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Dynamic noise modeling at roundabouts

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Abstract

Modeling spatial and temporal noise variations at roundabouts is a tedious task. Indeed, noise levels are strongly influenced by the complex vehicle interactions taking place at the entries. An accurate modeling of the merging process and its impact on vehicle kinematics, waiting time at the yield signs and queue length dynamics is therefore required. Analytical noise prediction models disregard those impacts since they are based on average flow demand patterns and predefined kinematic profiles. The only way to capture all traffic dynamics impacts on noise levels is to combine a traffic simulation tool with noise emission laws and a sound propagation model. Yet, such existing dynamic noise prediction packages fail in representing vehicle interactions when the roundabout is congested and are difficult to calibrate due to their numerous parameters. A new traffic simulation tool, specifically developed for roundabouts, is therefore proposed in this paper. It has few easy-to-calibrate parameters and can be readily combined with noise emission and propagation laws. The obtained noise package is able to produce relevant dynamic noise contour maps which can support noise emission assessment of local traffic management policies. Results are validated against empirical data collected on a French suburban roundabout on two different peak periods.

1 Introduction

Space-time variations in vehicle kinematics due to intersections were shown to greatly influence noise levels in urban areas (IMAGINE, 2004 [1]; Steele, 2001 [2]). Although several noise estimation procedures have been developed to catch these effects, few of them are suitable for improving noise mapping at roundabouts.

In most existing prediction tools, like the FHWA Traffic Noise Model (Fleming et al., 1995 [3]) or the German RLS-90 model (RLS, 1990 [4]), noise impacts of interrupted traffic flows are roughly taken into account through empirical correction factors. More sophisticated statistical models have been proposed to improve noise appraisal at signalized intersections in terms of traffic volumes, traffic speeds, traffic composition, geometry or pavement surface textures (Abo-Qudais and Alhiary, 2007 [5]). Derivation of similar regression formula at roundabouts is possible, yet unfruitful since spatial variations in noise levels cannot be captured.

To circumvent this deficiency, advanced analytical methodologies based on the computation of average noise signatures for each vehicle class have been introduced (Hunt and Samuels, 1992 [6], Certu, 1980 [7]). They combine a mean vehicle kinematic pattern per class with a noise emission law depending on vehicle speed and running conditions. A single class-specific vehicle is assumed to cross the junction. At each instant, the resulting sound power level and vehicle noise
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position are input into a propagation model to calculate the instantaneous sound pressure level at a given receptor point. The sound pressure exposure of the class-specific vehicle is then obtained by integrating the instantaneous sound pressure levels over the time needed to cross the junction. Repeating this process for a set of receivers along the infrastructure gives the average noise signature for the corresponding vehicle class. Influence of traffic flow can then be accounted for by multiplying the noise signature by the number of vehicles passing during a given interval and so, for each class. Such a method has already been applied to roundabouts (Picaut et al., 2005 [8]). It has also been implemented in the Harmonoise traffic source model to draw noise difference contour maps at road crossings between situations where all vehicles maintain constant speed without stopping and situations where they all have to decelerate, stop and then accelerate away from the junction (Watts, 2005 [9]). Spatial kinematic variations appeared to either increase or decrease average noise levels depending on the traffic volumes. The major drawback of this analytical procedure is to neglect the noise impacts of vehicle interactions. Particularly, it cannot catch noise variations due to: (i) random occurrence of stopping or freely-entering vehicles into the roundabout; (ii) fluctuating waiting times on the roundabout approaching links; (iii) fluctuating queue lengths; and (iv) potential hindrance effect due to a congestion spilling back from downstream. As a result, the output noise contour maps are not completely satisfactory.

The only way to catch the noise impacts of vehicle interaction consists in coupling a traffic flow simulation tool with vehicle noise emission laws and a sound propagation model. As underlined within the Imagine project (IMAGINE, 2007 [10]) this technique allows for evaluating how short-term transient queue events as well as spatial and temporal variations in vehicle kinematics influence noise estimates. It has already been implemented into softwares like DRONE (Bhaskar et al., 2007 [11]), MOBILEE (De Coensel et al., 2005 [12]), TUNE (Goodman, 2001 [13]) or ROTRANOMO (Volkmar, 2005 [14]). However, to the authors' knowledge, no model calibration or validation study was specifically conducted to assess noise estimates at roundabouts. This is troublesome since the corresponding traffic simulation tools included into these softwares (respectively AVENUE or M+P, PARAMICS, DRACULA, VISSIM) have a large number of parameters that are often difficult to calibrate and may affect noise outputs to a large extent. Moreover, these tools were recently shown to badly represent vehicle interactions and kinematics in highly congested situations (Hidas, 2005 [15]; Chevallier and Leclercq, 2008 [16]). By predicting unrealistic queue length and delay values, they may, therefore, lead to irrelevant noise estimates at roundabouts.

