Performance Analysis of Best-Effort Service in Saturated IEEE 802.16 Networks

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Abstract—The IEEE 802.16 standard is a promising technology for the fourth-generation mobile networks and is being actively promoted by the Worldwide Interoperability for Microwave Access (WiMAX) Forum: an industry-led consortium. Among the various service classes supported in the standard, best-effort service is expected to be the major service class subscribed to by users due to its operational simplicity and lower charging rate compared with other service classes. In this paper, we investigate the throughput and packet-access-delay performance of best-effort service using the contention-based bandwidth request mechanism in a saturated IEEE 802.16 network. In particular, we develop a simple fixed-point analysis to approximate the failure probability of a bandwidth request and derive analytical expressions for network throughput and packet access delay. Furthermore, the uniqueness of the fixed point is established. The accuracy of the analytical model is validated by comparison with simulation over a wide range of operating conditions. The implications of various different parameter configurations on the system performance are investigated using the analytical model. The utility of the model is further demonstrated by finding the maximum achievable throughput and obtaining its corresponding optimal initial contention window.

Index Terms—Best effort, IEEE 802.16, performance analysis, Worldwide Interoperability for Microwave Access (WiMAX).

I. INTRODUCTION

WORLDWIDE Interoperability for Microwave Access (WiMAX) was first developed for wireless broadband access and has evolved into one of the candidate technologies for fourth-generation mobile communication networks. Based on the IEEE 802.16-2004 Air Interface standard [1] and the IEEE 802.16e amendment [2], WiMAX can support 1) a very high capacity, for example, a theoretical peak data rate of 60 Mb/s for a totally downlink operation (without uplink) and 28 Mb/s for a totally uplink operation can be achieved with the use of two antennas at 10-MHz channel bandwidth; 2) wide-area mobility, for example, a speed of up to 120 km/h with handoff; and 3) multimedia services with different traffic characteristics and various quality-of-service (QoS) requirements.

To support multimedia services, the IEEE 802.16 defines five service classes referred to as scheduling services: unsolicited grant service (UGS), extended real-time polling service (ertPS), real-time polling service (rtPS), non-real-time polling service (nrtPS), and best-effort (BE) service [3]. The main characteristics of these scheduling services are described as follows.

1) UGS is designed for constant-bit-rate real-time traffic such as E1/T1 circuit emulation and voice over Internet Protocol (VoIP) without silence suppression. The main QoS parameters are the maximum sustained rate (MSR), the maximum latency, and the tolerated jitter.

2) ertPS is designed to support VoIP with silence suppression. It has the same QoS parameters as UGS. However, it is allocated the MSR only during active periods and is not allocated any bandwidth during silent periods.

3) rtPS is designed to support real-time applications with variable bit rates, such as MPEG videos. Other than the MSR, its QoS parameters also include the minimum reserved rate.

4) nrtPS is designed for applications without any specific delay requirement but with the need for a minimum amount of bandwidth, such as a file transfer protocol.

5) BE service is designed for applications that are delay tolerant and do not require a minimum amount of bandwidth. Bandwidth will be granted to this service class if, and only if, there is leftover bandwidth from other classes.

The basic operating mode of WiMAX is the point-to-multipoint (PMP) mode, where a base station (BS) serves a set of subscriber stations (SSs) within the same antenna sector in a time-division-multiplexed fashion. In the downlink direction, the BS is the only sender. It has full control of which time slot to use to send data to which SS, and data are broadcast to all SSs. In the uplink direction, in which SSs send data to the BS, SSs share a common channel. To avoid collisions, only one SS should be permitted to transmit at any time slot. According to the IEEE 802.16 standard, such an exclusive access is achieved by requiring each SS to have granted time slots before it can transmit. For this purpose, several request/grant mechanisms at the medium access control (MAC) layer have been specified for various scheduling services [4]. They are unsolicited granting, unicast polling, and broadcast polling. Among these bandwidth-reservation mechanisms, the broadcast polling mechanism is contention based and requires SSs to use the binary exponential backoff (BEB) algorithm [5] for contention resolution. It is typically used by BE connections.
BE service is relatively simple to provide because it does not involve QoS negotiation and enforcement of traffic parameters. Moreover, since BE service offers no QoS guarantees, its service charge is expected to be cheaper than that of other scheduling services. As a result, it would be a major scheduling service to which to be subscribed by WiMAX users. Therefore, it is important to understand the network performance when delivering BE service.

However, recent research on MAC-layer performance in WiMAX networks just focuses on the performance of bandwidth-request mechanisms. In [6], the delay performance of the contention-free unicast polling request mechanism is analytically studied. The cases in which nodes are sequentially polled at the beginning or end of each uplink subframe are both modeled. In [7], the delay performance under broadcast polling is investigated. An analytical model based on a Markov chain is used to evaluate the average delay of a transmission request for saturated networks. An alternative modeling approach is also proposed in [8]. The model in [7] is later extended to investigate the case with Bernoulli request arrival [9]. In [10], the capacity of the contention slots in delivering bandwidth requests, and the average access delay are derived, taking into account the response time for a bandwidth assignment and the possible timeouts for lost messages. However, the work in [7]–[10] does not take into account the delay incurred by data-packet transmission after a request is successfully transmitted. Furthermore, the delay is measured in terms of transmission frames, instead of measuring the time between the first attempt of a request and the completion of the packet transmission. In [11], the performance of bandwidth-request mechanisms based on piggyback and broadcast polling is compared by simulation.

The first major contribution in this paper is an accurate analytical model for evaluating the network throughput and packet access delay of BE service. To this end, we study the performance of BE service based on broadcast polling in a saturated WiMAX network. Here, a saturated network means that each SS always has a packet to send. Considering such a condition allows us to obtain the upper bound of the network performance. Our analytical model is based on a set of fixed-point equations that calculates the failure probability of a bandwidth request in broadcast polling as a function of system parameters. Furthermore, we show that these fixed-point equations have a unique solution. Based on the obtained failure probability of a bandwidth request, the network throughput and the packet-access delay are then derived.

