

Estimation of PM₁₀ from Exhaust and Non-Exhaust Emission in Traffic Area, Klang Valley, Peninsular Malaysia Using Air Quality Dispersion Modeling

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Abstract: This paper reviews methods to estimate the concentration of PM₁₀ from vehicular emission sources (exhaust and non-exhaust) by using AERMOD dispersion model in the Klang Valley region. The ground level concentration was obtained by processing various meteorological parameters, terrain features and source emission inventory data (emission rate) for 2014 were used in simulations within 50 km x 50 km model domain over 24 hours averaging periods. The results showed the maximum concentrations of PM₁₀ were revealed in central, southeast and southwest of the model domain. The evaluation of performance of the model was done by comparing observed and simulated PM₁₀ concentrations using statistical tools such as correlation coefficient, Normalized Mean Square Error, Factor of two and index of agreement. Therefore, the AERMOD model evaluation results revealed an acceptable model for conducting dispersion modeling from vehicular sources (exhaust and non-exhaust) in the Klang Valley with good model skill for the estimation of PM₁₀ concentrations in Shah Alam station. This study considers the first for evaluation PM₁₀ using AERMOD dispersion model in the Klang Valley region in Peninsular Malaysia.

Key words: PM₁₀ • Emission Inventory • Exhaust and non-exhaust emission • AERMOD dispersion Model

INTRODUCTION

Atmospheric particulate matter is the most important group of environmental pollutants that affect the air at different concentrations. Furthermore, it affects human health when it enters the respiratory system [1]. Present studies on the health problems associated with exposure to particulate matter (PM) with an aerodynamic diameter of 10 µm or less (PM₁₀) have identified a range of health-related problems including deterioration in lung function, heart disease, chronic pulmonary disease and premature death as well as an overall increase in mortality [2, 3].

Traffic emissions which generate air pollutants are either exhaust or non-exhaust emissions. Accumulated road dust is responsible of a major portion of the aerosol mass. In the case of vehicular traffic and non-exhaust

emissions should be tightly controlled. Road dust re-suspension, brake and tire are significant sources of non-exhaust emissions [4, 5]. Emissions from heavy traffic areas have been found to generate and contribute substantially to overall concentrations of PM in urbanized areas and exposure to PM from vehicle emissions has been proved to adversely affect human health [6]. AERMOD is a USEPA recommended, steady-state air quality distribution plume model. It uses the effect of vertical variations in the planetary boundary layer (PBL) to disperse pollutants. The plume growth is established by turbulence profiles that fluctuate with height [7, 8]. This model, which is a steady-state Gaussian plume model is beneficial in calculating pollutant distribution that is particularly effective in rural and urban; areas flat and varied land; surfaces and in situations of high level

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emission ; and from several sources (point, area, line and volume) of emissions [9, 10]. AERMOD has been promoted by the USEPA as a desirable air distribution model to replace the ISCST3 (Lee & Keener, 2008). The AERMOD is considered accurate for dispersion modeling at distances of 50 km or shorter from the emission source [8]. The aims of this study is to present the special concentration maps of PM₁₀ from vehicles exhaust and non-exhaust emission source in the Klang valley region within the model domain 50 km² and to assess the efficiency of the model based on its simulation of PM₁₀.

MATERIALS AND METHODS

Study Area: The Klang Valley is located in central Peninsular Malaysia (between 3.139003 N and 101.686855 E). It is geographically delineated by the Titiwangsa Mountains and surrounded by hilly terrain creating a bowl-like topography with an opening to the Straits of Malacca [11] (Figure 1). The climate is characterised by average temperature (28°C) high humidity above 70%, and with wind speed of 3.1 m/s. Precipitation is lowest during the SW monsoon (6.27mm) and highest during the NE monsoon (15.1mm) [12].

Estimation of the Exhaust Particulate Matter (PM₁₀) Emissions: Total emission from different vehicle types was estimated based on emission factor, traffic volumes and length of road segment. This emission was estimated according to USEPA, AP- 42 and [13] and the formula for emission from traffic used in the study is given below:

$$E_i = \sum_k [EF_{ik} \times A_k]$$

where:

E_i is total emission of PM₁₀/g, EF is Emission factor of PM₁₀ g/km, i is type of pollutants, k is vehicle type, k is Activity level for each pollutant source.

The emission factors (g/km) for pollutant PM₁₀ from exhaust emission were obtained from the literature [14-16]. The Traffic volume of different vehicle types and road segment lengths were based on the road traffic volume survey conducted by the Malaysian Public Works Department [17].

