

MATHEMATICAL MODELING CONSIDERING AIR POLLUTION OF TRANSPORTATION: AN URBAN ENVIRONMENTAL PLANNING, CASE STUDY IN PETALING JAYA, MALAYSIA

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Abstract

This paper provides the findings on a project undertaken to develop a geo-spatial mathematical model relating landuse, road type and air quality. The model shows how spatial elements and issues were quantified to accurately represent the usual and unusual urban environment in the development of residential land-use. The mathematical relationship was based on the optimum distance between residential area and urban transportation network. This mathematical analysis would provide a better planning for urban transportation. The spatial data (urban land-use and urban network development) were generated using satellite images, aerial photos and land use maps. Geospatial analyses were performed to find the effect and impact of urban air quality with respect to urban transportation networks. The output of the study would assist the task to reduce negative transport environmental impacts particularly in the field of air pollution. It would also be useful in identifying the potential residential area with respect to urban transportation network towards achieving sustainable development.

Keywords: Transportation, Model, Air pollution, urban environment, land use

1. Introduction

Air quality, regarded a main infrastructure element in urban transportation system, is considered as a major criterion for human settlement. Therefore transport emitted air pollution appears related to the establishment of urban landuse in proportion to urban transportation network. Increasing demand for

residential areas, along with this development of cities, has given rise to some environmental issues (Bell and Blake, 2000; Ranjan, 2001). More than half of the world's population lives in urban areas (Colesca S., 2009), therefore increasing urban land-uses resulted in several impacts on various fields such as air quality, accessibility and land use. Air quality has been considered as one of the major environmental elements by many urban planners (Colville et al. 2000; Nicolas et al. 2005). Therefore, there is a need to look into urban transportation planning together with landuse development.

With respect to above note, paper presents development of a mathematical model to find suitable locations for landuses and urban transportation network for the urban transportation system of Petaling Jaya Municipal Council (MPPJ). The model essentially investigated the best air quality for residential area. It was based on the optimum distance from residential zones to urban networks as main element avoiding transport emitted air pollution.

Petaling Jaya Municipal Council (MPPJ), a developing city in Malaysia, was chosen as study area, Its based on choices of environmental factors such as: several routes to access to important public facilities, High transportation activities, Efforts to maintain its "garden city" concept, Efforts to establish well-developed infrastructures and excellent investment opportunities.

2. RESEARCH APPROACH

The works undertaken includes several statistical and mathematical analysis of urban planning, focusing on land use and air quality using Geographic Information System (GIS) as visualization platform. Distance of residential land use from urban transportation network is modeled based on determining factors of Carbon Monoxide (CO) emission from vehicles movement. This emission is useful for modeling suitable locations of residential land use in urban areas. The determination of factors is based on their analyzing individual role in CO emission. Relevant factors such as plume rise of CO, average atmospheric temperature and pressure, stack exiting velocity, estimated stack diameter, average wind speed and traffic volume are determined based on historical data (Highway Planning Unit, 2004) and standard definition of road types. Mathematical analysis methods, used to determine quantitative role of factors in the model, include programming by software of Excel and power function analyzing.

The air quality expected with respect to CO distribution is modeled using the basic theory of Primary Standard of CO Pollutant in 8 hours averaging time (WHO, 2000), which shows that the threshold of air pollution for human health in Malaysia is considered equal to 10 mg/m^3 .

The model development method is based on the maximum ground level concentration of CO, considering the spread of a plume in vertical and horizontal directions, which is assumed to occur by simple diffusion along the direction of the mean wind. CO like other pollutants accumulates based on stability time, independently or in conjunction with atmospheric temperature and pressure, wind speed, curb length, or area. Curb length of CO, as indicator of suitable distance to locate residential zones, can be a function of plume rise of CO, average atmospheric temperature and pressure, stack exiting velocity, estimated stack diameter, average wind speed and traffic volume. This function can be restructured for the study area (MPPJ) to plan future urban development and improvement.

2.1 MATHEMATICAL MODELING TRAJECTORY

The base of mathematical trajectory is combination of several relative formulas and models to calibrate a new model in relation to statement problem. The basic steps of the mathematical trajectory are summarized in this section. Assuming that the optimum Euclidean distance (D_{min}) from the residential area to the urban transportation network is defined based on good air quality, the element to be constructed in the model is listed as indicator of good air quality for residential area, the role of transportation in production of air pollution and acceptable distance between roads and residential area to meet good air quality. The main portion of the modeling is to determine main variables of these elements.

