

FUZZY-LOGIC ADAPTIVE QUEUING FOR A HEURISTIC TCP PERFORMANCE IN MOBILE WIRELESS NETWORKS

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ABSTRACT

In this paper, we propose a new Fuzzy-Logic Adaptive Queuing controller (FLAQ) based on a classical Random Early Detection (RED) algorithm in wireless cellular network. The controller predicts dynamically the packet dropping rate and the corresponding average queue length. It relies on the average queue length at the base station router and the packet loss rate caused by the channel variations in mobile environment; assuming there is no buffer overflow due to the congestion. Using this model, a heuristic TCP performance can be estimated over a time-varying channel under different conditions of user's mobility. The results show a significant improvement in TCP throughput performance when the user's mobility is below 5 m/s; and becomes constant (i.e., close to i.i.d) beyond this speed especially at 5% of predefined packet error rate.

KEYWORDS

Fuzzy Logic, Queue Management, TCP Throughput, Time-Varying Channel, Wireless & Mobile Network

1. INTRODUCTION

Computer networks like internet necessitate network resources (i.e. bandwidth, buffer spaces, etc.) in order to accommodate the arriving packets at router buffers [1][2]. When the arriving packets cannot be accommodated due to lack of network resources, this indicates occurring congestion at router buffers of networks. More specifically, a poor network performance due to congestion can be offered in terms of high dropping and queuing delay for packets, low throughput and not maintaining the average queue length which may not prevent the router buffers from building up, then dropping packets. Moreover, congestion may also trigger an unbalanced share among the network sources [3].

However, congestion control mechanism implemented through "Queue Management" algorithms, is still the key factor to solve this problem keeping TCP/IP networks efficient and reliable from the user's viewpoint. In particular, it is well-known that the rapidly emerging mobile data networks such as 3G, HSPA, and LTE and the integrated WAN, LAN and 3G cellular systems has created new and crucial challenges for the development of Internet applications [4][5]. Unlike the wired networks, the mobile networks exhibit some challenges basically related to two factors; (i) a highly variable bandwidth and (ii) base stations are often equipped with large buffers in order to absorb bandwidth fluctuations and consequently prevent the unnecessary packet loss. Therefore, TCP/IP needs to meet the challenges introduced by the wireless portion of the network in order to make it valid with the wired network protocols for providing efficient services by web accesses, e-mails, bulk data transfers, etc. To solve and

mitigate these challenges, many works have investigated methods to estimate the key wireless link parameters, for example, link bandwidth estimation [6], link buffer size estimation such as max-min, loss-pair and sum-of delays [7]-[9], and queue length estimation [10]. For this reason, the utilization of fuzzy logic (FL) has shown to be useful in designing new active queue management (AQM) methods [11]-[16] that can be used to alleviate congestions in wired and wireless networks. For example, Mallapur *et al.* [15] propose buffer manager located at the base station using a fuzzy controller for packet dropping in wireless cellular networks. The controller uses three fuzzy parameters, namely application priority, queue length and packet size.

The traditional technique for managing queue lengths is to set a threshold (in terms of packets) for each queue, accepts packets for the queue until the threshold is reached, then reject (drop) subsequent incoming packets until the queue decreases below the value of threshold. This technique is known as "tail drop". This threshold most of the time is the maximum capacity of buffer, then tail drop signals congestion only when the queue has become full. The solution to the full queues problem is for routers to drop packets before a queue becomes full, so that end nodes can respond to congestion before buffers overflow. By dropping packets before buffers overflow, AQM allows routers to control packet drops. Many researches have been devoted on the queue size management algorithms such as Random Early Detection (RED) [16] and Weighted RED [17], PID controller [18], and NLRED [3] to reduce the likelihood of global synchronization, as well as keeping queue sizes down in the face of heavy load and bursty traffic. Other related AQM schemes such as GRED, WRED, and ARED are summarized in [3]. The Random Early Detection (RED) mechanism can solve the Drop Tail's (DT) deficiencies well. It uses randomization to ensure that all connections encounter the same loss rate. By dropping packets before the router's buffers are completely exhausted, the RED mechanism try to prevent congestion. In [19], another approach known as Random Exponential Marking (REM) was also developed to measure the average queue size instead of congestion measure. It was able to achieve high link utilization, negligible packet loss and short queuing delay in a simple scalar manner.

Modeling the dynamics of TCP and AQM is another important research area in heterogeneous networks. Accordingly, it is found that the main degradation of Quality of Service (QoS) in the wired networks is the congestion due to buffer overflow at the routers. Meanwhile, the QoS degradation in wireless links occurs not only caused by the congestion, but also due to possible high bit errors of the radio channels. Although TCP is basically employed for end-to-end wired Internet, it can not achieve sufficient throughput in the wireless environment where packet are lost due to transmission errors. TCP assumes that losses of packets are derived from a buffer overflow at a BS router. It invokes the congestion control to reduce traffic unnecessarily when the packet loss occurs due to transmission errors on a wireless link. Thus one approach to mitigate this problem and distinguish packet loss due to congestion and others packet errors due to channel variations, is to employ TCP proxy called Performance Enhancing Proxy (PEP) [20].

