

Cross-Virtual Concatenation for Ethernet-over-SONET/SDH Networks

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Abstract

Ethernet-over-SONET/SDH (EoS) is a popular approach for interconnecting geographically distant Ethernet segments using a SONET/SDH transport infrastructure. It typically uses *virtual concatenation* (VC) for dynamic bandwidth management. The aggregate SONET/SDH bandwidth for a given EoS connection is obtained by “concatenating” a number of equal-capacity virtual channels. Together, these virtual channels form a *virtually concatenated group* (VCG). In this paper, we introduce a new concatenation technique, referred to as *cross-virtual concatenation* (CVC), which involves the concatenation of virtual channels of *heterogeneous* capacities. We show that CVC can be implemented through a simple upgrade at the end node, thus utilizing the existing legacy SDH infrastructure. By employing CVC for EoS systems, we show that the SDH bandwidth can be harvested more efficiently than in conventional VC. We consider two problems associated with routing CVC connections: the connection establishment problem and the connection upgrade problem. The goal of the first problem is to compute a set of paths between two EoS end systems such that a total bandwidth demand and a constraint on the differential delay between the paths are satisfied. Among all *feasible* sets, the one that consumes the least amount of network bandwidth is selected. For this problem, we develop an integer linear program (ILP) and an efficient algorithm based on the sliding-window approach. For the connection upgrade problem, the goal is to augment an existing set of paths so as to increase the aggregate bandwidth while continue to meet the differential-delay constraint. We model this problem as a flow-maximization problem with a constraint on the delay of the virtual channels with positive flow. We then consider the problem of path selection under imprecise network state information. Simulations are conducted to demonstrate the advantages of employing CVC and to evaluate the performance of the proposed algorithms.

1. Introduction

The current optical transport infrastructure is dominated by the SONET/SDH technology [1, 2]. SDH uses a bandwidth hierarchy indicated by STM- n , where $n = 1, 4, 16, 64, \dots$. The basic unit in this hierarchy is the STM-1 channel (155.52 Mbps), which can support various smaller payloads, including VC-11 (1.5 Mbps), VC-12 (2 Mbps), and VC-3 (45 Mbps). SDH was originally developed to support voice traffic. As the demand for “data” services (IP and Ethernet traffic) continued to rise, it became imperative to provide cost-effective solutions for supporting such services over the SDH infrastructure. One popular solution is based on Ethernet-over-SONET/SDH (EoS), which enables a network provider to interconnect Ethernet LANs over regional and even transcontinental distances [3, 4, 5].

EoS facilitates the provisioning of various data services, such as the Ethernet-private-leased-line service (which supports dedicated point-to-point bandwidth) and the Ethernet-virtual-private-line service (which uses statistical multiplexing for bandwidth sharing among various traffic streams).

In EoS, the aggregate bandwidth needed to interconnect two Ethernet segments is typically obtained by virtually concatenating several SDH payloads (VC- n channels) that possibly belong to different optical carriers (STM- ns) [2]. An example of two optical carriers containing virtually concatenated payloads is shown in Figure 1. The two carriers are routed along two different paths, whose delays are denoted by D_1 and D_2 , respectively. The differential delay between these two paths is $|D_1 - D_2|$. The two paths represent two *virtual channels*. Together, they form a *virtually concatenated group* (VCG).

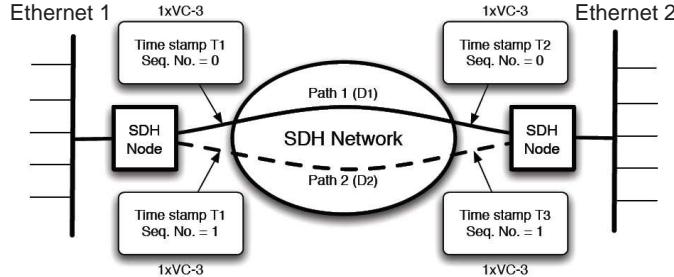


Figure 1: EoS setup with conventional virtual concatenation.

concatenated group (VCG). Different members of a VCG may be assigned noncontiguous slots within an SDH frame, and they can be routed independently. In conventional virtual concatenation, the virtual channels of a VCG are of the same SDH payload capacity. For a VCG of k virtual channels, the source node splits the traffic equally among the k channels. The data is byte-interleaved, e.g., if the first byte is transmitted on the r th virtual channel, then the next byte is transmitted on the $(r + 1)$ th virtual channel, and so on. Control information is embedded in overhead bytes (known as path overhead) associated with each virtual channel. The receiving end-point is responsible for re-assembling the original byte stream and compensating for the differential delay between the different paths taken by the virtual channels. This is done using a very high-speed memory. Once the SDH frames of all virtual channels that have the same frame number (i.e., time-stamp) have arrived at the destination node, Ethernet frames can be reconstructed. It is obvious that the higher the differential delay, the larger the memory requirement. Only the two end-points need to support VC [2]. By using the link capacity adjustment scheme (LCAS) [6], VC can support dynamic link upgrade and on-demand bandwidth adjustment without requiring additional hardware.

While the use of virtually concatenated EoS circuits for Ethernet transport enables efficient utilization of the SDH bandwidth, it may result in a large number of virtual channels between the two end-points. For example, consider the two Ethernet LANs in Figure 2, with an aggregate traffic demand of 100 Mbps. The two LANs can be connected by employing fifty VC-12 channels. This large number of circuits involves high maintenance overhead. The SDH bandwidth that is needed to support fifty VC-12 channels is actually greater than 100 Mbps. Specifically, the hierarchical implementation of an SDH frame [2] allows for a maximum of 21 VC-12 channels over a TUG-3 (tributary unit group) of an STM-1 frame. So, fifty VC-12 channels are transported using $45+45+(8/21)\times45 = 107$ Mbps of actual bandwidth. An alternative is to employ three VC-3 channels (3×45 Mbps = 135 Mbps), at the expense of excessive bandwidth wastage.

To improve bandwidth utilization while employing a relatively small number of connections, an EoS system needs to be able to combine virtual channels of different payloads in the same VCG. For example, the 100 Mbps traffic can be supported using two VC-3 channels (90 Mbps) and five VC-12 channels (10 Mbps). The actual SDH bandwidth

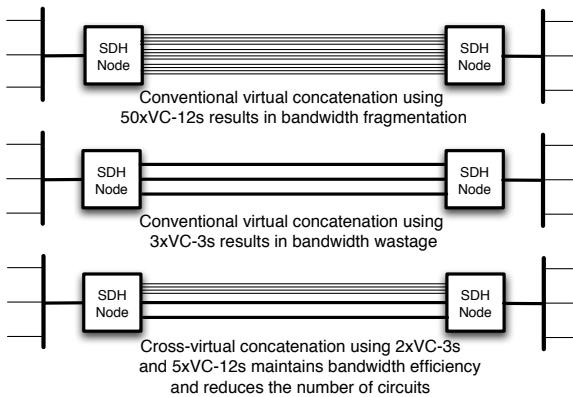


Figure 2: Example that demonstrates the advantages of CVC.

that will be consumed in this case is 101.9 Mbps ($2 \times 45 + (5/21) \times 45$). We refer to the concatenation of SDH channels of different payload capacities as *cross-virtual concatenation* (CVC). By employing CVC, we are able to harvest the SDH bandwidth more efficiently. We later show that the gain achieved by employing CVC increases linearly with the VCG bandwidth demand.

In this paper, we outline the general structure of the proposed CVC approach (Section 2) and describe its advantages over conventional virtual concatenation. We show a typical implementation of CVC based on VC-12 and VC-3 payloads. This implementation exploits control information already available in the SDH frame. It requires a simple end-point upgrade. We study two path selection problems associated with interconnecting two EoS endpoints via CVC. The first problem is that of establishing an EoS connection with a given bandwidth demand (Section 3). In this problem, the goal is to compute a set of paths that satisfy the given bandwidth demand and a given differential delay constraint, and that consumes the least amount of network bandwidth. We present an integer linear programming (ILP) formulation and an efficient algorithm to solve the connection establishment problem. The presented algorithm converts the problem into a flow-maximization problem and uses a modified version of the sliding-window approach [7] to find a feasible set of paths. We then consider the problem of augmenting an existing EoS connection in order to increase its aggregate bandwidth (Section 4). This problem is modelled as a flow-maximization problem with a two-sided delay constraint. We use the modified-link-weight K -shortest path (MLW-KSP) algorithm described [8] to compute the two-sided constrained paths. From these paths, we sequentially choose a set of paths that can satisfy the bandwidth and delay constraints. We also study the path selection problem when the information about link delays and bandwidth availability is not precisely known at the time of path selection (Section 5). We propose an algorithm based on the “backward forward” approach to solve this problem. We study the performance of the proposed algorithms (Section 6) and compare them with path selection under conventional virtual concatenation.

