Aerosol-CFD modelling of ultrafine and black carbon particle emission, dilution, and growth near roadways

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Abstract

Many studies have shown that on-road vehicle emissions are the dominant source of ultrafine particles (UFP; diameter < 100 nm) in urban areas and near-roadway environments. In order to advance our knowledge on the complex interactions and competition among atmospheric dilution, dispersion, and dynamics of UFPs, an aerosol dynamics-CFD coupled model is developed and validated against field measurements. A unique approach of applying periodic boundary conditions is proposed to model pollutant dispersion and dynamics in one unified domain from the tailpipe level to the ambient near-road environment. This approach significantly reduces the size of the computational domain, and therefore, allows fast simulation of multiple scenarios. The model is validated against measured turbulent kinetic energy (TKE) and pollution gradients near a major highway. Through a model sensitivity analysis, the relative importance of individual aerosol dynamical processes on the total particle number concentration (N) and particle number-size distribution (PSD) near a highway is investigated. The results demonstrate that (1) coagulation has a negligible effect on N and particle growth, (2) binary homogeneous nucleation (BHN) of H$_2$SO$_4$-H$_2$O is likely responsible for elevated N closest to the road, (3) N and particle growth are very sensitive to the condensation of semi-volatile organics (SVOCs), particle dry deposition, and the interaction between these processes. The results also indicate that, without the proper treatment of atmospheric boundary layer (i.e., its wind profile and turbulence quantities), the nucleation rate would be underestimated by a factor of 5 in the vehicle wake region due to overestimated mixing. Therefore, introducing ABL conditions to activity-based emission models may potentially improve their performance in estimating UFP traffic emissions.

1 Introduction

Many studies have shown that vehicle emissions are the dominant source of ultrafine particles (UFP; diameter < 100 nm) in urban areas and near-roadway environments.
For example, about 95 % of UFPs (diameter = 50 ∼ 100 nm) observed near a US freeway were apportioned to fresh vehicular emissions (Toner et al., 2008). In the more confined environment of a street canyon, over 99 % of particles in number were found to be below 300 nm and number concentrations of particles in this size range were found to be linearly correlated with the traffic volume (Kumar et al., 2008). Due to their small size and abundance in number, recent toxicological and epidemiological studies suggest a strong correlation between adverse health effects and personal exposure to UFPs (e.g. Brugge et al., 2007; Ruckerl et al., 2007; Valavanidis et al., 2008). A recent review study by Schlesinger (2007) pointed out that the health impacts of chemical constituents, such as sulphate, seem to be inconsistent across all epidemiological studies. Comparing epidemiological studies of heart rate variability in humans, Grahame (2009) suggests that differences in accuracy of exposure information for health-relevant emissions may explain conflicting study results. This has lead to an urgent need to study the temporal and spatial variations of local traffic emission in the vicinity of roadways.

With the growing concern of adverse health effects from exposure to UFPs, the gradients of vehicle-emitted pollutants (such as CO, NOx, and UFPs) have been measured in the ambient atmosphere near roadways (e.g. Beckerman et al., 2008; Reponen et al., 2003; Pirjola et al., 2006; Zhu et al., 2009). For example, Zhu et al. (2009) found that elevated particle numbers decay exponentially on the downwind side of three different types of roadways with increasing distance and reach background levels within a few hundred meters. Karner et al. (2010) synthesized field measurements of near-roadway pollutants from over 40 monitoring studies and investigated the concentration–distance relationship. The variation of UFP concentrations near roadways among studies is likely affected by factors including meteorological conditions (wind speed, ambient temperature, relative humidity, and atmospheric stability), traffic characteristics (volume and fleet composition), the geometry of roadways, and aerosol transformation processes (nucleation, coagulation, condensation/evaporation, and dry deposition). However, field measurements alone are often associated with such limitations as low spatial or tempo-
eral resolution in sampling, conclusions restricted by local meteorology, and difficulties in separating the effects of interactive processes.

Therefore, numerical modelling of UFPs has been conducted to address these limitations. Due to the challenge of resolving processes with very different scales, a two-stage dilution modelling strategy, including “tailpipe-to-road” and “road-to-ambient” dilutions, has been proposed (Zhang and Wexler, 2004; Zhang et al., 2004). In the first stage (i.e. “tailpipe-to-road”), strong vehicle induced turbulence (VIT) results in fast and strong dilution (dilution ratio $\sim 1000$ in 1 s) and triggers nucleation and condensation/evaporation. While in the “road-to-ambient” stage, atmospheric boundary layer turbulence (ABLT) continues to dilute exhaust particles with ambient air accompanied with particle size changes due to condensation/evaporation. A review study by Carpentieri et al. (2011) has shown that, with recent advances in numerical modelling, computational fluid dynamics (CFD) models can be valuable tools for nanoparticle dispersion in the first stage of dilution. In addition to the limited spatial scale of the dispersion investigated, other limitations in these most recent modelling studies include PSD and chemical composition not being explicitly resolved (Chan et al., 2010). Recent modelling studies of UFP dispersion on street level, on the other hand, have crudely simplified treatment of vehicular emission, VIT and aerosol dynamics (Gidhagen et al., 2004a, 2003; Kumar et al., 2009).

Most recently, Wang et al. (2013) proposed a two-stage simulation approach to integrate the “tailpipe-to-road” dispersion into the “road-to-ambient” dispersion stage for the first time. As the authors noted, however, the proposed approach remains computationally demanding, especially when both particle size and chemical composition need to be resolved. To effectively model UFP dynamics and dispersion near roadways in a single unified “tailpipe-to-ambient” domain, a unique approach of applying periodic boundary conditions to the computational domain is proposed in this paper. Compared to a “road-to-ambient” dispersion modelling approach, the advantage of a unified domain is that the uncertainty due to a simplified or non-existent treatment of VIT can be greatly reduced by explicitly modelling VIT. With VIT being explicitly modeled, aerosol
dynamics (such as nucleation, condensation and evaporation) triggered by the rapid first-stage dilution can be properly incorporated into dispersion models to study their effects on roadside air quality. From a modelling perspective, such a unified model provides a tool to link individual tailpipe emissions (controlled laboratory measurements) to roadside air quality (ambient field measurements), which is a noted challenging task (Keskinen and Ronkko, 2010). As the main focus of this paper, we present the development and validation of a multi-component sectional aerosol dynamics-CFD coupled model to account for the complex dilution, dispersion and dynamics of UFPs immediately after tailpipe emission to ambient background.

