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Influence of tree crown characteristics on the local PM_{10} distribution inside an urban street canyon in Antwerp (Belgium) : a model and experimental approach

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1 **Influence of tree crown characteristics on the local PM₁₀ distribution** 2 **inside an urban street canyon in Antwerp (Belgium): a model and** 3 **experimental approach**

4 **Keywords:**

5 Particulate matter (PM); Urban green; Light detection and ranging (LiDAR); Leaf deposition; Tree crown; Leaf
6 area density (LAD); Particle size fraction; Air pollution; Atmospheric modelling

7 **Abstract**

8 Apart from influencing the amount of leaf-deposited particles, tree crown morphology will influence
9 the local distribution of atmospheric particles. Nevertheless, tree crowns are often represented very
10 rudimentary in three-dimensional air quality models. Therefore, the influence of tree crown
11 representation on the local ambient PM₁₀ concentration and resulting leaf-deposited PM₁₀ mass was
12 evaluated, using the three-dimensional computational fluid dynamics (CFD) model ENVI-met[®] and
13 ground-based LiDAR imaging. The modelled leaf-deposited PM₁₀ mass was compared to gravimetric
14 results within three different particle size fractions (0.2 - 3, 3 - 10 and >10 μm), obtained at 20 locations
15 within the tree crown. Modelling of the LiDAR-derived tree crown resulted in altered atmospheric PM₁₀
16 concentrations in the vicinity of the tree crown. Although this model study was limited to a single tree
17 and model configuration, our results demonstrate that improving tree crown characteristics (shape,
18 dimensions and LAD) affects the resulting local PM₁₀ distribution in ENVI-met. An accurate tree crown
19 representation seems, therefore, of great importance when aiming at modelling the local PM
20 distribution.

21 **Capsule**

22 An accurate tree crown representation seems of great importance when aiming at modelling the local
23 PM distribution in urban environments.

24 1. Introduction

25 In contemporary cities, vegetation exerts many so-called ecosystem services. Apart from carbon
26 sequestration, micro-climate regulation, noise reduction, rainwater drainage, psychological and
27 recreational values (Bolund and Hunhammar, 1999; Jim and Chen, 2009; Li et al., 2010; Ulrich, 1984),
28 a significant amount of research has focussed on the air pollution mitigation potential of urban
29 vegetation (Beckett et al., 1998; Litschke and Kuttler, 2008; McPherson et al., 1997; Nowak et al., 2006;
30 Vos et al., 2013). Because of its high surface area, relative to the ground it covers, vegetation (especially
31 trees) can influence local atmospheric particulate matter (PM) concentrations through both direct and
32 indirect pathways. While vegetation can lower ambient particle concentrations by stimulating
33 deposition on its surfaces (direct pathway), it can also affect wind flow and, therefore, the dispersion
34 of PM polluted air (indirect pathway) (Langner, 2008). Research indicated that this indirect effect is
35 able to reduce, but also increase local atmospheric particle concentrations (Gromke and Ruck, 2007;
36 Vos et al., 2013). The impact of urban vegetation on air pollution is frequently estimated using three-
37 dimensional (3D) air quality models in which urban vegetation is often represented very rudimentary,
38 resulting from a lack of species-specific information (e.g. leaf area density (LAD), crown dimensions,
39 tree height, ...). Especially in urban environments, where species-specific tree crown morphology will
40 depend on local environmental conditions (light-, water-, nutrient- and space availability) and
41 experienced pruning frequency. Therefore, experimental datasets, including a thorough description of
42 the vegetation inside urban areas, are needed to improve existing dispersion models (Janhäll, 2015).
43 Most dispersion models include urban vegetation as spherical shapes or rectangular blocks with
44 theoretical leaf area densities (Buccolieri et al., 2011; Gromke and Ruck, 2007; Litschke and Kuttler,
45 2008; Tong et al., 2016; Vos et al., 2013; Wania et al., 2012).

46 Nevertheless, as previous studies already showed a clear influence of tree foliation (Hofman et al.,
47 2013; Salmond et al., 2013) and spatial implantation of urban green on local pollutant dispersion
48 (Buccolieri et al., 2011; Gromke and Ruck, 2012, 2007; Vos et al., 2013; Wania et al., 2012), an accurate

49 representation of urban tree crowns is hypothesized vital for pollutant dispersion modelling.
50 Therefore, this study evaluates the effect of a detailed tree crown representation on the ambient PM₁₀
51 concentration and the resulting leaf-deposited PM₁₀ mass, using the 3D computational fluid dynamics
52 (CFD) model ENVI-met[®]. Moreover, modelled leaf-deposited PM₁₀ results were compared to
53 gravimetric results obtained from leaf samples of a London plane tree (*Platanus x acerifolia* Willd.).

54 2. Material and Methods

55 2.1 Sampling site

56 A typical urban London plane tree (*Platanus x acerifolia* Willd.) was selected in the densely populated
57 city center of Antwerp, Belgium (Figure 1). The solitary tree is located on a roundabout (51°11'47.72"N,
58 4°25'24.81"E) in line with the *De Villegasstraat* street canyon, described in a previous study (Hofman
59 et al., 2014a). While the roundabout itself is relatively quiet in terms of traffic volume (50 vehicles h⁻¹
60 (SGS, 2010)), it is located in the vicinity (200 and 400m) of two busy thoroughfares of Antwerp (*Singel*
61 and *R1*), as can be seen from Figure 1.

