

On Multiplexing Gain for Networks with Deterministic Delay Guarantees

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Abstract—Multiplexing gain has been studied extensively in the context of statistical characterizations of traffic streams with quality-of-service criteria such as packet loss probability, mean delay, and delay variance. In this paper, we demonstrate that multiplexing gain can also arise in the context of deterministic traffic constraint functions, service curve scheduling, and quality-of-service requirements based on deterministic delay constraints. We show how to evaluate this multiplexing gain via the use of deterministic network calculus for both worst-case and time-averaged delay constraints. We show that significant multiplexing gain can be achieved in a deterministic setting using numerical examples drawn from a number of well-known MPEG video traces. Our results have application to provisioning services with tight, real-time constraints on end-to-end delay performance.

I. INTRODUCTION

Current proposals for supporting network-level deterministic quality-of-service (QoS) guarantees can be distinguished by scheduling policies, traffic regulation schemes, and the degree of multiplexing of individual connections or flows. Some policies, such as IntServ [1], support traffic on a flow by flow basis, while other policies, including DiffServ [2], require that traffic from multiple flows be multiplexed together. Deterministic QoS performance can be analyzed using network calculus theory (see [3][4]).

An important result of the network calculus theory under min-plus algebra concerns the *end-to-end deterministic gain* [3]: If an input stream A_i is relayed through n network elements, then the maximum delays incurred at each network element are not necessarily specific to a particular packet. This fact is referred to as the “pay bursts only once” phenomenon [4]. In this theory, deterministic end-to-end delay results are obtained by convolving service curves associated with nodes on the end-to-end path.

In the present paper, we develop a new concept of *deterministic multiplexing gain*, whereby a greater number of connections can be admitted under a given deterministic delay constraint if they are multiplexed together in one aggregate stream. Although statistical multiplexing gain with stochastic traffic models has been well-studied, it is not obvious that multiplexing gains can be achieved with respect to deterministic QoS constraints. Deterministic multiplexing gain may be explained intuitively as follows. The maximum delays encountered by two packets from two input streams A_1 ,

respectively A_2 , do not necessarily occur during the same time slot. We show analytically that deterministic multiplexing gain can be achieved and propose a heuristic to maximize the achievable gain among a set of connections. We present numerical examples of MPEG video traces which demonstrate that substantial gains can be achieved. Unlike statistical multiplexing gain, the deterministic multiplexing gain applies directly to the end-to-end case via the end-to-end convolved service curve result from the network calculus theory.

The paper continues as follows: Section II provides basic background in network calculus. Section III demonstrates how to obtain tight traffic constraint functions for multiplexed traffic. Sections IV and V discuss multiplexing gain with respect to worst case delay and time-averaged delay, respectively. Section VI shows the results of our numerical experiments using different MPEG-4 trace files, and Section VII offers some conclusions and observations.

II. TRAFFIC ENVELOPES AND SERVICE CURVES

Consider a network element E , for example a switch or multiplexer, and an input arrival process specified by a non-negative, non-decreasing function $A(\cdot)$, where $A(t)$ denotes the cumulative arrivals from time 0 to time t . The process A is usually called the arrival process. The output process is denoted by B and the output capacity of E is assumed to be $c > 0$.

Definition 1: A function $f : \mathbf{N} \rightarrow \mathbf{N}$ with the property that

$$A(t) - A(s) \leq f(t - s) \quad (\forall) \quad 0 \leq s \leq t, \quad (1)$$

is a traffic constraint function or an *envelope* of A [5]. The relation between A and f will be denoted by $A \prec f$.

The expression (1) may be also written as $A \leq A * f$, where $A * f$ represents the convolution between A and f , and $(A * f)(t) = \inf_{0 \leq s \leq t} (A(s) + f(t - s)) \quad (\forall) \quad t \geq 0$.