2 Aerosol dynamics-CFD coupled model

2.1 Multiphase approach to the mixture of atmospheric gas and aerosol

Following previous studies (e.g. Li et al., 2006b; Uhrner et al., 2007; Wang et al., 2013; Albriet et al., 2010), the commercial CFD code ANSYS FLUENT is used to model turbulent flow around realistically-shaped vehicles. An Euler–Euler approach to multiphase atmospheric air flow is coupled with new particle formation, transformation and dry deposition processes. Among the general multiphase models available in FLUENT, the mixture multiphase model is chosen for the present work due to its superior numerical efficiency compared to the Eulerian multiphase model (ANSYS, 2009a). The concept of “phase” in the FLUENT multiphase model is defined in a broad sense as an identifiable class of material that has a particular inertial response to and interaction with the flow (ANSYS, 2009b). The gas phase and the particulate phase in the real atmosphere are represented in the model by the “primary phase” and a number of “secondary phases”, respectively. The number of “secondary phases” is determined by the number of discrete size bins used to resolve the particle number-size distribution. The flow field of the “mixture phase” is obtained by numerically solving Reynolds-averaged Navier–Stokes equations (RANS) including conservation equations of mass, momen-
tum and energy with the standard $k-\varepsilon$ turbulence model. For the transport of gas and particulate species, FLUENT predicts the local mass fraction of each species by solving the advection–diffusion equations. The volume fraction of each secondary phase in a control volume (equivalent to the number concentration of particles of the same size) is obtained by numerically solving the continuity equation for the secondary phase with a specified source term due to aerosol dynamic processes. Thus, the advection, the turbulent mixing, and the diffusion of gases and particles are inherently treated by FLU-ENT through the continuity equation for each phase. Aerosol dynamic processes, which change the chemical components in particle and gas phases, are integrated through the source terms in continuity equations, and incorporated into FLUENT through User-Defined Functions (UDF).

2.2 Aerosol dynamics

Each secondary phase is a particulate phase composed of mixed chemical components within a specified size range. The density of particles in a given size bin is dynamically computed by FLUENT based on the volume-weighted mixing law. From the continuity equation for each secondary phase $\rho$, the volume fraction of the secondary phase ($\alpha_p$) is obtained by solving

$$\frac{\partial}{\partial t}(\alpha_p \rho_p) + \nabla \cdot (\alpha_p \rho_p \mathbf{u}_m) = -\nabla \cdot (\alpha_p \rho_p \mathbf{u}_{dr,p}) + S_\alpha,$$

where $\mathbf{u}_{dr,p}$ is the drift velocity of secondary phase, and $S_\alpha$ is the source of mass transfer due to aerosol dynamic processes. Number concentration of particles in size bin $\rho$ ($N_p$, in particles cm$^{-3}$) is computed from the ratio of the phase volume fraction solved by FLUENT to the particle volume of a certain size:

$$N_p = 10^{-6} \cdot \frac{\alpha_p}{(4/3)\pi(D_p/2)^3},$$
where $D_p$ is the diameter (in m) for particles in size bin $p$. And the local mass concentration of chemical component $i$ from particles in size bin $p$ is calculated from the phase volume fraction ($\alpha_p$) and the local mass fraction ($Y_i$) as:

$$m_i = 10^9 \cdot \alpha_p \rho_p Y_i.$$ (3)

The underlying implementation of aerosol dynamics is a multi-component, size-resolved, sectional aerosol model, as described as follows.

### 2.2.1 Nucleation

Immediately after tailpipe emissions, new particles form by homogeneous nucleation with initial particle size around 1.5–2.0 nm in the first few milliseconds of exhaust cooling and dilution (Kulmala et al., 2007). A qualitative investigation by Zhang and Wexler (2004) found that sulphuric acid-induced nucleation could be the dominant new particle production process. The experimental study conducted by Arnold et al. (2006) observed a positive correlation between gaseous sulphuric acid and particle number in the exhaust of a passenger diesel car burning ultra-low sulphur fuel, indicating an important role of sulphuric acid-induced nucleation. The sulphuric acid gas emission rate is estimated based on fuel sulphur content following Uhrner et al. (2007). The parameterization of BHN of $\text{H}_2\text{SO}_4-\text{H}_2\text{O}$ (Vehkamaki et al., 2003) developed specifically for engine exhaust dilution conditions is implemented in this study. This parameterization has already been successfully used in a number of different aerosol-CFD applications (e.g. Uhrner et al., 2007, 2011; Albriet et al., 2010; Wang and Zhang, 2012).

### 2.2.2 Coagulation

Particles in the exhaust plume collide due to random (Brownian) motion and turbulent mixing to form larger particles, which is called coagulation. The coagulation process reduces $N$ (mainly in the smaller size range) while preserving the aerosol total mass.
However, it modifies the particle number size distribution, and internally mixes particles of different chemical composition over the population. Coagulation may be driven by Brownian motion, turbulent flow conditions, gravitational collection, inertial motion and turbulent shear respectively. Individual coagulation rate coefficients (or coagulation kernels) due to the above driving forces are calculated in this work based upon (Jacobson, 2005), with consideration of particle flow regimes and convective Brownian diffusion enhancement. The overall coagulation rate coefficient is the summation of individual coefficients.

### 2.2.3 Condensation and evaporation

A complex mixture of condensable gases, including water vapor, sulphuric acid and semi-volatile organics (SVOCs), is emitted from the tailpipe after fuel combustion in the engine. During the strong dilution and cooling stage of up to a few seconds after emission (Zhang and Wexler, 2004), supersaturation of these condensable gases occurs and favors the diffusion-limited mass transfer process from the gas phase to the pre-existing particle phase. Following primary emission and nucleation, condensation of SVOCs was suggested by a number of studies (e.g. Clements et al., 2009; Wang and Zhang, 2012; Albriet et al., 2010; Uhrner et al., 2011; Mathis et al., 2004) to be responsible for the rapid growth of nanoparticles in the exhaust plume. As the reverse of condensation, evaporation occurs due to further dilution of condensable gases to subsaturation level in the air surrounding exhaust particles. It was suggested by field measurements of freeway emissions from predominantly gasoline vehicles that lower ambient temperature may favor the condensation of organic species to the particle phase (Kuhn et al., 2005). In this study, the net mass transfer rate of a condensable gas from/to an existing particle with multiple components is driven by the difference between the bulk partial pressure and the saturation vapor pressure above the particle surface (Jacobson, 2005). The calculation of mass transfer rate implements corrections to the diffusion coefficient, the thermal conductivity of air, and the saturation...
vapor pressure over curved particle surfaces to reflect its dependence on particle size and chemical composition.

