

1 A large eddy simulation of the dispersion of traffic emissions by moving vehicles at an intersection

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10

11 Abstract

12 Traffic induced flow within urban areas can have a significant effect on pollution dispersion, particularly
13 for traffic emissions. Traffic movement results in increased turbulence within the street and the
14 dispersion of pollutants by vehicles as they move through the street. In order to accurately model urban
15 air quality and perform meaningful exposure analysis at the microscale, these effects cannot be ignored.
16 In this paper we introduce a method to simulate traffic induced dispersion at high resolution. The
17 computational fluid dynamics software, Fluidity, is used to model the moving vehicles through a domain
18 consisting of an idealized intersection. A multi-fluid method is used where vehicles are represented as a
19 second fluid which displaces the air as it moves through the domain. The vehicle model is coupled with
20 an instantaneous emissions model which calculates the emission rate of each vehicle at each time step.
21 A comparison is made with a second Fluidity model which simulates the traffic emissions as a line source
22 and does not include moving vehicles. The method is used to demonstrate how the effect of moving
23 vehicles can have a significant effect on street level concentration fields and how large vehicles such as
24 buses can also cause acute high concentration events at the roadside which can contribute significantly
25 to overall exposure.

26 **Keywords:** air pollution; dispersion; traffic; emissions; exposure; CFD.

27 1. Introduction

28 Commuters and residents in urban areas are often exposed to high concentrations of pollution while
29 they travel due to their proximity to traffic emissions and the tendency to travel at peak hours. These
30 periods of high exposure correspond to periods of higher inhalation rates for pedestrians and cyclists (de
31 Nazelle et al., 2012). This results in a disproportionately high intake of pollutants during daily commutes
32 which require consideration in order to fully understand the impact of poor urban air quality on health.
33 For example, a study by de Nazelle et al. (2013) involving 36 subjects in Barcelona found that travel
34 activities contributed to 24% of total intake of NO₂, despite accounting for only 6% of time. Many
35 studies using portable sensors have investigated the exposure of pedestrians and cyclists to particulate
36 matter from vehicles (Berghmans et al. 2009, Int Panis et al. 2010, Kingham et al. 2013, Ragetti et al.
37 2013, Hankey and Marhsall 2015, Yang et al. 2015, Ham et al. 2017, Rivas et al. 2017). The
38 concentrations experienced by pedestrians and cyclists is highly variable during any given journey, with
39 the standard deviation of measurements often of the same order as the mean and a few acute
40 concentration events contributing significantly to overall exposure. This is particularly true for areas with
41 lower population density and therefore lower background PM concentration levels, for example as seen
42 by Kingham et al. (2013). Even greater heterogeneity would be expected for the concentrations of NO₂
43 as it is primarily emitted by vehicles and therefore concentrations near roads tend to be significantly

44 higher than the background. Higher inhalation rates for cyclists make these peak concentration events
45 more significant. For example, Int Panis et al. (2010) estimate a correction of 4.3 to account for the
46 increased inhalation rate of cyclists relative to pedestrians based on field measurements for commuter
47 cyclists. In addition, deeper inhalation during cycling leads to higher deposition of ultrafine particles in
48 the lungs (Daigle et al., 2003). The highly variable nature of pollution concentrations within the urban
49 environment is due to many factors including weather conditions, traffic flow rates, proximity to passing
50 vehicles, vehicle type, isolated high pollution sources including highly polluting vehicles. The use of on-
51 board measurement systems (Irwin et al. 2018) and portable emission measurement systems (PEMS)
52 (O'Driscoll et al. 2016) have shown that vehicle emissions are dominated by high emission peaks during
53 vehicle acceleration or gear changes, further contributing to the heterogeneity of concentrations within
54 urban streets. In order to accurately evaluate the health impact of pollutants inhaled within urban areas
55 beyond measurement field studies, the occurrence of these acute concentration events must be
56 modelled.

57
58 It is known that traffic induced flow contributes significantly to pollution dispersion in urban areas,
59 particularly at low wind speeds (Qin 1993). This was demonstrated in an experiment undertaken as part
60 of the DAPPLE project (Arnold et al. 2004) in which an inert tracer was released in a street in London,
61 concentrations of which were detected at upwind locations. It was hypothesized that these upwind
62 concentrations were due to the entrainment of the tracer by vehicles moving upwind. While there are
63 many modelling studies focused on understanding the dispersion of tailpipe emissions within the near-
64 wake region of moving vehicles (e.g. Baker 2001, Dong and Chan 2006, Carpentieri et al. 2010, 2012,
65 Tientcheu-Nsiewe et al. 2016), efforts to numerically simulate traffic induced dispersion within urban
66 scenarios have so far been limited in number. Solazzo et al. (2007) simulated the effect of moving
67 vehicles on air flow using moving canyon walls relative to stationary blocks representing vehicles. An
68 averaged parameterization of the impact of traffic induced turbulence on dispersion was derived by Di
69 Sabatino et al. (2003) and Kastner-Klein et al. (2003). This parametrization was used within a Reynolds
70 averaged Computational Fluid Mechanics (CFD) simulation by Thaker and Gokhale (2016), however this
71 approach does not resolve the temporal variation in concentrations and magnitude of high
72 concentration peaks seen in field study measurements. Other Reynolds averaged simulations using a
73 moving mesh were undertaken by Kim et al. (2016) and Dong et al. (2017). A Large Eddy Simulation (LES)
74 approach was used by Zhang et al. (2017), where each vehicle's drag force was applied to the air using a
75 momentum source term. Relatively simple setups in a wind tunnel with a limited number of vehicles
76 (Pearce & Baker 1997, Ahmad et al. 2002, Kastner-Klein et al. 2001, Kastner-Klein et al. 2003) have
77 proven to be informative, however a method for simulating complex traffic flows has yet to appear. In
78 this paper, we investigate the combined effect of moving vehicles and time-dependent emissions at the
79 rear of the vehicles on emission dispersion.

80 In order to realistically model traffic induced dispersion, it is important to use a realistic emissions model
81 for the vehicles. It is well known that the emissions of petrol and diesel vehicles within an urban
82 environment are highly variable, with high emission peaks occurring during high engine loads (O'Driscoll
83 et al. 2016, Irwin et al. 2018), for example when a vehicle accelerates from standstill. The emissions
84 model COPERT (Computer Program to Calculate Emissions from Road Transport), which is widely used
85 across Europe, estimates the average emission factors as functions of a vehicle speed only, with the
86 average speed along a road often used. While the model attempts to account for the higher emission
87 rates expected at lower average speeds due to increased stopping and starting, the spatial and temporal
88 variability is lost. Furthermore, COPERT has limitations at low vehicle speeds (<10kmph) (O'Driscoll et al.

89 2016), which causes further problems for urban modelling where average vehicle speeds can often be
90 low, particularly during peak hours and near junctions.

91 Here, a Large Eddy Simulation (LES), computational fluid dynamics method is used to simulate the air
92 flow and dispersion of traffic emissions due to traffic movement within an urban scenario. The open
93 source CFD code Fluidity (<http://fluidityproject.github.io/>) is used to model the air flow at a crossroads
94 consisting of four lanes of traffic for a period of low wind speed. A low wind speed case is chosen such
95 that traffic-induced turbulence dominates as wind driven turbulence is very low within the street
96 canyons. Fluidity's traffic-induced dispersion model is coupled with an instantaneous emissions model,
97 where each vehicle's NO_x emission is calculated as a function of the vehicle's velocity and acceleration.
98 By considering the acceleration in addition to the velocity, a more realistic emissions model is obtained
99 which attempts to account for the emission peaks at high engine loads. In order to provide the Fluidity
100 traffic model and the emissions model with the required vehicle dynamics, the traffic simulation
101 software PTV Vissim (PTV Group) is used to simulate the traffic along the crossroads geometry. The
102 method is capable of simulating the effect of moving traffic on pollution dispersion within the street at
103 high temporal and spatial resolution. We demonstrate the potential of the method by simulating the
104 dispersion at the crossroads, formed by the intersection of two street canyons. Two cases are
105 compared, one with the coupled traffic-emissions model, and one without traffic movement where the
106 emissions are modelled as a constant line source. The impact of moving vehicles on the dispersion of
107 emissions is investigated in addition to the effect of vehicles on the occurrence of acute high
108 concentration events at the roadside. This work was carried out as part of the MAGIC project
109 (<http://www.magic-air.uk/>).

110 Section 2 describes the methodology implemented, including the Fluidity traffic model, the traffic
111 simulation (PTV Vissim) and emissions model. Section 3 describes the setup of the simulations, whilst
112 the comparison of the two crossroads simulations is given in Section 4, along with an analysis of the
113 effect of traffic on dispersion. A discussion of the methodology is given in Section 5 and conclusions in
114 Section 6.

115 **2. Methodology**

116 The method used in this paper comprises of three parts. The first involves the computational fluid
117 dynamics (CFD) software, Fluidity, used to simulate the airflow within the domain. Section 2.1 provides
118 the details of the CFD methods used here. The second part described in Section 2.2 is the traffic
119 simulation using PTV Vissim (PTV Group), which provides Fluidity with the required vehicle dynamics.
120 The final part is the emissions model, described in Section 2.3, which is used by Fluidity to calculate the
121 emissions of each vehicle at any given time.

122 **2.1. Urban airflow model**

123 Fluidity is an open-source software, developed at Imperial College London. Fluidity is a general purpose,
124 finite-element CFD software, within which a Large Eddy Simulation (LES) methodology is implemented
125 with an anisotropic adaptive mesh. Using the LES method, Fluidity resolves the turbulent features of the
126 flow larger than a specified filter length by solving the filtered Navier-Stokes equations, while smaller
127 scale eddies are modelled as additional viscosity based on a Smagorinsky-type model. Its adaptive mesh
128 capability allows Fluidity to automatically adapt the mesh to regions of high gradients, such as evolving
129 eddy patterns, while using a coarser mesh at more stable regions. The level of refinement is controlled

130 by a desired interpolation error for each field, entered by the user, in addition to the chosen minimum
 131 and maximum edge lengths. The adaptive mesh technique is described in detail in (Pain et al., 2001). A
 132 variation of the Smagorinsky model, developed by (Bentham, 2003), is used to model the subgrid scale
 133 eddy viscosity. The second order scheme used here allows for anisotropic eddy viscosity where the filter
 134 length depends on the local element size, which is particularly suited for an unstructured, adaptive
 135 mesh.

136 Fluidity has an inbuilt traffic module capable of simulating the effect of individual vehicles on the airflow
 137 as they move through the domain. The traffic model treats the vehicles as a second fluid in a multi-fluid
 138 problem. Each vehicle is modelled as a second highly viscous fluid (essentially a solid), which displace the
 139 air around them as they flow through the domain. The problem is solved as a multi-fluid problem,
 140 however the vehicle dynamics are provided as an input and therefore the momentum equation for the
 141 vehicle “fluid” does not need to be solved. The continuity equation for this multi-fluid, incompressible
 142 flow is given by:

$$\frac{\partial}{\partial t}(\alpha_k \rho_k) + \nabla(\alpha_k \rho_k U_k) = 0, \quad \text{for } k = s, f.$$

143 Here s and f denote the fluid phase and stand for solid and fluid, respectively. α_k is the volume fraction
 144 of fluid k , where we have $\alpha_s + \alpha_k = 1$. ρ_k and U_k are the density and velocity of fluid k . For an
 145 incompressible problem the momentum equation for the air is given by:

$$\alpha_f \rho_f \left(\frac{\partial U_f}{\partial t} + (U_f - U_G) \cdot \nabla U_f \right) = -\alpha_f \nabla p + \nabla \cdot (\mu \nabla \cdot U_f + \tau_f) + \sigma_d (U_s - U_f),$$

146 where U_G is the finite element grid velocity due to the adaptive mesh, $U_s = f(\bar{x}, t)$ is the known velocity
 147 of the solid phase, p is the fluid pressure and τ_f is the unresolved turbulent stress tensor. σ_d is an
 148 absorption coefficient defined as $\sigma_d = \alpha_s \rho_f \beta / \tau_\sigma$, where $\beta \geq 1$ is a weighting factor (β is set to 1 in
 149 these simulations) and $\tau_\sigma = 1/\Delta t$ is used to relax the fluid velocity to the solid velocity in timestep Δt .
 150 The fluid equations were discretized using a continuous Galerkin discretization. The Crank-Nicholson
 151 scheme was used to discretize in time with an adaptive time step dependent on the user defined
 152 Courant number.

153 The last term on the right-hand side consists of both a momentum source and a sink which together
 154 apply the force of the vehicle fluid on the air. Where the vehicle velocity is greater than the fluid
 155 velocity, $U_s > U_f$, this term is positive and is therefore equivalent to a momentum source displacing the
 156 air around the solid. Where the vehicle velocity is less than the fluid velocity, $U_s < U_f$, this term is
 157 negative and therefore equivalent to an absorption term, decelerating the fluid velocity as it impacts the
 158 vehicle. It has been shown that with sufficient mesh refinement this method can be used to accurately
 159 model problems with complex solid geometries (Garcia et al, 2011).

