attempts to understand the complexager that collects topology and failure information from lifecycle management complexity, dramatically improved through the co-design of the netity of deploying a topology and expanding it. By analyzing the switches. It then periodically generates and dissemi-Contributions. To this end, our paper makes three contribu work topology, routing algorithm and failure detector. We ptimal-Performance Datacenters current practice in lifecycle management, we devise complexnates back to the switches and end-bosts alternate sets of tions. First, we design several complexity metrics (§3 and §4) Iell create an engineered network and routing protocol that diity metrics for lifecycle management, and show that existing routes to avoid failures. Centralized route management that can be indicative of lifecycle management costs (i.e., caprectly address the failure characteristics observed in data topology classes have low lifecycle management complexity Gal Shahaf[†] Michael Dinitz[‡] is both simple and flexible-a reasonable design choice ital expenditure, time and manpower required). These metrics centers. At the core of our proposal is a novel network by some measures, but not by others. Motivated by this, we Ankit S provided that failures do not occur very often. al.shahaf@mail.huii.ac.il mdinitz@cs.ihu.edu include the number of: switches, patch panels, bundle-types, topology that has many of the same desirable properties Maciei Besta design a new class of topologies, FatClique, that, while being Recent measurements of network-layer failures in data expansion steps, and links to be re-wired at a patch panel rack as FatTrees, but with much better fault recovery proplichael Schapira ETH Zurich performance-equivalent to existing topologies, is compara centers, however, have shown that failures are frequent erties. We then create a series of failover protocols that during an expansion step. ble to, or better than them by all our lifecycle management aniram@huii ac il naciei.besta@inf.ethz.ch and disruptive [10]. Network-layer failures can reduce the benefit from this topology and are designed to cascade We design these metrics by identifying structural elements complexity metrics. volume of traffic delivered by more than 40%, even when and complement each other. The resulting system, F10, of network deployments that make their deployment and ex-1 Introduction the underlying network is designed for failure resilience 1 Introduction can almost instantaneously reestablish connectivity and pansion challenging. For instance, the number of switches a high-nerformance cost-effectiv the network topology and exploiting the diversity of short As data centers grow, the probability of network failures load balance, even in the presence of multiple failures. Ty that approaches the theoreti Slim Fly is based on graphs in the topology determines how complex the network is in Over the past decade, there has been a long line of work on paths afforded by expanders for efficient delivery of data Data centers today form Our results show that following network link and switch and the consequent disruptions on the system as a whole terms of packaging - laying out switches into homogeneous designing datacenter topologies [2, 35, 31, 32, 3, 4, 20, 1]. raffic. Thus, our first contribution is shedding light on the o the degree-diameter problem are it to both traditional and stat tions. A well provisioned failures. F10 has less than 1/7th the packet loss of curwill likely increase, further exacerbating the problem. racks in a space efficient manner. Wiring complexity can perforunderlying reason for the empirically good performance of While most have focused on performance properties such as tant to ensure that servers rent schemes. A trace-driven evaluation of ManReduce Our goal is to co-design a topology and set of protobe assessed by the number of cable bundles and the patch is shows that Slim Fly has sign latency and throughput, and on resilience to link and switch previously proposed datacenter architectures, by showing necks to utilization: to isc performance shows that F10's lower packet loss yields a cols that admit near-instantaneous fine-grained localized s in latency, handwidth res panels a design requires. As these increase, the complexity of . Con- that these proposals are specific points in a much larger deand to gain more freedom failures, datacenter lifecycle management [30, 38] has largely network-level recovery and rebalancing for common-case Finally, we propose deadloc median application-level 30% speedup. manufacturing and packaging all the different cable bundles mance sign space of "expander datacenters". We observe, howthan having to tailor place al layouts for large computing co and power model. Slim Fly er and highly resilient datacenter been overlooked. Lifecycle management is the process of network failures. Because the network is already a signifiefficiently into cable trays, and then routing them from one enelits ever, that these points are either not sufficiently close to opbandwidth is available [] 1 Introduction building a network, physically deploying it on a data-center cant part of the cost of the data center, we limit ourselves patch panel to the next can be expected to increase. Finally om the timal performance-wise, are inherently not scalable, or face body of work has tackled floor and expanding it over several years so that it is available. Data center networks are an increasingly important comto not introducing any additional hardware relative to latency and high bandwidth because expansion is carried out in steps [38], where the net network capacity interco raphs' significant deployment and maintenance challenges (e.g., in for use by a constantly increasing set of services. ponent to the cost, reliability and performance of cloud PortI and Other work has shown that local renair is nosh as stencil or graph compu One crucial problem that work operates at degraded capacity at each step, the number terms of unpredictability and wiring complexity). With datacenters living on for years, sometimes up to a services. This has led to recent efforts by the network resible at the cost of significant added hardware relative to of expansion steps is a measure of the reduced availability how We argue that the quest for high-performance datacensigns is that of incremental decade [3], 12], their lifecycle costs can be high. A data a standard EarTree [9, 12, 13], so our work can be seen as adding servers and network search community to explore new topologies [11, 12, 13], in the network induced by lifecycle management. Wiring pect to ter designs is inextricably intertwined with the rich body center design that is hard to deploy can stall the rollout of new routing protocols [11] and new network manage either improving the speed of repair in FatTree and other s play an important role in to patterns also determine the number of links that need to be insurof research in mathematics and computer science on builddata center. This may be mo services for months; this can be expensive considering the rate ment layers [3, 4, 20], with a goal of improving network multi-tree networks or in reducing the hardware cost of as. The importance of the net ese in- ing good expanders. We seek a point in this design space rewired at a patch panel during each step of expansion, a base, which requires more at which network demands have historically increased [3] cost-effectiveness, fault tolerance and scalability fast repair in more general networks. A limitation of our per-node (multi-core) perfor tecture that offers near-optimal performance guarantees while proof more bandwidth-inten measure of step complexity [38]. work is that we assume that we can change both the netwe networks with tens of thor 231. A design that is hard to expand can leave the network sion can be made feasible A state of the art approach is taken by Al-Fares et al. [3] s a tan- viding a practical alternative for today's datacenters (in Our second contribution is to use these metrics to compare warehouse-sized HPC and work topology and the protocols used between network functioning with degraded capacity impacting the large array sioning of space and pow and its followup project PortLand [20]. In these systems, inder's terms of cabling, physical layout, backwards compatibility of such networks are dete the lifecycle management costs of two main classes of dataof services that depend on it. server base. The latter enal the data center network is constructed in a multi-rooted switches erature with today's protocols, and more). We present Xnander, a rement of nodes and cable enter topologies recently explored in the research literature It is therefore desirable to commit to a data-center network to be replaced by a larger r tree structure called a FatTree (inspired by fat-trees [17]) Our system is called F10 (the Fault-Tolerant Enginovel expander-datacenter architecture carefully engineered e taken into account while de (§2), Clos [2] and expander graphs [32, 35]. We find that design only after getting a sense of its lifecycle management at the same time, less nowe of inexpensive, commodity switches. These proposals neered Network), a network topology and a set of prototo achi.eve both these desiderata rst, high bandwidth is indisper neither class dominates the other: Clos has relatively lower and ar cost and complexity over time. Unfortunately, the costs of Industry experience indi provide scalability, both in terms of port count and the cols that can recover rapidly from almost all data center Importantly, utilizing expanders as network topologies orm all-to-all communicatio wiring complexity; its symmetric design leads to more unid. Our sion is an important prob overall bisection bandwidth of the network. They also network failures. We design a novel topology to make it the large array of components needed for deployment such as has been proposed in a large variety of contexts, rangunt for as much as 33% form bundling (and fewer cable bundle types); but expander of Facebook's data center deliver better performance at low costs, primarily due to easier to do localized repair and rebalancing after failures. 50% of the overall system er switches, transceivers, cables, racks, patch panels¹, and cable ing from parallel computing and high-performance computgraphs at certain scales can have simpler packaging require-Ve dis. 30,000 in November 2009 1 their use of commodity switches. This topology is applicable to the FatTree and other multiing [49] 15] 14 9] to ontical networks [41] and neer-to-neer they should be cost and trays, are proprietary and change over time, and so are hard ments due to their edge expansion property [32]; they end 2010 [24]. While Facebo The use of a large number of commodity switches, howtree networks. We then redesign the routing protocols to t-to-endpoint latency is imp to quantify. An alternative approach is to develop complexity networks [40] 39] Our main contributions are examining up using much fewer switches than Clos to achieve the same ter facilities too, much of th in high frequency trading. F ever, opens up questions regarding what happens when take advantage of the modified topology. To satisfy the measures (as opposed to dollar costs) for lifecycle managethe performance and operational implications of utilizing network capacity. Expander graphs also demonstrate better tal in existing facilities ("ad nt to link failures. expanders in the datacenter networking context, and seeklinks and switches fail. A FatTree has redundant paths need for extremely fast failover, we use a local recovment, but as far as we know, no prior work has addressed this expansion properties because they have fat edges (§4) which sis" [23]). For instance, Ea between any pair of hosts. If end host operating system ery mechanism that reacts almost instantaneously at the hat lowering network diame In part, this is due to the fact that intuitions about lifecvcle ing optimal design points in this specific domain (namely, permit more links to be rewired in each step double the size of its facili ut also the cost of a networ Xpander). Indeed, despite the large body of research on exchanges are possible between these end hosts, the network cost of additional latency and increased congestion. Some management are developed over time and with operations exearly 2012 [9]. Industry ex mes while maintaining Finally we design and synthesize a novel and practical class panders, many aspects of using expanders as datacenter netcan be set up to provide multiple paths. The end host manfailures are not short-term, so local rerouting eventually perience, and these lessons are not made available universally. cremental build-out as a use ing the diameter of a networ of topologies called FatClique (§5), that has lower overall works (e.g., throughput-related performance measures, speages packet loss and congestion across the paths using triggers a slightly slower pushback mechanism that redithe exup-front [20]. ergy consumption as each p lifecycle management complexity compared to Clos and excific routing and congestion control protocols, deployment A patch panel or a wiring aggregator is a device that simplifies cable MPTCP [22]. In many cases, the data center operator is rects traffic flows before they reach the faulty components. of SerDes. Another con Do current high-bandw pander graphs. We do this by combining favorable design costs, incremental growth, etc.) remain little understood. posals allow incremental is and moter buffers and will hit far-We next elaborate on expanders, expander datacenters, and other packets flowing throug proposal [1] as an illustrati Xpander. adace the number of corthy i ture is completely determin 10th USENIX Symposium on Networked Systems Design and Implementation (NSDI '13) 399 switches available. This is I USENIX Association ining high bisection bandy LISENIX Association 16th USENIX Symposium on Networked Systems Design and Implementation 235 First, it makes the design s topology [30] is an exam 1.2 Why Expanders? tion bandwidth fat-trees ca ection bandwidth. Still, 8192, 27648, and 65536 correspondence to the connections as it first has to Intuitively, in an expander graph the total capacity from any structured topology are not applicable) physical con up the tree to reach a core router and only then go down to show that state-of-the-art proposals for next-gener-set of nodes S to the rest of the network is large with re-"Firingly, a coin-toss decided the first among the first two authors. struction, and cabling layout. We discuss these chalation datacenters, e.g., low-diameter networks such as Slim spect to the size of S. We present a formal definition of MB is supported by the 2013 Goorde European Doctoral Fellowshin in Parallel ¹Numbers for random topologies are updated from values obtained using the