2. Implementation of Cross-Virtual Concatenation

In this section, we describe how CVC can be implemented via a simple end-point upgrade. For illustration purposes, we describe a particular instance of CVC, applied to the concatenation of VC-3 and VC-12 channels. However, the concept can be easily generalized for multiple payloads of different capacities. The implementation described in this section is divided into two parts: transmit side and receive side. This implementation is similar to a classic VC implementation but with a different mechanism for payload splitting and assembly.

2.1 Transmit Side

In a typical transmit-side implementation of EoS, Ethernet frames are encapsulated using the generic framing procedure (GFP) [9, 10] or the link access protocol-SDH (LAP-S) [2]. The resultant stream of bytes is then interleaved into the various members of the VCG. In the CVC implementation (see Figure 3), we use a special payload splitter and a buffer assembly, just after frame encapsulation. The splitter consists of a set of payload counters that are associated

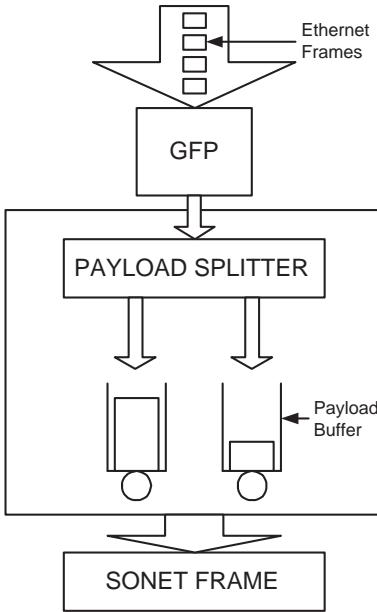


Figure 3: Transmit-side implementation of CVC.

with various payload types (VC-12, VC-3, etc.). A payload-select counter is used to choose the appropriate buffer for storing the incoming payload. Each payload counter maintains the number of bytes of a particular payload type present in an SDH frame. For example, if there are n VC-12 channels in a VCG, then the VC-12 payload counter counts from 0 to $34n - 1$ (there are 34 bytes of payload data in a VC-12 channel per SDH frame). Similarly, if there are m VC-3 channels in a VCG, the VC-3 payload counter counts from 0 to $765m - 1$ (there are 765 bytes of payload data in a VC-3 channel per SDH frame). The payload-select counter counts from 0 to $J - 1$, where J is the number of payload types participating in CVC; in this case two (VC-3s and VC-12s). The various states of the payload-select counter are used to generate enable signals for various payload counters. At any instant, only one payload counter is enabled.

There are J buffers (one for each payload type) for storing incoming payload bytes. When a byte is received from the GFP encapsulator, it is stored into the payload buffer for which the `write_enable` signal is high. The various virtual channels are fed from their respective buffers. Figure 4 depicts a timing diagram for the various counters. For example, in a CVC of one VC-3 and one VC-12 channels, there are 799 bytes per SDH frame. The first 765 bytes are stored in the VC-3 payload buffer and the next 34 bytes are stored in the VC-12 payload buffer. The splitting of payloads ensures that the ratio of traffic transmitted over different payload types is the same as the ratio of the number of bytes transmitted over each payload type in an SDH frame. As discussed later, such splitting does not add any complexity at the receiver. The multi-frame and sequence numbers are used in the same way as in conventional VC. For example, the sequence number for the n VC-12 channels ranges from 0 to $n - 1$, and for the

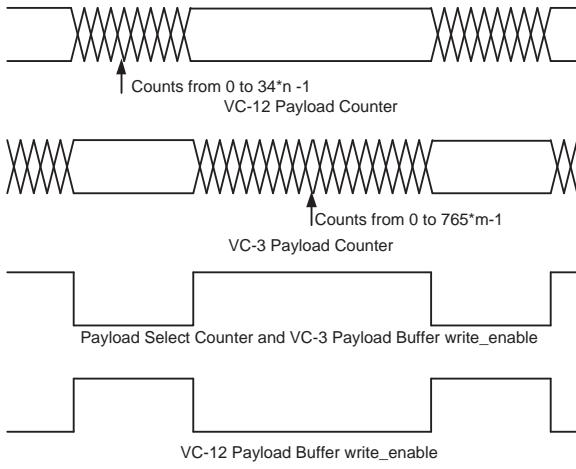


Figure 4: Timing diagram of various counters for the transmit-side implementation of CVC.

m VC-3 channels it ranges from 0 to $m - 1$. If there are multiple VC-3 or VC-12 channels, then as in conventional virtual concatenation, the payloads stored in different payload buffers are byte-interleaved when inserted into the SDH frame.

Each buffer is essentially a FIFO queue. After splitting, the data is written to the end of these queues. For a given VCG, the total amount of buffer needed to store the payload bytes is twice the number of bytes that are actually transmitted on the SDH frame. Two-frame worth of storage capacity is needed to maintain exclusion in read and write memory operations. In the worst case, buffering introduces an additional delay of $250 \mu\text{sec}$ (equivalent to the transmission time of two SDH frames). Since the complete data stream is delayed by a constant amount, such buffering does not add any extra complexity in the receive side.

2.2 Receive Side

In a typical receive-side implementation of EoS, the received SDH payload is stored in a *differential delay compensator*, which is essentially a buffer. Once the frames of all members of the VCG with the same frame number (time-stamp) have arrived, the payload bytes are sequentially removed from the buffer according to their sequence numbers, and are inputted to the GFP decapsulator. In a CVC implementation, we use different buffers for different payload types to compensate for the differential delay. Once the payloads of all the members of the VCG with the same frame number have arrived, the payload assembler sequentially extracts the various payload bytes. The assembler contains a set of counters that provide the appropriate addresses for fetching the payload bytes from the corresponding buffers (see Figure 5). The payload buffer also generates the payload-select signal, which is used as the select line of the multiplexer. The multiplexer chooses the correct payload bytes to be sent to the GFP module. An address generator generates appropriate read addresses for the payload buffers.

Note that the implementation of CVC described above does not require intermediate nodes along a path of a VC- n channel to be aware of the CVC connection. These intermediate nodes treat this VC- n as any other VC- n channel. Hence, implementing CVC requires hardware upgrade at the end nodes only.

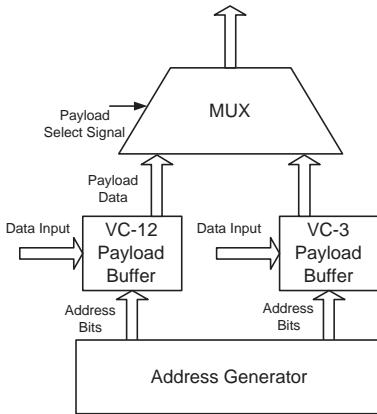


Figure 5: Receive-side implementation of CVC.

3. Connection Establishment Under CVC

In this section, we consider the problem of establishing a new CVC connection. Consider two remotely separated Ethernet LANs that need to be connected using EoS with a total bandwidth requirement \mathcal{R}_B . Let s and t be the two SDH end points. The goal is to find a set of paths between s and t that supports the required aggregate bandwidth, that satisfies a differential delay requirement, and that consumes the minimum possible network bandwidth. Note that while the paths that comprise the CVC connection may differ in their payloads, each path supports the same payload over all of its links. Before proceeding further, we define the following parameters:

- S_n : Actual bandwidth consumed by a VC- n channel within an SDH frame. For example, an STM-1 frame can transport three VC-3 channels. Hence, $S_3 = 155/3 = 51.66$ Mbps. Similarly, $S_{12} = 155/63 = 2.46$ Mbps for a VC-12 channel, and $S_4 = 155/1 = 155$ Mbps for a VC-4 channel. Note that S_n includes the overhead associated with transporting a given virtual channel within the SDH frame structure.
- \mathbb{B}_n : *Payload bandwidth* of a VC- n channel, i.e., the bandwidth associated with the user's data ($\mathbb{B}_3 = 45$ Mbps for a VC-3 channel, $\mathbb{B}_{12} = 2$ Mbps for a VC-12 channel, and $\mathbb{B}_4 = 140$ Mbps for a VC-4 channel).
- B_{uv} : Total payload bandwidth available over a link (u, v) ; $B_{uv} = \sum_n \mathbb{B}_n b_n(u, v)$, where $b_n(u, v)$ is the number of unused VC- n channels over link (u, v) .

The total amount of network resources consumed by a path P of $h(P)$ hops and of total connection bandwidth $f(P)$ is given by $h(P)f(P)$. Moreover, if a path P is chosen for routing a VC- n channel then this channel is reserved along all the links of path P .