The goal of this paper is to propose another dynamic noise emission model based on a new microscopic traffic simulation tool, specifically devoted to roundabouts. This tool solves the major drawbacks of classical micro-simulation packages underlined within the IMAGINE project: (i) it has few parameters; (ii) it is easy-to-calibrate with a limited set of data; (iii) it is able to model the number of stopped vehicles, the vehicle delays and the queue length dynamics on each approaching links accurately whatever the traffic conditions on the roundabout. Consequently, realistic vehicle kinematics can be obtained and fed into appropriate noise emission laws. Then, a sound propagation calculation is performed with respect to geometry and urban landscape. This paper will highlight that accurate dynamic noise contour maps can be simulated. They can be used to assess the noise impacts of traffic flow at roundabouts in both free-flow and saturated conditions.

The first part of this paper will outline the three key components of the proposed framework: the microscopic traffic simulation tool, the noise emission laws and the sound propagation model. For the study purpose, we will only focus on: (i) a single mean emission law for light duty vehicles; (ii) a very basic propagation model. However, it should be highlighted that the
framework can handle with any class-specific emission laws and very complex propagation models. In the second part of the paper, the overall framework will be validated against empirical sound pressure levels collected at four receptor points close to the road axis of a single-lane roundabout. Results in terms of equivalent sound pressure levels, statistical descriptors as well as noise levels distributions are really convincing.

2 Basic components of the roundabout noise emission model

2.1 Overview of the model

The purpose of the roundabout noise emission model is to accurately account for traffic dynamics in order to reproduce spatial and short-term variations in noise levels in the vicinity of the infrastructure. As mentioned in the introduction, this can be done by coupling a traffic flow simulation tool with noise emission laws and a sound propagation model. The modeling chain is depicted in figure 1. The traffic flow simulation needs some input data to fix the traffic demand scenario (traffic volumes, destination proportions, traffic composition) as well as some parameters which has to be calibrated beforehand. Outputs of the traffic model are position $x^t$, speed $v^t$, acceleration $a^t$ and vehicle type of each vehicle, at each time $t$. Those outputs are fed into noise emission laws to assign an instantaneous sound power level $L^w_t$ to each vehicle. Then, individual contributions of each vehicle at a given receptor point can be calculated according to a sound propagation model. The instantaneous sound pressure levels at the receiver are obtained by summing all the contributions at each time. Finally, those levels can be used to calculate a large array of noise descriptors.

2.2 Microscopic traffic simulation tool

Microscopic traffic simulation tools aim at modeling the progression of individual vehicles through a road network. The overall simulation period is usually broken down into a number of discrete time-steps $\Delta t$. Positions of all vehicles are updated at each time-step by specific algorithms. Speed and acceleration can then be deduced from positions at successive time-steps. Two algorithms are commonly used in order to model a roundabout: (i) a car-following algorithm which simulates vehicle trajectories on the approaching/departure links and inside the roundabout (the position of any vehicle in the network is given in terms of the one of its leader); (ii) an insertion decision algorithm which specifies whether vehicles arriving at the yield sign can enter or not the roundabout.

Most of the existing microscopic simulation tools for roundabouts were shown to have the following drawbacks:

- profusion of parameters that may be troublesome or data consuming to calibrate (Smith et al., 2007 [17]);
sensitivity of the simulated vehicle trajectories to parameters (Akçelik and Besley, 2001 [18]);
- failure in modeling insertion rates, vehicle delays and queue length dynamics when a congestion spills back on the roundabout (Hidas, 2005 [15]; Chevallier and Leclercq, 2008 [16]).

As a result, reliability of noise level estimates cannot be guaranteed when these traffic flow models are used in combination with a noise estimation process.

This has motivated the development of a new parsimonious, easy-to-calibrate microscopic traffic model for roundabouts able to capture two kinds of observed merging behaviors depending on traffic conditions on the roundabout:
- in free-flow conditions, approaching vehicles respect the yield rule and insert into the roundabout only if the time-interval before the arrival of the next circulating vehicle is sufficient (Kyte et al., 1996 [19]);
- in congested conditions, approaching vehicles do not respect the yield rule anymore and alternatively insert between circulating vehicles following a ratio $\gamma$ (Cassidy and Ahn, 2005 [20], Troutbeck, 2002 [21]).