Estimation of Particulate Matter (PM₁₀) from the Non-exhaust Emissions: The non-exhaust particulate matter from traffic emissions (E) was estimated using emission factors (EF) and data from vehicular activity (A). The basic equation is:

$$E_i = EF \times A$$

Estimation of Emission Factor for Re-Suspension (PM₁₀): The particulate matter (PM₁₀) emissions factor was estimated from re-suspension dust on a paved road surface by using US-EPA. AP-42 method:

$$EF_{pr} = K \times (SL)^{0.91} \times (W)^{1.02} \times (1 - P / 4N)$$

Where:

EF_{pr} is Particulate emission factor for a paved road (g/VKT), K is Particle size multiplier (g/VKT; from AP-42, table 13.2.1-1), SL is Road surface silt loading (g/m²),

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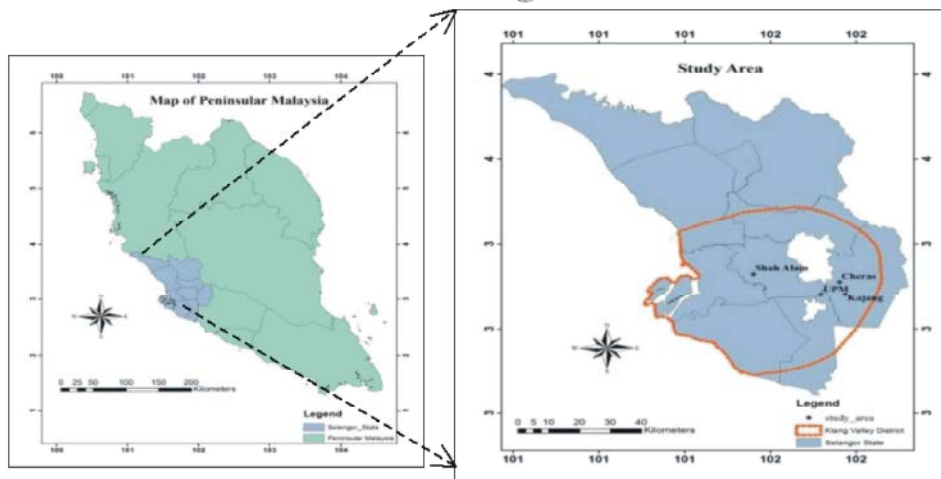


Fig. 1: Study area

W is Average weight of the vehicles using the road (ton), P is Number of wet days with at least 0.254 mm of rainfall during the average period, N is Number of days in the average period (365 for annual). The silt loading on the road surface is a critical factor for the purpose of estimating a paved road which was 0.1 g/m² to estimate the particulate emission factor for a paved road [18, 19].

Estimation of Emission Factor for Brake and Tire Wear PM₁₀: The particulate emissions factor from brake and tire wear was computed employing [20] formula:

$$E_{sj} = N_j \times M_j \times EF_{sj}$$

where:

E_{sj} is Total emission (g) from pollutant for each grid, N_j is Number of vehicle in defined class in each grid, M_j is Average of milege driven (km) per vehicle in each grid, EF_{sj} is Mass emission factor (g/km), s is Non-exhaust emission source (tire and brake wear), j is vehicle category. The emission factor for tire and brake wear PM₁₀ was collected from [20].

AERMOD Meteorological Preprocessor (AERMET) employs meteorological data and surface characteristics to compute boundary layer parameters, namely, mixing height and friction velocity. Surface and upper air were processed every hour in AERMET in the user guide, recommended data from [21] for guidance on the selection of these values. In this study, it is assumed that the land-use is urban area and the weather in Malaysia is similar to condition in summer. Therefore, albedo, Bowen ratio and Surface Roughness values were (0.16, 2 and 1). AERMOD Terrain Preprocessor (AERMAP) used gridded terrain data Digital Elevation Model (DEM) to compute a representative terrain-influenced height in association with each receptor location.

Model Validation: Data for observed ambient air concentrations of heavy metals in particulate matter (PM₁₀) were collected for the year 2014 for the average daily concentrations from the four stations of traffic areas in the Klang Valley region. This monitored daily average PM₁₀ data were compared with the models' estimated concentrations. Comparison was carried out by measuring some statistical parameters which had been used in earlier studies related to model validation [22-24] which is given by the equation below:

$$R = \frac{\sum_1^N (C^S - \bar{C}^S)(C^O - \bar{C}^O)}{\left(\sum_1^N (C^S - \bar{C}^S)^2 \sum_1^N (C^O - \bar{C}^O)^2 \right)^{0.5}}$$

where:

R is Correlation

$$NMSE = \frac{(C^S - C^O)^2}{C^S C^O}$$

NMSE is normalized mean square error

$$FA2 = \left(0.5 \leq \left(\frac{C^S}{C^O} \right) \leq 2.0 \right)$$

Fa2 is Factor of two is defined as a fraction within a factor of 2

$$IA = 1 - \frac{\sum_1^N (C^S - C^O)^2}{\sum_1^N (|C^S - \bar{C}^O| + |C^O - \bar{C}^S|)^2}$$

IA is Index of Agreement represents agreement between the average simulated and observed value.