The tightest envelope of A is given by the so-called *empirical envelope* [6] or minimal envelope process [3] and is defined as $\varepsilon_A(t) = \sup_{s \geq 0} (A(s + t) - A(s))$. In practice, all traffic constraint models try to approximate ε_A through an envelope $f \geq \varepsilon_A$. We will measure the accuracy of such an approximation through the following metric: $\Lambda(f, \varepsilon_A) =$

$\int_0^\infty |f(t) - \varepsilon_A(t)| dt$. Note that $\Lambda(f, \varepsilon_A)$ has the properties of a metric in an appropriate metric space.

In the following, we briefly review two well-known traffic envelope functions that have been discussed in the literature. The affine model (σ, ρ) is the simplest among the traffic constraint functions [5]. It is defined as $f(t) = \sigma + \rho t$, where $\sigma \geq 0$ represents the maximal allowable burst of the arrival process A , while $\rho > 0$ represents the long term average rate of A . The $(\vec{\sigma}, \vec{\rho})_n$ model [6] has an envelope function defined by $f(t) = \min_{i=1, n} (\sigma_i + \rho_i t)$, where $0 \leq \sigma_1 < \dots < \sigma_n$ and $0 < \rho_n < \dots < \rho_1$.

Besides the concept of traffic envelope, the concept of *service curve* [7] is necessary to derive performance results in the deterministic network calculus. We will adopt here the definition from [3]. Specifically, S denotes a service curve for an input arrival process A if $B(t) \geq (A * S)(t) (\forall) t \geq 0$. In terms of implementation, the SCED (Service Curve-based Earliest Deadline first) [7] scheduling algorithm is designed to guarantee service curves for the corresponding traffic flows.

Let us assume now that we have a set of n connections A_i , each with the envelope f_i and the service curve S_i . The next theorem gives a sufficient condition that must be held by a number of traffic flows in order to be accepted by the CAC (connection admission control) component.

Theorem 1 ([8][3]): If

$$\sum_{i=1}^n (f_i * S_i)(t) \leq ct (\forall) t \geq 0, \quad (2)$$

then the SCED policy guarantees connection A_i the service curve S_i for any $i = \overline{1, n}$.

Also, a stronger condition than (2), that is

$$\sum_{i=1}^n S_i(t) \leq ct (\forall) t \geq 0, \quad (3)$$

is given in [8]. If condition (3) holds, we say that the set of connections satisfies the condition *MPX*.

Suppose that the condition (2) holds and let $T \subseteq \{1, 2, \dots, n\}$. One may notice that it is not generally true that the SCED policy can guarantee the service curve $\sum_{i \in T} S_i$ for the multiplexed connection $\sum_{i \in T} A_i$. Motivated by this observation, we need the following definition for the rest of the paper:

Definition 2: Suppose that the condition (2) holds. For a subset $T \subseteq \{1, 2, \dots, n\}$, we say that the connections $A_i, i \in T$ may be multiplexed under condition *MPX_T*, if the following condition holds: $((\sum_{i \in T} f_i) * (\sum_{i \in T} S_i))(t) + \sum_{i \notin T} (f_i * S_i)(t) \leq ct (\forall) t \geq 0$. Note that the other connections $A_i, i \notin T$ are treated individually.

In other words, if the conditions (2) and *MPX_T* hold for $T \subseteq \{1, 2, \dots, n\}$, then the SCED policy guarantees the service curves S_i for the connections $A_i (\forall) i \notin T$, and the service curve $\sum_{i \in T} S_i(\cdot)$ for the multiplexed connection $\sum_{i \in T} A_i$.

III. AGGREGATE ENVELOPE CONSTRUCTION

This section demonstrates how to construct an envelope for a multiplexed stream, and derives some of its properties, which will be useful later in the paper. As the primary concern of this section is represented by traffic envelopes, our analysis is based on projections under min-plus algebra (for more details, see [3]).

