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Maximizing the stable throughput of high-priority traffic for wireless cyber-physical systems

Qinglin Zhao¹, Taka Sakurai², Jiguo Yu^{3*} and Limin Sun⁴

Abstract

In wireless cyber-physical systems (CPS), more and more traffic with different priorities is required to be timely transmitted over wireless networks such as 802.11. In order to make full use of 802.11 networks to provide quality of service for wireless CPS, we are concerned with the throughput stability issue (i.e., how much traffic load can be sustained in an 802.11 network). Recent studies on the stability in 802.11 networks have arrived at contradictory conclusions. In this paper, we first delve into the reasons behind these contradictions. Our study manifests that the maximum stable throughput is not simply larger than, less than, or equal to the saturation throughput as argued in previous works. Instead, there exist two intervals, over which the maximum stable throughput follows different rules: over one interval, it may be far larger than the saturated throughput; over the other, it is tightly bounded by the saturated throughput. Most existing related research fails to differentiate the two intervals, implying that the derived results are inaccurate or hold true partially. We then point out that for the parameter settings in the 802.11 enhanced distributed channel access (EDCA) standard, high-priority (HP) traffic can achieve a stable throughput far higher than the saturated throughput, according to the rules we find. This indicates that the prior recommendation by other authors, advocating operating a wireless LAN far below the saturation load to achieve stable throughput and avoid unbounded delay, might be too conservative for many settings. We next propose an idle-sense-based scheme to maximize the stable throughput that HP traffic can achieve, when they coexist with low-priority (LP) traffic. Finally, we ran extensive simulations to verify the effectiveness of the revealed rules and the proposed scheme. This study helps utilize the limited bandwidth of wireless networks fully.

Keywords: Cyber-physical systems, Stable throughput, Saturation throughput, 802.11, Idle sense

1 Introduction

Today, wireless cyber-physical systems (CPS) [1–6] have received a great deal of attention, since it provides great convenience in terms of information collection, distribution, and processing and control. In wireless CPS, more and more traffic with different priorities is required to be timely transmitted over wireless networks such as IEEE 802.11 (Wi-Fi), IEEE 802.15.1 (Bluetooth), and IEEE 802.15.4 (ZigBee). This is an extremely challenging task, as the wireless communication characteristics (such as random access, packet losses, time-varying delay, and

limited channel capacity) significantly affect the stability and the performance of wireless CPS. Therefore, much effort has been put to study the achievable capacity or throughput [7–13]. In this paper, we assume that wireless CPS adopts 802.11 protocols for communication and study how to maximize the stable throughput of high-priority traffic.

It has been well known that the stability of 802.11 networks is a notorious problem due to their distributed, random access nature. From Aloha [14] to IEEE 802.11 Wi-Fi networks [15], we cannot even answer a very simple problem: what is the maximum stable throughput (i.e., the network throughput equalling the aggregate input traffic load). For example, even for the simplest network version

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(such as buffered aloha networks), the throughput stability is still in discussion [16]. Therefore, this problem has attracted a great deal of attention such as [7, 8, 16–20].

Recent studies [19–22] considered the throughput stability problem in a one-hop 802.11 distributed coordination function (DCF) network and arrived at contradictory conclusions. In [21], the authors asserted that (a) the maximum stable throughput can only be achieved in the nonsaturation regime (where nodes do not always have packets to transmit) and that (b) it can be much higher than the saturation throughput while providing satisfactory quality of service (QoS). In [22], the authors also observed that the stable throughput may rise higher than the saturation throughput before the network is saturated. However, in [19], the authors argued that to ensure stability, the throughput should not be allowed to exceed the level of the saturation throughput. They recommended operating a DCF network far below the saturation load to achieve stable throughput and to avoid unbounded mean packet delay and delay jitter. Finally, the authors in [20] thought that the saturation throughput can be stable.

In this paper, we investigate this contradictory in general IEEE 802.11 enhanced distributed channel access (EDCA) [15] wireless LANs. 802.11 DCF (that provides a uniform channel contention access) is a special case of 802.11 EDCA (that provides a prioritized channel contention access). Compared with DCF, EDCA can support quality of service for real-time applications and therefore has received continuing attention [23, 24]. In EDCA, nodes belonging to high-priority (HP) access categories (ACs) are configured with a maximum contention window (CW) as small as 16, while nodes belonging to low-priority (LP) ACs are configured with a maximum CW as large as 1024. Such configurations enable HP nodes to enjoy a high opportunity to access the channel. In this paper, we consider an EDCA network with one HP AC and one LP AC (note that when the LP AC does not exist, the EDCA system reduces to the DCF system). Each AC behaves like a DCF network and has two configurable CWs: the minimum and maximum CWs (i.e., W_{\min} and W_{\max}). By default, we assume that $W_{\min} = W_{\max}$; [25] showed that the 802.11 system with such a configuration is similar to the slotted p -persistent CSMA system and can successfully emulate the 802.11 system with $W_{\min} \neq W_{\max}$. However, we also run an experiment (i.e., Experiment 3 in Section 4) to illustrate the performance similarity when $W_{\min} \neq W_{\max}$ and when $W_{\min} = W_{\max}$.

Our study manifests that the maximum stable throughput is not simply larger than, less than, or equal to the saturation throughput as argued in [19–22]. We show that given the node number, there exists a unique optimal HP CW. The HP throughput is only achieved at the optimal HP CW in the saturation regime. The optimal HP CW partitions the whole HP CW range into two intervals (as

shown in Figs. 2 and 4): over one interval, the maximum stable HP throughput may be significantly higher than the saturation throughput while the total packet delay is still very small; over the other, the maximum stable HP throughput is tightly bounded by the saturation throughput. For example, as illustrated in Fig. 4, when $CW = 20$, the simulated maximum stable throughput of 2.12 Mbps is far larger than the saturation throughput of 0.65 Mbps, while the mean total packet delay is only around 2 ms. Note that the assertion (a) in [21] generally does not hold true because this assertion is drawn from simulations based on constant-bit-rate (CBR) traffic. Most existing related research fails to differentiate the two intervals, implying that the derived results are inaccurate or hold true partially.

We then apply the revealed rules to study the CW settings of 802.11 EDCA, where HP ACs are configured with a CW as small as 16, while LP ACs are configured with a CW as large as 1024. We point out that such CW settings will enable an HP AC to achieve a stable throughput far higher than the saturated throughput. This indicates that the prior recommendation by other authors, advocating operating a wireless LAN far below the saturation load to achieve stable throughput and avoid unbounded delay, might be too conservative for many settings. Furthermore, we next propose an idle-sense-based scheme [26] to maximize the stable throughput that HP traffic can achieve, when they coexist with LP traffic.

