

Per-Flow Backlog Bounds for a FIFO Server Beyond Token-Bucket-Constrained Arrivals

Lukas Wildberger, Anja Hamscher, Jens B. Schmitt

June 5th 2025

Outline

- 1 Motivation
- 2 Backlog Bound
- 3 Numerical Evaluation
- 4 Conclusion

Motivation

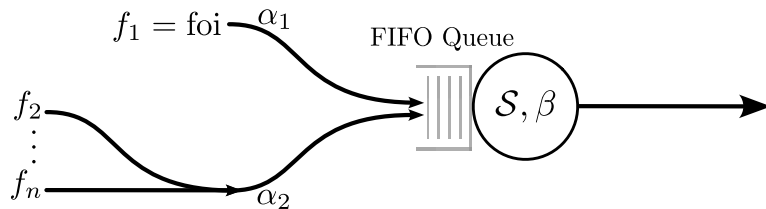
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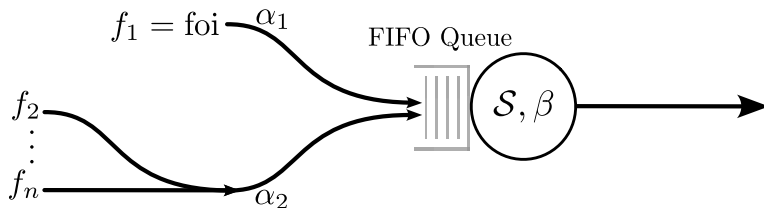
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What We're Up To



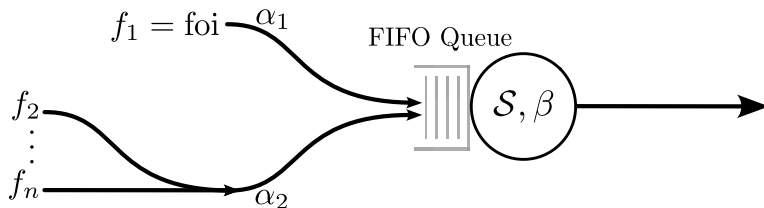
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- We have to solve the mathematical problem given by:

$$\theta_{\text{opt}} = \arg \min_{\theta \geq 0} \{v(\alpha_1, \beta_\theta^1)\}$$

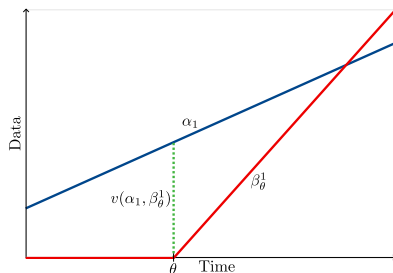
Function Set

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- Literature results primarily for token-bucket arrival and rate-latency service curves [Le Boudec and Thiran, 2001]

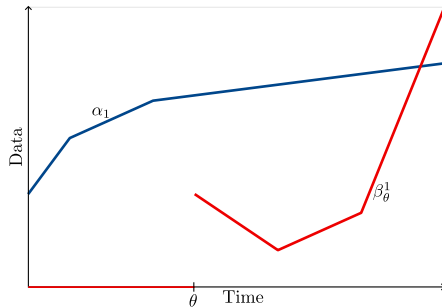
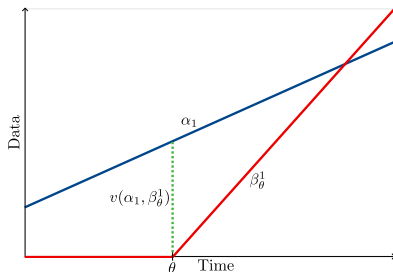
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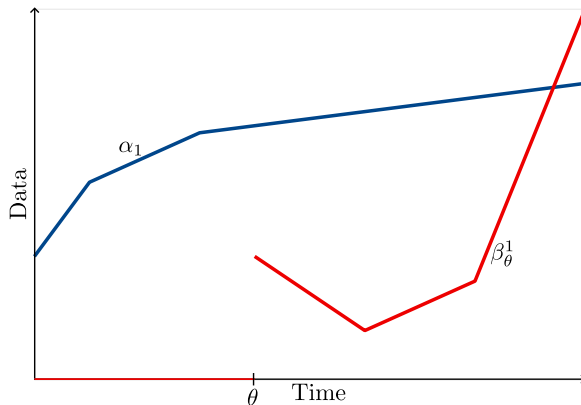