160 The vehicle emissions are modelled as a passive tracer with the following advection-diffusion equation
 161 used to model its dispersion:

$$\frac{\partial c}{\partial t} + U_f \cdot \nabla c = \nabla \cdot (\kappa \nabla c) + F,$$

162 where c is the tracer concentration, κ is the diffusivity tensor and F is a source term. This equation is
 163 discretized using a second-order coupled finite element/control volume method.

164 2.2. Traffic micro-simulation

165 The Fluidity traffic model requires the vehicle dynamics as an input. Specifically, Fluidity requires the
 166 length, coordinates, velocity, acceleration and vehicle type at each timestep for each vehicle. For the
 167 crossroads simulation presented here this information was obtained by running a traffic simulation
 168 using the software PTV Vissim ([PTV Group](#)). PTV Vissim is primarily used for traffic management
 169 purposes and is able to simulate the movement of vehicles through a predefined geometry in order to
 170 assess traffic management decisions. The model includes a car following model, gap acceptance model
 171 and traffic regulations at intersections to provide a reasonable simulation of real-life driving behavior.
 172 The default PTV Vissim version 10 driving model was used here. The required data for each vehicle was
 173 output at 0.5 second intervals. Linear interpolation is used to obtain values at any time. The vehicle type
 174 determined the dimensions of the vehicle in addition to the emissions model to be used. The
 175 coordinates provided are only the x and y coordinates, as PTV Vissim is a two-dimensional model. A
 176 perfectly flat surface was assumed for the floor of the geometries used for the simulations in this paper.

177 2.3. Vehicle emissions model

178 The NO_x emissions of each vehicle were calculated at each time step of the CFD simulation. The
 179 emissions models for diesel cars and buses developed by Panis et al. (2006) were used. The emissions
 180 calculated for each vehicle at each simulation time step are modelled as a release of a passive tracer at
 181 the rear of the vehicle. The higher temperature of the exhaust and its velocity are not considered within
 182 the model. While these are expected to affect the near-wake dispersion, they are less significant within
 183 the far-wake. The models are in the form of bivariate quadratic equations which are functions of the
 184 vehicle velocity, v_n , and acceleration, a_n , where n denotes the vehicle number:

$$E_n(t) = \max[0, f_1 + f_2 v_n(t) + f_3 v_n(t)^2 + f_4 a_n(t) + f_5 a_n(t)^2 + f_6 v_n(t) a_n(t)].$$

185 Here f_1, \dots, f_6 are coefficients derived from the measurement data using a non-linear multiple
 186 regression, where a different set of coefficients are derived for the diesel cars and the diesel buses.
 187 These coefficients are given in Table 1.

188 *Table 1: Emission functions for diesel car and bus*

Vehicle Type	f_1	f_2	f_3	f_4	f_5	f_6
Diesel car ($a \geq -0.5m/s^2$)	2.41e-03	-4.11e-04	6.73e-05	-3.07e-03	2.14e-03	1.50e-03
Diesel car ($a < -0.5m/s^2$)	1.68e-03	-6.62e-05	9.00e-06	2.50e-04	2.91e-04	1.20e-04
Bus	2.36e-02	6.51e-03	-1.70e-04	2.17e-02	8.94e-03	7.57e-03

189
 190 These models were empirically derived from emission measurements for five diesel cars and six diesel
 191 buses taken during real urban traffic situations. These vehicles comply with different emission
 192 standards. The five diesel cars consist of two Euro 1, one Euro 2 and two Euro 3 type vehicles, while the
 193 six buses consist of two Euro 1 and four Euro 2 type vehicles, where Euro 1, 2 and 3 are vehicle emission
 194 standards defined by EU directives (91/441/EEC, 94/12/EC, 98/69/EC). These emission standards no
 195 longer represent the majority of the fleet in Europe, new vehicles are now required to comply with Euro
 196 6 standards. However, the instantaneous emissions model derived from these measurements still

197 provide emission peaks for periods of high engine loads as is also seen for newer engines (O’Driscoll et
198 al., 2016). It is therefore expected that this instantaneous model can still provide a reasonable estimate
199 of the expected variation in emissions with vehicle velocity and acceleration. While the magnitude of the
200 total NOx emitted will be an overestimate, the total emissions can be scaled to match those expected
201 from vehicles which comply with more recent standards. Within the context of this paper, we are not
202 concerned with absolute values. Rather, it is our intention to demonstrate both the importance of
203 accounting for these highly variable emissions and the potential of this method for investigating their
204 impact on local air quality. Further work is required to derive emission models that are representative of
205 the current fleet.

206 **3. Model setup**

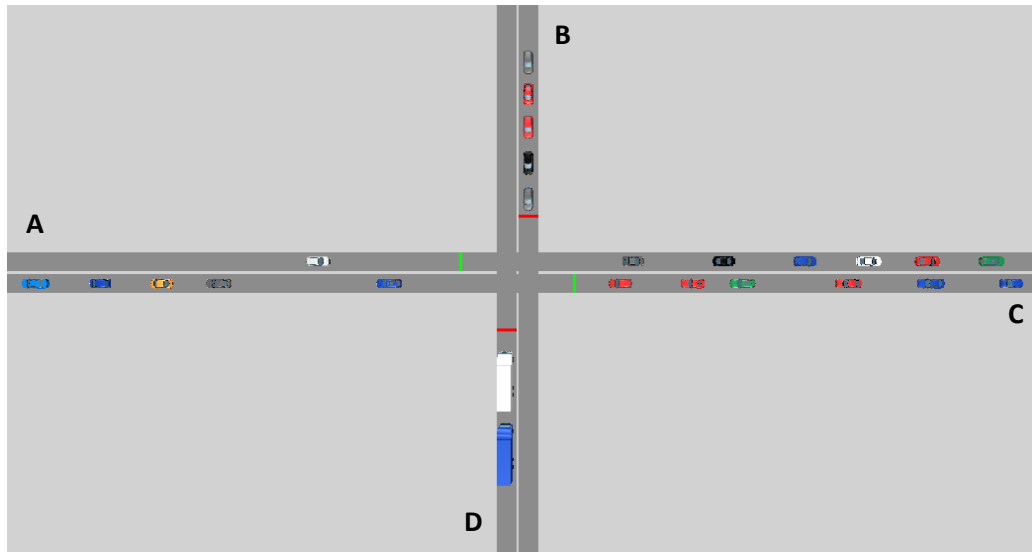
207 In order to capture the vehicle shapes as accurately as possible, Fluidity refines the mesh at the
208 interface between the vehicles and the air. These small mesh elements combined with the often high
209 velocities of the vehicles lead to smaller time steps than would otherwise be the case since the
210 maximum in-canyon velocities are likely to be lower in the absence of traffic. These small simulation
211 time steps combined with the relatively long simulation time required to model realistic traffic scenarios
212 currently limits the applicability of the method. Further work is required in order to investigate methods
213 for reducing simulation run times.

214 **3.1. Computational domain**

215 In order to reduce the run time, a coarse mesh was used relative to the size of the vehicles. A full size
216 geometry was used, where the vehicle size was 4.4 m x 1.5 m x 1.5 m for cars and 11.5 m x 2.55 m x 4.4
217 m for buses. A minimum element edge length of 0.5 m was used which sets the lower limit of the
218 element edge length used by the adaptive mesh. Each vehicle is modelled as a rectangular block. A
219 single vehicle simulation was used to assess the performance of the model at this resolution and the
220 suitability of the rectangular block geometry for the vehicles; the results of the simulation are provided
221 in Appendix A. It is shown that at this resolution the detailed flow dynamics around the vehicles is not
222 captured, particularly in the near-wake region. However, a better approximation of the far-wake is
223 achieved and a reasonable approximation of the overall effect of the vehicle on the flow is seen. An
224 adaptive time step was used for the simulation, the magnitude of which is limited by setting the Courant
225 number, which we set to equal 5. The performance of the model with this Courant number was
226 compared against the wind tunnel setup of Di Sabatino et al. (2003), as discussed in Appendix B. The
227 impact of the vehicles on the prevailing wind flow was in reasonable agreement for the traffic model
228 and the wind tunnel. Similarly, the turbulent velocities were also in reasonable agreement.

229 The crossroads traffic simulation used here is shown in Figure 1. The model was configured such that the
230 traffic flow rate from entry points A and C was 400 cars per hour. From entry point B the traffic flow rate
231 was set to 200 cars per hour, and for entry point D, the traffic flow was set to 200 buses per hour,
232 therefore simulating a bus lane. The average speed of the vehicles travelling through the domain was 14
233 km/h, reflecting the low speeds in busy urban areas. Each vehicle travelled directly across the
234 crossroads, with no vehicles turning. At the crossroads, a traffic light system was implemented. The
235 traffic lights, seen as green and red lines in Figure 1, allowed traffic to proceed along only one road at
236 any given time. The lights followed a signaling sequence as follows: 26 seconds Green, 3 seconds Amber,
237 30 seconds Red, 1 second Amber and Red. The first 15 minutes of the traffic simulation was discarded in

238 order to provide Fluidity with a fully developed traffic flow. The UK convention of vehicles driving on the left hand side of the road was adopted.
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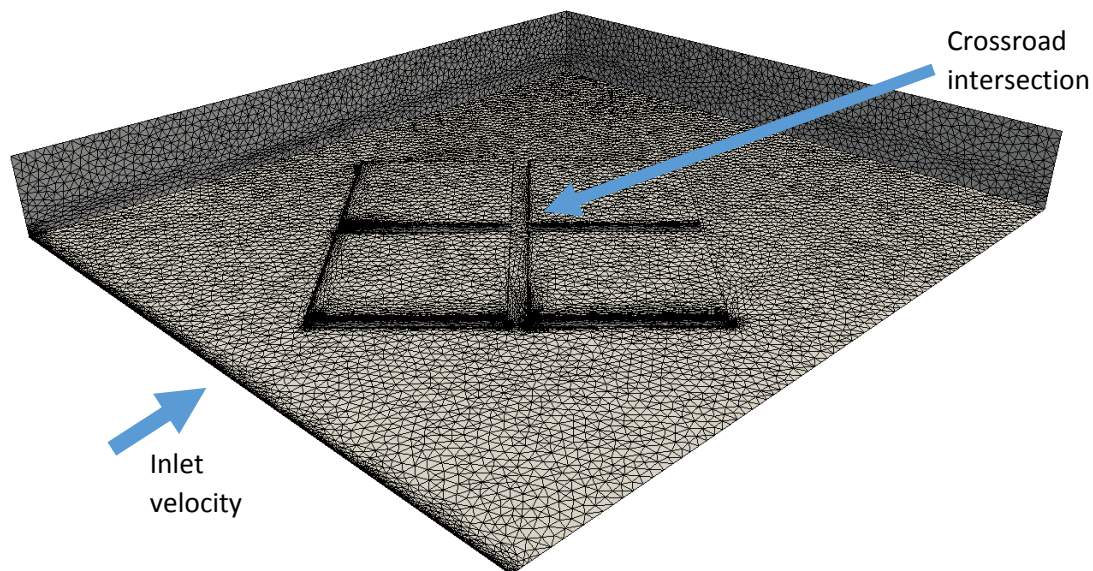
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241 *Figure 1: Snapshot of PTV Vissim traffic simulation of a crossroads consisting of three lanes of cars and one bus lane. Red and*
242 *green lines indicate location of traffic lights. The vehicle colours hold no significance.*

243 The crossroads geometry used in Fluidity is shown in Figure 2, where the crossroads are formed by the
244 intersection of two street canyons orientated at an angle of 45° to the average wind direction. The
245 domain size is 900 m x 800 m x 100 m. The intersecting canyons are formed by the inclusion of four 200
246 m x 200 m x 10 m buildings separated by 20 m. Each street canyon therefore has a width of 20 m and
247 height of 10 m and both roads have a length of 420 m. The distance from the canyon centerline to each
248 vehicle centerline is 2 m, simulating a road width of approximately 9 m. Each pavement is therefore 5.5
249 m wide with the exact distance from the canyon walls to the vehicle dependent on whether a car or a
250 bus is considered.

251 3.2. Boundary conditions

252 The Synthetic Eddy Method (Jarrin et al. 2006, Pavlidis et al. 2010) is used to apply a turbulent velocity
253 profile at the inlet. A profile representative of a neutral atmospheric boundary layer is applied with a
254 reference velocity of 1.5 m/s at roof height (10 m). A low wind speed was chosen in order to simulate a
255 scenario where traffic movement is likely to be at its most significant and dominates the turbulence
256 production within the canyon. No slip conditions were applied at the floor and building surfaces, “no
257 shear” conditions were applied to the two side walls and top surface and a pressure boundary condition
258 was applied to the outlet. As previously mentioned, a minimum element edge length of 0.5 m was used,
259 along with a maximum edge length of 10 m. The maximum number of nodes allowed for the mesh was
260 set to 1 million, however this number was never reached with the mesh size never exceeding 600,000
261 nodes. The average element edge length within the canyon was approximately 1 m.



262
263 *Figure 2: Initial mesh for crossroads geometry*

264 At each time step of the simulation, Fluidity uses the instantaneous emission models described in
265 Section 2.3 to calculate the magnitude of the instantaneous release of emissions from each car and bus.
266 Linear interpolation is used to calculate the vehicle velocity and acceleration at any given time from the
267 0.5 second resolution input from the traffic simulation. The passive tracer is released at either the rear
268 right-hand or left-hand side of the vehicle with a 50% chance of being either side. For buses, the
269 emissions were always on the left-hand side. These emission volume sources were cubes of 1 m height.

