# Multiple time-delays system modeling and control for router management

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Abstract—This paper investigates the overload problem of a single congested router in TCP (Transmission Control Protocol) networks. To cope with the congestion phenomenon, we design a feedback control based on a multiple time-delays model of the set TCP/AQM (Active Queue Management). Indeed, using robust control tools, especially in the quadratic separation framework, the TCP/AQM model is rewritten as an interconnected system and a structured state feedback is constructed to stabilize the network variables. Finally, we illustrate the proposed methodology with a numerical example and simulations using NS-2 [1] simulator.

## I. INTRODUCTION

In IP networks, active queue management (AQM), embedded in router, reports to TCP sources its processing load. The objective is to manage the buffer utilization as well as the queueing delay. This has motivated a huge amount of work aiming at understanding the congestion phenomenon and achieving better performances in terms of *Ouality of Service* (OoS). As a matter of fact, there has been a growing recognition that the network itself must participate in congestion control and ressource management [2].

The AQM principle consists in dropping (or marking when ECN option is enabled) some packets before buffer saturates. Hence, following the Additive-Increase Multiplicative-Decrease (AIMD) behavior of TCP, sources reduce their congestion window size avoiding then the full saturation of the router. Basically, AQM support TCP for congestion avoidance and feedback to the latter when traffic is too heavy. Indeed, an AQM drops/marks incoming packet with a given probability related to a congestion index (such as queue length or delays) allowing then a kind of control on the buffer occupation at routers. Various mechanisms have been proposed in the network community for the development of AQM such as Random Early Detection (RED) [3], Random Early Marking (REM) [4], Adaptive Virtual Queue (AVQ) [5] and many others [6].

As it has been highlighted in the litterature (see for example [7], [8] and reference therein), AQM acts as a controller supporting TCP for congestion control and can be reformulated as a feedback control problem. Then, a significant research has been devoted to the use of control theory to develop more efficient AOM. Using dynamical model developed by [8], some P (Proportional), PI (Proportional Integral) have been designed. In the same framework, other tools have been used to extend this preliminary work such as

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a PID controller [6] or robust control [9]. However, most of these papers do not take into account the delay and ensure the stability in closed loop for all delays which could be very conservative in practice.

The study of congestion problem in time delay systems framework is not new and has been successfully exploited. Nevertheless, most of these works have been dedicated to the stability analysis of networks composed of homogeneous sources (see for example [10], [11] and [12]). In this paper, networks with heterogeneous sources are considered introducing then several delays and increasing the model complexity. Regarding the AQM design problem, some works have already been done in [13] and [14]. In [14], the construction of the AQM required to invoke the Generalized Nyquist Theorem and [13] provides a delay dependent state feedback involving delay compensations with a memory feedback control. All these latter methodologies are interesting in theory but hardly suitable in practice. While these latter studies have considered the simplified model of TCP/AQM from [8], we use in this contribution a more accurate model presented in [7]. Indeed, contrary to [13] and [14], both forward and backward delays are taken into account (that is, we do not neglect forward delays). Then, congestion control of networks consisting in heterogeneous TCP sources is transformed into a stabilization problem for multiple time delays systems. Using a robust analysis framework and especially quadratic separation approach developed for time-delay systems by [15], a stabilizing AQM is designed. The proposed control mechanism enables QoS in terms of RTT (Round Trip Time) and delay jitter which are relevant features for streaming and real-time applications over IP networks. Note that the approach employed in this paper allows the formulation of the problem into matrix inequalities that provide systematic stability condition, easy to test.

### II. NETWORK DYNAMICS

In this paper, we consider a network consisting of a single router and N heterogeneous TCP sources. By heterogeneous, we mean that each source is linked to the router with different propagation times (see Figure 1). Since the bottleneck is shared by N flows, TCP applies the congestion avoidance algorithm to cope with the network saturation [16].