2.1.1. ELEMENT DESCRIPTION

There are some elements which are emitted by vehicle transportation. But the statistics of emissions of air pollution elements attributed to transportation shows main contribution (70-90% of total emission rate) of transportation producing CO (Clean Water Action Council, 2008; Department of Transport, 1996; Haughton et al. 2003; Meszaros et al. 2005; Rodrigue et al. 2006). Hence in this research CO has been considered as indicator of transportation air pollution. According to this consideration concentration level of CO for human health (10 mg/m^3 for 8 hours) is known as indicator of good air quality for residential area (WHO, 2000). The role of transportation in production of with respect to consideration of CO, air pollution can be quantified as CO emission rate for total vehicles. it is calculated by CO emission rate of one vehicle and number of vehicles passed during specific time and road length, called traffic volume (Vos, 2002)

With respect to above explanation, the minimum safe distance between roads and residential area to avoid emitted air pollution of vehicles (D_{min}) is the distance in which total CO emission rate of vehicles is reduced to 10 mg/m^3 (acceptable concentration level of CO, for human health). This safe distance

depends on a plume in vertical and horizontal directions is assumed to occur by simple diffusion along the direction of the mean wind, total CO emission rate of vehicles as expressed in Equation (1), developed by Turner (1995).

$$C_x = \frac{Q}{\pi \sigma_y \sigma_z U} e^{-1/2 \left(\frac{H}{\sigma_z} \right)^2} e^{-1/2 \left(\frac{Y}{\sigma_y} \right)^2} \quad (1)$$

Hence for calculating of this distance, it is required to apply total CO emission rate of vehicles, rise distance of emitted CO, mean wind speed and standard deviation of vertical and horizontal wind direction.

2.1.2. QUANTIFICATION OF ELEMENTS

CO pollution is very sensitive and traffic volume changes over times are considerably unpredictable. Therefore, for calculating total CO emission rate of vehicles, it is better to consider road capacity (maximum possible traffic volume) replacing traffic volume, as expressed in Equation (2)

$$Q = (RC)q_c \quad (2)$$

Where

Q = Total CO emission rate of vehicles

RC = Possible road capacity in stability time of CO

qc = Possible average CO emission rate for one vehicle

A simple Equation to calculate road capacity is developed by Li (1998) as follow:

$$RC = \left(\frac{(WL_p) / V_t}{C_m} \right) \quad (3)$$

Where,

W= Road width based on road type (m),

Lp= Passed road length by vehicle,

C_m = one vehicle's normal Average space time usage (m^2) and

V_t = average Vehicle speed based on road type (m/s) = Average passed length by car per second (m).

Since, the concentration of pollutant reaches a peak value within 5 minutes of the gas injection, the maximum reasonable specific time for calculating the number of passed car is considered as less than 15 minutes (Colorado Department of Public Health and Environment, 2006)

C_m is calculated for the study area, based on percentage of vehicle types (according to historical data of traffic volume) and their actual size. W and V_t were applied based on road types (table1). L_p (passed road length by vehicle) refers to selected greeed size to investigate and study of air pollution. And q_c was obtained trough mathematical process, considering some important elements like percentage of vehicle types, average fuel consumption of foreign and domestic employing vehicles, average normal age of employing vehicles, percentage of different ages of employing vehicles.

For computing rise distance of emitted CO by vehicle transportation (Δh), Equations (4) and (5) developed by Wayson (2000) were used as follows:

$$\Delta h = 1.6 \left(\frac{F_0 t^2}{U} \right)^{1/3} \quad (4)$$

$$F_0 = g v_s r_s^2 \left[1 - \left(\frac{T_a}{T_s} \right) \right] \quad (5)$$

Where,

Δh = Rise distance (m),

F_0 = Buoyancy factor (m^4/s^3),

t = Time (s),

U = Ambient horizontal wind speed (m/s),

g = Gravitational constant = 9.81 m/s^2 ,

v_s = Exit velocity (m/s),

r_s = Exit radius (m),

T_a = Ambient temperature (K) and

T_s = Exit temperature (K).

The parameters and also mean wind speed could easily be obtained through annual reports of the study area (DOE, 2004) actual measurement with respect to various vehicle types, standards and guidelines.

Horizontal and vertical dispersion (σ_y and σ_z) are determined from the graphs found in the Figure1 and in the attributes written in Table 1. In these figures vertical and horizontal dispersion coefficients for

different areas have been categorized in 6 weather stability classes (A, B, C for day and D, E, F for night). It is based on 3 atmospheric factors of wind speed, incoming solar radiation and thinly overcast.

2.1.3 FINALIZATION OF MODEL

By applying amount of $C_x = 0.01\text{g/m}^3$ (primary standard of CO pollutant in 8 hours), Equation (1) can be rewritten as follow:

$$\sigma_y (\sigma_z)^3 = \left(\frac{19.3Q\Delta h^2}{U} \right) \quad (6)$$

Where is D_{\min} ? In Equation (7), σ_y and σ_z were replaced by a mathematical function of D_{\min} .

