6.888 Lecture 5: Flow Scheduling

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Datacenter Transport

Goal: Complete flows quickly / meet deadlines

Short flows
(e.g., query, coordination) → Low Latency

Large flows
(e.g., data update, backup) → High Throughput
Low Latency Congestion Control (DCTCP, RCP, XCP, ...)

Keep network queues small (at high throughput)

Implicitly prioritize mice

Can we do better?
The Opportunity

Many DC apps/platforms know flow size or deadlines in advance

- Key/value stores
- Data processing
- Web search
Amy: “Many papers that propose new network protocols for datacenter networks (such as PDQ and pFabric) argue that these will improve "user experience for web services". However, none seem to evaluate the impact of their proposed scheme on user experience... I remain skeptical that small protocol changes really have drastic effects on end-to-end metrics such as page load times, which are typically measured in seconds rather than in microseconds.”
DC transport = Flow scheduling on giant switch

Objective?
- Minimize avg FCT
- Minimize missed deadlines

ingress & egress capacity constraints
Example: Minimize Avg FCT

<table>
<thead>
<tr>
<th>Flow</th>
<th>Size</th>
</tr>
</thead>
<tbody>
<tr>
<td>Flow A</td>
<td>1</td>
</tr>
<tr>
<td>Flow B</td>
<td>2</td>
</tr>
<tr>
<td>Flow C</td>
<td>3</td>
</tr>
</tbody>
</table>

arrive at the same time

share the same bottleneck link

 adapté from slide by Chi-Yao Hong (UIUC)
Example: Minimize Avg FCT

Fair sharing:
- 3, 5, 6
- mean: 4.67

Shortest flow first:
- 1, 3, 6
- mean: 3.33

Adapted from slide by Chi-Yao Hong (UIUC)
Optimal Flow Scheduling for Avg FCT

NP-hard for multi-link network [Bar-Noy et al.]

– Shortest Flow First: 2-approximation
How can we schedule flows based on flow criticality in a distributed way?

Some transmission order
PDQ

Several slides based on presentation by Chi-Yao Hong (UIUC)
PDQ: Distributed Explicit Rate Control

Sender — Switch — Switch — Receiver

Packet  hdr

criticality

rate = 10

Switch preferentially allocates bandwidth to critical flows

Traditional explicit rate control
Fair sharing (e.g., XCP, RCP)
Contrast with Traditional Explicit Rate Control

Traditional schemes (e.g. RCP, XCP) target fair sharing

- Each switch determines a “fair share” rate based on local congestion: \( R \leftarrow R - k \times \text{congestion-measure} \)
- Source use smallest rate advertised on their path
Challenges

PDQ switches need to agree on rate decisions

Low utilization during flow switching

Congestion and queue buildup

Paused flows need to know when to start
Challenge:
Switches need to agree on rate decisions

- What can go wrong without consensus?
- How do PDQ switches reach consensus?
- Why is “pauseby” needed?
Austin: “It is an interesting departure from AQM in that, with the concept of paused queues, PDQ seems to leverage senders as queue memory.”
Challenge:
Low utilization during flow switching

Goal:
How does PDQ avoid this?

Practice:
Early Start: Seamless flow switching

Throughput

Time

1

2 RTTs

Start next set of flows
Early Start: Seamless flow switching

Solution: rate controller at switches [XCP/TeXCP/D3]

increased queue
Discussion
Mean FCT

Mean flow completion time [Normalized to a lower bound]

Omniscient scheduler controls with zero control feedback delay
What if flow size not known?

Why does flow size estimation (criticality = bytes sent) work better for Pareto?
Other questions

**Fairness:** can long flows starve?

99% of jobs complete faster under SJF than under fair sharing

[Bansal, Harchol-Balter; SIGMETRICS’01]
Assumption: heavy-tailed flow distribution

**Resilience to error:** what if packet gets lost or flow information is inaccurate?

**Multipath:** does PDQ benefit from multipath?
pFabric
pFabric in 1 Slide

Packets carry a single priority #
  - e.g., prio = remaining flow size

pFabric Switches
  - Send highest priority / drop lowest priority pkts
  - Very small buffers (20-30KB for 10Gbps fabric)

pFabric Hosts
  - Send/retransmit aggressively
  - Minimal rate control: just prevent congestion collapse