In particular, the approaches devoted on improving TCP throughput over wireless cellular networks can be classified as follows. (i) Splitting connection at the wired-wireless border (e.g., BS). However, the major problem of such Indirect-TCP (e.g. Snoop protocol) and M-TCP is a hand-off latency due to mobility of MH [5][21]. Explicit loss notification (ELN) is another more precise approach with capability of distinguishing the packet loss due to packet loss due to congestion and the one from transmission errors. (ii) Developing data-link layer protocols, e.g. Automatic Repeat request (ARQ) and/or adaptive FEC [21] and (iii) Designing Cross-layer approaches [22] where some approaches focuses on RED AQM over wireless channels [23][24]. The channel variations are mainly caused due to the highly or correlated bit errors and this in consequence can cause to packet dropping and eventually reduce the network throughput. In this case, a TCP proxy at the router queue of the base station can be employed to distinguish

the packet error (dropping) due to the variations of wireless channel and the packet loss due to congestion of buffer overflow.

In this paper, therefore we develop a new Fuzzy-Logic Adaptive Queuing controller (FLAQ)-RED at the base station in order to tune the average queue length and the wireless packet error. The proposed model inclines to maintain small buffer size space over a time-varying channel which may exhibit significant degradation in the network bandwidth estimation. Then, we predict the lowest packet dropping received by the mobile receiver over Rayleigh fading environment when there is no congestion effect at the Base station buffer. By using this model, a heuristic TCP performance can be evaluated over a time-varying channel under different conditions of user's mobility.

2. NETWORK PRELIMINARIES

2.1 Classical RED

The RED scheme operates as follows. RED computes the average queue size Q_{ave} and a drop probability P_d based on the instantaneous queue size. In addition, RED maintains two thresholds of queue size Q_{min} and Q_{max} . **Figure 1** explains the basic operation of RED as follows [17]:

- If $Q_{avg} < Q_{min}$, no packet drop
- If $Q_{min} \leq Q_{avg} \leq Q_{max}$, drops each packet with the probability P_d .
- If $Q_{avg} > Q_{max}$, drop every arriving packet.

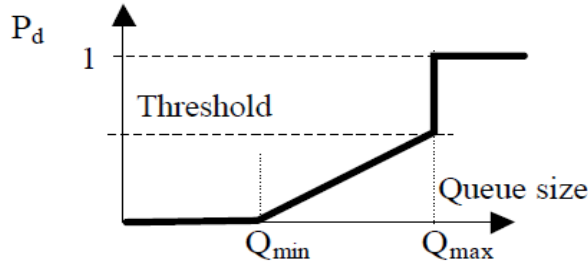


Figure 1. A Classical RED Algorithm

In the RED mechanism, the average queue size, Q_{ave} , is calculated and compared with two thresholds, a minimum threshold (min_{th}) and a maximum threshold (max_{th}). When Q_{avg} is less than the min_{th} , no packets are marked. When it is greater than the Q_{max} every arriving packet is marked.

When Q_{ave} is between the min_{th} and the max_{th} , each arriving packet is dropped with probability P_a , where P_a is a function of Q_{ave} . The packet-marking probability P_b is calculated as below [17]:

$$P_b = \frac{\max Q_{avg} - \max_{th}}{(\max_{th} - \min_{th})} \quad (1)$$

The final packet-dropping probability P_a increases slowly as the count increases since the last marked packet:

$$P_a = \frac{P_b}{1 - count \times P_b} \quad (2)$$

2.2. Network Model

We consider a common scenario of unicast network topology model of a wired-to-wireless network to investigate the TCP performance over wireless link as shown in **Figure 2**. The network model consists of the fixed host (FH) which is defined over the wired link, while the wireless last hop consists of a single pair of base station (BS) and one mobile host (MH) in wireless cellular network [21][24]. Within this model, the wireless channel feedback signal in terms of TCP ACK is used to return back the estimation of channel variations in terms of the wireless packet error rate (PER) over wireless channel.