The rest of this paper is organized as follows. Section 2 models the exact and asymptotic HP throughput. Section 3 investigates the maximum stable HP throughput. Section 4 illustrates the maximum stable HP throughput and verifies our augments via simulation. Section 5 concludes this paper.

2 HP throughput

The considered EDCA system, running in the basic mode and ideal channel conditions, consists of one HP AC and one LP AC. Each node has an infinite buffer size. All data packets from HP and LP nodes are transmitted to the AP, and the AP acts purely as the receiver of data packets.

The LP AC has n_0 nodes. Each LP node has the same packet size L_0 and always generates a random backoff count uniformly distributed in $[0, W_0]$ for each new transmission or retransmission, where $W_0 > 1$. We assume that each LP node is in saturation operation (i.e., the node always has packets to transmit) because here we study the maximum stable throughput that the HP AC can achieve, regardless of how the LP offered load varies.

The HP AC has n nodes. Each HP node has the same packet size L and packet arrival rate λ and always generates a random backoff count uniformly distributed in $[0, W]$ for each new transmission or retransmission, where $W > 1$.

When $n_0 = 0$, the considered EDCA system reduces to a DCF system without a backoff mechanism.

2.1 Exact HP throughput

We now express the total throughput of HP nodes.

Let $\beta_0 \in (0, 1)$ be the saturated attempt rate (i.e., the number of transmission attempts per slot) for each LP node and then we have $\beta_0 = 2/(W_0 + 1)$ from (8) in [27]; note that when $W_{\min} \neq W_{\max}$, the saturated attempt rate is given by (7) in [27]. Let $C_0 = (1 - \beta_0)^{n_0}$ be the probability that none of the n_0 LP nodes transmits packets.

Let $\beta \in (0, 1)$ be the general (i.e., saturated or nonsaturated) attempt rate for each HP node. Let Ω be the mean time that elapses for one decrement of the backoff counter. Then, we have

$$\Omega = P_e \sigma + P_b T_b + P_{b_0} T_{b_0} + P_c T_c, \quad (1)$$

where $P_e = (1 - \beta)^n C_0$,

$$P_b = [1 - (1 - \beta)^n] C_0,$$

$$P_{b_0} = (1 - \beta)^n [1 - C_0],$$

$$P_c = 1 - P_e - P_b - P_{b_0}.$$

In (1), P_e is the probability of an idle slot; P_b (P_{b_0}) is the probability that at least one of the n HP (n_0 LP) nodes transmits when none of the n_0 LP (n HP) nodes transmits packets; P_c is the probability of a collision involving both HP and LP nodes. σ is the length of one idle MAC slot and is set to 20 ms in IEEE 802.11b. T_b (T_{b_0}) $\gg \sigma$ is the mean time of a successful transmission for each HP (LP) node and is assumed to equal the mean time of a collision involving HP (LP) nodes only; we adopt this assumption for simplifying the analysis, and this assumption can be removed easily [28]; T_b (T_{b_0}) can be calculated by $T_{\text{pkt}}(L)$ in Table 1, where L denotes the packet size of HP or LP nodes. $T_c (= \max(T_b, T_{b_0}))$ is the mean time for a collision involving both HP and LP nodes.

2.1.1 Exact HP throughput

Given the HP node number n and the general HP attempt rate β , the total HP throughput, $\Gamma(n, \beta)$, is defined to be the number of bits transmitted successfully by HP nodes in the time interval of Ω . Then, $\Gamma(n, \beta)$ is expressed as

Table 1 Parameters for 802.11b basic mode

SIFS	10 μ s	IP	= 40 bytes
Slot	20 ms	MAC	= 30 bytes
DIFS	50 μ s	T_{ack}	= (PHY + 14) \times 8/ R_{basic}
R_{data}	11 Mbps	$T_{\text{data}}(L)$	= PHY \times 8/ R_{basic} + (IP + MAC + L) \times 8/ R_{data}
R_{basic}	1 Mbps	$T_{\text{pkt}}(L)$	= DIFS + $T_{\text{data}}(L)$ + SIFS + T_{ack}
PHY	24 bytes		

$$\Gamma(n, \beta) = \frac{P_s L}{\Omega}, \quad (2)$$

$$P_s = n\beta(1 - \beta)^{n-1} C_0$$

where P_s is the probability of a successful transmission from any of the n HP nodes, when none of the n_0 LP nodes transmits packets.

2.1.2 Saturated HP throughput

Let β_s be the saturated HP attempt rate, which is equal to $\beta_s = 2/(W + 1)$ from (8) in [27]. We then call $\Gamma(n, \beta_s)$ the saturated HP throughput.

2.2 Asymptotic HP throughput

We call $k \triangleq \lim_{n \rightarrow \infty} n\beta \in [0, \infty)$ the asymptotic total HP attempt rate and define a constant η as follows:

$$\eta \triangleq -\frac{(T_{b_0} - T_c) + C_0(\sigma - T_b - T_{b_0} + T_c)}{T_c + C_0(T_b - T_c)}. \quad (3)$$

2.2.1 Asymptotic HP throughput

Let $\Gamma(k) \triangleq \lim_{n \rightarrow \infty} \Gamma(n, \beta)$ be the asymptotic HP throughput. Theorem 1 below expresses $\Gamma(k)$ and shows that $\Gamma(k)$ has a unique maximum value $\Gamma(k_{\text{opt}})$, where $k_{\text{opt}} = \arg \max_{k \in [0, \infty)} \Gamma(k)$ is called the optimal total HP attempt rate.

Theorem 1. (a) $\Gamma(k) \geq 0$ is continuous in $[0, \infty)$ and is given by

$$\Gamma(k) = \frac{k}{e^k - \eta} \frac{C_0 L}{T_c + C_0(T_b - T_c)}, \quad (4)$$

where $\eta \in (0, 1)$.

(b) $\Gamma(k)$ is increasing in $[0, k_{\text{opt}}]$ and is decreasing (k_{opt}, ∞) , where

$$k_{\text{opt}} = \mathcal{W}_0\left(\frac{-\eta}{e}\right) + 1, \quad (5)$$

where $\mathcal{W}_0(\cdot)$ is one branch of the Lambert $W(z)$ function [29], $W(z)e^{W(z)} = z$, for any complex number z .

Proof. Please refer to the Appendix. \square

In general, $\Gamma(k)$ can well approximate $\Gamma(n, \beta)$ as shown in Fig. 1. Hereafter, we use $\Gamma(k)$ as a theoretical proxy for $\Gamma(n, \beta)$. The dashed circle curve in Fig. 1 illustrates $\Gamma(k)$. Note that (i) when $k = k_s \triangleq n\beta_s$, we call $\Gamma(k_s)$ the asymptotic saturated HP throughput and have $\Gamma(k_s) \leq \Gamma(k_{\text{opt}})$ obviously, and (ii) when $C_0 = 1$ and $T_b = T_{b_0}$ (i.e., no LP nodes exist), k_{opt} reduces to the solution to (10) in [26].