We now formally define the connection establishment problem.

Problem 1 *Connection Establishment (CE): Consider an SDH network that is modeled as a graph $\mathcal{G}(\mathcal{N}, \mathcal{L})$, where each link $(u, v) \in \mathcal{L}$ is associated with a total available payload capacity $B_{uv} = \sum_n \mathbb{B}_n b_n(u, v)$ and delay $w(u, v)$. Find a set of paths between source s and destination t such that: (1) the aggregate capacity of these paths is greater than or equal to \mathcal{R}_B , (2) the maximum differential delay between two paths in the selected set is less than a given constant Δ , and (3) the bandwidth of the selected paths is minimum among all feasible sets of paths that satisfy (1) and (2).*

The example in Figure 6 illustrates the CE problem. In this example, an EoS with $\mathcal{R}_B = 100$ Mbps is to be established between nodes 1 and 5. For the two paths $P_1 = (1 - 2 - 4 - 5)$ and $P_2 = (1 - 2 - 3 - 5)$, the number of available VC-3 (VC-12) channels are 2 (2) and 1 (3), respectively. Note that a single path between nodes 1 and 5 cannot support the required 100 Mbps. For simplicity, suppose that P_1 and P_2 satisfy the differential delay constraint. A feasible solution involving two VC-3 channels along P_1 and one VC-3 channel along P_2 satisfies the bandwidth requirement but consumes excessive network resources. A CVC solution of two VC-3 and two VC-12 channels along P_1 , and three VC-12 channels along P_2 provides an optimal solution in this case.

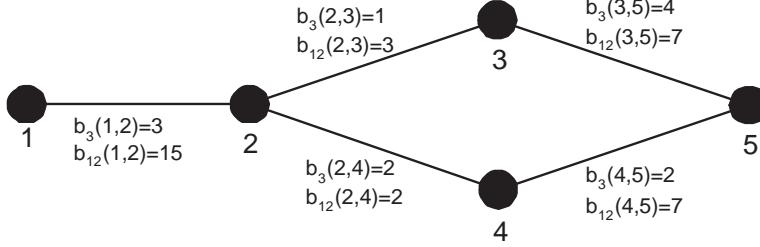


Figure 6: Example network for the connection establishment problem.

In [7], the authors considered the differential delay routing (DDR) problem with standard virtual concatenation. They showed that this problem is NP-complete and cannot be solved by any ϵ -approximation algorithm. The problem addressed in our paper is even more complicated than the DDR problem, and in the best case is equivalent to it (when only one type of payload is available). Hence, the CE problem is also NP-complete.

The CE problem can be addressed by splitting it into two parts. First, for the given network, we check the feasibility of the requested bandwidth. If it is feasible, we then compute a set of paths that satisfy the differential delay requirement. The first step can be solved by employing any standard maximum-flow algorithm (e.g., [11]). The second step requires the enumeration of all the paths, and is therefore NP-hard. Instead of computing these paths at the connection establishment time, we can precompute a set of K paths for each payload and then focus on the problem of choosing a subset of paths from these precomputed paths that satisfies the differential delay requirement and that has the desired total capacity with a minimum network bandwidth usage. We first propose an ILP formulation to solve the CE problem and later propose an efficient algorithm based on the sliding-window approach.

3.1 ILP formulation

For each payload type n , let $\mathbb{P}_n = \{P_1^{(n)}, P_2^{(n)}, \dots, P_K^{(n)}\}$ be a set of K precomputed paths that satisfy the differential delay requirement, i.e., $\max_{n,m} |D(P_i^{(n)}) - D(P_j^{(m)})| \leq \Delta$, where $D(P_j^{(m)})$ is the delay associated with path $P_j^{(m)}$. Define the indicator function $I_n^{(i)}(u, v)$ as follows: $I_n^{(i)}(u, v) = 1$, if $(u, v) \in P_i^{(n)}$, and is 0 otherwise. The set \mathbb{P}_n is an input to the ILP. Hence, $I_n^{(i)}(u, v)$ is known in advance $\forall (u, v) \in \mathcal{L}$. Let $N_n^{(i)}$ be an integer variable that indicates the number of VC- n channels along path $P_i^{(n)}$ that are in the solution set. Figure 7 depicts the ILP formulation. Constraint C1 ensures that the chosen paths and their capacities do not violate the capacities of various links. Notice that $I_n^{(i)}(u, v)S_nN_n^{(i)}$ is the total bandwidth consumed by $N_n^{(i)}$ VC- n channels over link (u, v) . Constraint C2 ensures that the solution satisfies the bandwidth requirement \mathcal{R}_B . The objective function minimizes the sum of the bandwidth reserved by the connection request over all the links in the network.

Minimize: $\sum_{(u,v) \in \mathcal{L}} \sum_n \sum_{i=1}^{|\mathbb{P}_n|} \{I_n^{(i)}(u,v)S_nN_n^{(i)}\}$
Subject to

$$\begin{aligned} \text{C1: } & \sum_n \sum_{i=1}^{|\mathbb{P}_n|} \{I_n^{(i)}(u,v)S_nN_n^{(i)}\} \leq B_{uv} & \forall (u,v) \in \mathcal{L} \\ \text{C2: } & \sum_{(u,v) \in \mathcal{L}: v=d} \sum_n \sum_{i=1}^{|\mathbb{P}_n|} \{I_n^{(i)}(u,v)\mathbb{B}_nN_n^{(i)}\} \geq \mathcal{R}_B \end{aligned}$$

Figure 7: ILP formulation for the connection establishment problem.

3.2 Sliding Window Algorithm

The set of paths returned by the ILP solution minimizes the total allocated bandwidth, but its worst-case complexity is exponential in the size of the network. Hence, we now propose a computationally efficient algorithm, which is a variant of the sliding-window approach presented in [7]. The algorithm uses the K-shortest path algorithm [12] to precompute a set of paths. It then sequentially tries these paths in search of a feasible solution. A pseudocode of the **Sliding-Window** algorithm is presented in Figure 8. It takes as input the graph $\mathcal{G}(\mathcal{N}, \mathcal{L})$, the source node s , the destination node t , the capacity requirement \mathcal{R}_B , and the differential delay constraint Δ . For each payload type n , the algorithm precomputes a set of k paths \mathbb{P}_n , ordered increasingly according to their delay values. Out of these paths, the algorithm forms a set $\mathcal{P} = \cup_n \mathbb{P}_n$, that contains the set of all paths ordered according to their delay values. In the first iteration, the algorithm considers all the paths P_1, P_2, \dots, P_r , where $P_r \in \mathcal{P}$ is the path with the highest delay value such that $D(P_r) - D(P_1) \leq \Delta$. After identifying these paths, the algorithm routes the maximum permissible flow along these paths, starting with the path that has the minimum hop count (in case of a tie, the path with the higher bandwidth is chosen). Note that the maximum flow $f(P)$ along a path P is given by $f(P) = \min_{(u,v) \in P} B_{uv}$. The algorithm routes an amount of flow $f(P)$ along the chosen path P . For each link $(u,v) \in P$, its capacity is recalculated as follows: $B_{uv} = B_{uv} - f(P)$ ¹. If the total added capacity along all the paths is greater than \mathcal{R}_B , the algorithm terminates.

In the j th iteration, the algorithm starts with the j th path (P_j) from the sorted list \mathcal{P} as the first path and then evaluates the path $P_{r'}$ as the maximum delay path in \mathcal{P} such that $D(P_{r'}) - D(P_j) \leq \Delta$. For the same iteration, we also maintain a set of paths \mathcal{P}_{itr} , which represents all the paths under consideration during the iteration. For example, for the j th iteration, $\mathcal{P}_{itr} = \{P_j, P_{j+1}, \dots, P_{r'}\}$.

Complexity: Precomputing a set of k paths for each payload type can be done in $\mathcal{O}(k|\mathcal{L}| \log(k|\mathcal{N}|) + k^2|\mathcal{N}|)$ time using the K-shortest path algorithm [12]. Let r be the number of payload types. The total complexity in computing the set \mathcal{P} is $\mathcal{O}(rk|\mathcal{L}| \log(k|\mathcal{N}|) + rk^2|\mathcal{N}|)$. Sorting \mathcal{P} based on hop count requires $\mathcal{O}(kr \log(kr))$ time. Each iteration in Step 4 requires updating the link capacities for a maximum of rk paths and each path can have $|\mathcal{N}|$ hops. Hence, the total complexity associated with each iteration is $rk|\mathcal{N}|$. The **Sliding-Window** algorithm can have a maximum of kr iterations. The total complexity associated with rk iterations is $\mathcal{O}(r^2k^2|\mathcal{N}|)$. Hence, the total complexity of the algorithm is $\mathcal{O}(rk|\mathcal{L}| \log(k|\mathcal{N}|) + rk^2|\mathcal{N}| + kr \log(kr) + r^2k^2|\mathcal{N}|)$.