62 {Figure 1}

63 **Figure 1:** Left panel: Location of the street canyon, “Singel” and ring road (R1) in Antwerp (source: Google). Right
64 panel: photographic image (upper left; source: Google) and LiDAR scan (upper right) of the considered tree crown,
65 with a sideview (lower) of the tree LiDAR hits and a plot of the number of LiDAR hits (x-axis) as a function of height
66 (y-axis in meters).

67

68 2.2 Gravimetric analysis of leaf samples

69 Quantitative information on the amount of leaf-deposited atmospheric particles was obtained by
70 means of a leaf sampling campaign, according to the protocol described in Hofman et al. (2014a). Since
71 differences in leaf Saturation Isothermal Remanent Magnetisation (SIRM) become more pronounced
72 as the growing season proceeds due to a prolonged exposure time (Hofman et al., 2014b; Kardel et al.,

73 2011), leaf samples were collected at the end of the growing season, i.e. on September 11, 2012. The
74 tree was sampled at the outer surface by means of a boom lift at three heights (3.5, 8.5 and 13.5 m)
75 and four azimuthal directions (NE, NW, SE, SW) around the tree crown. Next to leaf samples of the
76 outer tree crown surface, leaf samples were collected from within the tree crown as well, at half the
77 distance between the trunk and the crown outer surface, at two heights (3.5 and 8.5 m) and four
78 azimuthal directions (NE, NW, SE, SW), resulting in 20 leaf sampling locations. Three leaf samples were
79 collected for each leaf sampling location, with each sample consisting of five fully developed and
80 undamaged leaves. The leaf samples were separately collected in paper bags, labelled and transported
81 to the laboratory for analysis.

82 In the laboratory, the leaf samples were processed according to the procedure described in Hofman et
83 al. (2014a). Each leaf sample was hand-washed in ultrapure water ($<0.1 \mu\text{S cm}^{-1}$) (Eurowater, Belgium).
84 The washed leaf area was determined using a Li-3100 area meter (LI-COR Environmental, US). The
85 washing water was subsequently filtered over pre-weighed Nuclepore track-etched polycarbonate
86 filter membranes (Whatman, UK) with pore sizes of in succession $10 \mu\text{m}$, $3 \mu\text{m}$ and $0.2 \mu\text{m}$. Doing so,
87 we collected three particle size fractions on the filter membranes: large ($>10 \mu\text{m}$), coarse ($3\text{-}10 \mu\text{m}$)
88 and fine ($0.2\text{-}3 \mu\text{m}$). Loaded filter membranes were consequently dried at ambient temperature,
89 equilibrated for 24 h at 50% relative humidity and weighed using a $1 \mu\text{g}$ precision Mettler MT5 balance
90 (Mettler-Toledo International Inc., Switzerland). To avoid electrostatic charges on the filters, they were
91 passed through an ionizer antistatic system (Mettler-Toledo International Inc., Switzerland) before
92 weighing. Blank filter weights were determined for every size fraction by completing the entire
93 filtration procedure using only Ultrapure water. The blank filter weight was consequently subtracted
94 from the loaded filter weight. The resulting weights were finally normalized for washed leaf area
95 yielding the leaf-deposited particulate weight (mg m^{-2} leaf area) in three different size fractions (>10
96 μm , $3\text{-}10 \mu\text{m}$ and $0.2\text{-}3 \mu\text{m}$).

97 2.3 ENVI-met[®] model configuration

98 Two model studies were conducted in ENVI-met[®] to, respectively, (i) evaluate the effect of a detailed
99 three-dimensional tree crown representation on the modelled atmospheric PM₁₀ concentration and
100 resulting leaf-deposition, and (ii) quantify the amount of leaf-deposited PM₁₀ for the representative
101 exposure period of the collected leaf samples (May to September 2012).

102 ENVI-met[®] is a 3D computational fluid dynamics (CFD) model developed by Bruse & Fleer (1998). It is
103 Reynolds Averaged Navier-Stokes (RANS) equations based, non-hydrostatic micro scale obstacle-
104 resolving with advanced parameterizations for the simulation of interactions between urban surfaces,
105 vegetation and atmosphere (Bruse and Fleer, 1998). ENVI-met was selected because of its ability to
106 simulate the influence of plants on the atmosphere in a built environment at micro scale. This is useful
107 since plants will not only alter the distribution of particulates, but also intercept particles on their
108 surfaces. In ENVI-met[®], sedimentation of particles due to gravitational forces and particle deposition
109 to different surfaces (roofs, walls, vegetation and soils) is simulated, taking into account the
110 aerodynamic and sub-layer (quasi laminar) surface resistances (Bruse, 2007). The initial variables
111 required as an input in ENVI-met[®] are meteorological data, pollutant emissions and domain
112 characteristics (Nikolova et al., 2011).

113

114 2.3.1 Three-dimensional tree canopy structure

115 In ENVI-met[®], each plant is represented as a one-dimensional permeable column that is subdivided
116 into multiple LAD layers, expressed as $m^2 m^{-3}$ (Wania et al., 2012). To account for the influence of
117 vegetation on atmospheric processes, all prognostic equations in the model are extended into the
118 vegetation layers using source/sink terms describing heat, humidity and momentum exchanges (Wania
119 et al., 2012). The loss of wind speed due to vegetation friction is parameterized in airflow equations,
120 while additional turbulence and dissipation due to vegetation are integrated in the turbulence

121 equations. The sensible heat flux, evaporation flux and transpiration flux are calculated in the
122 vegetation module.