To model these merging behaviours, two distinct insertion decision algorithms are used:
- a gap-acceptance algorithm in free-flow conditions which specifies minimum distance and time-interval criteria to allow for an insertion;
- a rate-based algorithm in congested conditions which specifies the probability for an approaching vehicle to enter into the roundabout.

Figure 2 summarizes the main steps of the microscopic simulation model. Firstly, positions of all vehicles except those in conflict (the approaching vehicle who wants to enter the roundabout and the next circulating vehicle on the circulatory roadway who is going to arrive at the conflict point) are updated according to the car-following model. Then, traffic conditions on the roundabout are assessed before implementing the appropriate insertion decision algorithm. Those model components will be described in the sequel; more details can be found in Chevallier and Leclercq (2008 [16]).
a Car-following algorithm

In equilibrium traffic conditions on the roundabout, the velocity of vehicle $n$ at time $t$, $v_{tn}$, only depends on the spacing $s_{tn}$ with the vehicle ahead $n-1$. The relationship $v_{tn} = v_e(s_{tn})$ is classically referred to as the fundamental diagram. It is depicted in figure 3. It requires only three parameters: (i) the speed in free-flow conditions $u$; (ii) the minimum spacing when vehicles are stopped $s_0$; (iii) the speed at which a queue spills back on the network $w$. Free-flow conditions correspond to points on the fundamental diagram where both spacing and velocity are high while congested conditions match the other part of the curve.

![Fundamental diagram](image)

**Fig. 3** - Fundamental diagram

Because of aggressive insertions from the approaching links, $s_{tn}$ may shorten below its equilibrium value. In this case, experimental studies have demonstrated that vehicle $n$ adapts its speed to gradually increase the spacing with its leader. In other words, it tolerates short spacing for a while without braking sharply. This process is called a relaxation procedure. In the remaining, we denote $\beta_{tn}$ the ratio between the current spacing in front of vehicle $n$ and the equilibrium spacing (given by the fundamental diagram with respect to the speed of the leader). The relaxation process of vehicle $n$ starts as soon as another vehicle inserts just ahead with a very short spacing and lasts until $\beta_{tn} = 1$. In the chosen car-following model, the time evolution of $\beta_{tn}$ only depends on one additional parameter, the relaxation parameter, $\epsilon$, corresponding to the difference in speed drivers are willing to accept with the vehicle ahead in order to recover an equilibrium state (see Laval and Leclercq, 2008 [22] for more details). For instance, when two vehicles are closely spaced on the roundabout, $\epsilon$ is approximately the difference between the speed of the leader at the previous time-step and the desired speed of the follower for the current time-step.

Eventually, the position of vehicle $n$ at time $t + \Delta t$, $x_{tn}^{t+\Delta t}$, is given as the minimum between the position it is willing to reach when traffic is in free-flow condition and the position it cannot overpass due to the downstream vehicle $n-1$ when traffic is congested:

$$x_{tn}^{t+\Delta t} = \min \left[ x_{tn}^{t} + \min(u, v_{tn}^{t} + a \Delta t) \Delta t; \right.$$

$$\left. (1 - \alpha)x_{n-1}^{t} + \alpha x_{n-1}^{t} + v_{n-1}^{t} \Delta t - \alpha \beta_{tn}^{t+\Delta t} s_e(v_{n-1}^{t+\Delta t}) \right]$$

$$\min \left[ \right.$$"
vehicle trajectories on the approaching and departure arms as well as inside the roundabout. To account for different driver behaviours inside the roundabout, the parameters of the fundamental diagram can be modified (for instance, the speed in free-flow conditions can be reduced).

b Insertion decision algorithms

According to the current spacing of the first downstream circulating vehicle and the fundamental diagram, the prevailing traffic conditions on the roundabout (free-flow or congested) can be specified to chose the appropriate insertion decision algorithm.

Free-flow state

The gap-acceptance algorithm allows for an insertion into the roundabout during the current time-step if:
- a time $t_f$ has elapsed since the last circulating vehicle has passed by the entry. This criteria is equivalent to assuming a maximum inserting flow of $1/t_f$ which usually depends on the roundabout radius, the road gradient, or visibility constraints.
- the next circulating vehicle expected to pass by the entry is beyond a distance $d_{lag}$ which is sufficient to avoid collision in case of an insertion. This distance is given as the product of the free-flow speed $u$ and the difference between the minimum time-headway between two circulating vehicles that an approaching driver consider to be sufficient to enter the roundabout, $t_c$ (which is a model parameter), and $t_f$.
- the inserting vehicle has enough space to enter the roundabout, that is to say the position imposed by the next vehicle (second term of the minimum function in equation 1) do not lead the inserting vehicle to move backwards.