¹Note that the flow along a set of paths is not the sum of the flows along the individual paths. Augmenting a flow along a path may reduce the flow on other paths that share common links with the first path. Determining the maximum flow for a set of n paths is an NP-hard problem [13].

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Sliding-Window( $\mathcal{G}(\mathcal{N}, \mathcal{L})$ ,  $s, t, w(\cdot)$ ,  $B_{(\cdot)}$ ,  $\mathbb{B}_n$ ,  $\Delta$ ,  $\mathcal{R}_B$ )
1. Precompute  $K$  paths  $\mathbb{P}_n = \{P_1^{(n)}, P_2^{(n)}, \dots, P_k^{(n)}\} \forall n$ ,
   ordered increasingly according to their delays.
2. Set  $\mathcal{P} = \cup_n \mathbb{P}_n, n = 3, 4, 12, \dots$ 
3. Sort  $\mathcal{P} \equiv \{P_1, P_2, \dots, P_{|\mathcal{P}|}\}$  w.r.t. path delay
4. For  $i = 1, 2, \dots, |\mathcal{P}|$ ,
5.   Compute  $\mathcal{P}_{itr} = \{P_i, P_{i+1}, \dots, P_r : D(P_r) - D(P_i) \leq \Delta\}$ 
6.   Sort the paths in  $\mathcal{P}_{itr}$  based on hop count
      Resolve ties according to path capacity
7.    $F_{itr} = 0, \mathcal{S} = \Phi$ 
8.   For  $j = i, i + 1, \dots, r$ ,
9.     Route maximum flow  $f(P_j)$  along  $P_j$ 
10.     $F_{itr} = F_{itr} + f(P_j), \mathcal{S} = \mathcal{S} \cup \{(P_j, f(P_j))\}$ 
12.    If  $F_{itr} > \mathcal{R}_B$ ,
13.      Return  $\mathcal{S}$ 
      end For
14.  Set  $B_{uv} \forall (u, v) \in \mathcal{L}$  to its initial value
      end For

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Figure 8: Pseudocode for the **Sliding-Window** algorithm.

4. Connection Upgrade Problem

We now consider the connection upgrade problem. Given an established EoS connection between two LANs, we want to upgrade it by adding \mathcal{R}_B Mbps bandwidth. The additional bandwidth can be harvested by using a cross-virtually concatenated payloads such that the resulting flow is feasible and the consumed bandwidth is minimal.

In [8], the connection upgrade problem for the conventional virtually concatenated EoS system was studied. The problem was reduced to a two-sided constraint path (TSCP) problem, which was heuristically solved using a modified version of the standard K -shortest path algorithm, referred to as modified link weight K -shortest path (MLW-KSP). TSCP was shown to be NP-complete [8]. The problem at hand is more complicated than the standard TSCP problem because it involves multiple payload types and multiple paths. It can be thought of as a generalization of the TSCP problem. We now formally state the connection upgrade problem:

Problem 2 Connection Upgrade (*CU*): Consider a graph $\mathcal{G}(\mathcal{N}, \mathcal{L})$, a source-destination pair (s, t) , and an established EoS connection with m members (paths) in the VCG whose respective path delays are D_1, D_2, \dots, D_m . Let Δ be the maximum allowable differential delay. Each link $(u, v) \in \mathcal{L}$ is associated with a nonnegative capacity B_{uv} . Given a required upgrade bandwidth \mathcal{R}_B , we need to find a set of paths \mathcal{P} between s and t that satisfies the following:

- The maximum bandwidth that can be routed along the set \mathcal{P} is greater than \mathcal{R}_B .
- The delay associated with each path $P_i \in \mathcal{P}$ satisfies:

$$|D(P_i) - D(P_j)| \leq \Delta, \forall P_i, P_j \in \mathcal{P}. \quad (1)$$

- The delay associated with each path $P_i \in \mathcal{P}$ satisfies the differential delay constraint with all the m existing

members of the VCG:

$$\max_{1 \leq j \leq m} |D_j - D(P_i)| \leq \Delta, \forall P_i \in \mathcal{P}. \quad (2)$$

- The network resource consumed by \mathcal{P} is minimum among all feasible solutions.

In Figure 9, we pictorially show how (2) results in a two-sided constraint (represented by a lower bound C_1 and an upper bound C_2) on the delay of a new virtual channel due to the presence of existing virtual channels.

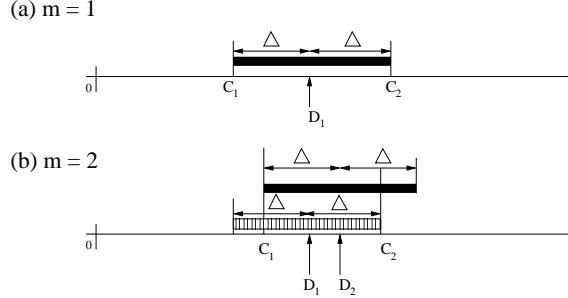


Figure 9: Pictorial representation of the differential delay requirement when the existing VCG contains: (a) one path, and (b) two paths.

4.1 Upgrade Algorithm

We now present an algorithm for solving the CU problem. A pseudocode for this algorithm, called **Upgrade** is shown in Figure 10. Its input is the graph $\mathcal{G}(\mathcal{N}, \mathcal{L})$, source node s , destination node t , the required capacity upgrade \mathcal{R}_B , the supported differential delay Δ , and two positive constants C_1 and C_2 that represent the minimum and maximum delays of the paths in the resulting solution set \mathcal{S} . The initial values of C_1 and C_2 are obtained using the m existing members of the VCG. Specifically, $C_1 = \max_{1 \leq i \leq m} (D_i - \Delta)$ and $C_2 = \min_{1 \leq i \leq m} (D_i + \Delta)$. For each payload type n , the algorithm precomputes a set of feasible paths \mathbb{P}_n by using the MLW-KSP algorithm described in [8]. The delay of each path $P \in \mathbb{P}_n$ satisfies $C_1 \leq D(P) \leq C_2$. The MLW-KSP algorithm may return less than k feasible paths. In that case, **Upgrade** explores fewer paths. The algorithm then computes the set of paths $\mathcal{P} = \cup_n \mathbb{P}_n = \{P_1, P_2, \dots, P_{|\mathcal{P}|}\}$, sorted increasingly according to the hop count. **Upgrade** sequentially considers paths in \mathcal{P} . In case of a tie, a path with higher available capacity is picked; further ties can be resolved randomly. For a path P_i , **Upgrade** performs the following:

- Routes the maximum flow $f(P_i)$ along P_i , where $f(P_i) = \min_{(u,v) \in P_i} B_{uv}$.
- Recomputes the new set of constraints $C_1 = \max(C_1, D(P_i) - \Delta)$ and $C_2 = \min(C_2, D(P_i) + \Delta)$.
- Updates the capacity associated with the links corresponding to path P_i .

Upgrade then removes from \mathcal{P} any path whose delay does not satisfy the differential delay constraint under the new values of C_1 and C_2 , or whose capacity is zero. The algorithm terminates if the bandwidth requirement is satisfied or if the set \mathcal{P} is empty.

