

# Capacity Provisioning for Schedulers with Tiny Buffers

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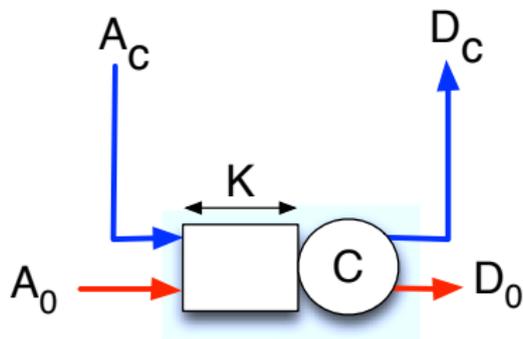
**Joint work with:**

J. Liebeherr

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# Resource Provisioning for Link Schedulers



- Input traffic: Through flows  $A_0$  and cross flows  $A_c$
- Output traffic: Through flows  $D_0$  and cross flows  $D_c$
- Link capacity  $C$ , buffer size  $K$
- Backlog  $b_0(t)$  and delay  $d_0(t)$  of the through flows at time  $t$

Size  $C$  and  $K$  such that:

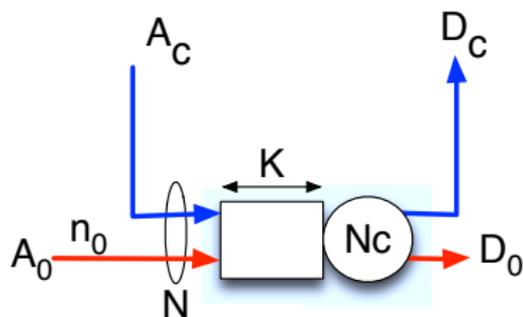
$P\{b_0(t) > K\} \leq \varepsilon^*$  and/or  $P\{d_0(t) > \bar{d}\} \leq \varepsilon^*$ , where  $\bar{d}$  is the delay bound

# Towards Small Buffers

There are arguments in favour of small buffers:

- Small buffers enable fast memory technologies (e.g., SRAM).  
(*Enachescu et al.' 05*)
- Small buffers might even mitigate traffic burstiness.  
(*Likhanov and Mazumdar' 98*), (*Mao and Panwar' 01*)
- In case of many sources, adding small buffers satisfies loss probability.  
(*Mao and Panwar' 01*)

# Asymptotic Observations



Define

- $c$ : per-flow capacity
- $\bar{a}$ : per-flow average rate  $:= \lim_{t \rightarrow \infty} \frac{1}{N} \frac{A_0(t) + A_c(t)}{t}$

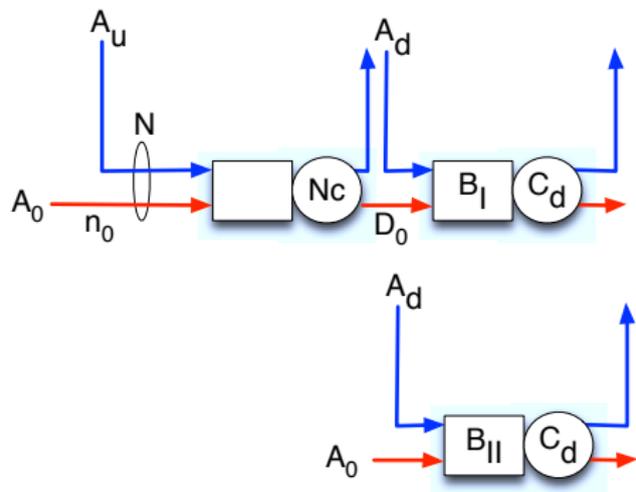
Given: a loss probability constraint (using large deviation techniques)

For any work-conserving scheduling  $\lim_{N \rightarrow \infty} c \rightarrow \bar{a}$ . (Eun and Shroff' 05)

The results hold for small buffers (i.e.,  $O(1)$ )  $\Rightarrow$  network decomposition