Consider a network element that consists of two arrival channels C_1, C_2 and a scheduler S . Each C_i consists of an input arrival process I_i , a traffic regulator R_i , and a regulated arrival process denoted by A_i . The empirical envelope ε_{I_i} of I_i is given by $\varepsilon_{I_i}(t) = \sup_{s \geq 0} (I(s+t) - I(s))$ and is usually determined experimentally. Lastly, the traffic regulator R_i is specified through an approximation of ε_{I_i} , that is the envelope function f_i of A_i . Since the actual inputs I_i may not conform to ε_{I_i} , they will have to go through R_i in order to yield a regulated output A_i .

An important property of the envelope functions is expressed as follows:

Proposition 1: If $A_1 \prec f_1$ and $A_2 \prec f_2$, then $A_1 + A_2 \prec f_1 + f_2$.

Proof: Since the $\inf(\cdot)$ functional is superadditive, the following holds: $\inf_{0 \leq s \leq t} (A_1(t-s) + f_1(s) + A_2(t-s) + f_2(s)) \geq \inf_{0 \leq s \leq t} (A_1(t-s) + f_1(s)) + \inf_{0 \leq s \leq t} (A_2(t-s) + f_2(s))$. ■

In view of the definition of a traffic envelope (1), we will show how to construct a better traffic envelope for $A_1 + A_2$ rather than $f_1 + f_2$. We now give some useful definitions pertinent to our study.

A. Projections under the min-plus algebra

We denote by \mathcal{F} the set of envelope functions $\{f \mid f \geq 0\}$. Two operations are introduced on \mathcal{F} : $f \oplus g = \min(f, g)$ and $(f * g)(t) = \min_{0 \leq s \leq t} (f(t-s) + g(s))$. Define $\overline{\mathcal{R}}^+ = \{x \in \mathcal{R} \mid x \geq 0\} \cup \{\infty\}$. On this set, two operations are introduced: $\alpha \oplus \beta = \min(\alpha, \beta)$ and $\alpha \otimes \beta = \alpha + \beta$. The elements of \mathcal{F} are also called vectors while the elements of $\overline{\mathcal{R}}^+$ are called scalars. We relate these sets with the operation \otimes such that $\alpha \otimes f = \alpha + f (\forall) \alpha \in \overline{\mathcal{R}}^+, f \in \mathcal{F}$. The set \mathcal{F} can be treated now as a moduloid (a vector space without inverse elements) over the semiring $(\overline{\mathcal{R}}^+, \oplus, \otimes)$. For two vectors $f, g \in \mathcal{F}$, the scalar projection of f onto g is defined by $\langle f, g \rangle = \sup_{t \geq 0} (f(t) - g(t))^+$, where $(x)^+ = \sup\{x, 0\}$. The vector projection of f onto g is denoted by $\langle f, g \rangle \otimes g$.

Consider a generic input arrival process I having the empirical envelope ε_I . As it is usually hard to implement directly a regulator for ε_I , we will give an upper $(\vec{\sigma}, \vec{\rho})_n$ approximation of it. Consider a subset $\{g_1, \dots, g_n\} \subset \mathcal{F}$ with $g_i(t) = \rho_i t$ and $\rho_1 > \rho_2 > \dots > \rho_n \geq 0$. The $(\vec{\sigma}, \vec{\rho})_n$ envelope of I is given by $f = \bigoplus_{i=1}^n ((\varepsilon_I, g_i) \otimes g_i)$. We call $\{g_1, \dots, g_n\}$ a *base characterization* set of I .