Deterministic fluid-flow models have been widely used (see [7], [8] and [17] and references therein) to describe congestion control and AQM schemes in IP networks. These models capture the mean behavior of the TCP dynamic. While many studies dealing with network control in the

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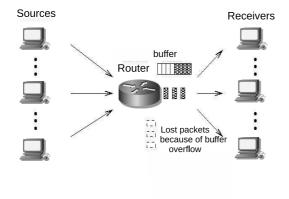


Fig. 1. Network topology

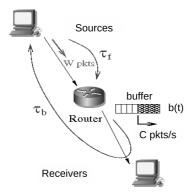


Fig. 2. A single connection

automatic control theory framework consider the model proposed by [8], we use in this paper the model introduced in [7] and described by (1). Contrary to the former, the model (1) takes into account the forward and backward delays and does not make the simplifying assumption that  $(W(t-\tau)/W(t))(1-p(t-\tau^b))=1$  [13]. The model and notations are as follow:

$$\begin{cases} \dot{W}_{i}(t) &= \frac{W_{i}(t-\tau_{i})}{\tau_{i}(t-\tau_{i})} (1 - p_{i}(t - \tau_{i}^{b})) \frac{1}{W_{i}(t)} \\ &- \frac{W_{i}(t-\tau_{i})}{\tau_{i}(t-\tau_{i})} \frac{W_{i}}{2} p_{i}(t - \tau_{i}^{b}) \end{cases} \\ \dot{b}(t) &= -C + \sum_{N} \eta_{i} \frac{W_{i}(t-\tau_{i}^{f})}{\tau_{i}(t-\tau_{i}^{f})} \\ \tau_{i} &= \frac{b(t)}{C} + T_{p_{i}} = \tau_{i}^{f} + \tau_{i}^{b} \end{cases}$$
(1)

where  $W_i(t)$  is the congestion window size of the source i, b(t) is the queue length of the buffer at the router,  $\tau_i$  is the round trip time (RTT) perceived by the source i. This latter quantity can be decomposed as the sum of the forward and backward delays ( $\tau_i^f$  and  $\tau_i^b$ ), standing for, respectively, the trip time from the source i to the router (the one way) and from the router to the source via the receiver (the return) (see Figure 2). C,  $T_{p_i}$  and N are parameters related to the network configuration and represent, respectively, the link capacity, the propagation time of the path taken by the connection i and the number of TCP sources.  $\eta_i$  is the number of sessions established by source i. The signal  $p_i(t)$  corresponds to the drop probability of a packet.

In this paper, the objective is to develop a method which computes the appropriate dropping probability applied at the router in order to regulate the queue length of the buffer b(t) to a desired level (Figure 3). Since control depends on the system state, it is required to have access to them. However, congestion windows  $W_i$  are not measurable. So that, we propose to reformulate the model (1) such that state vector can be measured. To this end, rates of each flow  $x_i$ , expressed as  $x_i(t) = \frac{W_i(t)}{\tau_i(t)}$ , will be considered. Hence, the dynamic of this new quantity is of the form  $\dot{x}_i(t) = \frac{d}{dt} \left( \frac{W_i(t)}{\tau_i(t)} \right) = \frac{\dot{W}_i(t) - x_i(t) \dot{\tau}_i(t)}{\tau_i(t)}$ . Based on the expressions of  $\dot{W}(t)$ ,  $\dot{b}(t)$ ,  $\tau_i(t)$  (see equation (1)) and  $\dot{\tau}(t) = \frac{\dot{b}(t)}{C}$ , a new model of the TCP behavior is derived

$$\begin{cases} \dot{x}(t) &= \frac{x(t-\tau)}{x(t)\tau(t)^{2}}(1-p(t-\tau^{b})) - \frac{x(t-\tau)x(t)}{2}p(t-\tau^{b}) \\ &+ \frac{x(t)}{\tau(t)} - \frac{x(t)}{\tau(t)C} \sum_{N} \eta_{i}x_{i}(t-\tau_{i}^{f}) \\ \dot{b}(t) &= -C + \sum_{N} \eta_{i}x_{i}(t-\tau_{i}^{f}) \end{cases}$$
(2)

**Remark 1** This model transformation allows us to use  $x_i$  instead of  $W_i$  which is more suitable to handle. Indeed, the measure of the aggregate flow has already been proposed and successfully exploited in [5] and [13] for the realization of the AVQ and a PID type AQM respectively.

