

C3PO: Computation Congestion Control (PrOactive)

an algorithm for dynamic diffusion of ephemeral in-network services

Liang Wang, Mario Almeida*, Jeremy Blackburn*, Jon Crowcroft

University of Cambridge, UK

Telefonica Research, ES*

Presented by Suzan Bayhan (TU Berlin, Germany)



Question: Where Computations Happen

- "Fat client and thin server" or "thin client and fat server".
- Trend changes driven by the changes in *usage pattern, advances in hardware and software technologies*, and even new *business models*.
- Nowadays, both fat clients and fat servers: *high processing power and storage capacity*
- Still difficult to keep their pace with the ever-growing user demands.

Observation I: Pervasive Mobile Apps

Pervasive mobile clients have given birth to complex mobile apps. These apps are continuously generating, disseminating, consuming, and processing all kinds of information, in order to provide us convenient daily services.

Observation II: Battery Is The Bottleneck

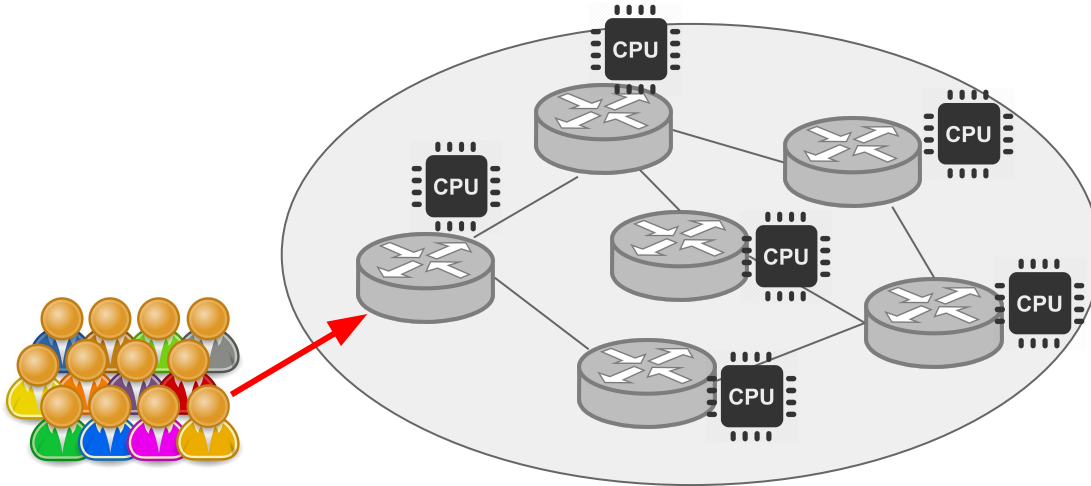
- Unfortunately, given current battery technology, these demanding apps impose a huge burden on energy constrained devices.
- Power hogging apps are responsible for **41%** degradation of battery life on average.
- Even popular ones such as social networks and instant messaging apps (e.g., Facebook and Skype) can drain a device's battery up to **9X** faster due only to maintaining an online presence.

Observation III: MiddleBoxes Grow Stronger

- Quite different from a decade ago, network middleboxes are no longer simple devices which only forward packets.
- ISPs' own network services have been shifting from specialized servers to generic hardware with the adoption of the NFV paradigm.
- E.g., Telefonica is shifting 30% of their infrastructure to NFV by 2016. Other providers such as AT&T, Vodafone, NTT Docomo, and China Mobile.

There are many in-network resources we can exploit. Besides, many of them are underutilised.

A New Paradigm: In-network Service Execution



- Ubiquitous in-network service execution
- Example services: ????

Challenge: How To Avoid Congestions

Because service execution consumes multiple resources on a router, especially demands CPU cycles for computation intensive tasks, it introduces a new type of "congestion" in a network - "computation congestion".

An obvious and important question:

- How to avoid such congestions?

i.e., what would be the mechanism of "Computation Congestion Control" for those ephemeral computation-intensive in-network services?

How Is It Different From Previous Settings

Traffic congestion: the solutions usually either

- try to reduce the transmission rate or
- take advantage of multiple paths.
- congestions are avoided by the cooperation of both communication ends.

But: In-network services do not necessarily impose a point-to-point paradigm!

