Credit-Based Flow Control for Multihop Wireless Networks and Stochastic Petri Nets Analysis

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Abstract—As the increase of data traffic is expected to grow faster than wireless capacity, dealing with congestion is unavoidable in the future. The capacity imbalance between wired and wireless links leaves a bottleneck on the wireless link while traffic keeps coming in through the bigger pipe. In current systems, the dynamics of a wireless channel and the variations in the higher priority real time traffic both make full link utilization impossible. Either the buffers are full, giving a full link but resulting in packet losses in the queues or the link is underutilized.

Closed loop credit based flow control can solve this problem by avoiding any buffer overflow completely and providing enough packets in the buffers to ramp up immediately when the channel has remaining capacity. In this paper a model using generalized stochastic Petri nets (SPN) is introduced. By structural analysis the boundedness of memory is proven while a Markov state analysis of the SPN provides numeric performance results. The studied scenario includes two wireless relayed hops, a wired backhaul and traffic class separation.

Keywords-stochastic Petri nets, wireless flow control, WFC, credit based, congestion, relay, multihop

I. INTRODUCTION

Wireless systems (cellular, ad-hoc and mesh) all have capacity limits which are significantly lower than what the fixed network backhaul can deliver. In the future the traffic generated and consumed by wireless terminals will increase faster than the capacity can grow due to new generations of radio technologies. Inevitably this will lead to congestion of the wireless links, at least during the busy hours. Realtime traffic will be saved from suffering by having a higher priority and limited load due to call admission control (CAC). So the remaining task needs to be solved: How is the non-real-time traffic protected from loss by buffer overflow, how is fairness among the congested flows guaranteed, how to avoid starvation of queues and how to best cooperate with the end-to-end TCP protocol for flow control and rate adaptation.

In cellular wireless systems we already have the notion of a flow [1] which is set up during connection establishment and whose QoS data is known to the MAC layer by means of crosslayer control plane signaling.

In this paper a flow control protocol is proposed for the traffic classes that are subject to congestion, especially the best effort (BE) class. This wireless flow control (WFC) is credit based [2], as this provides precision, stability and deterministic behavior [3] [4]. Known from the ATM ABR service, it is a new concept for wireless systems. However, it allows to safely drive the BE traffic into congestion and therefore cope with the future traffic load situation. Its structure is modeled using the Petri net approach, and performance results come from its stochastic analysis.

Petri nets [5] are an expressive and versatile tool for modeling and analysis of stochastic systems. They combine the flexibility of a Turing-complete automaton with the power of stochastic Markov chain analysis. Stochastic Petri nets (SPN) and generalized SPN (GSPN) [6] have become a useful tool for adept researches in computer science. Useful tools for GSPN analysis exist [7]. In recent years, GSPN have been used occasionally to model communications systems [8], [9], but a widespread use is not observed. Some examples are IEEE 802.16 [10] and IEEE 802.11 [11]. Links to higher layers exist in works for TCP models [12]. And there is still a high demand for recent technologies like the IMT-Advanced systems [13]. Especially ISO/OSI layer two (medium access control) can benefit from GSPN analysis methods. Most works study the maximum throughput only with a full buffer assumption. However, this is already in overload condition and either doesn't provide finite delays or drops lots of packets (finite buffer).

The paper is organized as follows: Section II introduces the basics of stochastic Petri nets. The following sections introduce the flow control protocol model, a scheduler and a channel model. A wireless system model using this flow control on two hops is treated in section VI. Performance results are presented in section VII before ending with the conclusion.

II. PETRI NETS

Petri nets (PN) is a graphical and mathematical tool suitable to model complex systems with a state. Systems can be be described and studied when they are concurrent, asynchronous, distributed, deterministic or stochastic. Subclasses of PN are finite state machines and marked graphs used for DSP algorithms [14]. Many aspects of flow charts

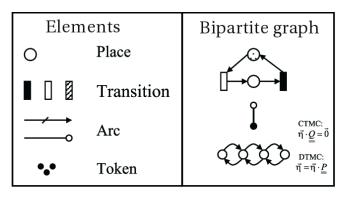


Figure 1: Elements of stochastic Petri nets.