123 To test for the effect of a detailed tree crown representation on the ambient PM₁₀ concentration and
124 the resulting leaf-deposited PM₁₀ mass, two model runs were performed. While the first run was
125 conducted with a theoretical tree crown representation, the second run was performed with the actual
126 tree crown representation assessed with ground-based Light Detection And Ranging (LiDAR).

127 Within ENVI-met[®], vertical LAD profiles are normalized from $z h^{-1} = 0.1$ (LAD1) to $z h^{-1} = 1$ (LAD10),
128 where z is the height of the LAD entry and h is the total plant height (m). Each vertical profile thus
129 consisted of 10 different horizontal LAD layers. For the theoretical tree crown representation, we used
130 8 unique vertical LAD profiles (Table 1) to be able to adjust the tree crown dimensions in relation to
131 the distance to the center of the crown. The applied LAD values ($0.1-1.15 \text{ m}^2 \text{ m}^{-3}$) were based on
132 standard deciduous LAD values provided in the local ENVI-met[®] database consisting of 27 different
133 plant structures (Bruse, 2012). The resulting average Leaf Area Index (LAI) of the theoretical tree crown
134 ($5.12 \text{ m}^2 \text{ m}^{-2}$) falls within the limits of, although scarce, reported averages ($0.3-7 \text{ m}^2 \text{ m}^{-2}$) for London
135 Plane (*Platanus x acerifolia* Willd.) within urban environments (OCA, 2007; Wu et al., 2008).

136

137 **Table 1:** Vertical LAD profiles in ENVI-met[®] are normalized from $z h^{-1} = 0.1$ (LAD1) to $z h^{-1} = 1$ (LAD10), where z
138 is the height of the LAD entry and h is the maximal height (m). Each vertical profile thus consisted of 10 different
139 horizontal LAD layers (LAD1-LAD10). Overview of the applied vertical LAD profiles (P1-P8) to describe the
140 theoretical tree crown structure with maximal height (m) of each vertical profile and LAD values ($\text{m}^2 \text{ m}^{-3}$) for the
141 individual horizontal LAD layers (LAD1-LAD10). The sum of all LAD layers over the maximal height results in the
142 LAI ($\text{m}^2 \text{ m}^{-2}$) of each LAD profile.

143

{Table 1}

144 Ground-based Light Detection and Ranging (LiDAR) was applied to improve the three-dimensional tree
145 crown representation inside the model. This is an active remote sensing technology which provides a
146 novel tool for generating a comprehensive and detailed 3D mathematical description of tree and
147 canopy structure in a non-destructive manner (Van Der Zande, 2008). These state of the art systems
148 are mobile, robust and small and can be used for diverse applications e.g. airborne topographic
149 mapping, surveying of buildings and plants, etc. (Van Der Zande, 2008). In this study we used the RIEGL
150 VZ-400 ground-based scanner (RIEGL Laser Measurement Systems GmbH, Horn, Austria). This time-of-
151 flight scanner has a range up to 350 m and a beam divergence of nominally 0.35 mrad and operates in
152 the near infrared (wavelength 1550 nm). The RIEGL VZ-400 scanner records multiple returns (up to
153 four returns per emitted pulse), with returns being derived from on-board waveform processing.
154 Multiple returns will lead to an improved sampling at greater canopy heights, which is of interest for
155 monitoring vegetation (Calders et al., 2015, 2014; Lovell et al., 2003). The angular resolution in both
156 zenith and azimuth direction was set to 0.06 degrees.

157 LiDAR data were acquired prior to the leaf sampling campaign, on September 7, 2012, to avoid
158 potential effects of structural tree crown damage during the leaf sampling campaign. Different scan
159 locations were used with reflecting targets, distributed throughout the scene. These targets were used
160 to register each individual scan location to a single registered pointcloud with the RiSCAN PRO software
161 (provided by RIEGL). The points belonging to the individual tree were manually selected from the
162 pointcloud. Next, this subset was filtered, by removing all individual points that had a distance of ≥ 10
163 cm to each other point were removed in order to reduce noise. Further, due to the setup of the
164 different scan locations some parts of the trees may be oversampled (e.g due to overlapping scans and
165 shorter distance to the scanner). Therefore, the pointclouds were converted to 5x5x5 cm, where a
166 voxel is considered filled if it contains a return. The tree crown domain was applied to the model grid
167 size resolution (1x1x1 m) and the number of filled voxels per grid cell was obtained. Theoretically, it is
168 possible to derive the LAD from the number of filled voxels using the approach of Hosoi and Omasa
169 (2006), but then additional information on the leaf inclination and the number of laser beams per voxel

170 is required which is difficult to assess when multiple laser scans are combined. We therefore
171 normalized the number of filled voxels to the LAD values of the theoretical tree crown by equating the
172 maximal number of filled voxels per grid cell (10401) to the maximal LAD ($1.15 \text{ m}^2 \text{ m}^{-3}$) used in the LAD
173 profiles of the theoretical tree crown (Table 1). As a result, we obtained 206 unique LAD profiles.
174 Because the number of unique vertical LAD profiles in ENVI-met is restricted to 150, we reduced the
175 number of profiles by averaging LAD profiles in the center of the tree crown (Figure 2). In fact, the
176 LiDAR-derived data reflects the Plant Area Index (PAD), as no distinction was made between woody
177 (stem, branches and twigs) and leaf material. The use of PAD is justified as woody tree material will
178 influence local air flows and has shown to act as accumulation surface for atmospheric particles
179 (Catinon et al., 2011, 2009; Huhn et al., 1995; Sawidis et al., 2011).