Recalling the metric Λ , we introduce an *f-pseudo-norm* on the moduloid \mathcal{F} , where $f \in \mathcal{F}$: $\|\varepsilon_I\|_f = \Lambda((\varepsilon_I, f) \otimes f, \varepsilon_I) = \int_0^\infty |\langle \varepsilon_I, f \rangle \otimes f - \varepsilon_I| dt, \varepsilon_I \in \mathcal{F}$

B. (σ, ρ) -envelopes

Let us suppose that the base characterization sets for I_1, I_2 are $\{g_1 = \rho_1 t\}$ and $\{g_2 = \rho_2 t\}$, respectively. Let $f_i = \langle \varepsilon_{I_i}, g_i \rangle \otimes g_i$ ($\forall i = 1, 2$), and $f = f_1 + f_2$. Let C be the aggregated arrival channel $C_1 + C_2$ and let $I = I_1 + I_2$. One may notice that the simplest way to constrain I is through f , while an alternative way to do it would be through $\tilde{f} = \langle \varepsilon_{I_1} + \varepsilon_{I_2}, g_1 + g_2 \rangle \otimes (g_1 + g_2)$ (see Fig. 1).

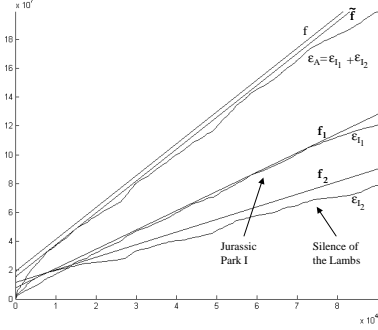


Fig. 1. Two trace files

The following two results reflect the properties of aggregate envelope \tilde{f} .

Proposition 2: For f and \tilde{f} defined above, we have that $\tilde{f} \leq f$.

Proof: Since the $\sup\{\cdot\}^+$ functional is subadditive, we have that $\langle \varepsilon_{I_1} + \varepsilon_{I_2}, g_1 + g_2 \rangle \leq \langle \varepsilon_{I_1}, g_1 \rangle + \langle \varepsilon_{I_2}, g_2 \rangle$. Then, $\tilde{f} = \langle \varepsilon_{I_1} + \varepsilon_{I_2}, g_1 + g_2 \rangle \otimes (g_1 + g_2) \leq \langle \varepsilon_{I_1}, g_1 \rangle \otimes g_1 + \langle \varepsilon_{I_2}, g_2 \rangle \otimes g_2 = f_1 + f_2 = f$. ■

The intuition behind the fact that $\tilde{f} \leq f$ is that it is possible that the burst times of ε_{I_1} differ from the burst times of ε_{I_2} . Hence, by aggregating C_1 and C_2 , the bursts of ε_{I_1} and ε_{I_2} will be interleaved, rather than summed up. As we will see in the next section, this property leads to possibly tighter bounds for the worst case delay of the aggregated flow C . That is, the worst case delay for C may be less than the minimum between the worst case delays of C_1 and C_2 .

Theorem 2: $\|\varepsilon_{I_1} + \varepsilon_{I_2}\|_{g_1 + g_2} \leq \|\varepsilon_{I_1}\|_{g_1} + \|\varepsilon_{I_2}\|_{g_2}$.

Proof: We have that:

$$\begin{aligned} & \int_0^\infty (\langle \varepsilon_{I_1} + \varepsilon_{I_2}, g_1 + g_2 \rangle \otimes (g_1 + g_2) - (\varepsilon_{I_1} + \varepsilon_{I_2})) dt \leq \\ & \int_0^\infty (\langle \varepsilon_{I_1}, g_1 \rangle \otimes g_1 - \varepsilon_{I_1}) dt + \int_0^\infty (\langle \varepsilon_{I_2}, g_2 \rangle \otimes g_2 - \varepsilon_{I_2}) dt \\ & \iff \sup_{t \geq 0} \{\varepsilon_{I_1}(t) - g_1(t) + \varepsilon_{I_2}(t) - g_2(t)\}^+ \leq \\ & \sup_{t \geq 0} \{\varepsilon_{I_1}(t) - g_1(t)\}^+ + \sup_{t \geq 0} \{\varepsilon_{I_2}(t) - g_2(t)\}^+, \end{aligned}$$

which is true because the $\sup\{\cdot\}^+$ functional is subadditive. ■

C. Generalization

A key feature of $(\vec{\sigma}, \vec{\rho})_n$ envelopes is the ability to capture the property that the arrival rate of a traffic stream decreases

as function of time. Since the (σ, ρ) model cannot capture this property, we will generalize the previous results to the $(\vec{\sigma}, \vec{\rho})_n$ model.