Our work focuses on the congestion control of a single router with a static topology (N and  $\eta_i$  are constant). Moreover, for the mathematical tractability, we make the usual assumption [7], [8], [13] that all delays ( $\tau_i$ ,  $\tau_i^f$  and  $\tau_i^b$ ) are time invariant when they appear as argument of a variable (for example  $x_i(t-\tau_i(t))\equiv x_i(t-\tau_i)$ ). This latter assumption is valid as long as the queue length remains close to its equilibrium value and when the queueing delay is smaller than propagation delays. Defining an equilibrium point

$$\begin{cases} \tau_{i_0} = T_p + b_0/C \\ \dot{b}(t) = 0 \Rightarrow \sum_{N} \eta_i x_{i_0} = C \\ \dot{x}_i(t) = 0 \Rightarrow p_{i_0} = \frac{2}{2 + (x_{i_0} \tau_{i_0})^2} \end{cases} , \tag{3}$$

model (2) can be linearized:

$$\begin{bmatrix} \dot{x}_{1}(t) \\ \vdots \\ \dot{x}_{N}(t) \\ \dot{b}(t) \end{bmatrix} = A \begin{bmatrix} \delta x_{1}(t) \\ \vdots \\ \delta x_{N}(t) \\ \delta b(t) \end{bmatrix} + A_{d} \begin{bmatrix} \delta x_{1}(t - \tau_{1}^{f}) \\ \vdots \\ \delta x_{N}(t - \tau_{N}^{f}) \\ \delta b(t) \end{bmatrix}$$

$$+ B \begin{bmatrix} \delta p_{1}(t - \tau_{1}^{b}) \\ \vdots \\ \delta p_{N}(t - \tau_{N}^{b}) \end{bmatrix}$$

$$(4)$$

where  $\delta x_i \doteq x_i - x_{i_0}$ ,  $\delta b \doteq b - b_0$  and  $\delta p_i \doteq p_i - p_{i_0}$  are the state variations around the equilibrium point (3). Matrices of

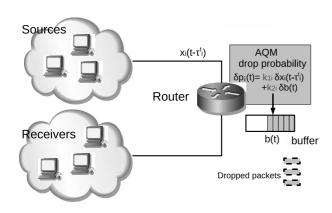


Fig. 3. Implementation of an AOM

the equation (4) are defined by

$$A = \begin{bmatrix} a_1 & 0 & 0 & h_1 \\ 0 & \ddots & 0 & \vdots \\ 0 & 0 & a_N & h_N \\ 0 & 0 & 0 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} e_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & e_N \\ 0 & 0 & 0 \end{bmatrix}$$
$$A_d = \begin{bmatrix} f_1\eta_1 & \dots & f_1\eta_N & 0 \\ \vdots & \vdots & \vdots & 0 \\ f_N\eta_1 & \dots & f_N\eta_N & 0 \\ \eta_1 & \dots & \eta_N & 0 \end{bmatrix},$$

with 
$$a_i=-\frac{1-p_{i_0}}{x_{i_0}\tau_{i_0}^2}-\frac{x_{i_0}p_{i_0}}{2},\ h_i=-\frac{2(1-p_{i_0})}{C\tau_{i_0}^3},\ f_i=-\frac{x_{i_0}}{\tau_{i_0}C}$$
 and  $e_i=-\frac{1}{\tau_{i_0}^2}-\frac{x_{i_0}^2}{2}$ . Remark that a multiple time delays system (4) is obtained with a particular form since each component of the state vector is delayed by a different quantity related to the communication path. Thus, in Section III an appropriate modeling of (4) with a structured form (in quadratic separation framework) is proposed in order to formulate the stability condition.

