Analysis of Contention Tree Algorithms
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Abstract—The Capetanakis–Tsybakov–Mikhailov contention tree algorithm provides an efficient scheme for multiaccessing a broadcast-communication channel. This paper studies statistical properties of multiple-access contention tree algorithms with ternary feedback for arbitrary degree of node. The particular quantities under investigation are the number of levels required for a random contender to have successful access, as well as the number of levels and the number of contention frames required to provide access for all contenders. Through classical Fourier analysis approximations to both the average and the variance are calculated as a function of the number of contenders \( n \). It is demonstrated that in the limit of large \( n \) these quantities do not converge to a fixed mode, but contain an oscillating term as well.

Index Terms—Broadcast communication, collision resolution, contention tree algorithms, random multiple access.

I. INTRODUCTION

The allocation of a single broadcast-communication channel among a large number of independent transmitters usually requires more advanced medium-access protocols than time-division multiple access (TDMA). The reason is that TDMA provides a notoriously low performance with respect to channel utilization, unless all transmitters are continuously transmitting, and with respect to access delay, unless the number of users is low.

Introduced in the early 1970s as a solution to the problem sketched above, the ALOHA protocol yields an elegant scheme to provide immediate random access to the channel [1]. The concept of random access implies that two or more transmitters may be active at the same time, prohibiting error-free reception. If such a collision occurs, the transmitters try again later, each one after a randomly chosen time. However, the performance of the ALOHA protocol becomes very poor, if the channel occupancy number of users is low.

Basically, there are two strategies to improve the performance of random multiple-access protocols: carrier-sense multiple access [2] and collision-resolution algorithms [3], [4], [25]. This paper studies statistical properties of the basic collision-resolution algorithm: the contention tree algorithm. The outline is as follows: Section II overviews the development and explains the operation of the contention tree algorithm. In Sections III and IV, we investigate the number of levels and the number of contention frames, respectively, required to complete the tree algorithm. We present conclusions in Section V. Appendices A and B provide the details of our mathematical analysis. A brief account of this work was presented in [5].

II. THE CONTENTION TREE ALGORITHM

Let us now describe the multiple-access contention tree algorithm as first reported by Capetanakis [3] and by Tsybakov and Mikhailov [4], [25]. A large number of transmitters (stations, terminals, sources, etc.) share a single, slotted broadcast channel. The transmitters that contend for channel access are able to acquire ternary feedback on what happened during a contention slot, i.e., whether zero transmitters (an empty slot), one transmitter (a successful transmission), or more than one transmitter (a collision) has been broadcasting during the particular slot. The ternary feedback can either be detected by the stations themselves or by a central controller and is not required to be immediate, i.e., there may be a certain delay between the transmission during the contention slot and the reception of the feedback. Furthermore, the tree has nodal degree \( m \geq 2 \), and as a consequence (see below) \( m \) consecutive contention slots are grouped into a contention frame.

The contention tree algorithm utilizes the ternary feedback as follows. Let us assume that there are \( n \) contending transmitters at the start of a new tree algorithm, i.e., \( n \) transmitters want to broadcast a data packet. During the first contention frame, i.e., the frame at the root of the tree, each of the \( n \) transmitters picks at random a number (say \( k \)) between 1 and \( m \)—with equal probabilities—and transmits its packet during the \( k \)th contention slot. If after completion of the contention frame the ternary feedback becomes available, each transmitter knows whether its packet has been successfully broadcast or not. If not, a new contention frame is assigned to all transmitters that caused the collision during the particular slot. Therefore, if there were collisions in all contention slots, \( m \) new contention frames would become available. This leads to the formation of a tree with nodal degree \( m \). The expansion of the tree stops at either empty or successful slots. Upon completion of the tree algorithm, all the \( n \) contenders have successfully broadcast their data. Thereafter, a new tree algorithm may start again. To exemplify the contention tree algorithm, Fig. 1 depicts a possible contention tree for \( n = 13 \) contenders and \( m = 3 \) slots per frame. Note that the formation of the tree is a stochastic process, because in each frame each contender picks a slot at random. The contention tree depicted in Fig. 1 is just one realization out of an infinite number of possible trees for \( n = 13 \) and \( m = 3 \). The development of the algorithm is thus the result of an interplay between the exponential growth of the contention tree and the random choices made by the contenders.

Capetanakis [3] has shown that in the case of Poisson generated data packets the maximum throughput of the binary \((m = 2)\) contention tree algorithm equals 0.347 packets/slot.
Capetanakis also demonstrated that if the number of contention slots in the root frame is allowed to be variable, but all other frames still contain two slots, the maximum throughput equals 0.430 packets/slot. The most efficient contention tree algorithm demonstrated so far leads to a maximum capacity of 0.487 packets/slot, again in the case of Poisson generated data packets [6]–[10], [26].

A variant of the contention tree algorithm, the contention stack algorithm, has been introduced by Tsybakov and Vvedenskaya [11]–[13], [27], [28]. In the stack algorithm, a transmitter does not need to wait until the tree is completed, but is allowed to contend in the first available contention frame. An advantage is that this eliminates the requirement that all transmitters continuously monitor the status of the channel.

In the above described algorithms the contention process is being executed with the data packets themselves, in other words, a successful contention implies that the data is transmitted as well. Alternatively, one can make use of the contention tree algorithm to make reservations or requests for data transmission. As generally the length of a reservation packet is smaller than the length of a data packet, the reservation contention tree algorithm may lead to even larger channel utilization [14], [15], [29]. This mechanism is currently being proposed in several network standards [16], [17].

Apart from the channel utilization, the access delay, i.e., the time it takes before the data packet has successfully been transmitted, is also an important performance parameter. In order to minimize the access delay, one must take into account the round-trip delay between the transmission of a packet and the reception of the feedback [3], [18]. If the round-trip delay is negligibly small, it is most advantageous to perform a serial search, in which each branch is fully completed before returning to the root. However, in communication channels with a large round-trip delay, such as metropolitan cable networks and satellite networks, it may be advantageous to perform a parallel tree search, in which all the contention frames at a certain level are executed before proceeding to the next level [3].

In the present paper, we investigate statistical properties of the contention tree algorithm. In particular, we study as a function of the number of contenders \( n \) and the number of slots in a contention frame (or nodal degree) \( m \) the following statistical quantities.

- The number of levels \( d_n \) required for a random contender to have successful contention. This number is of importance to calculate the mean access delay in systems with a large round-trip delay. In Fig. 1, \( d_{13} \) is either 2, 3, or 4, depending on the contender. The average equals \( \frac{32}{10} \).
- The number of levels \( D_n \) required to complete the tree. This number is of importance to calculate the duration of the algorithm in systems with a large round-trip delay. In Fig. 1, \( D_{13} = 4 \).
- The number of contention frames \( L_n \) required to complete the tree algorithm. In Fig. 1, \( L_{13} = 11 \). This quantity determines how much of the channel capacity is needed for the tree algorithm. In the case of a negligible round-trip delay, it determines the duration of the algorithm, as well.