270 Large emission source volumes were required in order to ensure the presence of a mesh element within
271 the volume and therefore a continuous emission source. These were positioned such that the source lay
272 within the vehicle wake, with the centre of the cube 0.5 m from the vehicle's rear face, 0.5 m from the
273 side of the vehicle and 0.5 m from the ground. The emissions for each lane of traffic was considered as
274 one tracer field, so that the dispersion of emissions from each lane can be independently analysed,
275 allowing for the investigation of the dispersion of emission from one lane by the traffic from another. As
276 there are four lanes of traffic in this simulation, four separate tracer fields are considered. Each tracer
277 field has several moving sources, one at the rear of each of the vehicles in the corresponding lane at that
278 time.

279 The simulation was allowed to run for 1000 seconds (~16 minutes) before the introduction of traffic in
280 order to develop the street canyon flow. Once traffic was introduced, the first 10 minutes was
281 considered an initialization period, and the remaining 8 minutes involved 8 sets of one-minute traffic
282 light sequences. Thus, the overall simulation time without and with traffic was approximately 34
283 minutes. The simulation time with traffic was approximately 40 seconds/day running on 16 cores.

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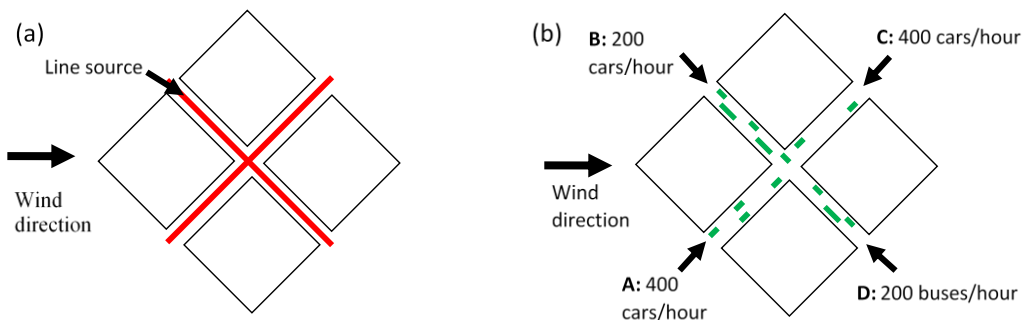
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289 3.3. Test Cases

290 The crossroad simulation with moving vehicles was compared to a crossroad simulation without, using
291 line sources to model traffic emissions. Two line sources were used, which are in reality volume sources
292 extending the length of each of the two roads, with a width of 8m and height of 2m, approximately the
293 width of two traffic lanes and the height of the cars respectively. The emission rate at the intersection
294 of the two volume sources was therefore twice of that elsewhere as the two volume sources overlapped. A
295 schematic of the two simulations is shown in Figure 3.



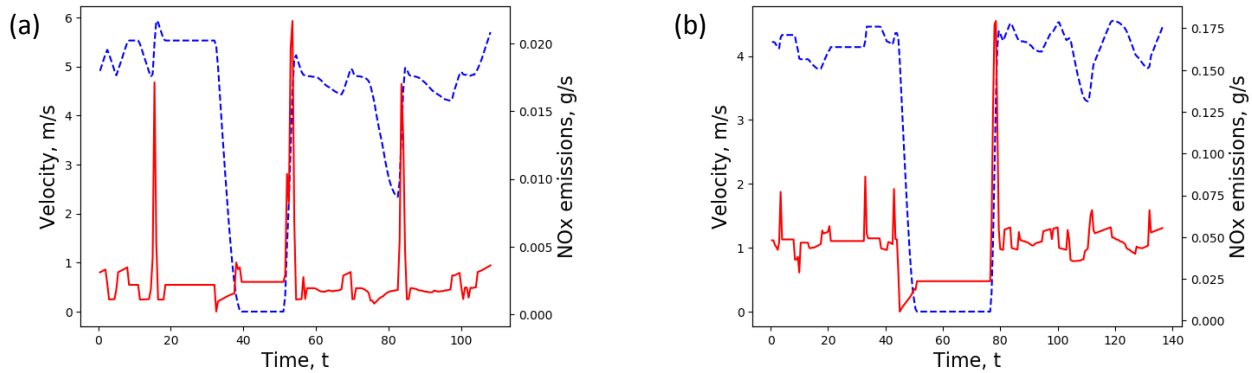
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297 *Figure 3: Schematic of (a) the line source simulation and (b) the traffic model simulation.*

298

299 **4. Results and Discussion**
300 **4.1. Emissions**

301 Figure 4 (a) and (b) show the instantaneous emissions and velocities for a car and a bus, respectively, as
302 they move through the crossroads domain. The zero velocity period for both cases is due to the period
303 the vehicle spends waiting for the lights to turn green at the crossroads. The large emission peaks during
304 periods of high acceleration are clearly visible and show similar behavior to that presented by O’Driscoll
305 et al. (2016) for measurements taken for a Euro 6 diesel car.



306
307 *Figure 4: Vehicle velocity (blue, dashed) and NOx emissions (red) against time for (a) a car and (b) a bus moving through the*
308 *crossroads geometry.*

309 Figure 5 shows the cumulative emissions along each road per meter length of road for the traffic
310 movement simulation for the 8 minute duration analyzed here. It is noted that the emissions model
311 used is based on an old vehicle fleet, however interesting insights can be obtained from looking at the
312 variations in emissions. The average emission factor for the cars travelling through the domain was 0.55
313 g/km, whereas the average emissions factor for a Euro 6 diesel car driving urban routes is likely to be
314 closer to 0.4 g/km (O’Driscoll et al. 2016). The average emission factors for the buses travelling through
315 the domain were an order of magnitude higher than for the cars. These emission factors, while perhaps
316 higher than the average for a modern fleet, are not beyond reasonable expectation. Further, as the
317 nonlinearity of chemistry is not considered, the concentrations can be scaled to reflect a desired average
318 emission factor.

319 The emission peaks that can be clearly seen for each lane are at the locations where the vehicles queue
320 at the traffic lights. This region of high emissions is partly due to the emission peaks at acceleration, such
321 as those seen in Figure 4; however they are also due to the accumulation of emissions while the engine
322 is idling and the vehicles are stationary. For this simulation, high emission rates as a result of
323 accelerations contribute between 20-30% to these peaks. The remaining 70-80% is due to emissions that
324 occur during idling. Despite the idling emission rates being significantly lower than the peak emission
325 rates whilst accelerating, the length of time spent idling leads to higher contributions to the total
326 amount emitted. This ratio is likely to be different for more modern vehicles where idling emissions in
327 particular would be expected to be lower.

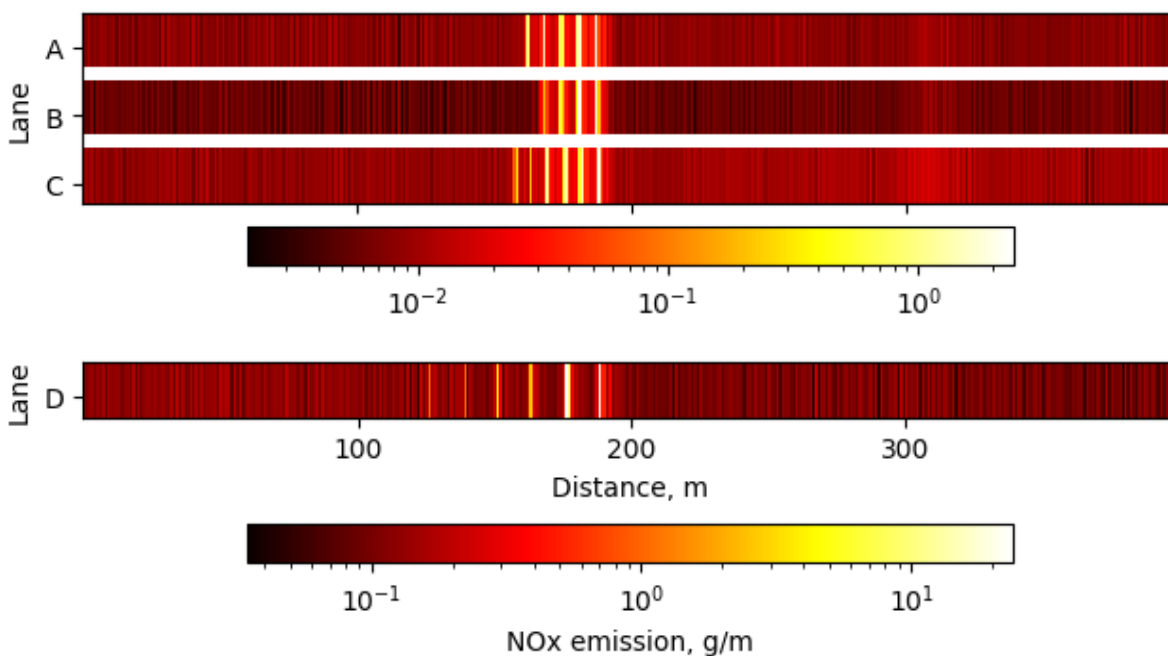
328 Comparing the plots for lanes A, B and C, despite the different vehicle counts between each lane, the
329 maximum emission peaks are all similar in magnitude (roughly 1 g/m). This is attributed to the vehicle
330 flow rate being sufficiently high along each road so that at least one vehicle is likely to be waiting at the

331 traffic lights each time they turn red. The lower traffic flow for case B is reflected in the higher rate in
 332 decrease of each peak with distance from the crossroads in addition to fewer peaks in total. The
 333 emissions away from the crossroads are also lower for case B than for cases A and C where the traffic
 334 flows are higher. Case D has significantly higher emissions than the others due to the different emissions
 335 model used, representative of an older bus fleet. Case D can also be seen to be a bus lane from the large
 336 distance between the emission peaks at the crossroads in comparison to those of the other cases with
 337 smaller vehicles. The vehicle count and total NOx emitted during the 8 minute period is shown in Table
 338 2. Lane D is the bus lane and has much higher emissions due to the different coefficients used for the
 339 buses as seen in Table 1.

340 *Table 2: Vehicle count and total NOx emitted along each lane of traffic over the 8 minute period.*

Lane	A	B	C	D
Vehicles	62	38	51	32
NOx emitted (g)	12.8	8.9	12.0	114.1

341
 342 In each case, away from the crossroads the accumulated emissions are two orders of magnitude lower
 343 than the highest peaks at the crossroad but remain highly variable.
 344



345
 346
 347 *Figure 5: Cumulative emissions per meter of road length for each lane of traffic.*

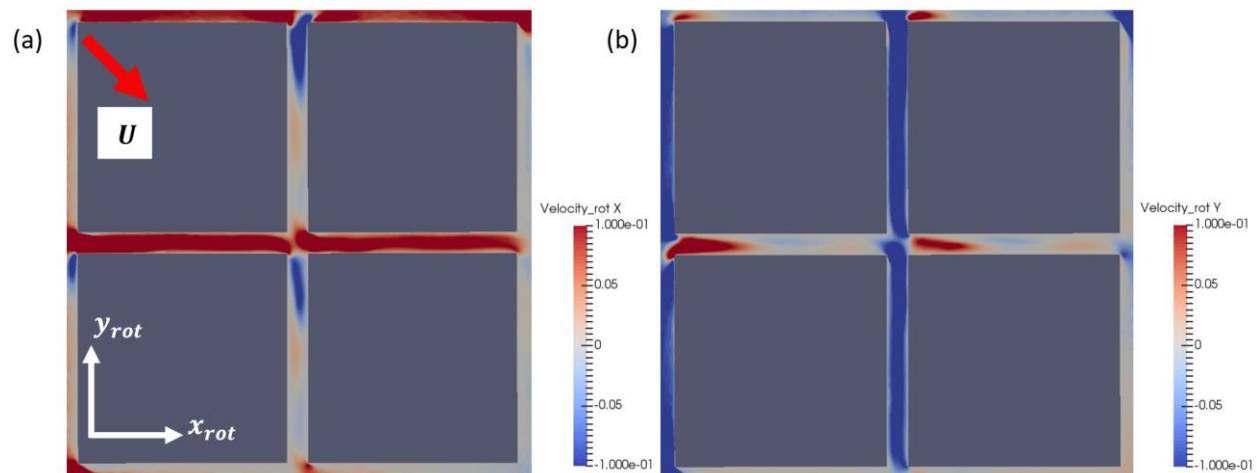
348 **4.2. Velocity fields**

349 Figure 6 shows average velocity fields on a z-plane at a height of 1 m for the line source case, where the
 350 velocities are shown for a rotated coordinate system to align with the two street canyons. Figure 6 (a)
 351 shows the velocity in the x_{rot} direction, as indicated by the coordinates in the diagram, and Figure 6 (b)

352 shows the velocity in the y_{rot} direction. It is clear that there exists a prevailing flow along both canyons,
353 dictated by the wind direction, driven mainly by the inflow of air at the entrance of the street canyon.
354 Higher velocities are also present immediately downwind of the intersection due to the downwash of air
355 into the intersection which again drives the flow along the canyon.