These conditions can be assessed thanks to the positions of all vehicles except those in conflict given by the car-following algorithm. In case of insertion, the positions of the two vehicles in conflict are updated according to the car-following model (equation 1) operating in equilibrium conditions ($\beta_n = 1$) since no aggressive insertion are assumed to occur in free-flow state.

Congested state

The rate-based insertion decision model aims at simulating an average insertion flow equal to $\gamma$ times the circulating flow on the roundabout passing by the entry. For this, on each approaching arm $i$, the average traffic volumes just upstream of the entry, $\Delta_i$, and just downstream of the entry on the roundabout, $\Omega_i$, should be computed (over the last 30s for instance). Then, as soon as an approaching vehicle is close enough to the yield-sign, the probability for it to enter the roundabout during the time-step, $p_{t+\Delta t}^i$, is computed from equation 2 (see Chevallier and Leclercq (2008) [16]):

$$p_{t+\Delta t}^i = \begin{cases} \frac{\Omega_i \gamma/(1 + \gamma) \Delta t}{1} & \text{if } \Delta_i \geq \Omega_i \gamma/(1 + \gamma) \\ 1 & \text{otherwise} \end{cases} \quad (2)$$

A Bernoulli process is then drawn to specify the outcome of the insertion process. In case of insertion, both vehicles in conflict may switch to a non-equilibrium state ($\beta_n^i < 1$) or stay in equilibrium ($\beta_n^i = 1$) depending on the position of their leader. Their new positions are then calculated according to the car-following equation [1].

c Input data

As any micro-simulation tool, the traffic flow model requires input data to specify the prevailing traffic demand scenario. Data can either be precisely measured or scaled from aggregated
traffic observations (see table 1).

<table>
<thead>
<tr>
<th>Input data</th>
<th>Precise on-site measurements</th>
<th>Aggregated traffic observations</th>
</tr>
</thead>
<tbody>
<tr>
<td>Traffic volume</td>
<td>Arrival times of all vehicles at the beginning of each approaching arm</td>
<td>Time-series of average flows at the beginning of each approaching arm</td>
</tr>
<tr>
<td>Destination</td>
<td>Destination of each vehicle entering the approaching arms</td>
<td>Turning proportions for each approaching stream</td>
</tr>
<tr>
<td>Traffic composition</td>
<td>Type of each vehicle entering the approaching arms</td>
<td>Vehicle type proportions for each approaching stream</td>
</tr>
</tbody>
</table>

**Table 1 - Input data depending on the level of detail of on-site measurements**

### 2.3 Noise emission laws

The traffic model outputs are fed into noise emission laws in order to calculate the instantaneous sound power levels, $L_{wn}^t$, for any vehicle $n$ on the roundabout. Any noise emission law in terms of vehicle type, vehicle speed and/or acceleration can be used. In this study, a single mean emission law is selected to characterize all vehicles on the roundabout (mainly light-duty vehicles as we will see later). This hypothesis was validated in Can et al. (2007 [23]) for classical descriptors estimation in urban areas, provided that traffic dynamics is precisely described. The chosen emission law is the average law for light-duty vehicles used within the French traffic noise prediction model (Hamet, 2008 [24]) and corresponding to the pavement surface texture of the studied experimental site (see section 3). It has two main advantages: (i) it is specifically devoted for urban traffic conditions; (ii) it has been validated in real life traffic conditions, for instance when vehicles are accelerating away from a signalized intersection (low speeds and high accelerations) at different points of a urban corridor (Leclercq et al., 2008 [25]).

This emission law, depicted in figure 4a, depends on vehicle speed and running conditions which are specified by the vehicle acceleration: (i) accelerating mode for $a > 0.5 \text{ms}^{-2}$; (ii) decelerating mode for $a < 0.5 \text{ms}^{-2}$; (iii) cruising mode otherwise. For speeds below 30km/h, the engine-exhaust noise is predominant (Peeters and Blokland, 2007 [10]). The value of $L_{wn}^t$ depends on the running conditions of $n$: in cruising or deceleration modes, it follows a logarithmic function truncated for low speeds; in acceleration mode it reaches a higher constant level. For speeds above 30km/h, the tire-pavement noise prevails upon the engine-exhaust noise: no distinction in running conditions is made for the calculation of $L_{wn}^t$ since, in urban areas, the acceleration rate is assumed to fall below 0.5 m.s$^{-2}$ for speeds between 30km/h and the speed limit (usually 50km/h). Running conditions are thus considered as cruising.