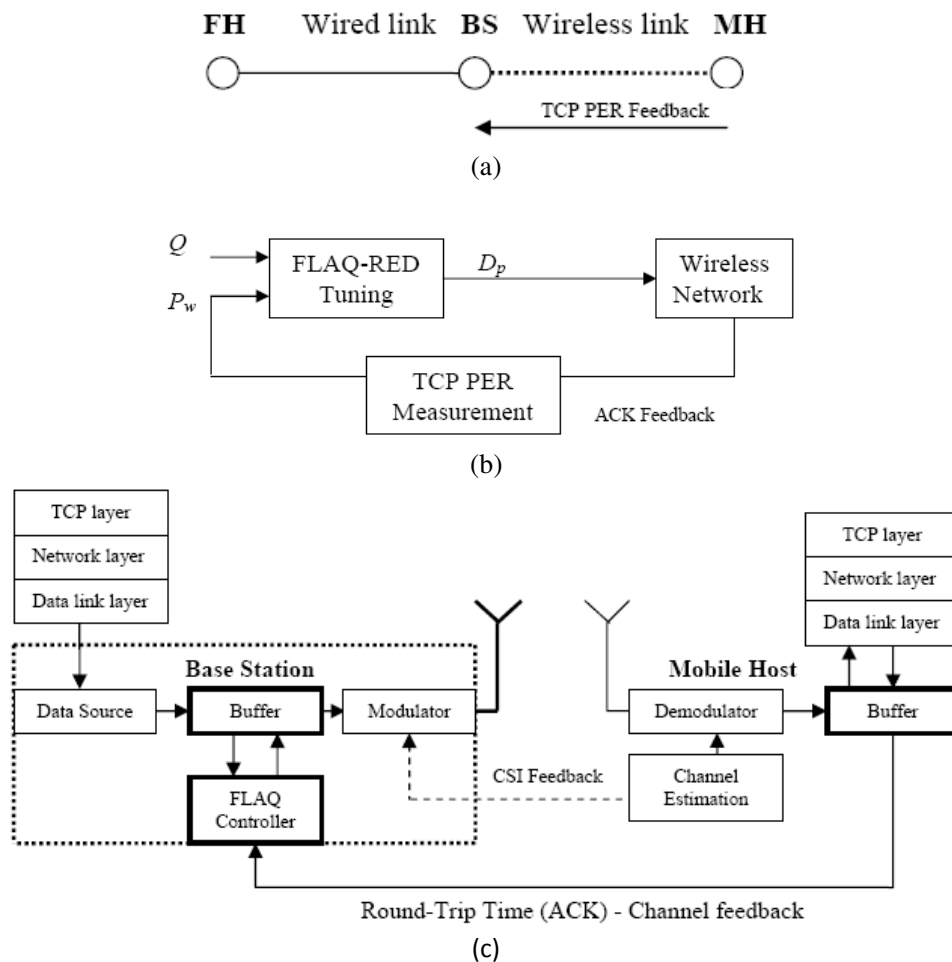


Figure 2. The proposed approach (a) Unicast Network Model (b) The process of proposed FLAQ-RED and (c) System model based on FLAQ-RED

2.3. System Scenario

We consider the key preliminaries required to predict the packet dropping rate based on fuzzy logic adaptive queuing scheme over Rayleigh fading channel as follows:

- A *single* TCP flow traffic is considered for only one mobile receiver.
- The network is assumed to be stable without heavy or bursty TCP traffic.
- The TCP packet error rate (PER), (i.e., P_w), is caused by the variations of wireless channel when only highly bit errors occurs during traffic transmission. Assuming there is no congestion at the router buffer of BS.
- P_w is measured by the channel estimator at mobile receiver and returned back via the ACK feedback of the round trip of TCP to indicate the sender about the channel bit errors, so we consider P_w changes from 5% to 30%.
- The TCP proxy at the router queue of the base station is required *if and only if* multiple TCP flows are present [20][21][24]. This proxy is mainly used to distinguish the packet error (dropping) due to the variations of wireless channel and the packet loss due to congestion of buffer overflow. In our system scenario, there is *no packet dropping* due to any buffer overflow. So, the link is under- utilization bandwidth.
- At the base station (BS), we consider the following assumptions:
 - Let the buffer size = 200 packets
 - The router queue with $Q_{min}=50$ [packet] ,and $Q_{max}=150$ [packet]
 - If the average queue length less than 50 no packets are dropped
 - If the average queue length more than 150 all the arriving packets are dropped
 - If the average queue length between Q_{min} and Q_{max} then the dropped packets controlled by the fuzzy logic controller depending on to inputs ($AVQL$ and P_w)
 - Packet loss rate at base station is ignored (no packet loss due to congestion)

3. PROPOSED FLAQ-RED CONTROLLER

Fuzzy logic controllers FLC, like expert systems, can be used to model human experiences and human decision-making behaviours [25]. In FLC the input-output relationship is expressed by using a set of linguistic rules for relational expressions. An FLC basically consists of four conceptual components including a fuzzifier, a defuzzifier , an inference engine and a rule base.

Figure 3 shows the Fuzzy Logic Controller architecture. It is evident that as in many fuzzy control applications, the input data are usually crisp, so a fuzzification is necessary to convert the input crisp data into a suitable set of linguistic value that is needed in inference engine. Singleton fuzzifier maps the crisp input to a singleton fuzzy set.