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Fly 9 or random networks like lellyfish 48 have an im-

Traditional Approach: Static Datacenter Topologies Which topology has better throughput?



Which topology has better throughput?

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> Measuring and Understanding Throughput of Network Topologies

High Throughput Data Center Topology Design

Ankit Singla, P. Brighten Godfrey, Alexandra Kolla University of Illinois at Urbana-Champaien

Abstract

With high throughput networks acquiring a crucial role in supporting data-intensive applications, a variety of data center network topologies have been proposed to achieve high capacity at low cost. While this work explores a large number of design points, even in the limited case of a network of identical switches, no proposal has been able to claim any notion of optimality. The case of hetvorks, incorporating multiple line-speed and port-counts as data centers grow over time, introuces even greater complexity In this paper, we present the first non-trivial upper-

ound on network throughput under uniform traffic pat terns for any topology with identical switches. We then show that random graphs achieve throughput surprisngly close to this bound, within a few percent at the scale of a few thousand servers. Apart from demonstrating that ous topology design may be reaching its lin its, this result also motivates our use of random graphs as ilding blocks for design of heterogeneous networks Given a heterogeneous pool of network switches, we explore through experiments and analysis, how the distribution of servers across switches and the interconnec tion of switches affect network throughput. We apply these insights to a real-world beteroreneous data center topology, VL2, demonstrating as much as 43% higher thput with the same equipment

1 Introduction

Data centers are playing a crucial role in the rise of Internet services and big data. In turn, efficient data center perations depend on high capacity networks to ensure that commutations are not bottlenecked on communication. As a result, the problem of designing massive hightant than ever. Numerous data center network architec ires have been proposed in response to this need [2, 10-15, 20, 23, 25, 26, 301, exploiting a variety of network pologies to achieve high throughput, ranging from fat trees and other Clos networks [2, 13] to modified renerlized hypercubes [14] to small world networks [21] and uniform random graphs [23]

However, while this extensive literature exposes several points in the topology design space, even in the lim-

ited case of a natural of identical switches, it does not answer a fundamental question: How far are we from muchmat-ontimal topology design? The case sensorie networks, i.e. networks composed of switches or servers with disparate capabilities, introduces even mater complexity. Hateronanaous natourk equipmen is, in fact, the common case in the typical data cen ter: servers connect to ton-of-rack (ToR) switches, which connect to aggregation switches, which connect to core witches, with each type of switch possibly having a dif ferent number of ports as well some variations in line speed. For instance, the ToRs may have both 1 Gbp and 10 Gbns connections while the rest of the network may have only 10 Gbps links. Further, as the network extrands over the years and new, more powerful equin ment is added to the data center, one can expect more neterogeneity - each year the number of ports sup ported by non-blocking commodity Ethemet switches in ases. While line-speed changes are slower, the mov to 10 Gbns and even 40 Gbns is happening now, and

higher line-speeds are expected in the near futur In spite of beterogeneity being commonplace in data center networks, very little is known about heterore neous network design. For instance, there is no clarify on whether the traditional ToR-aggregation-core organi ration is superior to a "flatter" network without such a switch hierarchy; or on whether powerful core switches hould be connected densely together, or spread more evenly throughout the network.

The roal of this paper is to develop an understand ing of how to design high throughput network topole vies at limited cost, even when heteroreneous compo nents are involved, and to apply this understanding to im noise real-world data center networks. This is pontriv ial: Network topology design is hard in general, becaus of the combinatorial explosion of the number of possible networks with size. Consider, for example, the related degree-diameter problem [9], a well-known graph the ory problem where the quest is to pack the largest possible number of nodes into a graph while adhering t constraints on both the degree and the diameter. Non trivial optimal solutions are known for a total of only

Sangeetha Abdu Jyothi*, Ankit Singla[†], P. Brighten Godfrey*, Alexandra Kolla* "University of Illinois at Urbana-Champaign 1ETH Zurich Advard—High throughput is of particular interest in data use-case and across time as applications spin up and down center and HPC networks. Although asynch activeck topologies or migrate [[23, [23]-42]. Although some applications maps that here proposed, is houris hard-head and the fight patterns [39, [39] ergs, and known versions are table topologies and across traffic patterns is about, and the fight patterns [39, [39] ergs, and head to see the strands of the stran

of multiple applications could still produce an unintended In this paper, we develop a framework to benchmark the difficult TM. In fact, [2] observes that network contention In this paper, we develop a framework is bandmark the difficult TM In fact, we using perturbations during aff TC space with the second space of the reperimental power of the second space of the second space of the space of the space space of the space with a start of the space work is start by the performance of the TM is a with energy of a space of a space of a space of the spa

topologies with scaling, robustness of performance across TMs, and the effect of scattered workload placement. Our evaluation responsible for it. code is freely available measure worst-case throughput, and (2) provide an expansive

I. INTRODUCTION

Throughput is a fundamental property of communication - TMs generated from real-world measurements, synthetic networks: at what rate can data be carried across the network. TMs, and finally our new near-woest-case TMs. We discuss between desired end-points? Particularly for data centers and each of these in more detail. high performance computing, an increase in throughput denand among compute elements has reinvigorated research on bisection bandwidth and sparsest cat, solve the problem network topology, and a large number of network topologies of estimating worst-case throughput. A number of studies

have been proposed in th network weproces, and a large number of network reproducts of estimating west-case throughput. A number of statists have been proposed in the past few years to achieve high (e.g. (1), (10, (4), (4), (6)), (6)) employ cut metrics. It has been capacity at low cost (3), (10, (11, (13, (16)-(18, (11, (19))))), noted (47) that bisection bundwidth does not always predict 40. 44. 45. However, there is little order to this large and ever-growing measure worst-case throughput? We show that it does not,

topologies, and there is no open, public framework available B, where A has a higher cut-metric even though B supports for testing and comparing topology designs. The absence of a well-specified benchmark complicates research on network that the mismatch between cuts and worst-case throughput

design, making it difficult to evaluate a new design against the exists even for highly-structured networks of small size = a 5erous past proposals, and difficult for industry to know ary 3-stage butterfly with only 25 nodes - where the sparsestwhich threads of research are most promising to adopt. Our goal is to build a framework for accurate and consistent than the worst-case throughput.

measurement of the throughput of network topologies, and use this framework to benchmark proposed data center and HPC

efficient algorithm to generate a near-worst-case TM for any To accomplish this, we need metrics for comparison of given topology. We show empirically that these near-wors t, and this turns out to be a subtle problem. Throughput can be measured by testing particular workloads, or traffic matrices (TMs), but the immediate question is what TMs to a TM for a somewhat different reason, which could be recon newside insight into the effect of topological structure for

test. One approach is to test a variety of common TMs, which purposed as a near-worst-case TM. Compared with [26], our methodology finds TMs that are just as close to the worst particular use cases reflected by those specific TMs. However

Our key contributions, then, are to (1) develop a heuristic to

chmarking of a variety of topologies using a variety of TMs

tically higher worst-case throughput. Further, we show

cut found through brute force computation is strictly greater

(1b) Since cut metrics don't achieve our roal, we develop a

heuristic to measure worst-case throughput. We propose an

average-case throughput (in a limited setting:

an arrue it is useful to ro beyond this. In HPC and data pper-bounding throughput performance of a routing scheme using an U niation. In addition, [29] did not use their TM to benchmark topologies center networks. TMs may vary widely depending on the

× - III 1 of 21 / A Throughput-Centric View of the Performance of Datacenter T... 57.4% - Q =



A Throughput-Centric View of the Performance of Datacenter Topologies

Pooria Namyar Sucha Supittayapornpong Mingyang Zhang University of Southern California Vidvasirimedhi Institute of Science University of Southern California and Technology

