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# IMPACT OF ROAD TRAFFIC ON AIR EMISSIONS: CASE STUDY KAFR EL-SHEIKH CITY, EGYPT

#### Mohamed Ragab<sup>1</sup>, Ibrahim H. Hashim<sup>2</sup>, Gaber M. Asar<sup>3</sup>

<sup>1</sup> Department of Civil Engineering, Higher Institute of Engineering and Technology in Kafr El-Sheikh, Egypt

- <sup>2</sup> Department of Civil Engineering, Faculty of Engineering, Menoufia University, Egypt
- <sup>3</sup> Department of Mechanical Power Engineering, Faculty of Engineering, Menoufia University, Egypt

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Abstract: Road traffic is one of the major sources of air emissions. The latest studies have indicated that, some traffic management measures used to improve the traffic operations may increase traffic emissions. Therefore, the objective of this paper is to evaluate the effect of road traffic on air emissions. In order to evaluate the traffic management measures, it is necessary to develop a microscopic traffic simulation model. Kafr El-Sheikh City, Egypt is used to build a traffic simulation network. The micro-simulation software VISSIM was used to model and analyze the selected network. The model was calibrated and validated using the collected data. The study investigated three scenarios using the developed model; Scenario 0 (original scenario), Scenario 1 (improvement of traffic flow) and Scenario 2 (promotion of public transportation). The evaluation was conducted based on travel time, fuel consumption and air emissions including Carbon Monoxide (CO), Nitrogen Oxides  $(NO_x)$  and Hydrocarbons (HC). The comparison of simulation results indicated that, the original scenario (Scenario 0) exhibited higher air emissions than other scenarios. Also, Scenario 1 exhibited lower travel time than other scenarios. The paper concluded that, the improvement of traffic flow in the study network can reduce air emissions as well as travel time whereas; the promotion of public transportation in the study network can reduce air emissions but cannot reduce travel time.

**Keywords:** air emissions, fuel consumption, traffic management, traffic simulation, travel time, VISSIM.

## 1. Introduction

In rapidly urbanizing countries like Egypt, the road traffic sector is growing rapidly. This has led to overcrowded roads and pollutions. On the congested roads, vehicles slow down and almost stop when stuck in a bottleneck; therefore, creating more air pollutant emissions. Some important vehicle emissions may release chemicals that are harmful to the environment and human health. Nowadays, the environmental analysis and the impact for any project are fundamental before proceeding to implementation of proposed designs (Olarte, 2011).

Consequently, traffic management measures are widely applied to improve the operational performance of road traffic. Such measures are rarely considered for being used to improve the environmental performance. Therefore, the implementation of these measures should not consider improving the operational performance of roads only,

<sup>&</sup>lt;sup>1</sup> Corresponding author: mohamedragab@sh-eng.menofia.edu.eg

but also the environmental performance (Kun and Lei, 2007).

The study investigated three scenarios to reduce air emissions due to road traffic. The scenarios are:

- Scenario 0: Original scenario (The selected network was analyzed with actual traffic volumes and speeds);
- Scenario 1: Improvement of road traffic flow; and
- Scenario 2: Promotion of public transportation.

According to an Environmental Protection Agency (EPA) report in 2005, road traffic contributes to 58.8% of Carbon Monoxide (CO), 35.5% of Nitrogen Oxides  $(NO_v)$ , and 25.8% of Hydrocarbons (HC) to the total emissions (Environmental Protection Agency, 2005). Therefore, the analysis was carried out using the micro-simulation platform VISSIM V. 7.00 (PTV AG, 2014) software to measure the levels of vehicle air emissions including Carbon Monoxide (CO), Nitrogen Oxides  $(NO_y)$  and Hydrocarbons (HC) due to road traffic for the selected network for each scenario. The analysis was carried out also to determine the travel time and to measure the vehicle fuel consumption.

The paper starts with the first section which presents a brief introduction followed by the second section which discusses previous studies. The third section illustrates the methodology of the study. The results of the analysis are presented in section four. Finally, conclusions are presented in section five.

## 2. Literature Review

Many years ago, traffic management has been used especially to improve traffic flow

efficiency. However, with the increase of environmental concerns, traffic management can also be used to reduce the negative impacts of traffic on the environment. Microscopic traffic simulation models play an important role in the evaluation of traffic management measures (Jeihani et al., 2015). Examples of commercially available and widely-used micro-simulation packages include VISSIM, AIMSUN, CORSIM, SimTraffic, Paramics, among others. Model calibration and validation are two necessary steps to ensure the reliability of the developed model. The accuracy of the model outputs mainly depends on the quality of the calibration and validation process (Milam, 2000).