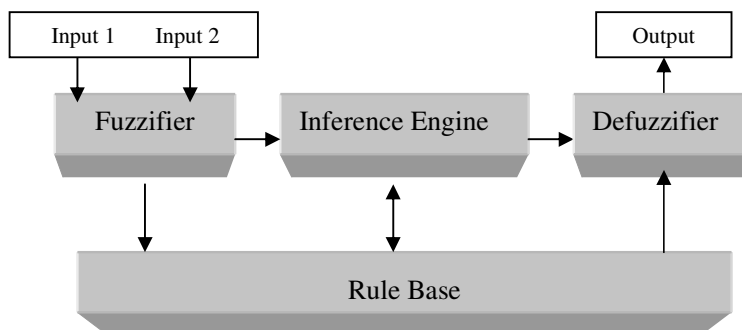


Figure 3.Fuzzy Logic Controller Architecture

In the rule-base of an FLC, a set of fuzzy control rules, which characterize the dynamic behaviour of system, are defined. The inference engine is used to form inferences and draw conclusions from the fuzzy control rules. The output of inference engine is sent to defuzzification unit. Defuzzification is a mapping from a space of fuzzy control actions into a space of crisp control actions.

To implement FLAQ-RED scheme, we use a combination of two methods [11] to construct the fuzzy linguistic rules: (1) *The trail and error* and (2) *The theory method*. This is because the first method depends on domain expert knowledge and experience, and the second method tunes the input and output linguistic parameters to accurate values. Thus using both methods together, the output D_p can be obtained in more accurate.

In proposed FLAQ-RED, FLC consists of (1) fuzzification (2) evaluation of the rules, (3) aggregation the outputted rules, and (4) defuzzification, to compute the output D_p at the BS router queue. Then, the FLAQ employs two input linguistic variables ($AVQL, P_w$) with aim to evaluate a single output linguistic variable (D_p). The general bell membership function is used to represent all linguistic variables. The generalized bell function depends on three parameters a, b , and c as given by

$$f(x; a, b, c) = \frac{1}{1 + \left| \frac{x-c}{a} \right|^{2b}} \tag{3}$$

The parameter b is usually positive and the parameter c locates the centre of the curve. Then, every linguistic variable is linked to a fuzzy set of the input and output linguistic variables as defined in **Table 1**:

Table (1): Linguistic variables ranges

Variables ranges	Linguistic variables
$AVQL = 50-150$ [packet]	$AVQL = \{low, medium, high, very high\}$.
$P_w = 5\% - 30\%$	$P_w = \{low, medium, high, very high\}$.
$D_{predict} = 0\% - 30\%$	$D_{predicted} = \{very low, low, medium, high, very high\}$.

Fuzzy Rules Base:

Fuzzy rules are linguistic IF-THEN constructions that have the general form "IF A THEN B" where A and B are propositions containing linguistic variables. A is called the premise and B is the consequence of the rule. The type of any fuzzy system is specified by the type of the rules used in it. Generally there are two types of fuzzy systems known as Mamdani and Takagi-Sugeno-Kang (TSK) fuzzy systems [26].

An example of fuzzy rule for m inputs ($x1, x2, \dots, xm$) and only one output y (Multi Input Single Output (MISO)) could be defined in many forms like:

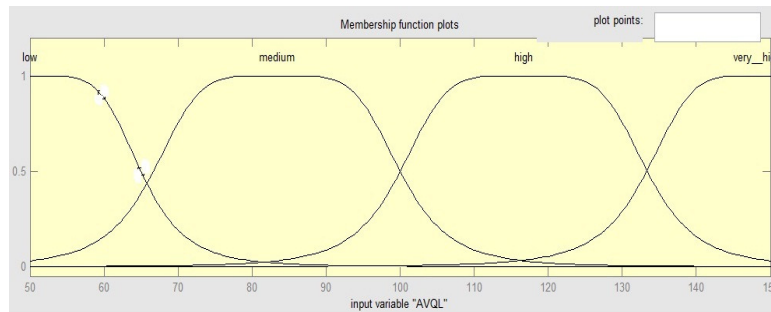
Mamdani Type: *R: If $x1$ is $\mu(x1)$ and $x2$ is $\mu(x2)$ and ... and xm is $\mu(xm)$ then y is $\mu(y)$... Or*

Sugeno Type: *R: If $x1$ is $\mu(x1)$ and $x2$ is $\mu(x2)$ and ... and xm is $\mu(xm)$ then $y = f(x1, \dots, xm)$...*

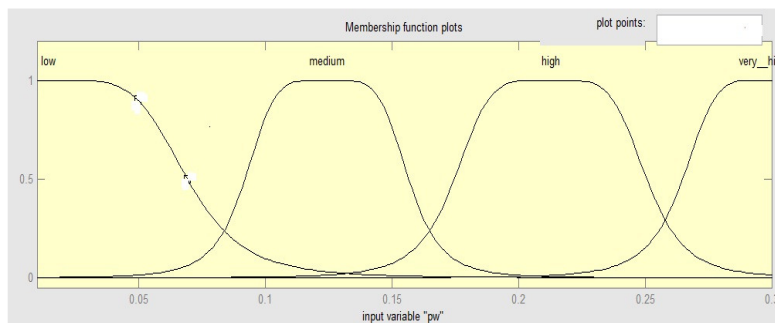
Because $AVQL$ has four membership functions and also P_w has four membership functions, then there are 16 fuzzy rules can be generated as listed in **Table 2**. As a result, **Figure 4** depicts the details of the proposed Fuzzy-Logic Adaptive Queuing based RED scheme using Mamdani Fuzzy –Logic when General Bell function of (3) is considered. The memberships of linguistic variables of $AVQL$, P_w and D_p are adjusted depending on the wireless environment in order to introduce the resultant fuzzy logic system decision surface.