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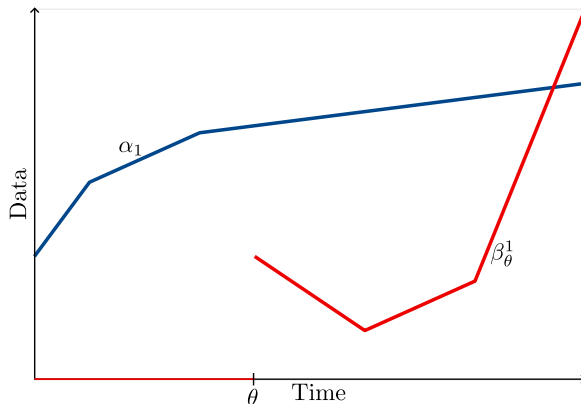
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Function Set - PWL Concave/Convex Curves

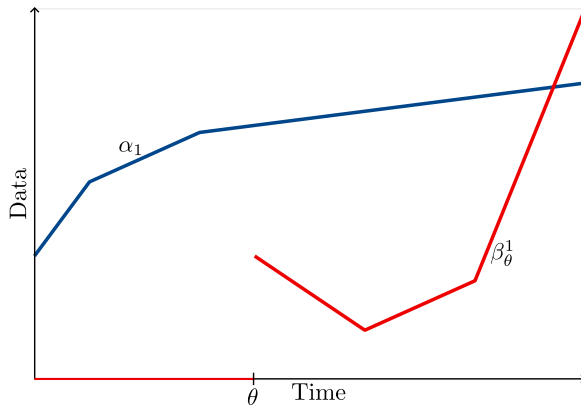


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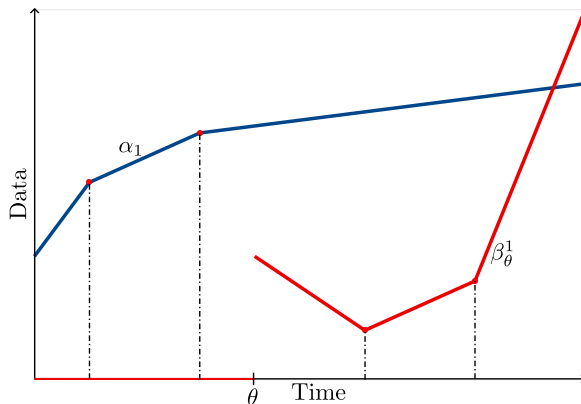


- FIFO residual service curve is also a service curve for the flow **even if it is not wide-sense increasing** [Bouillard et al., 2018]

Terminology



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- We call time where the slope of the linear segment changes “breakpoint”
 - ▢ In the literature also referred to as “intersection point” or “changepoint”

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Backlog Bound: Non-Wide-Sense Increasing Property

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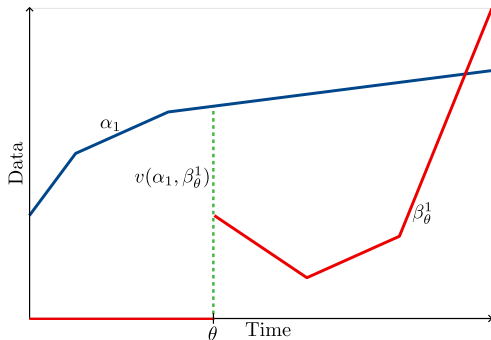
Lemma

Let $\alpha, \beta \in \mathcal{F}$ be given. Then it holds that $v(\alpha, \beta) = v(\alpha, \beta_{\downarrow})$, with $\beta_{\downarrow} := \beta \overline{\oslash} 0$.