Our second major contribution is to show the utility of the proposed analytical model in designing a better network. In particular, we use the model to investigate the optimal achievable throughput under different parameter settings at the MAC layer. This leads us to develop a mechanism for adjusting the initial contention window according to system parameters to achieve this optimality. This paper provides operators with analytical tools to evaluate the network performance and to gain insights into the optimal configuration of system parameters.

The rest of this paper is organized as follows. Section II describes the details of the broadcast polling mechanism. Section III presents our analytical model. Section IV validates the model by comprehensive comparisons with simulation results and evaluates the performance of BE service under different configurations of system parameters. Section V investigates how the optimal throughput depends on various system parameters and presents a mechanism for adaptively achieving the optimal throughput. Finally, conclusions are drawn in Section VI.

II. BROADCAST POLLING

The MAC frame structure defined by the IEEE 802.16 standard for time-division duplexing (TDD) in PMP mode is shown in Fig. 1. In this mode, the channel is time slotted into fixed-length frames; each consists of a downlink and an uplink subframe. The durations of the downlink and uplink subframes are dynamically controlled by the BS by broadcasting the so-called downlink map (DL Map) and uplink map (UL Map) messages at the beginning of each frame (as indicated in Fig. 1). The UL Map contains data or information elements that inform the SSs about transmission opportunities for bandwidth requests and data packets. Consider a frame $i$; when an SS has a data packet to send, it first sends a bandwidth request for transmitting its data in one of the transmission opportunities within the request interval of the uplink subframe. Upon receiving the bandwidth requests, the BS then allocates bandwidth or data slots for data transmission in the uplink data interval of frame $i + 1$ based on its scheduler.

In the following, we consider a scenario where broadcast polling is used by the BS with $m$ (fixed) transmission opportunities for bandwidth requests, which are referred to as request slots. In this case, if there is only one request submitted to a request slot, the request is successful. On the other hand, if there are two or more SSs submitting their requests to the same request slot, a collision occurs, and truncated BEB is used to resolve the contention. In particular, when sending a request at backoff state $i$, $i > 0$, an SS must carry out the backoff process by randomly selecting a backoff time in the range $[0, W_i - 1]$, where $W_i$ is the contention window for backoff state $i$. Here, the backoff time represents the number of request slots that must pass before the request can be submitted. At backoff state 0 (which is the first attempt), $W_0$ is set equal to $W$, which is the initial contention window. If the request is unsuccessful, then the contention window size is multiplied by $\alpha = 2$, and another backoff period is initiated. The process of doubling the
contention window continues until the maximum possible value, i.e., $CW_{\text{max}} = \alpha' W$, $r \geq 1$, is reached. Here, $r$ is referred to as the truncation value. If the request is unsuccessful after $r$ reattempts, the contention window is no longer expanded but maintained at $CW_{\text{max}}$ for the remaining attempts until either the request is successful or the maximum allowable number of attempts, i.e., $R \geq r + 1$, has been made. If the request is still unsuccessful after $R$ attempts, the packet is discarded. In summary, $W_i$ is given by

$$W_i = \begin{cases} \alpha^i W, & 0 \leq i \leq r \\ \alpha^i W, & r < i < R. \end{cases} \quad (1)$$

In this paper, the SSs are only allowed to request bandwidth to transmit one packet per request, and all packets are assumed to have the same length. Let $t$ be the length of a request (or backoff) slot. Furthermore, we assume that the BS always allocates the same amount of uplink capacity consisting of $d \leq m$ data slots in every uplink subframe for uplink traffic. Each data slot is of length $T \gg t$, which is the transmission time of a packet. As the standard does not define scheduling algorithms for both the BS and SSs, we assume here that the BS uplink scheduler will uniformly allocate bandwidth to SSs whose bandwidth requests were successful in the previous frame. Let $j$ be the number of requests that do not collide. If $j < d$, then in the next frame, there will be $(d - j) > 0$ data slots that remain unused and are wasted. However, if $j > d$, then $j - d > 0$ requests must be declined because there are only $d$ slots available in the next frame; these $j - d$ requests are also considered unsuccessful.

### III. ANALYTICAL MODEL

Consider an IEEE 802.16 network consisting of $N$ saturated SSs operating in the PMP mode. Our objective is to first develop a fixed-point approximation to compute the probability that a request is unsuccessful. The expressions for network performance metrics, such as the network throughput and the packet delay, will then follow.

#### A. Unsuccessful Request Probability

Let $p$ be the probability that a request sent by an SS is unsuccessful. As in [8], a request is regarded as unsuccessful when either the request experiences a collision during transmission (with probability $p_c$) or the request is successfully transmitted but the BS could not allocate bandwidth to it due to insufficient data slots (with probability $p_d$). For simplicity, we model these two events as independent. Then, $p$ can be expressed as

$$p = 1 - (1 - p_c)(1 - p_d). \quad (2)$$

Since both $p_c$ and $p_d$ can, in turn, be expressed as a function of the probability $p$, fixed-point equations can be established to calculate the individual probabilities. The fixed-point analysis is detailed as follows.

A request is successful on the first attempt with probability $1 - p$. Recall that the contention window is initially set to $W$, the average number of elapsed backoff slots before such a request is $(m/2) + (W - 1)/2$. The first term is due to the fact that an SS cannot start a new backoff period for its next request immediately after the previous one in the same frame but has to wait until the next frame; because requests slots are uniformly chosen among the $m$ request opportunities in each frame, the average number of backoff slots wasted until the next frame is $m/2$. The second term represents the average number of backoff slots an SS has to wait before attempting to send a request based on the contention-resolution mechanism described in Section II.

If its first attempt fails, a request is successful on the second attempt with probability $p(1 - p)$. The average number of elapsed backoff slots in this case is $(m/2) + (\alpha W - 1)/2$. Continuing this argument until the $R$th attempt yields the average number of elapsed backoff slots ($B_{\text{avg}}$) before the request is successful. We have

$$B_{\text{avg}} = \frac{m}{2} + \eta \sum_{i=0}^{r-1} p^i \left( \frac{\alpha^i W - 1}{2} \right) + \eta \left( \frac{\alpha^i W - 1}{2} \right)^{R-1} \sum_{i=r}^{R-1} p^i$$

$$= \frac{m}{2} + \eta W \frac{(1 - (\alpha p)^r)}{2(1 - \alpha p)} - \frac{1 - (\alpha W)^r}{2(1 - p^R)}$$

$$+ \frac{(\alpha^i W - 1)(p^r - p^R)}{2(1 - p^R)} \quad (3)$$

where $\eta = (1 - p)/(1 - p^R)^{-1}$, and $(1 - p^R)$ is a normalization factor.

