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LUPMAC: A Cross-Layer MAC Technique to Improve the Age of Information Over Dense WLANs

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Abstract—Age of Information (AoI) is a relatively new metric introduced to capture the freshness of a particular piece of information. While throughput and delay measurements are widely studied in the context of dense IEEE 802.11 Wireless LANs (WLANs), little is known in the literature about the AoI in this context. In this work we study the effects on the average AoI and its variance when a sensor node is immersed in a dense IEEE 802.11 WLAN. We introduce a new cross layer MAC technique, called Latest Update MAC (LUPMAC), aimed at modifying the existing IEEE 802.11 in order to minimize the average AoI at the receiver end. This technique lets the MAC layer keep only the most up to date packets of a particular piece of information in the buffer. We show, through simulation, that this technique achieves significant advantages in the case of a congested dense IEEE 802.11 WLAN, and it is resilient to changes in the variance of the total network delay.

Index Terms—Age of information, Dense WLANs, IEEE 802.11.

I. INTRODUCTION

The concept of the Age of Information (AoI) was first introduced in [1], and then formalized in [2]. This new metric answers the question: how fresh is that particular information stored at the receiver? It is different from the delay, since it includes the time from when a destination has received the last update about a particular piece of information in the buffer. These measurements are relevant, e.g. the current water level in a sewer pipe in order to ensure it does not exceed a given threshold. In this sense, the AoI metric is of crucial importance. Especially in the context of the new paradigm of the Internet of Things (IoT), or the Smart City paradigm [3], a typical application scenario might be sensor nodes continuously measuring and sending data, using a dense IEEE 802.11 Wireless Local Area Network (WLAN) shared amongst numerous other devices. For example, the sensor node might be interested in uploading the measured information to a remote unit, for storing or further processing. If the remote server is only interested in the freshest possible piece of the information sent by the sensor node, it is interested in the AoI at the receiver. In this work we will study a scenario where a sensor node is immersed in a dense IEEE 802.11 WLAN, where a number of devices are subscribed. It tries to send information to a remote destination. Dense WLANs are a specific scenario that will be covered in the forthcoming IEEE 802.11ax HEW (High Efficiency WiFi) standard [4]. The IEEE 802.11ah standard is also specifically designed for the IoT [5]. In this standard, an Access Point (AP) can cover up to 1 km in range, and is possible to foresee that overlapping networks with hundreds of devices would not be uncommon. Devices will have to compete for the channel with possibly hundreds of other devices, with a very heterogeneous population of traffic patterns. For example, there could be devices trying to offload traffic from the existing cellular infrastructure, further congesting existing IEEE 802.11 WLANs, as in the 5G HET-ereogeneous NETwork (HETNET) paradigm [6]. Competing with numerous devices degrades both throughput and delay performance, due to the increasing number of collisions, and in case of traffic burstiness, increases the idle time [7]. In case also of a high number of small frames, it deteriorates further [8]. The effects on the AoI are, however, not entirely clear.

In this paper, we extend the work in [1], [9] with a more practical implementation by introducing a new cross layer approach between the application layer and the MAC layer, called Latest Update MAC (LUPMAC), aimed at modifying the existing IEEE 802.11 in order to minimize the average AoI at the receiver end. We let the MAC know about the “freshness” of a packet received from the application layer, along with the particular application that generated it, in order to develop a strategy to minimize the AoI at the receiver. Briefly, it will always try to send the packets carrying the
freshest update of that particular information, without trying to transmit (or re-transmit) older packets.

The remainder of this paper is organized as follows. In Section II we will present an overview of the related work. In Section III we describe the concept of Age of Information. In Section IV we describe LUPMAC. In Section V we describe the scenario under test. In Section VI we present our simulation results and in Section VII we conclude the paper.

II. RELATED WORK

The Age of Information in IEEE 802.11 systems was first addressed in [1]. The authors study the age of information in a vehicular network (VANET) via simulation and with a VANET testbed. In their scenario, each vehicle acts as a node. Each node beacons a particular piece of information to nearby vehicles, and it is interested in the other vehicles having the most up to date piece of that information. Each node broadcasts its information, so no acknowledgements are involved. The authors introduce a cross layer MAC technique called “Latest state Out” (LO), in which the application sensing informations fills the packet at the front of the MAC buffer with the latest available piece of information whenever the opportunity of transmitting a frame arises. They show how this technique efficiently minimizes the average AoI in all the nodes in the VANET. They also show that using the optimal Contention Window (CW) from the Bianchi model [10] the average AoI is further minimized. They then show how neither maximizing the throughput nor minimizing the delay automatically minimizes the average AoI. Finally they introduce a cross-layer rate control mechanism that works with a normal FIFO queue and no CW adaptation in order to minimize the average AoI at the nodes.

