CTMC modelling for H2H/M2M coexistence in LTE-A/LTE-M networks
Ahmad Hani El Fawal, Mohamad Najem, Ali Mansour, Frédéric Le Roy, Denis Le Jeune

To cite this version:

HAL Id: hal-02055530
https://hal-ensta-bretagne.archives-ouvertes.fr/hal-02055530
Submitted on 4 Mar 2019

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L’archive ouverte pluridisciplinaire HAL, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d’enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Distributed under a Creative Commons Attribution - ShareAlike 4.0 International License
CTMC modelling for H2H/M2M coexistence in LTE-A/LTE-M networks

Ahmad Hani El Fawal1,2, Mohamad Najem3, Ali Mansour4, Frédéric Le Roy4, Denis Le Jeune6

1Lab-STICC, UMR 6285 – CNRS, ENSTA Bretagne, 29806 Brest, France
2AUL University, Beirut, Lebanon
3LIU University, Beirut, Lebanon
4E-mail: ahmad.el-fawal@ensta-bretagne.org

Abstract: Machine-to-machine (M2M) devices with their expected exponential booming in the near future, will be one of the significant factors to influence all mobile networks. Inevitably, the expected huge number of M2M devices causes saturation problems, and leads to remarkable impacts on both M2M and human-to-human (H2H) traffics, services, and applications. The research free-space lack requires creating an appropriate model which describes the functionality of long-term evolution-advanced (LTE-A) and long-term evolution for machine (LTE-M), through mathematical frameworks to evaluate relevant performance metrics. In this study, we bridge this gap by proposing a continuous-time Markov chain (CTMC) model as a stochastic process tool to characterise the H2H/M2M coexistence based on analytical equations. Afterwards, the authors validate the proposed model through extensive Monte Carlo simulations. Eventually, it becomes approachable to characterise the impact of H2H/M2M coexistence in one LTE-A/LTE-M radio resource allocation in dense areas and under disaster scenarios. The simulation results show that using a prioritise LTE-A system for both M2M and H2H traffics is convenient in dense area scenarios, while in emergency cases, it is more appropriate to use a non-prioritise traffic strategy to keep H2H and M2M traffics working properly at the same time.

1 Introduction
Traditionally, long-term evolution-advanced (LTE-A) network was developed to better serve human-to-human (H2H) services such as voice calls, video-streaming, and data traffics. However, with the novel paradigm so-called machine-to-machine (M2M) communications (e.g. actuators, sensors, smart meters etc.), an unprecedented innovation for the current LTE-A becomes a must [1]. Offering an attractive M2M services in a utopian autonomous internet of things (IoT) world seems tempting, but with massive connectivity, significant challenges arise on how to manage large number of devices, typically transmitting only small payloads, across wide range applications. In 2020, there will be around 50 billion connections with unavoidable coexistence among H2H and M2M traffics in one LTE-A network [2]; consequently, an efficient radio access strategy becomes one of the most challenges for mobile operators, researchers and the third generation partnership project (3GPP) community [3]. This community sounds ultimately keen on conducting several studies and researches to identify the mutual impact among M2M and H2H communications. In this paper, we mathematically characterise the key performance characteristics of M2M and H2H communications, then we propose an analytical methodology using continuous-time Markov chain (CTMC) model with the following objectives:

- Studying the mutual impact of both M2M and H2H traffics.
- Measuring and analysing the M2M congestion solutions.
- Modelling a new framework called coexistence analyser and network architecture for long-term evolution (CANAL).

The CANAL framework optimises prioritise radio resource allocation procedures in a LTE-A/long-term evolution for machines (LTE-M) network and achieves understanding of the system performance to reach good balance between M2M and H2H communications. Then, we simulate different dense area and disaster scenarios, which can be studied, analysed, and measured. Finally, by comparing the calculated results with the simulated ones, we come to the conclusion that with a noticeable results’ correlation we can validate our assumptions, models, and proposed architecture.

2 State of the art and motivations
Since the dawn of long-term evolution (LTE) in Rel-8 [4] till the recent Rel-15 [5], 3GPP has made major enhancements on LTE perspectives in term of reducing latency, power consumption, improving measurement accuracy, improving random access reliability, and small cell support [6].