### 2.2.4 Dry deposition

Driven by mechanisms such as Brownian diffusion, turbulent diffusion, sedimentation, and advection, dry deposition removes particles at the air–surface interface when they contact and remain on the surface (Jacobson, 2005). Brownian diffusion is more effective in removing smaller particles due to their larger diffusion coefficient, while sedimentation is more important for larger particles whose fall speeds are much higher. In the current study, parameterization of particle dry deposition follows the size-resolved dry deposition scheme developed by (Zhang et al., 2001). The effect of turbulent mixing on particle dry deposition is taken into account by the locally calculated friction velocity. This parameterization has been successfully validated and implemented in a number of air quality and climate studies (e.g. Gong et al., 2003; Pye and Seinfeld, 2010), and it has recently been improved and extended (Petroff and Zhang, 2010). The recent development accounts for more detailed characteristics of the surface canopy, and suggests possible overestimation of dry deposition velocity for particles in the fine mode. Thus, our current study is likely biased to overestimate the removal of UFPs by dry deposition.

### 2.3 Modelling turbulence

For turbulence modelling, although the large eddy simulation (LES) approach has been reported to be a more promising solution, the standard \(k-\varepsilon\) turbulence model is implemented in this work for several reasons. First, the high computational demands of the LES approach prevent its application for modelling the dispersion and transformation of multiple pollutants with complex geometry (i.e. gas and particle emissions and aerosol dynamics from multiple vehicles in this study). Compared to RANS closures, the LES approach is at least one order of magnitude more computationally expensive.
Aerosol-CFD modelling of ultrafine and black carbon particle emission

L. Huang et al.

ACPD 14, 12235–12278, 2014

(Rodi, 1997). Secondly, a proper treatment of the atmospheric boundary layer (ABL) has proven to be crucial to dispersion modelling studies (Blocken et al., 2007b; Zhang, 1994). Recent advances made by (Balogh et al., 2012; Parente et al., 2011a, b) permit a general and practical means to include the ABL using the standard \( k-\varepsilon \) model. To achieve this in LES simulations, on the other hand, inflow conditions would have to be carefully generated with additional, significant computational overhead (Xie and Castro, 2008; Li et al., 2006b). Finally, RANS models agree reasonably well with experimental data in predicting mean flow and pollutant concentrations (e.g. Labovsky and Jelemsnky, 2011; Sklavounos and Rigas, 2004). Kim et al. (2001) successfully modeled the dispersion of a truck exhaust plume in a wind tunnel using \( k-\varepsilon \) turbulent closure focusing on rapid dilution and turbulent mixing of exhaust \( \text{CO}_2 \).

Although CFD codes have been widely adopted in pollutant dispersion modelling, the accuracy of such simulations can be seriously compromised when wall functions based on experimental data for sand-grain roughened pipes and channels are applied at the bottom of the computational domain (Blocken et al., 2007b; Riddle et al., 2004). Attempts have been made to better predict ABL flow by changing turbulent model constants, tuning boundary profiles, and modifying wall functions and turbulent transport equations (Blocken et al., 2007a; Pontiggia et al., 2009; Li et al., 2006a; Alinot and Masson, 2005; Hargreaves and Wright, 2007; Labovsky and Jelemsnky, 2011; Balogh et al., 2012; Parente et al., 2011a, b). Among the aforementioned studies, recent advances made by Balogh et al. (2012) and Parente et al. (2011a, b) are implemented in this work, which permits a general and practical means to include ABL using the standard \( k-\varepsilon \) model in the CFD code, FLUENT. The ABL profiles of mean velocity, TKE, and dissipation rate for atmospheric flow under neutral stratification conditions
(Richards and Hoxey, 1993) are:

\[ u = \frac{u^*}{k} \ln \left( \frac{Z + Z_0}{Z_0} \right) \]  \hspace{1cm} (4)

\[ k = \frac{u^{*2}}{\sqrt{C\mu}} \]  \hspace{1cm} (5)

\[ \varepsilon = \frac{u^{*3}}{k(Z + Z_0)} \]  \hspace{1cm} (6)

A modified wall function for turbulent mean velocity following (Parente et al., 2011b):

\[ u = \frac{u^*}{k} \ln (E' z') \]  \hspace{1cm} (7)

is implemented through UDF and applied to wall adjacent cells, where \( E' = \frac{\nu}{z_0 u^*} \) and \( z' = \frac{(z + z_0) u^*}{\nu} \). To keep the default constant value of \( \sigma_{\varepsilon} \) in the standard \( k-\varepsilon \) model, a source term is added to the dissipation rate equation as follows:

\[ S_\varepsilon(z) = \frac{\rho_m u^{*4}}{(Z + Z_0)^2} \left( \frac{(C_2 - C_1)\sqrt{C\mu}}{k^2} - \frac{1}{\sigma_{\varepsilon}} \right). \]  \hspace{1cm} (8)

3 Simulation setup

3.1 Computational domain and flow boundary conditions

The Fast Evolution of Vehicle Emissions from Roadway (FEVER) study was conducted to monitor pollutant gradients perpendicular to a major highway north of Toronto,
Canada (Hwy-400) when the predominant wind was perpendicular to the highway (Lig-gio et al., 2012; Gordon et al., 2012a, b; Stroud et al., 2014). The model developed and tested in this paper was setup to simulate the observations from FEVER. This was a 6-lane (25 m across from the lane edges) highway site, which was mainly surrounded by flat agricultural fields and some trees lining on the side roads, with negligible local pollution sources other than vehicular emissions. Therefore, it is ideal for monitoring, as well as modelling, the dispersion of fresh vehicular emissions without interference of complex terrain or multiple sources.

The computation domain for the base case simulation in this study is a cuboid, as shown in Fig. 1a. The top of the domain is set to 50 m above ground so that the turbulent flow near the surface is not affected by the top boundary (the x-y plane in purple mesh). The horizontal dimension of 375 m perpendicular to the highway (x axis) is determined by considering the availability of measurements and the extent of pollutant dispersion (Gordon et al., 2012a). To reduce computational overhead, first, the actual two-way, 6-lane traffic fleet is represented by a one-way, 3-lane traffic fleet (as shown in Fig. 1b) while conserving the total traffic volume; Then, translational periodic boundary conditions are applied to the x-z planes (in Cyan mesh) to account for the effects of the stable, continuous traffic fleet on the highway by numerically repeating the computational domain in the direction of y axis. Based on the measured traffic volume of about 104.3 passenger vehicles/min, travelling speed of approximately 120 km h⁻¹ (or 33.3 m s⁻¹), and the assumed average vehicle length of 4.5 m, the average y axis distance (bumper to bumper) between two vehicles travelling in adjacent lanes is calculated as 11.5 m assuming all 3 lanes are evenly occupied. Thus, the horizontal dimension of 48 m along the highway (y axis) is calculated based on the measured traffic volume for weekday early morning rush hours between 06:00–08:00 a.m. The whole domain is meshed into 871 065 unstructured hexahedral cells with the finest ones concentrated around the moving vehicles, tailpipes, and their wake regions and immaterially above the ground.