3 Maximum stable HP throughput

Let us define the stable HP throughput first.

Stable HP throughput: The HP throughput $\Gamma(k)$ is said to be stable if for a given aggregate input traffic load, $n\lambda L$, (a) there exists a theoretical $k > 0$ so that the total

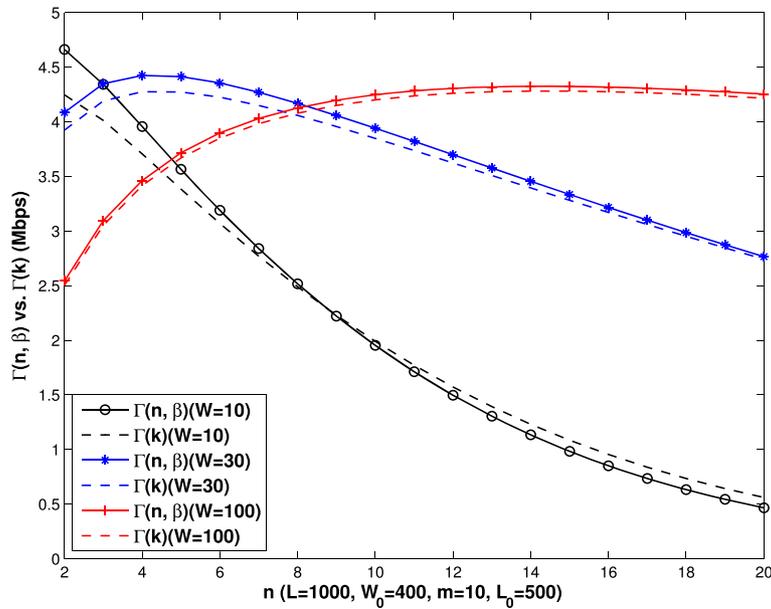


Fig. 1 $\Gamma(n, \beta)$ and $\Gamma(k)$ as n and W vary, where $k = n\beta$ and $\beta = 1/(W + 1)$

HP throughput $\Gamma(k) = n\lambda L$ and (b) the theoretical k is achievable, namely, all HP nodes are able to jointly and spontaneously tune their respective CWs to produce such a k .

Remarks. (i) The statement that k is achievable implies $\Gamma(k) = n\lambda L$, but the converse is not true; the difference is that the k in the original statement represents a realistic value while the k in the converse statement would represent a theoretical value. (ii) We give an example where k is unachievable. Consider the settings $n = 50$, $W = 5$, and $L = 1000$, as shown in Fig. 2. We have $k_s = 2n/(W + 1) \approx 16.7 > k_{opt} = 0.2866$. Theoretically, k may assume the value k_{opt} since $k \in [0, k_s] = [0, k_{opt}] \cup (k_{opt}, k_s]$. However, $k = k_{opt}$ is unachievable because the throughput is increasing in $k \in [0, k_{opt}]$ and the simulated maximum stable throughput is about 2.18 Mbps, which is far less than 4.3 Mbps corresponding to $k = k_{opt}$. (iii) The statement that k is achievable implies that the nodes can produce such a k , but not vice versa, because producing a k (say, $k = 0.2866$) possibly requires that the offered load should be far larger than the throughput $\Gamma(k)$. \square

Proposition 1 below presents a conjecture about the achievable interval of k .

Proposition 1. There exists a $k_{max} \in (0, k_{opt}] \cap (0, k_s]$ so that $k \in [0, k_{max}]$ is achievable, $k = k_{max}$ is not necessarily achievable, and $k \in (k_{max}, k_s]$ is unachievable.

From Proposition 1, $\Gamma(k_{max})$ is a tight upper bound on the stable HP throughput, namely, any traffic load below $\Gamma(k_{max})$ is stable. Note that (i) the throughput $\Gamma(k_{max})$ might be unstable, meaning that we possibly need to inject a much higher traffic load than $\Gamma(k_{max})$ in order to attain $\Gamma(k_{max})$; (ii) $\Gamma(k_{max}) \leq \Gamma(k_{opt})$ and $\Gamma(k_{max})$ is different from $\Gamma(k_s)$, but they might be equal sometimes, for example, when $(0, k_s] \subset (0, k_{opt}]$, which will be explained later. In the following, we investigate $\Gamma(k_{max})$ when the HP CW is either statically configured or dynamically adjustable.

3.1 Maximum stable HP throughput for static HP CW configurations

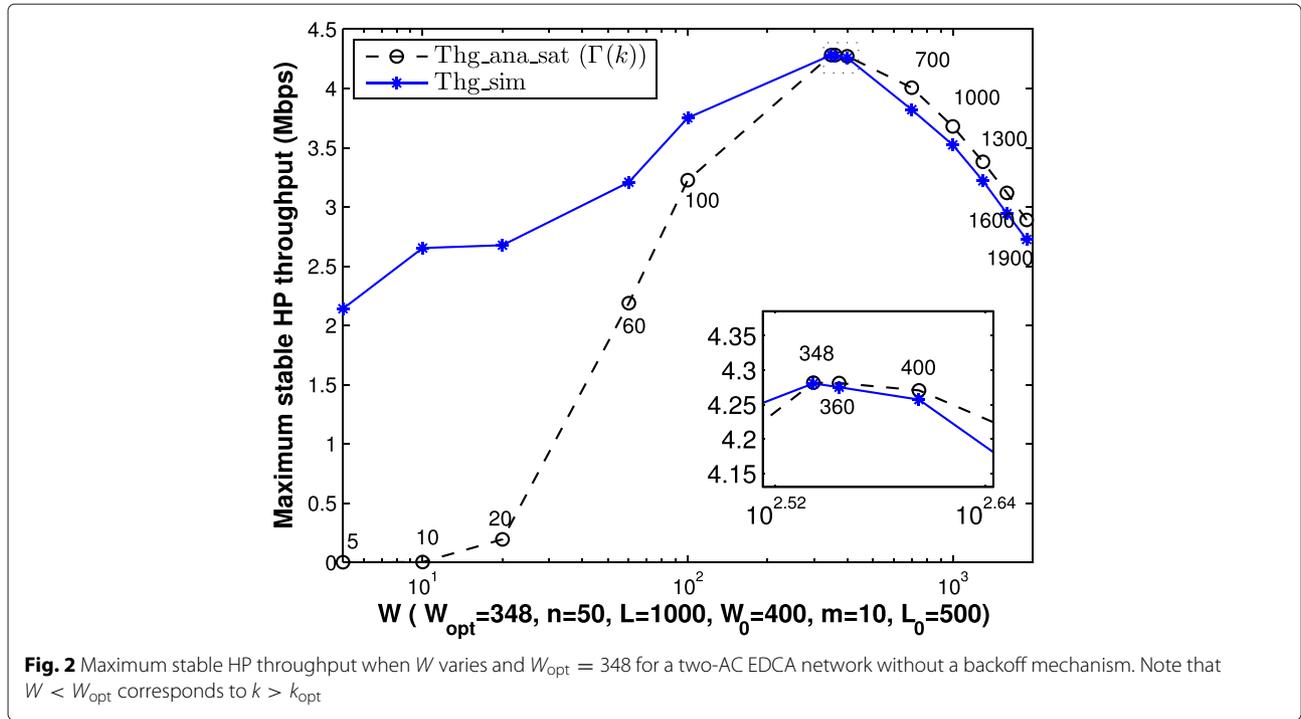
From Theorem 1, given n , there exists an optimal attempt rate, $\beta_{opt} = k_{opt}/n$, where k_{opt} is given by (5). Further, we can calculate the optimal CW, W_{opt} , as follows.