# Network Decomposition in an Asymptotic Regime

- Convergence of  $D_0$  to  $A_0$ :  
(Wischik' 99), (Ying et al.' 94)
- Convergence of  $B_I$  to  $B_{II}$ :  
(Eun and Shroff' 05), (Ciucu and Hohlfled' 09), (Ciucu and Liebeherr' 09)



# Does Link Scheduling Matter if $N$ is Finite?

Some existing non-asymptotic results for schedulers:

- $D_0 \rightarrow A_0$  for FIFO scheduling even when  $N$  is few hundreds under some statistical independence assumptions. (*Ciucu and Liebeherr' 09*)
- A non-asymptotic capacity size is computed for a given per-flow delay bound constraint in a FIFO scheduler. It scales by  $c = O(\frac{1}{N})$ . (*Ciucu and Hohlfled' 09*)

## Open question:

How does link scheduling impact capacity requirement and decomposition for finite  $N$ ?

# Contributions

We show that for finite  $N$ , the choice of link scheduling has a big impact on

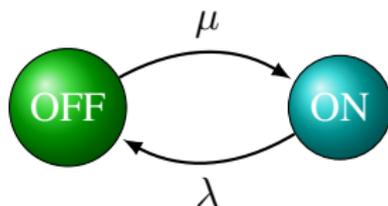
- Buffer overflow probability
- Capacity provisioning
- Viability of network decomposition

## In particular

$c - \bar{a}$  ranges from  $O\left(\sqrt{\frac{\log N}{N}}\right)$  to  $O\left(\frac{1}{N}\right)$  depending on the scheduling algorithm.

# Traffic Source (MMOO)

Markov-modulated On-Off (MMOO) source:



- $P$  Kbps in ON state, idle in OFF state
- Average time to return to the same state:  $T^* = \frac{\lambda + \mu}{\lambda \mu}$
- The larger the  $T^*$ , the more bursty the traffic

# Exponentially Bounded Burstiness

Exponentially Bounded Burstiness (EBB) sources (Yaron, Sidi'93)

An arrival process  $A$  is EBB with parameters  $(M, \rho, \alpha)$  if for any  $s \leq t$

$$P(A(s, t) > \rho(t - s) + \sigma) \leq Me^{-\alpha\sigma} := \varepsilon(\sigma).$$

We write it by  $A \sim (M, \rho, \alpha)$ .

**Suppose:**  $A$  is the aggregate of  $n$  iid MMOO flows with parameters  $\lambda$ ,  $\mu$ , and  $P$ .

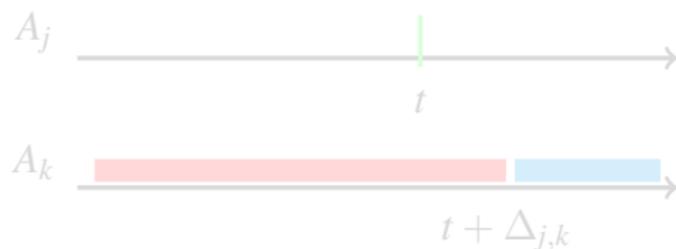
Then,  $A \sim (1, nr(\alpha), \alpha)$  for any  $\alpha \geq 0$ , with

$$r(\alpha) = \frac{1}{2\alpha}(P\alpha - \lambda - \mu + \sqrt{(P\alpha - \mu + \lambda)^2 + 4\mu\lambda}).$$

We use this flexibility (a family of EBB characterizations) to get new insights.

# $\Delta$ -Schedulers

A scheduler whose operation is entirely determined by a matrix of constants  $(\Delta_{j,k})_{j,k \in \mathcal{N}}$ .