Note that $B_{\text{avg}}$ is the average number of backoff slots an SS has to wait before sending requests, i.e., it is the average interarrival time of the requests in this system. Therefore, the probability that an SS attempts to send the request in a slot is given by

$$\tau = 1/(B_{\text{avg}} + 1). \quad (4)$$

Given that there are $N$ saturated SSs, the probability $p_c$ that a request sent by an SS collides with other requests can be expressed as

$$p_c = 1 - (1 - \tau)^{N^{-1}}. \quad (5)$$

Let $\xi$ be the probability that a collision-free request is made in a given slot, given that there are $N$ SSs, each attempting to send requests with probability $\tau$. Under the approximation that requests are independent, we have

$$\xi = N\tau(1-\tau)^{N^{-1}}. \quad (6)$$

The probability that there are $j$, $0 \leq j \leq k = \min(m, N)$ collision-free requests among $m$ request slots is then given by a truncated binomial distribution

$$Q(j) = \frac{\sum_{i=0}^{k} \binom{m}{i} \xi^i (1 - \xi)^{m-i}}{\sum_{i=0}^{k} \binom{m}{i} \xi^i (1 - \xi)^{m-i}}. \quad (7)$$
The probability that a collision-free request is unsuccessful due to lack of bandwidth in the subsequent frame can then be expressed as

\[ p_d = \frac{\sum_{j=d+1}^{k} (j - d)Q(j)}{\sum_{j=0}^{k} jQ(j)}. \]  

(8)

Equations (2)–(8) create a fixed-point formulation from which \( p \) can numerically be computed. The existence and uniqueness of a solution for the aforementioned formulation are provided in the next theorem, which is proven in Appendix A.

**Theorem 1:** Let \( g(\tau(p_c, p_d)) : [0, 1] \rightarrow [0, 1] \) be defined by the right-hand side of (8), and let \( f(\tau, p_c, p_d) = g(\tau(p_c, p_d)) - p_d \).

For \( m \leq N \)

1) for a given \( p_d, 0 \leq p_d \leq 1 \), the function \( \tau(p_c, p_d) : [0, 1] \rightarrow [0, 1] \) defined in (4) and \( p_c \) defined in (5) have a unique fixed-point solution;

2) the equations \( f(\tau, p_c, p_d) = 0 \), (4), and (5) have a unique fixed-point solution.

**B. Throughput Analysis**

From (7), the probability that there are \( j \) collision-free requests in each uplink subframe is given by \( Q(j) \). If \( j \leq d \), each request is allocated a data slot, and thus, \( j \) packets will be transmitted in the subsequent uplink subframe. On the other hand, if \( j > d \), only \( d \) requests are allocated data slots. As a result, only \( d \) packets will be transmitted in the subsequent uplink subframe. Therefore, the normalized throughput \( \Gamma \) is given by

\[ \Gamma = \frac{\sum_{j=1}^{d} jQ(j) + \sum_{j=d+1}^{k} dQ(j)}{d}. \]  

(9)

**C. Delay Analysis**

The delay of a packet is defined as the time duration from its first bandwidth request until the packet transmission is completed. Note that if the bandwidth reservation of a packet is successful, the packet will be removed from the head of the queue into a transmission queue waiting for its transmission in the coming uplink subframe. Under the saturation condition, an SS will then start a new backoff process for its next packet (which is now at the head of the queue) in the subsequent frame.

Referring to Fig. 1, for any data packet of a tagged SS, the time epoch at the end of the request interval in which its first request is sent is denoted as \( \tau_a \), and the time epoch at the beginning of the uplink data interval in which the data packet is sent is denoted as \( \tau_c \). Let \( U \) and \( V \) be the random variables (RVs) representing the time durations from the time of sending the first request until \( \tau_a \) and from \( \tau_a \) until the packet has been transmitted, respectively. The duration of the uplink subframe is given by \( T_{UL} = T_{RE} + T_{DA} \), where \( T_{RE} \) is the length of the request interval and \( T_{DA} = dT \) is the length of the uplink data interval, respectively.

Given that the tagged SS is successful in its first attempt of sending a bandwidth request, the packet delay \( X^{(0)} \) is therefore given by

\[ X^{(0)} = U + \Delta + V \]  

(10)

where \( \Delta \) is the duration of an IEEE 802.16 frame [including both downlink \( (T_{DL}) \) and uplink subframes \( (T_{UL}) \) and their guard times]. Note that the backoff period of this first attempt is not included as part of the delay due to the way we have defined the packet delay period, which starts from the time epoch when the first request is sent.

For the case when the tagged SS is not successful in its first attempt but is successful in the second attempt of sending bandwidth request, the packet delay can be expressed as

\[ X^{(1)} = U + Y^{(1)} + V \]  

where \( Y^{(1)} \) is the random time that a packet has to wait between the epochs \( \tau_a \) and \( \tau_c \) due to the first and the second request attempt and is given by

\[ Y^{(1)} = \sum_{i=0}^{1} K^{(i)} \Delta \]  

where \( K^{(i)} \) is the number of frames delayed in attempt \( i \). Using this notation, we can rewrite (10) as

\[ X^{(0)} = U + Y^{(0)} + V \]  

where, by definition, \( Y^{(0)} = K^{(0)} \Delta \) and \( K^{(0)} \) is always equal to one. In the second request attempt, the SS will uniformly choose its backoff in \([0, W_1 - 1] ; W_1 = \alpha W \), and thus, \( K^{(1)} \) is a discrete RV with the following probability mass function (pmf):

\[ K^{(1)} = \begin{cases} 
1, & \text{w.p. } m/W_1 \\
2, & \text{w.p. } m/W_1 \\
\vdots & \vdots \\
A_1 - 1, & \text{w.p. } m/W_1 \\
A_1, & \text{w.p. } 1 - \frac{(A_1 - 1)m}{W_1} 
\end{cases} \]  

(11)

where \( A_1 = [W_1/m] \), \( \lfloor x \rfloor \) gives a minimum integer value that is greater than or equal to \( x \), and w.p. stands for “with probability.”