and description languages can be modeled with PN. There is plenty of literature on the underlying graph theory, liveness analysis, reachability set and other properties [5]. A PN is defined as a directed, weighted, bipartite graph having two sets of nodes called *places* (P_i) and *transitions* (T_i) . Places are drawn as circles, transitions as boxes, see Figure 1. Arcs are directed and called input arcs if they connect from a place to a transition, output arc otherwise. A convenience notation is the notion of a disabling arc, drawn with a small circle at the end of the arc. Input arcs connect certain P_i to T_j with weight w_{ij} , output arcs connect T_j to P_i with multiplicity v_{ij} . Together they form the incidence matrix $\mathbf{D} = [v_{ij} - w_{ij}]$ When a PN consists of M transitions and N places, **D** is a $M \times N$ matrix (m rows, n columns). Places can contain an integer number of tokens (dots), all of which constitute the state called *marking* \vec{m} . The initial marking \vec{m}_0 is the start state. The notation $\#P_i = m_i$ means the current number of tokens in place P_i . An atomic action is the *firing* of a transition T_i which changes the marking to

$$\vec{m}_k = \vec{m}_{k-1} + \vec{t}_{k-1} \cdot \mathbf{D} \tag{1}$$

with the firing vector \vec{t}_{k-1} which is all zero except a single '1' at the j.th index. The firing rule is basically: A transition can be fired if all places connected to input arcs contain enough (w_{ij}) tokens and all places connected via disabling arcs are empty. For further details of firing rules, liveness and reachability see [5].

A. Stochastic Petri Nets

Stochastic PN (SPN) extend the paradigm to model time, essentially by assigning each transition a firing rate (in the continuous time case (CT)) or a firing probability (in the discrete time case (DT)). Here we will focus on the CT case only. Then the firing rates are given by $\vec{\Lambda} = \lambda_1, ..., \lambda_m$ which can be marking-dependent. λ_j is the inverse of the average firing time t_j . Firing times are exponentially distributed, therefore memoryless, and the resulting reachability graph (RG) forms a Markov chain (MC) [6]. All arcs in the RG are annotated with the λ_j of the transition T_j responsible

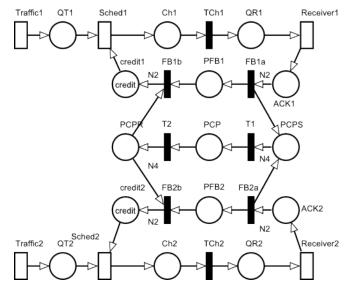


Figure 2: credit-based flow control of one link with two connections

for the state change. The matrix \mathbf{R} of all arc rates can then be used to calculate the steady-state solution of the MC and from that all other performance metrics can be obtained.

Generalized SPN (GSPN) combine both immediate (thick bar) and timed transitions (empty box), which also reduce to MC after condensing all tangible states, but offer a much higher modeling power. Priorities and weights can easily be annotated to immediate transition to model the outcome likelihood of deterministic and random decisions. Deterministic and stochastic SPN (DSPN) also allow one enabled transition with deterministic (fixed) firing time, denoted with a filled box.

In this paper we will not use colored PNs (CPN), as each token color extends the state space by huge amounts and an analytic/numeric treatment is often impossible. There is decent tool support for SPN and the MC is automatically determined and solved [7]. When the state space for numeric matrix solvers exceeds the current limit of around 10^6 states on a standard PC, simulation can be used, based on an exhaustive repetition of the *token game*.

Queueing models and networks are a subset of the modeling power of SPN, and results for queue length and waiting times can easily be derived [15], that is why SPNs are becoming increasingly popular for modeling communication networks [8]. SPN can be used as well for modeling packet level and flow level, where a constant fluid of data in bits/sis assumed while being in a state.