$$W_{opt} = \left\lceil \frac{2}{\beta_{opt}} - 1 \right\rceil = \left\lceil \frac{2n}{k_{opt}} - 1 \right\rceil, \quad (6)$$

where $\lceil \cdot \rceil$ is the round function.

We point out that W_{opt} partitions the whole contention window range into two intervals: $[1, W_{opt})$ and (W_{opt}, ∞) ; for $W \in [1, W_{opt})$, the maximum stable HP throughput $\Gamma(k_{max})$ may be significantly higher than the saturation throughput, while for (W_{opt}, ∞) , the maximum stable HP throughput $\Gamma(k_{max})$ is tightly bounded by the saturation throughput.

We now explain our arguments as follows. Given n and W , the per-node and total saturated attempt rates are $\beta_s = 2/(W + 1)$ and $k_s = n\beta_s$, respectively. $\Gamma(k_{max})$ varies depending on the relationship between k_s and k_{opt} .



3.1.1 The case of $k_s \leq k_{opt}$

When $k_s \leq k_{opt}$, we conjecture $\Gamma(k_{max}) = \Gamma(k_s)$ because (i) $\Gamma(k) \leq \Gamma(k_s)$ for any $k \leq k_s$ from Theorem 1 (b) and (ii) any HP node can really transmit packets at the attempt rate of β_s in saturated operation.

3.1.2 The case of $k_s > k_{opt}$

When $k_s > k_{opt}$, there exists $k_0 \in [0, k_{opt}]$ such that $\Gamma(k_0) = \Gamma(k_s)$ from the intermediate value theorem, since $\Gamma(k) \geq 0$ is continuous and $0 \leq \Gamma(k_s) \leq \Gamma(k_{opt})$. Consequently, we have $\Gamma(k) \geq \Gamma(k_0) = \Gamma(k_s)$ for any $k \in [k_0, k_{opt}] \subset [0, k_s]$ since $\Gamma(k)$ is increasing over $[k_0, k_{opt}]$.

Consider the possible situation where $k_{max} \in [k_0, k_{opt}]$. This would imply that $\Gamma(k_{max}) \geq \Gamma(k_s)$ and that $\Gamma(k_{max})$ would appear before the saturation operation since $k_{max} \leq k_s$. Such a phenomenon has been observed in Fig. 5 in our previous paper [28], and such a maximum throughput is called the “presaturation throughput peak” in [19].

However, contrary to the opinion expressed in [19] that such a presaturation throughput peak might not be sustainable and therefore the traffic load should be maintained far below the saturation load for DCF with a backoff mechanism, our simulation results show that for some CW settings, such a presaturation throughput peak is sustainable. Furthermore, it can be far above the saturation load, while the total packet delay can be very low. For example, as illustrated in Fig. 4, when $CW = 20$ (which is a case of $k_s > k_{opt}$), the simulated maximum stable throughput = 2.12 Mbps is far larger than the saturation

throughput of 0.65 Mbps, while the mean total packet delay is only about 2 ms.

An intuitive explanation is as follows. A very large attempt rate such as that required for saturated operation can cause too many collisions and lead to a reduced throughput. By limiting the attempt rate, we can potentially achieve a higher throughput than the saturated throughput, while maintaining stability. Nevertheless, finding $\Gamma(k_{max})$ is a challenging task in the case when $k_s > k_{opt}$ and we leave this topic for future research.

3.2 Maximum stable HP throughput for adjustable HP CW configurations

If the HP CW W is dynamically adjustable, we can make the maximum total HP attempt rate converge to the optimal rate k_{opt} , then $\Gamma(k_{max}) = \Gamma(k_{opt})$, thereby maximizing the stable HP throughput.

To this end, we envisage two possible schemes to attain the optimal throughput. The first scheme is a centralized scheme. In this scheme, whenever an existing HP node leaves or a new HP node joins the network, the AP updates the number of HP nodes, n , and then broadcasts it. Upon receiving the updated value of n , each HP node sets $W = W_{opt}$, where W_{opt} is given by (6).

The second scheme is a distributed scheme based on the idle sense idea in [26]. The core of the idle sense idea is an AIMD algorithm, which has the property of converging to equal values of the control variable [30]. This distributed scheme can automatically adapt to the change

in the number of HP nodes and therefore requires less control overhead than the centralized scheme. In the original idle sense scheme, all nodes have the same priority. Each node constantly estimates the mean number of consecutive idle slots between two transmission attempts and then compares the estimated mean number with an optimal mean number (maximizing the system throughput). From the comparison result, each node either increases or decreases its CW so as to make the mean number converge to the optimal value, consequently approximating the maximum throughput. Unlike the original idle sense scheme, our application aims to maximize the HP throughput and therefore we apply the original scheme to each HP node only.

Let θ_{opt} represent the optimal mean number of consecutive idle slots between two transmission attempts observed by a HP node. Note that the number of consecutive idle slots follows a Geometric distribution with parameter $1 - P_e$ and $\lim_{n \rightarrow \infty} P_e = e^{-k} C_0$. We have

$$\theta_{\text{opt}} = \lim_{n \rightarrow \infty, k=k_{\text{opt}}} \frac{P_e}{1 - P_e} = \frac{C_0}{e^{k_{\text{opt}}} - C_0}. \quad (7)$$

Again, when $C_0 = 1$, θ_{opt} reduces to (12) in [26].

We call the second scheme the idle-sense-based scheme, which is presented below.

3.2.1 The idle-sense-based scheme

Each HP node runs the idle sense scheme in Fig. 6 in [26], except replacing the calculation method of the optimal mean number (i.e., (12) in [26]) by θ_{opt} in (7).

