

Optimal Congestion Control of TCP Flows for Internet Routers

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Fast and fair transmission of TCP data \Rightarrow avoidance of network congestion

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- Validate the model with simulations in Network Simulator 3 (ns-3)

- 1 Problem Description
 - Formulation of Markov Decision Process
- 2 Solution
 - Analytical Results
 - Numerical Results
- 3 Simulations in ns-3
- 4 Conclusions and Future Work

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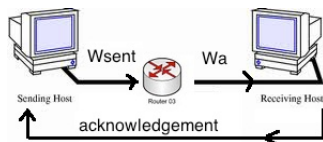


Figure: Example of one user sending TCP data

Problem Description: Formulation of MDP

R_n is function of W_n :

$$R_n^a := \begin{cases} \frac{(1 + W_n^a)^{1-\alpha} - 1}{1 - \alpha}, & \text{if } \alpha \neq 1, \\ \log(1 + W_n^a), & \text{if } \alpha = 1; \end{cases}$$

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- We consider *additive increasing* always

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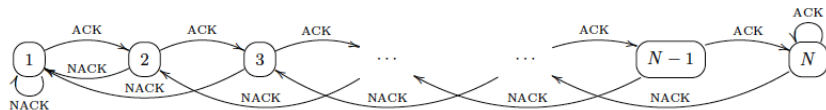
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- Maximizing the multiflow problem

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If for a given parameter ν , each policy π_k^* for $k \in \mathcal{K}$ optimizes the individual-flow problem then π^* optimizes the multi-flow problem (1).

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- 2 Combinatorial problem $\max_{\mathcal{S} \subseteq \mathcal{N}} \mathbb{R}_n^{\mathcal{S}} - \nu \mathbb{W}_n^{\mathcal{S}}$, where

$$\mathbb{R}_n^{\mathcal{S}} := \mathbb{E}_n^{\mathcal{S}} \left[\sum_{t=0}^{\infty} \beta^t R_{X(t)}^{a(t)} \right], \quad \mathbb{W}_n^{\mathcal{S}} := \mathbb{E}_n^{\mathcal{S}} \left[\sum_{t=0}^{\infty} \beta^t W_{X(t)}^{a(t)} \right]$$

Definition

We say that the above problem is **indexable**, if it exists real numbers ν_n , $n \in \mathcal{N}$ such that for all states the following holds:

- 1 if $\nu_n \geq \nu$, is optimal transmitting in state n
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We say that the above problem can be **solved under threshold policies** if $\nu_1 \geq \nu_2 \geq \dots \geq \nu_N$.

Main results: Analytical Results

From previous work, always indexable and solvable under threshold policies:

- 1 1-state and 2-state TCP flows
- 2 3-state TCP flow with decrease factor γ less than $\frac{2}{3}$

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Proposition

Three state TCP flow with $\gamma > \frac{2}{3}$ is indexable and:

- *if $\alpha < 1$, the threshold policies are optimal and the values of the indices are*

$$\nu_{k,1} = \frac{R_{k,1}}{W_{k,1}}, \nu_{k,2} = \frac{R_{k,2} - \beta R_{k,1}}{W_{k,2} - \beta W_{k,1}}, \nu_{k,3} = \frac{R_{k,3} + \beta(R_{k,3} - R_{k,2})}{W_{k,3} + \beta(W_{k,3} - W_{k,2})}.$$

- *if $\alpha \geq 1$, threshold policies are not optimal in general ($\nu_{k,1} > \nu_{k,3} > \nu_{k,2}$) and the values of the indices are*

$$\nu_{k,1} = \frac{R_{k,1}}{W_{k,1}}, \nu_{k,2} = \frac{R_{k,2} + \beta(R_{k,3} - R_{k,1}) + \beta^2(R_{k,3} - R_{k,2})}{W_{k,2} + \beta(W_{k,3} - W_{k,1}) + \beta^2(W_{k,3} - W_{k,2})},$$

$$\nu_{k,3} = \frac{R_{k,3} - \beta^2 R_{k,1}}{W_{k,3} - \beta^2 W_{k,1}}.$$

Numerical Results

Indexability of the problem tested over a large number of flows with different parameters \Rightarrow **always indexable**.

Conjecture: the scheme is always indexable.