As far as we know, the quantities \( d_n \) and \( D_n \) have not been studied in detail before. The quantity \( L_n \) has been thoroughly investigated from the moment of its introduction [3], [4], [6]–[9], [19]–[22], [25], [26], [30]. The reason for reinvestigating \( L_n \) is that the techniques taken at hand to calculate various statistical properties of \( d_n \) and \( D_n \) can be readily applied to the quantity \( L_n \). This allows us to confirm and state precisely various results and conjectures presented earlier [8], [20], [21].

The aim of our calculations is to analyze both the expectation values \( \mathbb{E}_n, \mathbb{D}_n, \text{and} \mathbb{L}_n \) as well as the variances \( \text{var}(d_n) \), \( \text{var}(D_n) \), and \( \text{var}(L_n) \). Through classical Fourier analysis we have found analytical expressions for these quantities in the limit of large \( n \). In a nutshell, the results can be summarized as follows:

\[
\begin{align*}
\mathbb{d}_n & \approx \log_m (n - 1) \\
\mathbb{D}_n & \approx 2 \log_m n \\
\mathbb{L}_n & \approx \frac{n}{\log_m n}
\end{align*}
\]

where the logarithm base \( m \) is given by \( \log_m n \equiv \log n / \log m \) and \( \log m \equiv \ln m \). More precise results are presented in Sections III and IV, with the mathematical details given in Appendices A and B. From comparison with the exact results it follows that the expressions obtained are already quite accurate for rather small values of \( n \). Furthermore, we demonstrate that the averages and variances obtained do not converge for large \( n \) to the laws (1)–(3), but contain oscillating terms as well, reflecting the discrete-level nature of the contention tree. We learned through the kindness of a referee that the oscillatory behavior of \( L_n \) was noted without proof by B. Hajek in 1980 and N. D. Vvedenskaya in 1984.

This paper considers the statistical properties of the contention tree algorithm only. However, our results can be combined with arbitrary traffic models in order to make predictions on the performance of various cases. In many situations, (1)–(3) suffice to make back-of-the-envelope estimates on the performance.
The probability distribution of the number of levels \( d_n \) required for a random transmitter to have successful contention when \( n \) transmitters contend in a tree algorithm with \( m \) contention slots per frame reads

\[
P_d(d|n) = \begin{cases} 
\delta_{1,d}, & n = 1 \\
\alpha_{n-1}(m^d) - \alpha_{n-1}(m^{d-1}), & n \geq 2 
\end{cases}
\] (4)

with the Kronecker delta function \( \delta_{1,d} = 1 \) if \( d = 1 \) and \( \delta_{1,d} = 0 \) otherwise. The function \( \alpha_n(M) \) is given by

\[
\alpha_n(M) = (1 - 1/M)^n
\] (5)

if \( M > 1 \), \( \alpha_n(M) = 0 \) otherwise. Equations (4) and (5) can be understood as follows, where we borrow an argument from [21], [23]: We consider an infinite, complete tree of nodal degree \( m \). The number of slots in level \( d \) amounts to \( m^d \). In the first level, each of the \( n \) contenders picks at random one of the \( m^d \) slots. This process is repeated for each subsequent level. As a result, the \( n \) contenders in level \( d \) are independently and identically, randomly distributed with equal probabilities over the \( m^d \) slots. Therefore, the probability that a slot in level \( d \) occupied by a random contender is not occupied by any of the \( n - 1 \) other contenders equals \( \alpha_{n-1}(m^d) \). The difference between \( \alpha_{n-1}(m^d) \) and \( \alpha_{n-1}(m^{d-1}) \) equals the probability that the random contender requires precisely \( d \) levels to be the single occupant of a contention slot. The fact that in the implementation of the contention tree algorithm the tree is not expanded upon an empty or successful slot, does not change this argument.

Similarly, we have for the probability distribution of the number of levels \( D_n \) required to complete the tree

\[
P_D(D|n) = \begin{cases} 
\delta_{1,d}, & n = 1 \\
b_n(m^D) - b_n(m^{D-1}), & n \geq 2 
\end{cases}
\] (6)

where \( b_n(m^D) \) denotes the probability that all \( n \) contenders in level \( D \) occupy different slots

\[
b_n(M) = \frac{M!}{(M-n)!M^n}
\] (7)

if \( M \geq n \), \( b_n(M) = 0 \) otherwise. Fig. 2 provides a histogram of \( P_d(d|n) \) and \( P_D(D|n) \) for \( n = 5 \), \( 25 \), and \( m = 3 \). Indeed, it takes a few levels more to have successful transmission for all contenders than for a random contender.

Given the probability distribution \( P_d(d|n) \) one can readily calculate the average

\[
\bar{d}_n = \sum_{d=1}^{\infty} dP_d(d|n)
\] (8)

the second moment

\[
\bar{d}_n^2 = \sum_{d=1}^{\infty} d^2P_d(d|n)
\] (9)

and the variance

\[
\text{var}(d_n) = \overline{d_n^2} - \overline{d_n}^2,
\] (10)

Similarly, \( \overline{D_n}, \overline{D_n^2}, \) and \( \text{var}(D_n) \) can be evaluated.

The moments and variance of \( d_n \) and \( D_n \) as a function of \( n \) and \( m \) can be computed up to arbitrary precision. However, much more insight in the tree algorithm can be obtained if the general behavior of \( d_n \), \( \text{var}(d_n) \), \( \overline{D_n} \), and \( \text{var}(D_n) \) is known as a function of \( n \) and \( m \). In Appendix A, we derive accurate analytical approximations for these values in the limit of large \( n \).

In short, our derivation for \( d_n \) proceeds as follows. First, we expand the function \( \alpha_n(m^p) \) in a series with terms \( n^{-k}f_k(n/m^p) \). Secondly, we note that

\[
f_k(n/m^p) - f_k(n/m^{p-1}) \approx 0
\]

except when \( m^p \approx n \), so that the summation in (8) and (9) can be replaced by a summation from \(-\infty\) to \( \infty \). Finally, we utilize classical Fourier analysis to approximate the summation up to the aimed accuracy. This approach can be directly applied to \( D_n \), as well. As expected, we find a logarithmic dependence on \( n \) for both \( \overline{d_n} \) and \( \overline{D_n} \). However, around this “DC value” there is a Fourier series of oscillations. For \( n \ll 20 \) only the first oscillation is significant, the others being exponentially small.