III. CREDIT-BASED FLOW CONTROL

In this paper a credit-based flow control protocol is proposed as known from the ATM domain [2] [3]. For the basic operation each link is considered to be a closed loop between adjacent nodes (BS, RN or UT). Peers are called S for

sender and **R** for receiver here. For each loop, the receiving node **R** has an available total buffer of size l_l (*limit*_{link}). This buffer is partitioned into individual memory l_v (limit_v) for each flow or connection v out of C BE connections through that node. This partitioning is only logical and can be overlapping, i.e. buffer sharing is possible with $\sum l_v > l_l$. Both buffers are protected by separate flow control loops, one per flow and one total. The sender S knows about each limit l_v (l_l) and installs this as the initial credit c_v (c_l) . Packets (or e.g. LTE-A resource elements) may now be sent arbitrarily (after the scheduler has served all higher-priority packets) as long as the current credits c_v (c_l) are positive (non-zero). For each sent packet a counter of transmitted packets t_v is incremented (from zero initially). **R** increments a counter r_v of packets that are received and forwarded to the next hop. In regular intervals (each $N2_{VC}$ values) this counter is sent back to the source S which updates its state for all included connections v by replacing the previous values of forwarded packets, f_v , with the new contents f_v^{new} of each record. The current credit for connection v is given by eq. 2 and similar for the link l in eq. 3):

$$credit_v = c_v = l_v - t_v + f_v \ge 0 \tag{2}$$

$$credit_l = c_l = l_l - t_l + f_l \ge 0 \tag{3}$$

By using absolute counters instead of tokens in the real implementation, a loss of credit messages on the wireless link is not a serious problem, as the next counter update is enough to continue.

In [3] a Petri net model was introduced which models all necessary details of one flow control loop. Here it is applied for wireless links with fading channels and in a multihop environment. Figure 2 shows the SPN model of one flow controlled link for two flows. Additional hops seamlessly integrate to the right of this model, as shown in section VI.

The main components in the forward direction are the packet queue memory QT_v (the instantaneous contents denoted by $m_v := \#QT_v$ packets) and the k_v packets in flight on the link in place Ch_v ($k_v := \#Ch_v$). A packet of flow v is only allowed to be sent by the scheduler (see section IV) if there are tokens in the credit place $credit_v$ ($c_v := \#credit_v$). In backward (feedback) direction f_v forwarded tokens queue in place ACK_v ($f_v := \#ACK_v$) until a record in PCPS is complete (pr := #PCPS). N4 of such records build a credit update message packet in PCP (cp = #PCP), which is again decomposed in the upstream node into records in PCPR (rr := #PCPR). PFB_v exists for distinguishing the flows and holds the number of credit update records for this flow v. It holds $r_v := \#PFB_v$ tokens.

There is one per-flow flow control loop (here C = 2 connections), using the feedback quantization constants N2 discussed in [3] to decrease the feedback frequency. Each link is can also be controlled by a *link* flow control loop,

quantized with $N2_{link}$ (N2L). This is omitted here but described in [4]. The packing into a feedback message consisting of N4 records reduces the flow control bandwidth overhead, which is discussed into details in [3].

Let this system be implemented in each hop and the connection between them be made by arcs which have buffering places to hold tokens which are currently on the fly. The consistency of each loop implies [16] that the weighted sums of tokens (Eq. 4-5) on them are constant:

$$\forall v : k_v + m_v + f_v + N2 \cdot r_v + c_v = l_v \tag{4}$$

$$\sum \left(k_v + m_v + f_v + c_v\right) + \tag{5}$$

$$N2 \cdot \left(\sum cr_v + rr + pr + N4 \cdot cp\right) = \sum l_v$$

Therefore there cannot be more tokens in any place of the loop than there are initial credits $(c_v = l_v)$ in the credit places. Thus the buffer places m_v can also at most contain l_v packets each and the whole queue (buffer memory) usage l_{max} is bounded by both the sum of all relevant connection limits and the link limit.

$$l_{max} := \max(\sum m_v) = \min(\sum_{\forall v} l_v, l_l)$$
(6)

The boundedness of buffer contents implies that no packet loss can occur due to buffer overflow. Packet losses on the link are accounted for by an underlying hop-by-hop ARQ mechanism. Actually ARQ and WFC can use and share the same sequence numbers. For the discussion of dimensioning and quantization effects see [3].

The complexity of the algorithm is not very high. It only requires three counters per flow for each intermediate node, and two of them are already present due to ARQ sequence numbers. The number of flows on wireless links is usually not high. Each UT might have 5 flows in parallel, which is the count on the last link. A link BS-RN might carry 500 flows in a cellular network, so the memory consumption is a few kilobytes which is small compared to the packet buffer itself. Each node should have a reasonable packet buffer of at least several times the bandwidth-delay product of the hop before and after.