Their work differs from the work carried out in this paper, since it studies a vehicular network, while we study a dense IEEE 802.11 WLAN of static nodes; we are interested in minimizing the AoI in a remote server instead of distributing the information to a set of nodes in the same network. Also they do not address the problem of other contenders (i.e. other devices trying to access the same wireless channel) in the network. Additionally, they are broadcasting the information, thus using only the first CW, not retrying to send the frame in case of a missing acknowledgment. Finally, in our work the MAC layer should be aware only of the application that generated the packet and the packet’s age, while in LO the MAC layer should signal the application whenever a transmission opportunity arises. In our work also, if the packets are sent by the application in order, the MAC layer will automatically infer the new packet is the freshest, thus not even needing an additional field with the packet’s age. The proposed LO technique is impractical. The time needed for the MAC layer to signal the application when it is ready to transmit, and then wait for the application layer to fill the MAC buffer is bigger than one IEEE 802.11 slot time (≈ 10μs), that is the time granularity in an IEEE 802.11 MAC. In addition, with this approach, the application must be allowed to write in the MAC buffer. This is in most of the cases, impractical. In short, this approach requires very close coupling between the MAC and the application that is both difficult and undesirable in practice. Finally, we will not use the optimal CW from the Bianchi’s model, since it is not possible in current hardware to change it at run time [10].

In [2], [9], [11]–[13] the authors study the AoI in different simple queuing systems with multiple classes of service, modeling the channel as a single server. For example, in [2] the authors derive a lower bound for the AoI given any service distribution in a simple queuing system with only one server. In [12] the authors study the minimization of the AoI under energy constraints, particularly a sensor that harvests energy from the environment via numerical simulation. While important properties of the AoI are derived, the effects on the AoI in a real life scenario such as a dense IEEE 802.11 WLAN are not investigated.

The only other study that uses a real network scenario in order to study the AoI, as far as the authors are aware, is [14]; there the authors study the AoI in an emulated WLAN with 2 nodes and compare their results with the theoretical results for various simple queuing systems. The study focuses on a small WLAN, and an IEEE 802.11 stack is not used, whereas we consider a dense WLAN with many more nodes and conduct simulations using a full 802.11 implementation.

As a final note, our cross-layer MAC technique is also a continuation of the work in [9], where the authors study the AoI in a system with N sources, a single queue and a delay channel. They introduce a new queuing discipline based on the age of information. It only holds the freshest packets of each class of information in the queue. On the other hand, the authors study the AoI in an abstract queueing system with N sources, where we make use of a full 802.11 implementation.

III. AGE OF INFORMATION

We will now give an overview of the concept of Age of Information. Consider a transmitter sensing and sending updates of the information I over a channel to a receiver. The receiver is interested only in the freshest update of information I. An example curve of the age of information I over time is depicted in Fig. 1.

Assume a packet with the desired information I is generated at time $t_{i-1}$ s from a a source sensing that information. The receiver receives it at time $t_{i-1}$ s. The packet will then have an age of $\epsilon_{i-1} = t_{i-1} - t_{i-1}$ s, so the age of the information I will be at time $\epsilon_{i-1}$ s. Then, if it is not receiving new packets, the AoI will increase over time with slope 1. The next packet carrying the updated information I is generated from the transmitter at time $t_i$ s. It is received at time $t_i$ s. The packet’s age at that point would then be $\epsilon_i = t_i - t_{i-1}$ s. If this packet is fresher than the current AoI (i.e. $\epsilon_i < t_i - t_{i-1}$ s) then the AoI will jump down to $\epsilon_i$ seconds, otherwise it will continue increasing. The AoI will continue to have this characteristic sawtooth behaviour, and it is possible to reconstruct its curve by interpolating between the various samples when packets are received. Then it is possible to reconstruct various metrics; for example, it is possible to reconstruct the average AoI by
The scenario considered in our work is depicted in Fig. 2. It models a sensor node immersed in a dense IEEE 802.11 WLAN with no hidden nodes, in order to better inspect the effects of LUPMAC on the average AoI. Such scenarios occur, for example, in a city, where a public hotspot serves a large number of users, and sensor nodes, such as smartgrid sensors or water flow sensors, use the existing infrastructure to send information remotely. Another example that could be modeled by this scenario is an industrial one, where sensor nodes have to send status updates about machinery to a central server while competing for the channel with other devices.
These sensor nodes send information fairly frequently, and the last reading of this information is what counts, i.e. we are interested in minimizing the average AoI.