LTE-A was coined initially to support H2H communications (e.g. web-browsing, phone calls, internet televisions etc.). M2M devices (e.g. smart cities, mobile health, smart meters etc.), with their sporadic small packet sizes, pour their payloads into the same LTE-A channel but with different aspects and specifications. Due to these differences, it would be an imperfect match for these two traffics to share the same LTE-A bandwidth. To solve this issue, the 3GPP introduces a licensed spectrum technology so-called LTE-M to serve M2M traffics, which occupies 1.4 MHz from the LTE-A bandwidth. However, with different perspectives and a diversity of applications of M2M and H2H traffics, many challenges are expected as a result of this coexistence. In [7], the main challenges raised by the M2M vision are spotted and by focusing in particular on the problems related to the support of massive M2M access in current cellular communication systems. Then, the most common approaches proposed in the literature to enable the coexistence of H2H and M2M services in the current and next generation of cellular wireless systems are analysed. In [8], the authors study the coexistence of H2H traffic and M2M traffic originating from wireless sensors. By using experimental measurements of real-world smart grid applications, the impact of variable H2H traffic loads on the sensor end-to-end delay performance is investigated. The obtained results show that a conventional Ethernet may cause a bottleneck and increase the delay for both H2H and M2M traffics. In [9], the author addresses the high overhead signalling load problem of M2M devices using a priority-based data aggregation scheme at the M2M gateway to maintain a good trade-off between the power consumption and delay requirement. According to this work, three types of priority are assigned to
M2M devices based on their related applications. Therefore, the data from M2M devices with higher priorities are served faster than the ones with lower priorities. An analytical model considering the idle and busy states behavior of the aggregator using preemption queuing with priority disciplines is developed and analysed. In [10], a shared channel resource allocation in an H2H/M2M coexistence scenario was considered in order to formulate the resource sharing problem among M2M and H2H communications. In [11], the author proposes a cognitive-based radio access model with a priority queuing scheme. The model is applied for a LTE-A networks with M2M/H2H coexistence distinguishing M2M devices based on their traffic quality of service (QoS) requirements, in which M2M communications have real-time (M2M-RT) and non-real-time (M2M-NRT) traffic. Radio access gives the highest priority to H2H, while M2M-RT has higher priority than M2M-NRT. Although an analytical methodology is developed in normal scenarios, but the expected surge number of M2M devices which might have higher priority during disaster scenarios and how to deal with this sticky situation have not been discussed.

Additionally, Markov chains are also found in the literature to characterise M2M and/or H2H traffics in LTE-A/LTE-M networks, due to the specificity of LTE-A/LTE-M networks as time-variant networks. Thus many works can be spotted based on this stochastic method. In [12], an efficient radio access strategy is proposed to manage an LTE network system where M2M devices and H2H users coexist. In this paper, a CTMC model is developed to evaluate the system performance in terms of service completion rate (SCR), blocking and forced termination probabilities, and mean queuing delay of the M2M traffic. Although the proposed model can be used to improve the system performance of M2M communication with different priorities, but the model design does not allow the analysis of the system behavior when facing different H2H applications with different priorities. Now, with a close look to many works found in the literature, many research flaws can be spotted: RACH (random access channel) congestion in a LTE-A/LTE-M network, high and low priority for both M2M and H2H traffics, queuing for high and low priority M2M and H2H mechanisms, adaptive resource allocation, and emergency scenario solutions.

As for to the congestion and overload problems, which may occur when a large number of M2M devices attempt to access the LTE-A network using RACH, a serious degradation of performance for both M2M and H2H devices is more likely to happen [13]. In our previous work [14], we compare the most common mechanisms found in the literature that deal with the RACH procedure issues and challenges by analysing the existing solutions and approaches to avoid RACH overload congestion in the M2M communications. In [15], the overload of physical RACH caused by M2M delay sensitive was addressed. The authors proposed a non-preemptive queuing model in order to investigate the performance of different M2M and H2H traffic classes under dynamic access grant time interval controlled scheduling scheme. This was done by using a queuing model focusing on the resource utilisation and the QoS. The results show a delay reduction and a less percentage of packet loss.

Based on all previous work outcomes, we are motivated to propose a CTMC model to study the H2H and M2M coexistence through a mathematical framework. Additionally, a new architecture is proposed which can help in studying and analysing the mutual impact between M2M and H2H traffic coexistence while considering high and low priority traffics for both M2M and H2H devices. Moreover, an adaptive resource allocation is proposed also to scale the network bandwidth especially during disaster scenarios.

### 3 Coexistence analyser and network architecture for LTE

Our aim is to measure the mutual impact of M2M and H2H traffics in different scenarios while maintaining a sustainable level of services with minimal congestion during emergency events by allowing both M2M and H2H devices to access the network resources efficiently. In this context, we propose a new architecture so-called CANAL, which extends the classical functionality of the eNodeB and includes an adaptive control of bandwidth based on the M2M load status throughput, as shown in Fig. 1.