Complexity: Precomputing a set of k paths for each payload type can be done in $\mathcal{O}(k|\mathcal{L}| \log(k|\mathcal{N}|) + k^2|\mathcal{N}|)$ time using the K-shortest path algorithm [12]. Let r be the total payload types. The total complexity in computing the set \mathcal{P} is $\mathcal{O}(rk|\mathcal{L}| \log(k|\mathcal{N}|) + rk^2|\mathcal{N}|)$. Sorting \mathcal{P} based on hop count requires $\mathcal{O}(kr \log(kr))$ time. Each iteration

```

Upgrade( $\mathcal{G}(\mathcal{N}, \mathcal{L})$ ,  $s, t, w(.)$ ,  $B_{(.)}$ ,  $C_1, C_2$ ,  $\mathbb{B}_n$ ,  $\Delta$ ,  $\mathcal{R}_B$ )
1. Precompute  $k$  paths  $\mathbb{P}_n = \{P_1^{(n)}, P_2^{(n)}, \dots, P_k^{(n)}\}$ ,  $\forall n$ 
   using MLW-KSP with  $C_1$  and  $C_2$  as the two-sided
   constraints.
2.  $\mathcal{P} = \cup_n \mathbb{P}_n, n = 3, 4, 12, \dots$ 
3. Sort  $\mathcal{P} \equiv \{P_1, P_2, \dots, P_{|\mathcal{P}|}\}$  w.r.t. path delay
4.  $F = 0$ ,  $\mathcal{S} = \Phi$ 
5. While  $\mathcal{P}$  is not empty,
6.   Remove path  $P$  from  $\mathcal{P}$  with minimum hop count; in
      case of a tie, pick a path with maximum bandwidth
7.   Route maximum flow  $f(P)$  along path  $P$ 
8.   Update  $C_1$  and  $C_2$ 
9.    $\mathcal{S} = \mathcal{S} \cup \{(P, f(P))\}$ ,  $F = F + f(P)$ 
10.   $B_{uv} = B_{uv} - f(P), \forall (u, v) \in P$ 
11.  Remove path  $P' \in \mathcal{P}$  if
12.     $f(P') = 0$  or  $D(P) < C_1$  or  $D(P) > C_2$ 
13.  If  $F > \mathcal{R}_B$ , /* feasible solution found */
14.    Return  $\mathcal{S}$ 
   end If
15. end While

```

Figure 10: Pseudocode for the **Upgrade** algorithm.

in Step 5 requires updating $\mathcal{O}(|\mathcal{L}|)$ link capacities and identifying paths with zero capacities and which are infeasible under then new values of C_1 and C_2 . This can be done in $\mathcal{O}(rk|\mathcal{L}|)$ time. Hence, the total complexity of the algorithm is $\mathcal{O}(rk|\mathcal{L}| \log(k|\mathcal{N}|) + rk^2|\mathcal{N}| + kr \log(kr) + rk|\mathcal{L}|)$.

5. Path computation under Inaccurate Network State Information

In Sections 3 and 4, we considered the path selection problem under the assumption that the delay and bandwidth (available number of VC- n 's) over each link are known exactly. In practice, however, these quantities may not be precisely known to the source node due to several factors [14, 15, 16], including sampling errors, propagation delays, topology state aggregation in hierarchical networks, and trigger-based advertisement policies. In this section, we provide a mechanism to solve the path selection problem with CVC when the delay and bandwidth information for a link is not precisely known at the time of path selection. Specifically, we solve the CU problem under imprecise knowledge of network parameters. A similar approach can also be devised for the CE problem.

5.1 Problem Formulation

To model uncertainties in the delay parameter, we follow a probabilistic approach, where we assume that for each link (u, v) the delay parameter $w(u, v)$ is a nonnegative random variable with mean $\mu_d(u, v)$ and variance $\sigma_d^2(u, v)$. We assume that link delays are mutually independent. Each node u computes the parameters $\mu_d(u, v)$ and $\sigma_d^2(u, v)$ for each outgoing link (u, v) , and disseminates them throughout the network. For more discussion on the feasibility and effectiveness of capturing and disseminating such information, we refer the reader to [14]. We use the same probabilistic model in [8] to deal with the most-probable two-sided constrained path (MP-TSCP) problem. In our work, however, the problem at hand involves an additional constraint on the path bandwidth. Uncertainty

in bandwidth information is attributed to the nonzero delay in disseminating this information to all the nodes. To model this uncertainty, we again follow a probabilistic approach where we assume that for each link (u, v) the number of available VC- n channels ($b_n(u, v)$) is a nonnegative integer random variable with a uniform distribution over the range $\{0, 1, \dots, b_n^{\max}(u, v)\}$. We assume that the number of available VC- n channels of different links are mutually independent. The delay of a link is assumed to be independent of the bandwidth available over that link. The CU problem under inaccurate state information can be formally stated as follows.

Problem 3 Most-Probable Connection Upgrade (MP-CU): *Consider a network $\mathcal{G}(\mathcal{N}, \mathcal{L})$. Each link $(u, v) \in \mathcal{L}$ is associated with an additive delay parameter $w(u, v)$ and a set of VC- n channels $b_n(u, v)$. Assume that the $w(u, v)$'s are nonnegative independent random variables with mean $\mu_d(u, v)$ and variance $\sigma_d^2(u, v)$. For each payload type n , assume that $b_n(u, v)$ is uniformly distributed over $\{0, 1, \dots, b_n^{\max}(u, v)\}$. Let \mathbb{P} be the set of all paths from s to t . For a path $P \in \mathbb{P}$, let $W(P) \stackrel{\text{def}}{=} \sum_{(u,v) \in P} w(u, v)$ and $B(P) \stackrel{\text{def}}{=} \min_{(u,v) \in P} \sum_n \mathbb{B}_n b_n(u, v)$. Given two constraints C_1 and C_2 on the delay of the path and a bandwidth constraint $C_3 (= \mathcal{R}_B)$, the problem is to find a path that is most likely to satisfy these constraints. Specifically, the problem is to find a path r^* such that for any other path $P \in \mathbb{P}$*

$$\pi(r^*) \geq \pi(P) \quad (3)$$

where

$$\pi(P) \stackrel{\text{def}}{=} \Pr[C_1 \leq W(P) \leq C_2, B(P) > C_3] = \Pr[C_1 \leq W(P) \leq C_2] \Pr[B(P) > C_3].$$

Notice that the problem statement focusses on finding a path (and not a set of paths) that satisfies the delay and bandwidth constraints in a probabilistic sense. If the bandwidth requirement is so high that no single path can satisfy it, then the proposed solution to the MP-CU problem can be executed multiple times with successively smaller values for C_3 . Each execution will use different delay and bandwidth constraints, and will result in a path. The aggregate bandwidth of the set of paths that are computed from all iterations will form a multi-path solution. The algorithm terminates when the bandwidth of this multi-path solution satisfies C_3 .

5.2 Backward Forward (BF) Algorithm

We now present a heuristic solution for MP-CU, referred to as the BF algorithm. The basic idea in this algorithm is similar to that of the BF-Inaccurate algorithm discussed in [8].

By exploiting the central limit theorem (CLT), we approximate the probability density function (pdf) of the delay of a path P ($W(P) = \sum_{(u,v) \in P} w(u, v)$) by a Gaussian distribution. For a given path $P \in \mathbb{P}$, let $\mu_d(P) \stackrel{\text{def}}{=} \sum_{(u,v) \in P} \mu_d(u, v)$, $\sigma_d^2(P) \stackrel{\text{def}}{=} \sum_{(u,v) \in P} \sigma_d^2(u, v)$, and let $F(x, \mu_d(P), \sigma_d^2(P))$ be the cumulative distribution function (CDF) of a Gaussian random variable with mean $\mu_d(P)$ and variance $\sigma_d^2(P)$ evaluated at x . For such a path, $\pi(P)$ can now be calculated as follows:

$$\begin{aligned} \Pr[B(P) > C_3] &= \prod_{(u,v) \in P} \Pr[b(u, v) > C_3] \\ \Pr[C_1 \leq W(P) \leq C_2] &= F(C_2, \mu_d(P), \sigma_d^2(P)) - F(C_1, \mu_d(P), \sigma_d^2(P)) \\ &\stackrel{\text{def}}{=} \Phi(C_2, C_1, \mu_d(P), \sigma_d^2(P)) \end{aligned}$$

Hence,

$$\pi(P) = \Phi(C_2, C_1, \mu_d(P), \sigma_d^2(P)) \prod_{(u,v) \in P} \Pr[b(u,v) > C_3].$$

The basic idea of the BF algorithm is to find an $s \rightarrow t$ path at any node u based on the already traversed segment $s \rightarrow u$ and the estimated remaining segment $u \rightarrow t$. The estimated remaining segment $u \rightarrow t$ can be one of the following:

- shortest $u \rightarrow t$ path w.r.t. $\mu_d(\cdot)$,
- shortest $u \rightarrow t$ path w.r.t. $\sigma_d^2(\cdot)$, or
- most-probable $u \rightarrow t$ path w.r.t. $b(\cdot)$ (an algorithm for finding such a path is presented in [14]).

Among the three possible sub-paths, the algorithm chooses the one that maximizes $\pi(s \rightarrow u \rightarrow v \rightarrow t)$. Notice that the subpath $s \rightarrow u \rightarrow v$ of path P is already known at v . Only subpath $v \rightarrow t$ is chosen from the three paths described above. This way of estimating the $u \rightarrow t$ path offers greater flexibility in choosing complete paths.

Because the BF algorithm considers complete paths, it can foresee several paths before reaching the destination.