$$F(D_{\min}) = \sigma_y (\sigma_z)^3 \quad (7)$$

Therefore, based on Equation (6)

$$F(D_{\min}) = \left(\frac{19.3Q\Delta h^2}{U} \right) \quad (8)$$

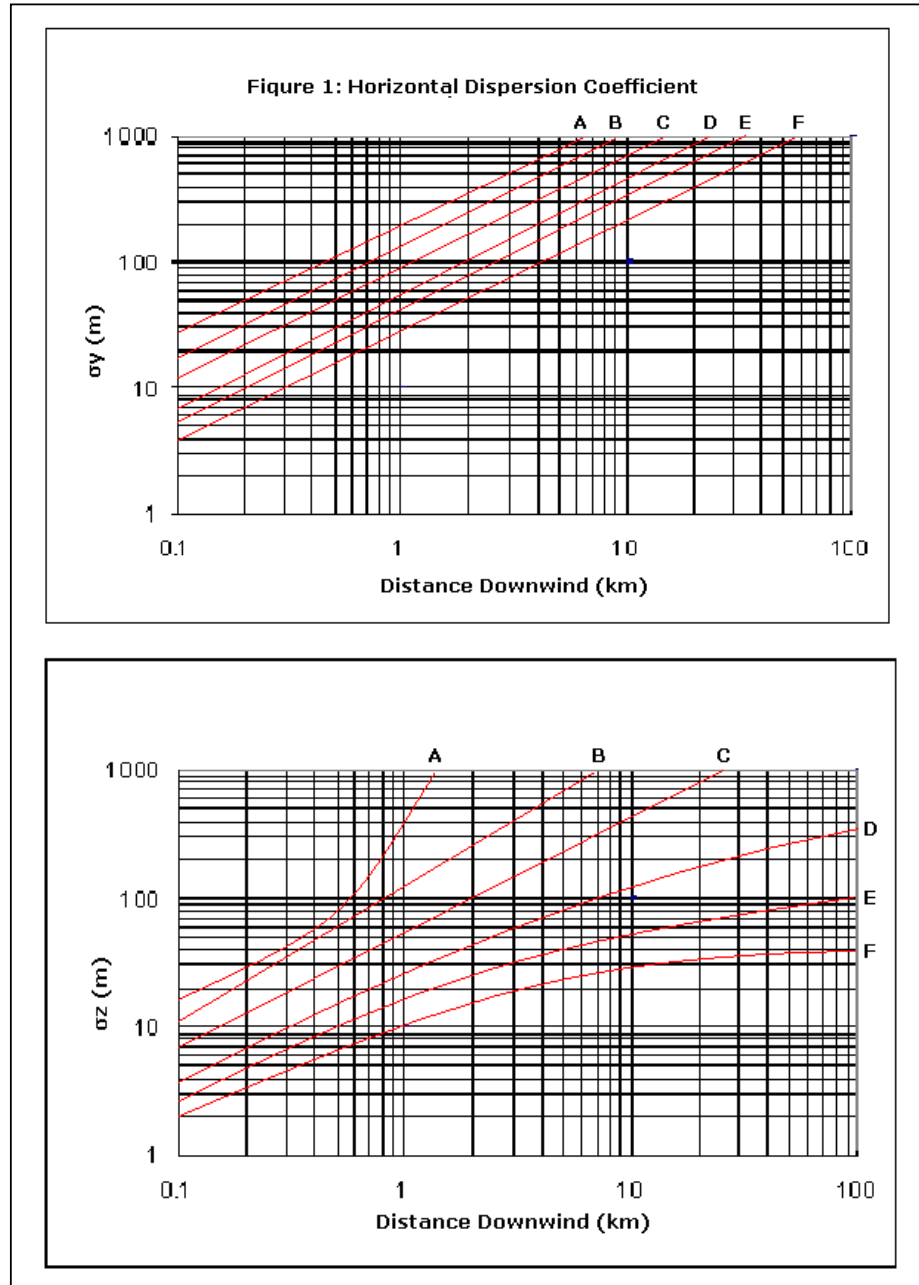
So, it was necessary to develop $F(D_{\min})$. Since 85-88% of vehicle transportation was done during the day (FHWA, 2000), thin overcast as weather indicator for night (Table 1) was ignored.

TABLE 1 - KEY TO WEATHER STABILITY CLASSES

Wind Speed (m/s)	Day			Night	
	Incoming Solar Radiation			Thinly Overcast	
	Strong	Moderate	Slight	More than 50 % Cloud	Less than 50 % Cloud
Less than 2	A	A-B	B	E	F
2-3	A-B	B	C	E	F
3-5	B	B-C	C	D	E
5-6	C	C-D	D	D	D
More than 6	C	D	D	D	D

Source: Turner, 1995

Solar radiation and average wind speed in the study area played main role in this process. Based on historical data and Table 1, weather conditions in the study area were more close to situation A in Figure 1.