The problem of regulation is tackled in Section III with the design of a stabilizing state feedback for multiple time delays systems in order to guarantee a QoS. Hence, the dropping probability  $p_i$  associated to source i will be computed at the router by

$$p_i(t) = p_{i_0} + k_1 \delta x_i(t - \tau_i^f) + k_2 \delta b(t)$$
 (5)

with  $k_1$ ,  $k_2$  are components which have to be designed (see Figure 3). Note that the dropping probability  $p_i$  perceived by the source i will be delayed  $p_i(t-\tau_i^b)$  because of the backward delay. Thus, a structured state feedback of the form (5) is proposed:

- to avoid unnatural signals with different delay combinations in the state feedback:  $\delta x_i(t-\tau_i^f-\tau_j^b)$  for  $i,\ j\in\{1,...,N\}$  introducing then many additional delays,
- it provides light computations (with less operations) reducing then the processing time at router.

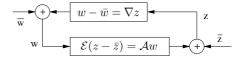


Fig. 4. An interconnected system

Applying the structured state feedback type control law  $p_i(t)$  to each source  $i, \ \forall i \in \{1,\ldots,N\}$ , the following interconnected system is obtained

$$\begin{bmatrix} \delta \dot{x}_{1}(t) \\ \vdots \\ \delta \dot{x}_{N}(t) \\ \dot{b}(t) \end{bmatrix} = A \begin{bmatrix} \delta x_{1}(t) \\ \vdots \\ \delta x_{N}(t) \\ \delta b(t) \end{bmatrix} + A_{d} \begin{bmatrix} \delta x_{1}(t - \tau_{1}^{f}) \\ \vdots \\ \delta x_{N}(t - \tau_{N}^{f}) \\ \delta b(t) \end{bmatrix} + BK_{1} \begin{bmatrix} \delta x_{1}(t - \tau_{1}) \\ \vdots \\ \delta x_{N}(t - \tau_{N}) \end{bmatrix} + BK_{2} \begin{bmatrix} \delta b(t - \tau_{1}^{b}) \\ \vdots \\ \delta b(t - \tau_{N}^{b}) \end{bmatrix}$$
(6)

where matrices gains  $K_1$  and  $K_2$  are structured as  $K_1 = diag\{k_{11}, \ldots, k_{1N}\}$  and  $K_2 = diag\{k_{21}, \ldots, k_{2N}\}$ .

Equation (6) represents thus the mean behavior of TCP regulated by a structured state feedback type AQM around an equilibrium point.  $K_1$  and  $K_2$  can be derived from the stability analysis of the interconnection (6). To this end, we propose to design these latter gains by a suitable modeling of (6) applying then the quadratic separation principle [18], [15] for the stability condition.

# III. STABILIZATION AND QOS GUARANTEE: DESIGN OF AN AQM

The stability of the interconnected system (6) depends on matrices  $K_1$  and  $K_2$ . This section aims to develop a method that provides such stabilizing matrices. Thus, the quadratic separation framework is considered and the following theorem will be employed [18].

**Theorem 1** Given two possibly non-squared matrices  $\mathcal{E}$ ,  $\mathcal{A}$  and an uncertain matrix  $\nabla$  belonging to a set  $\Xi$ . The uncertain system represented on Figure 4 is stable for all matrices  $\nabla \in \Xi$  if and only if it exists a matrix  $\Theta = \Theta^*$  satisfying conditions

$$\begin{bmatrix} \mathcal{E} & -\mathcal{A} \end{bmatrix}^{\perp *} \Theta \begin{bmatrix} \mathcal{E} & -\mathcal{A} \end{bmatrix}^{\perp} > 0 \tag{7}$$

$$\left[\begin{array}{cc} 1 & \nabla^* \end{array}\right] \Theta \left[\begin{array}{c} 1 \\ \nabla \end{array}\right] \leq 0. \tag{8}$$