- The followings are  $\Delta$ -schedulers:

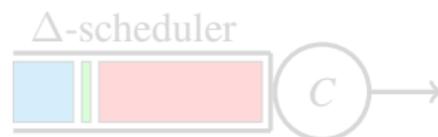
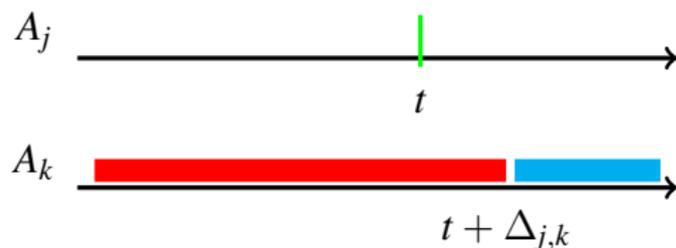
- ▶ FIFO:  $\Delta_{j,k} = 0$
- ▶ SP, BMux:  $\Delta_{j,k} = \begin{cases} -\infty \\ +\infty \end{cases}$
- ▶ EDF:  $\Delta_{j,k} = d_j^* - d_k^*$

if flow  $j$  has higher priority  
if flow  $k$  has higher priority

- GPS is not a  $\Delta$ -scheduler.

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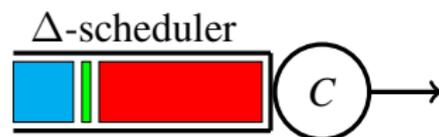
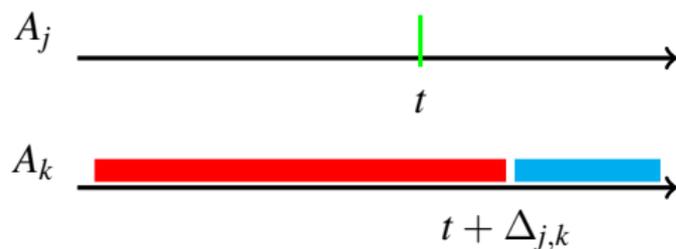
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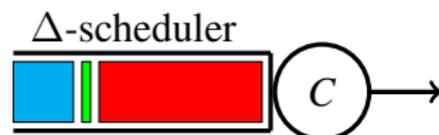
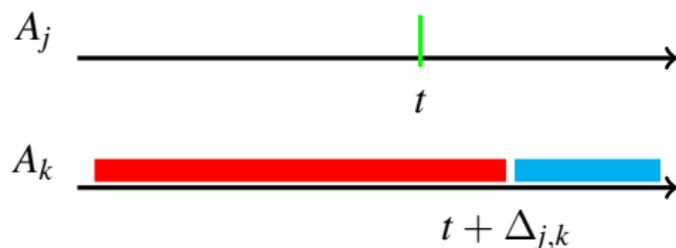
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# A Backlog Bound for EBB flows in $\Delta$ -Schedulers

## A backlog bound for $\Delta$ -schedulers [Ghiassi, Liebeherr, Burchard' 11]

- $A_0 \sim (M_0, \rho_0, \alpha_0)$  and  $A_c \sim (M_c, \rho_c, \alpha_c)$ .
- $\Delta_{0,c} = \Delta$  and capacity  $C$ .

For any  $\sigma_0, \sigma_c \geq 0$  and  $0 \leq \gamma \leq \frac{C - \rho_c - \rho_0}{2}$

$$\theta^* = \min \left( \frac{\sigma_c}{C - \rho_c - \gamma}, \frac{[\sigma_c + (\rho_c + \gamma)\Delta]_+}{C} \right)$$

$$b(\sigma_0, \sigma_c) = \sigma_0 + (\rho_0 + \gamma)\theta^*$$

$$\varepsilon(\sigma_0, \sigma_c) = M_0 e \left( 1 + \frac{\rho_0}{\gamma} \right) e^{-\alpha_0 \sigma_0} + M_c e \left( 1 + \frac{\rho_c}{\gamma} \right) e^{-\alpha_c \sigma_c} .$$

Then,

$$\Pr\{B_0(t) > b(\sigma_0, \sigma_c)\} \leq \varepsilon(\sigma_0, \sigma_c) .$$

# Capacity Sizing of a $\Delta$ -scheduler

## Corollary (Per-flow capacity scaling properties)

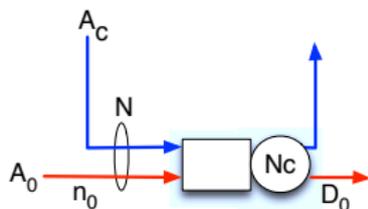
*The per-flow capacity of a  $\Delta$ -scheduler with a fixed (arbitrary small) buffer size, a target loss probability, and MMOO input flows satisfies*

$$c - \bar{a} = \begin{cases} o\left(\sqrt{\frac{\log N}{N}}\right) & \Delta \geq 0 \\ o\left(\frac{1}{N}\right) & \Delta < 0 \end{cases}$$

$\lim_{N \rightarrow \infty} c \rightarrow \bar{a}$  for all work-conserving schedulers.