The graph (A) was used to develop F (Dmin). Crowl and Louvar (2002) developed following equations for graph A:

$$\sigma_z = 0.24D(1+0.0001D)^{-1/2} \quad (9)$$

$$\sigma_y = 0.32D (1+0.0004D)^{1/2} \quad (10)$$

Since the numerical values of 0.0004 Dmin and 0.0001 Dmin which given in Equations (9) and (10) are very small, they can be eliminated and these equations can be rewritten as:

$$\sigma_y = 0.32(D_{\min}) \quad (11)$$

$$\sigma_z = 0.24(D_{\min}) \quad (12)$$

Based on Equation (7)

$$F(D_{\min}) = 0.32(D_{\min}) [0.24(D_{\min})]^3 \quad (13)$$

Then:

$$F(D_{\min}) = 4.42 \times 10^{-3} (D_{\min})^4 \quad (14)$$

And then based on Equations (8) and (14):

$$\left(\frac{19.3\Delta h^2 Q}{U} \right) = 4.42 \times 10^{-3} (D_{\min})^4 \quad (15)$$

$$D_{\min} = 8.13\Delta h^{1/2} \left(\frac{Q}{U} \right)^{1/4} \quad (16)$$

Generally final model can be rewritten as:

$$D_{\min} = k \frac{\Delta h^{1/2} Q^{1/4}}{U^{1/4}} \quad (17)$$

2.2 SENSITIVITY ANALYSIS

During process of sensitivity analysis, the effects of three main variables (Δh , Q and U) on calculated and field Dmin were compared to test accuracy and ability of the model. It was done to determine which

of these three variables are critical and has great effect on the model outputs. The results (Figure 2) show non-acceptable level of accuracy in most of samples.

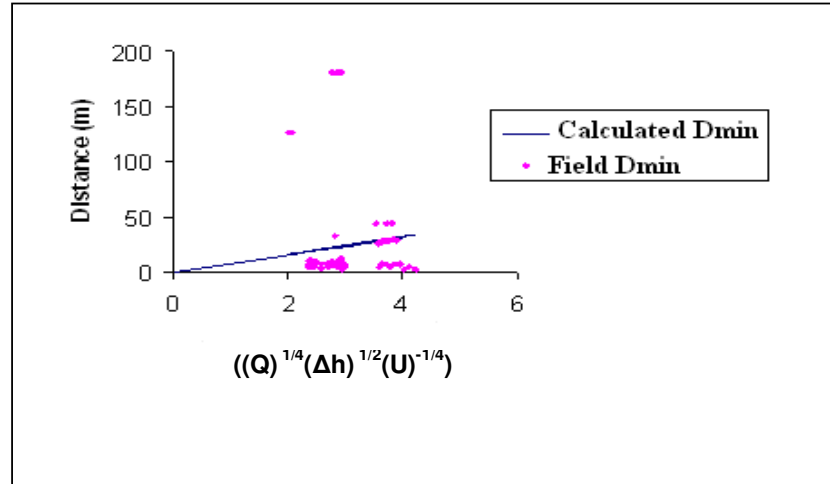


FIGURE 2 - FIELD AND CALCULATED D_{MIN} BY MODEL

Applying and analyzing errors of the samples show that there isn't any reasonable relationship between errors and rise distance and wind speed, While Figure 3 shows more concentration of errors in higher values of Q. It means the model is more sensitive to Q and its parameters.

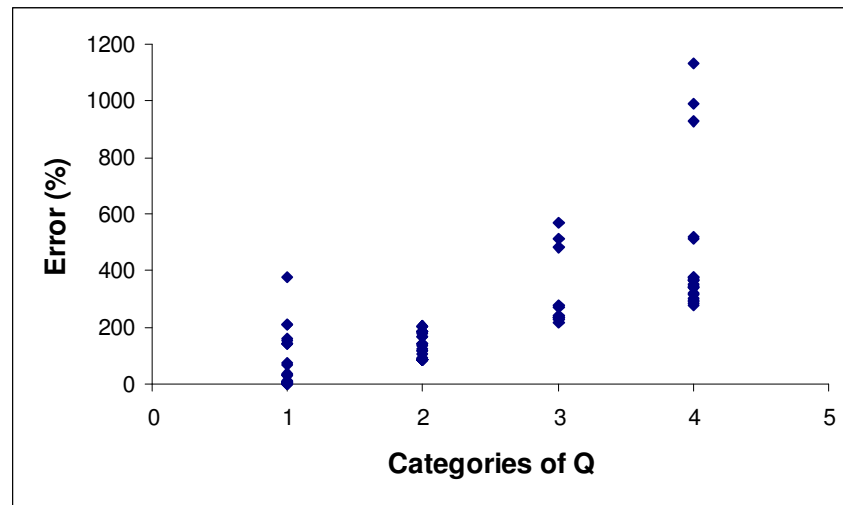


FIGURE 3 - RELATIONSHIP BETWEEN TOTAL VEHICLES EMISSION RATE AND ERRORS OF FIELD D_{MIN}

Two parameters of Q are flexible and controllable, that can be changed to achieve more accuracy. These are grid size of transport air pollution (passed road length by vehicle), and stability time period of CO to record transportation data. In Figure 2 these two elements have been considered equal to 100 minutes (Figure 4).