The proposed CANAL architecture manages automatically both H2H and M2M traffics. Based on the instantaneous requests of M2M, a part of the total bandwidth is dedicated to fulfill these sudden needs. To this end, the model operates at different conditions depending on the M2M load state throughout the network. The CANAL architecture consists of a single uplink cell in one eNodeB dedicated for a LTE-A/LTE-M network.

Let \( RB_{H2H} \) be the maximum resource blocks reserved for LTE-A network and \( RB_{M2M} \) be the one reserved for LTE-M network. There are two types of traffics: M2M and H2H traffics. Each traffic has LP and HP arrival rates \( (\lambda_i) \) assumed to be following Poisson’s distributions [16]. Let \( (\lambda_{HI}, \lambda_{HL}) \) be the average arrival rates for

![Fig. 1 CANAL architecture](http://creativecommons.org/licenses/by/3.0/)

---

This is an open access article published by the IET under the Creative Commons Attribution License.
applying a special priority strategy, we end up with an efficient process is initiated when the system reaches its cut-off point (the full state) and a new M2M/H2H device request contends to access the system. The new arrival request might be either H2H/M2M HP or M2M LP. The contention resolution is the core of the QCU by differentiating between two cases:

- **H2H/M2M LP new arrival**: When a H2H/M2M LP requests an access to the system, the QCU checks the RAC for available resources. If there is at least one resource available, the RAC grants the request access to the network. If there is no available resources the request will be forwarded to the H2H/M2M LP-Q. However, if the queue is full the request is terminated; otherwise, the request is enrolled in the appropriate queue, as shown by the flowchart presented in Fig. 2.

- **H2H/M2M HP new arrival**: When a H2H/M2M HP requests an access to the system, the QCU checks the RAC for available resources. If there is at least one resource available, then the RAC grants the request access to the network. If there is no available resources, the QCU verifies whether all resources are reserved by H2H/M2M HP or not. If the answer is ‘yes’, the request will be forwarded to the H2H/M2M HP-Q. However, if the queue is full, then the request is terminated, otherwise the request is enrolled in the appropriate queue. If the answer is ‘No’, an interruption command will be initiated asking for an immediate evacuation of a H2H/M2M HP resource reservation which will be pushed to the H2H/M2M LP-Q. Similarly, if the queue is full the request is terminated; otherwise, the request is enrolled in the appropriate queue, as shown by the flowchart presented in Fig. 2.

### 3.1 Queueing control unit

The CANAL architecture contains four feedback queues for the four traffic types: H2H LP, H2H HP, M2M LP, and M2M HP. By applying a special priority strategy, we end up with an efficient radio management which dominates the H2H/M2M coexistence.

The system is considered in the ‘full state’ when the number of reserved resource blocks in a LTE-A network peaks to RB\textsubscript{H2H} and/or in a LTE-M network peaks to RB\textsubscript{M2M}. In Fig. 1, the queuing process is initiated when the system reaches its cut-off point (the full state) and a new M2M/H2H device request contends to access the system. The new arrival request might be either H2H/M2M HP or H2H/M2M LP. The contention resolution is the core of the QCU by differentiating between two cases:

- When a H2H/M2M LP requests an access to the system the QCU checks the RAC for available resources. If there is at least one resource available the RAC grants the request access to the network. If there is no available resources the request will be forwarded to the H2H/M2M LP-Q. However, if the queue is full the request is terminated; otherwise, the request is enrolled in the appropriate queue, as shown by the flowchart presented in Fig. 2.

### Table 1 Description of notations used in CANAL architecture

<table>
<thead>
<tr>
<th>Notation</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>R\textsubscript{B\textsubscript{H2H}}</td>
<td>maximum resource blocks reserved for LTE-A</td>
</tr>
<tr>
<td>R\textsubscript{B\textsubscript{M2M}}</td>
<td>maximum resource blocks reserved for LTE-M</td>
</tr>
<tr>
<td>(\lambda\textsubscript{H})</td>
<td>average arrival rate for H2H high priority</td>
</tr>
<tr>
<td>(\lambda\textsubscript{L})</td>
<td>average arrival rate for M2M low priority</td>
</tr>
<tr>
<td>QCU</td>
<td>queuing control unit</td>
</tr>
<tr>
<td>HP</td>
<td>high priority</td>
</tr>
<tr>
<td>LP</td>
<td>low priority</td>
</tr>
<tr>
<td>HP-Q</td>
<td>high priority queue</td>
</tr>
<tr>
<td>LP-Q</td>
<td>low priority queue</td>
</tr>
<tr>
<td>RAC</td>
<td>resource allocation control unit</td>
</tr>
<tr>
<td>(\mu\textsubscript{H})</td>
<td>completed H2H high priority service rate</td>
</tr>
<tr>
<td>(\mu\textsubscript{L})</td>
<td>completed M2M low priority service rate</td>
</tr>
<tr>
<td>n</td>
<td>H2H low priority queue size</td>
</tr>
<tr>
<td>o</td>
<td>H2H high priority queue size</td>
</tr>
<tr>
<td>p</td>
<td>M2M low priority queue size</td>
</tr>
<tr>
<td>q</td>
<td>M2M high priority queue size</td>
</tr>
</tbody>
</table>