In our studied scenario, a sensor node is sending various information streams to a remote server. A packet from the sensor has to be sent first via the wireless channel, then it is routed via a normal fixed-link connection, labelled as “Network” in the diagram, to a remote server. A sensor node is formed by an application layer stack, where there are a number of applications, labeled in the figure as sources, each one of those measures a particular piece of information, and sends updates about their own information to a remote server. In case LUPMAC is used, the applications running in the application layer insert their unique ID in a field in the packet header, that is propagated all the way to the MAC layer, in order to let LUPMAC know which application generated that particular packet. Then, there is a network layer, and then an IEEE 802.11 MAC, that holds the packets generated by the various sources in its buffer. Next there is an IEEE 802.11 PHY to access the channel.

A sensor node is competing for the channel with a number of contenders, each one requesting content from the remote server. The contenders send requests to the remote server, then the server fullfills those requests by sending back content to them. They send relatively small packets for the request, and receive packets of various sizes back. This models a variety of users that stream content, offload traffic using the IEEE 802.11 WLAN and simply browse the web.

The remote link introduces a delay according to a random distribution. Since the metro (or backbone) part of the network is usually reliable, at least in big cities, we will assume the remote link to be reliable, so no packet is dropped there. This models, for example, a routed path to a remote destination via the internet. On the other end, we tested the reliability of LUPMAC by measuring the average AoI both with high and low variance in the network part of the simulation.

VI. RESULTS

We have conducted our simulation studies using OMNeT++ and the INET package [20]. The parameters used in the simulations are summarized in Table I.

All the plots are presented with 95% confidence, allowing for a sufficient warm-up period before taking measurements. The scenario simulated is the one described in Section V.

Table I

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Frequency</td>
<td>2.4 GHz</td>
</tr>
<tr>
<td>Noise Power</td>
<td>-110 dBm</td>
</tr>
<tr>
<td>SINR Threshold</td>
<td>4 dB</td>
</tr>
<tr>
<td>Physical</td>
<td></td>
</tr>
<tr>
<td>Transmission Power</td>
<td>20 mW</td>
</tr>
<tr>
<td>Reception Threshold</td>
<td>-85 dBm</td>
</tr>
<tr>
<td>Data Rate</td>
<td>54 Mbps</td>
</tr>
<tr>
<td>Slot Time (µs)</td>
<td>9 µs</td>
</tr>
<tr>
<td>Scenario</td>
<td></td>
</tr>
<tr>
<td>Scenario dimensions</td>
<td>600 x 400 m</td>
</tr>
<tr>
<td>Channel model</td>
<td>Free space</td>
</tr>
<tr>
<td>Free space exponent</td>
<td>2</td>
</tr>
<tr>
<td>number of sensor nodes</td>
<td>1</td>
</tr>
<tr>
<td>number of contenders</td>
<td>variable</td>
</tr>
<tr>
<td>information generation</td>
<td>every 0.1 s</td>
</tr>
<tr>
<td>request generation</td>
<td>(\sim \exp(0.01)) s</td>
</tr>
<tr>
<td>Packet length (sensors)</td>
<td>10 bytes</td>
</tr>
<tr>
<td>Packet length (contenders)</td>
<td>(\sim \exp(10)) bytes</td>
</tr>
<tr>
<td>Requested packet length</td>
<td>(\sim U(14, 1000)) bytes(^1)</td>
</tr>
<tr>
<td>MAC</td>
<td></td>
</tr>
<tr>
<td>type</td>
<td>802.11g (AC1)</td>
</tr>
<tr>
<td>buffer length (packets)</td>
<td>100</td>
</tr>
</tbody>
</table>

There is a sensor node uploading data to a remote server. It has an application (source) running, taking measurements. The sensor node is using an IEEE 802.11g WLAN with a number of contenders varying from 0 to 60. It is uploading small packets deterministically at a fairly slow rate (10 pk/s). The contenders are issuing requests to a remote server with exponentially distributed interarrival times, with an average rate of 100 pk/s, in order to increase the traffic load on the WLAN and congest it. The request packets are small (10 bytes on average, exponentially distributed), while the reply packets are uniformly distributed from small packets (14 bytes, a control frame) to big packets (1000 bytes). The delay on the wire connecting the access point to the remote server is considered to be a reliable metro/backbone connection. The average roundtrip time is however considered to be challenging with respect to VoIP traffic (150ms).