RESULTS AND DISCUSSIONS

Estimation Emission Inventory: The estimation of PM₁₀ emission inventory for traffic emissions (exhaust and non-exhaust) was based on the traffic information such as traffic volume, road length and emission factor for PM₁₀ of each road in each grid of 1km×1 Km within model domain Table 1.

Table 1: PM₁₀ emission from exhaust and non-exhaust (resuspended dust, brake wear and tire wear) according to 1km × 1km grid

Source	Emission g/s
Exhaust	0.000203233
Resuspended dust	0.000114034
Brake wear	7.94557E-05
Tire wear	6.94228E-05

Meteorological Observations: Figure 2 showed wind roses were generated using the WRPLOT and wind classes frequency distribution feature within the AERMET model, which indicated the wind roses of the prevailing winds noted in the study area during the study period showing that the strongest intensities were northeasterly (NE) and west northerly (WN) and the proportion of calm winds was only 16.87 % during the study period.

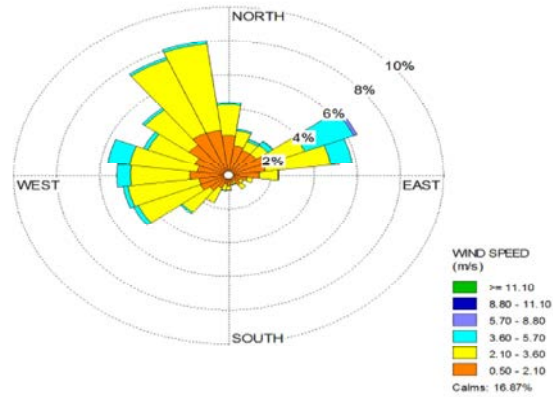


Fig. 2: Wind rose diagram of study area during the study period (blowing from)