This is an open access article published by the IET under the Creative Commons Attribution License (http://creativecommons.org/licenses/by/3.0/)
Similarly, if the CD data rate reaches the next threshold \( Th(1) \) again caused by an additional M2M storm, the CANAL architecture adapts gradually BL1 till it reaches:

\[
BL(M) = (M + 1) \times BL0.
\]

Finally, if the CD data rate exceeds the last threshold \( Th(M) \) an overload problem occurs. Once the number of M2M connections starts dwindling, the RAC keeps receding the LTE-M bandwidth iteratively until reaching the initial phase at the end of the disastrous events, consequently, resumes operating similar to any legacy e-NodeB. Following this methodology, the CANAL architecture has the adaptability to lend progressively a temporary bandwidth up to \( BL(M) = (M + 1) \times BL0 \) of the total H2H bandwidth to M2M devices tentative use. By implementing this proposed solution for the limited bandwidth saturation in LTE-M networks, an adaptive reallocation of the bandwidth leads for an acceptable resolution for any presumable M2M storm.

Based on our previous work in [17], we found out that by leasing some folds of LTE-M classic bandwidth, we can achieve a significant gain in the M2M traffic with the same ratio.

4 CTMC analytical methodology

Our CTMC analytical methodology consists of four steps:

- First, we use CTMC as a stochastic method to describe the sequence of possible events M2M or H2H built on different states and various priorities.
- Second, we generate the equilibrium equations which rule the probability of transitioning from one state to another.

### 4.1 Representing the system as a set of states

In our system states, two variables \((i, j)\) are considered to denote the number of two ongoing services one is high-priority traffic and the other is low-priority traffic.

- Third, we turn these probabilities and states into a linear system and by solving it we can calculate each state probability under certain conditions.
- Fourth, using the aforementioned probabilities, we propose some metrics to characterise the performance of the network, with the aim to evaluate M2M and H2H traffics.

![Fig. 2 H2H/M2M HP and LP queuing](image2)

![Fig. 3 RAC phases for CANAL architecture, where CD: connected devices, Th: threshold, INI: initial state, EMG: emergency state, BL: bandwidth limited, m: current state number, M: maximum state number](image3)
one state to another if an event occurs (increase/decrease of \(i\) or \(j\)). The two traffics have two average arrival rates \((\lambda_1, \lambda_2)\) assumed to follow Poisson distribution and two service rates \((\mu_1, \mu_2)\) assumed to follow an exponential distribution. The generic CTMC model is shown in Fig. 4. All notations used in the analytical methodology are summarised in Table 2.

The numbers of states \((s)\) based on the number of resources \(C\) should satisfy the following constraints:

\[
0 \leq i, j \leq C \tag{1}
\]

\[
0 \leq i + j \leq C \tag{2}
\]

where \(i\) represents the number of ongoing services for high-priority traffic and \(j\) represents the number of ongoing services for low-priority traffic.

Based on (1) and (2), \(s\) can be derived as follows:

\[
s_i = \sum_{0 \leq i + j \leq C} 1 = \sum_{i=0}^{C} \sum_{j=0}^{C-i} 1 = \sum_{i=0}^{C} (C-i+1) = \sum_{i=0}^{C} (C+1) - \sum_{i=1}^{C} j = (C+1)^2 - \sum_{i=1}^{C} i = (C+1)^2 - \frac{C(C+1)}{2} = (C+1)(C+2)/2 \tag{3}
\]

where \(1\) represents an indicator function defined on a set \(X\) that indicates membership of an element in a subset \(A\) of \(X\), having the value \(1\) for all elements of \(A\) and the value \(0\) for all elements of \(X\) not in \(A\). A numerical example can be found in [18].