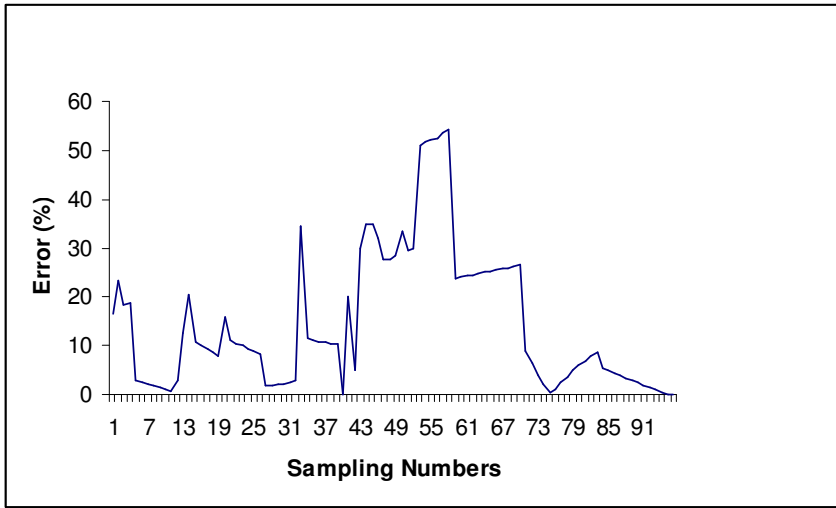


FIGURE 4 - MATHEMATICAL ERRORS OF FIELD D_{MIN} AFTER CALIBRATION OF STABILITY TIME OF CO AND GRID SIZE

But there are still can not be ignored. Analyzing mentioned samples shows all the samplings have been done in far distance (more than 30m distance). The result is shown in Figure 5.

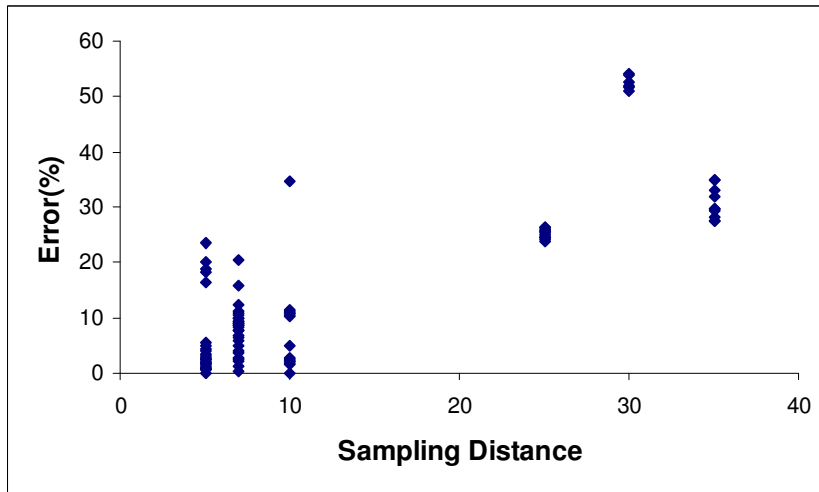


FIGURE 5 - RELATIONSHIP BETWEEN SAMPLING DISTANCES AND ERRORS OF FIELD D_{MIN}

Therefore the results shown in Figure5 must affect main parameters (Q, U and Δh). With respect to the nature of 3 main parameters of the model, this error depends on Q, because there is no relationship between Δh and sampling distance (D). Also the lengths of sampling distance are very small and can not change wind speed. With respect to unobservable sources of CO, Samplings be done in far distances from roads. There are some unobservable sources of CO along sampling distances such as smoking, burning, and boiling that can have interactions with Q. Normally in sampling distances of greater lengths larger numbers of these sources can be found. In this model samplings in far distances are invalid and sampling should be done in distance of less 30m. Based on above results, of calibration validation and testing of the model were done.

2.3 MODEL CALIBRATION

Comparing Equations (16) and (17), the value of k can be taken as 8.13. But it must be calibrated for the study area, using field data. Therefore, the model calibration should include comparisons between model-calculated conditions and field conditions, using available data on wind speed (U), total vehicle emission rate for CO (Q), and rise distance of CO (Δh). These four parameters (Δh, Q, U, and calculated D_{min}) can be computed by using applied primary data in the fieldwork such as samples are shown in Table2.