In general, for a packet transmission requiring \( i \) request attempts, the packet delay \( X^{(i)} \) is given by

\[ X^{(i)} = U + Y^{(i)} + V, \quad 0 \leq i < R \]  

(12)

with

\[ Y^{(i)} = \sum_{j=0}^{i} K^{(j)} \Delta \]  

and \( K^{(i)}, 0 < i < R \) has the following pmf:

\[ K^{(i)} = \begin{cases} 
j, & \text{w.p. } m/W_i, \quad j = 1, 2, \ldots, A_i - 1 \\
A_i, & \text{w.p. } 1 - \frac{(A_i - 1)m}{W_i} 
\end{cases} \]  

(13)

where \( A_i = [W_i/m] \). Note that if \( A_i = 1 \), i.e., \( m \geq W_i \), then \( K^{(i)} = 1 \) with probability one.
Overall, the packet delay $X$ can be expressed as

$$X = X^{(i)} \text{ w.p. } \eta p^i, \quad 0 \leq i < R.$$  \hspace{1cm} (14)

To complete the expression of $X$, we now determine the pmf of the $U$ and $V$ RVs. As the tagged SS uniformly chooses the backoff before sending its request, the pmf of $U$ can be approximated as

$$U = it \text{ w.p. } 1/m, \quad i = 1, 2, \ldots, m.$$  \hspace{1cm} (15)

Equation (15) assumes that an SS always sends its request at the beginning of the chosen slot.

As the BS uniformly allocates data slots among successful requests, the pmf of $V$ can be expressed as

$$V = iT \text{ w.p. } \sum_{j=i}^{k'} \frac{q(j-1)}{j}, \quad i = 1, 2, \ldots, k'$$  \hspace{1cm} (16)

where $k' = \min\{k, d\}$, and $q(j)$ is the probability that there are $j \geq 0$ successful requests other than the tagged SS in a frame. The probability $q(j)$ follows a truncated binomial distribution

$$q(j) = Q(j+1)/(1-Q(0)), \quad 0 \leq j \leq k-1$$  \hspace{1cm} (17)

where $Q(j)$ is given in (7).

From (14), we obtain

$$E[X] = \eta \sum_{i=0}^{R-1} p^i E[X^{(i)}]$$

$$\text{Var}[X] = \eta \sum_{i=0}^{R-1} p^i \left( \text{Var}[X^{(i)}] + \left( E[X^{(i)}] - E[X] \right)^2 \right).$$  \hspace{1cm} (18)

The mean and variance of $X^{(i)}$ are derived from (12)

$$E[X^{(i)}] = E[U] + E[Y^{(i)}] + E[V]$$

$$\text{Var}[X^{(i)}] = \text{Var}[U] + \text{Var}[Y^{(i)}] + \text{Var}[V]$$  \hspace{1cm} (19)

where

$$E[Y^{(i)}] = \Delta \sum_{j=0}^i E[K^{(j)}]$$

$$\text{Var}[Y^{(i)}] = \Delta^2 \sum_{j=0}^i \text{Var}[K^{(j)}].$$

From (13), it can be shown that

$$E[K^{(j)}] = \begin{cases} 1, & j = 0 \\ A_j - A_j(A_j - 1)(1 + 4A_j) & \frac{m}{60}\text{ms/W}, \quad j = 1, \ldots, R - 1 \\ A_j - A_j(A_j - 1)(1 + 4A_j) & \frac{m}{60}\text{ms/W}, \quad j = r, \ldots, R - 1 \end{cases}$$  \hspace{1cm} (20)

and

$$\text{Var}[K^{(j)}] = E[K^{(j)}]^2 - \left( E[K^{(j)}] \right)^2$$

where $K^{(j)}$ is the second moment of $K^{(j)}$ and is given by

$$E[K^{(j)}] = \begin{cases} 1, & j = 0 \\ A_j - A_j(A_j - 1)(1 + 4A_j) & \frac{m}{60}\text{ms/W}, \quad j = 1, \ldots, r - 1 \\ A_j - A_j(A_j - 1)(1 + 4A_j) & \frac{m}{60}\text{ms/W}, \quad j = r, \ldots, R - 1 \end{cases}$$

It remains to determine $E[U], \text{Var}[U], E[V]$, and $\text{Var}[V]$ from (15) and (16), which can be expressed as

$$E[U] = (m + 1)T/2$$

$$\text{Var}[U] = \frac{(m + 1)(m + 2)}{36} \left( \frac{T}{2} \right)^2,$$  \hspace{1cm} (21)

and $q(j)$ is given in (17). The delay and variance of packet delay are then calculated by substituting (19)–(21) into (18).

IV. MODEL VALIDATION AND PERFORMANCE ANALYSIS

In this section, we verify our analytical model using simulation and study in detail the network performance, including the throughput and the mean and standard deviation of delay, as functions of $N$, $m$, $d$, $r$, $R$, and $W$. To this end, we have developed a simulator [12] to simulate the broadcast polling with the BEB contention resolution mechanism, as described in Section II. The simulator is event driven and developed using C++. The duration of each simulation run is 3000 s, with a warm-up period of 300 s. The MAC- and physical-layer parameters were configured in accordance with default parameters taken from the standard [1]. In particular, the frame duration is 1 ms, consisting of 5000 physical slots or 2500 minislots, each having 0.4-μs length. The data rate is 120 Mb/s employing 64-ary quadratic-amplitude modulation at 25 MHz. Each bandwidth request consists of six minislots, including three minislots for an SS transition gap, two minislots for preamble, and one minislot for a bandwidth request message of 48 bits. The length of a data slot including the preamble and transition gap is 37.6 μs (i.e., 94 minislots), which allows the transmission of an approximately 0.5-kB packet per data slot. Using this simulator, we have validated our analytical model with different sets of typical system parameters. As shown by the results presented below, the model accurately predicts the performance of the contention-based BE service of the IEEE 802.16 standard and is therefore suitable to study the impact of different parameters on the overall system performance.