The option to reduce l_{max} below the sum of the individual credits ("buffer sharing") allows a tradeoff between memory and peer blocking probability [3].

The timing behavior is determined by the type of the transitions. Except the "schedulers" all of them are immediate and do not contribute to the latency. The sources of latency are the link propagation delay and the waiting for multipleweighted input arcs to gather enough tokens for the adjacent transition to fire [3].

Regarding the transition *Sched*1, we see the necessity separate the traffic into real-time (RT) and best effort (BE) by means of a static priority scheduler [17], which leads to the next section. Only BE traffic needs to be treated by flow

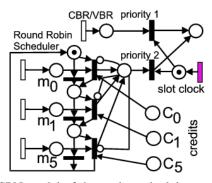


Figure 3: SPN model of the packet scheduler separating RT (CBR/VBR) and NRT (BE) traffic by static priority. Fairness in BE is established by RR subscheduling and credits tokens enable flow control from downstream

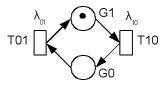


Figure 4: SPN of the Gilbert-Elliot channel. The state #G1 can be used in state-dependent formulations of SPN properties.

control, since the load of RT is below congestion by the help of CAC.

IV. HIERARCHICAL SCHEDULING

Figure 3 shows the SPN model of the packet scheduler assumed in each node. Note that packet scheduling is different and separable from resource scheduling on the wireless link [18]. Most important is the separation of RT and NRT (BE) traffic by a static priority scheduler (having O(1) complexity). Within the priority class there are subschedulers per class that treat all flows within the class according to their QoS requirements [17]. The RT traffic and scheduler is not treated here further, but there exist simple [19] and heavy solutions [20], FCFS being still the easiest with O(1) complexity and a good delay-fair behavior among flows. Within the BE class fairness is usually most important, especially in the congested region (full buffers). When no further weights are required, a simple round-robin (RR) scheduler does the job pretty well in O(1) complexity. In Figure 3 also the interfaces to the flow control are shown by the credit places.

V. CHANNEL MODELS FOR SPN ANALYSIS

Wireless channels possess a number of properties and are influenced by many effects. Usually they are decomposed into antenna gain, pathloss, slow fading, fast fading, shadowing, multipath propagation effects, The simplest model suitable in a packet level SPN approach is the two-state

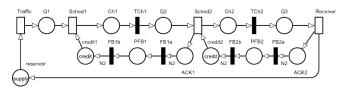


Figure 5: SPN of the wireless system with two hops using credit-based flow control

Gilbert-Elliot (GE) channel, also seen as on-off MC. Its SPN equivalent is shown in Figure 4 using places G_1 and G_0 , the token position in which marks the state as 'good' or 'bad'. The transition firing rates λ_{10} (into fading) and λ_{01} (recovery) model the state changes from 'good' to 'bad' and opposite. They are translated from the common parameters P_{error} and T_{period} by Eq. 7-8.

$$\lambda_{10} = (T_{period} \cdot (1 - P_{error}))^{-1} \tag{7}$$

$$\lambda_{01} = (T_{period} \cdot P_{error})^{-1} \tag{8}$$

The state itself can the be used in another part of the SPN to influence a scheduler, e.g. switch between transmitting or deleting packets which are currently transmitted on the channel or by switching the available short-term capacity.

Other SPN models like the finite-state Markov channel (FSMC) [21] can be found in [22].

VI. WIRELESS MULTIHOP LINKS USING FLOW CONTROL

In this section a scenario is analyzed which consists of two wireless hops, i.e. a relayed link. Figure 5 shows the basic SPN for one flow. Here we assume a system which can have wired links towards the source and destination, where the wired links are assumed to have a capacity much larger than the wireless links.

Sched1 is an exponential transition which models the fluctuations in remaining capacity due to channel variations and higher priority RT traffic that is not modeled here explicitly. The traffic (packet) interarrival time is exponential, putting packets (tokens) into queue Q1 before the first hop (imagine Q1 in the BS). After admitted to the channel by the scheduler, packets traverse the channel place Ch1 and are queued in Q2 on the intermediate hop (relay, RN). The scheduler Sched2 there controls access to the second hop. If admitted, the packet flows through Ch2 to be queued in Q3 shortly before the next link (wired) or application can make use of it.