The speed of convergence is highly affected by the scheduling algorithm.

# Network Decomposition ( $D_0 \rightarrow A_0$ )



## Output EBB characterization

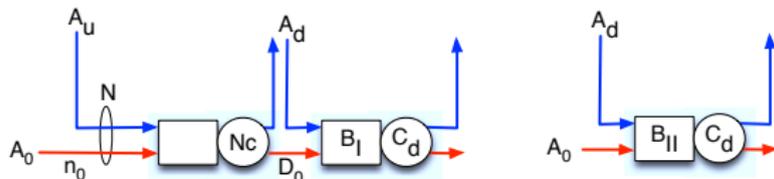
Given:  $A_0 \sim (1, \rho_0, \alpha_0)$  and  $A_c$  are MMOO input flows to a  $\Delta$ -scheduler.  
Then,  $D_0 \sim (M_0^{out}, \rho_0, \alpha_0^{out})$ , with

$$\alpha_0^{out} = \alpha_0 - O\left(\frac{1}{N}\right); \quad M_0^{out} = \begin{cases} L(N)N^{\frac{1}{N}} & \Delta \geq 0 \\ L(N)(Ne^{-N\beta})^{\frac{1}{N}} & \Delta < 0 \end{cases} .$$

where  $\lim_{N \rightarrow \infty} L(N) = 1$ .

- $D_0 \rightarrow A_0$  as  $N \rightarrow \infty$  for any work-conserving schedulers.
- The speed of convergence is substantially affected by the schedulers.

# Network Decomposition ( $B_I \rightarrow B_{II}$ )



## Theorem (a.s. convergence of $B_I$ to $B_{II}$ )

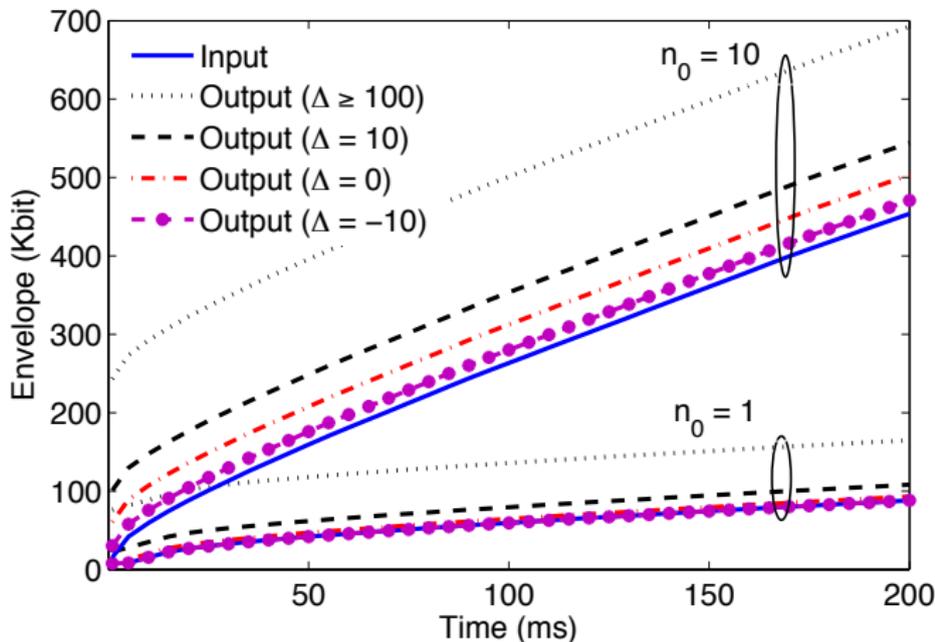
For MMOO traffic sources and  $\Delta$ -schedulers, there exists a constant  $\alpha > 0$  and a non-negative function  $L$  such that for any  $\sigma \geq 0$

$$\Pr\{|B_I(t) - B_{II}(t)| > \sigma\} = \begin{cases} O(N^2)e^{-N\alpha\sigma} & \Delta \geq 0 \\ O(N^2e^{-N\beta})e^{-N\alpha\sigma} & \Delta < 0 \end{cases}$$