During the execution of the algorithm, each node u maintains the following set of labels and uses them to compute the path that maximizes $\pi(\cdot)$:

- $D[u] = \{D_b[u], D_1[u], D_2[u]\}$, where $D_b[u]$ is the probability that the bandwidth on the shortest $u \rightarrow t$ path w.r.t. $\mu_d(\cdot)$ is greater than C_3 , $D_1[u]$ is the mean delay, and $D_2[u]$ is the variance of the delay of the shortest $u \rightarrow t$ path w.r.t. $\mu_d(\cdot)$.
- $E[u] = \{E_b[u], E_1[u], E_2[u]\}$, where $E_b[u]$ is the probability that the bandwidth on the shortest $u \rightarrow t$ path w.r.t. $\sigma_d^2(\cdot)$ is greater than C_3 , $E_1[u]$ is the mean delay, and $E_2[u]$ is the variance of the delay of the shortest $u \rightarrow t$ path w.r.t. $\sigma_d^2(\cdot)$.
- $F[u] = \{F_b[u], F_1[u], F_2[u]\}$, where $F_b[u]$ is the probability that the bandwidth on the most probable $u \rightarrow t$ path w.r.t. $b(\cdot)$ is greater than C_3 , $F_1[u]$ is the mean delay, and $F_2[u]$ is the variance of the delay of the most probable $u \rightarrow t$ path w.r.t. $b(\cdot)$.

Each node also maintains $\pi_s[u]$, the predecessor of u on the path from s to t . This parameter is used for the reconstruction of the path after the termination of the algorithm. A detailed explanation about how these variables are updated is described later in the paper.

We define $P_D[B(s \rightarrow u \rightarrow v \rightarrow t) > C_3]$ as the probability that the bandwidth along the path $s \rightarrow u \rightarrow v \rightarrow t$ is greater than C_3 , where the subpath $v \rightarrow t$ is the shortest $v \rightarrow t$ path w.r.t. $\mu_d(\cdot)$. The subpath $s \rightarrow u \rightarrow v$ is known in advance at node v , and an estimate of $v \rightarrow t$ is used in the computation. Similarly, $P_E[B(s \rightarrow u \rightarrow v \rightarrow t) > C_3]$ and $P_F[B(s \rightarrow u \rightarrow v \rightarrow t) > C_3]$ are defined as the probabilities that the bandwidth along the path $s \rightarrow u \rightarrow v \rightarrow t$ is greater than C_3 , where the path segment $v \rightarrow t$ is the shortest $v \rightarrow t$ path w.r.t. $\sigma_d^2(\cdot)$ and $b(\cdot)$, respectively.

The BF algorithm has two phases: (i) a backward phase from the destination node t to all other nodes, in which each node u is assigned the labels $D_b[u]$, $D_1[u]$, $D_2[u]$, $E_b[u]$, $E_1[u]$, $E_2[u]$, $F_b[u]$, $F_1[u]$, and $F_2[u]$ using reverse Dijkstra's algorithm (RDA) [11], and (ii) a forward phase to find the most likely path that minimizes $1 - \pi$. A pseudocode for the BF algorithm is presented in Figure 12. RDA is first executed over the graph w.r.t. $\mu_d(\cdot)$ to calculate $D_b[u]$, $D_1[u]$, and $D_2[u]$; then w.r.t. $\sigma_d^2(\cdot)$ to calculate $E_b[u]$, $E_1[u]$, and $E_2[u]$; and finally w.r.t. $b(\cdot)$ to calculate $F_b[u]$, $F_1[u]$, and $F_2[u]$. To find the shortest path w.r.t. $b(\cdot)$, we use the MP-BCP version of Dijkstra's

algorithm discussed in [14]. The forward phase is a modified version of Dijkstra's algorithm. It uses information provided by the RDA to find the next node from the already travelled segment. For example consider Figure 11, in which the search has already traversed from s to u . The next node v is determined from the unexplored nodes by

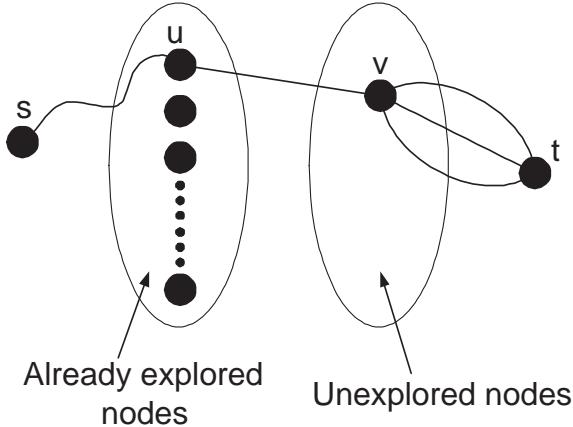


Figure 11: Pictorial representaion of the forward phase of the **Backward-Forward** algorithm.

finding the node that has the smallest cost function $1 - \pi$ for the path $s \rightarrow u \rightarrow v \rightarrow t$. Note that in this case, for each such node v , there are three possible paths from v to t , as calculated in the backward step. A pseudocode for the forward phase is shown in Figure 13. The performance of this phase can be further improved by using it with K-shortest path (KSP) implementation of Dijkstra's algorithm.

```

Backward-Forward( $\mathcal{G}, s, t, b(\cdot), \mu_d(\cdot), \sigma_d^2(\cdot), C_1, C_2, C_3$ )
1a.  $D = \text{Reverse\_Dijkstra}(\mathcal{G}, t, \mu_d(\cdot))$ 
1b.  $E = \text{Reverse\_Dijkstra}(\mathcal{G}, t, \sigma_d^2(\cdot))$ 
1c.  $F = \text{Reverse\_Dijkstra}(\mathcal{G}, t, b(\cdot), C_3)$ 
2.  $P = \text{Forward-Phase}(\mathcal{G}, s, t, b(\cdot), \mu_d(\cdot),
\sigma_d^2(\cdot), C_1, C_2, C_3, D, E, F)$ 

```

Figure 12: Pseudocode for the general structure of the **Backward-Forward** algorithm.

Complexity: The backward phase is performed by executing three instance of Dijkstra's algorithm which can be done in $\mathcal{O}(|\mathcal{N}| \log(|\mathcal{N}|) + |\mathcal{L}|)$ time. The forward phase is also essentially an instance of Dijkstra's algorithm. Hence, the overall complexity of BF algorithm is $\mathcal{O}(|\mathcal{N}| \log(|\mathcal{N}|) + |\mathcal{L}|)$.