The flow control works by sending feedback (backpressure) messages back to the sender. Only after a packed is further forwarded onto the next link or the application has consumed it, an ACK token is created and put into place ACK1 (or ACK2 on the second hop). As mentioned in section III, there is a quantization operation with N2 in order to reduce the feedback overhead. But basically all ACK messages assemble again in the *credit* place after some time. The scheduler transition may only fire if there

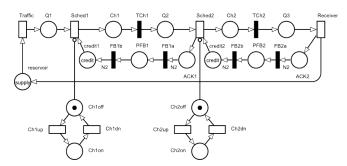


Figure 6: SPN of the wireless system with two hops using credit-based flow control and a fading channel on each hop

is at least one credit token left. By this way a packet loss by buffer overflow (due to a slower link downstream) is impossible.

The supply place P_{sup} is necessary to bound the state space. As a rule-of-thumb, the supply s should be ten times larger than the expected queue occupancy in Q.

Figure 6 shows the system model including two independent wireless fading channels. The channels are modeled as two separate subnets of type as in Figure 4, where the marking #Ch1off controls the disabling arc to the scheduler place, effectively stopping all transmissions. Both channels are independently fading.

For measurements the following equation derived from the *p*-invariant of the main loop is useful in order to simply determine the total delay:

$$#Q1 + #Q2 + #Q3 + #Ch1 + #Ch2 + #P_{sup} = s \Rightarrow (9)$$

$$E[\#Q1 + \#Q2 + \#Q3 + \xi] = s - E[\#P_{supply}] \quad (10)$$

VII. NUMERIC PERFORMANCE RESULTS

The system model has been analyzed by Markov chain analysis of the underlying MC (defined by the RG) of the SPN. The parameters used for this numeric example are: credit = 10, N2 = 5; the channel coherence time is specified by $T_{Ch1up} = T_{Ch2up} = 2 \cdot T$, $T_{Ch1dn} = T_{Ch2dn} = 2 \cdot T$; The channel capacity is defined by $T_{Sched1} = T_{Sched2} = T$. The traffic is varied between 'no traffic' ($\rho = 0$) and 'congestion' ($\rho = 1$) by $T_{Traffic} = T/\rho$. T itself does not play a role since all transition rates are defined relative to each other, so $T = 1\mu s$ is a good choice.

Figure 7 shows results for the study of different traffic arrival rates to the systems, normalized to the capacity of the system. The results reveal the queueing nature of random traffic with a server given by the wireless links. At a certain maximum load, the queue Q1 grows asymptotically to infinity, a condition usually called unstable overload. However, the flow control cares for the fact, that there is no overflow in the wireless queues Q2 and Q3.

Because all the MC state probabilities are known after analysis, it is easy to also obtain the probability density

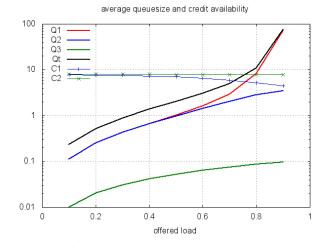


Figure 7: Performance results of the two hop flow controlled scenario by SPN analysis.

function (PDF) or CDF of the number of tokens in any place and therefore PDF and CDF of packet queue occupancy and packet delays.

Figure 8 shows results of the multihop model of Figure 6 with fading channels on both hops. Due to the parameter settings, both wireless channels can only carry 50% of the capacity of the example before. Therefore the offered load has been normalized to account for this fact ($\mu_{max} = 1/(2T)$).

The results show that only the first queue Q1 grows with increased traffic, because it is not within a flow control loop in this example. All other queues on the wireless nodes are bounded by construction (here: $\#Q2 \le l_v = 10$). The total number of packets in flight Qt is obviously dominated by Q1. The number of tokens in the credit places #credit(C1,C2 in the graph) do not deplete so soon and even have 5 on average in the congested situation. Even if the second link is completely blocked, the scheduler *Sched*2 would stop and soon the credits would drop to zero to also stop *Sched*1.

Figure 9 shows the complementary cumulative distribution functions (CCDF) of the token distribution in the queues Q1, Q2 and the credit place C1. More than the average results in Figure 6, this shows that the buffer memory in Q2 does never exceed the upper bound given by the initial credit.