It is then required to transform the initial system (6) into a feedback system of the form of Figure 4 where a linear equation is connected to a linear uncertainty  $\nabla$ . The key idea (proposed by [15] for the single time-delay systems case) consists in associating the delay operator as an uncertainty which must be bounded. Hence, Theorem 1 may be applied to the multiple time delays system (6) by rewritting it as an interconnected system (see Figure 4) with  $w = \nabla z$ ,

$$\nabla = Diag\left(s^{-1}\mathbf{1}_{4}, \mathcal{D}_{-f}, \mathcal{D}_{-b}, \mathcal{D}_{\tau}, \mathcal{I}_{-f}, \mathcal{I}_{-b}, \mathcal{I}_{\tau}\right) \tag{9}$$

and the feedforward equation  $\mathcal{E}z = \mathcal{A}w$ ,

where

$$\hat{b}(t) = \begin{bmatrix} \delta b(t) \\ \vdots \\ \delta b(t) \end{bmatrix}, \hat{b}^{\tau^b}(t) = \begin{bmatrix} \delta b(t - \tau_1^b) \\ \vdots \\ \delta b(t - \tau_N^b) \end{bmatrix},$$

$$\hat{x}^{\tau^f}(t) = \begin{bmatrix} \delta x_1(t - \tau_1^f) \\ \vdots \\ \delta x_N(t - \tau_N^f) \end{bmatrix}, \hat{x}^{\tau}(t) = \begin{bmatrix} \delta x_1(t - \tau_1) \\ \vdots \\ \delta x_N(t - \tau_N) \end{bmatrix},$$

$$\hat{x}(t) = \begin{bmatrix} \delta x_1(t) \\ \vdots \\ \delta x_N(t) \end{bmatrix}, \mathcal{D}_{\Diamond} = \begin{bmatrix} e^{-\lozenge_1 s} & 0 \\ \vdots \\ 0 & e^{-\lozenge_N s} \end{bmatrix}.$$
We also note  $\bar{A}_d = A_d \begin{bmatrix} 1_N \\ 0_{1 \times N} \end{bmatrix}, E_1 = \begin{bmatrix} 1_N & 0_{N \times 1} \end{bmatrix}, E_2 = \begin{bmatrix} 0_N & \vdots \\ 1 & 1 \end{bmatrix}$  and  $w_1(t) = \hat{x}(t) - \hat{x}^{\tau^f}(t), w_2(t) = \hat{b}(t) - \hat{b}^{\tau^b}(t), w_3(t) = \hat{x}(t) - \hat{x}^{\tau}(t), X(t) = \begin{bmatrix} \hat{x}' & b(t) \end{bmatrix}'.$  The delay matrix operators  $\mathcal{D}_{\Diamond}$  ( $\Diamond$  represents  $\tau^f$ ,  $\tau^b$  or  $\tau$ ) must be defined to create the delayed signals  $\hat{x}^{\tau^f}$ ,  $\hat{x}^{\tau}$  and  $\hat{b}^{\tau^b}(1)$ .  $\mathcal{I}_{\Diamond} = diag(\frac{1-e^{-\lozenge_1 s}}{s}, \dots, \frac{1-e^{-\lozenge_N s}}{s}), \diamondsuit = \{f, b, \emptyset\}$  are additional operators related to the delay enhancing the description of the artificial uncertain set [19][15].

We aim at proving the stability which problem can be recast in the present framework as the well-posedness of the feedback system for all values of the delays  $(\tau_i, \tau_i^f \text{ and } \tau_i^b, i = \{1, \dots, N\})$  and all admissible uncertainties  $\nabla \in \Xi$ . Delay operators  $\mathcal{D}_{\Diamond}$  are thus isolated as well as operators  $\mathcal{I}_{\Diamond}$  ensuring a better bound on each delays. Expressing the covering set on every block of  $\nabla$ , a conservative choice of separator satisfying the second inequality Then a conservative choice of quadratic separator that fullfils (8) is of the form

$$\Theta = \begin{bmatrix} \Theta_{11} & \Theta_{12} \\ \Theta'_{12} & \Theta_{22} \end{bmatrix}$$
 (12)