> Minlan Yu Ramesh Govindan Harvard Universit University of Southern California

> > 1 INTRODUCTION

ABSTRACT

While prior work has explored many proposed datacenter designs. only two desirns. Clos-based and expander-based are renerally concause they can scale using con chins. Prior work has used two different metrics, bisection ban width and throughout, for evaluating these topologies at scale. Little is known, theoretically or practically, how these metrics relate to each other. Exploiting characteristics of these topologies, we prove upper bound on their throwshout, then show that this upper bound better estimates worst-case throughput than all previous m. Using this upper bound, we show that for expan topolories, unlike Clos, beyond a certain size of the network, no idth; in fact, even relatively small expander-based topologies fail to achieve full throughput. We conclude by showing that using nowight to evaluate datacenter performance instead of bisection ndwidth can alter conclusions in prior work about datacenter cost, manageability, and reliability

CCS CONCEPTS

modeling: Network manageability: Topology analysis and cemeration - General and reference -> Metrics

KEYWORDS

Data centers, Throughput, Clos topologies, Network management ACM Reference Format-

Pooria Namvar, Sacha Samittavanerunour, Minervane Zhanz, Minlan Yu Ramesh Govindan. 2021. A Throughput-Centric View of the Perfor or of Datacenter Topologies. In ACM SIGCOMM 2021 Conference (SIG IM '22), August 23–28, 2022, Virtual Event, USA. ACM, New York, NY USA, 21 pages, https://doi.org/10.1145/3452296.347291

of servers. Our focus, in this paper, is on the design and evaluation Datacenter topology designs. Two distinct classes of topology designs have emerged in recent years. Clos [8] based designs Fat-tree [1], VL2 [15], Juniter [42] and Facebook Fabric [3], and such as F10 [36]. These hierard signs are bi-regalar, in which a switch either connects to H servers, or none at all (Figure 1). More recent alternative designs targe lower installation costs and/or incur lower management costs the Clos-based topologies. These designs employ an expander-graph to witches, and include Jellyfish [44]. Xpander [47], and FatClique [52]. These topologies are ani-regular: every switch connects to H servers (Figure 1). In both classes, each server connects

A primary contributor to the success of cloud computing is the dat-

commodity servers. The performance of distributed applic

tions running inside a datacenter, like search, reliable storage, and

social networks, is strongly determined by the design of the dat-

acenter network. This network consists of a topology in which

switches interconnect servers. Today, datacenters routinely have

to exactly one switch. Measures of topology capacity. The capacity of the data cente a topology with enough capacity to permit every server to send traffic at full line rate simplifies cloud application design: operators can place application instances anywhere in the network without cting performance, and this placement flexibility enables a plications to be more cost efficient and more robust to correlates failures (e.r., of an entire rack or power domain) [15, 21, 35, 42]. Most prior work [1, 3, 15, 42, 52] has used the network's bisectio bandwidth, the smallest aggregate capacity of the links crossing the vorst-case cut among all the cuts that divide the topology to two halves, as a measure of its capacity. A topology has fall hisection handwidth if its hisection handwidth is at least equal to half of the total servers; for Clos-based designs, such a topology permits arbitrary application instance placement

Other work [24, 26, 27, 59, 51] has explored an alternative mea sure of network capacity, throughput, defined as follows. The throughout under traffic matrix T is the highest scaling factor $\theta(T)$ such that the topology can support the traffic matrix, $T \cdot \theta(T)$,

Other topology designs, such as DragonPly [39], and SlimPly [6], do not scale to the inst of modern data context, so we do not consider them in this resources as 57

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- Generalization of the design space: *Topology can change over time*
- Static networks are a special case



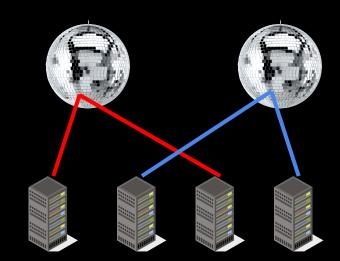
- Generalization of the design space: *Topology can change over time*
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- Generalization of the design space: *Topology can change over time*
- Static networks are a special case





ProjecToR [Sigcomm 2016]





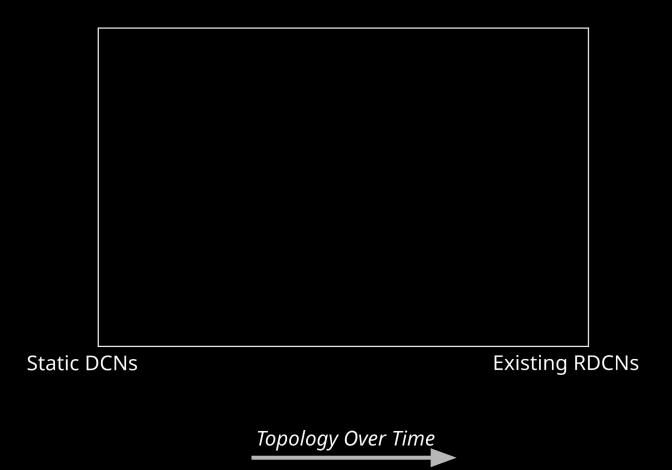
Reconfigurable Datacenter Networks Which topology has better throughput?



Which topology has better throughput?