Let us now quote our results. From (64), (65), and (76) we have

\[
\overline{d_n} = \overline{d_n} + \delta d_n + O(n^{-2}).
\] (11)

The first term on the right-hand side (RHS) denotes the “DC value,” which is given by

\[
\overline{d_n} = \log_m(n-1) + \left( \frac{1}{2} + \frac{\log m}{\log^2 m} \right) + \frac{1}{2n \log m}
\] (12)

with Euler’s constant \( \gamma \approx 0.5772 \). We find an approximate logarithmic dependence of \( \overline{d_n} \) as a function of \( n \), as is not unexpected for the tree algorithm. Multiplying the number of contenders by a factor of \( m \), leads to an increase of one required level. Around this “DC value” there are oscillations given by

\[
\delta d_n = \kappa_1 \sin[2\pi \log_m(n-1) - \phi_1]
\] (13)
TABLE I
THE PARAMETERS REQUIRED TO CALCULATE THE AVERAGE AND THE
VARIANCE OF THE NUMBER OF LEVELS $d_{\alpha}$ AND $D_{\alpha}$ FOR VARIOUS $m$
of CONTENTION SLOTS PER FRAME. THE OSCILLATIONS IN THE AVERAGES $\delta d_{\alpha}$
AND $\delta D_{\alpha}$ HAVE AMPLITUDE $\kappa_{1}$ AND PHASE $\phi_{1}$, SEE (13) AND (16). THE
LARGE-$n$ LIMIT OF THE “DC VALUE” OF THE VARIANCES $\tilde{v}_{d_{\alpha}}$ AND $\tilde{v}_{D_{\alpha}}$ IS
GIVEN BY $\tilde{v}_{d_{\alpha}}$, SEE (22). THE OSCILLATIONS AROUND THIS VALUE $\delta \tilde{v}_{d_{\alpha}}$
AND $\delta \tilde{v}_{D_{\alpha}}$ CAN BE DESCRIBED BY $\kappa_{2}$ AND $\phi_{2}$, SEE (21).

<table>
<thead>
<tr>
<th>$m$</th>
<th>$\kappa_{1}$</th>
<th>$\phi_{1}$</th>
<th>$\kappa_{2}$</th>
<th>$\phi_{2}$</th>
<th>$\tilde{v}<em>{D</em>{\alpha}}$</th>
</tr>
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<tr>
<td>2</td>
<td>$1.375 \cdot 10^{-5}$</td>
<td>-0.873</td>
<td>$1.465 \cdot 10^{-3}$</td>
<td>2.798</td>
<td>3.5071</td>
</tr>
<tr>
<td>3</td>
<td>$2.394 \cdot 10^{-4}$</td>
<td>-1.258</td>
<td>$1.244 \cdot 10^{-3}$</td>
<td>2.503</td>
<td>1.4462</td>
</tr>
<tr>
<td>5</td>
<td>$3.423 \cdot 10^{-3}$</td>
<td>2.177</td>
<td>$1.099 \cdot 10^{-2}$</td>
<td>-0.246</td>
<td>0.7184</td>
</tr>
<tr>
<td>10</td>
<td>$1.813 \cdot 10^{-2}$</td>
<td>0.765</td>
<td>$3.731 \cdot 10^{-2}$</td>
<td>-1.543</td>
<td>0.3936</td>
</tr>
</tbody>
</table>

where $\kappa_{1}$ and $\phi_{1}$ are defined according to (65). The values of $\kappa_{2}$ and $\phi_{2}$ are listed in Table I for several values of $m$. The magnitude of the oscillation increases with $m$.

Fig. 3(a) compares the approximation $\tilde{d}_{\alpha} + \delta d_{\alpha}$ with the exact value $\tilde{d}_{\alpha}$. We note that our large-$n$ approximation is remarkably accurate. Already, for $n = 5$ the deviation is below 1%. Fig. 4 displays the same data but now versus $\log(n-1)$. Indeed, one notices the logarithmic behavior, but there is a deviation from this behavior as can be clearly observed. This is due to the $1/n$ term in (12) and the oscillation (13).

The oscillation in $\tilde{D}_{\alpha}$ can be studied in more detail in Fig. 5, which compares $\tilde{D}_{\alpha} - \tilde{d}_{\alpha}$ with $\delta D_{\alpha}$. Note, that for $m = 3$ the oscillation is at least three orders of magnitude smaller than the “DC value,” and is therefore barely visible in Figs. 3(a) and 4. The approximation becomes excellent when $n$ increases. In mathematical terms, the exponential increase of the period of the oscillations with $n$ reflects the nature of the tree expansion, while the oscillations themselves are a consequence of the discreteness of the number of levels in the contention tree.

Similarly to $\tilde{d}_{\alpha}$, we find for the approximation of $\tilde{D}_{\alpha}$ using (78) instead of (76)

$$\tilde{D}_{\alpha} = \tilde{D}_{\alpha} + \delta D_{\alpha} + \mathcal{O}(n^{-2})$$

(14)

The “DC value” is given by

$$\tilde{D}_{\alpha} = 2 \log_{m} n + \left(\frac{1}{2} + \frac{\gamma - \log 2}{\log m}\right) - \frac{1}{3n \log m}.$$ 

(15)

It takes approximately twice as much levels to have successful transmission for all contenders than for a random contender. The oscillation $\delta D_{\alpha}$ around $\tilde{D}_{\alpha}$ has the same amplitude as $\delta d_{\alpha}$, at—approximately—a doubled frequency

$$\delta D_{\alpha} = \kappa_{1} \sin\left[2\pi \log_{m}(n^{2}/2) - \phi_{1}\right].$$

(16)

Again, we have found a “DC value” of magnitude

$$\tilde{v}_{d_{\alpha}} = \frac{1}{12} + \frac{\pi^{2}}{6 \log^{2} m} - \frac{1}{n \log^{2} m}.$$ 

(17)
Fig. 6. The variances \( \text{var}(d_n) \) and \( \text{var}(D_n) \) as a function of \( n \) for \( m = 10 \). Symbols denote the exact value calculated from (4)–(10), lines the approximation according to (19)–(21). For clarity, \( \text{var}(D_n) \) has been offset by a value of 0.2.

\[
\hat{\sigma}_{D_n} = \frac{1}{12} + \frac{\pi^2}{6 \log^2 m} - \frac{4}{3n \log^2 m}
\]

and an oscillation around this value according to

\[
\delta \hat{\sigma}_{d_n, D_n} = -\kappa_2 \sin(2\pi z - \phi_2)
\]

where \( z = \log_m (n-1) \) for \( \delta \hat{\sigma}_{d_n} \) and \( z = \log_m (n^2/2) \) for \( \delta \hat{\sigma}_{D_n} \). The parameters \( \kappa_2 \) and \( \phi_2 \) are defined in (67), see also Table I.