We first validate our model for the probability of unsuccessful REQ ($p$) using the aforementioned simulation. We set $r = 3$, $R = 5$, and $m = d = 10$ and obtain $p$ for different
Fig. 2. Unsuccessful request probabilities \((r = 3, R = 5, m = 10, \text{ and } d = 10)\). 

Fig. 3. Unsuccessful request probabilities \((r = 3, R = 5, W = 16, \text{ and } m = 10)\). 

Fig. 4. Unsuccessful request probabilities \((r = 3, R = 5, W = 16, \text{ and } d = m)\). 

Fig. 5. Normalized throughput \((r = 3, R = 5, m = 10, \text{ and } d = 10)\). 

numbers of SSs with \(W = 8, 16, 32, 64\), respectively. The results are shown in Fig. 2. As expected, a larger \(N\) leads to more request contention and, thus, a larger \(p\). While other system parameters are fixed, \(p\) decreases as \(W\) increases. This is because when \(W\) increases, there are more choices of a request slot in each backoff stage. As a result, the probability that an SS transmits a request in a request slot \((\tau)\) becomes smaller. Hence, \(p_c\) and \(p\) decrease. We now set \(W = 16, r = 3, R = 5, \text{ and } m = 10\) and obtain \(p\) for different numbers of SSs with \(d = 4, 6, 8,\) and 10. As can be seen from Fig. 3, the curves for \(d = 6, 8,\) and 10 essentially overlap each other, while the curve for \(d = 4\) exhibits a higher \(p\). This is because, given \(m\) available request slots, the average number of successful requests is less than \(m\). In this example case, \(d = 6\) is quite sufficient to serve the successful requests, and therefore, increasing \(d\) does not significantly change \(p\). However, when \(d\) becomes too small, e.g., \(d = 4\), there are not enough data slots available in the subsequent subframe, which then results in a larger \(p\). Next, we set \(W = 16, r = 3, R = 5, \text{ and } m = d\) and obtain \(p\) for different numbers of SSs with \(m = 10, 15,\) and 20, respectively. The results are shown in Fig. 4. It can be seen that \(p\) increases when \(m\) decreases, as a smaller \(m\) causes a higher contention probability in each request slot and, hence, a higher \(p\).

We now investigate the throughput performance under different system parameters. First, we set \(r = 3, R = 5, \text{ and } m = d = 10\) and plot the throughput against different numbers of SSs with \(W = 8, 16, 32,\) and 64, respectively, in Fig. 5. It can be seen that, for a given \(W\), the throughput varies with \(N\) such that there exists an optimal \(N\) that maximizes the throughput. The reason is that when \(N\) is small, the demand for data slots is low, and hence, the throughput is low. As \(N\) gradually increases, the throughput also increases. However, further increasing \(N\) will also cause more contention, which at some point results in fewer successful requests and a lower throughput. Thus, there exists an \(N\) value where the throughput is maximized. Alternatively, for a fixed \(N\), e.g., at \(N = 40\), the throughput increases and then decreases when \(W\) varies from 8 to 64, which indicates that there also exists an optimal \(W\).
at which the throughput is maximized. Furthermore, it is interesting to see from the plot that the maximum throughput value is more or less the same, irrespective of the \( N \) and \( W \) values.

Next, we set \( W = 16, r = 3, R = 5, \) and \( m = 10 \) and obtain the throughput for different numbers of SSs with \( d = 6, 8, \) and \( 10 \), respectively. The results are shown in Fig. 6. As \( d \) decreases, the throughput increases. Recall that the throughput is also a measure of the efficiency of the data slots in each uplink subframe. Since the average number of collision-free requests in a request interval is smaller than \( m \) [from (7)], a smaller \( d \) gives higher efficiency. Notice from Fig. 5 that the throughput does not exceed 37.5%. This motivates us to investigate the throughput when \( d \) is set to be 40\% of \( m \). The result is shown in Fig. 7, with \( W = 16, N = 40, \) and \( m \) varying between 5 and 30. It can be seen that a throughput of at least 80\% can be achieved with a relatively small \( p_d \). However, it should be noted that if \( d \) becomes even smaller, \( p_d \) and, thus, packet delay will increase, despite the high throughput value.

Finally, we investigate the delay performance under different system parameters. First, we set \( r = 3, R = 5, \) and \( m = d = 10 \) and plot the mean packet delay against different numbers of SSs with \( W = 8, 16, 32, \) and 64, respectively, in Fig. 8. Again, as expected, for a given \( W \), a larger \( N \) incurs a larger delay, because \( p \) is larger, and each request would need more attempts before being successful. For a fixed \( N \), when \( W \) increases, the average window size in each backoff stage increases, and hence, the mean packet delay increases. Furthermore, as \( W \) increases, a larger range of request slots across several subframes is available to be chosen, resulting in a larger standard deviation of packet delay, as shown in Fig. 9. Although the increase in \( W \) leads to the increase in mean and standard deviation of packet delay, it also means that more requests are successful, and thus, the probability of packet loss due to exceeding the maximum allowable number of attempts decreases, as shown in Fig. 10. Here, the probability of packet loss is expressed as \( p^R \), where \( p \) is given in (2), and \( R \) is the maximum allowable number of attempts per packet. Note from Fig. 10 that by doubling \( W \)
Fig. 10. Probability of packet loss \((r = 3, R = 5, m = 10, \text{ and } d = 10)\).

Fig. 11. Mean packet delay \((r = 3, R = 5, \text{ and } W = 16)\).

from 8 to 16 or from 16 to 32, the probability of packet loss is significantly reduced.