6. Simulation Results

We conduct extensive simulations to evaluate the performance of our algorithmic solutions. Our interest here is not only to assess the goodness of these solutions, but to also demonstrate the effectiveness of the CVC approach in general. Our simulations are based on random topologies that obey the recently observed Internet power laws [17]. These topologies were generated using the BRUTE software [18]. In a given topology, each link (u, v) is assigned a delay value $w(u, v)$ that is sampled from a uniform distribution $\mathcal{U}[0, 50]$ msec. To model a meaningful EoS scenario,

```

Forward-Phase( $\mathcal{G}(\mathcal{N}, \mathcal{L})$ ,  $s, t, b(\cdot), \mu_d(\cdot), \sigma_d^2(\cdot), C_1, C_2, C_3, D, E, F$ )
1. for all  $i \in \mathcal{N}$ ,  $i \neq s$ ,
    $S[i] = \infty$ 
    $\pi_s[i] = \text{NIL}$ 
   end for
2.  $S[s] = 0$ ,  $\mu_d(s, s) = 0$ ,  $\sigma_d^2(s, s) = 0$ ,
3. Insert_Heap( $s, S[s], Q$ )
4. while  $Q$  is not empty,
5.    $u = \text{ExtractMin}(Q)$ 
6.   if ( $u == t$ ),
      return: The path traversed  $s \rightarrow t$ 
   end if
7.   for each edge  $(u, v)$  outgoing from  $u$ ,
8a.      $\mu_D(v) = \mu_d(s \rightarrow u) + \mu_d(u, v) + D_1[v]$ 
8b.      $\mu_E(v) = \mu_d(s \rightarrow u) + \mu_d(u, v) + E_1[v]$ 
8c.      $\mu_F(v) = \mu_d(s \rightarrow u) + \mu_d(u, v) + F_1[v]$ 
9a.      $\sigma_D^2(v) = \sigma_d^2(s \rightarrow u) + \sigma_d^2(u, v) + D_2[v]$ 
9b.      $\sigma_E^2(v) = \sigma_d^2(s \rightarrow u) + \sigma_d^2(u, v) + E_2[v]$ 
9c.      $\sigma_F^2(v) = \sigma_d^2(s \rightarrow u) + \sigma_d^2(u, v) + F_2[v]$ 
10.    if  $\{S[v] > 1 - \Phi(C_2, C_1, \mu_D(v), \sigma_D^2(v))P_D(B(s \rightarrow u \rightarrow v \rightarrow t) > C_3)\}$ ,
         $S[v] = 1 - \Phi(C_2, C_1, \mu_D(v), \sigma_D^2(v))P_D(B(s \rightarrow u \rightarrow v \rightarrow t) > C_3)$ 
         $\pi_s[v] = u$ ,  $\mu_d(s \rightarrow v) = \mu_d(s \rightarrow u) + \mu_d(u, v)$ 
         $\sigma_d^2(s \rightarrow v) = \sigma_d^2(s \rightarrow u) + \sigma_d^2(u, v)$ 
        Insert_Heap( $v, S[v], Q$ )
    end if
11.   if  $\{S[v] > 1 - \Phi(C_2, C_1, \mu_E(v), \sigma_E^2(v))P_E(B(s \rightarrow u \rightarrow v \rightarrow t) > C_3)\}$ ,
         $S[v] = 1 - \Phi(C_2, C_1, \mu_E(v), \sigma_E^2(v))P_E(B(s \rightarrow u \rightarrow v \rightarrow t) > C_3)$ 
         $\pi_s[v] = u$ ,  $\mu_d(s \rightarrow v) = \mu_d(s \rightarrow u) + \mu_d(u, v)$ 
         $\sigma_d^2(s \rightarrow v) = \sigma_d^2(s \rightarrow u) + \sigma_d^2(u, v)$ 
        Insert_Heap( $v, S[v], Q$ )
    end if
12.   if  $\{S[v] > 1 - \Phi(C_2, C_1, \mu_F(v), \sigma_F^2(v))P_F(B(s \rightarrow u \rightarrow v \rightarrow t) > C_3)\}$ ,
         $S[v] = 1 - \Phi(C_2, C_1, \mu_F(v), \sigma_F^2(v))P_F(B(s \rightarrow u \rightarrow v \rightarrow t) > C_3)$ 
         $\pi_s[v] = u$ ,  $\mu_d(s \rightarrow v) = \mu_d(s \rightarrow u) + \mu_d(u, v)$ 
         $\sigma_d^2(s \rightarrow v) = \sigma_d^2(s \rightarrow u) + \sigma_d^2(u, v)$ 
        Insert_Heap( $v, S[v], Q$ )
    end if
end for

Function  $u = \text{ExtractMin}(Q)$  removes and returns the vertex  $u$  in the heap  $Q$  with the least key value.
Function  $\text{Insert\_Heap}(v, x, Q)$  inserts the node  $v$  in the heap  $Q$  with a key value  $x$ . If the node is already present in the heap, then its key is decreased to  $x$ .

```

Figure 13: Pseudocode for the **Forward-Phase** of **Backward-Forward** algorithm.

we use VC-12 (2 Mbps) and VC-3 (45 Mbps) SDH payloads over each link. The available number of VC-12s and VC-3s over link (u, v) , indicated by $b_{12}(u, v)$ and $b_3(u, v)$, are sampled from $\mathcal{U}[1,5]$ and $\mathcal{U}[1,50]$, respectively. We assume that each node has a cross-connect that supports a minimum cross-connection granularity of VC-12, i.e., each VC-3 over a link can be used as 21 VC-12 channels.

6.1 Results for the Connection Establishment Problem

For the CE problem, we study the performance of the **Sliding-Window** algorithm described in Section 3. We compare its performance with standard virtual concatenation. For a given bandwidth demand \mathcal{R}_B and a given differential delay constraint Δ , we randomize the selection of the source-destination pair. For a given (s, t) pair, if the algorithm finds a set of paths that supports the required bandwidth, then we call it a *hit*; otherwise, we call it a *miss*. For the standard virtual concatenation case, we separately consider the VC-12 and VC-3 payloads over the links, and execute the **Sliding-Window** algorithm for each type. Note that the number of available VC-12s over a link (u, v) in a standard VC-12 concatenation is $21b_3(u, v) + b_{12}(u, v)$. Our performance metrics are the consumed network bandwidth and the probability of a miss. For a set of paths \mathbb{P} , the former metric is given by $\sum_{P \in \mathbb{P}} h(P)f(P)$.

Figure 14 depicts the consumed bandwidth versus \mathcal{R}_B . Clearly, the performance of CVC is better than the other two standard concatenation techniques. Notice that the gain achieved by CVC increases with the required bandwidth. Figure 15 depicts the miss probability as a function of \mathcal{R}_B . Again, CVC performs better than the two

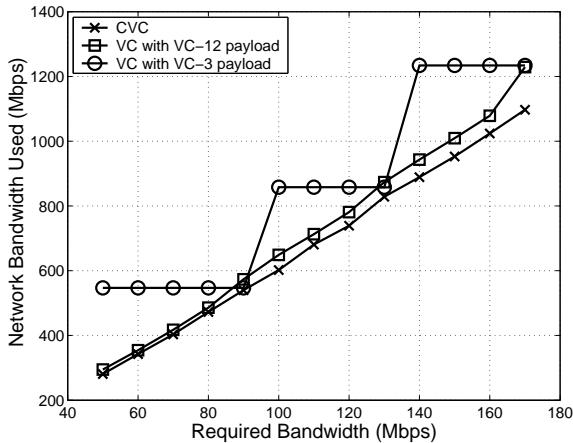


Figure 14: Network bandwidth used versus \mathcal{R}_B ($\Delta = 70$ msec, 15 precomputed paths, and a network of 100 nodes).

standard concatenation techniques. The gain in the miss probability increases with \mathcal{R}_B because CVC is able to efficiently utilize network bandwidth.

Consider now an EoS interconnection between two LANs that are connected using a single hop, as in Figure 2. For such a link Figure 16 depicts the additional bandwidth occupied by the standard concatenation techniques with VC-12, VC-3, and VC-4 relative to CVC. Notice that the extra bandwidth used by virtual concatenation with VC-12s and VC-3s increases roughly linearly with \mathcal{R}_B . This is due to the path overheads of VC-12/VC-3s. For standard concatenation using VC-4, the additional bandwidth over the link follows a saw-tooth pattern. If \mathcal{R}_B is an integral multiple of \mathbb{B}_4 ($= 140$ Mbps), then the CVC solution is same as the standard concatenation with VC-4. Note that for $\mathcal{R}_B = n\mathbb{B}_4 + \delta$ with $\delta < 140$ Mbps, standard concatenation with VC-4s will require $n + 1$ VC-4s over the link. This explains the observed saw-tooth behavior.

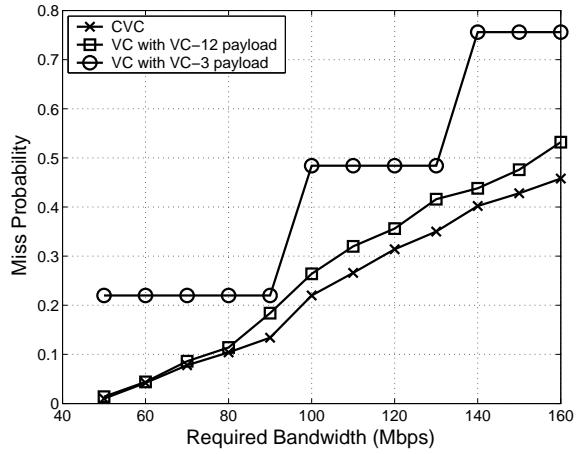


Figure 15: Miss probability versus \mathcal{R}_B ($\Delta = 70$ msecs, 15 precomputed paths, and a network of 100 nodes).

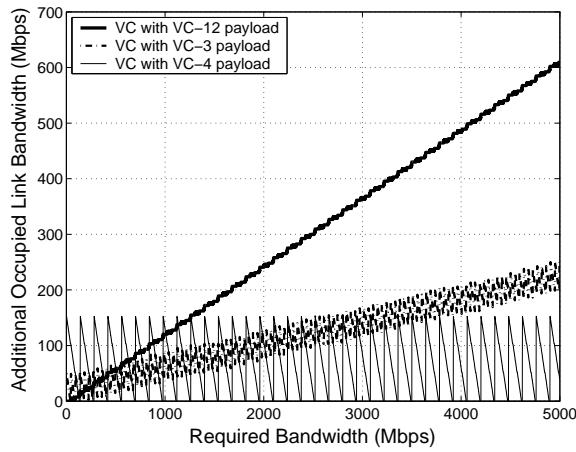


Figure 16: Additional bandwidth over CVC that is required by standard concatenation with VC-12, VC-3, and VC-4.