Load balancing: a cluster often has a regular structure, i.e., network topology, central coordination, homogeneous configurations, uniform demands, and etc. Therefore, fully centralised control is the norm in the cluster. The jobs are often able to tolerate long scheduling delay.

Characteristics Of Our Context

Our context in an ISP network is more complicated:

1. the underlying topology is not regular
2. the node configurations can be heterogeneous
3. demands distribution is highly skewed hence the resources in a neighbourhood needs to be well utilised
4. central coordination is often expensive and reduces responsiveness of those networked nodes
5. services (and clients) are intolerable to long scheduling delay.

Our Solution:

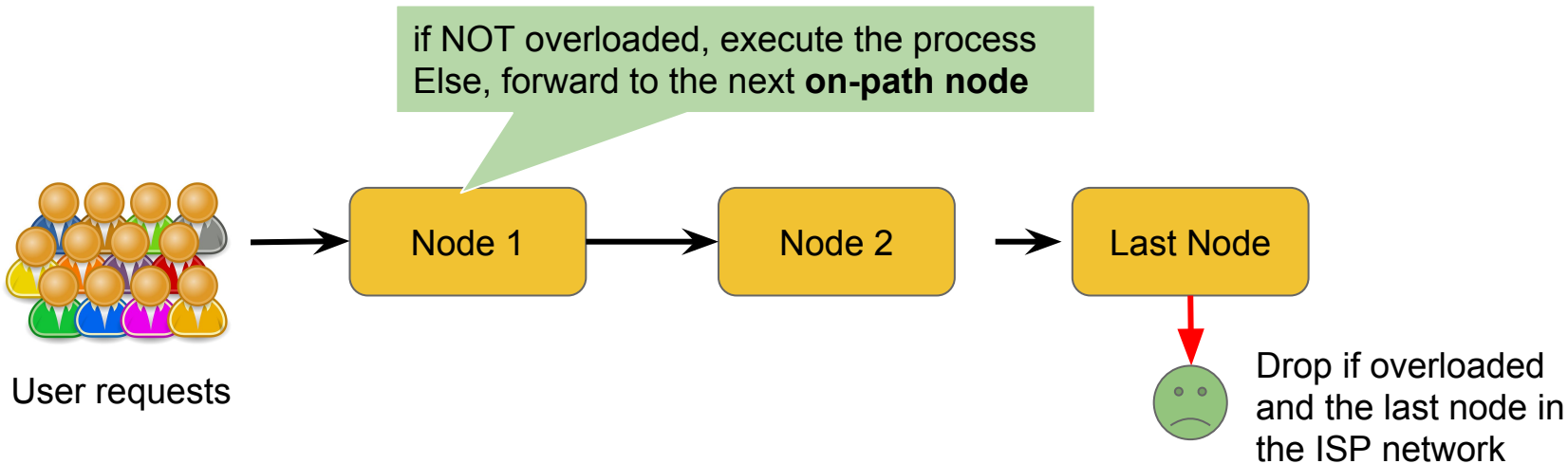
C3PO: Computation Congestion Control (PrOactive)

C3PO: a low-complexity **distributed** load balancer based on the **pessimistic prediction** of the service queue.

Why do we go for a distributed one instead of a centralised solver?

- A central solver needs global knowledge of all the nodes in a network
- the optimal strategy needs to be calculated periodically given the dynamic nature of a network and traffic
- there is a single point of failure
- there is often only marginal improvements over a smartly designed heuristic.