4 Model verification

In this section, we validate the effectiveness of the prediction of the maximum stable HP throughput. We use the TU-Berlin 802.11e simulator (http://www2.tkn.tu-berlin.de/research/802.11e_ns2/) in ns2 version 2.28 as a validation tool. In the 802.11e simulator, we disable the binary exponential backoff algorithm by letting the maximum CW be equal to the minimum CW (i.e., $W_{\text{max}} = W_{\text{min}}$), set the retry limit to 7 (setting a larger retry limit, say 100, just produces a negligible impact on the simulation results by our experiments), and use the DumbAgent routing protocol, whose header is 40 bytes. We set the buffer size to 1000 packets, which is used to mimic an infinite buffer. The other protocol parameter settings are listed in Table 1. In addition, we set $W_0 = 400$, $n_0 = 10$, and $L_0 = 500$ bytes for the LP AC unless otherwise specified.

The target of the simulation is to obtain the maximum stable throughput as the CW varies. In simulation, given the input traffic load $n\lambda L$, let Γ_{sim} denote the obtained throughput. Γ_{sim} is said to be stable if the error between Γ_{sim} and $n\lambda L$ is less than 1 %, namely,

$$\frac{|\Gamma_{\text{sim}} - n\lambda L|}{n\lambda L} < 1\%. \quad (8)$$

For each simulation value, the running time is 200 s when $k \leq k_{\text{opt}}$, whereas it is 1000 s otherwise to exclude the phenomenon that the system once evolves to saturation operation and will never get out of it again, as explained in Fig. 6 in [19]. Note, however, that we do not observe a distinct change in simulation results when the simulation time is set to 1000 s, in comparison with 200 s. In addition, for readability, we only plot the theoretical saturated throughput in Figs. 2 and 4, but its accuracy has been widely validated in [27, 28, 31].

We ran four experiments to verify our augments. We now explain the four experiments.

Experiment 1 to illustrate the error between $\Gamma(k)$ and $\Gamma(n, \beta)$: In the first experiment, we demonstrate that the asymptotic throughput $\Gamma(k)$ can well approximate the exact throughput $\Gamma(n, \beta)$. Theoretically, the approximation condition is $n \gg \beta$. In practice, the approximation is already good when n is moderately larger than β , which is readily satisfied. Figure 1 plots $\Gamma(k)$ and $\Gamma(n, \beta)$ for a two-AC wireless LAN when $n = 2, 3, \dots, 20$, $W = 10, 30, 100$ and $L = 1000$ bytes, where $k = n\beta = 2n/(W + 1)$. From this figure, we can see that the $\Gamma(k)$ curve closely matches the $\Gamma(n, \beta)$ curve for each W even when $n = 2$. For example, for $n = 2$, $\beta = 0.1818, 0.0645$, and 0.0198 , and the approximation error = 9, 4, and 1.5 % when $W = 10, 30, 100$, respectively. This indicates that (i) the approximation condition is not restrictive and (ii) the approximation accuracy increases as W increases.

Experiment 2 to illustrate the relationship between the maximum stable throughput and the saturation throughput: In the second experiment, we consider Poisson arrivals and demonstrate the maximum stable HP throughput and the mean total packet delay when the HP CW is statically configured. In this experiment, we set $n = 50$ and $L = 1000$ bytes. The optimal total HP attempt rate k_{opt} is 0.2866 and therefore the optimal per-node HP CW is $W_{\text{opt}} = 348$. We ran this experiment for the EDCA and DCF systems; Figs. 2 and 4, respectively, explain our observations.

Figure 2 plots the saturated HP throughput and the simulated maximum stable HP throughput, when $W = 5, 10, 20, 60, 100, 348, 360, 400, 700, 1000, 1300, 1600$, and 1900. In two subfigures, we plot W on a log-scale for readability, but we label the corresponding W value in linear units near the curve points. Note that since $k_s = 2n/(W + 1)$, $W < W_{\text{opt}}$ implies $k_s > k_{\text{opt}}$ and, conversely, $W > W_{\text{opt}}$ implies $k_s < k_{\text{opt}}$. We now explain how the experiment was conducted and discuss its result.

- When $W \geq 348$, for each W , we increase the input offered load according to $(0.9 + j0.05)\Gamma(k)$ Mbps as j increases from 1 to 3 and then find the maximum

stable throughput and the corresponding total delay. From the figure, we see that when W increases from 348 to 1900, the simulated maximum stable HP throughput decreases while the corresponding delay increases quickly. The simulation curve of the maximum stable throughput is slightly below the theoretical curve of the saturation throughput, confirming that the saturation throughput is a tight upper bound on the stable throughput.

- When $W < 348$, for each W , we increase the input offered load according to $j\Gamma(k_{opt})/8$ Mbps as j increases from 1 to 8, and then find the maximum stable throughput and the corresponding total delay. From the figure, we see that when W increases from 5 to 100, the simulated throughput increases from 2.1 to 3.8 Mbps while the corresponding delay increases from 2 to 16 ms. In contrast, the theoretical saturation throughput underestimates the simulation result, especially when W is small. For example, when $W = 20$, the predicted throughput value is 0.19 Mbps while the simulated throughput is 2.68 Mbps and the simulated delay is 3.8 ms only. This observation suggests that for some CW settings, the recommendation in [19] calling for operating a wireless LAN far below the saturation load might be too conservative.

The recommended settings for voice transmission in EDCA are backoff factor = 2, $W_{min} = 8$, and $W_{max} = 16$. As a result, the corresponding mean CW is less than

20. Even if we were to adopt such W_{min} and W_{max} settings (including a backoff mechanism) for HP nodes in our simulations, we can safely deduce that the saturation throughput will still significantly underestimate the maximum stable throughput, from the huge gap between the simulated stable throughput and the theoretical saturation throughput when $W \leq 20$ in Fig. 2.

Figure 3 plots the simulated total packet delay, corresponding to the simulated maximum stable HP throughput in Fig. 2. From this figure, we have the following observations.

- When $W \geq 348$, the total delay increases as W increases. There are two reasons. Reason 1: a larger W will cause a larger contention delay, thereby increasing the total delay. Reason 2, as explained in Fig. 2, over this interval, the maximum stable throughput approximates the saturated throughput; it indicates that the traffic load is heavy and therefore more collisions will occur, thereby increasing the total delay as well.
- When $W < 348$, the total delay is low. Particularly, when W is far less than 348, the total delay is much low. For example, when W decreases from 100 to 5, the total delay decreases from tens of milliseconds to several milliseconds. The main reason is, as explained in Fig. 2, over this interval, the system traffic load is very light and is far less than the saturated load; therefore, each arrived packet can be delivered to the receiver quickly, leading to a much low delay.