The average AoI and its variance are measured with an increasing number of contenders in the case that the delay has narrow variance, i.e. the one-way delay is uniformly distributed between 74ms and 76ms (so as to have an average roundtrip time of 150ms) with LUPMAC or the standard IEEE 802.11 FIFO approach. Then it is tested in the case it has a large variance, i.e. the one-way delay is uniformly distributed between 0s and 150ms (still an average round trip time of 150ms) with LUPMAC or the standard IEEE 802.11 FIFO approach. In Fig. 3 the AoI for all the cases is presented.

As we can see from Fig. 3, the difference between high and narrow variance in the standard case (i.e. IEEE 802.11 FIFO)

\(^1U(a, b)\) is the uniform distribution between \(a\) and \(b\).
is quite small, only a fraction of the average AoI even with a totally saturated network with 60 contenders. In both cases the average AoI grows almost two tenths of a second from 10 to 60 contenders. This is quite a high increase, considering that the source on the sensor node is generating one packet every tenth of a second.

Then, we tested LUPMAC (introduced in Section IV). As we can see, LUPMAC significantly improves the AoI in case of a highly saturated scenario (when the number of contenders grows over 30), with an improvement of almost a tenth of a second with 60 contenders on the average AoI. Also, the AoI appears more stable, as the variance grows much more slowly when LUPMAC is on. The improvement over the average AoI is extremely good, considering that the source on the sensor node generates one packet every tenth of a second.

The improvement can be explained by the number of replaced packets in the MAC buffer when LUPMAC is used. In Fig. 4 the percentage of the replaced packets according to Algorithm 1 over the totality of packets sent by the application layer in the sensor is presented. As we can see, LUPMAC starts to replace packets in the MAC buffer as soon as we have a sufficiently high number of contenders in the WLAN (in this case $\geq 15$), exactly when the average AoI starts to diverge from the one measured in the standard case (i.e. IEEE 802.11 FIFO).

If we allow for a faster update generation, the benefits are overwhelming. In Fig. 5, the source on the sensor node is allowed to generate up to 100 pk/s, i.e. one packet every hundredth of a second with 30 contenders and narrow variance on the one-way network delay. Notice that the y-axis is in log-scale.

When LUPMAC is enabled, the average AoI is improved by up to an order of magnitude compared with when the sensor is simply relying on an unmodified IEEE 802.11 MAC. In addition, when LUPMAC is used, the average AoI is fairly stable, and its variance limited.

VII. CONCLUSIONS AND FUTURE WORK

In this paper we investigated the effects of contenders on the age of information of a sensor node immersed in a dense IEEE 802.11 WLAN via simulation. We then investigated the effects of variance of the transmission delay on the age of information. We also introduced a new MAC technique called LUPMAC designed to improve the performance of the IEEE 802.11 MAC for sensor nodes in terms of the average AoI.

An extension to this work will be to dynamically assign different traffic priorities to different information sources (i.e. different ACs, as defined in the IEEE 802.11e EDCA) according to priority and traffic load, following the findings from [1]. Another approach would be to use the technique described in [21], in order to use a probabilistic technique on top of LUPMAC, thus approximating the throughput optimal CW in order to minimize further the average AoI at the receiver.
end. Since sensors are usually low power devices, we will investigate the effects of the contenders in terms of energy usage of the sensors, while trying to minimize the average AoI at the receiver end taking inspiration from the findings in [12] and extending them in a realistic environment. Further steps are also a real implementation of LUPMAC in a sensor node and a mathematical evaluation of its performances in terms of the average AoI at the receiver end.

LUPMAC can be integrated into the existing IEEE 802.11ah standard, and fits in the wider scope of the IoT and 5G. It shows substantial benefits in terms of both the average AoI and its variance compared to the normal, unmodified IEEE 802.11 when the WLAN becomes saturated with traffic. This technique is also resilient to changes in the variance on the experienced delay.

Fig. 5. Average Age of Information (a) and variance (b) measured at the destination both with and without LUPMAC with the sensor generating up 100 pk/s, 30 contenders and narrow variance on the one-way network delay. Notice that the y-axis is in log-scale.

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