180 {Figure 2}

181 **Figure 2:** Top view of the applied vertical LAD profiles for the theoretical tree crown (left), based on
182 eight unique vertical LAD profiles (P1-P8), and the LiDAR-derived tree crown (right), based on 150
183 unique vertical PAD profiles (A1-W9), in the ENVI-met[®] Area Input file Editor.

184 2.3.2 Model domain

185 Within this chapter, two separate model studies were evaluated. In a sensitivity study, we considered
186 the effect of tree crown structure within a theoretical three-dimensional model domain of 100x100x50
187 m and a 1x1x1 m equidistant grid cell resolution. One model run was performed with the theoretical
188 spherical tree crown consisting of 8 unique LAD profiles while a second run was conducted with the
189 improved tree crown representation based on 150 unique LiDAR-derived LAD profiles (Figure 2). While
190 the theoretical tree crown had an average Leaf Area Index (LAI) of $5.12 \text{ m}^2 \text{ m}^{-2}$, a LAI of $7.79 \text{ m}^2 \text{ m}^{-2}$
191 was obtained for the LiDAR-derived tree crown (Figure 3). This results from the applied procedure to
192 transform the filled voxel numbers to LAD values using the maxima equation, and the inclusion of the
193 woody tree material using the LiDAR procedure.

194

{Figure 3}

195 **Figure 3:** Model domain of the sensitivity study in which the effect of tree crown structure on the ambient PM_{10}
196 concentration and PM_{10} leaf deposition was tested. A theoretical spherical shaped tree crown (left) was compared
197 to a detailed LiDAR-derived tree crown (right).

198 In the second model study (real-life study), the leaf-deposited PM_{10} mass was simulated throughout
199 the entire in-leaf season using a model domain that represented the actual situation. As the size of the
200 model domain is directly related to its computational time, we first evaluated the need to incorporate
201 two busy thoroughfares of Antwerp *R1* and *Singel* inside the real-life model domain. To do so, we
202 considered a slim three-dimensional model domain of 750x250x50 m with an equidistant grid cell
203 resolution of 50 m³ (5x5x2 m). To simulate the maximal potential influence, we modelled maximal
204 (rush hour) traffic intensities at 90% of the maximal hourly wind speed obtained during the growing
205 season (3.04 m s⁻¹), perpendicular on the *R1* and *Singel* in the direction of the considered tree crown.
206 As can be seen from Figure 4, the emission plume of the *R1* and *Singel* only resulted in a limited (<0.30
207 $\mu\text{g m}^{-3}$) local increase of the atmospheric PM_{10} concentration without any vertical stratification at the
208 considered tree crown location. Considering the recessed nature of the *R1* road below ground level,
209 the actual influence of the emission plume will even be overestimated in this model scenario. We,
210 therefore, assumed that both roads do not influence the spatial leaf-deposited PM_{10} variation within
211 the considered tree crown, and can be excluded from the final model domain.

212

213

{Figure 4}

214 **Figure 4:** Model results to evaluate the influence of the nearby two busy thoroughfares of Antwerp, i.e. ring road
215 and *Singel*, on the atmospheric PM_{10} concentration in the vicinity of the considered tree crown.

216 After evaluating the effect of tree crown structure (sensitivity study) and vicinity of *R1* and *Singel*, we
217 constructed the model domain for the real-time study. We, therefore, made use of the detailed LiDAR-

218 derived tree crown in a three-dimensional model domain of 239x239x30 m with a 1x1x1 m grid cell
219 resolution (Figure 5). The equidistant mesh consisted of grid cells with identical heights ($H = 1$ m),
220 except for the lowest five grid cells (near soil/pavement simulation) which had a smaller vertical
221 extension ($0.2 \times H = 0.2$ m). The building height inside the model domain was obtained from the LiDAR
222 based Digital Height Model (DHM) of Flanders (Flemish Environment Agency). To avoid edge effects, 5
223 nesting grids were considered at each border of the model domain, which contained a total of 1.71
224 million grid cells.

225 {Figure 5}

226 **Figure 5:** Model domain of the final model study with the detailed LiDAR derived tree crown and the urban
227 morphology of the surrounding buildings.