At this point it becomes clear that an ideal wireless network should also extend the flow control loops to the links (and hops) delivering the packets to the first wireless node (base station). This would totally avoid any buffer overflow in any node of the network and provide a reliable communication for BE traffic in a congested network. This will become an important aspect in the future, when traffic grows beyond capacity.

As this credit-based flow control is rather new in this

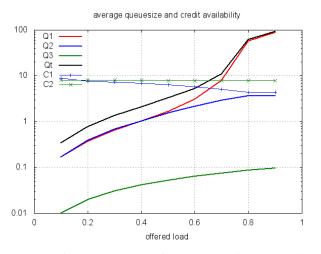


Figure 8: Performance results of the two hop flow controlled scenario with fading channels by SPN analysis.

context, it cannot be expected that this concept will be adopted also for the wired links so soon. The classical endto-end TCP control is not very suitable for the wireless domain [23]. However, a solution exists by extending the BS (eNB in LTE-A) functionality to ISO/OSI layer 4, where TCP ACK messages can be created by the BS (instead of the end point in the UT). These TCP ACK messages contain flags that can be used to throttle the incoming traffic flow or even exert a finer grain rate control.

VIII. CONCLUSION

This paper addresses the problem of congested traffic of the best effort category in future wireless networks. The proposed solution consists of a traffic separation of real-time (RT) and non-real-time (NRT) or best-effort (BE) traffic, so that RT traffic can be treated with standard QoS-aware scheduling algorithms and limits by CAC, and BE traffic is allowed to use up the available capacity completely. This can be achieved by fair schedulers, e.g. round robin, and a flow control protocol that avoids buffer overflows completely. A modeling approach using stochastic Petri nets (SPN) was proposed for studying the performance of the protocol in wireless systems. The main contribution is the proposal to use this protocol know from the ATM domain also in wireless networks and its stochastic Petri net (SPN) model. More complex system components can be added easily in the SPN paradigm. It is convenient to plug in this flow control model into a bigger system model and concentrate on other interesting tasks on the MAC layer like resource allocation, scheduling, and the study of packet delays. Also it is straightforward to extend the two hops analyzed here to multiple hops and even towards mesh networks, as long as a flow establishment [1] is used which exchanges the WFC parameters in the initial handshake protocol. Future work



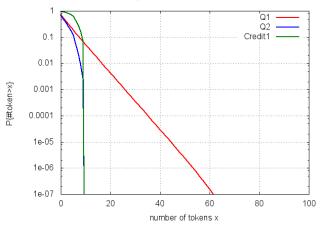


Figure 9: complementary CDF of queue contents and credit token count for scenario(Figure 6) at a load of $\rho = 0.6$.

will make more use of this and can provide analytic/numeric results where otherwise only simulation would have been used. Also, the sustainability aspect of wireless communications requires a rethinking [24] towards coping with the limitations instead of overprovisioning and flatrate tariffs.

REFERENCES

- A. Otyakmaz, D. Bültmann, R. Schoenen, and I. Durmaz, "On Flow Management for Future Multi-Hop Mobile Radio Networks," in *IEEE WiCom*, Beijing, China, Sep 2009.
- [2] H. Kung and S. Wang, "Zero Queueing Flow Control and Applications," in *Proceedings of the IEEE INFOCOM*, 1998.
- [3] R. Schoenen, G. Post, and A. Müller, "Analysis and Dimensioning of Credit-Based Flow Control for the ABR Service in ATM Networks," in *Proceedings of the IEEE GLOBECOM*, 1998, vol.4 p.2399.
- [4] R. Schoenen and A. Dahlhoff, "Closed Loop Credit-Based Flow Control with Internal Backpressure in Input and Output Queued Switches," in *Proc. of ATM Workshop*'99, 2000.
- [5] T. Murata, "Petri Nets: Properties, Analysis and Applications," *Proceedings of the IEEE*, vol. 77, no. 4, pp. 541–581, April 1989.
- [6] M. Marsan, Modelling with Generalized Stochastic Petri Nets. Wiley, 1996, iSBN 0-471-93059-8.
- [7] R. German, "A toolkit for evaluating non-Markovian stochastic Petri nets," *Performance Evaluation*, vol. 24, pp. 69–87, 1995.
- [8] J. Billington et al., Application of Petri Nets to Communication Networks. Springer, 1999, ISBN 3-540-65870-X.
- [9] L. Lei, C. Lin, J. Cai, and X. Shen, "Performance Analysis of Wireless Opportunistic Schedulers using Stochastic Petri Nets," *IEEE Transactions on Wireless Communications*, vol. 8, no. 4, April 2009.