In our simulation, the vehicles are set to be stationary, nonslip walls with a roughness length of 0.0015 m (Wang and Zhang, 2009), while the blowing air has two velocity
components: the first component towards the vehicles (or the negative $y$ axis direction) of 33.3 m s$^{-1}$ in magnitude; and the second component perpendicular to the highway (or the negative $x$ axis direction) in a form given by Eq. (4). The first component of wind velocity accounts for the relative movement between the moving vehicles and still air, and the second component describes the observed wind speed according to a fully developed ABL wind profile under neutral stratification. Thus, the upwind side boundary parallel to the road (the $x$-$y$ plane in purple mesh) and the top boundary are set as velocity inlets. The ground surface is set to have the same velocity magnitude as the running vehicles but in the opposite direction. The modified wall function (Eq. 7) and the additional source term to dissipation rate (Eq. 8) as described in Sect. 2.3 are applied to the ground surface to account for a fully developed ABL turbulent flow. Translational periodic boundary conditions are set to the $x$-$z$ planes of the domain, and a pressure outlet boundary is applied to the boundary (in red mesh) at the far end side to the highway. Each tailpipe, 52 mm in diameter, is specified as mass flow inlet with a mass flow rate of 0.055 kgs$^{-1}$ and an exhaust temperature of 480 K (Uhrner et al., 2007). An O-Grid composed of 7 inflation layers, whose thickness gradually increase from 0.003 m, around vehicles is used to allow the standard wall functions to apply to the fully-turbulent layer around moving vehicles. Crucial but not mentioned in previous CFD models, a high spatial resolution applied here results in a dimensionless wall distance ($y^+ = \rho u_* y / \mu$) of about 90 at the vehicle surface, which is well within the suggested range of 30 to 300 for the standard wall functions to apply (ANSYS, 2009b).

3.2 Chemical boundary conditions: background concentrations and traffic emissions

In addition to meteorological and traffic data, chemical data of gases and particles are required as part of CFD boundary conditions. According to the source type, the required chemical data are divided into two categories: background concentrations and traffic emission rates. The mass concentrations of background gaseous and particulate species from the FEVER field measurements are listed in Table 1, with their corre-
sponding values used as model input. The background gas phase includes dry air (O\textsubscript{2} and N\textsubscript{2}), water vapor and CO\textsubscript{2}. The CO\textsubscript{2} volume fraction and RH value are converted to mass fractions to specify the species input values for the velocity inlet boundaries. For the particulate phase, a bi-modal log-normal particle size distribution is assumed (as summarized in Table 3). The parameters of background particles are obtained from the FEVER measurements at about 100 m upwind of Hwy-400. Given the total background N and PSD, the volume fractions of individual size bins are obtained, and the mass of black carbon (BC) and organic aerosol (OA) are distributed into each size bin according to the ratio of their background mass concentrations listed in Table 1.

Vehicles driving on the highway continuously emit a complex mixture of gases and particles. It is not possible to include a complete set of gaseous and particulate species in the model, which also is not numerically practical. In this study, the tailpipe emission rates of the gaseous and particulate species are summarized in Table 2. Currently, the exhaust gas is composed of CO\textsubscript{2}, H\textsubscript{2}O, H\textsubscript{2}SO\textsubscript{4}, SVOCs, and N\textsubscript{2}, which are key species to the aerosol dynamics and dispersion. The treatment of H\textsubscript{2}SO\textsubscript{4} as direct emission rather than a mixture of SO\textsubscript{2} and SO\textsubscript{3} followed by hydrolysis has been explained in Sect. 2.2.1. It was suggested by modelling single exhaust plumes (e.g. Albriet et al., 2010; Uhrner et al., 2007, 2011) that SVOCs are likely to be responsible for the rapid growth in particle size when they condense on UFPs. Following Albriet et al. (2010), pyrene (C\textsubscript{16}H\textsubscript{10}), n-nonadecane (C\textsubscript{19}H\textsubscript{40}), and n-pentacosane (C\textsubscript{25}H\textsubscript{52}) are introduced to represent the polycyclic semi-volatile organic compounds, the semi-volatile alkanes between C\textsubscript{14} and C\textsubscript{22}, and semi-volatile alkanes between C\textsubscript{23} and C\textsubscript{29}, respectively. The mass fractions of the above three groups of SVOCs are based on Albriet et al. (2010), and the total mass emission rate of SVOCs is set as 0.0186 gkm\textsuperscript{-1} (Pye and Seinfeld, 2010). All SVOCs initially from the tailpipe are assumed to exist only in the gas phase, but are subject to interactions with the particle phase through condensation/evaporation upon immediate dilution with the surrounding air. To reduce the number of species considered in the model, the non-volatile fraction of primary organic aerosol (POA) from tailpipes is assumed to share the properties of the back-
ground OA, i.e. with an average molecular mass of 300 g mol$^{-1}$ and an average density of 1.5 g cm$^{-3}$.

This assumption is not likely to affect our results because the amount of the non-volatile fraction of POA from tailpipes is very small compared to the background OA. N and PSD for tailpipe emissions are based on a recent study by Nikolova et al. (2011b), which provides an emission rate according to traffic volume and type. As pointed out by Nikolova et al. (2011a), however, their original proposed parameterization implicitly accounts for fast nucleation process. As indicated by laboratory measurements (Ronkko et al., 2007; Kirchner et al., 2009), the nucleation mode particles have a nonvolatile core in the exhaust of a heavy duty diesel vehicle; however, they are completely volatile under 280°C in the exhaust of a diesel passenger car. Thus, we assume in this study that N of nucleation mode particles from all passenger cars are from BHN, while those from heavy duty vehicles have a solid core of BC and nonvolatile POA. Given the mass flow rate of the tailpipe exhaust, the mass fraction of each individual species can be estimated from its mass emission rate listed in Table 2. These mass fractions are used to specify chemical boundary conditions for tailpipes.