228 2.3.3 Meteorology

229 The in-leaf period (May 1st to September 11, 2012) was characterised by a mean air temperature of 16
230 °C, relative air humidity of 73%, vector averaged wind direction and speed of 277° and 1.40 m s^{-1} and
231 a total cumulative precipitation of 370 mm. As dry deposition of atmospheric particles on leaf surfaces
232 has been shown to occur throughout the entire in-leaf season, ENVI-met[®] simulations should ideally
233 cover all environmental variation that is experienced during this period. Meteorological data were,
234 therefore, obtained from the closest measuring station of the Flemish Environment Agency (VMM),
235 located at 7 km from the considered tree crown, for the period from May 1st to September 11, 2012.
236 Model simulations of an entire in-leaf season are, however, not feasible due to calculation time
237 restrictions. We therefore, simulated the entire in-leaf season with average atmospheric temperature
238 (°C) and relative atmospheric humidity (%) at the hour of model initialisation (8am), while average
239 wind direction and speed were considered for different wind sectors. As it is widely established by now
240 that wind is the most important meteorological factor determining local particle dispersion in urban
241 environments (Hofman et al., 2016, 2013; Janhäll, 2015; Kozawa et al., 2012; Kumar et al., 2011), we
242 conducted model runs for eight wind sectors (N, NE, E, SE, S, SW, W, NW) with averaged wind speed

243 within each wind sector (see Table 2). Wind speed measurements were obtained at a height of 30 m
244 while ENVI-met requires wind speed at 10 m. Therefore, wind speed at 10 m was derived from the
245 measured wind speed using the logarithmic wind velocity profile originally described by Priestley
246 (1959) and adapted for flow over a disturbed surface (equation 1):

$$247 \quad \frac{u}{u^*} = \frac{1}{k} \ln \left(\frac{z-z_d}{z_0} \right) \quad (1)$$

248 where u is the wind speed (m s^{-1}), u^* is the friction velocity, k is the von Karman constant (≈ 0.41), z is
249 the considered height (m), z_d zero displacement height (m) and z_0 is the roughness length (m). The
250 roughness length was considered to be 0.1 m following the parameterization of Britter and Hanna
251 (2003) for a moderately open urban surface. The displacement height is set to be equal to 0.5 times
252 the nearest building height (6 m) to the considered tree, following Grimmond and Oke (1999).

253 **Table 2:** Wind flow characteristics for each of the eight considered wind sectors (N, NE, E, SE, S, SW, W, NW).
254 Sector range ($^\circ$), considered wind direction inside the model ($^\circ$), average wind speed (m s^{-1}) during the in-leaf
255 season and time (h) that wind blew in this sector throughout the in-leaf season.

256 {Table 2}

257 Average measured meteorological data are only used in the initialization phase of the model during
258 which meteorological parameters are calculated up to a height of 2500 m (1D model). Thereafter,
259 initial 3D model parameters are calculated based on surface solar exposition and inclination (sky-view
260 factor) of the 3D terrain resulting in thermal stratification, surface temperature and disturbed wind
261 flow field.

262 After the model runs were performed for each wind sector, the modelled hourly leaf-deposited PM_{10}
263 loads for the 20 leaf sampling locations were summed up over the entire in-leaf season using the
264 number of hours within each wind sector (Table 2).

265 2.3.4 Emissions

266 Pollution sources in ENVI-met[®] are defined in the model domain and can be line, point or area sources
267 with hourly emission rates. Traffic emissions are calculated by multiplying an emission factor (g km^{-1})
268 with the prevailing traffic intensity (vehicle s^{-1}) (Ketzel et al., 2007). Emissions were calculated
269 separately for every street inside the model domain using vehicle type-specific traffic intensities and
270 emission factors. The traffic intensity was obtained from a traffic model (SGS, 2010) specifically
271 developed for the city of Antwerp and previously applied in a study of Lefebvre et al. (2011). The model
272 generates street-specific traffic intensity, expressed as $\text{vehicles hour}^{-1}$, for light, medium and heavy
273 traffic. Emission factors were obtained from the most recent Environmental Impact Report (LNE, 2012)
274 of the Flemish Environment, Nature and Energy department. In this report, emission factors for PM_{10}
275 (g km^{-1}) are provided for each vehicle type (light, medium and heavy), taking into account driving
276 speed. The resulting hourly averaged line source emissions ($\mu\text{g s}^{-1} \text{m}^{-1}$) were then calculated for every
277 street in the model domain (Table 3) and appended in the source database. As described in 2.2.3, the
278 traffic-intensive *R1* and *Singel* roads were not included in the final domain since they showed a limited
279 influence on the local atmospheric- and leaf-deposited PM_{10} concentrations at the considered tree
280 crown location.

281 **Table 3:** *The applied PM_{10} emission factors for light/medium/heavy traffic (Emission factor in g km^{-1}), kind of*
282 *traffic intensity (Traffic), sum of light, medium and heavy traffic intensity (Total Traffic intensity in veh h^{-1}), and*
283 *the resulting PM_{10} emission strength (PM_{10} emission in $\mu\text{g s}^{-1} \text{m}^{-1}$) of the considered streets in the model domains.*

284 {Table 3}

285 3. Results and Discussion

286 Two model studies were conducted in ENVI-met[®] to, respectively, (i) a sensitivity study to evaluate
287 the effect of a detailed tree crown representation on the modelled atmospheric PM_{10}
288 concentration and resulting leaf-deposition, and (ii) a real-life study to quantify the amount of leaf-
289 deposited PM_{10} for the representative exposure period of the collected leaf samples (May to
290 September 2012).