Consider two arrival channels C_1 and C_2 with the base characterization sets $\{g_1, \dots, g_n\}$ and $\{h_1, \dots, h_n\}$, respectively, with $n > 1$. The envelopes g and h corresponding to C_1 and C_2 , respectively, are constructed as before. The envelope for the aggregated channel $C = C_1 + C_2$ is given by $\tilde{f} = \bigoplus_{i=1}^n (\langle \varepsilon_{I_1} + \varepsilon_{I_2}, g_i + h_i \rangle \otimes (g_i + h_i))$.

We now extend the definition of the norm $\|\cdot\|$ in the following way. For any $g_1, \dots, g_n \in \mathcal{F}$, we define $\|\varepsilon_I\|_{(g_1, \dots, g_n)} = \int_0^\infty \left[\bigoplus_{i=1}^n (\langle \varepsilon_I, g_i \rangle \otimes g_i) - \varepsilon_I \right] dt$. The generalization of Proposition 2 is the following:

Proposition 3:

$$\begin{aligned} \tilde{f} & \triangleq \bigoplus_{i=1}^n (\langle \varepsilon_{I_1} + \varepsilon_{I_2}, g_i + h_i \rangle \otimes (g_i + h_i)) \leq \\ & \bigoplus_{i=1}^n (\langle \varepsilon_{I_1}, g_i \rangle \otimes g_i + \langle \varepsilon_{I_2}, h_i \rangle \otimes h_i) \triangleq f. \end{aligned}$$

The proof is based on the following lemma:

Lemma 1: Let $g_1, \dots, g_n \in \mathcal{F}$ and $h_1, \dots, h_n \in \mathcal{F}$ two sequences of non-decreasing real functions such that $g_i \leq h_i$ ($\forall i$). Then, $\bigoplus_{i=1}^n g_i \leq \bigoplus_{i=1}^n h_i$.

Proof: Let $t \geq 0$, $i_g = \sup\{j \mid g_j(t) \leq g_l(t) \ (\forall l = \overline{1, n})\}$ and $i_h = \sup\{j \mid h_j(t) \leq h_l(t) \ (\forall l = \overline{1, n})\}$. We have the following: $\bigoplus_{i=1}^n g_i(t) = g_{i_g}(t) \leq g_{i_h}(t) = \bigoplus_{i=1}^n h_i(t)$. ■

Theorem 2 can be generalized as follows:

Theorem 3: If $\|\varepsilon_{I_1} + \varepsilon_{I_2}\|_{g_n + h_n} < \|\varepsilon_{I_1}\|_{g_n} + \|\varepsilon_{I_2}\|_{h_n}$, then

$$\|\varepsilon_{I_1} + \varepsilon_{I_2}\|_{(g_1 + h_1, \dots, g_n + h_n)} < \|\varepsilon_{I_1}\|_{(g_1, \dots, g_n)} + \|\varepsilon_{I_2}\|_{(h_1, \dots, h_n)}.$$

The proof is omitted here. Note that the theorem's condition $\|\varepsilon_{I_1} + \varepsilon_{I_2}\|_{g_n + h_n} < \|\varepsilon_{I_1}\|_{g_n} + \|\varepsilon_{I_2}\|_{h_n}$ gives a sufficient condition in order to consider the envelope multiplexing of C_1 and C_2 for the general case.