### 2.4 Sound propagation model

The proposed approach can handle with any sound propagation model depending on the urban landscape. In this paper, multiple reflections and diffractions are neglected since the studied roundabout is in open area. The sources associated with each vehicle are mapped on a set of moving line sources whose angle $\Theta_n^t$ is defined by the positions of vehicle $n$ at the beginning and the end of the time-step. Although this noise source representation requires an attenuation calculation at each time-step, it was selected because of its accuracy. Note that other classical source representations such as point source or fixed line source could have also been chosen (Can et al., 2008a [26]). Subsequently, the instantaneous noise pressure level $L_{Aeq,\Delta t}^t$ at a receptor point $P$ is given by:
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\[ L'_{A_{eq,\Delta t}} = 10 \log \left( \sum_n \frac{\Theta_n^t}{2\pi d \|l_n\|} \|l_n\|_{10}^{\frac{L'_w}{10}} \right) \]  

(3)

where \(\|l_n\|\) is the length of the line source; \(d\) is the distance between \(P\) and the line source (see figure 3).

3 Noise emission model validation

3.1 Study site

To make a validation of the proposed dynamic noise emission model, a single-lane roundabout in the suburban area of Toulouse, France, was chosen as a study area. The goal is to check the relevance of the simulated dynamic noise contour maps by comparing observed and simulated sound pressure levels at several receivers. The emissions comparison is focused on one quarter of the roundabout with one receiver along the approach 4, one receiver between the approach 4 and the next departure link 1 and two receivers along the departure link 1 (see figure 5). Due to symmetry effects, it is assumed that if the simulated dynamic noise levels are accurate at these receivers, so they are along the other legs of the roundabout.

3.2 Data collection process

The studied site is a four-leg roundabout in an open area without high buildings in the surroundings as shown in figure 5. Traffic flow is moderate and essentially composed of light duty vehicles. Buses and heavy duty vehicles represent only 1.9% (respectively 0.9%) of the overall traffic in the morning (respectively in the evening) while motorcycles represent 1.3%. These figures justify the use of a single mean emission law to characterize all vehicles on the roundabout. The main movements are from entry 1 to exit 3 in the morning and from entry 3 to exit 1 in the evening. Note that receivers 1 and 2 were placed to measure how noise levels are affected by the disturbance effect of the main movement (1-3) on the stream coming from approach 4.

Two sets of traffic and acoustical data were collected on 2h-periods in the morning and in the evening peaks in October 2005. Traffic data was collected from a 15m-high video camera which has recorded movements of all vehicles. An image processing software (called AUTOSCOPE) was used to extract passing times and vehicle identities at about 4m before each entry (\(A_1\) to \(A_4\)) and after each exit (\(D_1\) to \(D_4\)) (see figure 5). Two additional positions were analysed on link 1 at respectively 30m and 60m from the yield line (\(A_{30}, A_{60}, D_{30}, D_{60}\)). In the microscopic
traffic simulation tool, incoming traffic on each approaching link is reproduced from the passing times at $A_{60}$, $A_2$, $A_3$ and $A_4$. Moreover, each created vehicle is assigned a destination thanks to the vehicle identities tracking at each entry and exit. The four microphones were located in the acoustic field at 1.2m high and at 5m to 6.5m from the road axis (see figure 5b). The sound pressure levels were recorded every one second ($\Delta t = 1s$) in global and in third-octave band spectra over the range 63-10 000Hz.

Fig. 5 – Data collection process

3.3 Parameters calibration

Table 2 summarizes the parameter values chosen for the microscopic traffic flow simulation model. They were fitted for matching observed traffic data or falling in the range of classical recommended values for urban conditions.

The free-flow speed $u$ on the approaching and departure arms has been set equal to the speed limit with an extra of 5% (to account for non-complying vehicles!). Inside the roundabout, it has been chosen according to the recorded vehicle passing times when traffic volume is low; note that it matches classical recommended values for single lane roundabouts usually comprised between 20km/h and 24km/h (Robinson, 2000 [27]). The desired acceleration rate $a$ has been calibrated from observed vehicle trajectories on the departure arm 1. Parameter $w$ is characteristic of a one-way urban link and can be measured by estimating the starting wave speed at a signalized intersection when signal turns green, for instance by video-recording the location the back of the queue in time (Leclercq, 2005 [28]). Parameter $s_0$ is also characteristic of a one-way urban link and can be determined from the average spacing between two stopped vehicles. An average value of 2km/h for the parameter $\epsilon$ was found to provide accurate vehicle trajectories in Leclercq et al. (2007b) [29] and has been chosen in this study since the relaxation phenomenon seems to be non-site-specific.