Next, setting \(r = 3, R = 5, W = 16, \text{ and } m = d\), we plot the mean and standard deviation of packet delay against different numbers of SSs with \(m = 10, 15, \text{ and } 20\), respectively, in Figs. 11 and 12. For a fixed \(N\), when \(m\) increases, it is more likely to have a successful request; hence, both the mean and the standard deviation of packet delay decrease. When \(r = 3, R = 5, W = 16, m = 10, \text{ and } d = 6, 8, \text{ and } 10\), we can see from Fig. 13 that the mean packet delay does not significantly vary. This is consistent with the result in Fig. 3 that \(p\) is more or less the same over this range of \(d\). However, for \(d = 4\), \(p\) is larger, and so is the delay.

V. THROUGHPUT OPTIMIZATION

From the results shown in Fig. 5, we have observed that, for fixed \(N, m, d, r, \text{ and } R\), when the contention window \(W\) is varied, there exists an optimal value such that throughput is maximized. This observation has motivated us to find the optimal \(W\) and the corresponding optimal throughput for a given set of system parameters.

In Appendix B, the optimal \(W\) is derived and given by (35). In the special case, where \(r = R - 1\) and \(N\) is large, the optimal \(W\) becomes

\[
W_{\text{opt}} = \frac{(N - m/2)(1-p^R)(1-\alpha p)}{(1-p)(1-(\alpha p)^R)}. \tag{22}
\]

Note that, for a large \(R\), i.e., \(R \to \infty\), the asymptotic value of \(W_{\text{opt}}\) in (22) is

\[
W_{\text{opt}}^{R \to \infty} = \frac{(N - m/2)(1-\alpha p)}{(1-p)}.
\]

In the following, we will investigate how the system parameters affect the optimal \(W\) and throughput values.

Fig. 14 shows the size of the optimal \(W\) versus the number of SSs in the system. For a given number of request slots \((m)\), increasing the number of subscribers increases the optimal \(W\) since that avoids excessive collision. Conversely, for a fixed
number of subscribers \( N \), increasing \( m \) causes the optimal \( W \) to decrease, which avoids excessive idle request slots. These trends are both consistent with the intuition that the throughput of a large contention-based system is maximized when the expected number of attempts per slot is fixed at one [13].

The impact of the truncation value \( r \) defining \( CW_{\text{max}} \) and the maximum allowable attempts \( (R) \) on the optimal \( W \) is shown in Fig. 15. Observe that, for the same number of subscribers and request slots, a smaller \((r, R)\) pair requires a larger \( W \) to maximize the throughput. This is because the optimal attempt probability that maximizes the throughput would be the same for any \((r, R)\) values, provided that the numbers of subscribers and request slots are fixed. Since a smaller \((r, R)\) would yield a smaller average backoff window, which would then result in a greater attempt probability, a larger \( W \) is required to maximize the throughput as compared with that of a larger \((r, R)\).

The impact of various system parameters on the optimal \( W \) and the resulting network performance is illustrated in Table I. It is clear that the optimal throughput only depends on the ratio \( d/m \) but is independent of \( N \). However, the corresponding optimal \( W \) increases with \( N \). Therefore, for a given \( d/m \), when \( N \) is known, \( W \) can be adjusted to the optimal value to achieve the optimal throughput. Since the optimal \( W \) increases with \( N \), the resulting delay also increases. On the other hand, as also shown in the table, when \( N \) and \( m \) are fixed, \( d \) can be set smaller than \( m \) to achieve a higher optimal throughput without greatly affecting the packet delay and loss probability. Finally, the effect of \( d/m \) on the optimal throughput is depicted in Fig. 16.

For a given number of SSs \((N)\), Algorithm 1 will optimize the system performance. In particular, the algorithm adjusts the number of available data slots in the uplink subframe in a small step size during its iterations and is terminated after the packet loss probability \( (p^R) \) exceeds a certain threshold \((\epsilon)\). The loss threshold \( \epsilon \) is a design parameter and can be set according to the actual applications’ requirements. Finally, the algorithm outputs the optimal initial contention window and the corresponding maximum throughput. A similar algorithm can be developed where the delay is considered as a design parameter. This is because, while achieving the optimal throughput, packets may experience different delays using different \((r, R)\) parameter pairs, as described earlier. Note that, in both cases (i.e., loss versus delay), the objective is to adjust the system parameters so that the system performance (e.g., throughput) is optimized based on results of the derivation in this section.

**Algorithm 1 Find \( W_{\text{opt}} \) and maximum throughput**

**Require:** \( N, m, (r, R) \), and \( \epsilon \)

1: \( d = m \) // Initialization
2: Set \( r = 1/N \)
3: Set \( p = 1 - (1 - p_c)(1 - p_d) \), where \( p_c \) and \( p_d \) are in (4) and (7)
4: **while** \( p^R \leq \epsilon \) **do**
5: Set \( d = d - 1 \)
6: Update \( p \) according to (2)
7: **end while**
8: Calculate \( W_{\text{opt}} \) according to (21)
9: Calculate \( \Gamma \) according to (8) // maximum throughput

VI. CONCLUSION

In this paper, we have developed an analytical model to evaluate the performance of BE service in a saturated IEEE 802.16 network. Explicit expressions for the network throughput and the mean and standard deviation of the packet delay are derived. A comparison with simulation confirms the accuracy of our analytical model over a wide range of operating conditions. Using the proposed model, we have studied the impact of various parameters on the overall network performance such as throughput and packet access delay. We have shown that the initial contention window of the broadcast polling mechanism can be suitably chosen to maximize the network throughput. A simple asymptotic value for this optimal window is then provided, assuming the number of allowable retransmissions is large. Furthermore, the maximum achievable throughput is found to be only dependent on the ratio between the number
TABLE I

<table>
<thead>
<tr>
<th>$N$</th>
<th>$m$</th>
<th>$d$</th>
<th>optimal $W$</th>
<th>optimal throughput</th>
<th>mean delay (ms)</th>
<th>std dev of delay (ms)</th>
<th>packet loss prob</th>
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<td>6</td>
<td>23</td>
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<td>6.8110</td>
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<td>3.9719</td>
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<td>3.9693</td>
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<td>6.3821</td>
<td>0.0983</td>
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<td>5.4761</td>
<td>6.3821</td>
<td>0.0983</td>
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</tbody>
</table>

Fig. 16. Optimal throughput against $d/m$ ($N = 50, m = 10, r = 3, \text{ and } R = 5$).

of data slots available on the uplink subframe and the number of request slots for bandwidth requests. We have also provided a mechanism that dynamically adjusts the initial contention window to achieve the optimal performance. Further work will be carried out to extend our model to consider the unsaturated case.