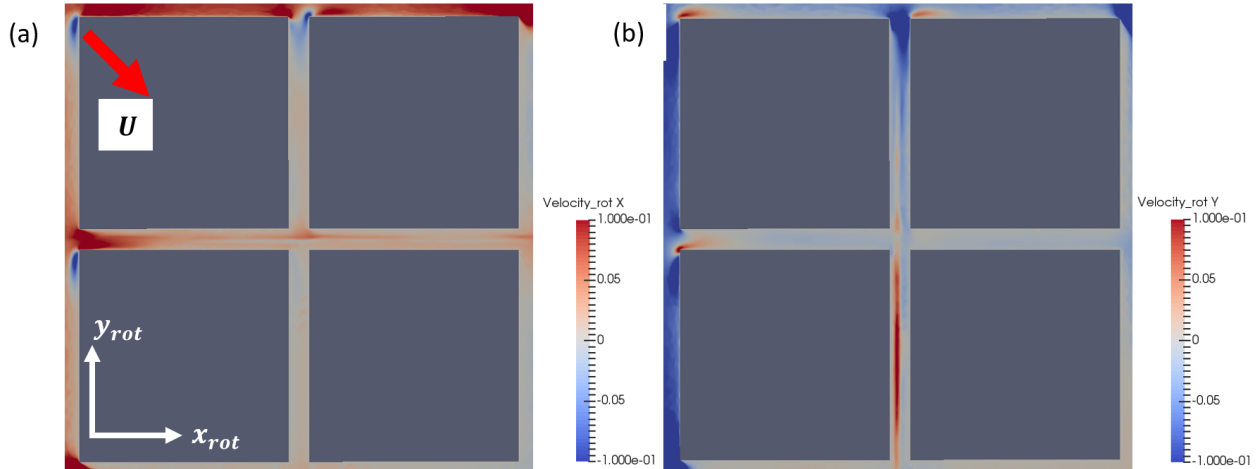
356 Figure 7 shows the equivalent average velocity fields for the case with traffic-induced dispersion. These
357 velocity fields are averaged over the 8-minute period. There is a clear difference between the case with
358 traffic-induced dispersion and that without. In contrast to the line source case, there doesn't seem to be
359 an obvious direction of prevailing flow along either canyon when traffic movement is included other
360 than near the wind facing ends of the two canyons where the inflow of air is still significant. This
361 suggests that, at least at this height of 1 m from the ground, the traffic movement has a significant
362 impact on the flow for the low wind speed case considered. This reflects the wind tunnel results of Di
363 Sabatino et al. (2003) where moving plates, representing vehicles, were found to induce a prevailing
364 flow. This wind tunnel setup was simulated using the Fluidity traffic model and good agreement was
365 found between the two as discussed further in Appendix B. In the case of the bus lane, a region of
366 upwind average velocity can be seen in Figure 7 (b). The buses have a larger impact on the air flow
367 within the canyon due to their larger size.

368



369

370 *Figure 6: Average velocity field for line source simulation rotated to align with street canyons at a height of $z=1m$.*



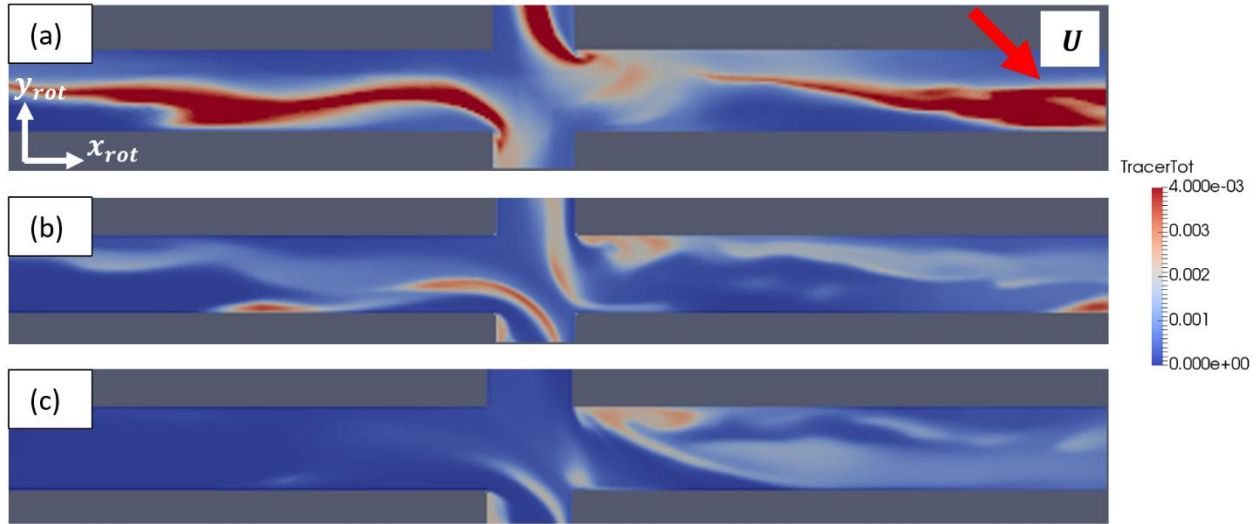
371
 372 Figure 7: Average velocity fields for traffic simulation rotated to align with street canyons at a height of $z=1\text{m}$.

373 **4.3. Tracer dispersion**

374 Figure 8 shows an instantaneous tracer concentration field for the line source case on a plane at heights
 375 $z=1\text{ m}$, 4 m and 8 m . Only one canyon is shown as the concentrations in the two canyons are very similar
 376 due to the symmetry of the geometry. The concentrations shown have been scaled using the total
 377 emissions of the line source and instantaneous emissions simulations as follows:

$$C_{line}^* = C_{line} \frac{E_{traffic}}{E_{line}},$$

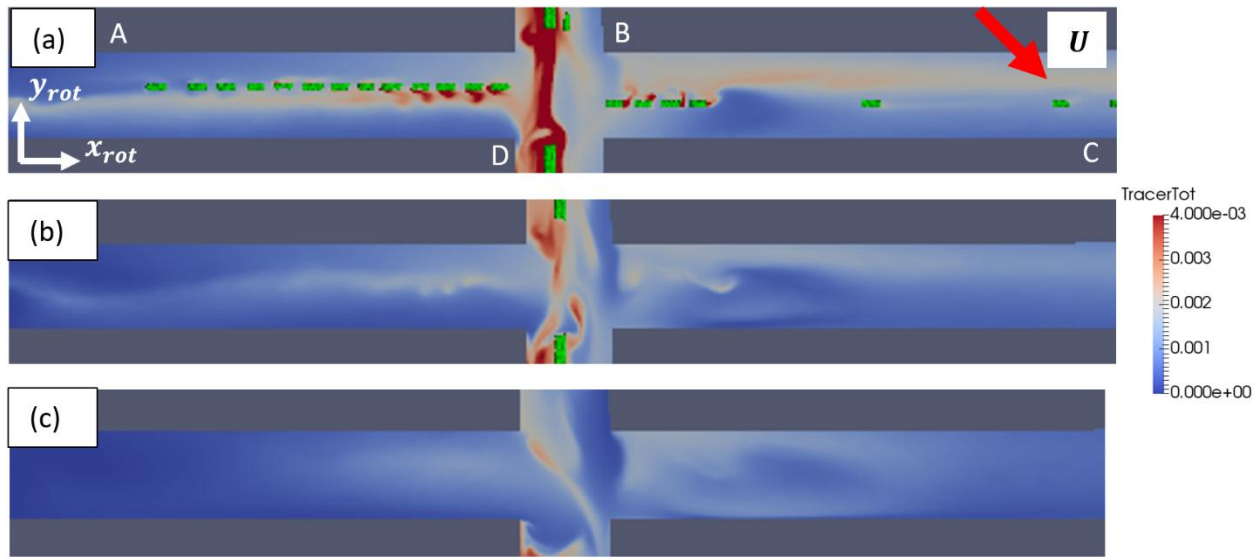
378 where C_{line}^* is the scaled concentration, C_{line} is the original concentration due to the line source
 379 emissions E_{line} , and $E_{traffic}$ is the total traffic emissions for the simulation period. C_{line}^* therefore
 380 represents the concentrations due to emissions from a line source equating to the emissions from the
 381 traffic-induced dispersion simulation. The effect of the inflow of air at the wind-facing ends of the
 382 canyons is clear as the tracer is dispersed down the canyon towards the intersection. This is particularly
 383 true higher up the canyon, where concentrations at $z=8\text{ m}$ are very low upwind of the intersection.
 384 Vertical mixing generated by the inflow of air at the intersection leads to higher concentrations higher
 385 up the canyon downwind of the intersection. While a turbulent flow is applied at the inlet, the street
 386 canyons are shielded from this turbulence to a degree due to a boundary layer that forms along the flat
 387 roofs of the four buildings. This leads to the tracer dispersion at $z=1\text{ m}$ being dominated by larger scale
 388 turbulent motions generated by the canyon geometry rather than small scale turbulence.



389
 390 *Figure 8: Instantaneous scaled tracer concentrations (g/m^3) for the line source simulation at (a) $z=1\text{m}$, (b) $z=4\text{m}$ and (c) $z=8\text{m}$.*

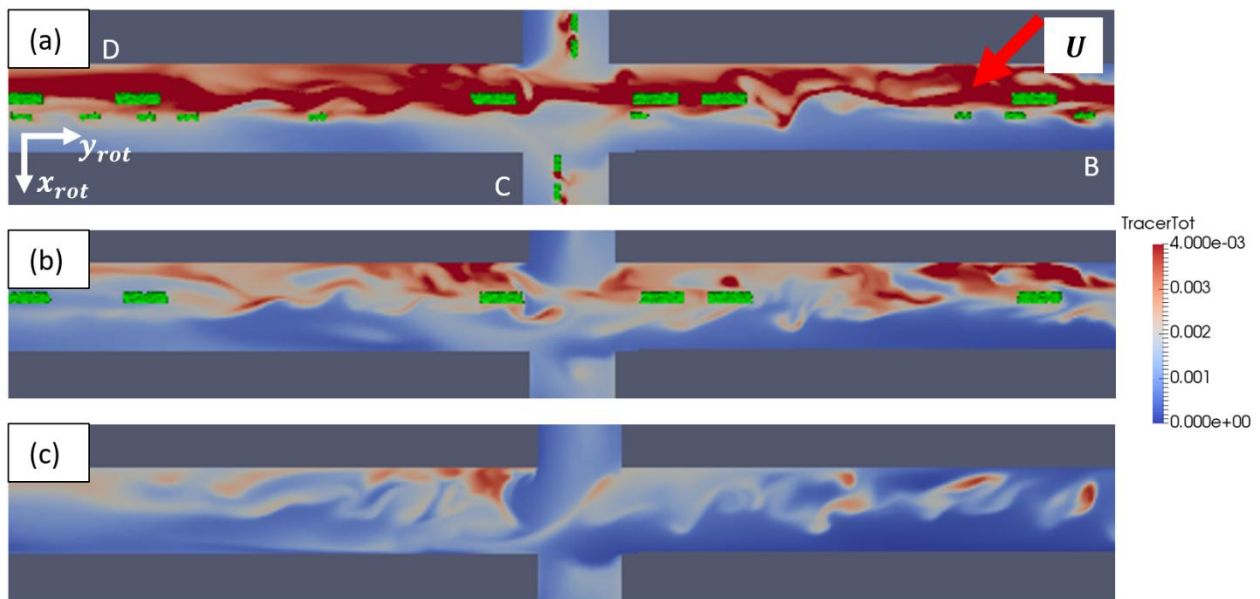
391 Figure 9 and Figure 10 shows an instantaneous concentration field at three different heights for the
 392 traffic emissions along canyons A and C and canyons B and D, respectively. Here Figure 10 has been
 393 rotated horizontally. The effect of the vehicles on the tracer dispersion is immediately evident. For this
 394 particular point in time, the vehicles in lanes B and D are moving across the crossroads while the
 395 vehicles in lanes A and C are waiting at the red lights. The higher concentrations behind the buses are
 396 due to the higher emission rates for these vehicles. The inclusion of moving vehicles leads to greater
 397 mixing of the emissions across the canyons as compared to the line source concentration fields seen in
 398 Figure 8. There are also significantly higher concentrations at height $z=8\text{ m}$ upwind of the intersection
 399 when vehicle movement is included. This is particularly true for the bus lane. The concentration field
 400 along the bus lane suggests a highly turbulent flow with significant vertical dispersion.

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408

409 *Figure 9: Instantaneous tracer concentration (g/m^3) along lanes A and C due to emissions from all traffic lanes at (a) $z=1\text{m}$, (b)*
 410 *$z=4\text{m}$ and (c) $z=8\text{m}$.*



411

412 *Figure 10: Instantaneous tracer concentration (g/m^3) along lanes B and D due to emissions from all traffic lanes at (a) $z=1\text{m}$,*
 413 *$z=4\text{m}$ and (c) $z=8\text{m}$.*

414 As the bus emissions are significantly greater than those of the cars, a clearer picture of the situation
 415 can be obtained by analysing the emissions from the cars and the buses separately. Figure 11 shows the
 416 emissions from the cars only (i.e. lanes A, B and C) at four different times. Arrows are used to show
 417 which lanes are moving across the intersection and a red cross is used to indicate a red light for that
 418 lane, with each lane corresponding to the symbol to its clockwise direction. An amber arrow indicates
 419 that the lights are about to turn red, while the green arrow cases show the situation shortly after the
 420 light turns green.