$\lim_{N \rightarrow \infty} B_I \rightarrow B_{II}$  for all work-conserving schedulers.

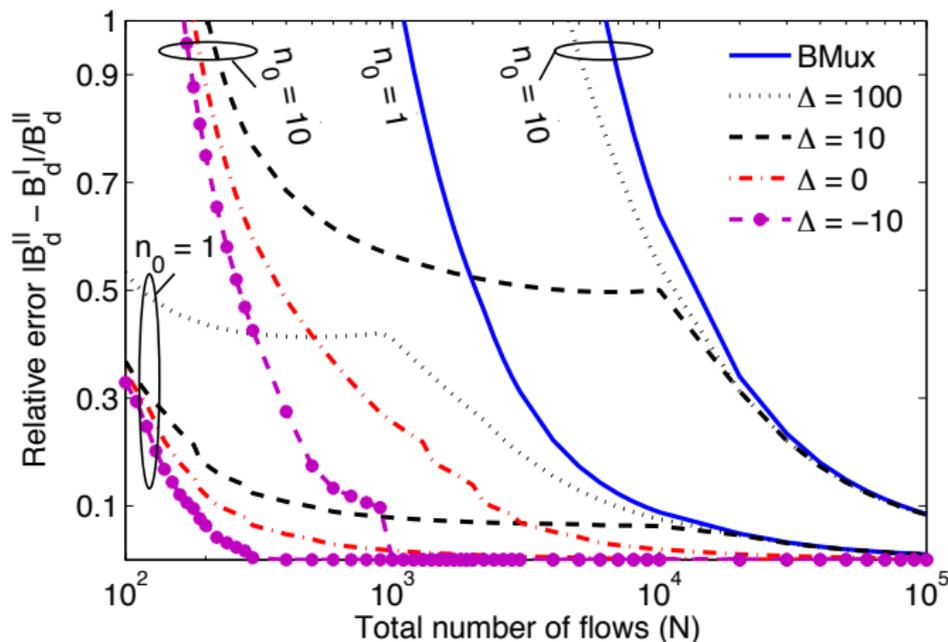
The speed of convergence is highly affected by the scheduling algorithm.

# Example 1: Network Decomposition ( $D_0 \rightarrow A_0$ )



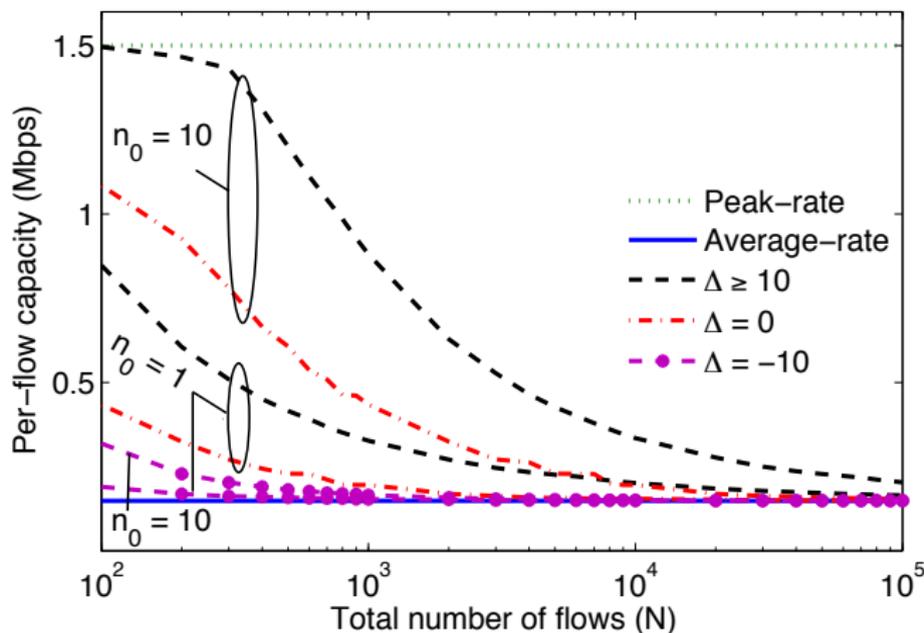
- $n_0 = 1, 10$ ,  $C = 100$  Mbps,  $U = 90\%$ , and  $\varepsilon^* = 10^{-6}$
- MMOO iid flows each with  $P = 1.5$  Kbits and  $T^* = 10$  ms