291 3.1 Sensitivity study

292 The sensitivity study consisted of an evaluation of the effect of considering a detailed LiDAR-derived
293 tree crown (150 unique LIDAR profiles) on the resulting modelled atmospheric and leaf-deposited PM₁₀
294 concentrations in ENVI-met. As can be seen from Figure 6, the LiDAR-derived tree crown resulted in an
295 altered local atmospheric PM₁₀ concentration when compared to the theoretical tree crown. Applying
296 the LiDAR-derived tree crown, local PM₁₀ concentrations are increased up to 34% at a height of 3.5 m
297 (and 62% at a height of 0.1 m) relative to the theoretical tree crown structure. As can be seen from the
298 relative flow differences (vectors) in Figure 6, the increased PM₁₀ concentration was due to a wind field
299 alteration at the downwind side of the tree crown. By using the LiDAR based methodology, not only
300 the total LAI of the tree crown increased (from 5.12 to 7.79 m² m⁻²), making it less permeable to wind
301 flow, but also the roughness of the crown surface increased giving rise to disturbance of the flow field
302 (turbulence) along the edges of the canopy, especially at the leeward side of the tree crown.

303 {Figure 6}

304 **Figure 6:** Relative difference (%) in atmospheric PM₁₀ concentration and wind flow between the LiDAR-derived
305 tree crown and the theoretical tree crown (reference) at a height of 3.5 m. The wind rose depicts the north
306 azimuth (black arrow) and simulated wind direction (red arrow).

307 The capacity of urban green to alter local pollutant concentrations was previously demonstrated in
308 several modelling and wind tunnel studies (Buccolieri et al., 2011, 2009; Gromke and Ruck, 2012, 2007;
309 Janhäll, 2015; Ries and Eichhorn, 2001; Vos et al., 2013; Wania et al., 2012). Trees in urban street
310 canyons have shown to obstruct the wind flow, thereby reducing pollutant dilution (ventilation),
311 resulting in higher pollutant concentrations. Vos et al. (2013) modelled the effect of 17 different urban
312 vegetation scenarios, including trees, hedges and green barriers, on the resulting atmospheric PM₁₀,
313 PM_{0.2} and NO₂ concentrations in ENVI-met. They predominantly observed a deteriorating air quality
314 due to the considered vegetation scenarios, mainly determined by the type, height and permeability
315 (=LAD) of the vegetation. Besides inhibition of the street canyon ventilation, vegetation was also found

316 to reduce wind speed at crown-height and to disrupt the flow field in close vicinity to the canopy in an
317 ENVI-met model study by Wania et al. (2012). Previous wind tunnel experiments and numerical
318 simulations (FLUENT) by Gromke and Ruck (Gromke and Ruck, 2012, 2009, 2007) and Buccolieri et al.
319 (Buccolieri et al., 2011, 2009) confirmed the aerodynamic effect of vegetation on in-canyon pollutant
320 concentrations, with denser tree crowns leading to lower wind speeds and higher resulting pollutant
321 concentrations. Increased foliage density (LAD) typically results in higher pollutant concentrations,
322 obtained at the leeward side of the street canyon, which is in agreement to our results.

323 It is now commonly believed that the aerodynamic effect exerted by urban vegetation outweighs its
324 pollutant removal capacity, which pleads for thoughtful consideration during the spatial planning of
325 urban green. Nevertheless, observed air quality outcomes due to vegetation are still less pronounced
326 than outcomes due to street canyon geometry (e.g. height-width ratio) or wind fields.

327 Looking at the resulting leaf-deposited PM_{10} load, the elevated atmospheric PM_{10} concentration results
328 in elevated deposition loads of up to 18% at a height of 3.5 m ($0.44 \mu\text{g m}^{-2}$ after 6 hours) on the leaves
329 at the downwind side of the tree crown (Figure 7).

330 {Figure 7}

331 **Figure 7:** Relative difference (%) in leaf-deposited PM_{10} load and wind flow between the LiDAR-derived tree crown
332 and the theoretical tree crown (reference) at a height of 3.5 m. The dark blue color (-100%) is due to the microscale
333 difference in tree crown dimensions. The wind rose depicts the north azimuth (black arrow) and simulated wind
334 direction (red arrow).

335 As can be seen from Figure 8, a clear relation ($R^2 \approx 1$) exists between the atmospheric PM_{10}
336 concentration and the leaf-deposited PM_{10} mass when model results are extracted for the tree crown
337 grid cells at eight heights ($n=1075$). This relation was previously described in Hofman & Samson (2014)
338 and is clearly demonstrated by the relative tree crown differences in Figure 6 and 7. We must, however,
339 note that no resuspension of leaf-deposited particles, nor saturation of the particle loading capacity of

340 the leaf surface is simulated by ENVI-met. The deposited PM_{10} mass might, therefore, be
341 overestimated in the model simulations.

342 {Figure 8}

343 **Figure 8:** Relation between the modelled atmospheric PM_{10} concentration ($\mu g m^{-3}$) and the amount of leaf-
344 deposited PM_{10} ($\mu g m^{-2}$) after six hours of simulation time with associated regression line (full line). Modelled
345 results ($n=1075$) were extracted for the tree crown grid cells at heights of 0.5, 1.5, 2.5, 3.5, 6.5, 8.5, 10.5 and 13.5
346 m.

347 Although this model study was limited to a single tree and model configuration, our results show that
348 improving tree crown characteristics (shape, dimensions and LAD) affects the local distribution and
349 resulting leaf-deposition of PM_{10} in ENVI-met. An accurate tree crown representation seems,
350 therefore, of great importance when aiming at modelling local PM distributions.