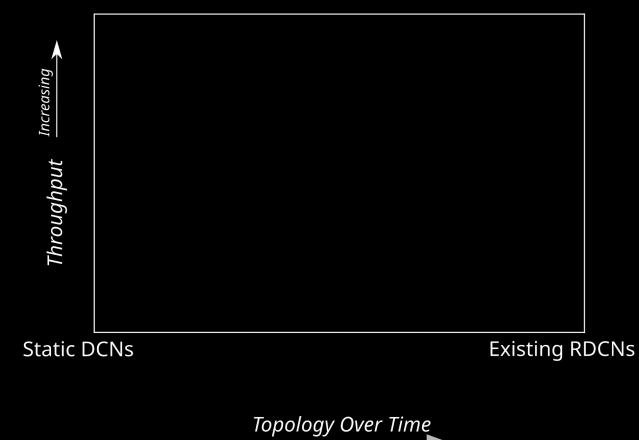
Main focus of this paper. *Stay tuned!*



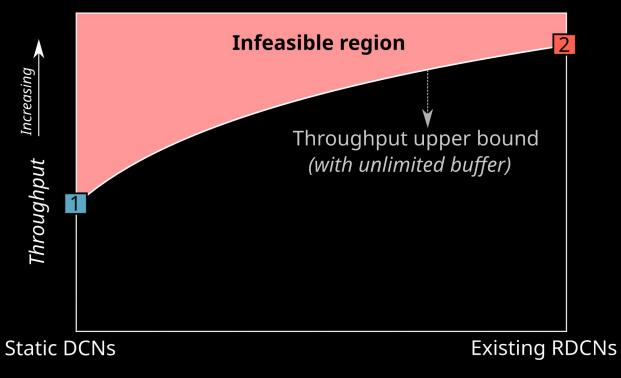






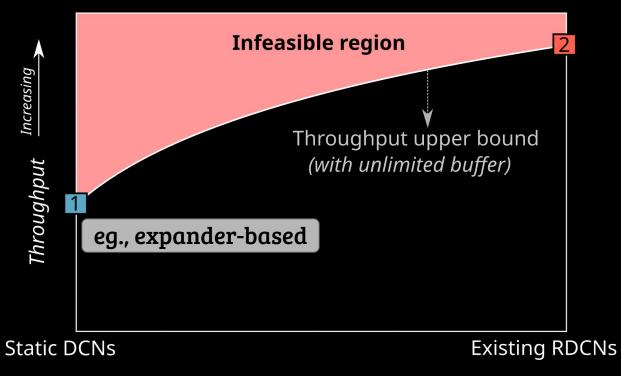




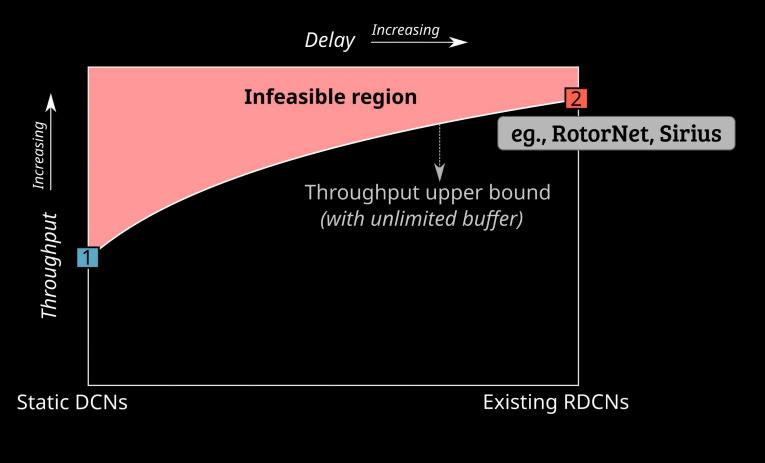






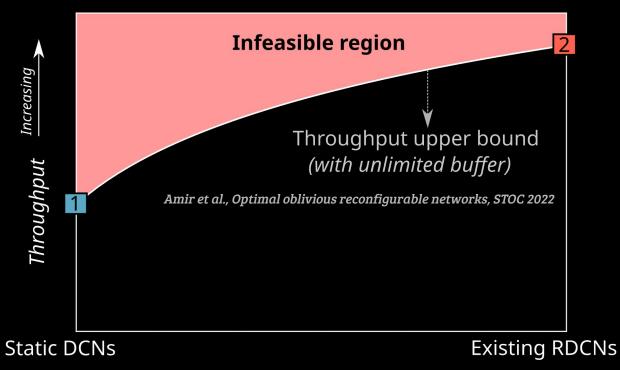






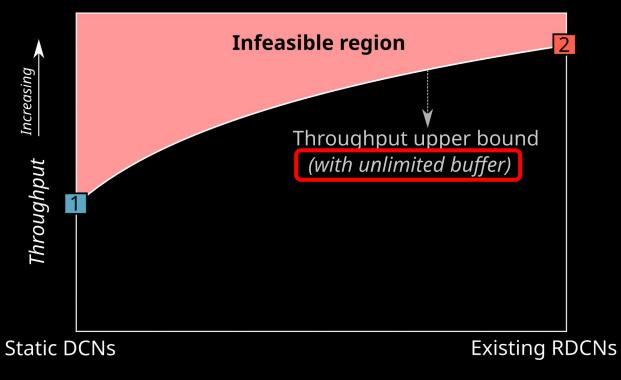






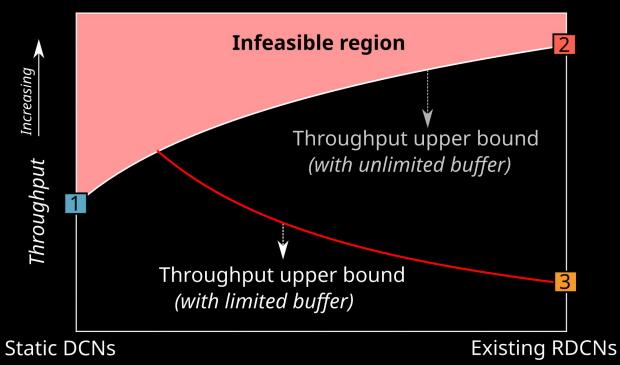






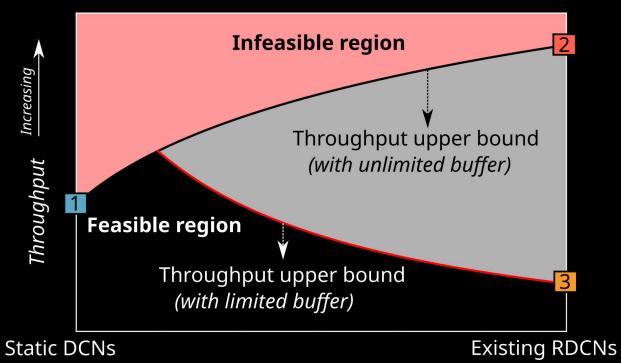






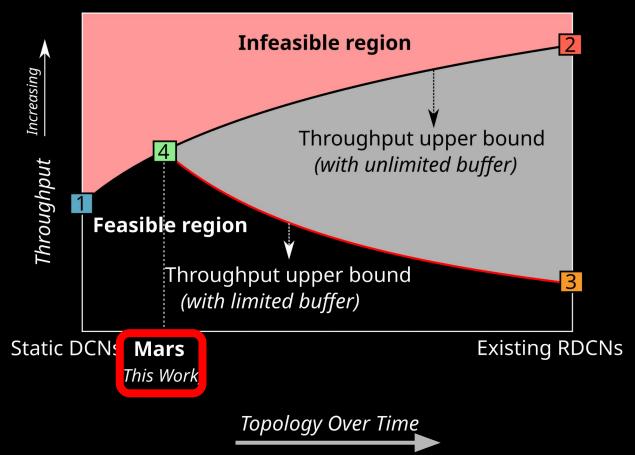


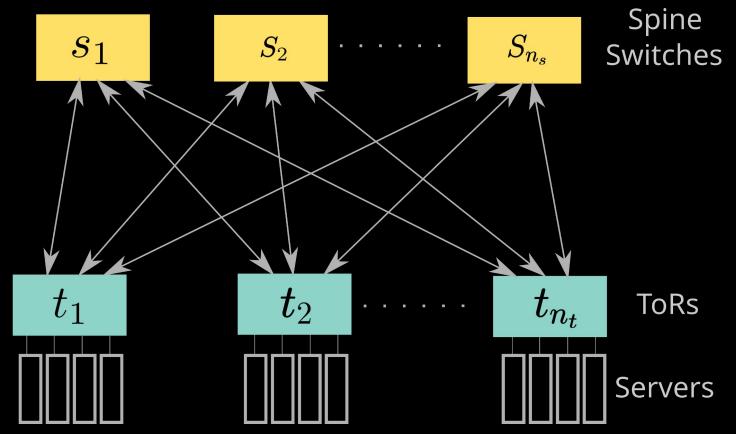












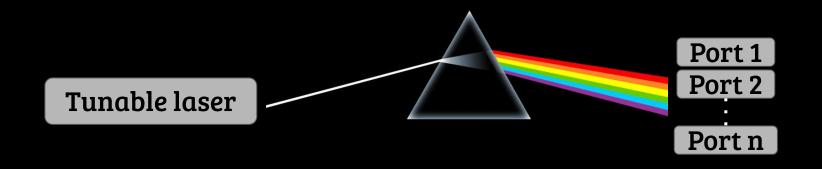




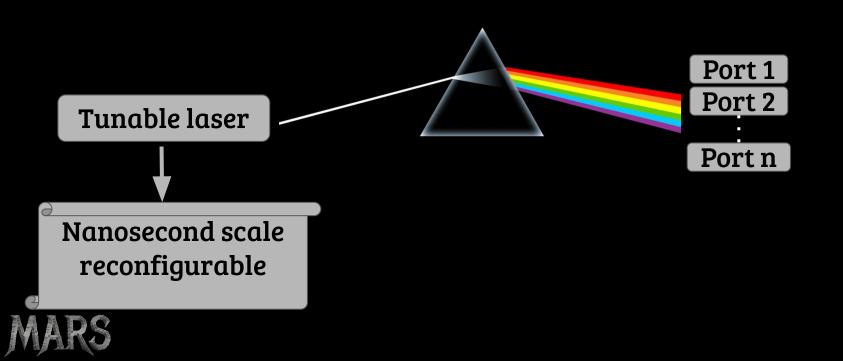


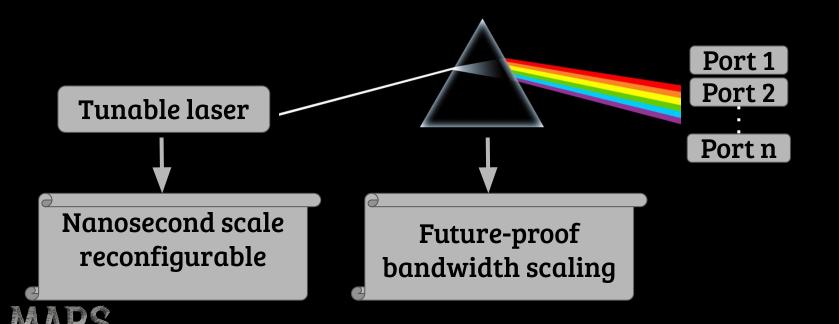


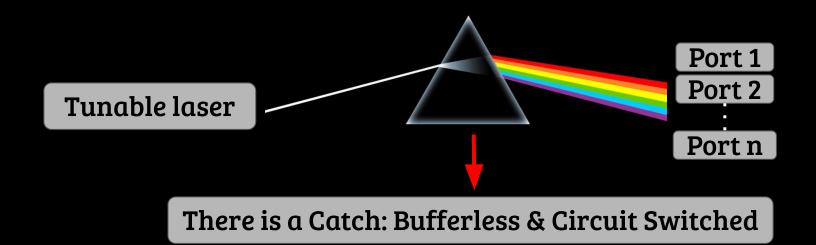




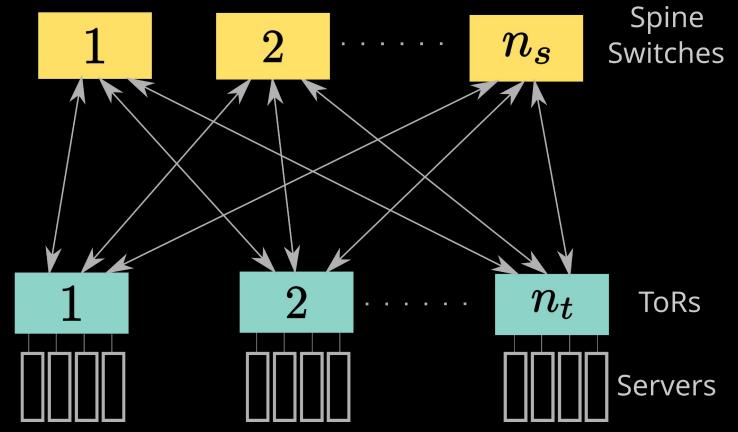




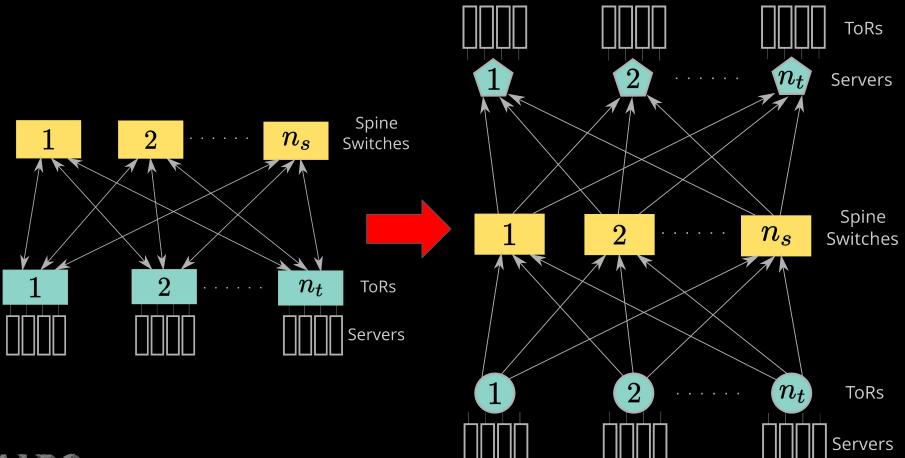




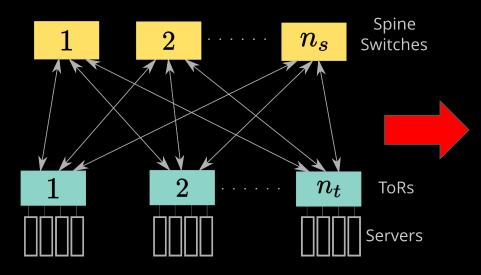


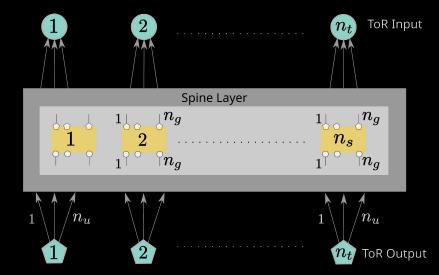




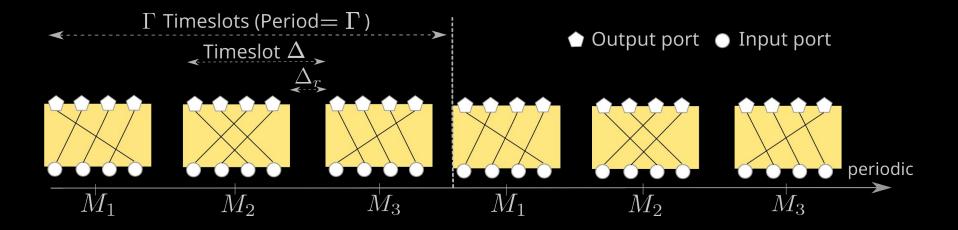


MARS

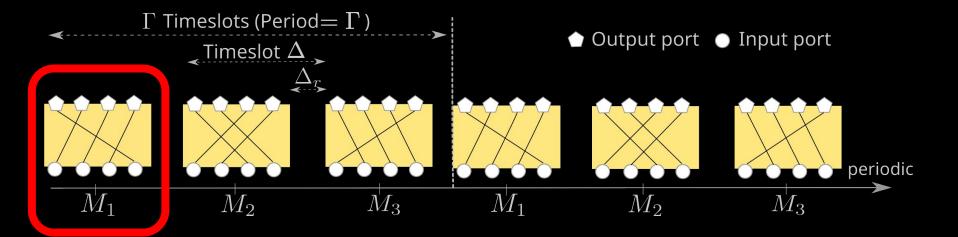




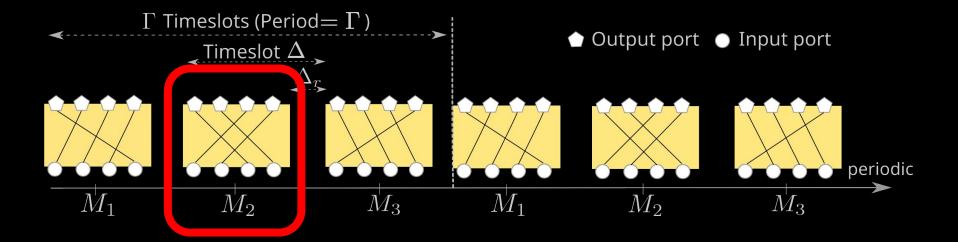