4 Results and discussions

Turbulent mixing of tailpipe emissions with the ambient air largely determines the initial dilution and the three-dimensional distribution of the traffic pollutants downwind of roadways. Thus, the modeled TKE is first compared against on-road and near-road TKE measurements reported by Gordon et al. (2012b). The modeled concentration gradients of CO$_2$, BC and PSDs are compared with the FEVER field measurements. Finally, the impacts of individual aerosol dynamical processes on UFPs and model sensitivity to the treatment of ABL are investigated.
4.1 Model validation

4.1.1 Turbulent kinetic energy

Both theoretical (e.g. Zhang and Wexler, 2004; Ketzel and Berkowicz, 2004) and monitoring (e.g. Zhou and Levy, 2007) studies have concluded that dilution is the dominant mechanism governing N of UFPs. And the turbulent kinetic energy (TKE) measures the strength of mixing and dilution. Figure 2 compares the modeled TKE with the measurements from the FEVER chasing experiments of passenger cars on highways. The x axis in Fig. 2 is the following time behind passenger cars and the y axis is 10 s average TKE. The modeled TKE values in Fig. 2 are calculated for individual vehicle wakes super-imposed on an estimated background on-road TKE of 2.4 m² s⁻² (Gordon et al., 2012b). The average travelling speed during the FEVER chasing experiments was about 20 m s⁻¹, and 84.5% of the measurements were taken at a chasing speed between 15 and 25 m s⁻¹. Therefore, model simulations are conducted for passenger cars travelling at 15, 20, and 25 m s⁻¹. As shown in Fig. 2, the modeled TKE in the wake of a vehicle travelling at speed of 15–25 m s⁻¹ agrees well within the 25th and 75th percentile of the measurements. And the variations among modeled TKE in Fig. 2 show the sensitivity of on-road TKE to vehicle type and travelling speed. With these scenarios agreeing within the 25th and 75th percentile of the measurements, it is clear that the turbulent mixing within individual vehicle wakes on the highway can be reasonably well modeled.

The decay of the turbulent mixing strength under perpendicular wind conditions is also modeled and compared to the turbulence data collected at a fixed tower adjacent to the highway. Two time periods (05:00–06:00 and 06:00–08:00 a.m. local time) with distinctly different traffic volumes are considered. For both time periods, the atmospheric boundary layer was neutrally stratified. The average traffic volume, however, increased from about 55 to 105 vehicles per minute, as shown in Fig. 2 of (Gordon et al., 2012a). Although the tower is stationary, the distance and time along the wind trajectory from the highway center vary with changing wind direction and speed. Therefore,
the evolution of turbulence with distance can be investigated based on measurements at a fixed location in a Lagrangian sense. There are 120 and 240 measurements of TKE taken for the 05:00–06:00 and 06:00–08:00 a.m. time periods, respectively, and they are binned by distance as shown in Fig. 3. The obtained wind trajectory distances vary between 20 to 80 m with about 93% of them concentrating on the first bin (20–40 m). For the period of 05:00–06:00 a.m., the measured (modeled) TKE at a distance of 20–40 m from the highway center is in the range of 0.46–0.80 (0.58–0.73) m² s⁻². Similarly, for the period of 06:00–08:00 a.m., the measured (modeled) TKE lies in the range of 0.55–0.90 (0.65–0.95) m² s⁻². Although the observed TKE decay is limited in spatial resolution for both time periods, the comparison in Fig. 3 shows an adequate agreement with the field measurements and suggests turbulent mixing in a roadside environment can be successfully modeled even with varying traffic volumes.

4.1.2 Near-road concentration gradients: CO₂ and BC

As a chemically passive gas species in vehicular emissions, CO₂ is an ideal indicator of atmospheric mixing of tailpipe exhaust with ambient air. In a previous study (Kim et al., 2001), CO₂ was experimentally measured inside a single turbulent plume of heavy-duty truck exhaust and successfully modeled with the standard $k-\varepsilon$ model in the CFD code FLUENT. The focus of this study, however, is the concentration gradient on the downwind side of a highway.

Figure 4a shows the concentration of CO₂ (ppmv) as a function of downwind distance from the center of Hwy-400 for the morning period of 06:00–08:00 a.m. FEVER measurements were first corrected to wind trajectory distance, grouped into 20 m bins between 50 and 350 m, and then plotted in median concentrations and 25th and 75th percentiles. Modeled CO₂ concentrations closely follow the decreasing trend of the median values of the FEVER measurements, and agree well within the 25th and 75th percentiles. However, the model tends to underestimate CO₂ concentrations by about 6 ppmv in the first 50 m (i.e. 50–100 m) of downwind distance and overestimate by about 8 ppmv in the last 50 m (i.e. 250–300 m). Similarly, the concentration–distance
relationship for particulate BC is shown in Fig. 4b. Modeled BC concentrations are also within the 25th and 75th percentiles exhibiting a trend with distance similar to the median of the measured values. Similar to CO₂, minor underestimations (15 %) between 50–100 m from Hwy-400 and slight overestimations (20 %) after 100 m were observed for particulate BC.

This behavior of the model suggests slightly overestimated pollutant concentrations between 0–3 m above the ground within 100 m distance to the road, possibly due to underestimated vertical mixing by the model. Beyond 100 m away from the road, vertical diffusion of the near surface pollutants results in the overestimations. There are two factors that may explain the underestimated vertical mixing closest to the road. Firstly, the modeled road structure is missing a 1-m-high barrier at the highway center, which could potentially lift near surface pollutants under cross wind conditions (Ning et al., 2010; Hagler et al., 2011). Secondly, midsize vehicles and trucks are neglected due to their small fractions in total traffic, which emit pollutants at a greater height up to 4 m than 0.5 m for passenger cars (Gordon et al., 2012a).

### 4.1.3 Particle size distribution

The fate of atmospheric particles depends strongly on PSD, which is the result of the complex influences of mobile emissions, atmospheric dilution and transformation processes. In Fig. 5, the predicted PSDs are compared to SMPS measurements for two selected periods of the early morning rush hours (i.e. 05:00–06:00 and 06:00–08:00 a.m.) at two fixed locations: Sites B and C, located 34 and 300 m from Hwy-400, respectively. The early morning rush hours are subdivided into the above two periods based on hourly averaged traffic flow. According to Gordon et al. (2012a), the traffic flow of 06:00–08:00 a.m. (105 veh min⁻¹) almost doubled the average traffic of 05:00–06:00 a.m. (55 veh min⁻¹), while the ambient conditions (such as atmospheric stability class, incoming solar radiation, and wind velocity) remained approximately constant.