Table 2: Some Samples of Calculated Parameters by Fieldwork Data

U	Δh	Q	Calculated D _{min}	(Δh) ^{1/2} (Q) ^{1/4} (U) ^{-1/4}	Field D _{min}	Justified K
10.53	0.15	1717.65	11.07	1.38	9.83	7.12
10.80	0.15	1717.65	11.11	1.39	8.72	6.27
11.08	0.15	1717.65	10.98	1.37	12.62	9.21

Field D_{min} also is interpolated using amount of Q and D (distance of sampling). In Table 2 amount of justified k is computed by Equation (18).

$$K_j = \frac{FieldD_{min}}{(U^{-1/4} \Delta h^{1/2} Q^{1/4})} \quad (18)$$

Where

Δh = Rise distance of CO,

Q = Total vehicle emission rate for CO,

U = Wind speed, and

Kj = Justified constant value for every sample.

After calculating Kj for all the samples, justified constant value for final model is calculated by averaging total amounts of Kj. The results show average total amount of Kj is equal to 8.68. Hence the model for study area can be written as follow:

$$D_{\min} = 8.68 \frac{\Delta h^{1/2} Q^{1/4}}{U^{1/4}} \quad (19)$$

Where,

Δh = Rise distance of CO,

Q = Total vehicle emission rate for CO,

U = Wind speed, and

Kj = Justified constant value for every sample.

2.4. MODEL VALIDATION

The model validation is done by comparing its predictions with field data. With respect to basic knowledge of sensitivity analysis and model calibration, the validation is done by applying values of parameters of calculated D_{min} (after calibration) and field D_{min} and analyzing mathematical relationship of these two parameters. This process was done for two series of observations, which were randomly selected from field data. Some samples are shown Table 3. Fortunately this mathematical relationship shows acceptable level of accuracy, 87.89% (Figure 6) and 87.10% (Figure 7).

TABLE 3: SOME SAMPLES OF COMPARING FIELD D_{min} AND CALCULATED D_{min} AFTER CALIBRATION

Sample Numbers	Calculated D _{min}	Field D _{min}	Error (%)
1	16.27	14.38	13.15
2	15.44	15.21	1.57
3	15.86	17.98	11.76

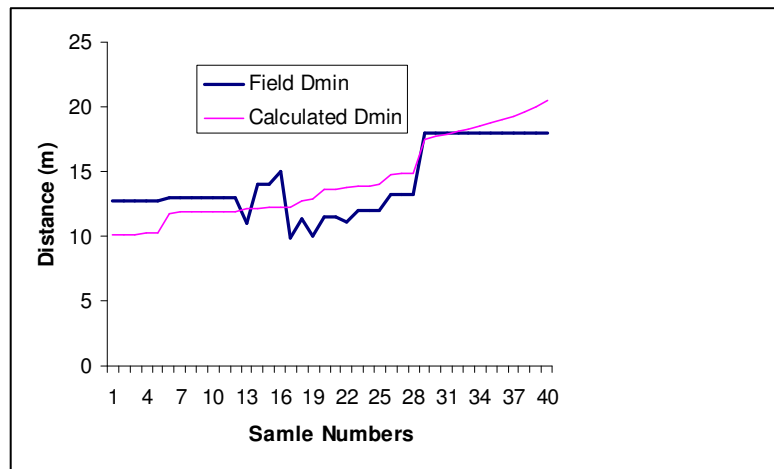


FIGURE 6 - MATHEMATICAL ERROR BETWEEN FIELD AND CALCULATED D_{MIN} BY MODEL (SERI 1)

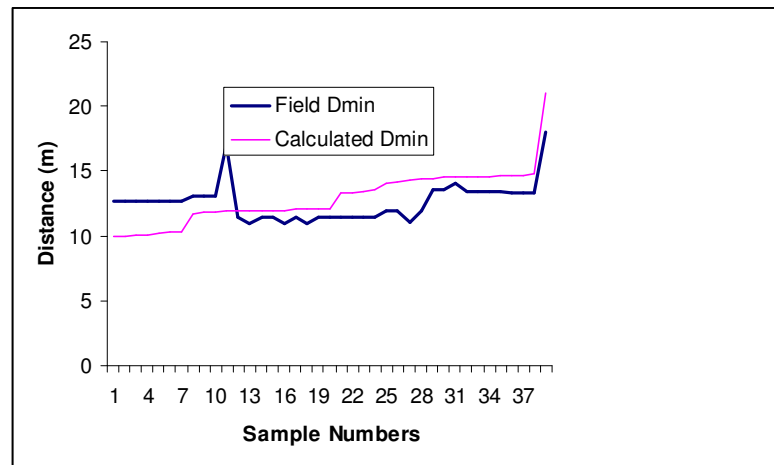


FIGURE 7 - MATHEMATICAL ERROR BETWEEN FIELD AND CALCULATED D_{MIN} BY MODEL (SERI 2)

2.5. MODEL TESTING

In the analysis of data, field and calculated values were compared to verify model for a range of idealized and real condition with respect of model purposed. For example, if the calculated minimum safe distance for one station was more than the actual distance, the sensors monitored some pollution in this station. Also, if the calculated distance was less than the field distance, there should not be any pollution. Four possible situations in the mentioned process are shown in Figure 8. Some samples of applied data in fieldwork are shown in Table 4.