Two Strategies Are Studied: Passive Control



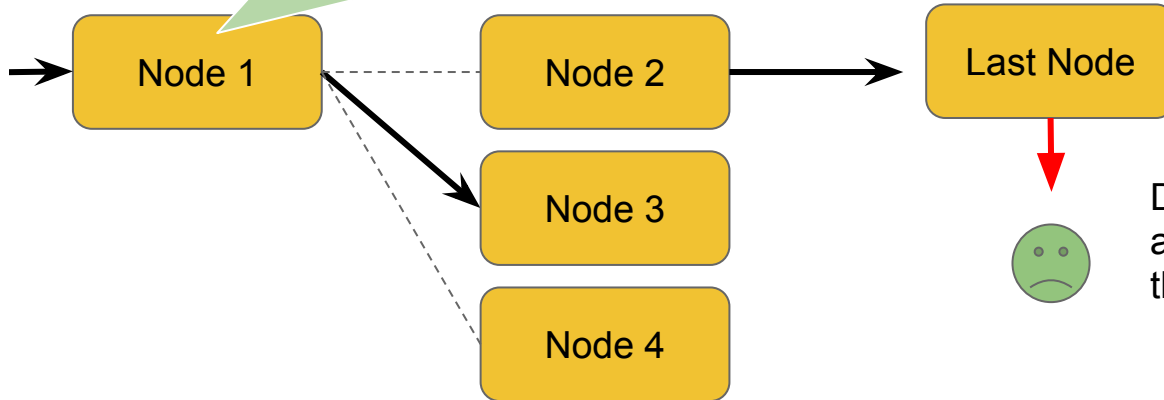
- + Very simple
- Reactive

Proactive Control

Estimate the request arrival rate
Estimate potential consumption
If may become overloaded,
forward to the neighbor with **the least load**
Else, execute the process

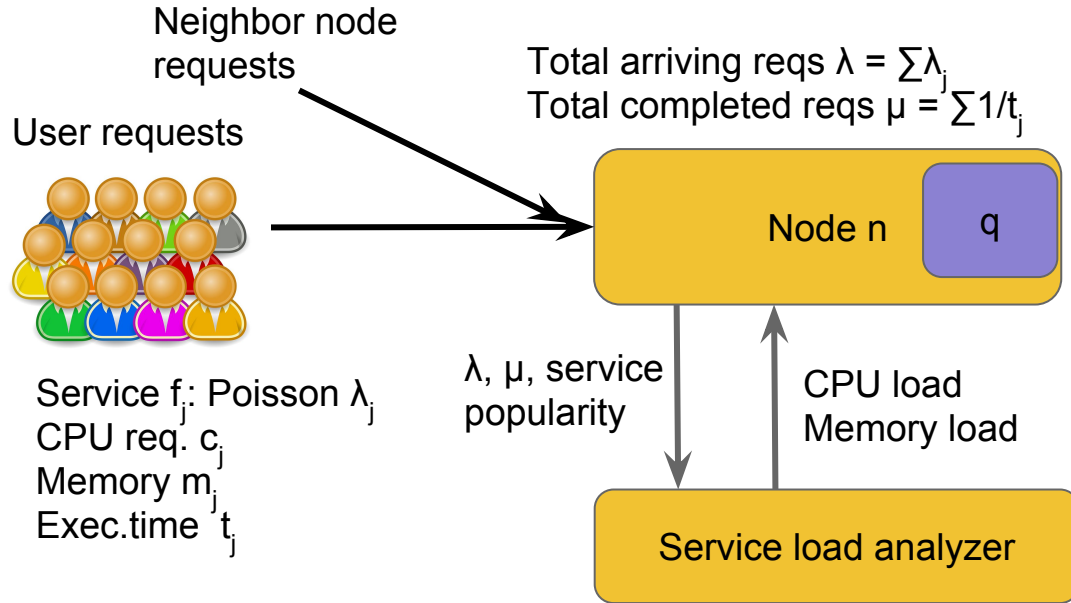


User requests



- + Proactive, anticipatory
- Requires 1-hop information_{1,2} exchange

Proactive Control: A Closer Look



Birth-death system

For a stable system,
workload < node capacity

Probabilistically (q) execute requests to ensure stability

Tune q based on expected load

Intuitive Explanations of C3PO

Stationary analysis of M/M/1-PS model on the service queue on a router.

Intuitively:

- Try to estimate the future incoming request rate based on the previous observations,
- If a router thinks it is going to be overloaded based on its capacity and service requirements, it only probabilistically executes some of the service requests.
- When estimating the future request rate, a router takes into account the second-order information (i.e., the derivative) of the rate. Essentially, it is smoothing. But it only considers the positive derivatives (line 4 in the code), so it is “pessimistic”, therefore “proactive”. Why :)