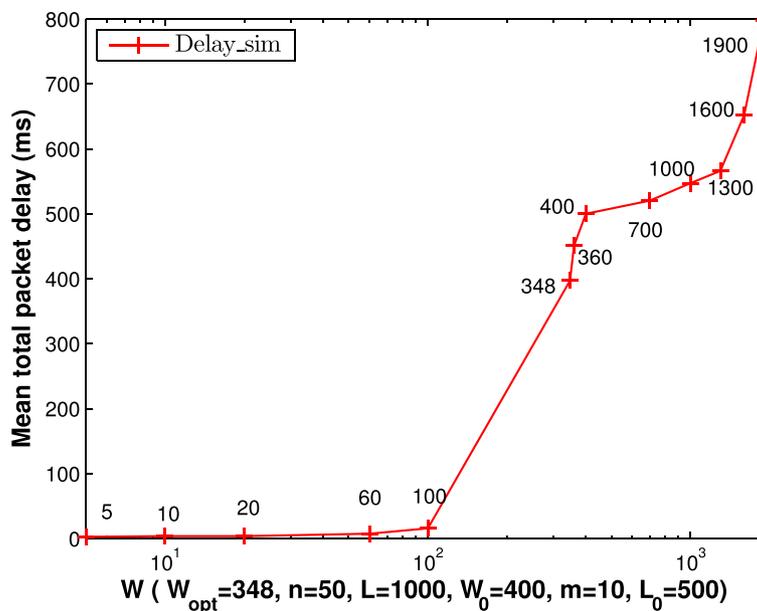


Fig. 3 Total packet delay in ms when W varies and $W_{opt} = 348$ for a two-AC EDCA network without a backoff mechanism. Note that $W < W_{opt}$ corresponds to $k > k_{opt}$

Figures 4 and 5, respectively, repeats Figs. 2 and 3, except that only the HP AC exists (i.e., it is the DCF system without a backoff mechanism), $n = 30$, $L = 500$ bytes, $k_{opt} = 0.1904$, and $W_{opt} = 315$. As expected, the results observed in the DCF system are very similar to those in the two-AC system, when $k_s \leq k_{opt}$ and $k_s > k_{opt}$.

Experiment 3 to illustrate the performance similarity when $W_{min} \neq W_{max}$ and when $W_{min} = W_{max}$: To illustrate that a DCF system with $W_{min} \neq W_{max}$ exhibits throughput and delay properties that are very similar to that with $W_{min} = W_{max}$, we consider the default EDCA CW configuration for voice traffic, namely, $[W_{min}, W_{max}] = \gamma[8, 16]$. We map the CW configuration with $W_{min} \neq W_{max}$ to that with $W_{min} = W_{max} \triangleq \lceil 2/\beta_s - 1 \rceil$, where β_s is the saturated attempt rate for $W_{min} \neq W_{max}$. For example, when $n = 30$, $L = 500$ bytes, we have the CW mapping: $[W_{min}, W_{max}] = [8, 16] \rightarrow [W_{min}, W_{max}] = [13, 13]$ and $k_s = n\beta_s = 30 \times 2/(13 + 1) = 4.2857 \gg k_{opt} = 0.1904$. We then consider Poisson traffic and run simulations to obtain the following simulation results. Table 2 compares the maximum stable throughput, the saturated throughput, and the total delay before and after mapping and shows that the mapping pair has very similar performance. From this table, when $W_{min} \neq W_{max}$, the corresponding simulation results are respectively 1.9731 and 0.2038 Mbps, 1.7398 ms; when $W_{min} = W_{max}$, they are respectively 1.9751 Mbps, 0.2042 Mbps, and 1.9743 ms. This comparison indicates that like $W_{min} = W_{max}$, when k_s is sufficiently greater

than k_{opt} , the DCF system when $W_{min} \neq W_{max}$ may achieve a stable throughput far higher than the saturated throughput, while providing satisfactory QoS.

Experiment 4 to verify the effectiveness of the proposed idle-sense-based scheme: The fourth experiment considers CBR voice traffic and determines the maximum number of admitted HP nodes, n_{max} , that ensures throughput stability, when the idle-sense-based scheme is used, and for the static cases of $W = 300$ and 20. We take the theoretical value of n_{max} to be $\lfloor \Gamma(k_{opt})/(\lambda L) \rfloor$ for the adjustable HP CW case and $\lfloor \Gamma(k_s)/(\lambda L) \rfloor$ for the static HP CW case. The simulation value of n_{max} is the maximum number of HP nodes satisfying (8). In this experiment, we calculate n_{max} when all HP nodes are assigned, alternately, the G.711-100, G.711-50, iLBC, G.729, and G.723a voice codecs (which are shown in Table 3 and are from Table IV (b) in [31]), where each type of voice traffic is modeled as CBR traffic. The theoretical and simulation values of n_{max} are listed in Table 4. We now explain the obtained results.

- In the case of the idle-sense-based scheme, Table 4 shows that each simulation value of n_{max} is exactly equal to its theoretical value (except for the G.711-50 voice codec) and is the maximum (except for the G.723a voice codec) when compared with the cases of $W = 300$ or $W = 20$. At the same time, we also observe that the total delay for each voice codec is less than 25 ms. This manifests that (i) the idle-sense-based scheme can maximize the stable HP

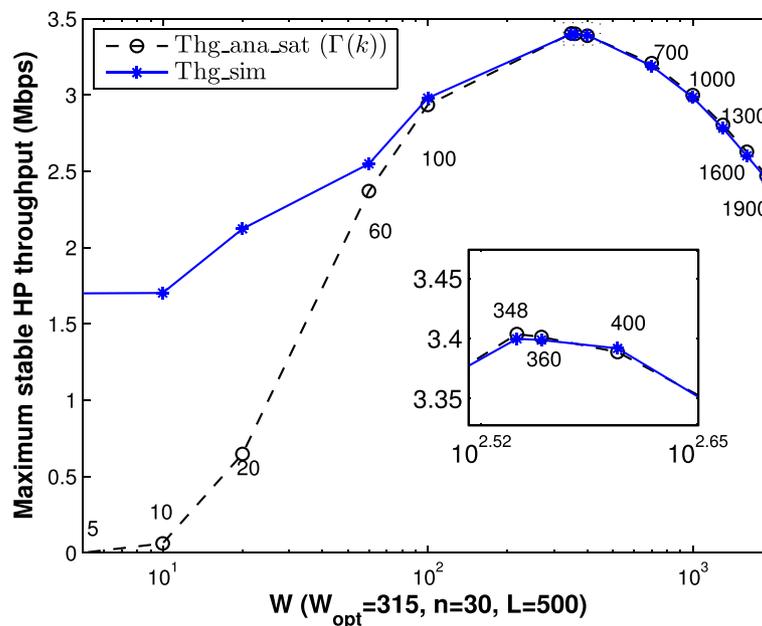


Fig. 4 Maximum stable HP throughput when W varies and $W_{opt} = 315$ for a DCF network without a backoff mechanism. Note that $W < W_{opt}$ corresponds to $k > k_{opt}$