Comparing the measured PSDs at two sites, it was found that for all measured particle sizes, the number concentrations decreased significantly when particles were trans-
ported from 34 to 300 m downwind of the highway. The observed total particle number concentrations decreased by a factor of about 2.5 and 5.7 between these two locations for the periods of 05:00–06:00 and 06:00–08:00 a.m., respectively. Similar to many previous roadside monitoring studies reviewed by Pant and Harrison (2013), the measured PSDs showed distinct multi-modal size regimes (i.e. nucleation, soot and accumulation modes) in the measured size range of 15–700 nm. Tri-modal lognormal curve fitting of the observed PSDs revealed a nucleation mode at 20–25 nm, a soot mode at 65–75 nm and an accumulation mode at 160–380 nm. In agreement with Zhu et al. (2002b, a), the nucleation mode particles dominated N and decreased much faster (by a factor of 8 in this study) than that of the accumulation particles (only a factor of 2 in this study). It was also found that the geometric mean diameter of the nucleation mode increased from 20.0 (at Site B) to 24.7 nm (at Site C), which may be attributed to the condensation of SVOCs (Clements et al., 2009) and the coagulation of nucleation mode particles (Zhu et al., 2002b).

The comparison in Fig. 5 demonstrates an adequate agreement between the modeled and the observed PSDs at both distances under different traffic conditions. For the peak traffic hours during 06:00–08:00 a.m., the model estimated total particle number concentrations are $6.25 \times 10^4$ and $1.66 \times 10^4$ particles cm$^{-3}$ at Sites B and C, respectively with approximately 10% underestimations compared to the observations. Secondly, the dominant nucleation mode was properly captured by the current model, as well as its decreasing trend with increasing distance from the highway. Furthermore, the nucleation mode particles were modeled to grow from 19.8 to 24.3 nm in geometric mean diameter with increasing distance away from the highway. This agrees exceptionally well with the observations. Similar conclusions can be drawn from the 05:00–06:00 a.m. comparison.

However, the model clearly underestimated the number concentrations of particles of 100–730 nm in diameter. This discrepancy can be attributed, at least partially, to the missing non-tailpipe emissions in the current model, such as brake wear, road-tyre interaction and re-suspension of road dust as reviewed by Kumar et al. (2013). Although
road dust particles formed mechanically by frictional contact between road surface and tire or between break system components are assumed to be primarily coarse particles, both laboratory experiments (Dahl et al., 2006; Gustafsson et al., 2008) and real-world measurements (Mathissen et al., 2011) recently observed a significant portion of particles of 6–700 nm in diameter. On the other hand, the estimated emission factors for sub-micrometer particles generated by the road-tire interaction under steady driving condition based on these studies vary significantly, indicating that the emission strength tends to be very site specific. Thus, the underestimated particles larger than 100 nm might be a result of missing estimates of non-tailpipe emissions for the underlying site.

4.2 Model sensitivity analysis

4.2.1 Role of aerosol dynamical processes

Along with dilution, aerosol dynamical processes (i.e. condensation/evaporation, coagulation, nucleation, and dry deposition) may interact with one another and modify N and PSD in near-road environments. In this section, the relative importance of the above aerosol dynamical processes is investigated by conducting simulations with individual processes removed and comparing against the base case simulation, in which all dynamical processes are considered by the model. The obtained N and the geometric mean diameter of nucleation mode particles from this sensitivity analysis are summarized in Table 4. The base case simulation demonstrates that in moving from Site B to Site C, the nucleation mode particles decreases by approximately a factor of 3, and the geometric-mean diameter increases by 4.5 nm. The modeled soot mode and accumulation mode particles are excluded from the analysis due to significant underestimations compared to the measurements, as discussed in the previous section. The results of excluding particle dry deposition process are investigated first because its impact on N can interact with particle nucleation and condensation processes, as discussed later in this section.
When particle dry deposition is deactivated in the model, nucleation mode particle numbers increase significantly (~23 and 53% at Site B and C, respectively), resulting in 1–5 nm smaller geometric mean diameters compared to the base case as listed in Table 4. The modeled particle dry deposition velocity is up to 0.2 ms$^{-1}$ for the smallest particles of 3–5 nm in diameter due to strong Brownian diffusion. Our results show that particle dry deposition plays a significant role in governing N in the vicinity of roadways between 30 and 300 m. Gidhagen et al. (2004b) estimated that dry deposition removes only about 12% of total particles near a Swedish highway, in contrast to our estimation of 15–35%. This discrepancy may be due to the different treatment of atmospheric boundary layer turbulence in both studies. Gidhagen et al. (2004b) introduced an artificial source of turbulence into their model to mimic the observed atmospheric dilution of NO$_x$ near the road, while the theoretically-based method by Parente et al. (2011b) combined with the measured ABLT was implemented in this study. The modeled VIT and ABLT have been validated against the measurements in Sect. 4.1.1.

Compared with the base case simulation, the predicted mean diameters at both sites without condensation remain nearly unchanged from the tailpipe emission of 15 nm. At the same time, the predicted particle number concentrations without condensation are about 1 order of magnitude lower than the base case, and are the lowest among all scenarios. The implication of this is two-fold. In agreement with previous modelling studies (i.e. Wang and Zhang, 2012; Uhrner et al., 2011, 2007; Albriet et al., 2010), it strongly suggests that the condensation of SVOCs is responsible for the growth of nucleation mode particles during their atmospheric transport. It also reflects the strong interactions between particle growth and removal processes in governing the simulated particle number. Without the condensational growth of nucleation mode particles, new particles formed due to the BHN mechanism remains in the smallest size bin of 3–5 nm in diameter. Immediately after formation, these particles are subject to efficient removal by dry deposition due to their small particle sizes, resulting in the lowest N among scenarios. This result implies that controlling tailpipe SVOC emissions may indirectly help reduce UFP number concentrations in the vicinity of roadways.
For the scenario without BHN of H$_2$SO$_4$-H$_2$O, the geometric mean diameters at both sites are similar to the base case with slightly more condensational growth in size. However, the particle number concentrations are underestimated by 36 and 10% at Site B and C, respectively, compared to the base case. This implies that over 60 and 90% of the nucleation mode particles at Site B and C are attributed to HDV emissions with non-volatile cores. The result also shows that the BHN of H$_2$SO$_4$-H$_2$O has the greatest impact on the particle population closest to the road. This is because the particles formed through BHN are much smaller in size than those directly emitted with non-volatile cores around 15 nm in diameter. Thus, particles of BHN origins are subject to faster dry deposition removal, and contribute less to N at greater distances in the near–road environment. However, it should not be ignored in air quality modelling studies of mobile emissions, especially within the first 100 m of the roadways.