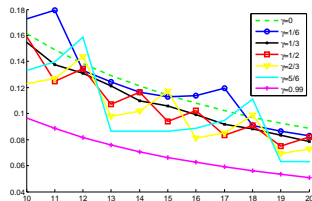
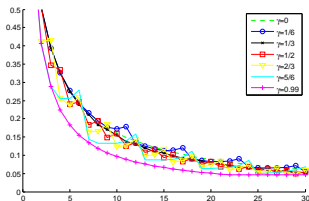


Figure: Seven Heterogeneous TCPs

Simulations Scenario Description

Network Simulator-3:

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Implementing the model:

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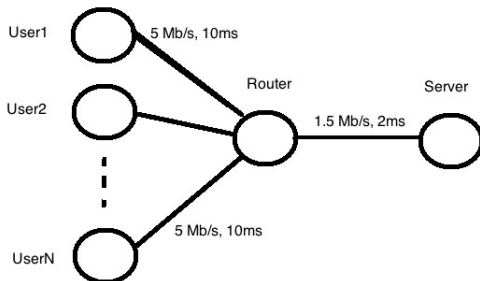
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Packet Size: 536 Bytes

Buffer size = Bandwidth-Delay Product = 14

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Packet-level heuristic index policy: Upon a packet arrival,

- if the buffer is not full, then accept the packet
- otherwise, drop the packet (either the new one or from the queue) with *smallest index* value
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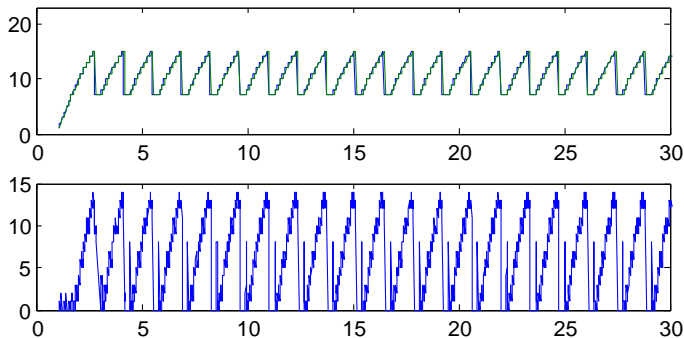
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Implementation in ns3:

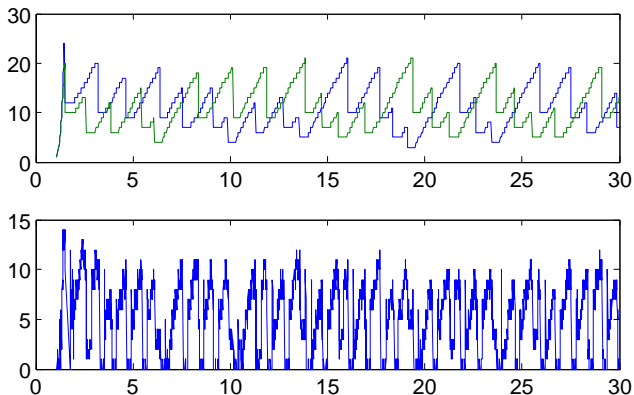
- 1 We calculate the indices for each user when program starts.
- 2 We get the congestion window of the user that want to send a packet.
- 3 We send the packet with the corresponding index, according to the congestion window.
- 4 In the queue of the router the index is read and it is taken the decision of transmitting it or not.

Droptail policy



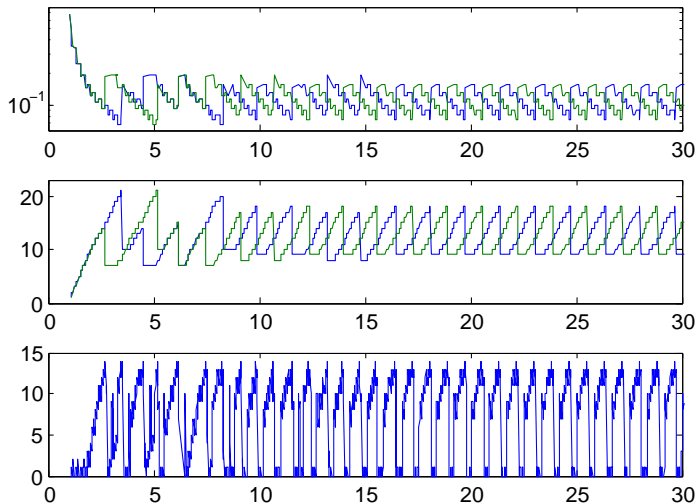
Simulation Results: 2 users and $\gamma = \frac{1}{2}$

RED



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Index policies model with $\alpha = 1$.



Main conclusions:

- Throughput increases
- More efficient buffer management
- Developed a packet implementation of index-policy

Future Work:

- Development new TCP models (Slow-start, users with different decrease factor...)
- Calculation of the index in the router \Rightarrow not needed to assume compliant end-users (index estimating and learning techniques)
- Investigate more complicate topologies.

Thank you for your attention

Thank you!!!