421 In Figure 11 (a), lanes B and D have been moving across the intersection for the green light period of 26
422 seconds, and the lights are about to turn red. Meanwhile, the vehicles in lanes A and C have been idling
423 for this 26 second period while waiting for the lights to turn green. The build up of emissions due to the
424 idling vehicles is clear to see. The emissions from the idling vehicles in lane A are not dispersed across
425 the intersection by the prevailing in-canyon wind direction due to the perpendicular flow generated by
426 the passing buses. Instead, concentrations build up across the street ahead of the queueing vehicles,
427 where you may expect to find waiting cyclists and pedestrians.

428 In Figure 11 (b) the vehicles in lanes B and D have now stopped and the vehicles in lanes A and C are
429 moving across the intersection. The concentration hotspot that formed ahead of the queueing vehicles in
430 lane A seen in Figure 11 (a) is now dispersing across the intersection. High concentrations are present
431 behind the cars as they accelerate across the intersection and entrain some of the emissions built up
432 while idling.

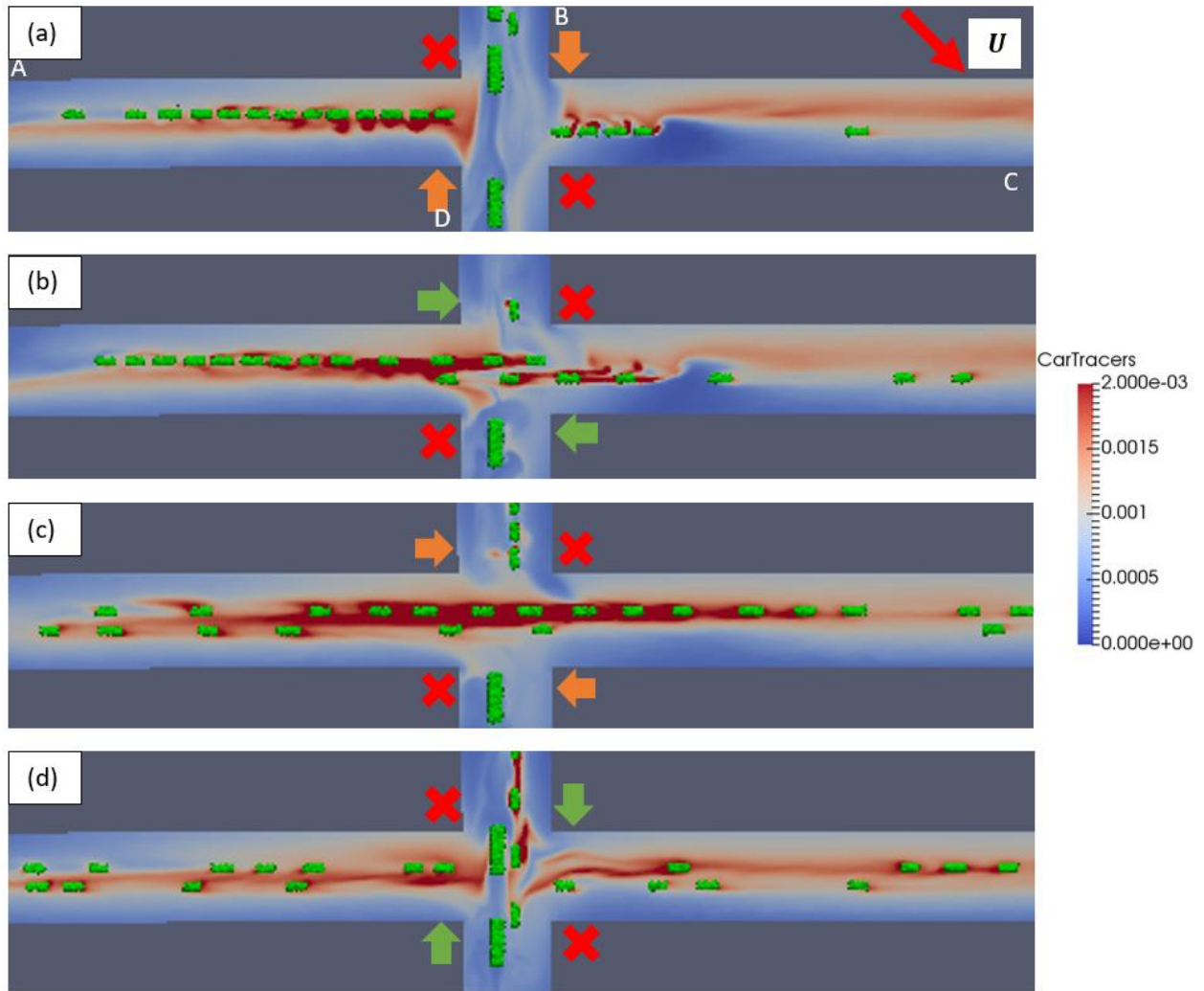
433 In Figure 11 (c) the cars in lane A and C have been moving across the intersection for the duration of the
434 green light period. High concentrations are clearly seen within the wake of the two lines of traffic, with
435 higher emissions in lane A due to a higher volume of traffic at this time. In Figure 11 (d), lanes A and C
436 have stopped and lanes B and D are now moving across the intersection, clearing away the emissions
437 due to lanes A and C as they do so. The shear layer blocking the dispersion of the idling emissions seen
438 in Figure 11 (a) forms again as the buses cross the intersection. High concentrations can be seen behind
439 the cars in lane B due to the period spent idling.

440 Figure 12 shows the the bus emissions for the same period as Figure 11. In Figure 12 (a) the buses have
441 been moving across the intersection for the duration of the green light period. The bus emissions are
442 contained within the wake of the buses to a greater extent than for the cars due to the larger size of the
443 vehicles leading to a stronger wake. It can be seen in Figure 12 (b) and (c) that as the cars move across
444 the intersection they clear away the bus emissions from the intersection. In Figure 12 (c) the emissions
445 due to the idling buses are dispersed in the direction of travel of the buses, which is upwind relative to
446 the in-canyon wind direction. This is despite the buses having been stationary for up to 26 seconds,
447 however as shown in Figure 7 (b), the bus lane induces an upwind flow within the canyon. Figure 12 (b)
448 and (c) also show that as the buses decelerate the high emissions within their wakes are dispersed
449 across the street leading to high concentrations on each side of the street where pedestrians are likely
450 to be walking. High concentrations are again seen in Figure 12 (d) as the buses accelerate once the light
451 turns green.

452

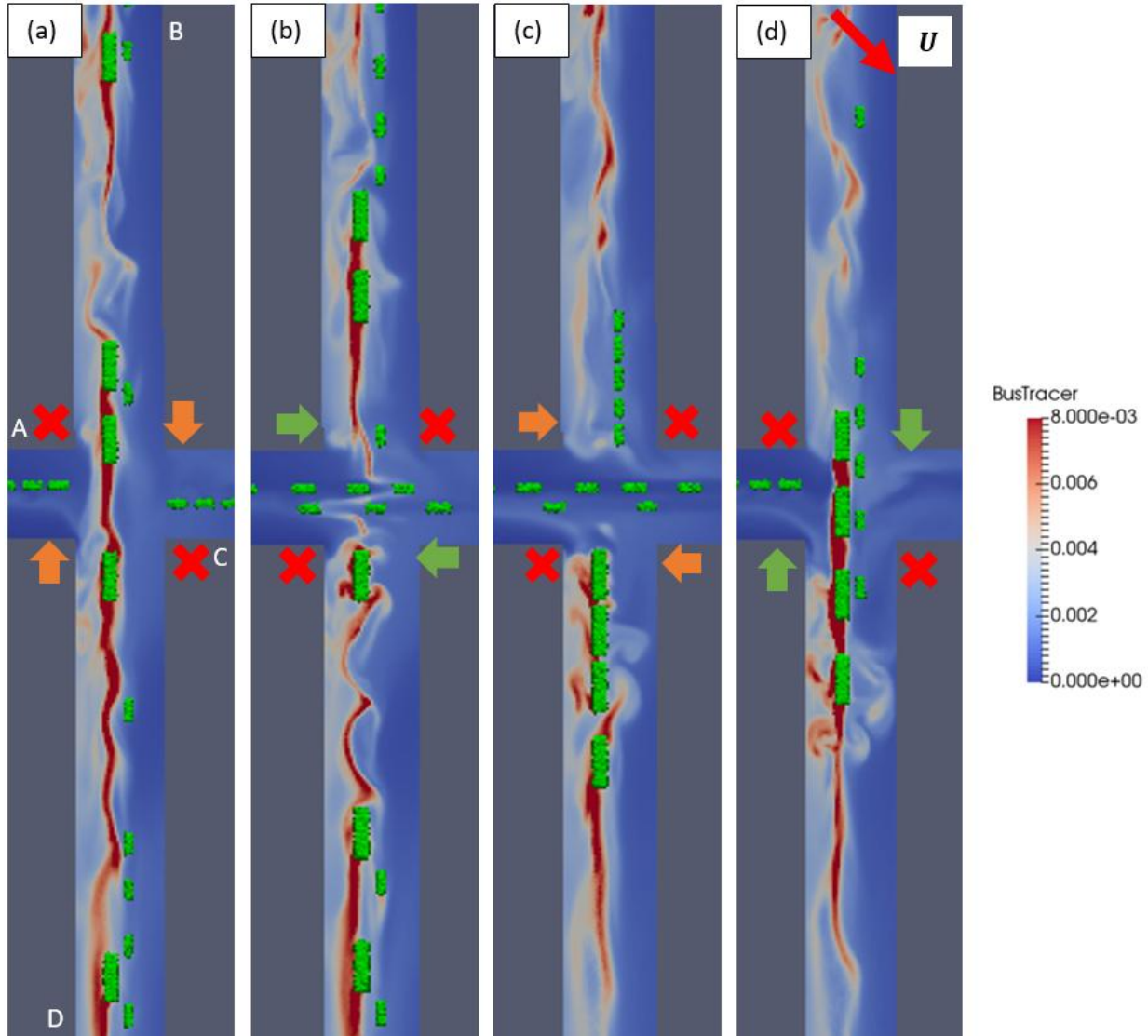
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455

456 *Figure 11: Tracer concentrations (g/m^3) due to emissions from lanes A, B and C at a height of $z=1\text{m}$ at different stages of the*
 457 *traffic lights signaling.*



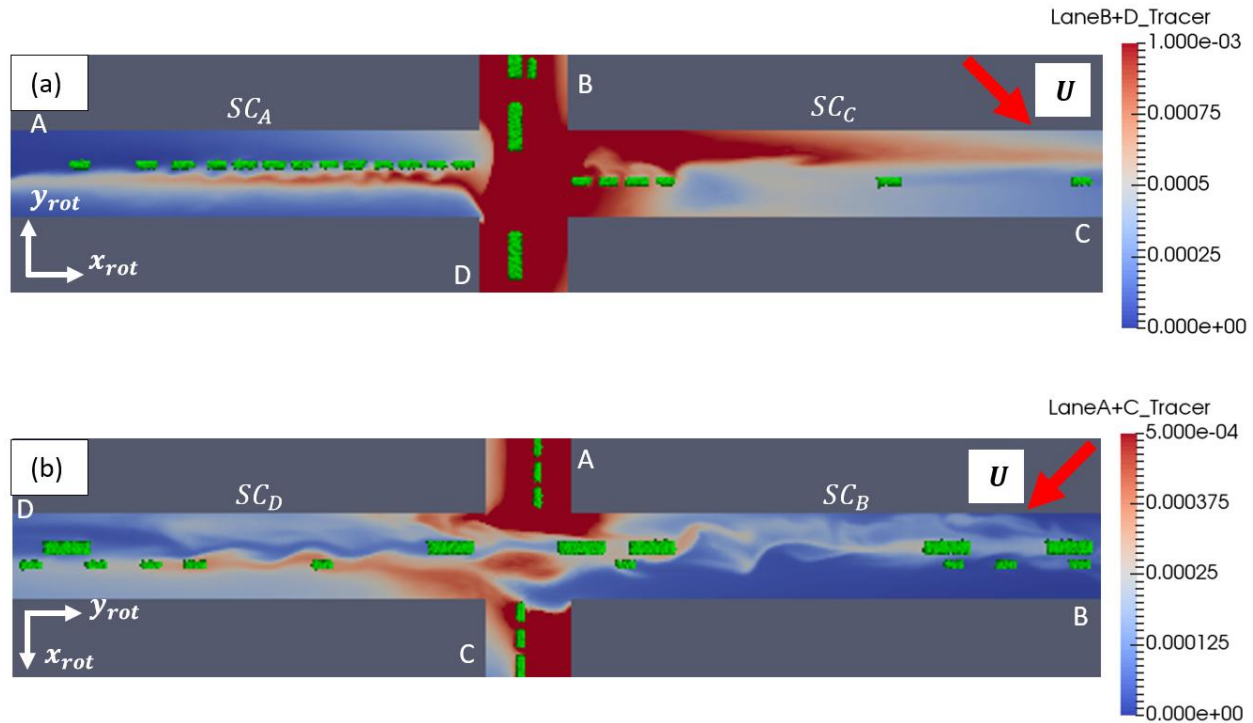
458

459 *Figure 12: Tracer concentrations (g/m^3) due to emissions from lane D at a height of $z=1m$ at different stages of the traffic lights*
 460 *signaling.*

461 Figure 13 (a) and (b) show the tracer concentration fields for the emissions from lanes B and D, and
 462 lanes A and C, respectively. A different scale is used to highlight the dispersion of emissions from one
 463 canyon to the other. In both cases a larger amount is dispersed in the downwind direction, however
 464 significant upwind dispersion of emissions from the intersecting road can also be seen.