Simulation models can be classified into small network models and medium/large network models. The calibration process of small network models, involving static routing decision, usually focuses on driver behavior and lane change parameters. Medium/large network models, on the other hand, are more intensive as they use dynamic traffic assignment that requires demand matrix estimation and route choice calibration (El Esawey and Sayed, 2011).

Air emissions due to road traffic include Nitrogen Oxides, Carbon Monoxide and Hydrocarbon. Carbon Monoxide (CO) is formed during combustion when there is insufficient oxygen to fully oxidize the fuel. CO is a potentially dangerous emission and gasoline engines contribute to more than 90% of the total emissions. In urban areas, CO concentrations follow a diurnal pattern, which mainly depends on traffic volume. Nitrogen Oxides (NO<sub>x</sub>) are released into the atmosphere by vehicle exhausts especially diesel engines. Nitric Oxide (NO) is the main emission from vehicles. Nitric Oxide is converted into Nitrogen Dioxide  $(NO_2)$ by reaction with oxidants presents in ambient air. Hydrocarbon (HC) is chiefly released into the atmosphere by vehicle exhausts especially vehicles using petrol. Hydrocarbon components due to vehicle exhaust are mainly Ethylene, Acetylene and Benzene (Srinivasan and Subramaniam, 1979).

Many studies investigated the impact of traffic management measures on air emissions due to road traffic. In a study conducted by Rakha et al. (2000), the authors investigated the effect of signal coordination on air emissions. The results showed that, efficient signal coordination can reduce air emissions up to 50% in a highly simplified scenario.

Kun and Lei (2007) evaluated the impact of setting bus exclusive lane and the effect of optimization of signal timing plan on air emissions in Beijing, China. The study showed that, setting bus exclusive lane can improve the traffic operation of the roads in the study network, and reduce the emissions of CO, HC, and  $NO_v$  of buses by 2.58 %, 5.02 %, and 2.67 % respectively. However, the setting bus exclusive lane increases the CO emissions of cars and Light Good Vehicles (LGVs), by 13.26% and 16.52% respectively. Also, the study illustrated that; the optimization of signal timing plan can improve the operational and the environmental performance of road traffic.

Neunhauserer and Diegmann (2010) used a microscopic traffic simulation model (VISSIM) to simulate an arterial road in Cologne-Mulheim, Germany, containing several signalized intersections over a length of 1 km. Two scenarios, without and with coordinated traffic signals, were investigated. Average  $NO_x$  emissions were subsequently estimated for each street section, Depending on the considered section, they found changes in  $NO_x$  emissions ranging from a decrease by 45% to an increase by 18 %.

Zallinger et al. (2010) investigated the effect of signal coordination along an existing arterial road with 12 signalized intersections in Graz, Austria. Simulation results showed that, optimized signal settings could reduce fuel consumption,  $NO_x$  and PM (particulate matter) emissions by 14%, 19% and 17% respectively.

## 3. Methodology

The proposed methodology of this study consists of the following subsections. The first section describes the studied network. Next, the procedure for model calibration and validation is discussed. Then the developed model is obtained. Lastly, traffic management measures are investigated using the developed model.

## 3.1. Study Network

The study network is a part of Kafr El-Sheikh City, Egypt. This road network is located near the capital of Egypt, Cairo, which is a typical representation of urban road network. This network covers two classes of roads including arterials and local roads as shown in Fig. 1. These roads are two way traffic directions. The geometric characteristics of these roads are given in Table 1. Geometric characteristics include road type, number of lanes, lane width, median width and sidewalk width.

ID	Road Name	Road Type	No. of Lanes	Section Length (km)	Average Lane Width (m)	Average Median Width (m)	Average Sidewalk Width (m)
1	Stad	Arterial	6	0.9	3.65	4.00	1.85
2	El-Khalefa	Arterial	4	1.10	3.55	2.55	1.9
3	El- Masnaa	Arterial	6	1.00	3.50	4.50	1.5
4	Ibrahim El- Moghazy	Local Road	4	0.55	3.25	1.80	2.10

 Table 1

 Geometric Characteristics of the Study Network



**Fig. 1.** *The Selected Network for the Study (From Google Maps)* 

# 3.2. Development of the Micro-simulation Model

VISSIM is a microscopic simulation model that was developed by German company Planung Transport Verkher (PTV) Vision Suite. This software has the ability to simulate multimodal traffic flows including cars, trucks, buses, etc. VISSIM is versatile and provides the modeler with the ability to model a wide range of traffic operations in both the interrupted and uninterrupted traffic environment (PTV AG, 2014). VISSIM was chosen to simulate the study network where the behavior type was set to be urban traffic which uses Wiedemann 74 car following model. The following subsections provide more details on the model development.