4.2 Generating the equilibrium equations

In the following, we assume that the observation time intervals are so small that at one time interval only one event \((i+1, i-1, j+1, j-1)\) may occur. In this case, the system falls into one of the following three cases:

- **Case 1**: ‘empty state’, where \(i = j = 0\), includes one state \(S(0, 0)\) and has the following equilibrium relationship:
  
  \[
  (\lambda_1 + \lambda_2)\pi_{0,0} = \mu_1\pi_{1,0} + \mu_2\pi_{0,1} \tag{4}
  \]

  where \(\pi_{i,j}\) is the probability to be in the state \(S(i, j)\).

- **Case 2**: ‘occupied state’, where \(0 < i + j < c\) includes \((c-1)(c+2)/2\) states and has the following equilibrium equation:

### Table 2 Description of notations used in the analytical methodology

<table>
<thead>
<tr>
<th>Notation</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>(i)</td>
<td>number of ongoing services for HP traffic</td>
</tr>
<tr>
<td>(j)</td>
<td>number of ongoing services for LP traffic</td>
</tr>
<tr>
<td>(\lambda_1)</td>
<td>average arrival rate for HP traffic</td>
</tr>
<tr>
<td>(\lambda_2)</td>
<td>average arrival rate for LP traffic</td>
</tr>
<tr>
<td>(\mu_1)</td>
<td>completed service rate for HP traffic</td>
</tr>
<tr>
<td>(\mu_2)</td>
<td>completed service rate for LP traffic</td>
</tr>
<tr>
<td>(c)</td>
<td>number of resource blocks used in the network</td>
</tr>
<tr>
<td>(s)</td>
<td>numbers of states</td>
</tr>
<tr>
<td>(\pi_{i,j})</td>
<td>the probability to be in the state (S(i, j))</td>
</tr>
<tr>
<td>(A)</td>
<td>the coefficient matrix in a linear system</td>
</tr>
<tr>
<td>(\Pi)</td>
<td>the steady-state probability vector</td>
</tr>
<tr>
<td>SCR(_{HP})</td>
<td>SCR for HP traffic</td>
</tr>
<tr>
<td>SCR(_{LP})</td>
<td>SCR for LP traffic</td>
</tr>
<tr>
<td>(R_u)</td>
<td>resource utilisation</td>
</tr>
<tr>
<td>(E(i))</td>
<td>the expected value of ongoing services for HP traffic</td>
</tr>
<tr>
<td>(E(j))</td>
<td>the expected value of ongoing services for LP traffic</td>
</tr>
</tbody>
</table>

\(J.\) Eng.

This is an open access article published by the IET under the Creative Commons Attribution License (http://creativecommons.org/licenses/by/3.0/)
As we consider the service rate (μ) as constant in our model, (13) can be written as

\[ \text{SCR} = \mu \sum_{ij} i \pi_{0,i} \]  

As the sum of conjoint probability leads to the marginal probability, the previous equations can be re-written as follows:

\[ \text{SCR} = \mu \sum_{i} i \pi_{0,i} = \mu E(i) \]  

where \( E(i) \) represents the expected value of an ongoing service denoted by \( i \).

ii. Resource utilisation (\( R_u \)): This metric gives the probability of the system to be busy serving the arrivals in terms of the number of utilised RBs in each state:

\[ R_u = \sum_{ij} \frac{i+j}{c} \pi_{0,j} \]

\[ = \sum_i i \pi_{0,i} + \sum_j \frac{j}{c} \pi_{0,j} \]

\[ = \frac{E(i) + E(j)}{c} \]  

### 5 Simulations and result discussions

In this section, we present our developed simulation model which can generate both H2H and M2M traffics with full flexibility to add queuing or priority for any traffic in order to study the mutual impact for H2H and M2M traffics.

#### 5.1 M/M/1 simulations

An M/M/1 queue made by having a single server, where arrivals are determined by a Poisson process and job service times have an exponential distribution.

The proposed architecture is based on Matlab libraries presented in [19]. Many enhancements are made in order to match our CANAL architecture presented in Section 3:

- The architecture consists of a single-queue and single-server system with a double-traffic source. In order to omit the role of the queuing for the moment, we consider an infinite storage capacity (in this case the queue sizes \( n = o = p = q = 0 \) for the different traffics, see Table 1).
- The architecture follows a CTMC with transition rate matrix similar to the one in Section 4.
- Arrivals occur at rate \( \lambda \) according to a Poisson process and move the process from state \( i \) to \( (i+1) \) and/or \( j \) to \( (j+1) \).
- Service times have an exponential distribution with rate parameter \( \mu \) in the M/M/1 queue, where \( 1/\mu \) is the mean service time.
- A single server serves H2H and M2M traffics one at a time by allowing one event only (either \( i \) or \( j \)) in the smallest time interval.
- The queue could be configured to work according to a FIFO (first input first output), LIFO (last input first output) or by priority discipline, with a flexible buffer size.
- When the service is completed, the served request leaves the system and the number of ongoing services in the system reduces by one (\( i \) to \( (i-1) \) and/or \( j \) to \( (j-1) \)).
- The architecture makes it easy to compare empirical results with the corresponding theoretical results.