We find that \( \text{var}(d_n) \approx \text{var}(D_n) \). This is not unexpected, since \( D_n \) is in fact the largest value of the \( n \) values of \( d_k \), for each realization of the tree algorithm. For large \( n \), we have

\[
\lim_{n \to \infty} \hat{\sigma}_{d_n} = \lim_{n \to \infty} \hat{\sigma}_{D_n} = \hat{\sigma}_{d, D} = \frac{1}{12} + \frac{\pi^2}{6 \log^2 m}.
\]

Values are displayed in Table I. The variances \( \text{var}(d_n) \) and \( \text{var}(D_n) \) are plotted in Fig. 6 for \( m = 10 \) (the oscillations are less prominent for smaller \( m \)). Note, the difference in the period of the oscillation.

IV. THE NUMBER OF CONTENTION FRAMES

Let us now address the number of contention frames \( L_n \) required to complete the contention tree algorithm with \( n \) contenders. This quantity has been studied extensively upon the introduction of the contention tree algorithm itself [3], [4], [6]–[9], [14], [19]–[22], [25], [26], [29], [30]. We note that the exact definition of \( L_n \) varies a bit from author to author. The differences are due to whether the root frame consists of one or \( m \) contention slots and to whether the number of contention slots are counted instead of the number of contention frames. We will follow the definition as given in [21], which corresponds to counting the number of contention frames. For the tree of Fig. 1 this implies \( I_{13} = 11 \). The expectation value of \( L_n \) can be expressed recursively according to [8], [22]

\[
\bar{L}_n = (1 - m^2 (m - n))^{-1} \left[ 1 + \sum_{k=2}^{n-1} \binom{n}{k} \frac{(m-1)^{n-k}}{m^k \cdot I_k} \right]
\]

for \( n \geq 3 \). The first two values are \( \bar{L}_1 = 1, \bar{L}_2 = m/(m - 1) \). It has been demonstrated in [8], [21] that \( \bar{L}_n \) increases proportionally with \( n \). It has been suggested by Massey [8] that for a binary tree (\( m = 2 \)) the constant of proportionality equals \( 1/\log 2 \). However, this suggestion was rebutted by the observation that \( \bar{L}_n/n \) does not really converge to a fixed value, but rather oscillates weakly around some value [8], [21]. Below, we reinvestigate this issue and obtain exact expressions for the “DC value” as well as the magnitude and the phase of the oscillation.

Equation (23) allows easy calculation of the values of \( \bar{L}_n \). However, for further analysis it is more convenient to start from the expression by Kaplan and Gulko [21]

\[
\bar{L}_n = 1 + \sum_{i=1}^\infty c_n(m^i)
\]

where the function \( c_n(M) \) is given by

\[
c_n(M) = M \left[ 1 - (1 - 1/M)^n \right] - n(1 - 1/M)^{n-1}.
\]

The term \( c_n(m^i) \) in the summation of (24) equals the expected number of contention slots with collisions in level \( p \) and thus the expected number of contention frames in the next level.

In Appendix B, we analyze and approximate the infinite series (24) in a similar fashion as used for the evaluation of \( d_n \) and \( D_n \). The result can be written as

\[
\bar{L}_n = \bar{L}_n + \delta \bar{L}_n \pm o(n^{-1})
\]

where \( \bar{L}_n \) denotes the “DC value,” given by

\[
\bar{L}_n = \frac{n}{\log m} \left( 1 - \frac{1}{m-1} \right).
\]

Note that \( \bar{L}_n \) increases linearly with \( n \). Indeed, this result confirms and generalizes the conjecture by Massey [8] and the results by Mathys and Flajolet [20] that the constant of proportionality equals \( 1/\log m \). The fluctuation around this linear behavior can be approximated with excellent accuracy by

\[
\delta \bar{L}_n = -\lambda_1 \cos(2\pi \log m n + \phi_1) + \lambda_2 \sin(2\pi \log m n + \phi_2)
\]

where the parameters are given by

\[
\lambda_1 = \sqrt{\frac{2\pi^2/\log m}{4\pi^2 + \log^2 m}} \sinh(2\pi^2/\log m)
\]

\[
\theta_1 = \arg \left[ \frac{1 - 2i\pi/\log m}{0} \right]
\]

\[
\lambda_2 = \left( \frac{2\pi}{\log^2 m} \right)^{1/2} \sqrt{\frac{2\pi^2/\log m}{\sinh(2\pi^2/\log m)}}
\]

\[
\theta_2 = \arg\Gamma(1 - 2i\pi/\log m).
\]

Note that \( \theta_2 = -\phi_1 \). In Table II we display the numerical value of these parameters for selected values of \( m \). We remark that, similarly to \( \bar{L}_n \), the quantity \( \delta \bar{L}_n \) contains a part which rises linearly with \( n \) as well. However, for not too large values of \( m \) the amplitude of the fluctuation \( \lambda_1 \) is at least an order of magnitude smaller than the leading coefficient \( 1/\log m \). The limiting behavior of \( \bar{L}_n/n \) for \( m = 2 \) is in complete agreement with the numerical values found by Kaplan and Gulko [21].

Fig. 7 compares the exact values of \( \bar{L}_n \) as calculated from (23) with our approximation \( \bar{L}_n + \delta \bar{L}_n \) calculated from (27) and (28). Deviations are only observable for a small number of contenders \( n \leq 2 \) and are smaller than 1% if \( n \geq 3 \). For the values of \( m = 2, 3, 5, 10 \), respectively. Note further that the oscillating behavior is barely visible on this plot. The oscillation is depicted in more detail by Fig. 8, which compares the exact value minus the “DC value” \( \bar{L}_n - \bar{L}_n \) with the “AC approximation” \( \delta \bar{L}_n \) for \( m = 3 \). Note, that the vertical scale is three orders of magnitude smaller than in Fig. 7. Clearly, the
TABLE II

The Parameters Required for the Calculation of the Average and the Variance of the Number of Contention Frames $L_n$, Enlisted for Various Values of $m$, the Number of Contention Slots per Frame. The Constant of Proportionality for the “DC Value” of $\hat{L}_n$ equals $1/\log m$, See (27). The Quantities $\lambda_{1,2}$ and $\theta_{1,2}$ Describe the Fluctuation Around This Value, See (28)–(32). The Last Column Lists the Number $\tau_L$, Which Approximates $\text{var}(L_n)/n$ for Large $n$, According to (33) and (34).