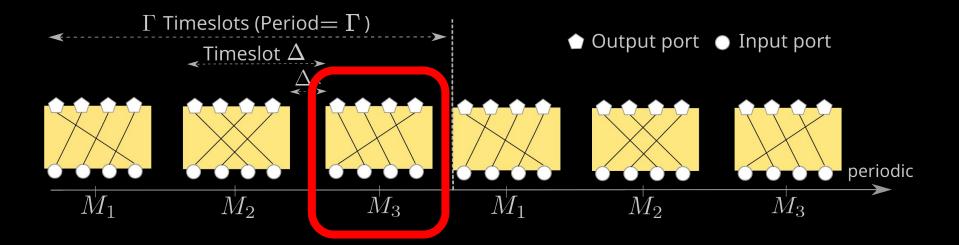




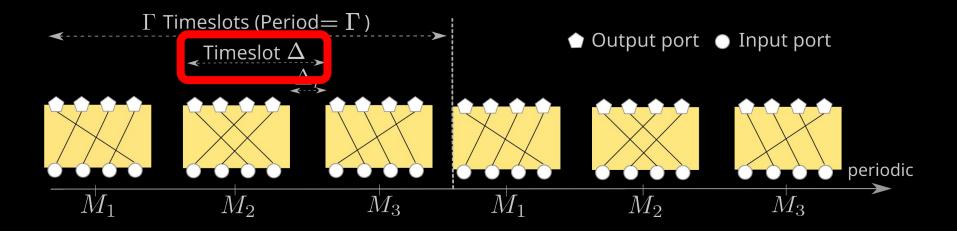




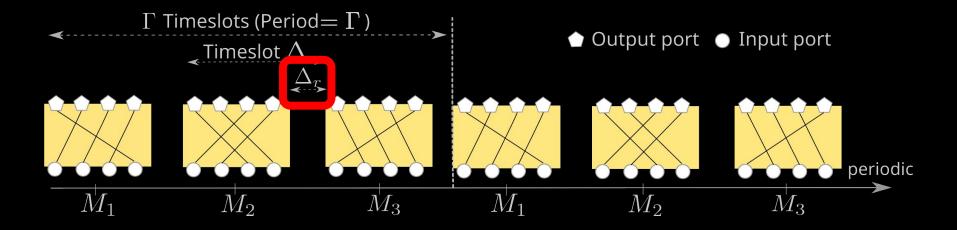




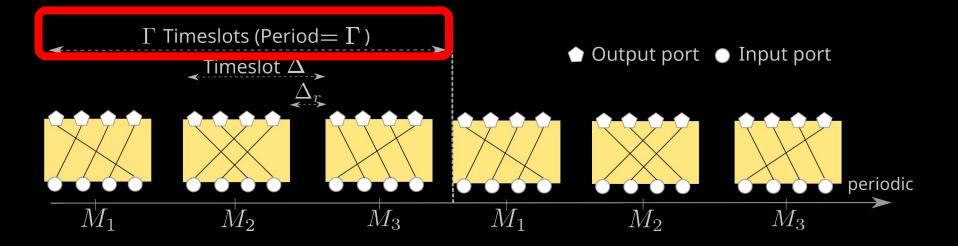




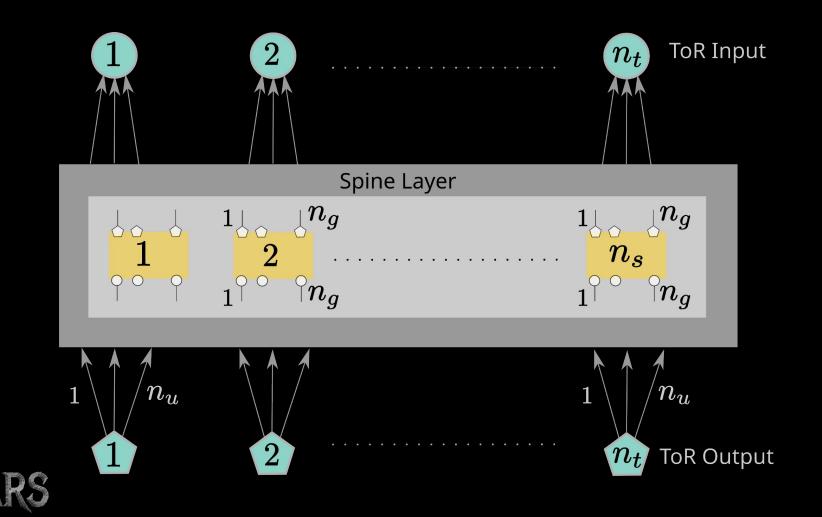


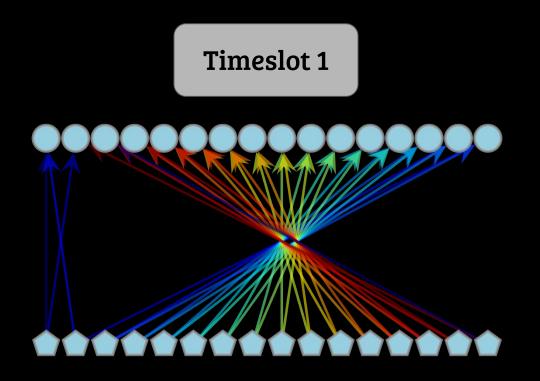




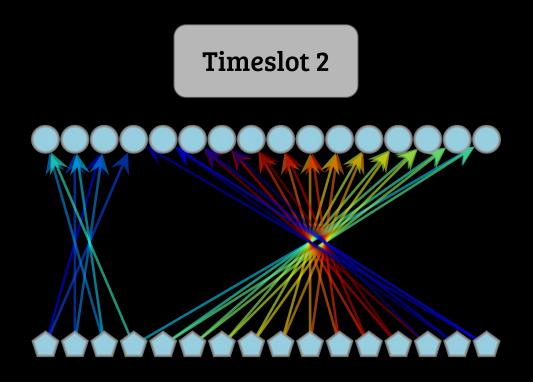




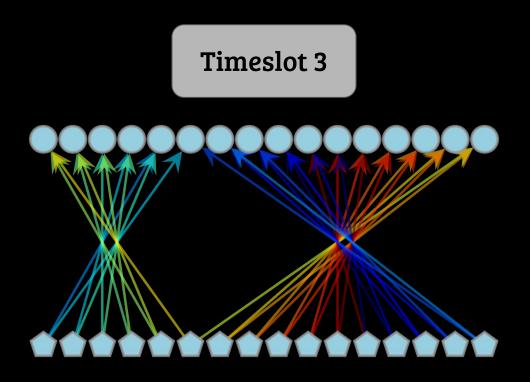




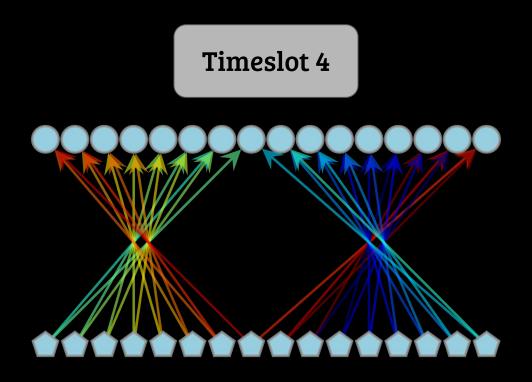




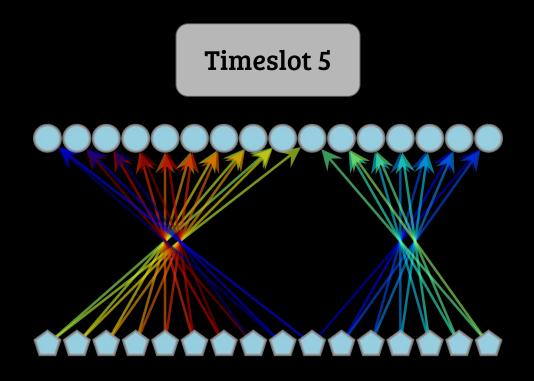




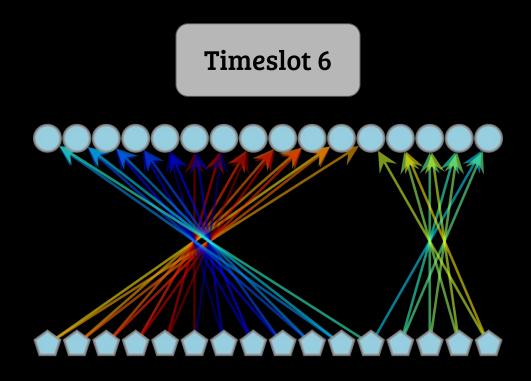




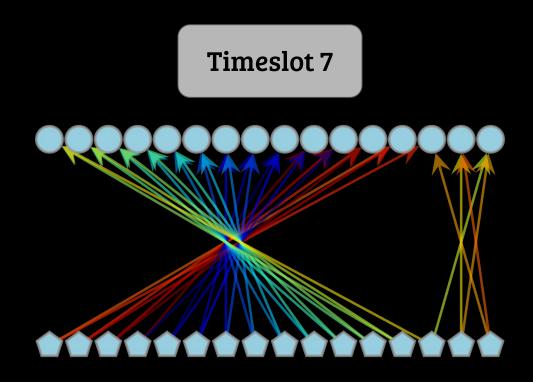




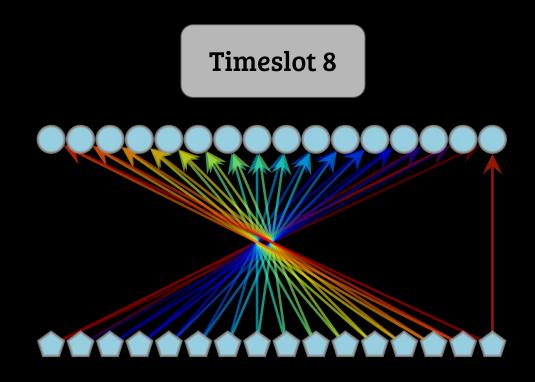






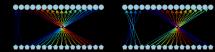




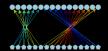


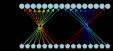


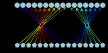
Periodic Graph

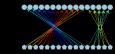


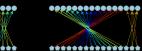


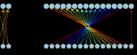














Optimal Topology

Input: Demand Matrix \mathcal{M}