Parameters $t_f$ and $t_c$ were chosen so as to be in agreement with the recommended values of the American National Cooperative Highway Research Program (Rodergerdts et al., 2007 [30]) based on extensive validation studies on single-lane roundabouts:

- $t_f$ is usually comprised between 2.6s and 4.3s with an average of 3.2s and a standard deviation of 1.1s;
- $t_c$ is usually comprised between 4.2s and 5.9s with an average of 5.1s and a standard deviation of 1.3s.
Finally, parameter $\gamma$ has been set equal to 1 since, when the roundabout is congested vehicles have been observed to share priority with circulating vehicles given a one-by-one process.

| car-following | $u$ [km/h] | 44.6 | 54 | 19 |
| $s_0$ [m] | 4.8 |
| $w$ [km/h] | -10.6 |
| $\epsilon$ [km/h] | 2 |
| $a$ [m/s$^2$] | 1.3 |

| insertion decision | $t_f$ [s] | 3 |
| $t_{c} (d_{lag})$ [s] [m] | 5 (10.6) |
| $\gamma$ | no | 1 |

Tab. 2 - Calibration of the traffic flow simulation tool

3.4 Noise descriptors selection for the validation study

Prediction of the instantaneous noise pressure levels $L_{A_{eq},1s}$ offers a substantial breakthrough compared to analytical estimation models. Firstly, the energy-based descriptors can be calculated easily over any aggregation period. In the sequel, the equivalent sound pressure level, $L_{A_{eq}}$, will be computed over the entire 2h-simulation periods and over successive 15min-intervals. Secondly, statistical descriptors, $L_{N,T}$, reflecting the A-weighted pressure level exceeded during $N\%$ of the aggregation period $T$, can be evaluated. We will study the $L_{5,T}$, $L_{10,T}$, $L_{50,T}$ and $L_{90,T}$ on 2h and 15min-periods. Note that shortest aggregation periods were shown to be unreliable in Can et al. (2008b [31]) because of the descriptors’ sensitivity to noise peaks and the lack of statistical significance of the $L_{A_{eq}}$ sample. Thirdly, the distribution of the $L_{A_{eq},1s}$ over the entire simulation period can be obtained. This is a valuable tool for assessing the ability of the proposed model to reproduce dynamic on-field noise measurement. Finally, the time-series of $L_{A_{eq},1s}$ can be used to predict when the noisiest periods will occur. Particularly, one is able to pinpoint which traffic situations have the greatest impacts in noise levels. This can help in the choice of traffic management policies or infrastructure design.