<table>
<thead>
<tr>
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<th>$\lambda_1$</th>
<th>$\theta_1$</th>
<th>$\lambda_2$</th>
<th>$\theta_2$</th>
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<tr>
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<td>-0.765</td>
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</tr>
</tbody>
</table>

Fig. 7. The average number of contention frames versus the number of contenders $n$ for $m = 2, 3, 5, 10$ contention slots per frame. The symbols denote the exact value $\hat{L}_n$ according to (23), the lines the approximation $\hat{L}_n + \delta L_n$ according to (27) and (28).

Fig. 8. The oscillations in the average number of contention frames $\overline{L}_n$ as a function of the number of contenders $n$ for $m = 3$ contention slots per frame. The symbols denote $\overline{L}_n$ according to (23) and (27) and the line denotes $\delta L_n$ according to (28).

approximation becomes better upon increasing $n$. More details upon the accuracy of (25)–(28) can be found in Appendix B.

Finally, we address the variance of the number of contention frames. In [21] an expression is given for $\text{var}(L_n)/n$ in the limit of large $n$. Again, it was found from numerical evaluation that this value does not converge to a fixed value, but oscillates with small amplitude around a “DC value.” In Appendix B, we derive from the expression in [21] an analytical approximation for the variance, again using classical Fourier analysis. The result can be written as

$$\lim_{j \to \infty} \frac{\text{var}(L_n)}{n} \approx \hat{\nu}_L + \delta \nu_L(l)$$

where $l$ is fixed. The second term $\delta \nu_L(l)$ oscillates as a function of $l$. We have only explicitly evaluated the “DC value” $\hat{\nu}_L$ since the exact magnitude of the oscillation, though computable, is not relevant and very small. The result is

$$\hat{\nu}_L = \frac{2}{\log m} \sum_{m=1}^{\infty} \frac{1}{1 + m^2} + \frac{1}{2 \log m} - \frac{1}{(\log m)^2}. \quad (34)$$

The summation can be easily carried out numerically. The magnitude of $\hat{\nu}_L$ agrees with the numerical values given in [8], [21]. In Table II the number $\hat{\nu}_L$ is given for some values of $m$. As expected, the variance decreases upon increasing number of contention slots per frame.

V. CONCLUSIONS

We have analyzed properties of the contention tree algorithm for multiaccessing a broadcast-communication channel as a function of the number of contenders $n$ and the number of contention slots per frame $m$. The quantities under study are the number of levels $d_n$ required for a random contender to have successful access, as well as the number of levels $D_n$ and the number of contention frames $L_n$ required to complete the tree algorithm. These quantities are of importance for the evaluation of the performance of the contention tree protocol in communication channels with both low and high round-trip delays.

We have presented the probability distribution of $d_n$ and $D_n$, which enables us to determine various statistical quantities, such as the average $\overline{d}_n$, $\overline{D}_n$ and the variance $\text{var}(d_n)$, $\text{var}(D_n)$. Through classical Fourier analysis we have derived accurate, analytical approximations for these quantities. Both $\overline{d}_n$ and $\overline{D}_n$ increase logarithmically with $n$. Around this increase there is a small oscillation with exponentially increasing period which reflects the discrete-level nature of the contention tree. The amplitude increases with $\eta$. In addition, it is found that $\text{var}(d_n) \approx \text{var}(D_n)$ apart from similar oscillations.

Starting from expressions given by Kaplan and Gulko [21], the average $\overline{L}_n$ and variance $\text{var}(L_n)$ have been evaluated as well. This has allowed us to confirm the conjecture by Massey [8] and the results by Mathys and Flajolet [20] that $\overline{L}_n$ increases linearly with $n$ with constant of proportionality equal to $1/\log m$. This surmise was under debate because it had been found numerically that $\overline{L}_n/n$ does not converge to a fixed value, but rather oscillates. We have identified this oscillation as well.

APPENDIX A

DETAILS FOR SECTION III

In this appendix we derive the results given in Section III for the expectation value and the variance of the number of levels in the tree algorithm required for a random contender and for all
the contenders. The results follow from the approximation and the (approximate) Fourier analysis of the quantities
\[
\overline{d}_n^k \equiv \sum_{j=1}^{\infty} p^k [a_{n-1}(m^p) - a_{n-1}(m^{p-1})] \tag{35}
\]
\[
\overline{D}_n^k \equiv \sum_{j=1}^{\infty} p^k [b_{n}(m^p) - b_{n}(m^{p-1})] \tag{36}
\]
for \( k = 1, 2 \), as well as of the quantities \( \var{d_n} \) and \( \var{D_n} \), when \( n \to \infty \). The functions \( a_n \) and \( b_n \) are defined in (5) and (7).

A. Approximation and Relevant Summation Ranges

To obtain convenient expressions for mean and variance of \( d, D \) we expand \( a_n, b_n \) as
\[
a_n(M) = e^{-n/M} - \frac{1}{2n} \left( \frac{n}{M} \right)^2 e^{-n/M} + \frac{1}{n^2} \left[ \frac{1}{8} \left( \frac{n}{M} \right)^4 - \frac{1}{3} \left( \frac{n}{M} \right)^3 \right] e^{-n/M} + \cdots, \quad M \geq 1 \tag{37}
\]
\[
b_n(M) = e^{-n^2/2M} + \frac{1}{2n} \left[ \frac{n^2}{2M} - \frac{2}{3} \left( \frac{n^2}{2M} \right)^2 \right] e^{-n^2/2M}
+ \frac{1}{n^2} \left[ \frac{2}{3} \left( \frac{n^2}{2M} \right)^2 - \frac{4}{9} \left( \frac{n^2}{2M} \right)^3 + \frac{2}{9} \left( \frac{n^2}{2M} \right)^4 \right]
\times e^{-n^2/2M} + \cdots, \quad M \geq n. \tag{38}
\]
Here the errors caused by truncating the series are of the same order as the first deleted term.