IV. MULTIPLEXING GAIN UNDER WORST-CASE DELAY

Consider two regulated arrival processes A_1, A_2 with $A_i \prec f_i$ and worst case delay d_i for any $i = 1, 2$. Without loss of generality, suppose that $d_1 \leq d_2$ and assume that the condition (2) holds. The minimum service curve corresponding to A_i such that the server guarantees the delays d_i is $S_i(t) = f_i((t - d_i)^+)$ ($\forall i = 1, 2$). [3]

Let us suppose that the condition $MPX_{\{1,2\}}$ holds, that is $S_1 + S_2$ is a service curve for $A := A_1 + A_2$. Also, let \tilde{f} , constructed as in the previous section, be the envelope of A . The worst case delay corresponding to the multiplexed connection A is simply: $d = \inf \left\{ d \geq 0 \mid \tilde{f}(t) \leq S_1(t + d) + S_2(t + d) \ (\forall t \geq 0) \right\}$.

Proposition 4: In the current context, we have that $d \leq d_2$.

Proof: From Propositions 2 and 3 we have that $\tilde{f}(t) \leq f_1(t) + f_2(t) \leq S_1(t + d_2) + S_2(t + d_2) (\forall) t \geq 0$. The result then follows immediately. ■

Informally, the above result states the following: Suppose that we have a server that guarantees the worst case delay d_1 for an input A_1 . Although it may be impossible to accept a new input A_2 requiring a worst case delay d_2 , it may be possible to accept A_2 , but at the expense of a possibly higher worst case delay for A_1 , if the inputs A_1 and A_2 are treated as the multiplexed input $A := A_1 + A_2$. Also, note the fact that the delay d for the multiplexed connection A may be less than d_1 !

Definition 3 (Deterministic Multiplexing Gain): Assume that we have a server and n input connections A_i , each requiring a worst case delay. Let N be the maximum number of connections accepted by a given CAC (see for example Equation (2)) when no multiplexing is done. Let M be the maximum number of connections accepted by the CAC when multiplexing is done according to a partition $\{T_k\}$ on a subset T of the connections as follows. The connections $A_j \in T_k$ are multiplexed together and treated by the CAC as a single aggregate connection with a worst case delay requirement less than or equal to the minimum of the worst case delays for the connections $A_j \in T_k$. Multiplexing according to the partition $\{T_k\}$ is called “good multiplexing” since the worst case delay requirements of all connections in T are satisfied. Then, the *deterministic multiplexing gain* is defined as:

$$G = \frac{M}{N}. \quad (4)$$

The previous definition is similar to the definition for statistical multiplexing gain given in [9].

In general, the problem of finding a partition of the set of connections to maximize the deterministic multiplexing gain G belongs to the class of NP-complete problems. We now give a heuristic, called H^{MG} , for this problem:

Algorithm 1 (H^{MG}): Loop among the connections A_i , $i = 1, \dots, n$. Assume that before each step i , there is a partition $\{T_k\}$ of the set $\{A_1, \dots, A_{i-1}\}$. At step i , the connection A_i is multiplexed sequentially with each set T_k in the partition, and it is checked if this multiplexing is “good”. If yes, add A_i to the current set T_k and proceed to the algorithm’s next step. If A_i cannot be multiplexed in a “good” way with any of the sets T_k , create a new set consisting just of A_i , add this set to the partition $\{T_k\}$, and then proceed to the next step.

One may deduce that the computational complexity for this algorithm is $O(n^2)$. The main intuition behind this algorithm can be expressed as follows: If A_i cannot be multiplexed in a “good” way with either T_{k_1} or T_{k_2} , it is less likely that a “good multiplexing” can be realized with the set $T_{k_1} \cup T_{k_2}$.

The following example shows a case when the worst case delay for A is less than the minimum between the worst case delay for the independent connections A_1 and A_2 . In other words, a multiplexed gain of $G > 1$ can be obtained.