3.5 Results

a Energy-based and statistical descriptors on 2h-periods

Table 3 sums-up the equivalent sound pressure level and the classical statistical descriptors aggregated over each 2h-simulation period, both for measurements and simulations. In all 40 cases but two, descriptors are estimated within a 2dB(A) error bound. The maximum deviation is 2.7dB(A). More specially, the deviation in $L_{A_{eq}}$ never exceeds 1dB(A) for all microphones except $M_3$ and $M_4$ in the morning peak where it reaches a 1.8dB(A) underestimation. A similar accuracy is observed for $L_{50}$. Higher noise levels corresponding to $L_5$ and $L_{10}$ are also reproduced with a precision of 1dB(A) for all microphones in the evening peak. The discrepancy is slightly higher in the morning peak with an underestimation lying from 0.8dB(A) to 2.1dB(A). Lowest noise levels characterized by $L_{90}$ fall within a 1dB(A) error bound in the morning peak. They are slightly overestimated from 0.8dB(A) to 2.7dB(A) in the evening peak.
\begin{table}[h]
\begin{center}
\begin{tabular}{|c|c|c|c|c|c| c|c|c|c|c|}
\hline
 & \multicolumn{5}{c|}{morning peak} & \multicolumn{5}{c|}{evening peak} \\
\hline
\text{db(A)} & \text{L}_{\text{Aeq}} & \text{L}_5 & \text{L}_{10} & \text{L}_{50} & \text{L}_{90} & \text{L}_{\text{Aeq}} & \text{L}_5 & \text{L}_{10} & \text{L}_{50} & \text{L}_{90} \\
\hline
\text{observed} & 62.4 & 67.2 & 65.7 & 60.2 & 56.1 & 62.9 & 67.3 & 66.0 & 60.7 & 56.4 \\
\text{M}_1\text{ simulated} & 62.6 & 66.4 & 64.2 & 59.8 & 55.1 & 63.0 & 67.3 & 65.3 & 61.5 & 57.2 \\
\text{deviation} & 0.2 & -0.8 & -1.6 & -0.4 & -1.1 & 0.04 & -0.02 & -0.8 & 0.8 & 0.8 \\
\hline
\text{observed} & 64.7 & 69.1 & 67.7 & 62.0 & 56.7 & 64.6 & 69.1 & 67.7 & 62.8 & 57.1 \\
\text{M}_2\text{ simulated} & 63.7 & 68.1 & 67.0 & 62.2 & 57.1 & 65.2 & 69.0 & 68.1 & 64.4 & 59.1 \\
\text{deviation} & -1.0 & -1.1 & -0.8 & 0.2 & 0.4 & 0.5 & -0.06 & 0.5 & 1.6 & 1.9 \\
\hline
\text{observed} & 66.7 & 71.7 & 70.0 & 63.4 & 55.4 & 65.8 & 70.8 & 69.3 & 63.1 & 55.8 \\
\text{M}_3\text{ simulated} & 65.1 & 70.5 & 69.1 & 62.9 & 55.5 & 65.6 & 70.5 & 69.5 & 63.7 & 57.3 \\
\text{deviation} & -1.6 & -1.2 & -0.9 & -0.6 & 0.04 & -0.2 & -0.3 & 0.1 & 0.6 & 1.5 \\
\hline
\text{observed} & 67.7 & 73.4 & 71.4 & 62.9 & 52.0 & 66.5 & 71.9 & 70.4 & 62.6 & 52.0 \\
\text{M}_4\text{ simulated} & 65.9 & 71.3 & 70.2 & 63.2 & 52.7 & 66.6 & 71.5 & 71.0 & 64.1 & 54.6 \\
\text{deviation} & -1.8 & -2.1 & -1.2 & 0.4 & 0.7 & 0.1 & -0.3 & 0.6 & 1.5 & 2.7 \\
\hline
\end{tabular}
\end{center}
\caption{Energy-based and statistical descriptors on 2h-periods}
\end{table}

\textbf{b Time-evolution of energy-based and statistical descriptors on 15min-periods}

The previous statistical descriptors are now aggregated over 15min-periods and their time-evolution is compared in figure 6.

\begin{figure}[h]
\begin{center}
\begin{subfigure}[b]{0.49\textwidth}
\centering
\includegraphics[width=\textwidth]{fig6a.png}
\caption{M$_1$}
\end{subfigure}\hfill
\begin{subfigure}[b]{0.49\textwidth}
\centering
\includegraphics[width=\textwidth]{fig6b.png}
\caption{M$_2$}
\end{subfigure}
\end{center}
\begin{center}
\begin{subfigure}[b]{0.49\textwidth}
\centering
\includegraphics[width=\textwidth]{fig6c.png}
\caption{M$_3$}
\end{subfigure}\hfill
\begin{subfigure}[b]{0.49\textwidth}
\centering
\includegraphics[width=\textwidth]{fig6d.png}
\caption{M$_4$}
\end{subfigure}
\end{center}
\end{figure}

\textbf{FIG. 6 – Time-evolution of noise descriptors over 15min-periods in the morning peak}
Dynamic noise modeling at roundabouts

Since flows are roughly constant during each 2h-peak period, statistical levels are not varying a lot from one aggregation period to the other. One can notice that the simulated results closely match the time-evolution of the observed levels. Particularly, the maximum deviation for all descriptors except $L_{90}$ never exceeds 3.5dB(A) in the morning peak and 2.7dB(A) in the evening peak. Most of the estimates usually fall within a 1dB(A) accuracy compared to measurement. Although trucks and buses are disregarded, the $L_5$ level is correctly estimated by the model, which confirms that such noisy vehicles were nearly absent from the study site. In other circumstances, they should be included to improve prediction of the highest noise levels. This could be achieved with the proposed dynamic noise prediction tool since: (i) the traffic flow model can be extended to represent the effects of heavy vehicles on traffic (Leclercq and Laval, 2007 [32]); (ii) specific French noise emission laws for heavy vehicles have been derived (Lelong, 2003 [33]). It should be worth also to notice that the model seems to be less effective in predicting the $L_{90}$ time-evolution on 15min-intervals even if the discrepancy never exceeds 4dB(A). This is not surprising since the model only accounts for road traffic noise and neglects disturbances coming from other sources such as wind, pedestrians or cyclists.