Therefore, \( a_n(M) \) is either close to 0 or close to 1, unless \( M \) is confined to a region \((en, e^{1/n})\). It follows that in the series (35) only those \( p \) contribute that satisfy
\[
p \approx \frac{\log n}{\log m} = \log_m n. \tag{39}
\]
Similarly, in the series (36) only those \( p \) contribute that satisfy
\[
p \approx \frac{\log(n^2/2)}{\log m} = 2 \log_m n - \log_m 2. \tag{40}
\]
The proof of (37) with truncation error assessment follows easily from the Taylor expansion of \( \log(1-t/M) \equiv \log(1-t) \) around \( t = 0 \) and the inequality
\[
(1-t)^n \leq e^{-nt}, \quad 0 < t \leq 1 \tag{41}
\]
so that the \( t \)-regime \((0, 1] \) for which (37) has to be established can be split up conveniently in \((0, n^{-1/2}] \) and \((n^{-1/2}, 1] \). The proof of (38) uses the approximation
\[
\log \Gamma(z) = \left( z - \frac{1}{2} \right) \log z - z + \frac{1}{2} \log(2\pi) + O(z^{-1}),
\]
\[
z \to \infty \tag{42}
\]
with the inequality
\[
\frac{M!}{(M-n)!n^n} \leq \exp \left[ - \frac{n(n-1)}{2M} \right], \quad M \geq n \tag{43}
\]
for conveniently splitting up the range for \( M \) in \([n, n^{3/2}) \) and \((n^{3/2}, \infty) \), and a lengthy but elementary computation.

B. Fourier Analysis of Leading Approximations

We replace in (35) and (36)
\[
a_n(m^p) \quad \text{by} \quad \exp(-n/m^p) \tag{44}
\]
and we extend the summation range of \( p \) to all integers, at the expense of errors of order \( \exp(-n) \). We thus arrive at the leading approximation
\[
\mu_{k,d}(n+1) = \sum_{j=\infty}^{\infty} p^k [\exp(-n/m^p) - \exp(-n/m^{p-1})] \tag{45}
\]
\[
\mu_{k,d}(n) = \sum_{j=\infty}^{\infty} p^k [\exp(-n^2/2m^p) - \exp(-n^2/2m^{p-1})] \tag{46}
\]
for the \( k \)th moment of \( d, D \), respectively. Observe that
\[
\mu_{k,d}(n) = \mu_{k,d}(\frac{n+1}{2}) + 1 \tag{47}
\]
so that we can restrict ourselves in the remainder of this subsection to the evaluation of \( \mu_{k,d} \).

We introduce the notation \( \alpha = \log m, z = \log_m (n-1), \) i.e., \( n = e^\alpha + 1 \), and we define
\[
f(z) = \exp(-e^{\alpha z}) - \exp(-e^{\alpha(z+1)}). \tag{49}
\]
Then the following holds:
\[
\mu_{k,d}(n) = \sum_{j=\infty}^{\infty} p^k f(z-p). \tag{50}
\]
For \( k = 1, 2 \) we have
\[
\mu_{1,d}(n) = z g_1(z) - g_1(z) \tag{51}
\]
\[
\mu_{2,d}(n) = z^2 g_2(z) - 2z g_1(z) + g_2(z), \tag{52}
\]
Here we have set for \( k = 0, 1, 2 \)
\[
g_k(z) = \sum_{j=\infty}^{\infty} (z-p)^k f(z-p), \tag{53}
\]
which are one-periodic, bounded, smooth functions of \( z \). Noting that \( g_k(z) \equiv 1 \) (as \( \exp(-e^{\alpha z}) \) decreases from 1 to 0 as \( z \) increases from \(-\infty \) to \( \infty \)), we get
\[
\mu_{1,d}(n) = z - g_1(z) \tag{54}
\]
\[
\sigma_{d}^2(n) = \mu_{2,d}(n) - \mu_{1,d}(n) = g_2(z) - g_2(z) \tag{55}
\]
for the leading approximations of mean and variance of \( d \), respectively. Hence, \( \mu_{1,d}(n) \) grows like \( \log_m(n-1) \) with oscillations due to the term \( g_1(z) \), and \( \sigma_{d}^2(n) \) is a bounded function of \( n \), which is one-periodic in \( z = \log_m(n-1) \).

Let us now analyze the functions \( g_k(z) \) a bit further. From the Poisson summation formula and some elementary properties of the Fourier transform it follows that
\[
g_k(z) = \left( \frac{-1}{2\pi i} \right)^k \sum_{q=-\infty}^{\infty} F^{(k)}(q)e^{2\pi izq} \tag{56}
\]
where \( F^{(k)}(\nu) \equiv \partial^k F(\nu)/d\nu^k \) is the \( k \)th derivative of the Fourier transform
\[
F(\nu) = \int_{-\infty}^{\infty} e^{-2\pi iz\nu} f(z)dz, \quad \nu \in \mathbb{R} \tag{57}
\]
of \( f \) in (49). The function \( F(\nu) \) can be expressed in terms of the \( \Gamma \)-function as
\[
F(\nu) = \frac{e^{2\pi i\nu}}{2\pi i\nu} \Gamma(1 - 2\pi i\nu/\alpha).
\] (58)

It thus follows upon some elementary but rather lengthy computations that
\[
g_1(z) = -\left( \frac{1}{2} + \frac{\gamma}{\alpha} \right) - \frac{1}{2\pi} \sum_{q \neq 0} \frac{1}{q} \Gamma(1 - 2\pi iq/\alpha) e^{2\pi iqz}
\] (59)
\[
g_2(z) = \frac{1}{3} + \frac{\gamma}{\alpha^2} \left( \frac{\pi^2}{6} + \gamma^2 \right)
- \frac{1}{2\pi^2} \sum_{q \neq 0} \frac{1}{q} \Gamma(1 - 2\pi iq/\alpha) e^{2\pi iqz}
\times \left[ \pi i - 1/q - (2\pi i/\alpha) \psi(1 - 2\pi iq/\alpha) \right]
\] (60)

with the digamma function
\[
\psi(z) = \frac{\Gamma'(z)}{\Gamma(z)}
\] (61)

and Euler’s constant \( \gamma = \psi(1) \approx 0.5772 \).

We now argue that the terms in the two series at the RHS of (59) and (60) can be largely ignored. To this end we quote the formulas, see [24]
\[
\Gamma(1 + \delta y)^2 = \frac{\pi y}{\sinh \pi y}, \quad y \in \mathbb{R}
\] (62)
\[
\psi(z) = \log z + \mathcal{O}(z^{-1}), \quad z \to \infty, \quad \Re(z) \geq 0.
\] (63)

It thus follows that the two series in (59) and (60) are negligibly small for not too large values of \( \alpha \) (\( m = e^\alpha = 2, 3, 4 \)), while it is sufficient to include only the terms with \( q = \pm 1 \) for moderately larger values of \( \alpha \) (\( m = 5, \ldots, 20 \)).