Our methodology to reach our final simulation model is to build the simplest model with one traffic first, then we develop the basic H2H and M2M traffic models in which we can fix a starting experimental point with some highlighted flaws. In the end, by

\[ (i\lambda + \lambda + i\mu + j\mu)\pi_{0,i,j} = (i+1)\mu\pi_{0,i+1,j} + (j+1)\mu\pi_{0,i,j+1} + \pi_{0,i} \]

\[ + (i+1)\mu \pi_{0,i+1,j} + (j+1)\mu \pi_{0,i,j+1} \]  

\[ \\text{Case 3:} \ \text{‘full state’, where } i + j = c, \text{ includes } (c+1) \text{ states and has the following equilibrium equation:} \]

\[ (i\mu + j\mu)\pi_{0,i,j} = \lambda_1 \pi_{0,i-1,j} + \lambda_2 \pi_{0,i-1,j-1} \]  

\[ + (i+1)\mu \pi_{0,i+1,j} + (j+1)\mu \pi_{0,i,j+1} \]  

\[ \sum_{i = 0}^{\infty} \sum_{j = 0}^{\infty} \pi_{0,i,j} = 1 \]  

\[ 0 \leq \pi_{0,i,j} \leq 1 \]  

The general equation in (7) can be written in linear form:

\[ A\Pi = 0 \]

where the square matrix \( A \) represents the coefficients in the linear system, and \( \Pi \) represents the steady-state probability vector:

\[ \Pi = \begin{bmatrix} \pi_{0,0} \\ \pi_{0,1} \\ \vdots \\ \pi_{0,c} \end{bmatrix} \]

By replacing the first row of \( A \) by the coefficients of (8), we obtain the following modified system:

\[ B\Pi = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \]  

Equation (7) can be solved while considering the numbers of states \( s_i \) as dimension for the steady-state probability vector \( \Pi \) and \( (s_i \times s_j) \) as dimensions for the square matrix \( B \). Knowing that \( s_i \) could be calculated using (3), \( B \) is a full rank \( (s_i \times s_j) \) matrix and linear system of (12) can be solved easily.

### 4.4 Performance and metrics

We use two performance metrics to validate our model:

i. **SCR**: It gives the number of completed requests per time interval and it is based on the service rate \( \mu \) and the number of ongoing requests for a certain traffic (e.g. SCR_{HP} and SCR_{LP} which represent the SCR for HP/LP traffics [11]):

\[ \text{SCR} = \sum_{ij} i\mu \pi_{0,i,j} \]  

As the sum of conjoint probability leads to the marginal probability, the previous equations can be re-written as follows:

\[ \text{SCR} = \mu \sum_{i} i \pi_{0,i} = \mu E(i) \]  

\[ \sum_{i = 0}^{\infty} \sum_{j = 0}^{\infty} \pi_{0,i,j} = 1 \]  

\[ 0 \leq \pi_{0,i,j} \leq 1 \]  

The above three equations can be summarised by the following generic equation:

\[ (\alpha \lambda_1 + \alpha \lambda_2 + i\beta \mu_1 + j\beta \mu_2)\pi_{0,i,j} = \beta \lambda_1 \pi_{0,i-1,j} + \beta \lambda_2 \pi_{0,i-1,j-1} + \pi_{0,i} \]

\[ + (i+1)\alpha \mu_1 \pi_{0,i+1,j} + (j+1)\alpha \mu_2 \pi_{0,i,j+1} \]

where \( \alpha = 0 \) indicates the ‘full state’ (otherwise \( \alpha = 1 \)) and \( \beta = 0 \) represents the ‘empty state’ (otherwise \( \beta = 1 \)).

### 4.3 Linear system solution

To recall, in our system states, \( (i,j) \) denotes the number of ongoing services, i.e. H2H and M2M. The system moves from one state to another, when a service is achieved or a new request arrives (by increasing or decreasing \( i \) or \( j \)) with a steady-state probability \( \pi_{0,i,j} \):

\[ \sum_{j = 0}^{s_j} \sum_{i = 0}^{s_i} \pi_{0,i,j} = 1 \]

\[ 0 \leq \pi_{0,i,j} \leq 1 \]  

This is an open access article published by the IET under the Creative Commons Attribution License (http://creativecommons.org/licenses/by/3.0/)
working on the drawbacks of the basic model, many enhancements
could be suggested as result of adding priority and queuing
strategies and compare it with the basic model.