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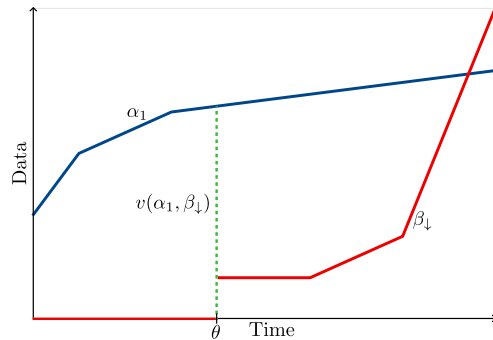
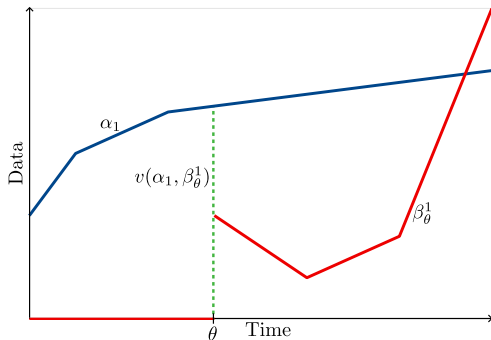
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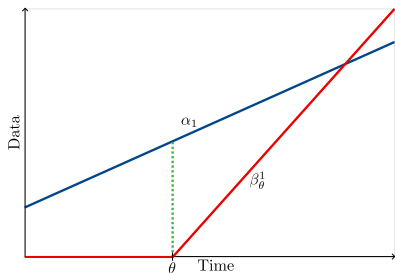
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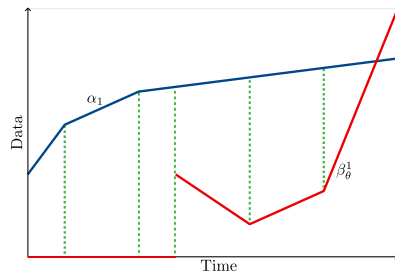
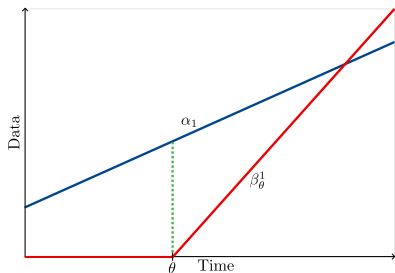
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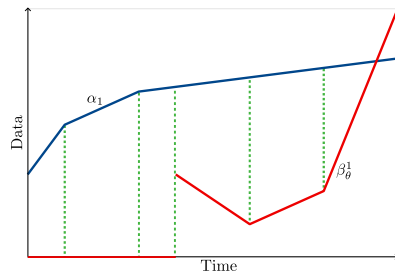
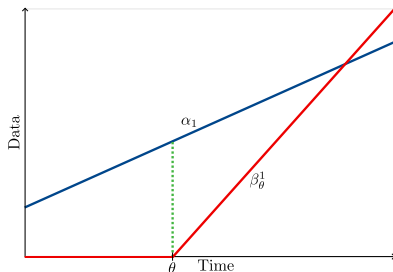
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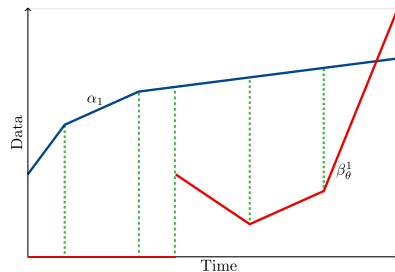
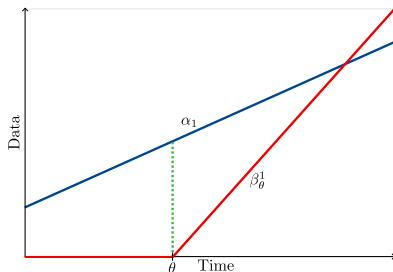


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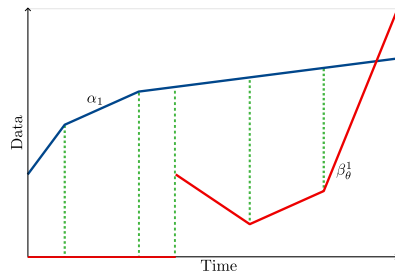
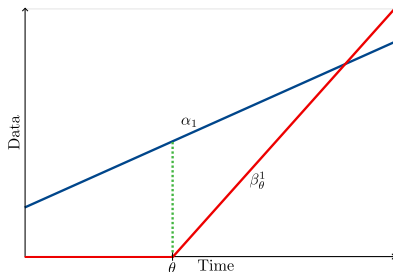
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- If we choose smaller/larger θ values the vertical distance will increase at some breakpoints and decrease at others
- We cannot split the problem into token-bucket cases and take the minimum
- We have to take care of all linear sections at once