Table (2): Fuzzy Rule-Base System

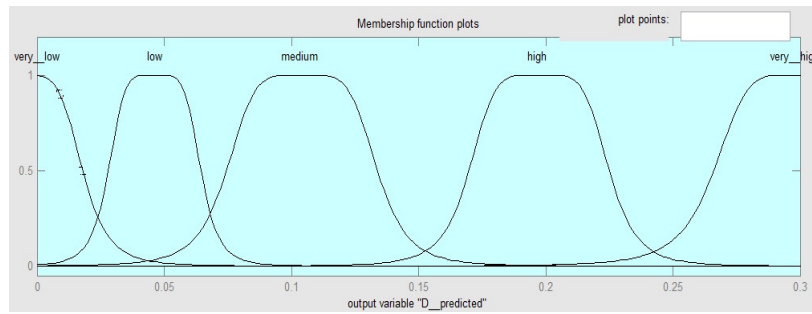
Fuzzy Rule Base System of FLAQ based RED		
If (AVQL is low) and (pw is low) then (D__predicted is very__low)		
If (AVQL is low) and (pw is medium) then (D__predicted is very__low)	If	
(AVQL is low) and (pw is high) then (D__predicted is very__low)	If	
(AVQL is low) and (pw is very__high) then (D__predicted is low)	If	
(AVQL is medium) and (pw is low) then (D__predicted is very__low)	If	
(AVQL is medium) and (pw is medium) then (D__predicted is very__low)	If	
(AVQL is medium) and (pw is high) then (D__predicted is low)	If	
(AVQL is medium) and (pw is very__high) then (D__predicted is medium)	If	If
(AVQL is high) and (pw is low) then (D__predicted is very__low)	If	
(AVQL is high) and (pw is medium) then (D__predicted is low)	If	
(AVQL is high) and (pw is high) then (D__predicted is medium)	If	
(AVQL is high) and (pw is very__high) then (D__predicted is high)	If	
(AVQL is very__high) and (pw is low) then (D__predicted is low)	If	
(AVQL is very__high) and (pw is medium) then (D__predicted is medium)	If	If
(AVQL is very__high) and (pw is high) then (D__predicted is high)	If	
(AVQL is very__high) and (pw is very__high) then (D__predicted is very__high)		



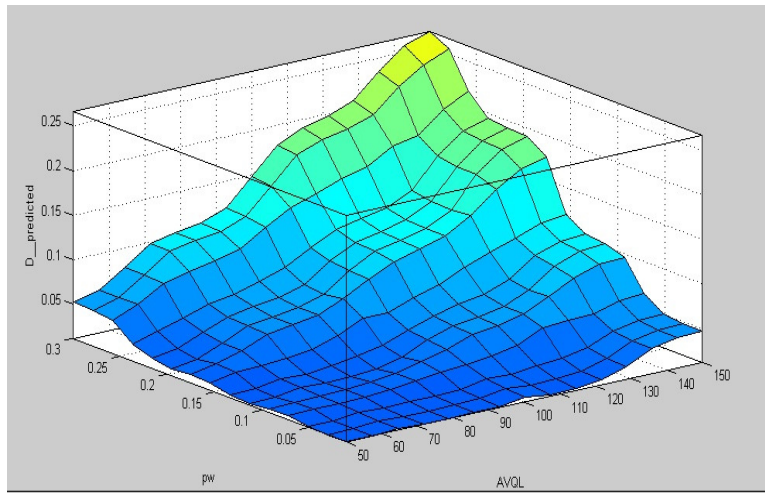
(a)



(b)



(c)



(d)

Figure 4. A Fuzzy-Logic Adaptive Queuing based RED scheme: Mamdani Fuzzy –Logic Based on General Bell function (a) membership of linguistic variable $AVQL$ (b) membership linguistic variable P_w (c) membership of output D_p , and (D) Fuzzy logic system decision surface

4. THROUGHPUT PERFORMANCE

4.1. Heuristic TCP Formula

There are several formulas that have been proposed to approximate the behaviour of TCP throughput [3][24][27]. In this framework, in order to estimate the network throughput, we can apply the heuristic TCP formula derived by Zorzi *et al.* [27] for the most accurate performance in wireless communications. According to it, the heuristic throughput of a single connection flow over Rayleigh fading channel can be defined by

$$R_k \approx R_s \begin{cases} 10^{X_s/(X_s+1)} & \text{if } \xi_w > \xi_s \\ 1 & \text{if } \xi_w \leq \xi_s \end{cases} \quad (4)$$

with,

$$X_s = 10^{\varepsilon \cdot \ln\left(\frac{1}{\xi_w - \xi_s} - 1\right)}$$

- R_k is the current average throughput at the sample k .
- R_s is the average throughput for error-free links and it can also denote the available or source bit rate in [bps].
- $\xi_w = 1 + \lceil \log_0(P_w) / 3 \rceil$
- $\xi_s = 1 / (\theta f_d + \beta) - K$.