Proactive Strategy: C3PO

Algorithm 1 C3PO - Proactive Computation Control

```
1: void on_arrival (request  $r$ ):
2:    $\text{buf}_\lambda[i] \leftarrow \text{timestamp}(r)$ 
3:    $\lambda \leftarrow \text{mean\_rate}(\text{buf}_\lambda)$ 
4:    $\Delta\lambda \leftarrow \max(0, \lambda - \lambda')$ 
5:    $\lambda \leftarrow \lambda + \Delta\lambda$ 
6:    $q \leftarrow \text{eq.4}(\lambda, \mu, c', c'', m', m'')$ 
7:   if  $\text{draw\_uniform}([0,1]) < q$  then execute ( $r$ )
8:   else forward_to_lightest_load_node ( $r$ )
9:    $i \leftarrow (i + 1) \bmod k$ 
10:  if  $i == 0$  then  $\lambda' \leftarrow 0.5 \times (\lambda' + \lambda - \Delta\lambda)$ 
11:
12: void on_complete (function  $s$ ):
13:   $\text{buf}_\mu[i] \leftarrow \text{execution\_time}(s)$ 
14:   $\text{buf}_{c''}[i] \leftarrow \text{CPU\_consumption}(s)$ 
15:   $\text{buf}_{m''}[i] \leftarrow \text{memory\_consumption}(s)$ 
16:   $i \leftarrow (i + 1) \bmod k$ 
17:  if  $i == 0$  then
18:     $\mu \leftarrow 0.5 \times (\mu + \text{mean}(\text{buf}_\mu))^{-1}$ 
19:     $c'' \leftarrow 0.5 \times (c'' + \text{mean}(\text{buf}_{c''}))$ 
20:     $m'' \leftarrow 0.5 \times (m'' + \text{mean}(\text{buf}_{m''}))$ 
21:  forward_result ( $s$ )
```

- The algorithm is simple yet effective. We need to make sure the load balancing itself will not cause too much overhead.
 - The algorithm maintains four circular buffer with fixed size.
 - **on_arrival** and **on_complete** two functions need to be performed whenever a request arrives or finished.
- Overall complexity**
- The “proactiveness” is achieved by “being conservative”. Technically, by smoothing the load curve, or getting the derivative of the load increasing rate.

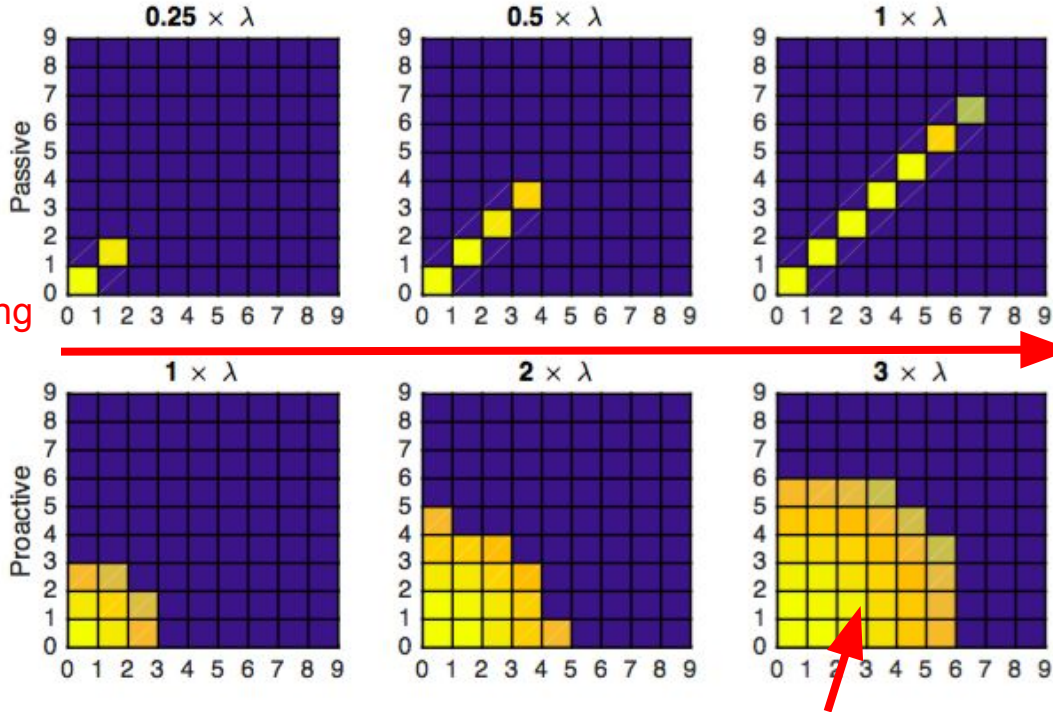
Implementation Details

- We use four fixed-size circular buffers instead of fixed time window to prevent the memory usage from being subject to service arrival/completion rate. The reason is the number of arrived requests can vary in a fixed time window.
- Parameter k (buffer size) represents a trade-off between **stability** and **responsiveness**.
 - Larger $k \rightarrow$ more stable estimates (longer history considered)
 - Smaller $k \rightarrow$ higher responsiveness to the changes in λ and μ .
- Although λ needs to be calculated at every request arrival (line 3), further optimisations can be easily done to reduce the complexity.