Backlog-Optimal θ

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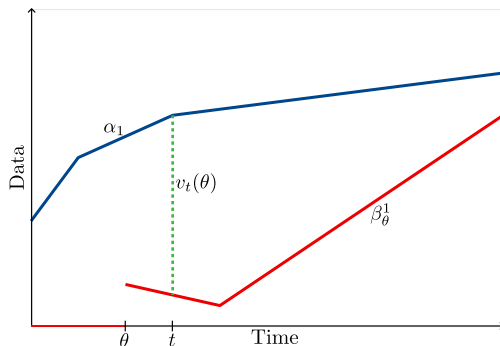
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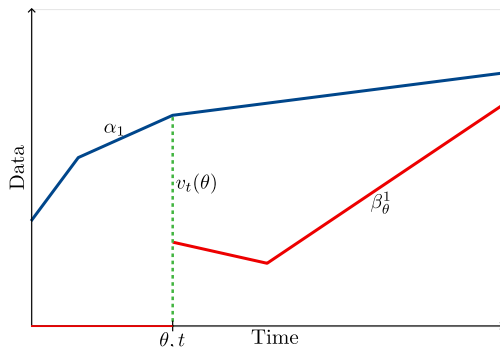
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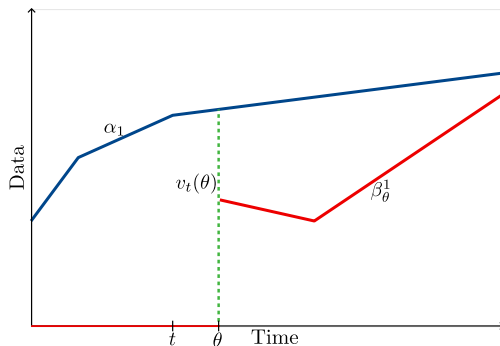
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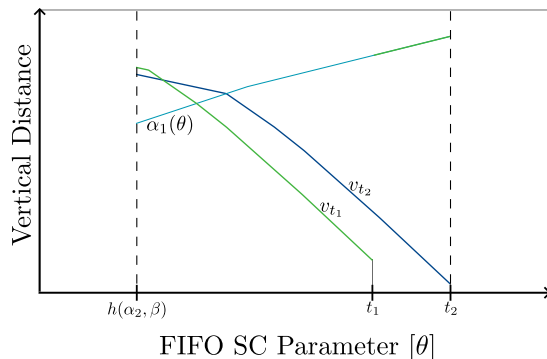
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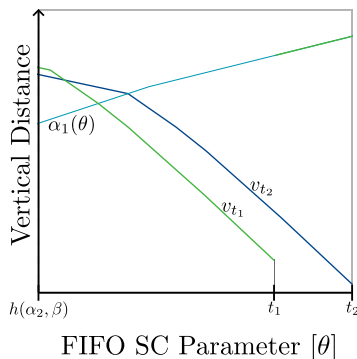
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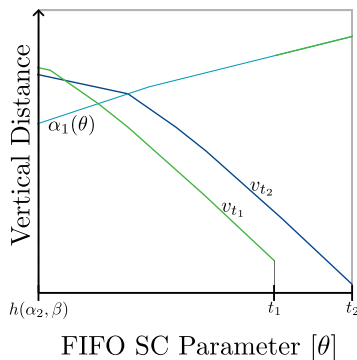


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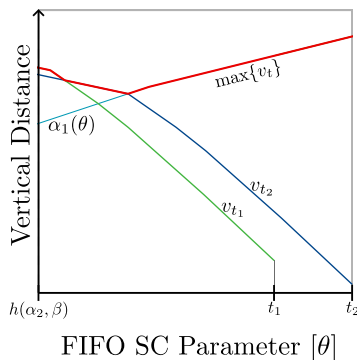
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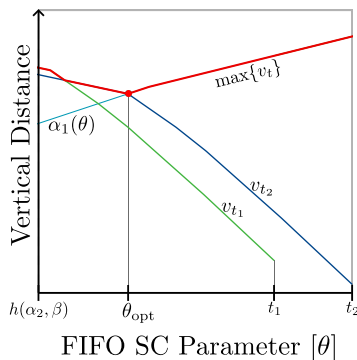
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Numerical Evaluation

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Numerical Evaluation: Setting

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- ▣ Multiple cross flows and foi constraint by T-Spec arrival curves

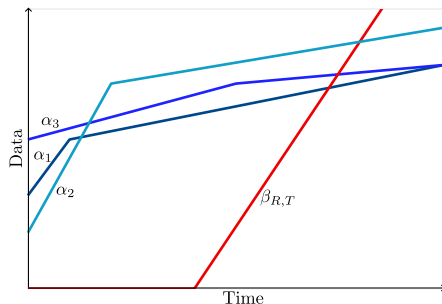
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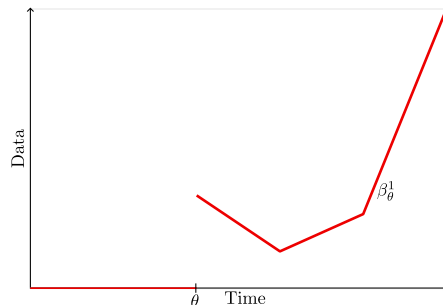
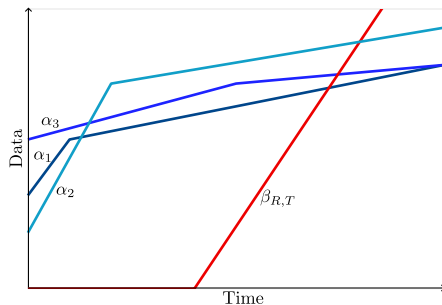