- [10] S. Geetha and R. Jayaparvathy, "Modeling and Analysis of bandwidth Allocation in IEEE 802.16 MAC: A Stochastic Reward net Approach," *Int. J. Communications, Network ans System Sciences*, vol. 3, no. 7, pp. 631–637, July 2010.
- [11] R. Jayaparvathy, S. Anand, S. Dharmaraja, and S. Srikanth, "Performance Analysis of IEEE 802.11 DCF with Stochastic Reward Nets," in *International Journal of Communication Systems*, vol. 20, no. 3, 2007.
- [12] R. Gaeta, M. Gribaudo, D. Manini, and M. Sereno, "On the use of Petri nets for the computation of completion time distribution for short TCP transfers," *Proceedings of the 24th international conference on Applications and theory of Petri nets*, vol. LNCS, pp. 181–200, 2003.
- [13] J. Monserrat, P. Sroka, G. Auer, J. Cabrejas, D. Martin, A. Mihovska, R. Rossi, A. Saul, and R. Schoenen, "Advanced radio resource management for IMT-Advanced in WINNER+ (II)," in *Proc. ICT-MobileSummit 2010*, Florence, Italy, Jun 2010.
- [14] V. Živojnović, R. Schoenen, and H. Meyr, "On Retiming of Multirate DSP Algorithms," in *Proceedings of the IEEE International Conference on Acoustics, Speech and Signal Processing*, vol. VI, Atlanta, May 1996.
- [15] L. Kleinrock, *Queueing Systems, Vol. I: Theory.* New York: John Wiley & Sons, 1975.
- [16] R. Schoenen, V. Živojnović, and H. Meyr, "An Upper Bound of the Throughput of Multirate Multiprocessor Schedules," in *Proceedings of the IEEE International Conference on Acoustics, Speech and Signal Processing*, München, Apr. 1997.
- [17] R. Schoenen and A. Otyakmaz, "QoS and Flow Management for Future Multi-Hop Mobile Radio Networks," in *IEEE VTC Fall*, Ottawa, Canada, Sep 2010.
- [18] R. Schoenen and F. Qin, "A Closed Loop Control Model separating Resource and Packet Scheduling in Multihop Cellular Networks," in *Proc. of the European Wireless*, Lucca, Italy, Apr 2010.
- [19] J. Peha and F. Tobagi, "Evaluating Scheduling Algorithms for Traffic with Heterogeneous Performance Objectives," in *Proceedings of the IEEE GLOBECOM*, 1990.
- [20] D. Niyato and E. Hossain, "Service differentiation in broadband wireless access networks with scheduling and connection admission control: a unified analysis," *Wireless Communications, IEEE Transactions on*, vol. 6, no. 1, pp. 293–301, 2007.
- [21] H. Wang and N. Moayeri, "Finite-State Markov Channel A Useful Model for Radio Communication Channels," *IEEE Transactions on Vehicular Technology*, vol. 44, no. 1, pp. 163–171, Feb 1995.
- [22] R. Schoenen, M. Salem, A. Sediq, and H. Yanikomeroglu, "Multihop Wireless Channel Models suitable for Stochastic Petri Nets and Markov State Analysis," in *Proceedings of IEEE VTC Spring*, Budapest, May 2011.

- [23] M. Malkowski and S. Heier, "Interaction between UMTS MAC Scheduling and TCP Flow Control Mechanisms," in *Proceedings of 2003 International Conference on Communication Technology*, Beijing, China, Apr 2003, pp. 1373–1376.
- [24] R. Schoenen, "On increasing the spectral efficiency more than 100% by user-in-the-control-loop," in *Proc. of the 16th Asia-Pacific Conference on Communications (APCC)*, Auckland, October 2010.