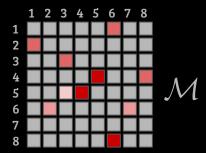
Available buffer size

Output: Periodic graph

Maximize: Throughput



Throughput

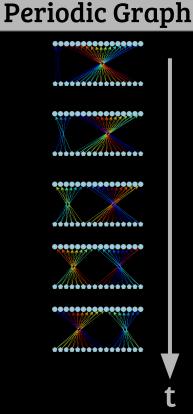


Demand Matrix

Throughput

 $X \quad \theta(\mathcal{M})$

Highest scaling factor such that the scaled demand θ (\mathcal{M}) is feasible in the periodic graph





Throughput of the Periodic Graph

Input: Periodic Graph *G*

Demand Matrix \mathcal{M}

Objective: Maximize $\theta(\mathcal{M})$

Output: $\theta(\mathcal{M})$ and a feasible *flow**

**subject to conservation, demand and capacity constraints*



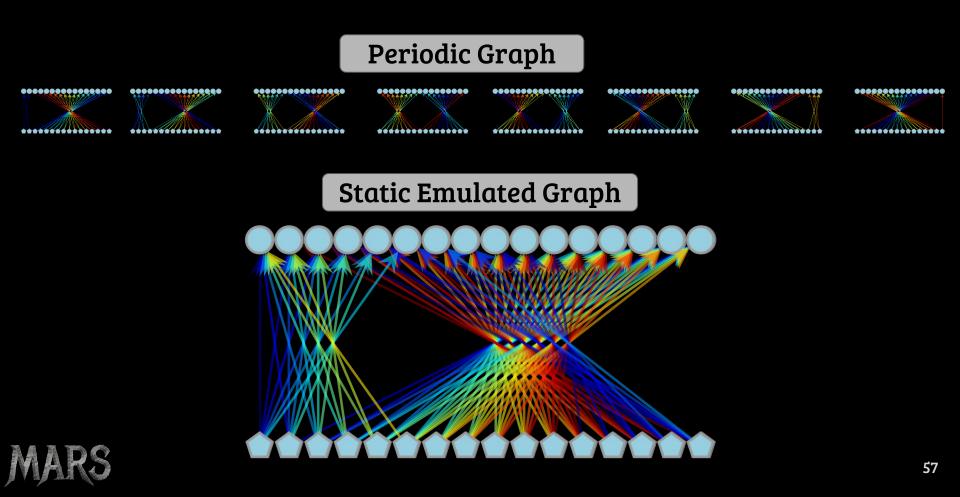
Theorem 1: Periodic Graph \iff Static Graph

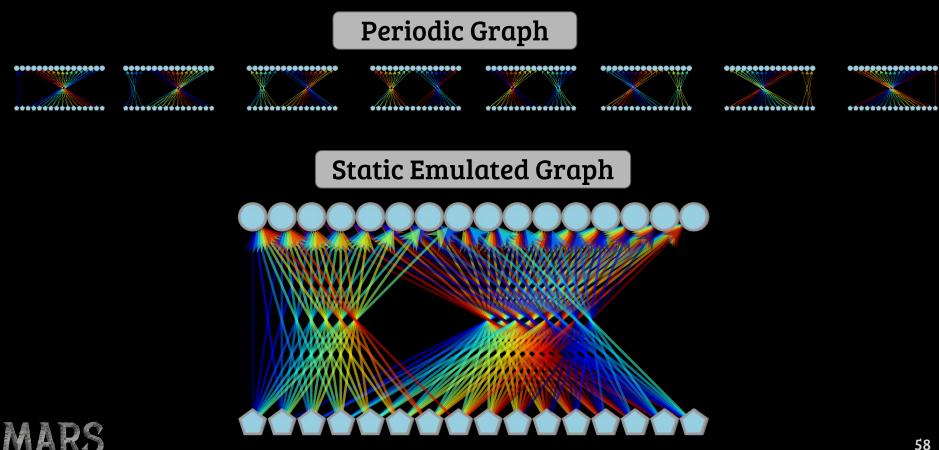
• The periodic graph has the same throughput as that of a static graph it emulates - *Static Emulated Graph*



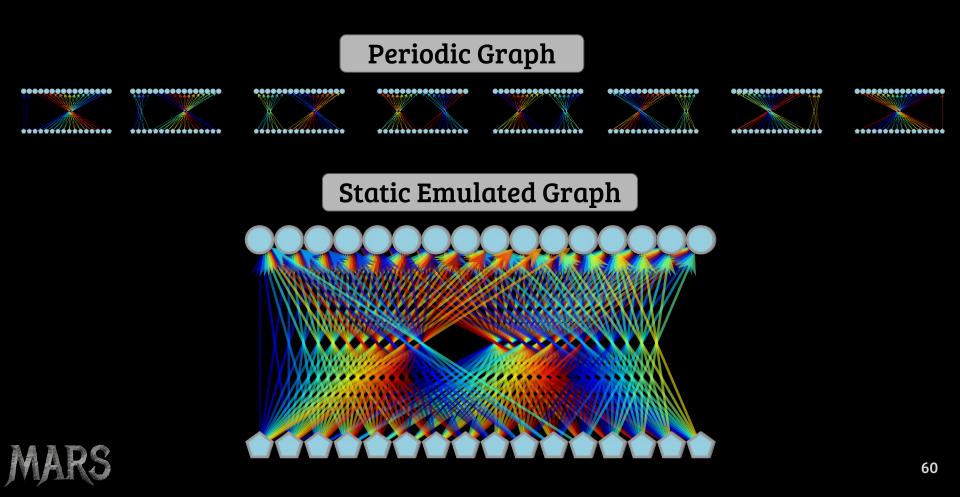
Periodic Graph ************ -----...... ********** -----********** -----************ -----************ ************ **Static Emulated Graph** 55