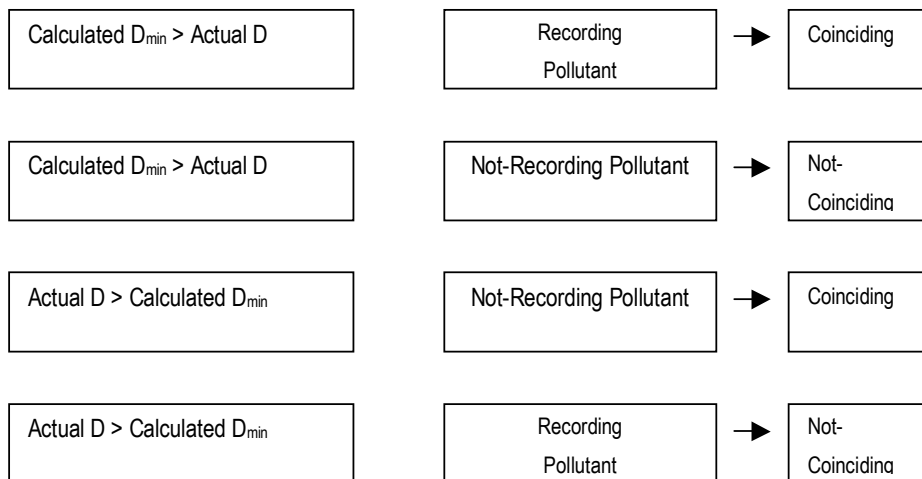


FIGURE 8 - POSSIBLE SITUATIONS FOR TESTING OF FIELD DATA

TABLE 4 - SOME SAMPLES OF TESTING OF APPLIED DATA IN FIELDWORK

Road Type	TV/h	U	Ts	Ta	D _{min}	D	CO	Coinciding
Local	279.75	14.14	395	301	7.84	25	1	c
High way	1441.5	9.42	395	301	11.54	5	13	c
Arterial	1045.25	8.58	395	301	11.10	7	7	n
Local	279.75	12.75	395	301	7.98	25	0	c

Note: c = coincided, n= not coincided

Parameters of this table are road capacity per hour (TV/h), wind speed (U), exhaust gas temperature (Ts), atmospheric temperature (Ta), calculated minimum distance by model (D_{min}), and actual selected distance for survey (D). The results, which have been done, show 89.9% of the cases were coincident in the observations confirming ability and usefulness of the model (Figure 9).

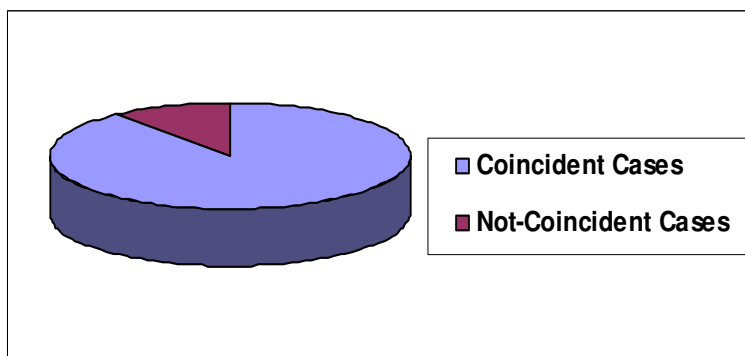


FIGURE 9 – GRAPHICAL RESULTS

3. CONTRIBUTION OF MODEL FOR THE STUDY AREA

The model when applied with the generalized values for the components of transportation system and weather condition had found to be fit for the study area. The main constants used are:

- a) T_s (General average exiting gas temperature from exhaust) = 395° k,
- b) V_s (General average CO exiting velocity from exhaust) = 0.4 m/sec,
- c) T_a (Average annual atmospheric temperature for MPPJ) = 301° k,
- d) P (Average annual atmospheric pressure for MPPJ) = 1000 and
- e) C_m (Average single vehicle's normal space time usage) = 7.3 m²

After applying these generalized values in the model, the D_{min} were calculated for road types in MPPJ, the results are shown in Table 5.

In this research, determination of residential landuse was done with determination of polluted areas of urban transportation. Areas with good air quality were obtained through overlaying analysis and deselecting of effective polluted zones by road types. Based on the standard road types (FHWA, 2000; PBD, 1998), potential polluted zones were sub-divided into five zones: potential polluted zones by collector roads, local roads, arterial roads, sub-arterial roads and highways. Applying the appropriate D_{min} values indicated in Table 5, five polluted areas have been zoned and presented in maps of varying road types.