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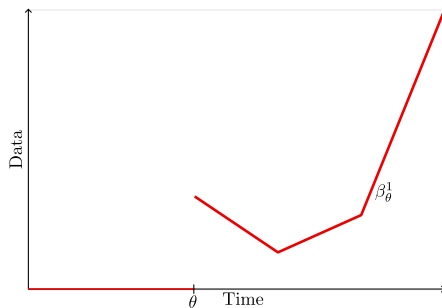
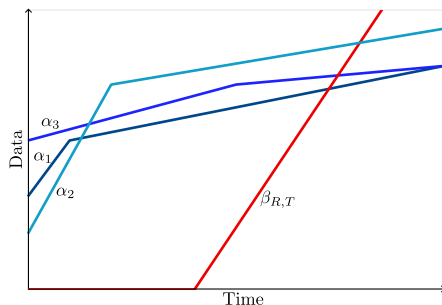


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- The parameters used, b_i^1, r_i^1, b_i^2 and r_i^2 , are randomly chosen out of a predefined interval

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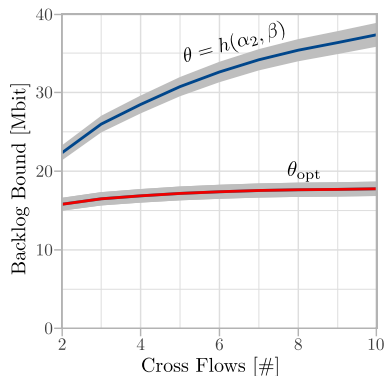
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What's next?

- Explore (further) applications
- Consider multi-node case
- ...

References

-  Bouillard, A., Boyer, M., and Le Corronc, E. (2018).
Deterministic Network Calculus: From Theory to Practical Implementation.
John Wiley & Sons.
-  Le Boudec, J.-Y. and Thiran, P. (2001).
Network Calculus: A Theory of Deterministic Queuing Systems for the Internet.
Springer.

Delay-Optimal θ (Theoretical Result)

Theorem

Let $\mathcal{S}$ be a system that multiplexes two flows f_1 and f_2 according to FIFO, where the arrivals of f_1 and f_2 , A_1 and A_2 , are constrained by α_1 and α_2 respectively. Further, assume that $\mathcal{S}$ offers a service curve β to the aggregate of the flows. If $\alpha_1 \in \mathcal{F}$ and $\alpha_2 \in \mathcal{F}$ are concave and $\beta \in \mathcal{F}$ is convex, then the optimal θ , that minimizes the delay bound, is given by

$$\theta_{opt} = \arg \min_{\theta \geq 0} \{h(\alpha_1, \beta_\theta^1)\} = h(\alpha_{agg}, \beta).$$

- Under FIFO scheduling the delay bound of the aggregate equals the per-flow delay bound
- It can be directly shown that $h(\alpha_{agg}, \beta) = h(\alpha_1, \beta_{h(\alpha_{agg}, \beta)}^1)$

$v_t(\theta)$ Curves

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$$\begin{aligned}
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 &= \begin{cases} \alpha_1(t) - \beta(t) + \alpha_2(t - \theta), & \text{if } h(\alpha_2, \beta) \leq \theta < t, \\ \alpha_1(\theta), & \text{if } \theta \geq t > h(\alpha_2, \beta). \end{cases}
 \end{aligned}$$

By definition we have $\beta_\theta^1 = [\beta(t) - \alpha_2(t - \theta)]^+ \wedge \delta_\theta(t)$. Since for $\theta < t$ it holds that $\delta_\theta(t) = 0$, and because $h(\alpha_2, \beta) \leq \theta$ ensures that $\beta(t) \geq \alpha_2(t - \theta)$ for all t , both the positive part and $\delta_\theta(t)$ can be omitted.

Application

We are now able to properly choose flow queue sizes for a setting with flows, having their own flow queue, getting FIFO scheduled