Note that f_d denotes the Doppler frequency used to model the Rayleigh fading, whereas θ , β and K are constants. For numerical values we can use $\theta=1.39$, $\beta=2.78$, $K=0.03$, and a typical value for f_d is $f_d=6$ Hz. In addition, it is worth noting that the throughput expression given in (4) is a heuristic function obtained by fitting the simulation curves in [27].

To examine the effect of time period of measuring the bit error rate, i.e., the burst errors which accomplishes TCP traffic flow over wireless channel, the smaller value of time period of fading is expected so it can be adapted to the time changing of packet loss rate. However, it must be large enough for the fading speed. The fading speed thus is given by f_d (Doppler frequency) as

$$f_d = \frac{v \cdot f_c}{c} \quad (5)$$

where f_c is the carrier frequency, v is the vehicle speed (m/sec), and c is light speed. For 1.9 GHz in 3G cellular wireless networks, the above formula yields a Doppler frequency of 6 Hz/m/s. Therefore, for the high-speed wireless transmissions (e.g., in the order of Mbps), the fading speed can be viewed as roughly constant value for the time period of measurement of bit error rate. As a consequence, such fading effect is required to be investigated for the measured packet loss rate due to the burst bit errors.

According to our proposed FLAQ-RED controller described in **Figure 2**, we adopt the TCP layer packet error rate (PER) as the feedback information on the current degree of wireless environment reliability. Once the TCP layer drops a certain number of packets, the receiver (MH) informs the sender (BS) to tune the queue size in terms of $AVQL$, and consequently finding the predicted D_p , which is the target packet loss rate required to optimize and maximize the TCP throughput performance at the mobile receiver. Thus **Figure 4** explains the process of FLAQ-RED tuning algorithm based on TCP PER measurement via ACK feedback.

4.2. Buffer Queue Model Based On RTT

In our FLAQ-RED scheme, firstly we let $B_k \leq B_w$ where B_w stands for available wireless channel bandwidth and B_k is the current available bandwidth [24]. Then, the buffer queue length (Q) is expressed at the sampling instant $(k+1)$ by the following equation [28],

$$Q_{k+1} = Q_k + RTT \times (R_{k-1} - B_k) \quad (6)$$

where RTT denotes the current round-trip time which is also the sampling interval and R_{k-1} is the TCP source rate of BS during the previous RTT . This allows for the round-trip delay in the control loop. In (6), B_k denotes the available network bandwidth. *Since the heuristic TCP throughput formula defined in (1) does not directly depend on RTT , therefore we are able to assume that the available network bandwidth equals the previous TCP source rate, then the resultant average queue length ($AVQL$) in (6) is being $Q_{k+1}=Q_k$.*

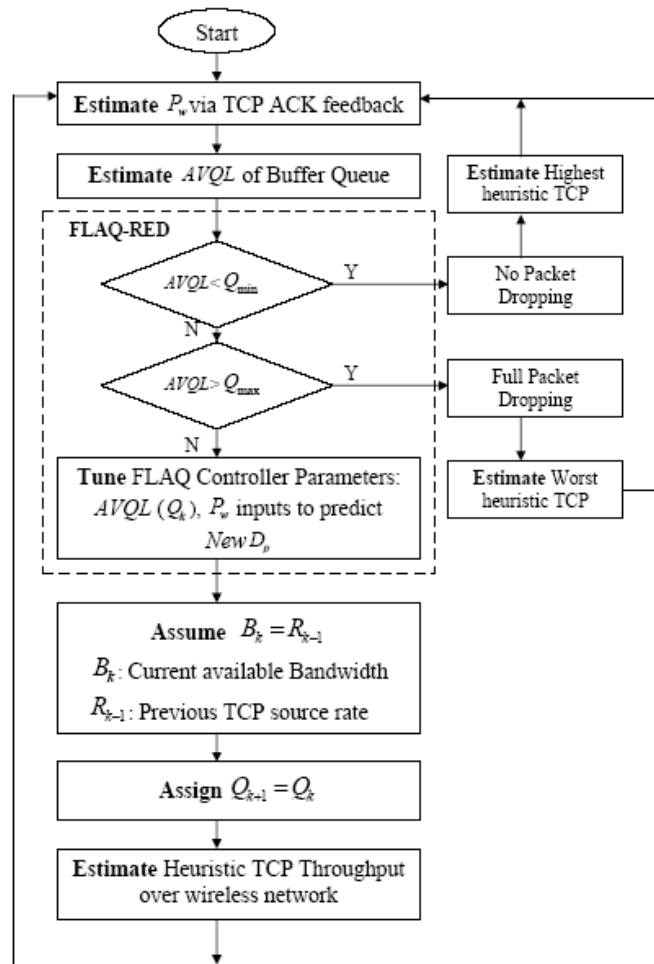


Figure 4. Flow chart of control loop of the proposed FLAQ-RED algorithm to estimate heuristic TCP throughput performance

5. SIMULATION RESULTS

5.1. Network Settings

Using **Table 3**, we have conducted our simulation in UMTS cellular wireless network in order to investigate the improvement in TCP performance over a wireless link of BS and MH nodes.