Periodic Graph -----************ ************ -----************ -----************ ************ **Static Emulated Graph** 56

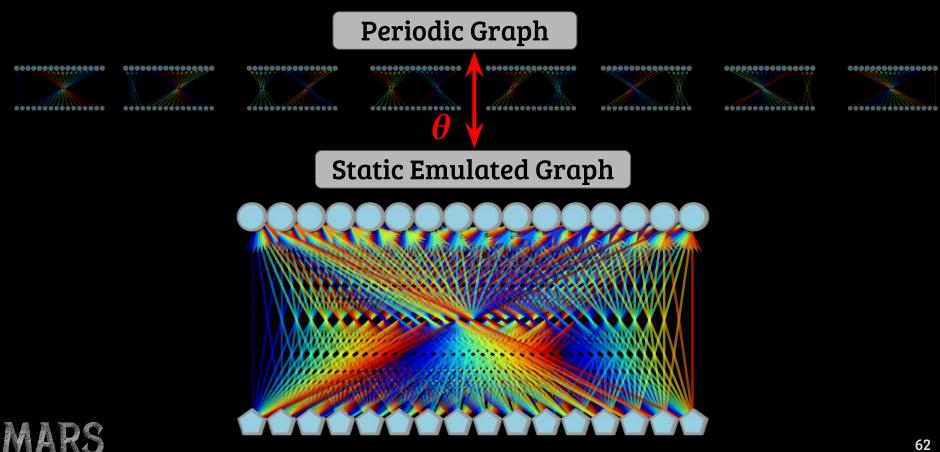




Periodic Graph -----************ ----********** ************ -----************ -----************ ************ Static Emulated Graph 59



Periodic Graph -----...... ************* -----********** ********** -----********** -----************ ************ Static Emulated Graph 61



Throughput of the Periodic Graph

Input: Periodic Graph *G*

Demand Matrix \mathcal{M}

Objective: Maximize $\theta(\mathcal{M})$

Output: $\theta(\mathcal{M})$ and a feasible *flow**

**subject to conservation, demand and capacity constraints*



Throughput of the Periodic Graph

Input: Periodic Graph& Static Emulated Graph G

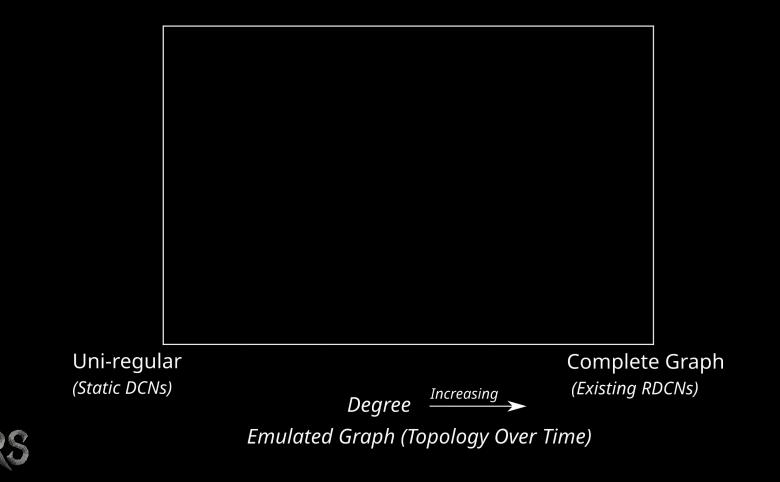
Demand Matrix ${\cal M}$

Objective: Maximize $\theta(\mathcal{M})$

Output: $\theta(\mathcal{M})$ and a feasible *flow**

**subject to conservation, demand and capacity constraints*





$$\theta(\mathcal{M}, F) \leq \frac{\hat{C}}{M \cdot \operatorname{ARL}(\mathcal{M}, F)}$$



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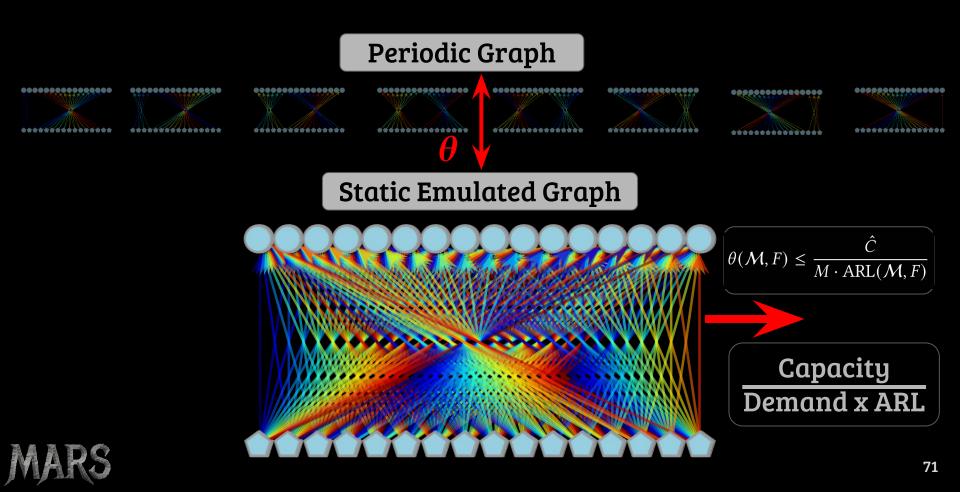


$$\theta(\mathcal{M}, F) \leq \frac{\hat{C}}{M \operatorname{ARL}(\mathcal{M}, F)}$$

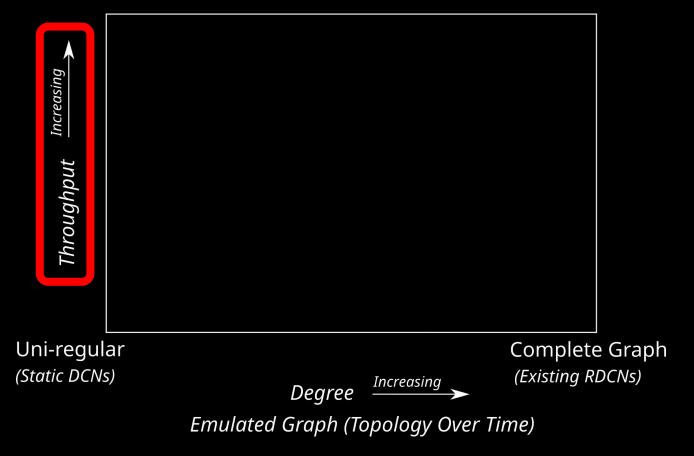


$$\theta(\mathcal{M}, F) \leq \frac{\hat{C}}{M \operatorname{ARL}(\mathcal{M}, F)}$$









Delay

Input: Periodic Graph *G*

Demand Matrix ${\cal M}$

Throughput heta and a feasible flow ${\mathcal F}$

Output: Minimum worst-case delay



- Delay bound is a function of:
 - Degree **d** of the emulated graph
 - Duration of the period $\mathbf{\Gamma} \cdot \mathbf{\Delta}$
 - Throughput **f**

$$L_{max} \ge \operatorname{ARD}(\mathcal{M}, \mathcal{F}) = \operatorname{ARL}(\mathcal{M}, \mathcal{F}) \cdot \Gamma \cdot \Delta$$
$$\ge \Omega\left(\frac{d \cdot \Delta}{n_u \cdot \theta(\mathcal{M}, \mathcal{F})}\right)$$



- Delay bound is a function of:
 - Degree **d** of the emulated graph
 - Duration of the period $\mathbf{\Gamma} \cdot \mathbf{\Delta}$
 - Throughput **f**

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- Delay bound is a function of:
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 - Duration of the period $\mathbf{\Gamma} \cdot \mathbf{\Delta}$
 - Throughput **f**

$$L_{max} \ge \operatorname{ARD}(\mathcal{M}, \mathcal{F}) = \operatorname{ARL}(\mathcal{M}, \mathcal{F}) \cdot \Gamma \cdot \Delta$$
$$\ge \Omega\left(\frac{d \Delta}{n_u \cdot \theta(\mathcal{M}, \mathcal{F})}\right)$$