465 The crossroads geometry can be considered as four street canyons linked by the intersection at the
 466 centre. These street canyons are 10 m high, 20 m across and 200 m long. Denoting each canyon by the
 467 lane of traffic which first enters the canyon, we have street canyons SC_A , SC_B , SC_C , and SC_D . Table 3
 468 shows the contribution of each lane of traffic to the total NOx in each of these street canyons. It can be
 469 seen that canyon SC_D contains the highest total NOx. This is to be expected as the buses travel along
 470 this canyon and it is downwind of the intersection. Similarly, canyon SC_C contains more NOx than
 471 canyon SC_A due to its downwind position relative to the intersection. The bus emissions contribute
 472 significantly to the total NOx in each canyon, including canyons SC_A and SC_C along which no buses

473 travel. This is true for canyon SC_A (19.5%) despite its upwind location relative to the intersection as the
 474 vehicles travelling along lane C entrain the bus emissions as they pass over the intersection. This
 475 demonstrates how concentrations along quieter roads could be significantly increased by the
 476 entrainment of pollutants from busier intersecting roads.



477
 478 *Figure 13: Tracer concentrations (g/m^3) in (a) canyons SC_A and SC_C due to emissions from lanes B and D and (b) canyons SC_B
 479 and SC_D due to emissions from lanes A and C. Emissions from each canyon are dispersed in both the downwind and upwind
 480 directions along the intersecting canyon.*

481 *Table 3: Contribution from each traffic lane to total NOx in each canyon.*

Canyon	Grams of NOx in canyon due to lane emissions				Total
	Lane A	Lane B	Lane C	Lane D	
SC_A	6.7 (35.8%)	0.4 (2.1%)	7.9 (42.6%)	3.6 (19.5%)	18.6
SC_B	1.0 (2.3%)	2.8 (6.2%)	0.9 (1.9%)	40.5 (89.7%)	45.2
SC_C	14.1 (31.7%)	2.4 (5.4%)	11.2 (25.2%)	16.7 (37.7%)	44.3
SC_D	3.3 (5.1%)	6.2 (9.5%)	3.9 (5.9%)	52.0 (79.6%)	65.3

482
 483 **4.4. Roadside concentrations**

484 Four locations were chosen to investigate the effect of traffic on the variation in concentration levels
 485 seen at the intersection. These locations are shown in Figure 14. The points were chosen as locations
 486 where pedestrians and cyclists may be expected to wait to cross the intersection. The points are
 487 approximately 2 m from the nearest passing lane of traffic.

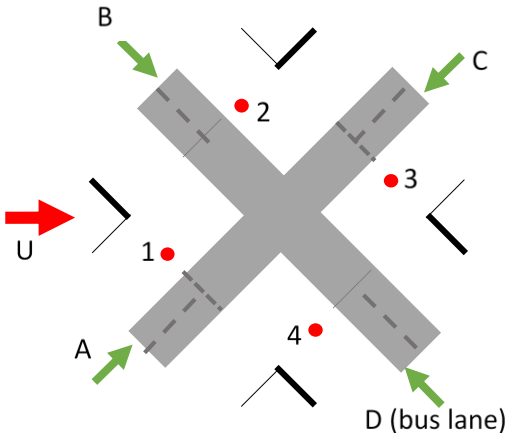
488 Figure 15 shows box plots of the concentrations seen at points 1 to 4 for the line source model and the
 489 traffic-induced dispersion model. For the line source model, the median concentration varies

490 significantly between each location, with the median at point 2 over 11 times higher than that at point
491 1. From Figure 8 it can be seen that point 2 is located in an area of high concentration formed by the
492 dominant in-canyon wind flow. However, for the case with traffic movement the median concentrations
493 are more consistent between each location, with the highest median, at point 4, less than twice that of
494 the lowest, at point 3. This lower variation between the median concentrations relative to the line
495 source model at points 1 to 4 is perhaps unexpected considering the different distances of each point to
496 the higher polluting bus lane and suggests that the moving vehicles are effective in mixing the emissions
497 at the crossroads.

498 The variation in concentrations at each point is also affected by the inclusion of traffic-induced
499 dispersion. The variation is increased at point 4, with an increase in the relative standard deviation (RSD)
500 from 0.29 for the line source to 0.72 with moving vehicles. Here we define the RSD as the standard
501 deviation, σ , over the mean, μ , such that $RSD = \sigma/\mu$. Despite the use of an instantaneous emissions
502 model, at points 1 and 3 the RSD is lower with the traffic dispersion model, at 0.06 and 0.13,
503 respectively, in comparison to 0.45 and 0.26, respectively, for the line source model. The RSD at point 2
504 is in relative agreement for the two models. With the inclusion of traffic-induced turbulence and
505 instantaneous emissions, point 4 experiences peak concentrations up to seven times greater than the
506 median. Point 4 is next to the bus lane and close to where the buses accelerate from standstill when
507 crossing the intersection. Higher concentrations are to be expected here relative to points 1, 2 and 3 due
508 to the higher emission rates for the buses. However the larger variation relative to the mean seen at this
509 point cannot be explained by the higher emissions alone as comparable RSD values would also be
510 expected at points 1, 2 and 3. Rather, the high RSD (0.72) and high number of outliers at point 4 is due
511 to the impact of the buses on the airflow as they drive past as seen in Figure 12. It can be seen from
512 Figure 12 that bus emissions tend to stay within the strong wake of the bus rather than disperse more
513 smoothly across the street as is the case for the car emissions in Figure 11. This leads to high
514 concentration gradients within the street. As the buses decelerate to stop for the lights, these areas of
515 high concentrations behind the buses are dispersed to the side of the road (as seen in Figure 12), leading
516 to the exposure of pedestrians and cyclists to large concentration peaks.

517 These results indicate that if the exposure of pedestrians and cyclists is to be modelled accurately,
518 exposure to acute concentration events must be considered. Let us estimate the exposure at any point
519 as $E_{NO_x} = \sum_i C_i t_i$, where C_i is the concentration at time t_i . At point 4, the top quartile (i.e. highest 25%)
520 of concentration values contribute to 48% of the overall exposure yet only occur for 23% of the total
521 time. Whereas the contributions of the outliers only, that is the values that exceed the upper quartile
522 plus 1.5 the interquartile range, contribute to 18% of the exposure and only 6% of the total time. These
523 outliers are therefore likely to contribute significantly to the exposure of pedestrians and cyclist while
524 waiting to cross the intersection. Although this analysis uses stationary points, while cyclists and
525 pedestrians will move across the intersection, it demonstrates the extent to which acute concentration
526 events can contribute to overall exposure. This is particularly significant for cyclists who have higher
527 inhalation rates and often share road space with buses.

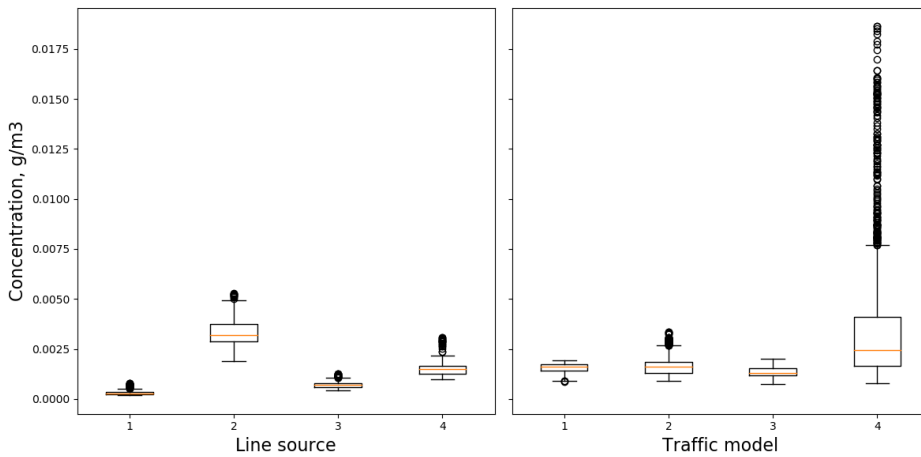
528 While this simulation represents one theoretical scenario, with a constant wind direction, the method
529 could be used for in depth analysis of pedestrian exposures at busy junctions.



530

531 *Figure 14: Location of points chosen for exposure analysis. Locations 1 to 4 are at a distance of 2 m from the nearest lanes of*
 532 *traffic.*

533



534

535 *Figure 15: Box plots of NOx concentrations at points 1-4 for the line source and traffic model. The lines show median values, the*
 536 *boxes represent the interquartile range and the whiskers (lines extending from the box) extend to the first point within 1.5 times*
 537 *the interquartile range.*

538 **5. Conclusions**

539 An idealized crossroads geometry is simulated using the CFD code Fluidity. The dispersion of pollutants
 540 is modelled in two ways: firstly wind driven dispersion using a line source to represent traffic emissions,
 541 and secondly using Fluidity's traffic model to capture traffic induced dispersion of instantaneous
 542 emissions. A low wind speed case is simulated in order to investigate a scenario where the impact of
 543 vehicle movement is likely to be at its most significant and dominates the turbulence within the canyon.
 544 For the second case, the emissions of each lane of traffic is simulated using individual passive tracers,
 545 with the emission rate of each vehicle calculated instantaneously as a function of the vehicle's velocity
 546 and acceleration. The emission model therefore captures emission peaks during high acceleration. A
 547 traffic simulation (PTV Vissum) is used to simulate the traffic flow dynamics which are used as an input
 548 to the Fluidity traffic model. A coarse mesh relative to vehicle size is used, with a minimum edge length

549 of 0.5 m, to limit the long run time required. Despite the low mesh resolution relative to vehicle size, the
550 method is still capable of simulating the effect of vehicles on pollution dispersion as demonstrated by
551 the validation studies presented in the appendix. This is a first attempt at using this method coupled
552 with traffic emissions to look at the dispersion of traffic emissions in an urban scenario. Further work is
553 required to reduce the run times of the simulation and to update the emissions model.

554 Comparison of the line source and traffic-induced dispersion simulations demonstrate the importance of
555 considering traffic induced flow and the dispersion of emissions by vehicles, along with instantaneous
556 emission rates, when considering urban concentrations. We observe that the inclusion of traffic
557 movement has an impact on the prevailing direction of air flow near ground level for this low wind
558 speed scenario, leading to notably different concentration fields to that given by the line source
559 simulation with no traffic movement. The traffic simulation demonstrates the ability of vehicles to
560 disperse emissions from other sources upwind and provides insights into the formation of pollution
561 hotspots at the intersection. The inclusion of moving vehicles and instantaneous emissions model has a
562 significant impact on the estimation of exposure of pedestrians and cyclists, particularly when large
563 vehicles such as buses are present. For the test case used here, it is shown that for a roadside location at
564 the intersection 2 m from the bus lane (point 4), the top quartile of concentration values contribute to
565 48% of the overall exposure yet account for only 23% of the total time. Similarly, the contribution of
566 extremely high concentration outliers, that is the values that exceed the upper quartile plus 1.5 the
567 interquartile range, contribute to 18% of the exposure while accounting for only 6% of the total time.
568 These acute high concentrations are not seen when the effect of traffic movement and instantaneous
569 emissions are not simulated. While these results are taken from a single simulation with a low wind
570 speed and one particular direction and therefore cannot be generalized, they serve as an example of the
571 importance of considering such effects in a typical urban scenario. Further, these results highlight the
572 limitations of using a line source to represent traffic emissions which are highly variable along the road.

573 The method presented in this paper can be used for detailed exposure analysis of pedestrians and
574 cyclists travelling through urban scenarios. The method is able to resolve the heterogeneous nature of
575 pollution dispersion within streets at high temporal and spatial resolution, therefore resolving the high
576 concentration peaks seen during measurement studies. In order to improve the accuracy of the
577 estimated exposure of active commuters these peak concentrations must be considered. The method
578 can also be used to improve our understanding of the effect of traffic movement on street level flow
579 features in urban areas and to improve the accuracy of simpler operational dispersion models.