## Example 2: Network Decomposition ( $B_{II} \rightarrow B_I$ )



- $n_0 = 1, 10$ ,  $C = 100$  Mbps,  $U = 90\%$ , and  $\varepsilon^* = 10^{-6}$
- MMOO iid flows each with  $P = 1.5$  Kbits and  $T^* = 10$  ms

## Example 3: Capacity Provisioning



- $n_0 = 1, 10$ ,  $b_0 = 1.5$  Kbits,  $U = 90\%$ , and  $\varepsilon^* = 10^{-6}$
- MMOO iid flows each with  $P = 1.5$  Kbits and  $T^* = 10$  ms

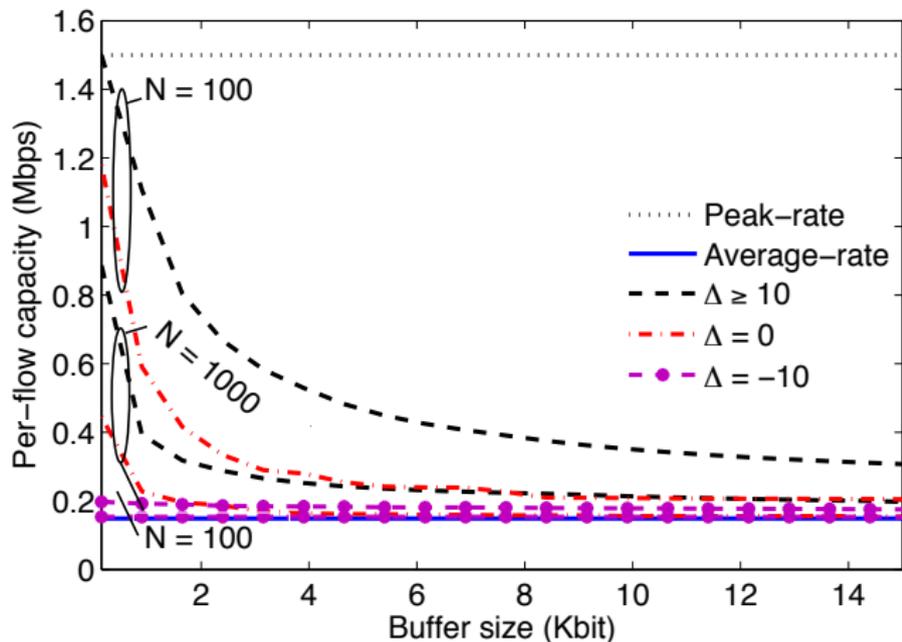
# Conclusions

- $c - \bar{a}$  ranges from  $O\left(\sqrt{\frac{\log N}{N}}\right)$  to  $O\left(\frac{1}{N}\right)$  depending on the scheduling algorithm.
- Capacity provisioning is highly affected by the scheduling algorithm.
- Network decomposition is valid for some schedulers even for moderate values of  $N$  (e.g., few hundreds).

Thank You

Questions?

## Example 4: Capacity Provisioning



- $n_0 = 1$ ,  $U = 90\%$ , and  $\varepsilon^* = 10^{-6}$
- MMOO iid flows each with  $P = 1.5$  Kbits and  $T^* = 10$  ms