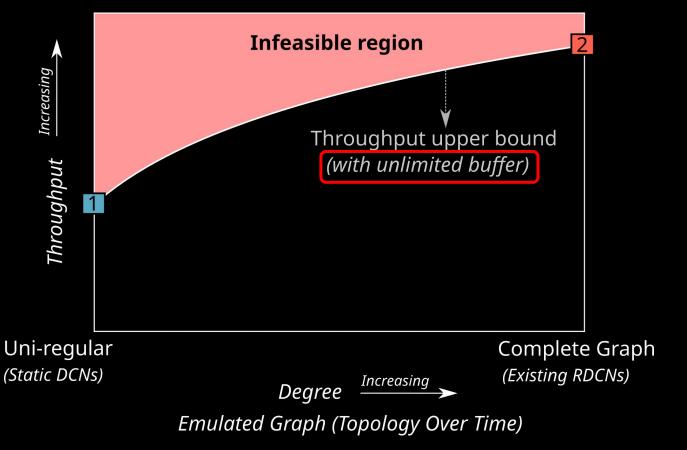


- Delay bound is a function of:
 - Degree **d** of the emulated graph
 - Duration of the period $\mathbf{\Gamma} \cdot \mathbf{\Delta}$
 - Throughput **f**

$$L_{max} \ge \operatorname{ARD}(\mathcal{M}, \mathcal{F}) = \operatorname{ARL}(\mathcal{M}, \mathcal{F}) \cdot \Gamma \cdot \Delta$$
$$\ge \Omega\left(\frac{d \cdot \Delta}{n_u \quad \theta(\mathcal{M}, \mathcal{F})}\right)$$







Buffer Requirements

Input: Periodic Graph *G*

Demand Matrix ${\cal M}$

Throughput heta and a feasible flow ${\mathcal F}$

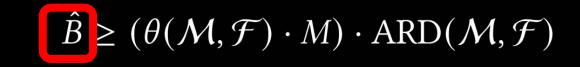
Output: Minimum required buffer to achieve throughput heta



• The required buffer is at least the *throughput delay product*

$\hat{B} \geq (\theta(\mathcal{M}, \mathcal{F}) \cdot M) \cdot \text{ARD}(\mathcal{M}, \mathcal{F})$







$$\hat{B} \ge (\theta(\mathcal{M}, \mathcal{F}) \cdot M) \cdot \operatorname{ARD}(\mathcal{M}, \mathcal{F})$$





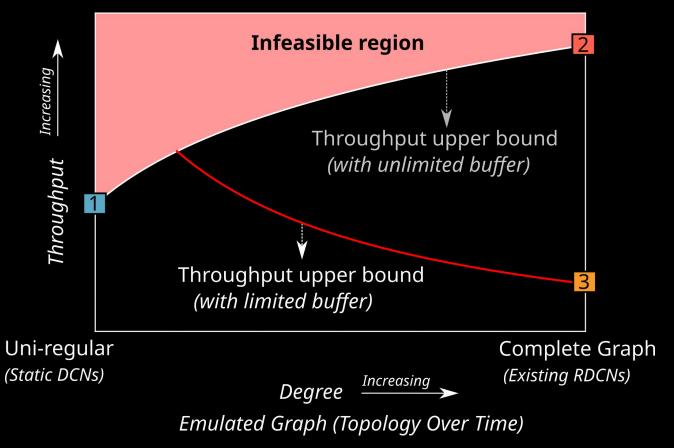


$$\hat{B} \ge (\theta(\mathcal{M}, \mathcal{F}) \cdot M) \cdot \text{ARD}(\mathcal{M}, \mathcal{F})$$

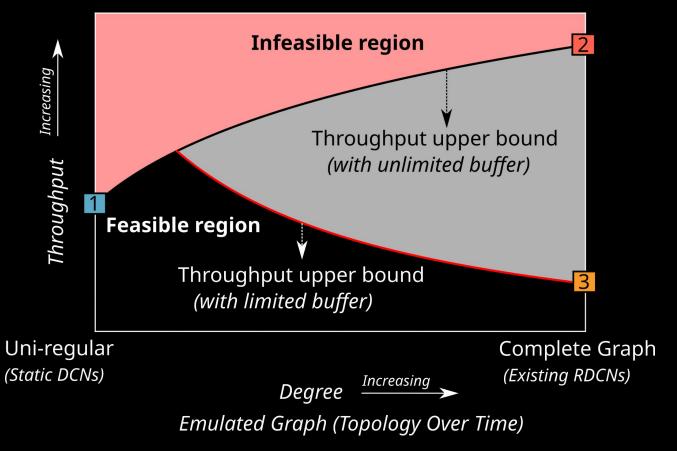
Buffer > Bandwidth x Delay











Goals

- Maximize Throughput
- Minimize Latency
- Minimize Buffer Requirements



Input: Periodic GraphG

Demand Matrix ${\cal M}$

Available buffer size B at each node

Output: Degree d of the emulated graph \longrightarrow Periodic graph **Maximize:** Throughput θ

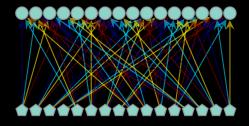


Output: $d = B/(c \cdot \Delta)$



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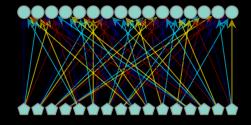
d-regular directed deBruijn graph





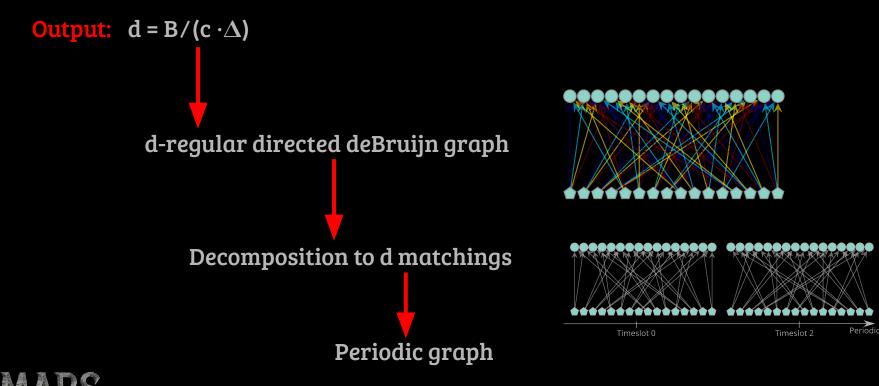
Output: $d = B/(c \cdot \Delta)$

d-regular directed deBruijn graph



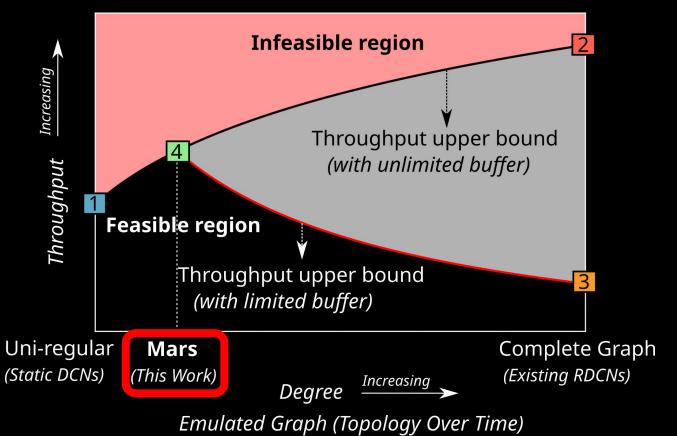
Decomposition to d matchings





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Optimal Topology Implications and Future Outlook

Output: $d = B/(c \cdot \Delta)$

If the reconfiguration technology (Δ) remains same:

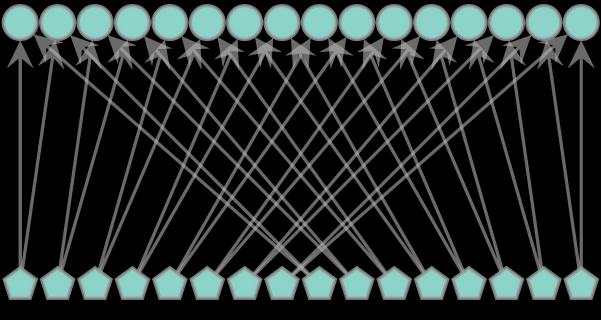
• Buffer sizes (B) must keep up with the increase in capacity (c)

If Buffer sizes (B) do not keep up:

- Increase in capacity (c) must be accompanied by decrease in reconfiguration times (Δ)
- If not, reducing the degree (d) of the emulated graph is inevitable to optimize throughput → eventually reaching the case of static topologies.



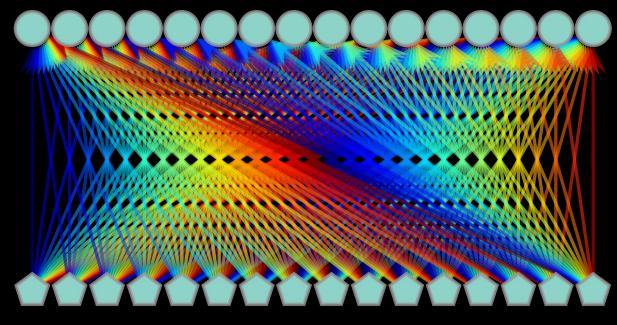
Static DCNs (uni-regular)



Low throughput but low delay and buffer requirements

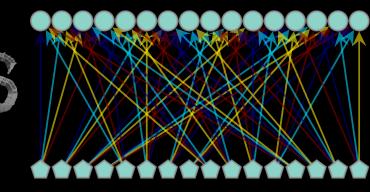


Existing RDCN designs (Emulating a complete graph)

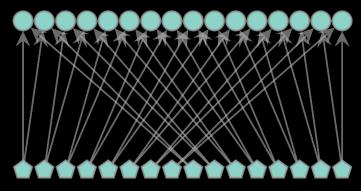


High throughput but high delay and buffer requirements

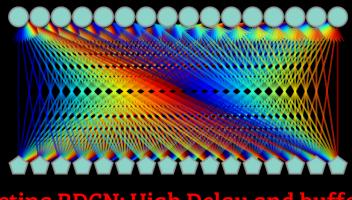




Near-optimal throughput within the available buffer



Static DCN: Low Throughput



Existing RDCN: High Delay and buffer



Thank you

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