580 **Acknowledgements**

581 This work was funded by the Engineering and Physical Sciences Research Council (EPSRC) Grand
582 Challenge grant 'Managing Air for Green Inner Cities (MAGIC) [grant number EP/N010221/1].
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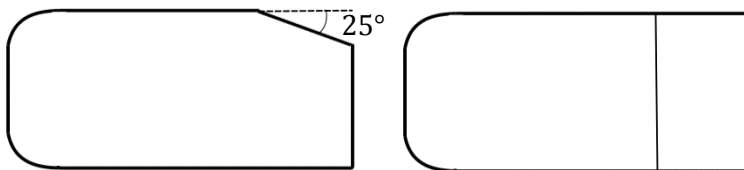
593 **Appendix A – Single vehicle simulation**

594 The single vehicle simulations consisted of single stationary car, modelled using the traffic model
595 described in Section 2, within a domain with a constant inlet and moving floor. This setup was designed
596 to simulate the flow over a vehicle at zero wind conditions.

597 The inlet velocity and floor velocity were set to 5 m/s in order to produce an equivalent simulation to a
598 vehicle moving at 5 m/s through a zero velocity flow field. The car size was the same as that used for the
599 crossroads simulation and the minimum edge length was set to 0.5 m.

600 Comparisons were made with two higher resolution simulations. These simulations were of the same
601 setup, however a physical, no-slip boundary was used to model the car rather than the traffic model.
602 The first of these simulations represented the car as an Ahmed body (Ahmed, 1981), the geometry of
603 which can be seen in Figure 16. The approaching wind velocity and floor velocity was set to 5 m/s. The
604 second simulation attempted to replicate the geometry used in wind tunnel experiments by Carpentieri
605 et al. (2012) (a 2004 Vauxhall AstraVan). While the Ahmed body model was full scale, as was the
606 crossroads simulation, the AstraVan model was a 1:20 model in order to replicate the wind tunnel setup.
607 The same velocity used in the wind tunnel experiment of 2.5 m/s was also used here. The minimum
608 edge lengths for these simulations was set to $\sqrt{A}/20$ where A is the cross-sectional area of the surface
609 facing the flow. For the full scale case, this results in a minimum edge length of 0.075 m. These mesh
610 size for these higher resolution simulations was approximately 100,000 nodes in comparison to 20,000
611 nodes for the low resolution vehicle model.

612 A passive tracer is emitted from the rear right-hand side of the vehicle at a constant emission rate for
613 each of the three simulations



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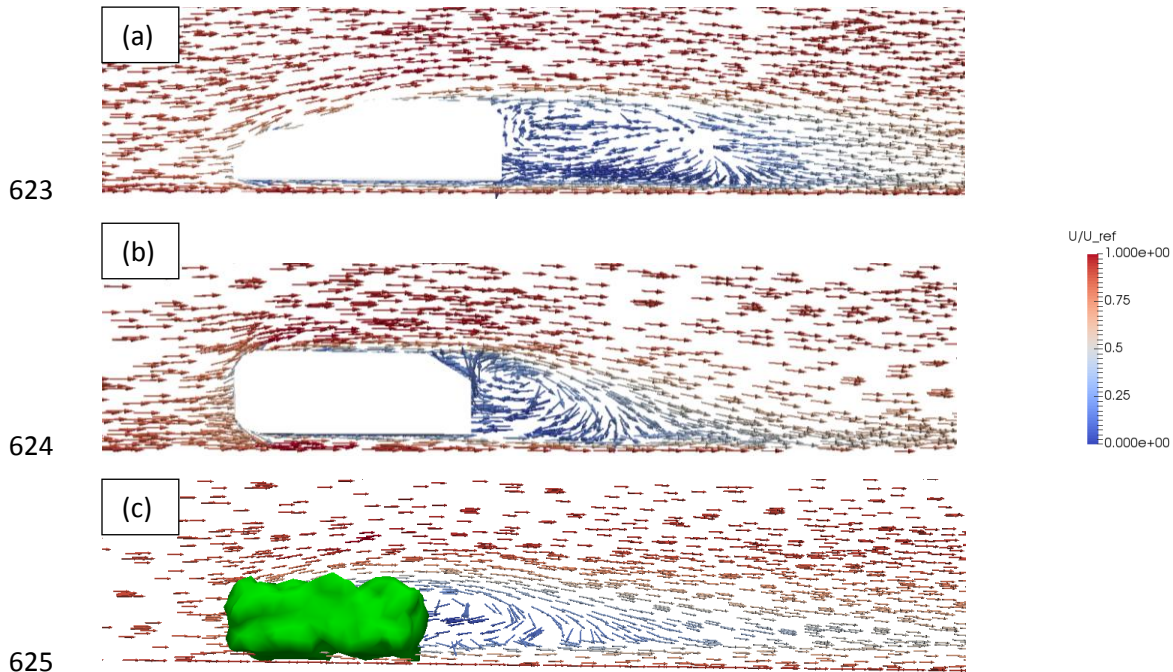
615 *Figure 16: Schematic of Ahmed body shape used for single vehicle simulation.*

616 Figure 17 and Figure 18 are vector plots of the average velocity around the vehicles. It can be seen that
617 despite the low resolution, the traffic model is able to capture the recirculation region behind the car.
618 However the full complexity of the flow is not resolved due to the relatively large minimum edge length
619 used.

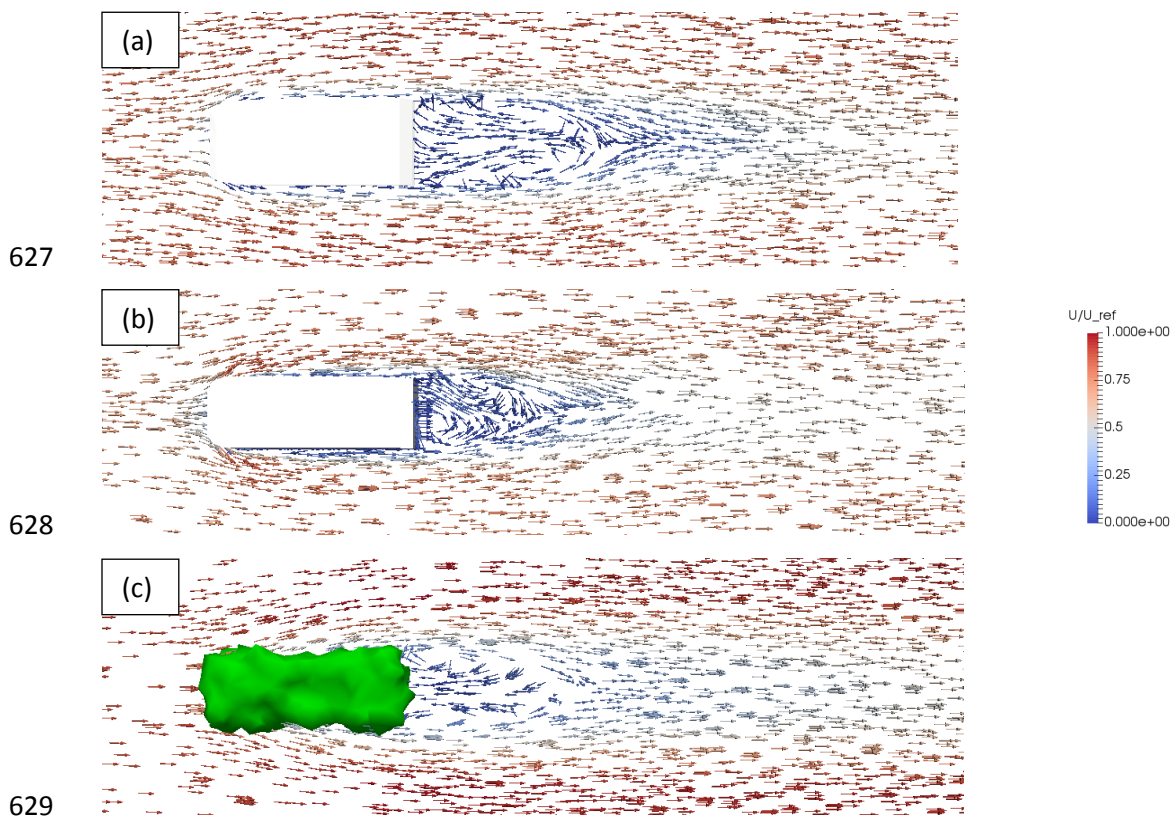
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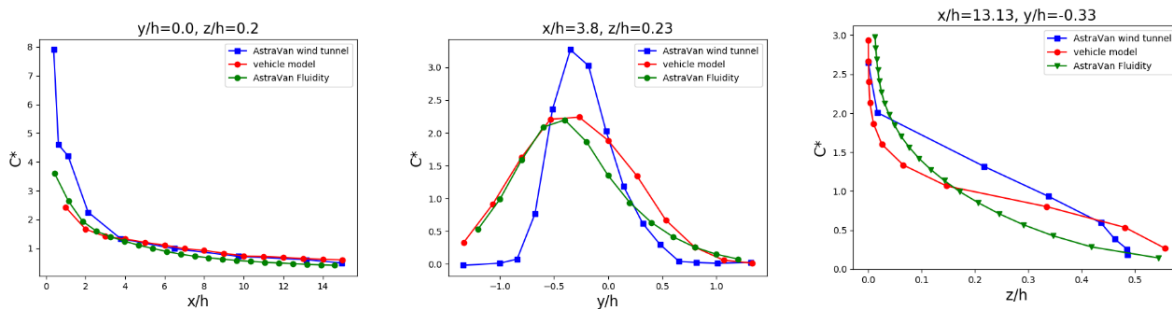
626 *Figure 17: Flow profile on $y=0$ plane for (a) the AstraVan (b) the Ahmed body and (c) the low resolution vehicle model*



630 *Figure 18: Flow profile on $z=0.5m$ plane for (a) the AstraVan (b) the Ahmed body and (c) the low resolution vehicle model*

631 Figure 19 show a comparison between the vehicle model, the wind tunnel experiment of Carpentieri et
 632 al. (2012) and the AstraVan Fluidity simulation for normalized exhaust tracer concentration, C^* . Here we

633 define $C^* = (C U h^2)/Q$, where C is the concentration, U is the approaching wind speed, h is the height
 634 of the vehicle and Q is the source mass flow rate. Differences in the profiles are to be expected for the
 635 low resolution traffic model, particularly in the near-wake region, as no attempt is made to replicate the
 636 shape of the AstraVan. There are also differences between the wind tunnel setup and that used in
 637 Fluidity. In the wind tunnel, the AstraVan model is positioned on a false floor above the ground of the
 638 wind tunnel and the tracer is released at a velocity of 0.13 times the inlet velocity. For the Fluidity
 639 simulations a moving floor is used and a zero-velocity tracer release. However, the objective for the low
 640 resolution model is not to perfectly replicate the flow pattern around each vehicle. Particularly as these
 641 profiles will vary from vehicle to vehicle due to different shapes and sizes. The low resolution traffic
 642 model is capable of providing an estimation of the impact of a typical vehicle on the dispersion of
 643 emissions as it moves through a domain. Further work is required to determine the optimal shape and
 644 size to represent the average car in the fleet.



645
 646 *Figure 19: Comparison of low resolution vehicle model with wind tunnel model of Carpentieri et al. (2012) and AstraVan Fluidity*
 647 *simulation.*

648 **Appendix B – Traffic induced turbulence for multiple vehicles**

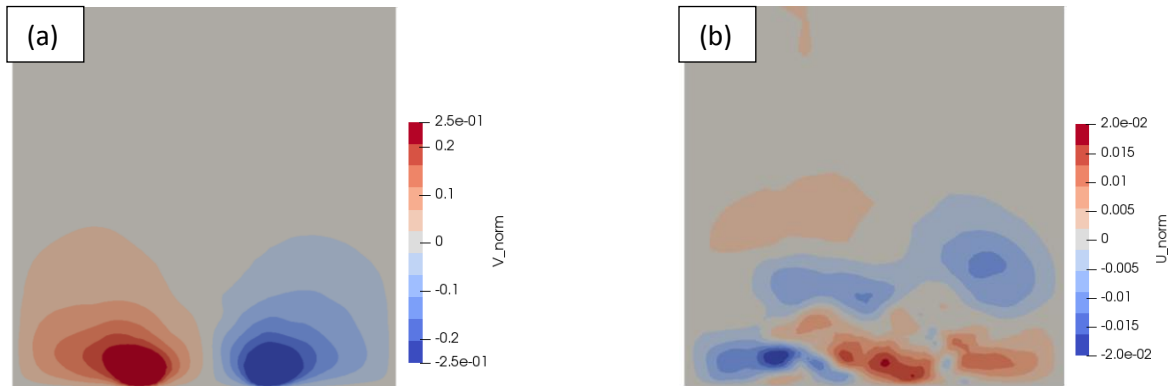
649 In order to assess the performance of the model when considering multiple vehicles, the wind tunnel
 650 experiment of Di Sabatino et al. (2003) was simulated. The wind tunnel experiment used moving metal
 651 plates on two belts to represent two lanes of vehicles moving along a street canyon in opposite
 652 directions, with no prevailing wind flow (i.e. zero wind conditions). A plate density of $20 m^{-1}$ was used
 653 with the plates moving at $12 m/s$, representative of a vehicle speed of $30 km/h$ in full scale. The street
 654 canyon was $120 cm$ long, and had equal height and width of $12 cm$. This setup was simulated using the
 655 Fluidity traffic model. Rather than using plates, the vehicle dimensions were kept as those used for the
 656 crossroad simulation however scaled down to wind tunnel size, with a ratio of 240:1 from full size to
 657 wind tunnel scale. A mesh of approximately 600,000 nodes was used, with a minimum edge length of
 658 $1.5 mm$, equivalent to $0.5 m$ in full scale as was used for the crossroads test case. This resulted in an
 659 average edge length within the canyon of approximately $3.4 mm$ which at full scale is equivalent to 0.8
 660 m . As for the crossroads simulation, a maximum Courant number of 5 was used to govern the size of the
 661 time step. 35 vehicles travelled along each lane during the averaging time.