Returning to (59) and (60) for the leading approximations, we see that \( \mu_{1,2}(n) \) has, within a negligible error, a sinusoidal oscillation around a monotonously increasing term
\[
\mu_{1,2}(n) = z + \left( \frac{1}{2} + \frac{\gamma}{\alpha} \right) + \kappa_1 \sin(2\pi z - \phi_1)
\] (64)
\[
\kappa_1 e^{j\phi_1} = \Gamma(1 + 2\pi i/\alpha)/\pi.
\] (65)

Similarly, we find for the variance
\[
\sigma_3^2(n) = \frac{1}{12} + \frac{\pi^2}{6\alpha^2} - \kappa_2 \sin(2\pi z - \phi_2),
\] (66)
\[
\kappa_2 e^{j\phi_2} = \Gamma(1 + 2\pi i/\alpha) \{2\pi \psi(1 + 2\pi i/\alpha) + \gamma/\alpha + i\} / \pi^2.
\] (67)

C. Error Analysis for Leading Approximations

We now briefly indicate how the leading approximations change when instead of per (44) \( a_0(M) \) is replaced by (see (37))
\[
e^{-n/M} = \frac{1}{2n} \left( \frac{n}{M} \right) ^2 e^{-n/M}.
\] (68)

An analysis similar to the one given in Appendix A-B shows that the RHS of (51) has to be changed into
\[
zg_0(z) - g_1(z) - \frac{1}{2n} \left[ zj_0(z) - j_1(z) \right]
\] (69)

and the RHS of (52) into
\[
z^2 g_0(z) - 2zg_1(z) - g_2(z) - \frac{1}{2n} \left[ z^2j_0(z) - 2zj_1(z) + j_2(z) \right]
\] (70)

where
\[
j_k(z) = \sum_{p=\infty}^{\infty} (z - p)^k h(z - p)
\] (71)

with
\[
h(z) = e^{2\pi z} \exp(-e^{2\pi z}) - e^{2\pi(z+1)} \exp(-e^{2\pi(z+1)}).
\] (72)

Since \( j_0(z) = 0 \), it follows that
\[
\mu_{1,2}(n) = z - g_1(z) + \frac{1}{2n} j_1(z)
\] (73)
\[
\sigma_3^2(n) = g_2(z) - g_1^2(z)
- \frac{1}{2n} \left[ j_2(z) - 2g_1(z)j_1(z) + g_1^2(z) \right].
\] (74)

Hence the corrected values differ by \( \mathcal{O}(n^{-1}) \) from (54) and (55).

One can make calculations for \( j_1 \) and \( j_2 \) in the same manner as was done in Appendix A-B for \( g_1 \) and \( g_2 \). It thus turns out that the function \( j_1(z) \), required for the average, and the function between the square brackets in (74), required for the variance, both oscillate with period 1 and small amplitude around the values \( 1/\alpha \) and
\[
\frac{2}{\alpha^2} \left( 1 + \frac{1}{4n} \right) - \frac{2\pi^2}{\alpha^3} \sum_{q \neq 0} \frac{2 - (1 + 4n^2 q^2/\alpha^2)/(2n)}{\sinh(2\pi^2 q/\alpha)}
\] (75)

respectively. The series in (75) is very small and will be neglected together with the \( 1/(4n) \). The first-order corrections to (54) and (55) are thus given by
\[
\mu_{1,2}(n) = z - g_1(z) + \frac{1}{2n} j_1(z)
\] (76)
\[
\sigma_3^2(n) = g_2(z) - g_1^2(z) - \frac{1}{n \alpha^2}.
\] (77)

In principle, one can continue the process of adding terms, see (37) and (68), so as to obtain higher order corrections. However, the expressions (76) and (77) approximate \( \Delta_0(n) \) and \( \text{var}(D_n) \) already up to sufficient accuracy. Inclusion of the third term in (37) only has the modest effect on \( \Delta_3(n) \) of magnitude \( 1/(12n^2) \).

We also note that the so-obtained series of approximations is asymptotic in nature, in the sense that the \( n \)-range where inclusion of the \( n \)-th term of the RHS of (37) and (38) yields a better approximation shifts toward \( \infty \) with increasing \( n \).

An evaluation of the corrections on mean and variance of \( D \) from (38) is almost similar. Note that the leading approximation of \( \overline{D_n} \) and \( \text{var}(D_n) \) can be obtained from the relation (48). Including the constant terms of the next order of approximation we have
\[
\mu_{1,2}(n) = z - g_1(z) - \frac{1}{3n \alpha} \exp(-e^{2\pi z})
\] (78)
\[
\sigma_3^2(n) = g_2(z) - g_1^2(z) - \frac{4}{3n \alpha^2}.
\] (79)

where now \( z = \log_\alpha (n^2/2) \).
In this appendix we derive the results presented in Section IV on the approximation of the average and the variance of the number of contention frames required for the tree algorithm in the limit of a large number of contenders $n$. We start from the expressions given in [21] for the average

$$
\bar{L}_n = \sum_{j=0}^{\infty} c_n(m^j)
$$  
(80)

where $c_n(M)$ is defined in (25) and for the variance

$$
\lim_{j \to \infty, n \to m^j} \frac{\text{var}(L_n)}{n} = 2 \sum_{r \geq p} \rho \left( \frac{1}{m^p} \right) \sum_{r \geq p} \chi \left( \frac{1}{m^r} \right) + \sum_{r \geq p} \rho \left( \frac{1}{m^p} \right) \chi \left( \frac{1}{m^r} \right) - \left[ \sum_{r \geq p} \tau \left( \frac{1}{m^r} \right) \right]^2
$$  
(81)

for fixed integer $l$, where

$$
\rho(x) = (1 + xe^{-x})
$$  
(82)

$$
\chi(x) = 1 - \frac{e^{-x}}{x} - e^{-x}
$$  
(83)

$$
\tau(x) = xe^{-x}
$$  
(84)

for $x \geq 0$. Our large-$n$ analysis makes use of the same type of approximations and Fourier analysis as the analysis in Appendix A.

**A. Approximation of $\bar{L}_n$**

The following approximation holds for (25):

$$
c_n(M) = n \chi \left( \frac{n}{M} \right) + \phi \left( \frac{n}{M} \right) + \frac{1}{n} \theta \left( \frac{n}{M} \right) + \frac{1}{n^2} \eta \left( \frac{n}{M} \right) + \cdots
$$  
(85)

where $\chi$ is given in (38) and

$$
\phi(x) = \frac{1}{2} x(x-1)e^{-x}
$$  
(86)

$$
\theta(x) = \left( -\frac{2}{3} x^2 + \frac{17}{24} x^3 - \frac{1}{8} x^4 \right) e^{-x}
$$  
(87)

$$
\eta(x) = \left( -x^2 + \frac{11}{12} x^3 - \frac{13}{48} x^4 + \frac{1}{48} x^5 \right) e^{-x}
$$  
(88)

for $x \geq 0$. The proof of (85) follows straightforwardly from (37).