351 3.1 Real-life study

352 For the real-life simulations, the atmospheric PM_{10} concentration in the vicinity of the LIDAR-derived
353 tree crown differed considerably between the different wind sectors (Figure 9). This is not surprising
354 as atmospheric PM_{10} emissions, originating from the considered roads (with differing emission
355 strengths) inside the model domain, will contribute differently under varying wind directions.

356 {Figure 9}

357 **Figure 9:** Modelled atmospheric PM_{10} concentration ($\mu g m^{-3}$) and wind vectors in the vicinity of the detailed LiDAR-
358 derived tree crown for the considered wind sectors (N, NE, E, SE, S, SW, W, NW). Grey areas in the model domain
359 represent buildings.

360 Modelled hourly PM_{10} depositions ($\mu g m^{-2} h$) at every leaf sampling location were integrated over the
361 entire in-leaf season using the number of hours within each wind sector (see Table 2). The resulting
362 modelled in-leaf season leaf-deposited PM_{10} load varied from 113.57 to 547.36 $\mu g m^{-2}$. The modelled
363 amount of leaf-deposited particles decreased with 74% between a height of 3.5 and 13.5 m (Figure 10)

364 and leaves at the inner crown locations accumulated less particles (on average 94%) compared to
365 leaves at the outer canopy.

366 {Figure 10}

367 **Figure 10:** Azimuthal (SE, SW, NW and NE) variation of modelled (black) and average gravimetric (grey) leaf-
368 deposited PM_{10} ($mg\ m^{-2}$) for the outer bottom (b), middle (m) and top (t) locations in the tree crown. Standard
369 deviations of gravimetric averages are denoted by the error bars. Note the different y-axis limits.

370 The obtained model results of the real-time simulation (§3.2) were compared to the gravimetric results
371 of the filter membranes obtained from the leaf sampling campaign (§2.1). The average gravimetrically-
372 determined mass of leaf-deposited particles for the 20 leaf sampling locations was 626.13, 28.72 and
373 33.69 $mg\ m^{-2}$ for respectively the $>10\ \mu m$, 3-10 μm and 0.2-3 μm size fraction. While the average total
374 leaf-deposited mass was 687.93 $mg\ m^{-2}$, the average mass within the PM_{10} size fraction (0.2-10 μm)
375 amounted 62.41 $mg\ m^{-2}$. These results are comparable to other studies that quantified the amount of
376 leaf surface-deposited particles (Dzierzanowski et al., 2011; Hofman et al., 2014a; Sæbø et al., 2012;
377 Terzaghi et al., 2013). Apart from the leaf surface-deposited particles, previous studies found that part
378 of the leaf-deposited particle load will be encapsulated inside epicuticular wax layers (Hofman et al.,
379 2014a; Terzaghi et al., 2013). In Hofman et al. (2014b), it was shown that the biomagnetic signal of this
380 fraction accounted for 33% of the total leaf *Saturation Isothermal Remnant Magnetisation* (SIRM). Our
381 results yielded a similar 34% representation of the leaf-encapsulated fraction (washed leaves),
382 compared to the total leaf biomagnetic (SIRM) signal. Moreover, due to aggregation of smaller (<10
383 μm) particles, part of the smaller size fractions will be retained on the largest (10 μm) filter membrane.
384 This hypothesis was confirmed by scanning electron microphotographs of the PM loaded filter
385 membranes (Figure 11). We can, therefore, assume that the amount of leaf-deposited $<10\ \mu m$ particles
386 is still underestimated with the methodology used in our study.

387 {Figure 11}

388 **Figure 11:** Scanning electron microphotographs (back-scattered electron images) of the PM loaded 10 μm (left)
389 and 0.2 μm (right) filter membranes. Note the clear 10 μm -sized filter pore (left).

390 Comparing the gravimetric results for PM_{10} (sum 3-10 and 0.2-3 μm) and PM_{total} (sum >10, 3-10 and
391 0.2-3 μm) with the modelled seasonal PM_{10} deposition for all 20 leaf sampling locations in the tree
392 crown, an overall weak agreement ($R^2 = 0.0004\text{-}0.14$) is observed (Figure 12). Spearman Rank tests
393 show no significant correlations between the modelled leaf-deposited PM_{10} mass and the gravimetric
394 PM_{10} ($r=-0.18$, $p=0.43$) and PM_{total} ($r=0.31$, $p=0.18$) results.

395 {Figure 12}

396 **Figure 12:** Relation between the modelled leaf-deposited PM_{10} mass (mg m^{-2}) and the gravimetric results for the
397 mass of the PM_{10} (sum 0.2-3 and 3-10 μm) and PM_{total} (sum 0.2-3, 3-10 and >10 μm) size fractions. Every point
398 represents a sampling location in the tree crown ($n=20$). Regression lines with associated functions and R^2 -values
399 are denoted by the full lines.