662 Figure 19 shows the along-the-canyon and transverse components of the average velocity field while
 663 Figure 20 shows the along-the-canyon and transverse turbulent velocities. The values are normalized by
 664 the vehicle velocity. A qualitative comparison with the results of Di Sabatino et al. shows that the model
 665 provides reasonable approximation of the impact of the vehicles on the prevailing flow directions. The
 666 magnitude of the prevailing along-the-canyon flow induced by the vehicles is comparable for the
 667 simulation and wind tunnel, with a maximum value around 25% of the vehicle velocity reported by Di

668 Sabatino et al. Similarly, the transverse component is of similar magnitude for the two cases with both
669 an order of magnitude lower than the along-the-canyon flow. The along-the-canyon and transverse
670 turbulent velocities are also similar for the traffic model and the wind tunnel, with magnitudes up to
671 15% of the vehicle velocity for the along-the-canyon turbulent velocity, and slightly lower values for the
672 transverse velocity.

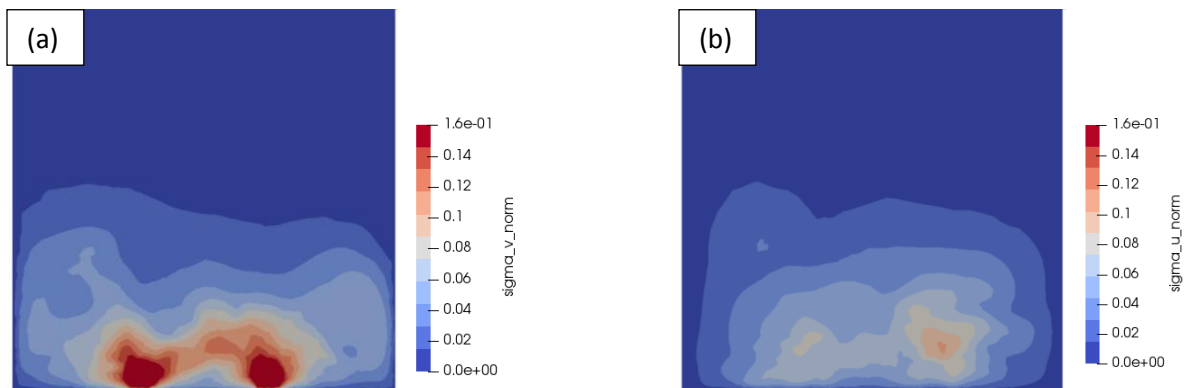
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676 *Figure 20: Traffic-produced normalised mean (a) along-the-canyon and (b) transverse velocity component in the central plane of an idealised street canyon.*
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678

679 *Figure 21: Traffic-produced normalised mean (a) along-the-canyon and (b) transverse turbulent velocity component in the central plane of an idealised street canyon.*
680

681 References

682 SR Ahmed, 1981. An experimental study of the wake structures of typical automobile shapes. Journal of
683 Wind Engineering and Industrial Aerodynamics. 9. 49-62.

684

685 SJ Arnold, H ApSimon, J Barlow, S Belcher, M Bell, JW Boddy, R Britter, H Cheng, R Clark, RN Colvile, S
686 Dimitroulopoulou, A Dobre, B Grealley, S Kaur, A Knights, T Lawton, A Makepeace, D Martin, M
687 Neophytou, S Neville, M Nieuwenhuijsen, G Nickless, C Price, A Robins, D Shallcross, P Simmonds, RJ
688 Smalley, J Tate, AS Tomlin, H Wang, P Walsh, 2004. Introduction to the DAPPLE Air Pollution Project.
689 Science of The Total Environment, 332, 139–153. doi:10.1016/j.scitotenv.2004.04.020

690

691 CJ Baker, 2001. Flow and dispersion in ground vehicle wakes. *Journal of Fluids and Structures*. 15. 1031-
692 1060. doi:10.1006/jfls.2001.0385

693

694 P Berghmans, N Bleux, L Int Panis, VK Mishra, R Torfs, M Van Poppel, 2009. Exposure assessment of a
695 cyclist to PM10 and ultrafine particles. *Science of the Total Environment*. 407. 12860-1298.

696

697 JHT Bentham, 2003. Microscale modelling of air flow and pollutant dispersion in the urban environment.
698 Ph.D. Thesis, Imperial College London.

699

700 M Carpentieri, P Kumar, A Robins, 2010. An overview of experimental results and dispersion modelling
701 of nanoparticles in the wake of moving vehicles. *Environmental Pollution*. 159. 685-693.

702

703 M Carpentieri, P Kumar, A Robins, 2012. Wind tunnel measurements for dispersion modelling of vehicle
704 wakes. *Atmospheric Environment*. 62. 9-25.

705

706 S Di Sabatino, P Kastner-Klein, R Berkowicz, RE Britter, E Fedorovich. 2003. The Modelling of Turbulence
707 from Traffic in Urban Dispersion Models – Part I: Theoretical Considerations. *Environmental Fluid*
708 *Mechanics*. 3. 129-143.

709

710 A de Nazelle, S Fruin, D Westerdahl, D Martinez, A Ripoll, N Kubesch, M Nieuwenhuijsen, 2012. A travel
711 mode comparison of commuters' exposures to air pollutants in Barcelona. 59. 151-159.

712

713 A de Nazelle, E Seto, D Donaire-Gonzalez, M Mendez, J Matamala, M J Nieuwenhuijsen, M Jerrett, 2013.
714 Improving estimates of air pollution exposure through ubiquitous sensing technologies. 176. 92-99.

715

716 CC Daigle, DC Chalupa, FR Gibb, PE Morrow, G Oberdörster, MJ Utell, MW Frampton, 2003. Ultrafine
717 particle deposition in humans during rest and exercise. *Inhalation Toxicology*. 15. 539-552.

718

719 G Dong, TL Chan 2006. Large eddy simulation of flow structures and pollutant dispersion in the near-
720 wake region of a light-duty diesel vehicle. *Atmospheric Environment*. 40. 1104-1116.

721

722 J Dong, Y Tao, Y Xiao, J Tu, 2017. Numerical simulation of pollutant dispersion in urban roadway tunnels.
723 *The Journal of Computational Multiphase Flows*. 9. 26-31.

724

725 X Garcia, D Pavlidis, GJ Gorman, JLMA Gomes, MD Piggott, E Aristodemou, J Mindel, JP Latham, CC Pain,
726 H ApSimon, 2011. A two-phase adaptive finite element method for solid-fluid coupling in complex
727 geometries. *International Journal for Numerical Methods in Fluids*. 66. 82-96. DOI: 10.1002/fld.2249

728

729 W Ham, A Vijayan, N Schulte, JD Herner, 2017. Commuter exposure to PM2.5, BC, and UFP in six
730 common transport microenvironments in Sacramento, California. *Atmospheric Environment*. 167. 335-
731 345.

732

733 S Hankey, JD Marshall, 2015. On-bicycle exposure to particulate air pollution: Particle number, black
734 carbon, PM2.5, and particle size. *Atmospheric Environment*. 122. 65-73.

735

736 M Irwin, H Bradley, M Duckhouse, M Hammond, MS Peckham, 2018. High spatio-temporal resolution
737 pollutant measurements of on-board vehicle emissions using ultra-fast response gas analyzers.
738 *Atmospheric Measurement Techniques*. 11. 3559-3567. <https://doi.org/10.5194/amt-11-3559-2018>

739
740 L Int Panis, S Broekx, R Liu, 2006. Modelling instantaneous traffic emission and the influence of traffic
741 speed limits. Science of the Total Environment. 371. 270-285.
742
743 L Int Panis, B de Geus, G Vandenbulcke, H Willems, B Degraeuwe, N Bleux, V Mishra, I Thomas, R
744 Meeusen, 2010. Exposure to particulate matter in traffic: A comparison of cyclists
745 and car passengers. Atmospheric Environment. 44. 2264-2270.
746
747 N Jarrin, S Benhamadouche, D Laurence, R Prosser, 2006. A synthetic-eddy-method for generating
748 inflow conditions for large-eddy simulations. International Journal of Heat and Fluid Flow. 27. 585-593.
749
750 P Kastner-Klein, E Fedorovich, MW Rotach, 2001. A wind tunnel study of organized and turbulent air
751 motions in urban street canyons. Journal of Wind Engineering and Industrial Aerodynamics. 89. 849-861.
752
753 P Kastner-Klein, E Fedorovich, M Ketzel, R Berkowicz, R Britter, 2003. The Modelling of Turbulence from
754 Traffic in Urban Dispersion Models – Part II: Evaluation Against Laboratory and Full-Scale Concentration
755 Measurements in Street Canyons. Environmental Fluid Mechanics. 3. 145-172.
756
757 Y Kim, L Huang, S Gong, CQ Jia, 2016. A new approach to quantifying vehicle induced turbulence for
758 complex traffic scenarios. Chinese Journal of Chemical Engineering. 24 (1). 71-78.
759
760 S Kingham, I Longley, J Salmond, W Pattinson, K Shrestha, 2013. Variations in exposure to traffic
761 pollution while travelling by different modes in a low density, less congested city. Environmental
762 Pollution. 181. 211-218.
763
764 R O'Driscoll, HM ApSimon, T Oxley, N Molden, MEJ Stettler, A Thiyagarajah, 2016. A Portable Emissions
765 Measurement System (PEMS) study of NOx and primary NO2 emissions from Euro 6 diesel passenger
766 cars and comparison with COPERT emission factors. Atmospheric Environment. 145. 81-91.
767
768 CC Pain, AP Umpheby, CRE de Oliveira, AJH Goddard, 2001. Tetrahedral mesh optimization and adaptivity
769 for steady-state and transient finite element calculations. Computer methods in applied mechanics and
770 engineering. 190. 3771-3796.
771
772 D Pavlidis, GJ Gorman, JLMA Gomes, CC Pain, H ApSimon, 2010. Synthetic-Eddy Method for Urban
773 Atmospheric Flow Modelling. Boundary-Layer Meteorology. 136 (2). 285-299.
774
775 W Pearce, CJ Baker, 1997. Wind-tunnel investigation of the effect of vehicle motion on dispersion in
776 urban canyons. Journal of Wind Engineering and Industrial Aerodynamics. 69-71. 915-926.
777
778 PTV Group. PTV Vissim. [Online]. Retrieved from [http://vision-traffic.ptvgroup.com/en-us/products/ptv-](http://vision-traffic.ptvgroup.com/en-us/products/ptv-vissim)
779 [vissim](http://vision-traffic.ptvgroup.com/en-us/products/ptv-vissim)
780
781 Y Qin, 1993. Dispersion of vehicular emission in street canyons, Gaungzhou City, South China (P.R.C.).
782 Atmospheric Environment. 27B (3). 283-291.
783
784 MS Ragettli, E Corradi, C Braun-Fahrländer, C Schindler, A de Nazelle, M Jerrett, RE Ducret-Stich,
785 N Künzli, HC Phuleria, 2013. Commuter exposure to ultrafine particles in different urban locations,
786 transportation modes and routes. Atmospheric Environment. 77. 376-384.

787
788 I Rivas , P Kumar, A Hagen-Zanker, M de Fatima Andrade, AD Slocic, JP Pritchard, KT Geurs, 2017.
789 Determinants of black carbon, particle mass and number concentrations in London transport
790 microenvironments. 161. 247-262.
791
792 E Solazzo, X Cai, S Vardoulakis, 2008. Modelling wind flow and vehicle-induced turbulence in urban
793 streets. Atmospheric Environment. 42. 4918-4931.
794
795 M Tientcheu-Nsiewe, D Tcheukam-Toko, F Murzyn, A Mehel, B Patte-Rouland, 2016. Numerical Study of
796 a Turbulent Flow in the Near-Wake of an Ahmed Body. American Journal of Environmental Engineering.
797 6(6). 157-163.
798
799 P Thaker, S Gokhale, 2016. The impact of traffic-flow patterns on air quality in urban street canyons.
800 Environmental Pollution. 208. 161-169.
801
802 F Yang, Daya Kaul, K Chun Wong, D Westerdahl, L Sun, Kin-fai Ho, L Tian, P Brimblecombe, Z Ning, 2015.
803 Heterogeneity of passenger exposure to air pollutants in public transport microenvironments. 109. 42-
804 51.
805
806 Y Zhang, Z Gu, C Wah Yu, 2017. Large eddy simulation of vehicle induced turbulence in an urban street
807 canyon with a new dynamically vehicle-tracking scheme. Aerosol and Air Quality Research. 17. 865-874.