Substitution of (85) into (80) yields

$$
\bar{L}_n = n \sum_{j=0}^{\infty} \chi \left( \frac{n}{m^j} \right) + \sum_{j=0}^{\infty} \phi \left( \frac{n}{m^j} \right) + E
$$  
(89)

where $E$ is the error due to the last two terms on the RHS of (85). This $E = O(n^{-1})$, since, for instance,

$$
\sum_{j=0}^{\infty} \frac{n}{m^{2j}} e^{-n/m^j} \leq \frac{1}{n} \sum_{j=-\infty}^{\infty} \left( \frac{n}{m^j} \right)^2 e^{-n/m^j}
$$  
(90)

the RHS-series being periodic (and whence bounded) in $\log_m n$. Furthermore, we extend the lower summation limit in (89) to minus infinity at the expense of an exponentially small error so that we obtain

$$
\bar{L}_n = \mu_{\bar{L}}, + O(n^{-1})
$$  
(91)

where

$$
\mu_{\bar{L}} = n \sum_{p=0}^{\infty} \chi \left( \frac{n}{m^p} \right) + \sum_{p=0}^{\infty} \phi \left( \frac{n}{m^p} \right) - \frac{1}{m-1}
$$  
(92)

where the last term on the RHS is due to the $1/x$ in (83). The highest order term of our result is equal to the function given in [21]. Below, it is demonstrated how from this term and the other series in (92) an analytical approximation to $\bar{L}_n$ can be obtained.

**B. Fourier Analysis of $\mu_{\bar{L}}$**

Similarly to Appendix A, we use the notation $\alpha = \log m$ and $z = \log_m n$, i.e., $n = e^{\alpha z}$. From the following Fourier transforms

$$
\int_{-\infty}^{\infty} e^{-2\pi iuv} \chi(\alpha u) du = \frac{1}{\alpha} \frac{\Gamma(1 - 2\pi i u/\alpha)}{1 + 2\pi i u/\alpha}
$$  
(93)

$$
\int_{-\infty}^{\infty} e^{-2\pi iuv} \phi(\alpha u) du = \frac{\pi u}{\alpha^2} \frac{\Gamma(1 - 2\pi i u/\alpha)}{1 + 2\pi i u/\alpha}
$$  
(94)

and from the Poisson summation formula it follows that

$$
\sum_{p=-\infty}^{\infty} \chi \left( \frac{n}{m^p} \right) = \frac{1}{\alpha} + \frac{1}{\alpha} \sum_{q \neq 0} \frac{\Gamma(1 - 2\pi i q/\alpha)}{1 + 2\pi i q/\alpha}
$$  
(95)

$$
\sum_{p=-\infty}^{\infty} \phi \left( \frac{n}{m^p} \right) = -\frac{\pi u}{\alpha^2} \sum_{q \neq 0} q \frac{\Gamma(1 - 2\pi i q/\alpha)}{1 + 2\pi i q/\alpha}
$$  
(96)

The result (95) has already been obtained by Mathys and Flajolet [20] on the basis of an asymptotic analysis. The assessment under what conditions and which terms in the RHS of (95) and (96) with $q \neq 0$ are significant is the same as in Appendix A-B. For not too large values of $m$, including only the $q = \pm 1$ terms, leads to sufficient accuracy. From (91), (92), (95), and (96) one can easily derive the results (26)–(32) presented in the main text.

In a similar fashion as in Appendix A-C, we can give corrections to the approximation just found by incorporating the higher order terms of (85) in the analysis. The term $\theta(n/M)/n$ yields a correction to $\bar{L}_n$ which oscillates with small amplitude around “DC value” 0, while the term $\eta(n/M)/n^2$ yields a correction term of magnitude $-1/\log n^2$. Hence, it is sufficient to consider only the first two terms in (85).

**C. Fourier Analysis of (81)**

We briefly outline how the RHS of (81) can be evaluated, the method being similar to the derivations above. We only explicitly calculate the “DC value,” but we have checked that the oscillations around this value are indeed at least an order of magnitude smaller.

Let us consider for $\beta > 0$ the quantities

$$
s(\beta; \beta) \equiv \sum_{p=-\infty}^{\infty} \rho \left( \frac{1}{m^p} \right) \chi \left( \frac{1}{m^p} \beta \right)
$$  
(97)
In the RHS of (81), the first term equals $2 \sum_{r \geq 1} s(l, m^{-r})$ and the second term equals $s(l, 1)$. Setting $z = \log_{m} l$, we obtain for the Fourier transform of $s(l, \beta)$

$$
S(\nu, \beta) = \frac{\Gamma(1 - 2\pi \nu \alpha) \beta}{\beta(\alpha + 2\pi \nu \alpha)} \left[ 1 + \frac{(1 - 2\pi \nu \alpha) \beta + \beta^2}{(1 + \beta)^2 - 2\pi \nu \alpha} - 1 \right].
$$

(98)

Note that $S(\nu, \beta) = O(\beta)$ as $\beta \to 0$. This ensures that the summation $s(l, m^{-r})$ over $r$ converges rapidly. The “DC term” of $s(l, \beta)$ is given by

$$
S(0, \beta) = \frac{1}{\alpha} \frac{\beta}{1 + \beta}.
$$

(99)

For the third term on the RHS of (81) we note that, as before,

$$
\sum_{j=\infty}^{\infty} \tau \left( \frac{1}{m^p} \right) = \frac{1}{\alpha} + \frac{1}{\alpha} \sum_{q=0}^{\infty} \Gamma(1 - 2\pi \nu q \alpha) e^{-2\pi \nu q \alpha}
$$

so that its contribution to the “DC value” of (81) is given by

$$
- \int_{0}^{\infty} \left[ \sum_{p} \tau \left( \frac{1}{m^p} \right) \right]^2 dz = \frac{-2}{\alpha^2} \sum_{q=1}^{\infty} \frac{2\pi q^2 / \alpha}{\text{sh}(2\pi q \alpha)}.
$$

(100)

The second term on the RHS is very small and will be neglected. Collecting the results, we arrive at our expression (34).

We conclude this appendix with the remark that the techni- ques used here can be applied to the series $d(\xi)$ given in [21, Theorem 7] (this quantity is of importance for contention tree algorithms where the number of slots in the root is variable). We immediately give the result

$$
d(\xi) = \sum_{p=\infty}^{\infty} \frac{\xi m^p}{(1 + \xi m^p)^2} = \frac{1}{\alpha} + 4\pi \sum_{q=1}^{\infty} \frac{q \cos(2\pi q z)}{\text{sh}(2\pi \sqrt{q^2 / \alpha})},
$$

(101)

where $z = 1/\log_{m} \xi$. As remarked in [21], the oscillations around the “DC value” are indeed substantially smaller than those in (95).

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