Table (3): Simulation Parameters [22][24]

Parameters	Value
RTT_{min}	168 ms
B_w	1 Mbps
Carrier frequency	1.9 GHz
Source bit rate	1 Mbps
Buffer Queue Size	200 [Packet]
Q_{min} , Q_{max}	50 [Packet], 150 [Packet]
Mobile speed	1m/s - 10m/s
P_w	5%-30%

5.2. Performance Evaluation

We performed our simulation experiments using the numerical values described in Table 1 as follows. MATLAB mfile and fuzzy logic tool box are used to simulate our proposed FLAQ-RED scheme. We conducted our experiments to investigate two cases of user's mobility as follows: (1) the stationary receiver (i.e., there is no Doppler effect) and (2) The mobile receiver (i.e., there is Doppler frequency) over time-varying wireless channel.

The first experiment was on predicting the dropping packet rate at the BS router using our proposed FLAQ-RED scheme. The wireless packet loss rate and the measured average queue length are considered as inputs of FLAQ-RED controller. The range of predefined wireless packet rate is changed over the range of 5% -30%. Using General Bell fuzzy logic model we were able to estimate the predicted dropping packet rate for the next transmission. In this model, we assume the maximum buffer size is up to 200 packets, and the two queue threshold levels are Q_{min} and Q_{max} as defined in Table 1, respectively.

Figure 6 displays the normalized throughput vs. the AVQL for various values of the predefined P_w for stationary receiver over wireless channel. It is found that our FLAQ-RED introduces an excellent performance especially for high values of the predefined P_w . The maximum achievable throughput of 0.9999 can be obtained when AVQL does not exceed the two queue thresholds and P_w equals 5%. In addition, the higher values of P_w , say 30%, the resultant throughput will not be less than 0.55.

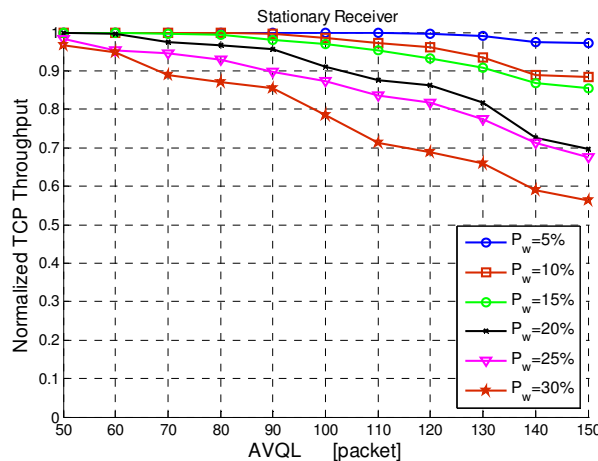


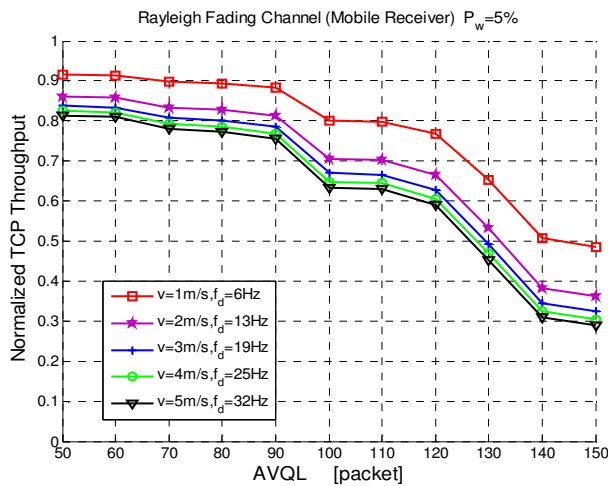
Figure 6. Normalized TCP throughput vs. AVQL for various values of packet loss rate

Table 4 summarizes the performance comparison of our proposed FLAQ-RED scheme. The normalized TCP throughput is evaluated at the predicted packet dropping rate for the two queue length thresholds, $Q_{min}=50$ and $Q_{max}=150$, respectively. The results obtained indicate that FLAQ-RED is able to tune its inputs to achieve the maximum throughput at Q_{min} and the minimum throughput at Q_{max} , respectively. For instance, once P_w increases to 30% the worst achievable heuristic throughput becomes 0.5638 at Q_{max} . Then, using RED algorithm, any increment beyond Q_{max} takes place; there is an increment in D_p and consequently the predicted throughput drops immediately down to be in a worst state over the network.