Evaluation Setup

- Aim: analyze how different strategy impacts **load distribution** as well as **latency, drop rate, responsiveness to jitters**.
- We test three strategies on both synthetic and realistic topologies using Icarus simulator
 - None (only edge routers execute services),
 - Passive,
 - Proactive
- Poisson request stream with $\lambda = 1000$ as arrival rate
- We assume CPU is the first bottleneck in the system for computation intensive services, and only present the results of using Exodus network in the following.

C3PO Exploits Its Neighbourhood



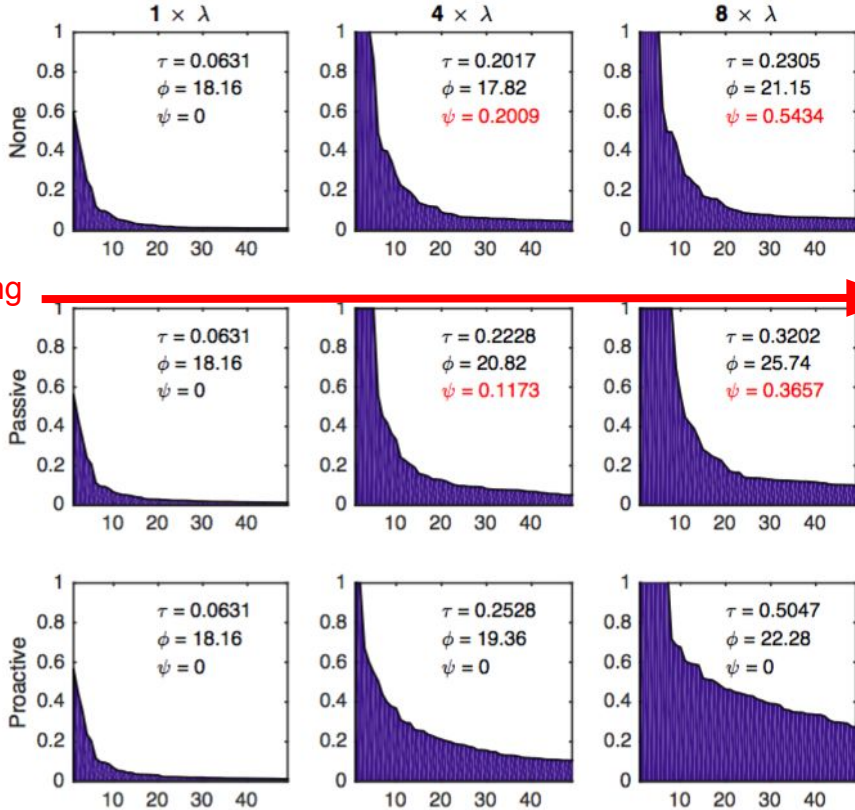
Client connected to router at (0,0)
Server connected to router at (9,9).

Proactive is more capable at utilizing the nearby resources within its neighborhood, leading to *better load balancing and smaller latency*.