with

$$\begin{split} \Theta_{11} = & diag\Big(\mathbf{0}_{N+1}, -\mathbf{Q}_{0}^{f}, -\mathbf{Q}_{0}^{b}, -\mathbf{Q}_{0}, -\mathbf{Q}_{1}^{f}\mathbf{T}^{f^{2}}, -\mathbf{Q}_{1}^{b}\mathbf{T}^{b^{2}}, -\mathbf{Q}_{1}\mathbf{T}^{2}\Big), \\ \Theta_{22} = & diag\Big(\mathbf{0}_{N+1}, \mathbf{Q}_{0}^{f}, \mathbf{Q}_{0}^{b}, \mathbf{Q}_{0}, \mathbf{Q}_{1}^{f}, \mathbf{Q}_{1}^{b}, \mathbf{Q}_{1}\Big), \\ \Theta_{12} = & diag\big(-P, \mathbf{0}_{6N}\big), T^{\diamondsuit} = diag\big(\tau_{1}^{\diamondsuit}, \dots, \tau_{N}^{\diamondsuit}\big), \\ \mathbf{Q}_{0}^{\diamondsuit} = & diag\big(q_{01}^{\diamondsuit}, \dots, q_{0N}^{\diamondsuit}\big), \mathbf{Q}_{1}^{\diamondsuit} = diag\big(q_{11}^{\diamondsuit}, \dots, q_{1N}^{\diamondsuit}\big), \end{split}$$

$$(13)$$

where  $P \in \mathbb{R}^{\mathbb{N}+1 \times \mathbb{N}+1} > 0$  and  $q_i^{\lozenge}$  are positive scalars for all  $i = \{1, \dots, N\}$  and for  $\lozenge = \{f, b, \emptyset\}$  (the proof is detailed in [15] for the single time-delay case). Since, the inequality (8) is satisfied by definition of the operator bounds of the uncertain matrix  $\nabla$ , it remains to verify the first one (7). After some algebraic manipulations, it can be shown that this latter inequality can be expressed as

$$\begin{bmatrix} \Xi_1 & \Xi_2 Q_1 T & \Xi_2 Q_1^f T^f \\ * & Q_1 T & \mathsf{0}_{\mathsf{N}} \\ * & * & Q_1^f T^f \end{bmatrix} > 0 \tag{14}$$

where

$$\Xi_{1} = N_{1}^{T} \Theta N_{1} + N_{2}^{T} \Theta N_{1} + N_{1}^{T} \Theta N_{2},$$

$$\Xi_{2} = \begin{bmatrix} 0_{N \times 2N+1} & E_{1}BK_{2} & E_{1}BK_{1} \end{bmatrix}^{T},$$

$$\frac{A \quad \bar{A}_{d} \quad 0_{N+1 \times 2N}}{E_{1}}$$

$$E_{2} \quad 0_{3N \times 3N}$$

$$E_{1}$$

$$E_{1}A \quad E_{1}\bar{A}_{d} \quad 0_{N \times 2N}$$

$$0_{N \times N+1} \quad E_{2}\bar{A}_{d} \quad 0_{N \times 2N}$$

$$E_{1}A \quad E_{1}\bar{A}_{d} \quad 0_{N \times 2N}$$

$$E_{1}A \quad E_{1}\bar{A}_{d} \quad 0_{N \times 2N}$$

$$E_{1}A \quad E_{1}\bar{A}_{d} \quad 0_{N \times 2N}$$

$$E_{1}BK_{1} \quad E_{2}$$

$$E_{1}BK_{2} \quad E_{1}BK_{1}$$

$$0_{N \times 2N+1} \quad E_{1}BK_{2} \quad E_{1}BK_{1}$$

$$0_{N \times 4N+1}$$

$$0_{N \times 2N+1} \quad E_{1}BK_{2} \quad E_{1}BK_{1}$$

$$0_{N \times 2N+1} \quad E_{1}BK_{2} \quad E_{1}BK_{1}$$

This latter condition gives the following theorem:

**Theorem 2** For given scalars  $\tau_i$ ,  $\tau_i^b$  and  $\tau_i^f$  for  $i = \{1, ..., N\}$ , if there exists a  $N+1 \times N+1$  positive definite matrix P and  $N \times N$  diagonal positive matrices  $Q_k^f$ ,  $Q_k^b$ ,  $Q_k$  with  $k = \{0, 1\}$  and  $K_1$ ,  $K_2$  such that the inequality (14) is satisfied, then the system (6) is stable.