The scenario excluding particle coagulation alone results in the least impact on both N and geometric mean diameter of nucleation mode particles near the road. The results strongly agree with both timescale analysis (Zhang and Wexler, 2004) and previous CFD modelling studies (Wang and Zhang, 2012; Albriet et al., 2010; Gidhagen et al., 2004b). However, the coagulation process was suggested to be important under mild to weak atmospheric dilution conditions, such as street canyons (Gidhagen et al., 2004a) and road tunnels (Gidhagen et al., 2003).

### 4.2.2 Role of atmospheric boundary layer

Previous studies have shown that accurate CFD simulation of the ABL (including its wind profile and turbulence quantities) is essential for atmospheric dispersion of inert pollutants. For example, Gorle et al. (2009) investigated the effect of atmospheric TKE on the dispersion of particles of 1 µm in diameter, and concluded the impact was significant. In their study, however, aerosol dynamical processes were not considered, nor were their interactions with the ABL. Here, a sensitivity analysis on the ABL is performed to investigate the impact of the ABL on UFP formation and dispersion in the near roadway environment. Specifically, the base case simulation is compared with...
a test simulation where only the wall-function modifications of Eqs. (7) and (8) are not applied.

Figure 6a shows the model predicted UFP concentration gradients of the base case and the test simulations, along with the integration of SMPS data at two fixed locations for the 06:00–08:00 a.m. morning rush hours. The predicted concentrations of UFPs from the test simulation are lower by about $1 \times 10^4$ particles cm$^{-3}$ (or 18 % of the background corrected peak concentration) near the center of the highway compared to the base case. However, the concentration difference for CO$_2$ (as shown in Fig. 6b) between the two simulations is only slightly different (~10 % of its background corrected peak value). The underestimated concentrations of both pollutants are due to unrealistic acceleration of the surface wind and changes in the TKE profile in the upstream region of the computational domain, as discussed in (Blocken et al., 2007b). Thus, the concentration underestimation for CO$_2$ is the result of the overestimated mixing near the surface, where vehicular exhaust occurs. In addition to the overestimated dilution effect on particle dispersion, the impact of the ABL on UFP number concentrations is enhanced by the reduced nucleation rate due to the underestimation of gaseous precursors (i.e. H$_2$SO$_4$ and H$_2$O) in the vehicle wake regions.

The maximum nucleation rate from both simulations is around $2.9 \times 10^{16}$ particles m$^{-3}$ s$^{-1}$, which are in the range of $1 \sim 6.3 \times 10^{16}$ particles m$^{-3}$ s$^{-1}$ from single vehicle exhaust plume simulations (Wang and Zhang, 2012; Uhrner et al., 2007). Although the maximum nucleation rate is not sensitive to ABL profiles, the area averaged nucleation rate of the cross-section in the exhaust pipe plane behind the vehicle is underestimated by a factor of 5 due to the overestimated mixing behind vehicles in the sensitivity run. This comparison strongly suggests that the concentration of UFPs from mobile sources may be even more sensitive to the ABL conditions than inert gaseous species. It also implies that introducing ABL conditions to activity-based emission models (such as Nikolova et al., 2011b) may potentially improve their performance in estimating UFP traffic emissions.
5 Conclusions

In this study, an aerosol dynamics-CFD coupled model is applied to a single unified computational domain to investigate the dynamics and dispersion of UFPs from tailpipe exhaust to the near-road environment. The interactions among individual exhaust plumes are explicitly modeled within the “tailpipe-to-ambient” computational domain. The unique application of translational periodic boundary conditions effectively reduces the size of the computational domain and allows fast multiple-scenario simulations of size- and chemical component-resolved aerosol dynamics. This paper has demonstrated that, together with field measurements, the model is an effective tool which can be used to advance our knowledge on the formation and dispersion of UFPs in the near-road environment. This information is needed to help develop parameterizations of sub-grid processes ultimately to improve air quality model simulations over urban areas.

The model was successfully validated with FEVER field study measurements of both on-road and near-road TKE. The results indicate that the strength of turbulent mixing of pollutants due to VIT and ABLT is properly captured by the model, leading to good agreements between modeled and measured concentration gradients for CO₂ and BC. For UFPs, the modeled PSDs demonstrated adequate agreement with measurements at two fixed locations near a major highway, under different traffic conditions. Sensitivity analysis indicated that the modeled N and PSD of UFPs are sensitive to H₂SO₄-H₂O binary homogeneous nucleation, condensation/evaporation of SVOCs, and particle dry deposition. However, for such an unconfined near-road environment as in this study, coagulation appears to have negligible effect on UFPs. Results also suggest that UFPs from mobile sources may be even more sensitive to ABL conditions than inert species because the average nucleation rate in vehicle wakes is very sensitive to the dilution of H₂SO₄. Therefore, introducing ABL conditions to activity-based emission models may potentially improve their performance in estimating UFP traffic emissions.
The next step of our study is to conduct multiple-scenario simulations and to provide parameterizations on mobile emissions to large scale air quality models. Ultimately, a unified model of UFP dynamics and dispersion can help predict exposures in the vicinity of roadways to support risk assessments and health effect studies. Better treatment of these processes in air quality models is also expected to allow for more accurate predictions of the impacts specific vehicle emission control strategies may have on near-road air quality and subsequent health and environmental benefits.

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L. Huang et al.


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Table 1. Background concentrations of particulate and gaseous species considered in the model, along with ambient relative humidity (RH) and temperature (T).

<table>
<thead>
<tr>
<th>Background PM</th>
<th>FEVER measured value</th>
<th>Model input</th>
</tr>
</thead>
<tbody>
<tr>
<td>BC (µg m⁻³)</td>
<td>0.298–0.53</td>
<td>0.39</td>
</tr>
<tr>
<td>OA (µg m⁻³)</td>
<td>0.676–1.50</td>
<td>1.04</td>
</tr>
<tr>
<td>PM₂.₅ (µg m⁻³)</td>
<td>~ 5.0</td>
<td>4.78</td>
</tr>
<tr>
<td>N (# cm⁻³)</td>
<td>4921–7335</td>
<td>5800</td>
</tr>
<tr>
<td>CO₂ (ppmv)</td>
<td>412.7–421.3</td>
<td>415</td>
</tr>
<tr>
<td>RH</td>
<td>84.1–89.1 %</td>
<td>87 %</td>
</tr>
<tr>
<td>T (K)</td>
<td>283.05–284.55</td>
<td>283.15</td>
</tr>
</tbody>
</table>
Table 2. Tailpipe gaseous and particulate species mass emission rates (g km\(^{-1}\) driven) for gasoline engines. *Canada’s National Inventory Report 1990–2009.