Example 1: Let $c = 4$, $\varepsilon_{A_1} = \{0, 12, 12, 12\}$ ($\varepsilon_{A_1}(0) = 0, \dots, \varepsilon_{A_1}(3) = 12$), and $\varepsilon_{A_2} = \{0, 8, 12, 16\}$. Also, let

$S_1(t) = 2t$, $S_2(t) = 2t$ and the base characterization sets for A_1 and A_2 be $\{g_1(t) = 2t\}$, respectively $\{g_2(t) = 2t\}$. We simply obtain the expressions for the corresponding envelopes of A_1 and A_2 as $f_1(t) = 2t + 10$, respectively $f_2(t) = 2t + 10$. We obtain that the worst case delays d_1 and d_2 that can be guaranteed for the connections A_1 , respectively A_2 , are $d_1 = 5$ and $d_2 = 5$, respectively. Observe now that the condition MPX holds. Hence, the multiplexed input $A_1 + A_2$ is guaranteed the service curve $S(t) = 4t (\forall) t \geq 0$. On the other hand, the envelope for the multiplexed input $A_1 + A_2$ is $\tilde{f}(t) = 4t + 16$. Now, since $\tilde{f}(t) \leq S(t + 4) (\forall) t \geq 0$, a worst case delay of 4 for packets from both A_1 and A_2 can be guaranteed!

V. MULTIPLEXING GAIN UNDER TIME-AVERAGED DELAY

For various network traffic, such as data transfer, one may request the required QoS parameter to be the time-averaged delay, instead of the worst case delay. Motivated by this, we extend the results from the previous section to time-averaged deterministic multiplexing gain.

Let us denote the average arrival rate of the process A by $\lambda = \lim_{t \rightarrow \infty} \frac{1}{t} A(t)$, and the average delay by $d_{avg}^f = \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t d_f(s) ds$. One can show that

$$d_{avg}^f \leq \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t (f(s) - (A * S)(s)) ds \quad (5)$$

Let us define $H(f, A, S) = \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t (f(s) - (A * S)(s)) ds$. A simple analysis of (5) shows that if $\int_0^\infty (f(t) - (A * S)(t)) dt < \infty$, then $\lambda d_{avg}^f = 0$; this is the intuitive case when the arrival envelope and the output curve tend to each other asymptotically. If this is not the case, then we obtain that $\lambda d_{avg}^f \leq \lim_{t \rightarrow \infty} (f(t) - (A * S)(t))$.

Let us consider now two input connections A_1 and A_2 with average arrival rates of λ_1 and λ_2 , respectively. Also, assume that $A_i \prec f_i$ and that the service curve S_i are given for $i = 1, 2$. As in the previous section, the condition $MPX_{\{1,2\}}$ must hold, while A represents the multiplexed connection $A_1 + A_2$. Let \tilde{f} be the envelope of A constructed as in Section III and let $S := S_1 + S_2$ be the service curve for A . Assume that f_i belongs to the (σ, ρ) -envelope model.

Proposition 5: In the current context, we have that $\frac{H(\tilde{f}, A, S)}{\lambda_1 + \lambda_2} \leq \max\left(\frac{H(f_1, A_1, S_1)}{\lambda_1}, \frac{H(f_2, A_2, S_2)}{\lambda_2}\right)$.

The proof is omitted. Note that a similar result may be obtained if f_i belong to the $(\vec{\sigma}, \vec{\rho})_n$ -envelope model. The consequence of the last result is exactly that of Proposition 4. That is, by aggregating two connections, one likely obtains a smaller time-averaged delay than the smallest between the time-averaged delays of the two flows, if those were obtained without aggregation.

VI. EXPERIMENTAL RESULTS

In this section, we consider several trace files for MPEG-4 encoded videos with a duration of 1 hour and a rate of 25 frames/sec (see [10]). The time unit in our analysis is $\tau = 40$ ms, and the data unit is 1 byte.