6.2 Results for the Connection Upgrade Problem

For the CU problem, we study the performance of the `Upgrade` algorithm proposed in Section 4. For a given simulation run, the values of C_1 and C_2 fall into one of the following three cases:

1. The delay of the shortest path between s and t w.r.t. $w(.,.)$ is greater than C_2 . In this case, there is no feasible solution to the problem.
2. The delay of the shortest path between s and t w.r.t. $w(.,.)$ is less than C_2 but greater than C_1 .
3. The delay of the shortest path between s and t w.r.t. $w(.,.)$ is less than C_1 .

Case 2 can be obtained from Case 3 by re-adjusting C_1 ($= W(P^*)$) and Δ ($= C_2 - W(P^*)$), where P^* is the shortest path between s and t w.r.t. $w(.,.)$. Case 3 is nontrivial and is the one considered in our simulations. Accordingly, C_1 and C_2 are generated such that they are always greater than the length of the shortest path between s and t w.r.t. $w(.,.)$. Specifically, we let $C_1 = W(P^*) + A + \mathcal{U}[0, 50]$ and $C_2 = C_1 + \Delta$, where A is a positive constant.

Figure 17 compares the `Upgrade` algorithm with the standard virtual concatenation (VC-12 and VC-3) in terms of required network bandwidth by varying \mathcal{R}_B . The performance of CVC is better than the other two concatenation techniques. The performance gain increases with \mathcal{R}_B .

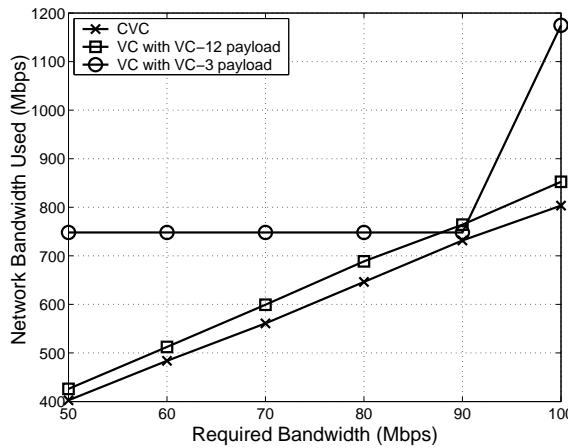


Figure 17: Network bandwidth used versus \mathcal{R}_B ($\Delta = 70$ msec, $A = 50$ msec, 15 precomputed paths, and a network of 100 nodes).

In Figure 18, we compare the performance in terms of the miss probability.

6.3 Result for Path Computation under Imprecise Network State Information

For a given topology, each link (u, v) is assigned a random delay $w(u, v)$. We assume that $w(u, v)$ is normally distributed with mean $\mu_d(u, v)$ and variance $\sigma_d^2(u, v)$. To produce heterogeneous link delays, we also randomize the selection of $\mu_d(u, v)$ and $\sigma_d^2(u, v)$ by taking $\mu_d(u, v) \sim \mathcal{U}[10, 50]$ msec and $\sigma_d^2(u, v) \sim \mathcal{U}[10, 50]$ msec². The true value of the link delay ($w^*(u, v)$) is sampled from a Gaussian distribution with mean $\mu_d(u, v)$ and variance $\sigma_d^2(u, v)$. This w^* is used to check if the path returned by the algorithm is feasible. We set C_1 and C_2 as in Section 6.2, where P^* is now the shortest path between s and t w.r.t. $w^*(.,.)$. The available number of VC- n channels over a link $b_n(u, v)$ is sampled from $\mathcal{U}[lb_n(u, v), ub_n(u, v)]$, where we take $lb_3(.,.) = 2$, $lb_{12}(.,.) = 10$, $ub_3(.,.) = 10$, and $ub_{12}(.,.) = 50$. The exact value of the number of VC- n channels available over a link ($b_n^*(u, v)$) is sampled from $\mathcal{U}[lb_n(u, v), ub_n(u, v)]$.

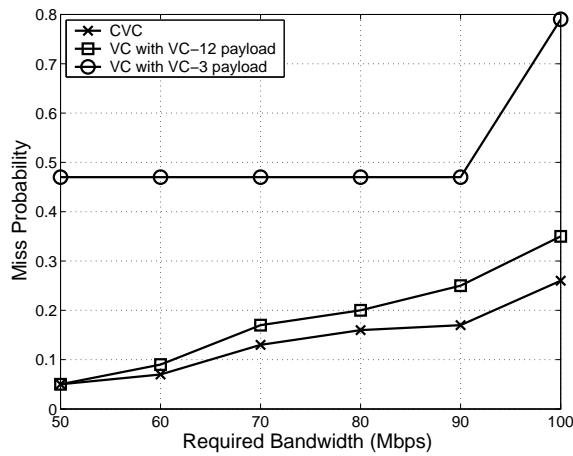


Figure 18: Miss probability versus \mathcal{R}_B ($\Delta = 70$ msecs, $A = 50$ msecs, 15 precomputed paths, and a network of 100 nodes).

Let $W(P_r)$ be the weight of this path P_r that is returned by the **Backward-Forward** algorithm w.r.t. $w^*(.,.)$, and let b^* be the available bandwidth. If $C_1 \leq W(P_r) \leq C_2$ and if $C_3 \leq b^*$, then we count it as a hit.

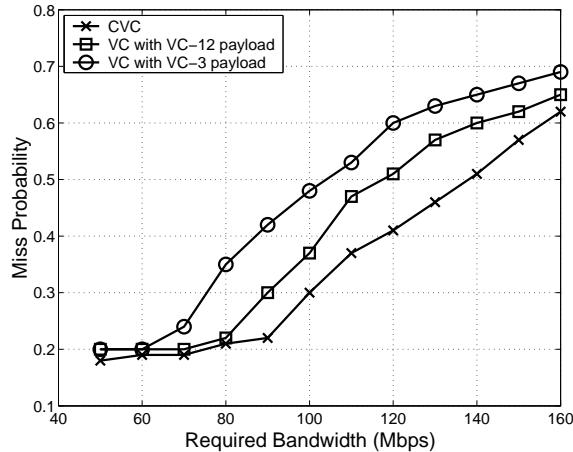


Figure 19: Miss probability vs. \mathcal{R}_B for path selection under imprecise information ($\Delta = 70$ msecs, $A = 30$ msecs, and a network of 100 nodes).

In Figure 19, we depict the miss probability for a connection request versus \mathcal{R}_B . As shown in the figure, CVC performs better than the other two concatenation techniques. Figure 20 shows the average number of iterations needed to find a feasible path when no limit is imposed on K . Clearly, CVC incurs less computational time than the two concatenation techniques considered.

7. Conclusions and Summary

In this paper, we introduced the general structure of cross-virtual concatenation (CVC) for EoS circuits. We proposed a simple implementation of CVC that involves an upgrade at the end nodes and that reuses the existing SDH overhead. We demonstrated the bandwidth gain of CVC over standard virtual concatenation. The algorithmic problems associated with connection establishment and connection upgrade were studied. For the establishment of a new EoS connection, we modeled the problem as a flow-maximization problem. An ILP solution as well as an

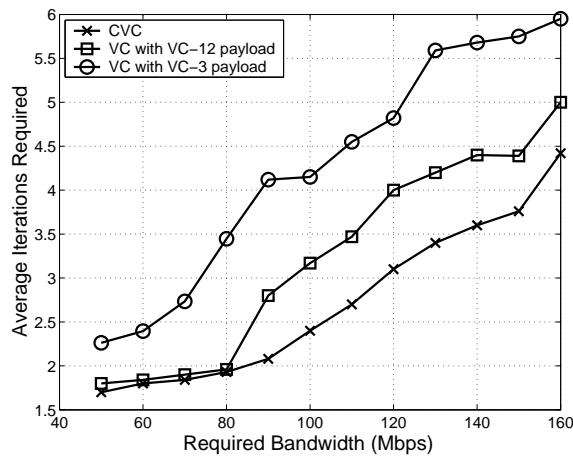


Figure 20: Number of iterations needed versus connection bandwidth ($\Delta = 70$ msec, $A = 30$ msec, and a network of 100 nodes).

algorithm based on the sliding-window approach were proposed. For the connection upgrade problem, we modeled it as a flow-maximization problem where the delay of each path in the solution set is required to satisfy the differential delay constraint of the existing members of the VCG. We proposed an efficient algorithm called **Upgrade** for solving this problem. We also considered the problem of path selection when the network state information is imprecisely available at the time of path selection. Extensive simulations were conducted to show the effectiveness of employing CVC for EoS systems. We contrasted the performance of CVC with the standard virtual concatenation in terms of the miss probability and consumed network bandwidth. The results show that CVC harvests the SDH bandwidth more efficiently, provides greater flexibility in terms of path selection, and reduces the miss probability for connection establishment.

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