400 While the leaf-deposited PM_{10} mass differs by two orders of magnitude between the modelled and
401 gravimetric results, the spatial variation between modelled and measured results seems to differ as
402 well. A 6% decrease of the gravimetric leaf-deposited mass is obtained between a height of 3.5 and
403 13.5 m and on average 44% higher leaf-deposits are found at the inner crown locations compared to
404 the crown's outer edge locations. Considering the azimuthal variation between the sampling locations
405 at the different heights (Figure 10), the gravimetric results appear to differ from the modelled results
406 as well. Significant higher gravimetric results at the SE side at the bottom and middle of the tree crown
407 might suggest an influence of the *Singel* and ring road (two busy thoroughfares of Antwerp in the
408 neighbourhood of the tree), as was also suggested by Hofman et al. (2014a), while the modelled results
409 suggest the *Statiestraat* as the main particle source at the NW side of the bottom sampling locations
410 (Figure 10). Although the model analysis in §2.2.2 estimated a limited influence of the ring road and
411 *Singel* ($<0.30 \mu\text{g m}^{-3}$), the low traffic intensity of the local streets seems to lead to rather low PM
412 concentrations ($<0.40 \mu\text{g m}^{-3}$) in the vicinity of the tree crown as well, as can be seen in Figure 9.

413 The lack of agreement between the modelled and gravimetric results can be explained by the modelled
414 sources and applied model parameters. Our model studies namely only included local traffic sources,
415 to simulate spatial PM variation within the considered tree crown. This explains the generally low
416 atmospheric PM₁₀ concentrations (<1.31 µg m⁻³) obtained in Figure 9. Considering an average seasonal
417 (from May to September) atmospheric PM₁₀ concentration of 25 µg m⁻³, measured at an urban
418 background monitoring station of the Flemish Environment Agency (VMM) at 1.5 km from the
419 considered tree crown, the local modelled PM₁₀ concentration at a height of 3.5 m (<1.31 µg m⁻³) would
420 account for only 5% of the average atmospheric PM₁₀. Previous studies already demonstrated that
421 local traffic emissions account for less than 20% of the urban PM₁₀ concentration in Antwerp (Deutsch
422 et al., 2006; TML, 2012; VMM, 2014, 2011). Taking a homogeneous PM₁₀ background of 25 µg m⁻³ into
423 account, the modelled PM₁₀ concentration reduction with increasing height decreases to 4.44%, which
424 is comparable to the 6% reduction with height obtained from the gravimetric results.

425 The lack of urban background concentration inside the model also explains the obtained smaller leaf-
426 deposited PM mass, when compared to the gravimetric results. Nevertheless, when we take the urban
427 background into account, the modelled leaf-deposited PM₁₀ weight amounts to an average 6.14 mg m⁻²,
428 which is still 15% of the averaged gravimetric results. Apart from the missing urban background
429 concentration, road dust resuspension is not simulated in the ENVI-met model, due to a lack of traffic-
430 induced turbulence. As PM₁₀ emissions due to road dust resuspension are approximately of the same
431 order of magnitude as tailpipe emissions (Amato et al., 2009; de la Paz et al., 2015; Gehrig et al., 2004),
432 depending on the considered environment, it might account for part of the observed discrepancy as
433 well. Additional uncertainty in the model outcome might be due to various assumptions in our model
434 setup, e.g. a potential underestimation of the deposition velocity, lack of local measurements of the
435 input parameters (meteorology, traffic counts, ambient PM concentration, ...) or oversimplification of
436 the simulation period (May to September). Nevertheless, such assumptions are all frequently applied
437 in contemporary air quality models.

438 Our findings emphasize the need for continuous validation efforts of modelled results against
439 experimental data. Moreover, an integrated modelling approach should be envisaged, where pollutant
440 sources at different spatial scales are integrated inside the model domain. As for many countries, more
441 than 50% of the atmospheric PM concentration is due to cross-boundary pollution (EEA, 2015), and as
442 the size of the model domain is directly related to the computational time, integration of sources over
443 different monitoring scales is often achieved by defining boundary conditions inside the model
444 domain. This is, however, not possible in the current ENVI-met[®] v4 version.

445 4. Conclusion

446 Overall, the application of ground-based LiDAR has shown to be very useful in generating a fast,
447 detailed 3D model description of tree crown morphology in a non-destructive manner. The altered 3D
448 shape and canopy density, improved by the LiDAR methodology, have shown to influence the modelled
449 pollutant dispersion and resulting leaf-deposition of particulates by changing local wind flow
450 characteristics in urban environments. With regard to the effect of urban green on local air flows, it
451 might be more appropriate to use tree PAD instead of LAD. Moreover, our study confirmed the effect
452 of urban vegetation on the natural ventilation in urban environments. An accurate tree crown
453 representation seems, therefore, of great importance when aiming at modelling local PM distributions.

454 Gravimetric determination of particle loaded filter membranes resulting from washed leaf samples,
455 showed a leaf area normalised particle loading of 626.13, 28.72 and 33.69 mg m⁻², for respectively the
456 >10 µm, 3-10 µm and 0.2-3 µm particle size fractions. The averaged total leaf-deposited mass was
457 687.93 mg m⁻², of which the PM₁₀ fraction (0.2-10 µm) amounted 62.41 mg m⁻². An overall bad
458 agreement was obtained between the gravimetric and modelled leaf-deposited particulate mass.
459 Apart from the need for an accurate tree crown representation in atmospheric models, our findings,
460 therefore, emphasize the importance of continuous validation of modelled results against
461 experimental data in order to guarantee reliable air quality model outcomes.

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464

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