

# Modeling TCP-Vegas under On/Off Traffic <sup>1</sup>

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There has been a significant amount of research toward modeling variants of the Transmission Control Protocol (TCP) in order to understand the impact of this protocol on file transmission times and network utilization. Analytical models have emerged as a way to reduce the time required for evaluation when compared with more traditional evaluations performed using event driven simulators such as *ns*. In addition, when designed carefully, analytical models help researchers make design decisions about novel TCP mechanisms.

A wide variety of techniques have been applied to the problem of TCP modeling with a fair amount of success. These techniques range over renewal theory [10, 17], Markov chains [11], and fluid models [1]. However, until the recent seminal work of papers using fix-point methods [3, 4, 5, 6, 7, 8, 14, 15, 18], no modeling technique was able to mimic both the structure of a TCP source and the interaction a source has with the network. The fix-point methods take the novel approach of separating the modeling of network behavior from the modeling of the behavior within a TCP source, and then allowing the two to tune each other via feedback. The fix-point framework allows general network topologies to be analyzed, and issues such as the interpretation of end-to-end loss rates in multiple bottleneck networks are addressed in [3, 7, 8, 15, 18].

In this paper, we generalize the framework based on a fixed point method introduced by Casetti and Meo in [4, 5] in order to allow us to model TCP-Vegas connections. The framework of Casetti and Meo, which uses a Markov chain to model the TCP source, has a few advantages over other fixed point methods: (1) it can model explicit details of TCP, making it possible to distinguish different flavors of TCP; (2) it allows modeling on-off traffic sources; (3) it gives the fraction of time that TCP spends in each state, from which we can evaluate the effectiveness of each mechanism of the protocol.

A majority of existing analytical models focus on TCP-Reno, the most widely deployed variant of TCP, and there has been little research on analytical models of TCP-Vegas, a more recently proposed variant. Analytical models of TCP-Vegas have been difficult to develop because TCP-Vegas uses observed delay to detect an incipient stage of congestion and try to adjust the sending rate before packets are lost. Prior studies on measurement and simulation of the performance of TCP-Vegas suggest that in many situations it is able to provide users higher throughput and lower loss rates than TCP-Reno. Hence, it is an important task to model the performance of TCP-Vegas in order to understand how this protocol performs in a network shared with other variants of TCP.

Because TCP-Vegas uses observed delay, as well as loss events, to adjust the congestion window size, while TCP-Reno uses only loss events, the extension made in this work to the framework of Casetti and Meo for modeling TCP-Reno is nontrivial. We need to make two major changes. First, analyzing the network model to determine the loss rate and the mean queueing delay is no longer enough; the full distribution of queueing delay must be obtained. Second, the Markov chain used to model a TCP-Reno source must be extended to use information from the delay distribution.

The model that results from these extensions represents a leap in our ability to model TCP-Vegas sources. In particular two major limitations of other TCP models are overcome in our model:

1. Our model is the first model of TCP-Vegas sources sending on-off traffic; all previous work on modeling TCP-Vegas only allowed bulk transfers [1, 2, 9, 12, 13, 16, 19]. Bulk transfer models are usually associated with FTP traffic, where

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a user sends a large amount of data without ever closing the connection. On-off traffic provides a more general model, one that actually includes bulk transfer as a special case (when the on periods are very long). On-off traffic applies both to FTP connections, the case of bulk transfer, and HTTP traffic, which inherently is made up of many short connections.

- Our model is the first model of TCP-Vegas that accurately predicts the operating point of the network in terms of throughput and loss rate for on-off traffic. In particular, all but one previous work assumes loss-free operation [1, 2, 9, 12, 13, 16]. Samios and Vernon [19] recently proposed the first analytical model to incorporate loss rate. Based on a renewal approach, they give a closed-form formula for the throughput achieved by a bulk transfer of a single TCP-Vegas sender as a function of the loss rate and the round trip time (RTT) of the network. However, since their closed form is dependent on knowledge of the achieved loss rate, they are not able to predict the full operating point of the network.

We validate our model under a single bottleneck network topology against *ns* simulations and existing analytical models. It is also possible to extend the network modeling to more general network topologies. We find that our model is accurate in a wide range of situations [20].