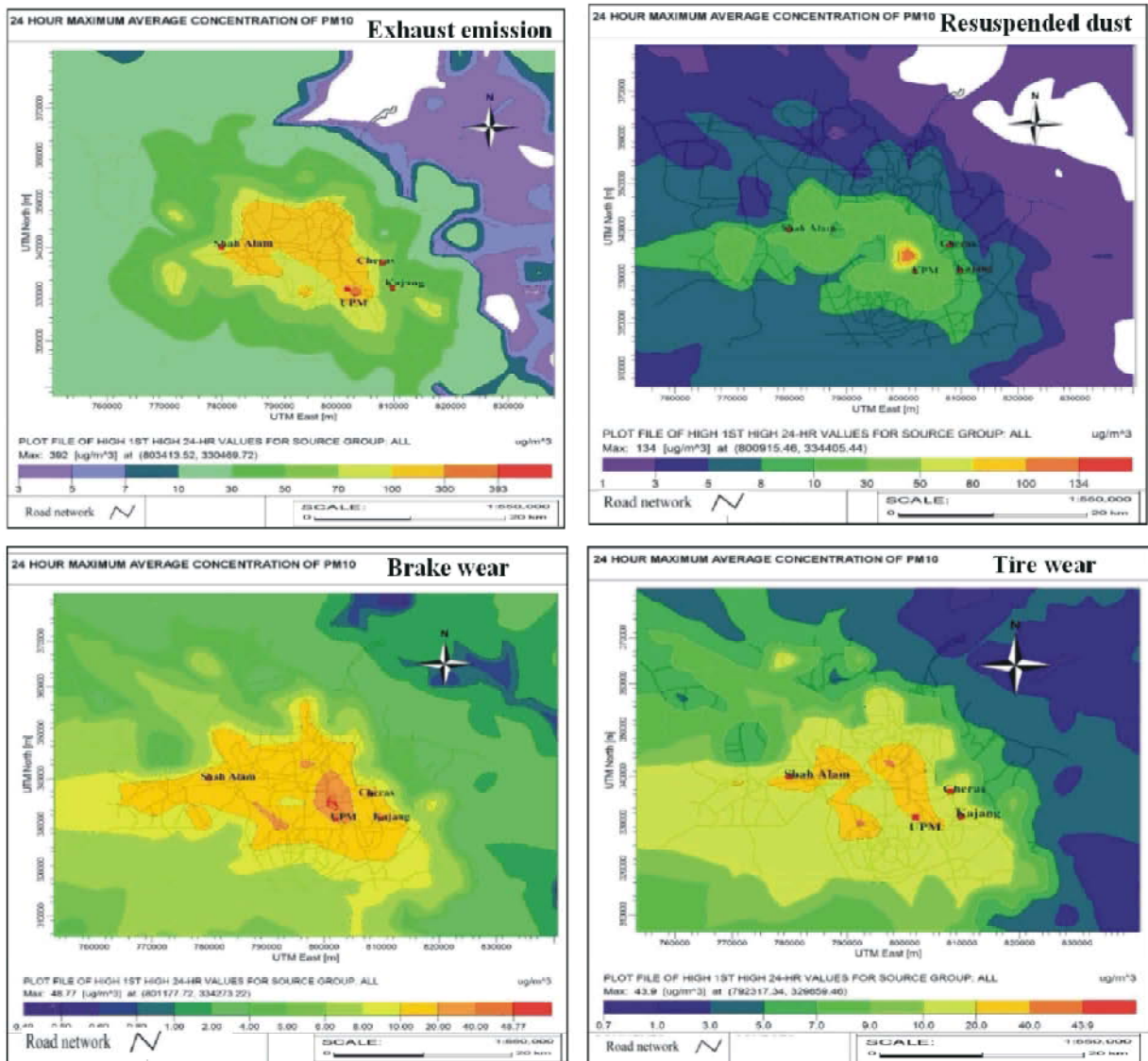


Fig. 3: Spatial distribution of PM_{10} concentration from exhaust emission (year 2014)

Table 2: AERMOD model performance statistics for PM₁₀ simulation

Station	R ²	NMSE	F2	IA
Cheras	0.53	0.85	1.01	0.11
Shah Alam	0.64	0.61	1.02	0.54
Kajang	0.7	1.52	1.3	0.2
UPM	0.54	2.25	1.27	0.23

AERMOD Model Simulation of PM₁₀ from Exhaust and Non-exhaust Emission: The AERMOD model used the hourly meteorological data, emission rate, terrain characteristics and receptor data for the prediction of PM₁₀ concentration. The results of AERMOD simulation presented in two parts, spatial concentration maps and statistical analysis for PM₁₀ in domain 50 × 50 km in Klang Valley region 2014. The spatial concentration maps of PM₁₀ for exhaust and non-exhaust emission were presented in (Figure 3). The spatial concentration maps of PM₁₀ for exhaust indicated that the highest PM₁₀ concentrations were 393 µg/m³ located at centre of the domain heading to UPM and Kajang stations, while the second average concentration ranged between 100-300 µg/m³ which covered all the stations. Generally, most of the domain areas were located at less concentration zones where the amount of PM₁₀ is between 3 to 30 µg/m³. The PM₁₀ was dispersed to the west of domain, possibly due to the Northeast monsoon wind which disperses this pollutant to the west. However, the maximum PM₁₀ concentrations that emit from non-exhaust namely resuspended dust, brake and tire wear were 134, 48.7 and 43.9 µg/m³ respectively which were found directly to the central, southeast and southwest of the domain close to highways of Cheras, Kajang and UPM, likely due to the heaviest traffic emissions from transportation that contribute largely in the center [26] (Figure 5, 6 and 7). Based on the 24 hours simulation results, the concentration of simulated PM₁₀ from exhaust emission in Klang Valley region within the domain was higher than the concentration guideline recommended by the DOE Malaysia which is 150 µg/m³ for 24 hours. Whereas, the concentration of PM₁₀ that release from non-exhaust was well below than this limits. Singh and Perwez [27] reported the predicted concentration level of SPM varied from 1000 to 2000 µg/m³ at all the receptor.

Table 2 demonstrates the results of AERMOD simulation at Cheras, Shah Alam, Kajang and UPM stations for 24-hour PM₁₀ concentrations. The correlation coefficients for the simulated versus observed 24-hour PM₁₀ concentrations were high and moderate.

The statistics also revealed NMSE of 0.85 and 0.6, suggesting approximately close to ideal value for Cheras and Shah Alam stations respectively (NMSE=1), while 24-hour PM₁₀ concentration declined to within a factor of two (FA2). The index of agreement showed good agreement between observed and simulated values for Shah Alam. These results in line of study carried out by Gulia, Shrivastava, Nema and Khare [28] evaluated the quality of air in the vicinity of a heritage site in Amritsar city in India employing AERMOD. The AERMOD performance was estimated for the forecast of PM₁₀ and the results exhibited that predicted pollutant levels were within acceptable limits and the estimated IA values for PM₁₀ was 0.50, which showed the performance of AERMOD was acceptable.

CONCLUSION

AERMOD model was used to predict the meteorological conditions, Traffic emissions and air pollution dispersions in the Klang Valley region. The spatial distributions revealed the maximum concentrations of PM₁₀ observed in central, west and southeast of the domain heading to Cheras, UPM, Kajang and Shah Alam stations, which are suitable for studying the high concentration of contributions by the prevailing wind direction during the Northeast monsoon season. Evaluation of AERMOD model showed there was good agreement between the simulated and observed concentrations of PM₁₀ in Shah Alam.

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