Main Idea: Decouple scheduling from rate control
pFabric Switch

Boils down to a sort
- Essentially unlimited priorities
- Thought to be difficult in hardware

Existing switching only support 4-16 priorities

pFabric queues very small
- 51.2ns to find min/max of ~600 numbers
- Binary comparator tree: 10 clock cycles
- Current ASICs: clock ~ 1ns
pFabric Rate Control

Minimal version of TCP algorithm

1. Start at line-rate
   - Initial window larger than BDP

2. No retransmission timeout estimation
   - Fixed RTO at small multiple of round-trip time

3. Reduce window size upon packet drops
   - Window increase same as TCP (slow start, congestion avoidance, ...)

4. After multiple consecutive timeouts, enter “probe mode”
   - Probe mode sends min. size packets until first ACK

What about queue buildup?
Why window control?
Why does pFabric work?

Key invariant:
At any instant, have the highest priority packet (according to ideal algorithm) available at the switch.

Priority scheduling
- High priority packets traverse fabric as quickly as possible

What about dropped packets?
- Lowest priority → not needed till all other packets depart
- Buffer > BDP → enough time (> RTT) to retransmit
Discussion
Overall Mean FCT

![Overall Mean FCT Graph]

- **Ideal**
- **pFabric**
- **PDQ**
- **DCTCP**
- **TCP-DropTail**

**FCT (normalized to optimal in idle fabric)**

**Load**

FCT values increase with load for all protocols, with TCP-DropTail showing the highest FCT as load increases.
Mice FCT (<100KB)

**Average**

**99th Percentile**

- **Ideal**
- **pFabric**
- **PDQ**
- **DCTCP**
- **TCP-DropTail**

![Graph showing performance comparison between different load balancing algorithms](image)
Elephant FCT (>10MB)

Average

Why the gap?
Loss Rate vs Packet Priority (at 80% load)

* Loss rate at other hops is negligible

Almost all packet loss is for large (latency-insensitive) flows
Next Time:
Multi-Tenant Performance Isolation

Application-Driven Bandwidth Guarantees in Datacenters

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Joon-Myung Kang  Puneet Sharma
HP Labs, "University of Edinburgh, "Databricks

ABSTRACT
Providing bandwidth guarantees to specific applications is becoming increasingly important as applications compete for shared cloud network resources. We present CloudMirror, a solution that provides bandwidth guarantees to cloud applications based on a new network abstraction and workload placement algorithm. An effective network abstraction would enable applications to easily and accurately specify their requirements, while simultaneously enabling the infrastructure to provision resources efficiently for deployed applications. Prior research has approached the bandwidth guarantee specification by using abstractions that resemble physical network topologies. We present a contrasting approach of deriving a network abstraction based on application communication structure, called Tenant Application Graph or TAG. CloudMirror also incorporates a new workload placement algorithm that efficiently meets bandwidth requirements specified by TAGs while factoring in availability considerations. Extensive simulations using real application traces and datacenter topologies show that CloudMirror can handle 40% more bandwidth demand than the state of the art (e.g., the Octopus system), while improving high availability from 20% to 70%.

Categories and Subject Descriptors: C.2.3 [Computer-Communication Networks]: Network Operations

Today, it is easy to share and virtualize compute and storage resources effectively. In contrast, implementing network virtualization with bandwidth guarantees on a shared network infrastructure is an inherently complex and challenging task [3–7, 18, 45], which requires three key technologies: 1) An easy-to-use network abstraction model for tenants to accurately express their bandwidth requirements; 2) A workload placement algorithm that efficiently allocates datacenter resources to meet the tenant requests, and 3) A scalable runtime mechanism to enforce the bandwidth guarantees and utilize unused bandwidth efficiently. In this paper, we propose CloudMirror, a solution that combines a new network abstraction with a new workload placement algorithm, while leveraging an existing mechanism [7] for enforcing guarantees.

An effective network abstraction model serves two purposes. One purpose is for tenants to specify their network requirements in a simple and intuitive yet accurate manner. The other purpose is to facilitate easy translation of these requirements to an efficient deployment on the low level infrastructure components. Most prior work, e.g., [4–9], has designed abstractions for specifying bandwidth guarantees that can be expressed as idealized physical network models, e.g., non-blocking switch (hose) [8] or two-level tree (hierarchical hose) [4, 6]. This is a natural approach since, for example, in cloud computing, tenants want to have the illusion of running their applications on dedicated hardware. But