5.2 Single traffic simulations and results

As a first step, we start by experimenting the behavior of a single
traffic system with the aim to prove the accuracy of the proposed
CTMC model. Additionally, a comparison for both analytical and
simulation results are conducted in order to validate the accuracy
of the system by calculating the mean absolute error (MAE): MAE = \(\frac{1}{n} \sum_{i=1}^{n} |x_{i0} - x_{i1}|\)
and the root mean squared error (RMSE): RMSE = \(\sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_{i0} - x_{i1})^2}\).

The analytical and simulation results for the probability to be in
a given state \(x_0\) and with a number of resource blocks \(c = 6\) RBs
are shown in Fig. 5, which shows that the accuracy of the proposed
system falls into the an error limit with a MAE\(_{\text{max}} = 0.0592\). Also, a
slight mismatch between the analysis results and simulation results
could be spotted especially in the full state P6 (where the arrival
average peaks to the maximum system capacity 6). This mismatch
is unavoidable because in our simulator, we added two random
generators for two arrivals \((i, j)\) following a Poisson
distribution. In some cases, \((i, j)\) exceeds the maximum number
of available resources \(c\). These cases should be removed in order to
respect the constraint: \((i + j ≤ c)\) which affects our random
variables supposed following the Poisson distribution. So, we solve
this issue by introducing a saturation function in our simulations.
Consequently, an acceptable approximation is considered for these
cases (where \(i + j > c\)), similar to occurrences of \((i + j = c)\),
knowing that in our example \((c = 6)\). In this case, the probability to
be in the full state is more likely to happen with a noticed
difference comparing to the analysis results. This non-linear
behavior can explain the small mismatch.

The results obtained with \((c = 6)\) drive us to do more
exploration by increasing \(c\) to 25 RBs. The simulated results are
shown in Fig. 6 in which we end up with MAE\(_{\text{max}} = 0.0177\). In
Fig. 6a, we can realise a minor mismatch at the end of the curves.
This mismatch is due to removing the unwanted values which do
not respect the following constraint: \(i + j ≤ C\) (as explained in
Fig. 5). Similarly, in Fig. 6c, a minor mismatch at the beginning of
the curves due to removing the unwanted values which do not
respect the constraint: \(0 ≤ i + j\). In Fig. 6b, the probability peaks
when the number of resources \((C = 24)\) because of the
high-arrival rate \((λ = 20)\) accompanied with a low-completion rate
\((μ = 5)\) and only a limited number of resources \((C_{\text{max}} = 25)\)
which results a fast network saturation.

To sum up, all previous exploration experiments are able to
achieve a good accuracy with the aforementioned MAE and RMS,
which might be considered as a valid starting point to take up our
simulation model by upgrading the single traffic simulation to a
basic simulation for H2H and M2M traffics.

5.3 Basic simulations and results

In order to study the mutual impact of H2H and M2M traffics in
dense area or emergency scenarios, we conduct several simulations
based on the proposed architecture in [19], with the following
parameters:

- We assume to have one LTE-A network using a bandwidth of 5
  MHz \((c = 25)\) in order to stress H2H and M2M traffics to the
  maximum.
- Each traffic has an average arrival rate \((λ, μ)\) with a service rate
  \((μ, μ)\).
- H2H and M2M traffics have the same priority.
- A FIFO queue type is used with an infinite capacity size.
- Simulation duration = 1000 s.

5.3.1 Dense area scenario: In urban and dense areas, a huge
number of M2M and H2H devices may exchange their payloads
exponentially. Consequently, we consider a fixed arrival rate of

\[λ_1 = 2; \ μ_1 = 4\]

H2H requests \((λ_1 = 5)\) and incremental arrival rate of M2M
requests \(2 < λ_2 ≤ 20\), while \(μ_1 = 0.5\) and \(μ_2 = 1\). From the results
shown in Fig. 7a and by calculating the percentage of served
requests comparing to the total arrivals, we can conclude the following:

- When \(λ_1 ≤ 15\): The system is able to serve all M2M (in red) and
  H2H (in blue) requests, because H2H arrivals, with \(λ_1 = 5\) and
  a service rate \(μ_1 = 0.5\), occupy an average of 5 resources from the
  25 total resources for 2 consecutive time intervals, which means
  on the second time interval it occupies 10 resources (5 previous
  arrivals and 5 new arrivals). Consequently, it remains around 15
  resources per time interval for M2M arrivals with a maximum
  \(λ_2 = 15\) and a service rate \(μ_2 = 1\).
- When \(15 ≤ λ_1 ≤ 20\): A degradation on both M2M and H2H SCR
  can be realised, because in our assumption they have similar
  priority and the total number of requests are much more than the
  available resources \(c = 25\). At the peak \((λ_1 = 20)\), only 4K of 5K
  total H2H requests are served \((80\%)\) and 17K of 20K total M2M
  requests are served \((85\%)\). As an average of 4 from the 25 total
  resources will be occupied by H2H traffic for 2 consecutive
time intervals (actually 4 previous arrivals and 4 new arrivals),
and the remaining 17 resources will be used by M2M traffic.
- The high utilisation of the system \((R_u = 100\%)\) requires a
  queuing strategy to fulfill the excessive requests and to minimise
  the number of forced terminated services, as suggested in
  Section 3.1.
5.3 Emergency scenario: In emergency cases, a M2M signalling storm is definitely expected as a result of connected devices, which leads inevitably to a M2M congestion. This congestion caused by the simultaneous synchronisation affects both H2H and M2M devices. In this scenario, we consider the following system parameters: $\lambda_1 = 50$, $50 \leq \lambda_2 \leq 200$, $\mu_1 = 0.5$ and $\mu_2 = 1$. From the results shown in Fig. 7b, when $\lambda_2 = 50$ a degradation on both M2M and H2H SCR can be realised, because they have similar priority and the total number of requests are much more than the available resources $c = 25$. Only 6K of 50K total H2H requests are served (16%) and 8 K of 50 K total M2M requests are served (16%), because an average of 8 from the 25 total resources will be occupied by H2H traffic for 2 consecutive time intervals, and the remaining 8 resources will be used by M2M traffic each time interval.

By increasing $\lambda_2 = 100, 150, 200$, we end up with the following results as shown in Table 3.

5.4 M/M/1 results

In Section 5.3, two different scenarios (dense area and emergency) are conducted to measure the H2H and M2M mutual impact but without neither considering a prioritise system nor applying queuing strategy to ensure the QoS of different traffics. Now, it is time to consider two scenarios with similar parameters using a prioritise system as follows.

5.4.1 Prioritise dense area scenario: In this simulation, we emphasise similar case to the one considered in Section 5.3.1; While assigning a higher priority for one traffic, we study the trade-off between cost or gain on H2H and M2M traffics.

From results shown in Fig. 7a, we realise that the system is able to serve all M2M and H2H requests as long as $\lambda_2 \leq 15$. After this point, the system maintains the higher priority traffic towards a degradation on the lower priority traffic.

5.4.2 Prioritise emergency scenario: Similar to the case in Section 5.3.2 but with different M2M and H2H priorities, the results shown in Fig. 7b spot that the system maintains the higher priority traffic with a total elimination of the lower priority traffic.

To sum up, the simulation results show that using a prioritise LTE-A system for both M2M and H2H traffics could handle more requests in dense area cases by maintaining the higher priority traffic without any degradation while reducing the completion rate of the lower priority traffic. Meanwhile, in emergency cases, the higher priority traffic takes precedence over the lower priority traffic with an improvement on its completion rate while totally eliminating the low priority traffic; in such cases, the non-priority traffic strategy has a competitive advantage because it keeps H2H and M2M traffics working at the same time.

6 Conclusion and perspectives

In this paper, we address the saturation problem caused by the expected huge number of M2M devices which leads to remarkable impacts on both M2M and H2H traffics. We have proposed an enhanced architecture designed for LTE-A/LTE-M networks in order to fulfill H2H/M2M traffic coexistence supported with various priority strategies to satisfy the QoS for each traffic. A queuing strategy is enforced by a QCU, which rule four queues allocated for different traffic types. A CTMC model is proposed as a stochastic process tool to characterise the H2H/M2M coexistence based on analytical equations. Our simulation results matched with the CTMC analytical model which corroborates the proposed model.

Moreover, they show that using a prioritise LTE-A system for both M2M and H2H traffics is more convenient in dense area scenarios; while in emergency cases, it is more appropriate to use a non-prioritise traffic strategy to keep both H2H and M2M traffics working properly at the same time.

In this work, a homogeneous number of resources has been considered for all applications; in our future work, the authors aim to study the heterogeneity in different applications which reflects the actual behavior of various IoT applications.
7 References