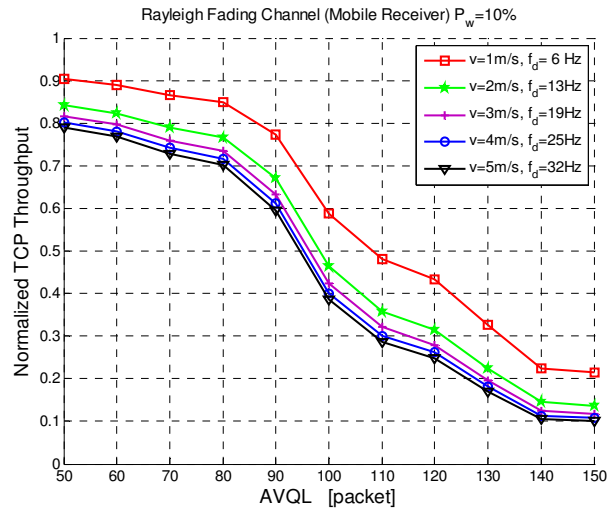
Figure 7 reveals the effect of user’s mobility on the TCP performance for the two predefined P_w of 5% and 10%, respectively. The results obtained uses UMTS cellular network assuming the operating frequency of 1.9 GHz and the mobile speed varies up to 10 m/s. It is clearly shown

Table (4): Heuristic TCP Throughput Performance of FLAQ-RED at Stationary Receiver

Predefined P_w	Predicted D_p , Heuristic TCP Throughput	
	$Q_{min} = 50$	$Q_{max} = 150$
5%	1.29% , 0.9999	4.96% , 0.9714
10%	1.40% , 0.9998	10.0% , 0.8835
15%	1.74% , 0.9992	11.4% , 0.8551
20%	1.91% , 0.9987	19.1% , 0.6976
25%	3.97% , 0.9835	20.2% , 0.6764
30%	5.24% , 0.9675	26.6 % , 0.5638



(a)



(b)

Figure 7. Normalized TCP throughput vs. AVQL over time-varying fading channel (mobile receiver) (a) $P_w = 5\%$ and (b) $P_w = 10\%$

that once the mobile speed increases up to 5m/s, the heuristic TCP throughput rapidly degrades especially when AVQL achieves Q_{\max} , where all arriving packets are dropped. Then, the lowest TCP throughput achieves 0.6 and 0.1, respectively, when P_w changes from 5% to 10% and AVQL becomes Q_{\max} . Moreover, TCP throughput performance becomes independent (i.e., close to i.i.d case) when Doppler frequency closes to 5 m/s or higher. Finally, **Tables 5 and 6** both can summarize some details on performance comparison of the proposed FLAQ-RED under different channel conditions.

Table (5): Heuristic TCP throughput performance of FLAQ-RED over wireless Rayleigh fading channel at $P_w = 5\%$

Mobile Speed v (m/s)	Doppler Frequency (Hz)	Predicted D_p , Optimal Heuristic TCP Throughput	
		$Q_{\min} = 50$	$Q_{\max} = 150$
1	6	1.29% , 0.9162	4.96% , 0.4843
2	13	1.29% , 0.8610	4.96% , 0.3620
3	19	1.29% , 0.8383	4.96% , 0.3246
4	25	1.29% , 0.8246	4.96% , 0.3046

Table (6): Heuristic TCP throughput performance of FLAQ-RED over wireless Rayleigh fading channel at $P_w = 10\%$

Mobile Speed v (m/s)	Doppler Frequency (Hz)	Predicted D_p , Optimal Heuristic TCP Throughput	
		$Q_{\min} = 50$	$Q_{\max} = 150$
1	6	1.40% , 0.9027	10.0% , 0.2145
2	13	1.40% , 0.8415	10.0% , 0.1375
3	19	1.40% , 0.8166	10.0% , 0.1174
4	25	1.40% , 0.8017	10.0% , 0.1072

6. CONCLUSIONS AND FUTURE WORK

In this paper, we have proposed an effective scheme called FLAQ-RED to improve the heuristic TCP performance based on active queue management (AQM) in mobile environment. The classical RED algorithm is considered to tune the queue length at the base station node when a predefined packet loss rate is fed via TCP ACK feedback from the mobile host. Fuzzy-logic controller is used to predict the dropping packet rate for the next transmission. The obtained results introduce a good improvement in the link reliability observed by a heuristic TCP throughput over Rayleigh fading channel under different conditions of user's mobility. For further work, this proposed scheme can be extended to investigate the congestion effect of multiple-TCP connection flows. Furthermore, the effect of dynamic packet sizes on this FLAQ-RED scheme can be taken into account in order to improve the multimedia traffic over wireless environment. The scheme can also be improved by applying the weighted RED to provide different multiple-queue threshold levels for traffic classes.

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