APPENDIX A

UNIQUENESS OF THE FIXED POINT

The proof of the uniqueness of the fixed point uses the following lemma.

Lemma 1: Let $Q(j; \xi)$ be given by (7), and let

$$H(\xi; n) = \sum_{j=1}^{n} jQ(j; \xi) + \sum_{j=n+1}^{m} nQ(j; \xi).$$

Then, for $\xi \in [0, 1]$ and $d = 1, \ldots, m - 1$, we have the following:

1) $H(\xi; d)$ is increasing in $\xi$.
2) $H(\xi; d)/H(\xi; m)$ is decreasing in $\xi$.

Proof: The proof will make repeated use of the fact that, for $a, b, X, X', Y, Y' > 0$

$$\frac{X}{X'} < \frac{Y}{Y'} \iff \frac{X}{aX' + bY'} \leq \frac{aX + bY}{Y'}. \tag{23}$$

To show monotonicity, we will perturb $\xi$ by some amount $\delta \in (0, 1 - \xi]$. We use the notation "↑ in (·)" to indicate that a function is increasing in a specified argument.

Let

$$\bar{F}(j; \xi) = \sum_{i=j}^{m} Q(i; \xi)$$

be the complementary cumulative distribution function corresponding to probabilities $Q$. Then

$$H(\xi; n) = \sum_{j=1}^{n} \bar{F}(j; \xi). \tag{24}$$

We first claim that

$$\frac{\bar{F}(j; \xi + \delta)}{\bar{F}(j; \xi)} \uparrow \text{in } j. \tag{25}$$

To see this, first note that

$$\frac{(\xi + \delta)^{j} (1 - (\xi + \delta))^{m-j}}{(\xi)^{j} (1 - \xi)^{m-j}} \uparrow \text{in } j. \tag{26}$$

This implies that

$$\frac{Q(j; \xi + \delta)}{Q(j; \xi)} < \frac{\bar{F}(j + 1; \xi + \delta)}{\bar{F}(j + 1; \xi)}$$

by backward induction on (23) from $j = m - 1$. Applying (23) to the fact that $\bar{F}(j; \xi) = Q(j; \xi) + \bar{F}(j + 1; \xi)$ gives

$$\frac{\bar{F}(j; \xi + \delta)}{\bar{F}(j; \xi)} < \frac{\bar{F}(j + 1; \xi + \delta)}{\bar{F}(j + 1; \xi)}.$$

This implies (25) and establishes the claim.
We can now prove the first part of the lemma. Note that \( F(0; \xi + \delta) = F(0; \xi) = 1 \). By the foregoing claim
\[
\frac{\hat{F}(j; \xi + \delta)}{\hat{F}(j; \xi)} > \frac{\hat{F}(0; \xi + \delta)}{\hat{F}(0; \xi)} = 1.
\]
Thus, for all \( j > 0 \), \( \hat{F}(j; \xi + \delta) > \hat{F}(j; \xi) \). By (24), this implies that \( H(\xi; d) \uparrow \) in \( \xi \), as required.

To prove the second part of the lemma, it is sufficient to show that
\[
\frac{H(\xi; m)}{H(\xi; d)} = 1 + \sum_{j=d+1}^{m} \frac{\hat{F}(j; \xi)}{\hat{F}(d; \xi)} \uparrow \text{ in } \xi,
\]
(27)

Let
\[
A(\xi) = \sum_{j=1}^{d} \hat{F}(j; \xi),
\]
\[
B(\xi) = \sum_{j=d+1}^{m} \hat{F}(j; \xi).
\]

Then, the foregoing claim and (23) together imply that
\[
\frac{A(\xi + \delta)}{A(\xi)} \leq \frac{\hat{F}(d; \xi + \delta)}{\hat{F}(d; \xi)} < \frac{B(\xi + \delta)}{B(\xi)}
\]
which establishes (27) and, hence, the lemma.

We are now ready to prove the uniqueness of the fixed point.

Proof (Theorem 1): Since \( m \leq N, k = m \) in the following proof. To prove the first part of Theorem 1, invert (5) to give
\[
\hat{\tau}(p_c, p_d) = 1 - (1 - p_c)^{1/(N-1)}.
\]
(28)

For any given \( p_d \), this is continuous on the domain \( p_c \in [0, 1] \), and since
\[
\frac{\partial \hat{\tau}}{\partial p_c} = \frac{1}{N-1}(1 - p_c)^{1/(N-1) - 1} > 0
\]
(28) is monotonic and increases from \( \hat{\tau}(0, p_d) = 0 \) to \( \hat{\tau}(1, p_d) = 1 \). On the other hand, \( \tau(p_c, p_d) \) defined in (4) is continuous on \( p_c \in [0, 1] \) and will now be shown to be monotonically nonincreasing. Rewrite (4) as
\[
\tau(p(p_c, p_d)) = \left( \frac{m+1}{2} + \frac{W}{2} \sum_{i=r}^{R-1} \alpha^i + \alpha^r \sum_{i=r}^{R-1} \beta^i \right)^{-1}
\]
\[
= \left( \frac{m+1}{2} + \frac{W}{2} \sum_{i=0}^{R-1} \beta^i \prod_{j=1}^{i-1} \beta_j \right)^{-1}
\]
(29)

where \( \beta_j = \alpha \) if \( j \leq r \) and 1 otherwise. It suffices to show that \( h(p) := \sum_{i=0}^{R-1} p^i \prod_{j=1}^{i-1} \beta_j / \sum_{i=0}^{R-1} p^i \) is nondecreasing in \( p \in [0, 1] \). Let
\[
A_k = \sum_{i=0}^{k-1} p^i \prod_{j=r}^{R-k-1} \beta_j
\]
\[
B_k = \sum_{i=0}^{k-1} p^i.
\]