TABLE 5 - CALCULATED D_{min} FOR ROAD TYPES OF MPPJ

Road type	D_{min}
Local	18.32
Collector	17.14
Sub-arterial	16.35
Arterial	15.43
Highway	12.77

For instance Figure 10 depicts the polluted zones for collector road. Finally, the potential pollution zone map of MPPJ (Figure 11) was generated by overlaying all maps of potential polluted zones by road types. This map shows that 20.07 of MPPJ are potentially polluted by transportation, which is considerably high by MPPJ management standard. This map is useful in delineating non-polluted zones. Figure 12, depicts a map resulting from combining map of residential area with potential pollution zone map for analyzing current residential landuse. It essentially shows 29.37% of current residential areas are located in area potential polluted by urban transportation.

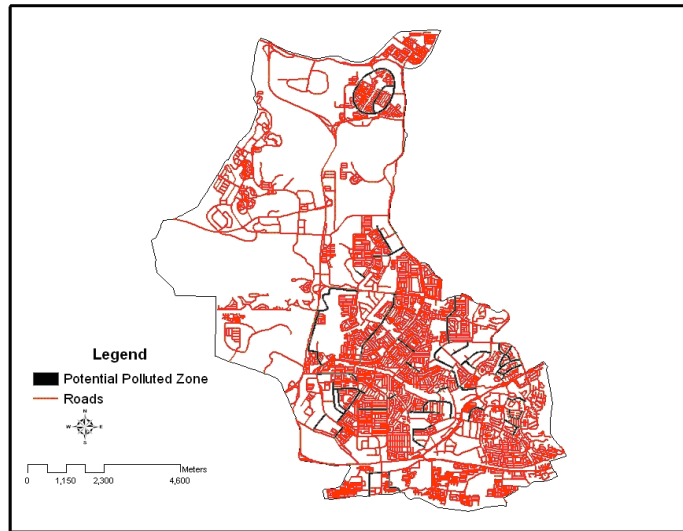


FIGURE 10 - POTENTIAL POLLUTED AREA ALONG COLLECTOR ROADS

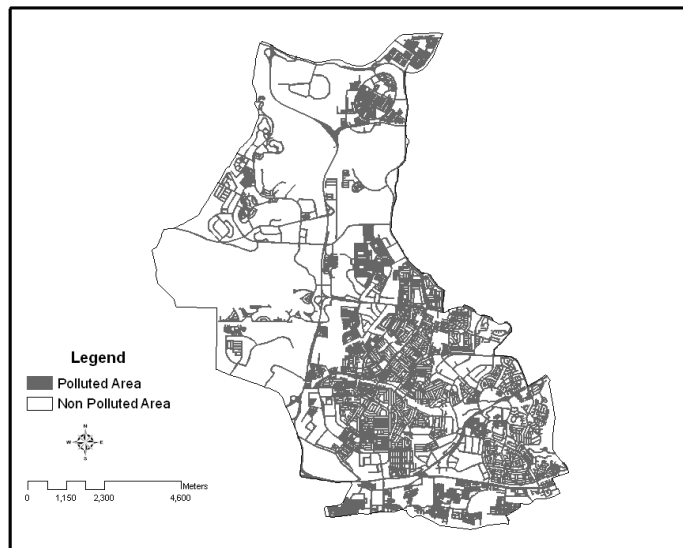


FIGURE 11 - POTENTIAL POLLUTED AREA BY ROADS TYPES

4. DISCUSSION

The mathematical model developed in this study comprised most of the criteria identified in the introduction of paper. Individual formulas and models of each criterion may not meet research objectives individually, but their preferences were incorporated in the final model. The structure of the model developed enables interactive modification by changing the weights and properties of parameters. Compared to other models, this model has some additional features that can support planners in decision making. The objectives of this model were elaborated in more detailed parameters

and parameters indicators were determined. Indicators are driven data to satisfy the purpose of the model. This detailed classification strengthens analysis of any situation and the result is normally more reliable. In this model, constraints were also considered for instance planners should avoid selecting a site which was defined as effective polluted area by transportation. This model also is used more friendly and easier than other models; it structured in simple mathematical formula and does not need to train any other skim.

It is expected that the urban planners who are studying the spatial distribution of transportation network and landuse will utilize the model. A user must be aware of the limitations of the model. Moreover, the result will depend on the quality of data that are used in this model. It is important to note that this model supports the planners by providing alternative sites in urban development plans.

5. CONCLUSION

This research has successfully managed to identify and develop a scientific based method understanding the relationship between landuse and urban network location, by modeling for transportation air pollution and analyzing the successful and non-successful development of landuses and urban networks based on the developed model. The research strategy is able to support urban planners with a range of options. Implementation of the model suggests that some areas can be more suitable than others for residential landuse and urban transportation network development, if performances and criteria are considered carefully. This suitability largely depends on the goals of the transportation projects, but importance of the main negative element (air quality) can not be ignored in all of transportation projects. The operational model can handle multiple types of transportation sources, by corresponding and averaging conditions, where each condition includes some important elements with respect to human health, all economic. The model was not sensitive to small changes of the values of input parameters. Therefore under specific meteorological conditions, areas far away from the emission sources (roads) can also be highly polluted.

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