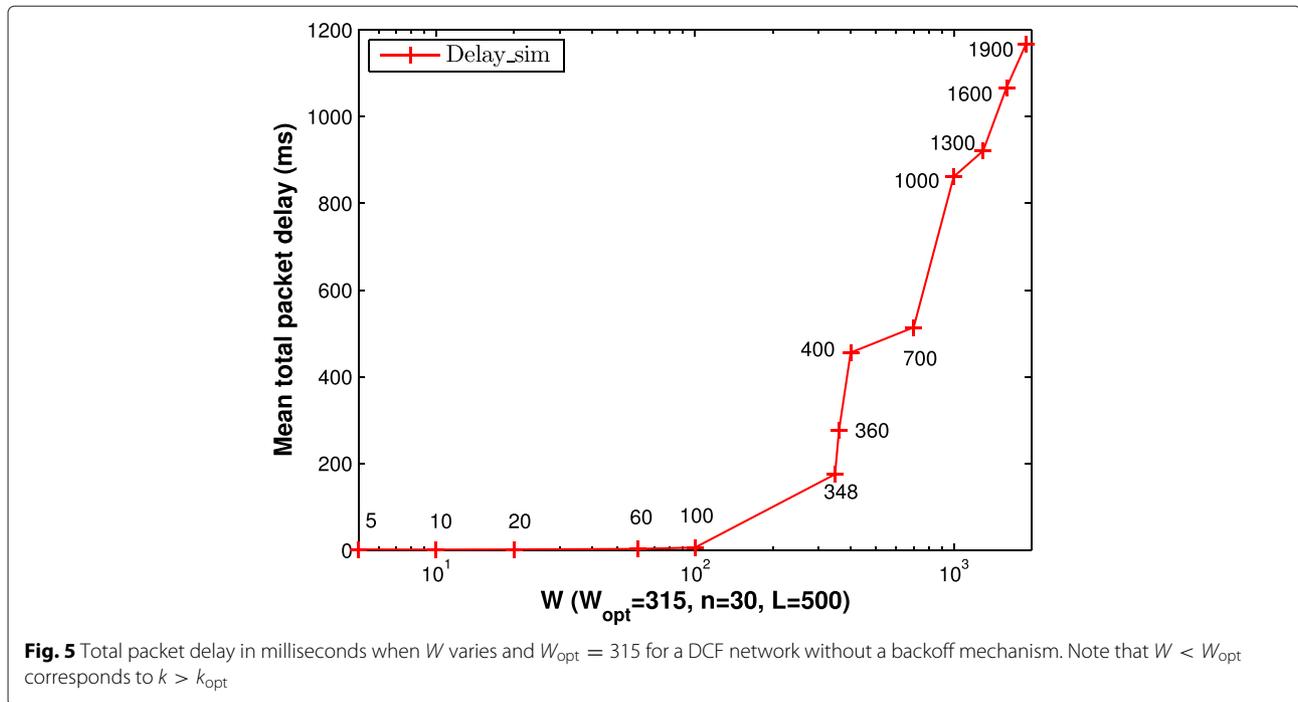


Fig. 5 Total packet delay in milliseconds when W varies and $W_{opt} = 315$ for a DCF network without a backoff mechanism. Note that $W < W_{opt}$ corresponds to $k > k_{opt}$

throughput while maintaining a low total delay and (ii) $\Gamma(k_{opt})$ is a tight upper bound on the stable throughput for the idle-sense-based scheme.

- For the case of $W = 300$, we have $W \geq W_{opt}$. From Table 4, the prediction error of n_{max} between the theoretical and simulation results for each voice codec is at most one. This indicates that $\Gamma(k_s)$ is a tight upper bound on the stable throughput, as expected in this case. However, the total delay for each voice codec except G.711-50 is significantly larger than 25 ms.
- For the case of $W = 20$, we have $W < W_{opt}$. From Table 4, the prediction of n_{max} is reasonably accurate for the G.711-100 and G.711-50 voice codecs since $W = 20$ approximates their respective optimal CWs, but the prediction is far less accurate for the iLBC, G.729, and G.723a voice codecs since $W = 20$ is far less than their respective optimal CWs. This indicates that the saturation throughput might significantly underestimate the maximum stable throughput, as expected in this case. Comparing the case of $W = 300$

with the cases of $W = 20$, we can see that each simulation value of n_{max} in the latter is reasonably less than that in the former. However, the total delay for each voice codec is less than 25 ms in the latter and hence is significantly less than that in the former. This indicates that it is attractive to set a small CW value in order to obtain a satisfactory quality of service.

5 Conclusions

In wireless cyber-physical systems, more and more data will be transmitted over wireless networks such as 802.11 networks. To fully utilize the bandwidth of 802.11 networks for supporting wireless CPS applications, in this paper, we study the maximum stable throughput in 802.11 networks. We reveal that the maximum stable throughput follows different rules in two partitioned intervals and point out that failing to differentiate the two intervals is the fundamental reasons causing the existing contradictory conclusions. Applying the rules, we assert that for the parameter settings in 802.11 EDCA standard, high-priority traffic can achieve a stable throughput far higher than the saturation throughput, which contradicts

Table 2 Performance comparison between IEEE 802.11 standard settings and the mapped CW configurations when $n = 30$ and $L = 500$

$[W_{min}, W_{max}]$	[8, 16]	[13, 13]
Maximum stable throughput (Mbps)	1.9731	1.9751
Saturation throughput (Mbps)	0.2041	0.2041
Total delay (ms)	1.7398	1.9743

Table 3 Parameters for voice packets

	G.711-100	G.711-50	iLBC	G.729	G.723a
λ (pkts/s)	100	50	33.33	25	16.67
L (byte)	80	160	50	40	48

Table 4 Comparison of n_{\max} between the theoretical results and the simulation results for voice codecs shown in Table 3

n_{\max}		G.711-100	G.711-50	iLBC	G.729	G.723a
Idle_sense	sim	9	18	29	39	58
	ana	9	17	29	39	58
Fixed_CW ($W = 300 > W_{\text{opt}}$)	sim	0	12	26	38	59
	ana	0	11	25	38	58
Fixed_CW ($W = 20 < W_{\text{opt}}$)	sim	9	16	25	32	47
	ana	8	12	16	18	22

the existing augments. Furthermore, we propose a CW-adjustable scheme to maximize the stable throughput of high-priority traffic. This study provides new insights on the unsaturation performance and helps utilize the limited bandwidth of wireless networks fully.