The monotonicity of \( h \) will be shown by induction on the following statements: 1) \( A_k/B_k \) is nondecreasing in \( p \); 2) \( A_k - B_k \) is nondecreasing in \( p \); and 3) \( A_k \geq B_k \). The base case \( k = 1 \) is immediate. For the inductive step, assume that the statement is true for some \( k \geq 1 \), and note that \( B_k \) is increasing in \( p \). Then, since \( \beta_{R-k} \geq 1 \), \( A_k - B_k = (1 + \beta_{R-k} p A_k - (1 + p B_k) > 0 \) is increasing in \( p \), which establishes statements 2 and 3 for \( k + 1 \). Moreover
\[
A_{k+1}/B_{k+1} = \frac{1 + \beta_{R-k} p A_k}{1 + p B_k} = \beta - \frac{(1 - \beta)(A_k - B_k)}{1 + p B_k}
\]
which is increasing in \( p \) by statements 2 and 3. This proves statement 1 for \( k + 1 \) and completes the induction. The monotonicity of \( h \) is implied by the case \( k = R \).

The uniqueness of the solution is proven by noting that \( \tau(p(p_c, p_d)) \) in (29) continuously and monotonically decreases from \( \tau(p(0, p_d)) = \tau(p(1, p_d)) \), where
\[
\tau(p(0, p_d)) = \left( \frac{m+1}{2} + \frac{W}{2} \sum_{i=0}^{R-1} p^i \right)^{-1} \hat{\tau}(0, p_d) = 0
\]
\[
\tau(p(1, p_d)) = \left( \frac{m+1}{2} + \frac{W}{2} \sum_{i=0}^{R-1} p^i \right)^{-1} \hat{\tau}(1, p_d) = 1.
\]

To prove the second part of Theorem 1, note that, for any given \( p_d \in [0, 1] \), the previous nonlinear equations (28) and (29) give a unique solution for the two unknowns \( \tau \) and \( p_c \). Denote this solution by \( (\tau^*(p_d); p^*_c(p_d)) \). Observe that both \( \tau^*(p_d) \) and \( p^*_c(p_d) \) are monotonically decreasing functions of \( p_d \in [0, 1] \) since \( \tau(p(p_c, p_d)) \rightarrow \tau(p(p_c, p_d + \delta)) \rightarrow \hat{\tau}(0, p_d) \) for a sufficiently small \( \delta > 0 \).

It remains to prove that, for a given \( d < m \), there is a unique solution to \( f(\tau, p_c, p_d) := g(\tau(p_c, p_d)) - p_d = 0 \), where \( g(\tau(p_c, p_d)) : [0, 1] \rightarrow [0, 1] \) is defined by the right-hand side of (8). Our task is to prove that \( f(\tau, p_c, 0) = 0 \) has a unique solution where the function \( g(\tau(p_c, p_d)) \) is calculated based on the solution \( (\tau^*(p_d); p^*_c(p_d)) \), which is unique for each \( p_d \).

To establish the existence of a solution, observe that \( f(\tau, p_c, 0) \) is continuous and that \( f(\tau, p_c, 0) > 0 \). This is because \( g(p(0, 0)) > 0 \) due to the fact that \( 0 \leq d < m \) and \( \hat{\tau}(0, p_d) \neq (0, 0) \). On the other hand, \( f(\tau, p_c, 1) < 0 \) since \( g(p(0, 1)) < 1 \) for any \( 0 \leq d < m \). The intermediate value theorem now guarantees a solution. Denote this solution by \( (p^*_d(d); \tau^*(p_d); p^*_c(p_d)) \) for a given \( 0 \leq d < m \).

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To prove the uniqueness of the solution \( (p_d^*(d); \tau^*(p_d^*); p_c^*(p_d^*)) \), suppose \( f(\tau^*(1), \tau^*(2), \tau^*(3); p_d^*(1), p_d^*(2), p_d^*(3)) = 0 \) for \( p_d^*(2) > p_d^*(1) \), where the corresponding solutions are \( (p_d^*(1); \tau^*(1); p_c^*(1)) \) and \( (p_d^*(2); \tau^*(2); p_c^*(2)) \), respectively. Because \( \tau^*(p_d) \) is a monotonically decreasing function in \( p_d \), we have \( \tau^*(1) > \tau^*(2) \). However, observe that \( p \) is a function of \( \tau \) and can be written as

\[
p = 1 - (1 - \tau)^N - \sum_{j=0}^{d-1} j Q(j) + \sum_{j=d+1}^{N} j Q(j) \]

where \( H \) is defined in Lemma 1, and \( H(\xi; d) = m \xi \) is the mean of the binomial distribution. To see that \( p \) is increasing in \( \tau \), first consider the case that \( \tau < 1/N \), in which case, \( \partial H(d)/\partial \tau > 0 \). By the second part of Lemma 1, the final factor in (30) is decreasing in \( \xi \) and, hence, decreasing in \( \tau \), whence \( p \) is increasing in \( \tau \). Conversely, if \( \tau \geq 1/N \), then \( \partial H(d)/\partial \tau \leq 0 \), whence, by the first part of Lemma 1, \( H(\xi; d) \) is decreasing in \( \tau \), and \( p \) is again increasing in \( \tau \).

Thus, \( \tau^*(2) < \tau^*(1) \) should correspond to \( p^*(2) < p^*(1) \), which is a contradiction because function \( \tau(p) \) in (29) is a monotonically increasing function. Therefore, \( (p_d^*(1); \tau^*(1); p_c^*(1)) = (p_d^*(2); \tau^*(2); p_c^*(2)) \).

Therefore, \( W_{opt} \) is given by

\[
W_{opt} \approx \frac{2}{\eta} \left( \frac{1}{1 - \alpha p + \alpha^r \sum_{i=1}^{R-1} p^i} \right)
\]

where \( p \) is obtained from (2), (5), and (8) using \( \tau = \tau_{opt} \). By substituting \( \tau_{opt} \) into (35), we obtain (22). Finally, we can obtain the optimal throughput by using (6), (7), and (9).

REFERENCES


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Dr. Andrew is a member of the Association for Computing Machinery. He was a corecipient of the best paper award at the 2007 IEEE International Conference on Mobile Ad Hoc and Sensor Systems.