On-path routers: a few due to small-world structure of networks

Yellow: high load
18

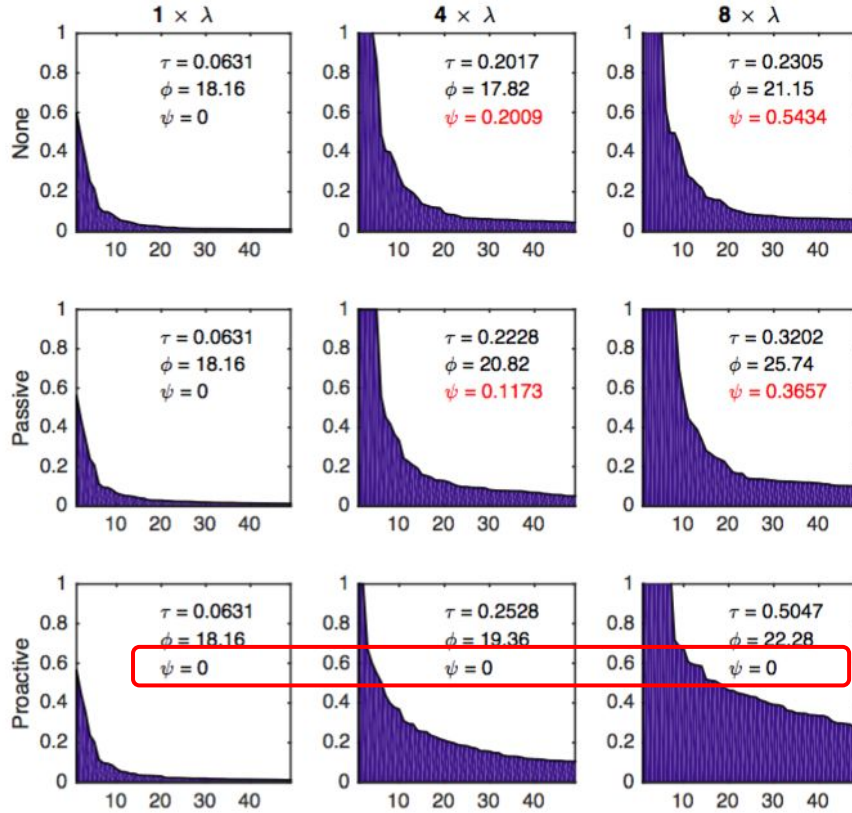
Scalable to Workload on Exodus ISP network



x-axis is node index and y-axis is load.
Top 50 nodes of the heaviest load are sorted in decreasing order

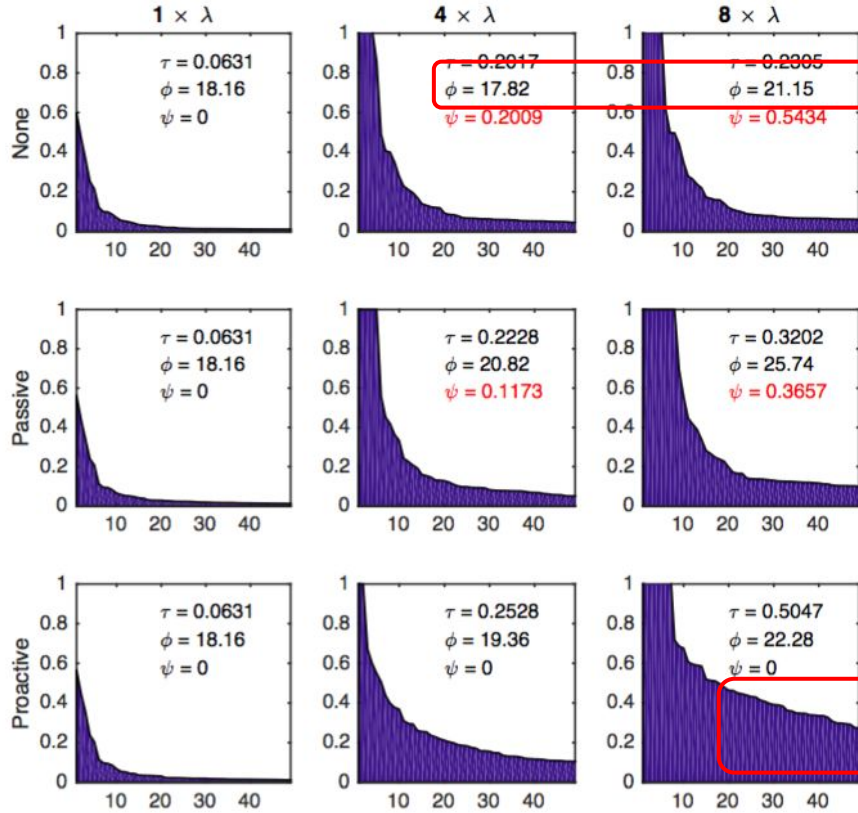
τ : average load;
 ϕ : average latency (in *ms*);
 ψ : ratio of dropped requests.