Appendix

Proof of Theorem 1. We first prove that $0 < \eta < 1$. Note that

$$\begin{aligned} T_c &= \max(T_b, T_{b_0}) \gg \sigma, \\ 0 &< C_0 < 1. \end{aligned}$$

If $T_b \geq T_{b_0}$, we have

$$\begin{aligned} T_c &= T_b \gg \sigma, T_{b_0} \gg \sigma, \\ \eta &= -\frac{(T_{b_0} - T_c) + C_0(\sigma - T_b - T_{b_0} + T_c)}{T_c + C_0(T_b - T_c)} \\ &= -\frac{(T_{b_0} - T_b) + C_0(\sigma - T_{b_0})}{T_b} \\ &= \frac{(T_b - T_{b_0}) + C_0(T_{b_0} - \sigma)}{T_b}. \end{aligned}$$

On one hand, since $T_b \geq T_{b_0}$ and $T_{b_0} \gg \sigma$, we have

$$\eta > 0.$$

On the other hand, since $0 < C_0 < 1$, we have

$$\begin{aligned} \eta &= \frac{(T_b - T_{b_0}) + C_0(T_{b_0} - \sigma)}{T_b} \\ &< \frac{(T_b - T_{b_0}) + (T_{b_0} - \sigma)}{T_b} \\ &< \frac{T_b - \sigma}{T_b} \\ &< 1. \end{aligned}$$

In short, when $T_b \geq T_{b_0}$, we have $0 < \eta < 1$.

If $T_b < T_{b_0}$, we have

$$\begin{aligned} T_c &= T_{b_0} \gg \sigma, T_b \gg \sigma, \\ \eta &= -\frac{(T_{b_0} - T_c) + C_0(\sigma - T_b - T_{b_0} + T_c)}{T_c + C_0(T_b - T_c)} \\ &= -\frac{C_0(\sigma - T_{b_0})}{T_{b_0} + C_0(T_b - T_{b_0})} \\ &= \frac{C_0(T_b - \sigma)}{(1 - C_0)T_{b_0} + C_0T_b}. \end{aligned}$$

On one hand, since $T_b \gg \sigma$ and $0 < C_0 < 1$, we have

$$\eta > 0.$$

On the other hand, since $0 < T_b - \sigma < T_b$, we have

$$\begin{aligned} \eta &= \frac{C_0(T_b - \sigma)}{(1 - C_0)T_{b_0} + C_0T_b} \\ &< \frac{C_0T_b}{(1 - C_0)T_{b_0} + C_0T_b} \\ &< 1. \end{aligned}$$

In short, when $T_b < T_{b_0}$, we have $0 < \eta < 1$.

We now prove (a). First, we note

$$\begin{aligned} k &\triangleq \lim_{n \rightarrow \infty} n\beta \\ &= \lim_{n \rightarrow \infty} \left[(n-1)\beta \times \frac{n}{n-1} \right] \\ &= \lim_{n \rightarrow \infty} (n-1)\beta. \end{aligned}$$

Then, applying the Poisson distribution to approximate the binomial distributions in (1) and (2), we have

$$\begin{aligned} \lim_{n \rightarrow \infty} P_e &= \lim_{n \rightarrow \infty} (1 - \beta)^n C_0 = e^{-k} C_0, \\ \lim_{n \rightarrow \infty} P_b &= \lim_{n \rightarrow \infty} [1 - (1 - \beta)^n] C_0 = (1 - e^{-k}) C_0, \\ \lim_{n \rightarrow \infty} P_{b_0} &= \lim_{n \rightarrow \infty} (1 - \beta)^n [1 - C_0] = e^{-k} (1 - C_0), \\ \lim_{n \rightarrow \infty} P_c &= \lim_{n \rightarrow \infty} (1 - P_e - P_b - P_{b_0}) \\ &= 1 - e^{-k} C_0 - (1 - e^{-k}) C_0 - e^{-k} (1 - C_0) \\ &= 1 - C_0 - e^{-k} (1 - C_0) \\ &= (1 - C_0)(1 - e^{-k}). \\ \lim_{n \rightarrow \infty} P_s &= \lim_{n \rightarrow \infty} n\beta(1 - \beta)^{n-1} C_0 = ke^{-k} C_0, \end{aligned}$$

and

$$\begin{aligned} \lim_{n \rightarrow \infty} \Omega &= \lim_{n \rightarrow \infty} (P_e \sigma + P_b T_b + P_{b_0} T_{b_0} + P_c T_c) \\ &= e^{-k} C_0 \sigma + (1 - e^{-k}) C_0 T_b \\ &\quad + e^{-k} (1 - C_0) T_{b_0} + (1 - C_0)(1 - e^{-k}) T_c \\ &= T_c + C_0(T_b - T_c) \\ &\quad + \frac{C_0(\sigma - T_b) + (1 - C_0)(T_{b_0} - T_c)}{e^k}. \end{aligned}$$

Next, we calculate $\Gamma(k)$ as follows.

$$\begin{aligned}
\Gamma(k) &= \lim_{n \rightarrow \infty} \Gamma(n, \beta) \\
&= \lim_{n \rightarrow \infty} \frac{P_s L}{\Omega} \\
&= \frac{ke^{-k} C_0 L}{T_c + C_0(T_b - T_c) + \frac{C_0(\sigma - T_b) + (1 - C_0)(T_{b_0} - T_c)}{e^k}} \\
&= \frac{k C_0 L}{[T_c + C_0(T_b - T_c)] e^k + C_0(\sigma - T_b) + (1 - C_0)(T_{b_0} - T_c)} \\
&= \frac{k C_0 L}{[T_c + C_0(T_b - T_c)] e^k + (T_{b_0} - T_c) + C_0(\sigma - T_b - T_{b_0} + T_c)} \\
&= \frac{k}{e^k - \frac{(T_{b_0} - T_c) + C_0(\sigma - T_b - T_{b_0} + T_c)}{T_c + C_0(T_b - T_c)}} \frac{C_0 L}{T_c + C_0(T_b - T_c)} \\
&= \frac{k}{e^k - \eta} \frac{C_0 L}{T_c + C_0(T_b - T_c)}.
\end{aligned}$$

We now prove (b). To maximize $\Gamma(k)$, we set the first derivative of $\Gamma(k)$ to zero and have

$$\begin{aligned}
1 - \eta e^{-k} &= k, \\
(k - 1)e^{k-1} &= -\eta e^{-1}.
\end{aligned}$$

Hence, we have

$$\begin{aligned}
k - 1 &= \mathcal{W}_0(-\eta e^{-1}), \text{ or} \\
k - 1 &= \mathcal{W}_{-1}\left(\frac{-\eta}{e}\right).
\end{aligned}$$

Finally, since only $\mathcal{W}_0(-\eta e^{-1}) > -1$ for $-\eta e^{-1} \in (-1/e, 0)$, we have

$$\begin{aligned}
k_{\text{opt}} &= \mathcal{W}_0(-\eta e^{-1}) + 1 \\
&\geq 0.
\end{aligned}$$

■

Competing interests

The authors declare that they have no competing interests.

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