c Distributions of the instantaneous sound pressure levels

Figure 8 shows the distribution of the instantaneous sound pressure levels over both 2h-simulation periods at the four receptor points. Whatever the receiver, the shape of the simulated distribution is very close to the observed one. Particularly, the most frequent simulated levels perfectly match the most frequent observed modes. For the microphones $M_3$ and $M_4$ mostly affected by accelerating vehicles on the main departure link, one can see that the noisiest levels are usually under-represented by the model. Two reasons can explain this phenomena: (i) heavy vehicles and motorcycles which are expected to be noisier than light-duty vehicles, especially in

Figure 7 – Time-evolution of noise descriptors over 15min-periods in the evening peak
accelerating mode, are not accounted for in this study; (ii) random events like klaxons, road works or human voices in front of a microphone are disregarded. Moreover, for all receivers, the lowest levels are slightly over-represented. Indeed, between the passing by of vehicles, the simulated level drops to the background noise level. This explains the little peaks in the simulated distribution for levels around 47dB(A). In reality, low levels are more scattered due to noise coming from other sources.

**Fig. 8** - Statistical distribution of the instantaneous sound pressure levels in the morning peak

**Fig. 9** - Statistical distribution of the instantaneous sound pressure levels in the evening peak
d Noise contour maps

Building a rectangular grid of receivers with spacing of 4m allows us to generate dynamic noise contour maps of sound pressure levels at each time-step. They can be used to draw the time-evolution of noise contour maps of the $L_{Aeq}$ and all statistical descriptors over a given sub-period. Maps of the $L_{Aeq}$ and $L_{10}$ over the first 15min-period of the morning peak period are illustrated in figure 10. One can see that the highest levels of both $L_{Aeq}$ and $L_{10}$ occurs on the main link 1, at the exits and entries. They reach values of 65dB(A) for the $L_{Aeq}$ and of 70dB(A) for the $L_{10}$ at 5m from the road axis of link 1. Those values decrease on the circulating roadway and farther from link 1. Note that noise levels around link 4 are much lower because of fewer arriving and departing vehicles on this link.

![Noise contour maps](image)

**Fig. 10** - Noise contour maps of descriptors over the first 15min-period in the morning peak

Analysis of dynamic noise contour maps can also be valuable to identify the noisiest traffic situations. For instance, two situations, with about the same number of vehicles on the roundabout, are depicted in figure 11. On the left side, only one vehicle is approaching fast on link 4 since it is not disturbed by circulating vehicles. Consequently, the noise levels received close to link 4 are important. On the right side of the figure, three vehicles are queuing because of hindrance of circulating vehicles. Due to their very low speeds, the noise levels are lower than previously even if more vehicles are present on the roundabout. This example shows that vehicle interactions have a substantial effect on noise levels. Analytical methods which neglect them are therefore inconsistent.
4 Conclusion

The proposed dynamic noise emission model fills the shortage of accurate noise estimation procedures at roundabouts. The noise emissions due to stochastic vehicle interactions at roundabout entries can be fully captured by combining a microscopic traffic flow model with noise emission laws and propagation calculation. Contrary to other existing noise simulation packages, the merging process and its impact on kinematics, vehicle delays and queue lengths is accurately reproduced, whatever the traffic conditions. The traffic outputs are fed into relevant noise emission laws depending on the vehicle type and the prevailing road pavement type. From vehicle speed and acceleration an instantaneous noise power level can be calculated for each vehicle. These levels are input into a propagation model to compute the noise levels received at different points in the vicinity of the roundabout. Accuracy of the obtained noise contour maps was demonstrated by comparing the simulated and the observed noise pressure levels at several receivers along a suburban roundabout. Such maps could be used to draw noise difference contours where the effects of the junction are compared with noise levels if vehicles were freely moving. This could help in deriving correction factors for simpler analytical noise prediction models to adjust for the extra noise induced by the junction. Moreover, noise difference contours could also be plotted for different junction layouts or control types to help practitioners for choosing the best design for the road-crossing.

The roundabout noise emission model will be integrated into a more general noise simulation package of a whole network called SYMUBRUIT (Leclercq and Lelong, 2001 [34]). It will allow an explicit modeling of roundabouts while, up to now, SYMUBRUIT just treats junctions as pointwise intersections. Extension of the roundabout model to other types of unsignalized intersections as well as to non-priority turning movements at traffic signals would be worth investigating. It will complete the detailed modeling of intersections within SYMUBRUIT. Further validation studies should also be conducted to check the accuracy of estimated noise levels when several local streets, urban arterials and intersections interact together. This will open the door to better noise impacts assessment of local transport management policies such as junction
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control device, intersection layout or signal coordination plans.

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