The state feedback gains  $K_1$  and  $K_2$  are thus derived solving the inequality (14) of Theorem 2. Regarding the synthesis problem, since  $K_1$  and  $K_2$  are decision variables, condition (14) is bilinear and a global optimal solution cannot be found. Nevertheless, the feasibility problem can still be tested to provide a suboptimal solution:

- ⊳ using a BMI solver.
- □ using a relaxation algorithm:

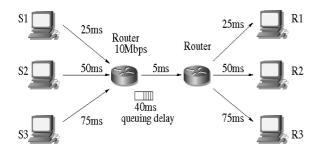


Fig. 5. Network topology: an example

- $step \ 0$ . Initialization:  $K_1 = K_{1_0}$ ,  $K_2 = K_{2_0}$ .
- step 1. Run LMI computation (inequality (14) with fixed  $K_1, K_2$ )  $\Rightarrow$  backup  $P_0 = P, Q_{1_0} = Q_1, Q_{1_0}^f = Q_1^f$ .
- step 2. Set  $P = P_0$ ,  $Q_1 = Q_{1_0}$ ,  $Q_1^f = Q_{1_0}^f$  and  $K_1$ ,  $K_2$  are free, run LMI computation.
- $step\ 3$ . If condition feasible: stop algorithm. Otherwise: return to  $step\ 0$

### **IV. NS-2 SIMULATIONS**

In this section, we perform simulations with the network simulator NS-2 [1] (release 2.30) to validate the exposed theory. Throughout this part, the efficiency of the proposed mechanism is evaluated and compared to few existing AQM. Consider the numerical example of the Figure 5. So, the objective is to regulate the queue length of the router to a desired level  $b_0 = 100$  packets while the maximal buffer size is set to 400 packets. Propagation times are as illustrated on Figure 5. The link bandwidth is fixed to 10Mbps, that is 2500 packet/s considering packet size of 500 bytes. Hence, at the equilibrium the queueing delay is equal to 40ms. Each of the three sources uses TCP/Reno and establishes 10 connections generating long lived TCP flows (like FTP connections). Upon these latter specifications, the equilibrium point (3) is derived:  $b_0 = 100pkt$ ,  $\tau_{1_0} = 150ms$ ,  $\tau_{2_0} = 250ms$ ,  $\tau_{3_0} = 350ms$ ,  $x_{1_0} = x_{2_0} = x_{3_0} = 83.33pkt/s$  (rate for each connections of the three sources) and  $p_0 = 10^{-3} [9.508 \ 3.444 \ 1.760]'$ . Then, we have to find matrices  $K_1$  and  $K_2$  such that the derived feedback system is stable and the regulation around the equilibrium point is ensured. Applying Theorem 2 exposed in section III, two stabilizing matrix gains

$$K_1 = 10^{-3} \begin{bmatrix} -0.09 & 0 & 0 \\ 0 & 0.61 & 0 \\ 0 & 0 & 0.76 \end{bmatrix}, K_2 = 10^{-3} \begin{bmatrix} 0.27 & 0 & 0 \\ 0 & 0.13 & 0 \\ 0 & 0 & 0.08 \\ (16) \end{bmatrix}$$

can be found. Considering the network of Figure 5, we have simulated the congestion phenomenon at a router applying the drop tail mechanism and different AQM. Effect of these latter dropping strategies on the queue length have been evaluated and are illustrated on Figure 6. Table I presents additional characteristics of the results. As expected, the drop tail mechanism maintains the queue size close to the