Scalable to Workload on Exodus ISP network



Proactive: No service drops due to load balancing capability

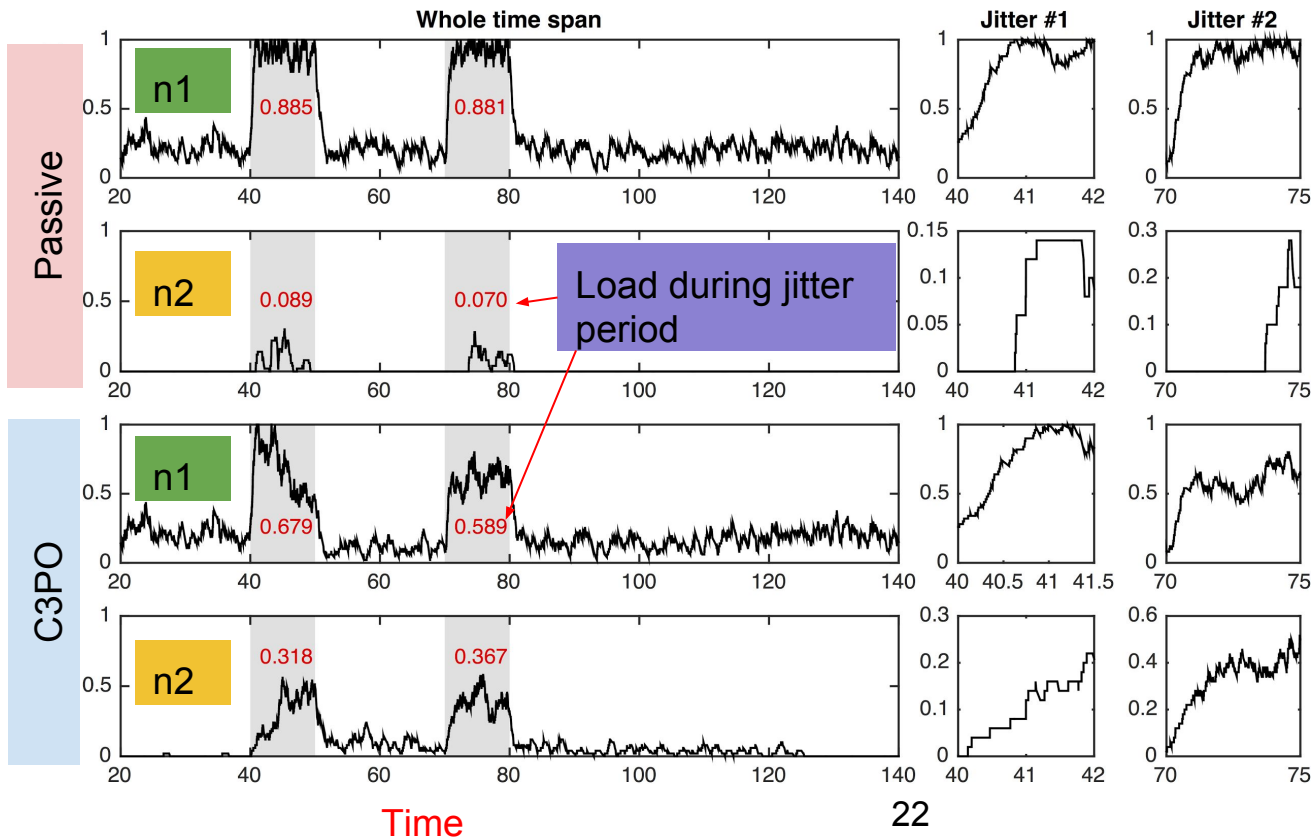
Scalable to Workload on Exodus ISP network



Lower latency because only edge routers execute services

Heavy Tail proves the success of load balancing in the network

Responsiveness to Jitters



Simple line topology:
client \rightarrow $n1$ \rightarrow $n2$ \rightarrow server.

Jitters (6λ arrival for 10ms) are injected at time 40 ms and 70 ms.

C3PO balances the load on $n1$ and $n2$ whereas Passive cannot!

Conclusion

- We studied two control strategies (Passive and Proactive).
- Based on the Proactive control, we proposed a fully distributed, low complexity, and responsive load controller to avoid potential computation congestions when executing in-network services.
- Our results showed that
 - the proposed solution C3PO can effectively take advantage of available resources in a neighbourhood to balance the service load and further reduce service latency and request drop rate.

Thank you. Questions?

For more information: Liang Wang
liang.wang@cl.cam.ac.uk