We conducted the experiments as follows. First, we took n traces A_i and computed the empirical envelopes ε_{A_i} . Second, for a given n , we built a $(\vec{\sigma}_i, \vec{\rho}_i)_n$ envelope for A_i , denoted by f_i , as discussed in Section III. Third, we assigned various service curves $S_i(t) = c_i t$ for the connections A_i and set the server capacity to be $c = \sum c_i$. Then, we computed the delays d_i that could be guaranteed by the server for each A_i . Finally, we aggregated all the connections A_i into one connection A and computed the $(\vec{\sigma}, \vec{\rho})_n$ envelope \tilde{f} of A . Note that the condition MPX holds. We then computed the guaranteed delay d for A and compared it with the guaranteed delays d_i .

Recall Fig. 1 which shows the corresponding empirical envelopes and the (σ, ρ) envelopes for two video traces. Fig. 2 presents an experiment involving the results obtained for three trace files.

Trace file	Avg. rate	# leaky buckets	S(\cdot)	Worst case delay	Time-averaged delay
Silence of the Lambs Medium quality (SLM)	876	10	1500*t	4672	7720
Jurassic Park I Medium quality (JPM)	1339	10	2500*t	2425	1352
Soccer High Quality (SOH)	5533	10	7000*t	3890	1139
SLM+JPM+SOH	7792	10	11000*t	1190	849

Fig. 2. Three trace files

Observe that the last row represents the actual multiplexed stream. The most notable fact is that the bound for the worst case delay under the multiplexed analysis dramatically improves upon the bound for the worst case delay for the independent connections. The same is true for the bound for the time-averaged delay. An interesting observation is that the time-averaged delay for the first trace file is larger than the worst case delay for the same trace, which indicates that the bound (5) may be very loose. We observed that this fact stems from the backlog bound, used in the derivation of (5), which may be loose. Another observation to be specified here is that the heuristic H^{MG} presented in Section IV generated the partition $\{\{SLM, JPM, SOH\}\}$.

The same conclusion can be drawn from Fig. 3, with the observation that no gain is obtained with respect to the time-averaged delay. Note that the heuristic H^{MG} generated, as expected, the partition $\{\{SLM, MBM, DHH, VVM\}\}$.

We also point out that for some sets of streams we analyzed, less dramatic multiplexed gains than shown in these examples have been observed. This was the case for the streams that lacked a sufficient degree of burstiness such as ‘‘Silence of the Lambs’’ (see Fig. 1). Moreover, no multiplexing gain occurs when identical streams are multiplexed together. Since the bursts for two identical streams occur at the same moments of time, the envelope \tilde{f} for the multiplexed stream does not differ from the sum f , of the envelopes corresponding to the considered streams.

Trace file	Avg. rate	# leaky buckets	S(\cdot)	Worst case delay	Time-averaged delay
Silence of the Lambs Medium quality (SLM)	876	15	1500*t	4632	6924
Mister Bean Medium quality (MBM)	920	15	2000*t	5133	2647
Die Hard III High quality (DHH)	3487	15	5000*t	10486	2362
VIVA TV Music Medium quality (VVM)	1738	15	3000*t	4088	2244
SLM+MBM+DHH+VVM	7021	15	11500*t	2853	2658

Fig. 3. Four trace files

VII. CONCLUSIONS

We have shown that multiplexing gain can be achieved in a deterministic network framework which involves traffic envelope functions, service curves, and the SCED scheduling algorithm. Our analysis is based on min-plus algebra. In particular, multiplexing gain arises when projections are done after multiplexing. In addition to worst-case delay as a deterministic QoS parameter, we showed analytically that multiplexing gain can also be achieved with respect to time-averaged delay. We point out that the results for deterministic multiplexing gain obtained in this paper apply directly to the end-to-end case via the end-to-end service curve convolution results from network calculus theory.

Using MPEG video traces, we presented numerical results demonstrating that multiplexing gains could be achieved under the worst-case delay criteria. We found that the (averaged) delay bounds obtained for the multiplexed streams were significantly smaller than those obtained for the individual streams when no multiplexing was done.

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