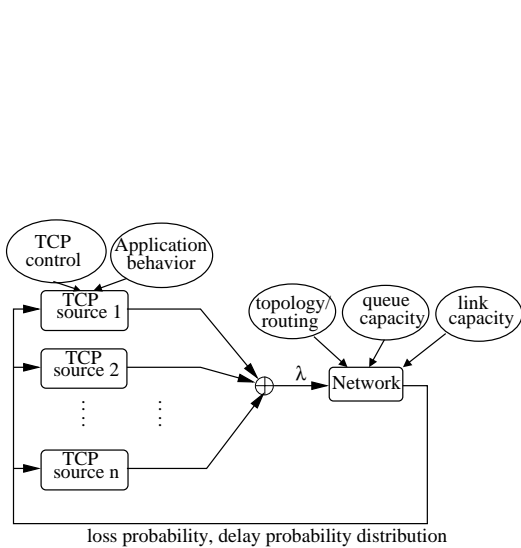


Figure 1: TCP analysis methodology

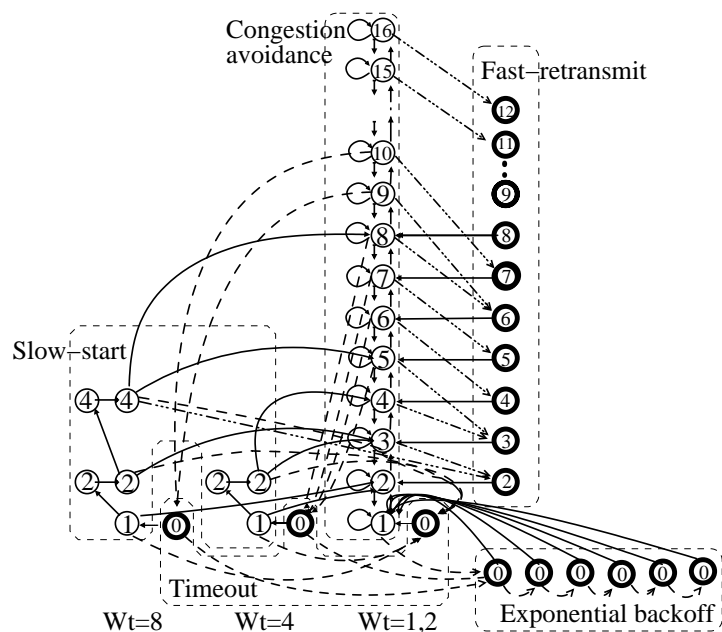


Figure 2: A continuous time Markov chain of TCP-Vegas for a maximum window size of 16, where states represents the window size and the phase of transmission: slow-start, congestion avoidance, fast-transmit, timeout, and exponential backoff.  $W_t$  denotes the slow-start threshold value.

## Model Overview

Our framework consists of two components: the TCP sources and the network (Figure 1). The throughput of each source is independently analyzed via a Markov chain, while the loss rate and the probability distribution of the delay of the network is analyzed separately using queueing models. For the Tahoe and Reno models in [4, 5], only the loss rate is analyzed, but we require the delay distribution for modeling TCP-Vegas.

Although the two components are analyzed separately, the parameters in the source models and the network model are tuned by iterative analysis with feedback from each other. That is, the source models are analyzed to obtain an aggregate

traffic, which is used as an arrival process of the network model. The network model, which for this work is an  $M/M/1/B$  queue, is analyzed to obtain the loss rate and delay probability distribution, which are used as the parameters in source models. Notice that a TCP source adjusts its sending rate (its congestion window size) based on the observation of events such as packet loss and delay of acknowledgments. The source models are analyzed again, and the procedure is continued until a stable solution is obtained. Given accurate source and network models, the fixed point of this iteration matches the operating point of the true network. We assume that the sources are independent. Therefore, when the sources are homogeneous, it suffices to analyze only one Markov chain; when the sources are heterogeneous, the number of Markov chains to be analyzed is the number of different types of sources.

Mimicking the structure of real-world network hosts, our model of an individual source is composed of two levels: an application level and a transport level. The application level alternates between a busy state and an idle state. While the application level is in a busy state, the transport level moves among TCP states, changing the window size and the phase of transmission (such as slow-start, congestion avoidance, fast-retransmit, and timeout).

A continuous time Markov chain for TCP-Vegas is shown in Figure 2. The states keep track of the congestion window size, as well as other information such as the slow-start threshold and whether we need to recover from loss or not. States with thin borders correspond to states where no loss event has occurred: slow-start states and congestion avoidance states. States with thick borders correspond to states where loss has occurred but has not yet been recovered from: fast retransmit states, timeout states, and exponential backoff states. The transition rate is determined by the loss rate and the distribution of the RTT, which are provided by the network model. By solving this Markov chain, we can determine the limiting probabilities of the congestion window size, which determines the throughput of the sender. See [20] for more details.

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