TABLE I SOME STATISTICS ON THE QUEUE LENGTH FOR DIFFERENT AQM

	DT	RED	REM	PI	SF
Mean(pkt)	317.6	103.8	94.8	99.4	102.3
Stand.dev.(pkt)	84	21.7	70.7	35.1	23.3
$\begin{array}{c} Average \\ queueing \ delay(ms) \end{array}$	127	41.5	37.9	39.8	40.9
$Average \ delay \ jitter(ms)$	33.6	8.7	28.3	14	9.3

buffer overflow (maximal buffer size: 400pkt) involving many uncontrolled dropped packets. Futhermore, a such large queue size implies a large queueing delay and also large oscillations providing an important delay jitter (see Table I). On Figure 6, we can observed that SF provides a better regulation causing less oscillations and good precision. These characteristics can be verified in Table I which analyses each response from Figure 6. In the simulation, AQM was designed to regulate queue size at 100pkt (equilibrium) avoiding severe congestion and ensuring then a stable queueing delay of 40ms. First, in Table I, the mean (and average queueing delay since the queueing delay is expressed by b(t)/C) shows the accuracy of each AOM according the desired prescribed queue length. Next, the standard deviation (and average delay jitter) shows the ability of each AQM to maintain the queue length (and queueing delay) close to the desired equilibrium providing then an efficient regulation. Note that REM has the lowest queueing delay, however the objective is to guarantee the queue length stability inducing then a queueing delay equals to 40ms. If a lower queueing delay is required, one just has to change the equilibrium point (reducing the desired queue length).

Besides, the stability of the congestion phenomenon keeping a stable queue length (and thus a stable queueing delay), allows to control the RTT ( $\tau_i$  for  $i=\{1,...,N\}$ ) for all sources to a desired value with low variations (see Table I). Thanks to the efficient regulation applied by SF, a QoS in terms of RTT and delay jitter is hence guaranteed. It provides thus an interesting feature which is relevant for streaming and real-time applications. Although RED performs a good regulation on the queue length and the RTT at the steady state, its response time is very slow which makes RED inefficient against short-lived traffic perturbations [20].

Table II presents statistics related to the packet arrival rates of each user when different AQM are implemented at the router and bears out the accuracy and the control efficiency of the proposed SF. The prescribed equilibrium rate for each connection (according to (3)) that establishes fairness is  $x_{1_0} = x_{2_0} = x_{3_0} = 83.33pkt/s$ . As it can be seen, only SF is able to keep arrival rates comparatively close to the equilibrium value. Futhermore, based on the Jain's fairness index  $=\frac{(\sum x_i)^2}{(n\sum x_i^2)}$  [21] (the more the index is close to 1, the more the distribution of the resources is fair), we observe that SF applies a fair strategy.

TABLE II SOME STATISTICS ON ARRIVAL RATES FOR DIFFERENT AQM

	RED			REM			PI			SF		
users	1	2	3	1	2	3	1	2	3	1	2	3
Mean (pkt/s)	147	83	148	147	81	60	146	88	109	104	91	88
Stand. dev. (pkt/s)	61	36	372	56	34	23	66	35	335	41	30	40
Jain's fairness index	0.9450		0.8703		0.9579			0.9946				

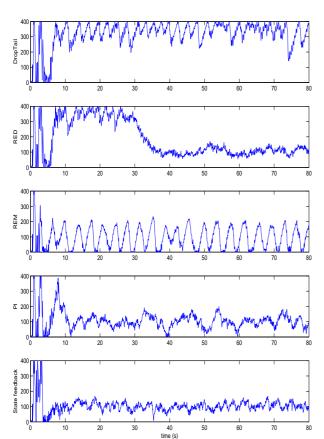


Fig. 6. Evolution of the queue length b(t) (pkt): the desired size is 100pkts

### V. CONCLUSION

In this paper the design of an AQM for congeston control of a single router has been presented. The considered topology consists in several TCP sources sending long-lived flows through a router to their respective receivers. To supply the TCP congestion control mechanism, an AQM must be implemented to the router. Based on a modified mathematical model of the protocol behavior, a such AQM has been developed with control theory tools. Indeed, in a time delay system framework, a control law has been proposed and then the stability analysis of the feedback system has been performed. Consequently, the regulation of flows and the queue size of at the router is ensured. At last, a numerical example and simulations have shown the effectiveness of the proposed methodology.

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