<table>
<thead>
<tr>
<th>Species</th>
<th>Reference emission rates (g km(^{-1}) driven)</th>
<th>Reference</th>
<th>Model input (g km(^{-1}) driven)</th>
</tr>
</thead>
<tbody>
<tr>
<td>CO(_2)</td>
<td>278</td>
<td>Estimated from fuel-based emission factors* and the observed vehicle composition</td>
<td>278</td>
</tr>
<tr>
<td>SVOCs</td>
<td>0.0186</td>
<td>Gasoline powered vehicles (Pye and Seinfeld, 2010)</td>
<td>0.0186</td>
</tr>
<tr>
<td>H(_2)SO(_4)</td>
<td>2.94–8.82 × 10(^{-3}) 0–3.4 × 10(^{-7})</td>
<td>Light-duty diesel vehicles (Uhrner et al., 2007)</td>
<td>6.25 × 10(^{-5})</td>
</tr>
<tr>
<td>H(_2)O</td>
<td>99</td>
<td>Light-duty diesel vehicles (Uhrner et al., 2007)</td>
<td>99</td>
</tr>
<tr>
<td>BC</td>
<td>0.0063</td>
<td>SP2 data (Liggio et al., 2012)</td>
<td>0.0063</td>
</tr>
<tr>
<td>Non-volatile POA</td>
<td>0.0020</td>
<td>MOBILE6.2C</td>
<td>0.0020</td>
</tr>
</tbody>
</table>
Table 3. Particle number-size distribution parameters assumed by the model.

<table>
<thead>
<tr>
<th>Sampling site</th>
<th>Particle mode</th>
<th>Number concentration (# cm(^{-3}))</th>
<th>Geometric mean diameter (nm)</th>
<th>(\sigma)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Background</td>
<td>Soot mode</td>
<td>5000</td>
<td>50</td>
<td>1.6</td>
</tr>
<tr>
<td></td>
<td>Accumulation mode</td>
<td>800</td>
<td>120</td>
<td>1.6</td>
</tr>
<tr>
<td>Tailpipe (raw exhaust)</td>
<td>Nucleation mode</td>
<td>(3.86 \times 10^7)</td>
<td>15</td>
<td>1.4</td>
</tr>
<tr>
<td></td>
<td>Soot mode</td>
<td>(9.42 \times 10^6)</td>
<td>60</td>
<td>1.6</td>
</tr>
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</table>
Table 4. Number concentration and mean diameter of the nucleation mode particles predicted by the model under different scenarios. *Normalized bias (in percentage) was calculated as (base case – scenario)/base case ×100%.

<table>
<thead>
<tr>
<th>Sampling site</th>
<th>Number concentration (# cm⁻³)</th>
<th>Geometric-mean diameter (nm)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Base case</td>
<td>Without deposition</td>
</tr>
<tr>
<td>B</td>
<td>5.94 x 10⁴ (N/A)</td>
<td>7.29 x 10⁴ (-23%)*</td>
</tr>
<tr>
<td>C</td>
<td>1.59 x 10⁴ (N/A)</td>
<td>2.44 x 10⁴ (-53%)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>B</th>
<th>C</th>
</tr>
</thead>
<tbody>
<tr>
<td>Geometric-mean diameter (nm)</td>
<td>19.8</td>
<td>24.3</td>
</tr>
<tr>
<td></td>
<td>18.8</td>
<td>19.6</td>
</tr>
<tr>
<td></td>
<td>15.5</td>
<td>15.2</td>
</tr>
<tr>
<td></td>
<td>20.2</td>
<td>25.4</td>
</tr>
<tr>
<td></td>
<td>19.5</td>
<td>23.3</td>
</tr>
</tbody>
</table>
Table A1. Nomenclature.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$u$</td>
<td>ABL mean velocity</td>
</tr>
<tr>
<td>$k$</td>
<td>Turbulent kinetic energy or TKE</td>
</tr>
<tr>
<td>$\varepsilon$</td>
<td>Dissipation rate of TKE</td>
</tr>
<tr>
<td>$u^*$</td>
<td>Friction velocity</td>
</tr>
<tr>
<td>$\kappa$</td>
<td>Von Karman constant</td>
</tr>
<tr>
<td>$z$</td>
<td>Height above the ground</td>
</tr>
<tr>
<td>$z_0$</td>
<td>Aerodynamic roughness length</td>
</tr>
<tr>
<td>$C_{\mu}$</td>
<td>Constant in the standard $k-\varepsilon$ model</td>
</tr>
<tr>
<td>$\sigma_\varepsilon$</td>
<td>Turbulent Prandtl number for $\varepsilon$</td>
</tr>
<tr>
<td>$E'$</td>
<td>Wall function constant</td>
</tr>
<tr>
<td>$z^+$</td>
<td>Non-dimensional wall distance</td>
</tr>
<tr>
<td>$\nu$</td>
<td>Kinematic viscosity</td>
</tr>
</tbody>
</table>
Fig. 1. Computational domain (a) and running vehicles and ground mesh (b).
Fig. 2. Comparison of the on-road TKE from the passenger vehicle chasing experiments of the FEVER project (black line) and model simulations (red, blue and purple lines). The error bars represent the 25th and 75th percentiles of the measured on-road TKE. PC stands for passenger vehicle and SUV stands for sport utility vehicle. $V$ is the average travelling speed ($\text{m s}^{-1}$) of PC in chasing experiments or the vehicle speed used in model simulation.
Fig. 3. Comparison of the TKE from the FEVER observations at a roadside tower and model simulations for morning rush hours: 05:00–06:00 a.m. (red) and 06:00–08:00 a.m. (blue). Measurement data are plotted in solid lines and model simulation results are plotted in dashed lines. The error bars on top of measurement data represent 25th to 75th percentiles.
Fig. 4. Comparison of modeled and measured near-road gradients of CO$_2$ (ppmv) and BC ($\mu$g/m$^3$) on the downwind side of Hwy-400. Median concentrations from FEVER measurements are plotted in black solid lines, and modeled concentrations are plotted in red dashed lines. The grey areas represent measurements within 25th and 75th percentiles.
Fig. 5. Particle number-size distributions at site B (34 m from the hwy center) and site C (300 m from the hwy center).
Fig. 6. Predicted UFP number (a) and CO$_2$ (b) concentrations as a function of distance to the center of Hwy-400.