# 3.2.1. Determination of Measures of Effectiveness

Determination of measures of effectiveness (MOE) is considered the first step in the calibration and validation process. In this study, the average travel speeds were used for the calibration process while the average travel times were used for the validation process. These parameters were selected because of their ease of collection from the field and VISSIM output files.

## 3.2.2. Field Data Collection

Two types of data are required to build a VISSIM simulation model for the selected network. The first type of parameters includes existing geometric characteristics and traffic counts used for network coding of the simulation model. The second type is the field data used for the calibration/validation process of model parameters including average travel time and average travel speed.

#### Geometry Inputs

Geometry and network inputs include all other inputs to the model that are not associated with the volumes, traffic compositions, or routing decisions in VISSIM. The selected network was coded in VISSIM V. 7.00 using the data collected from the field. The simulator in the VISSIM model is responsible for generating traffic and is where the network is graphically built. Aerial photos were obtained from Google Maps patched together and used as a background for network coding.

#### Traffic Inputs

Volume data and traffic compositions were extracted at 15-minute intervals by manual classified counts on links of the selected network in the study area.

Traffic was classified into five vehicle classes: motorcycles, passenger cars including private cars and taxis, light good vehicles including mini trucks and microbuses, heavy vehicles and buses. Average travel speeds and average travel times were obtained using moving car technique for network links, in the study area.

Collection of traffic data was carried out in normal working days during the daylight hours. During data collection periods, the weather was clear and the pavement was dry. Average values of traffic volumes for the study network are presented in Table 2. Whereas average values of travel speeds and travel times for the study network are presented in Table 3.

#### Table 2

Average Values of Traffic Volumes for the Network Links

Road		Direction	Traffic Volumes (veh/h)	
	Link 1	Eastbound	687	
Stad		Westbound	564	
Stad	Link 2	Eastbound	687	
	LINK Z	Westbound	587	
	Link 3	Eastbound	325	
El-Khalifa		Westbound	536	
El-Kitalita	Link 4	Eastbound	527	
	LINK 4	Westbound	583	
El- Masnaa	Linh C	Northbound	558	
EI- Masnaa	Link 5	Southbound	646	
Ihmhim El Maghany	Link 6	Northbound	188	
Ibrahim El-Moghazy	LIIKO	Southbound	211	

Road	Direction	Section Length (km)	Average Travel Speed (km/h)	Travel Time (sec)
C 1	Eastbound	0.0	38	76
Stad	Westbound	0.8	42	69
	Eastbound	1.0	33	109
El-Khalifa	Westbound	1.0	31	116
	Northbound	0.5	28	64
El- Masnaa	Southbound	0.5	26	69
	Northbound	0.5	29	62
Ibrahim El-Moghazy	Southbound	0.5	27	67

# Table 3 Average Values of Travel Speeds and Travel Times for the Study Network

# 3.2.3. Model Calibration

Model calibration is the process of modifying and determining the set of model parameters, based on modeling judgments and collected data. Such parameters accurately represent the prevailing field conditions of a given study area. In general, the target of calibration is to ensure that the simulation outputs of the evaluation variable match the observed data. Instead of calculating average errors to assess the quality of the calibration, a more robust approach is applied where the "matching" is evaluated using a suitable hypothesis statistical test. If the simulation output fits the observed data statistically then the model is said to be calibrated. Otherwise, the calibration variable would need further fine tuning (Manjunatha et al., 2013).

The desired speed distribution is one of the most influential parameters of a VISSIM simulation model. Therefore, average travel speed and traffic volume were chosen as the calibration parameters in the current analysis. The model would have been considered calibrated if a statistical agreement is found between the simulated and observed traffic volumes. In this study, the evaluation criteria used in the calibration process includes:

#### - GEH Statistic

The GEH Statistic is a formula used in traffic engineering, traffic forecasting, and traffic modeling to compare observed versus simulated link volumes (Oketch and Carrick, 2005). The GEH formula gets its name from Geoffrey E. Havers, who invented it in the 1970s while working as a transport planner in London, England. The GEH is represented by the equation as below (Eq. (1)):

$$GEH = \sqrt{\frac{(O-M)^2}{0.5(O+M)}}$$
(1)

where: O = Observed hourly traffic volume values; and M = Simulated hourly traffic volume values.

Various GEH values give an indication of a goodness of fit, as outlined below:

GEH < 5: Traffic volumes can be considered a good fit;

5 < GEH < 10: Traffic volumes may require further investigation; and

10 < GEH: Traffic volumes cannot be considered to be a good fit.

#### - Coefficient of Determination $(R^2)$

Coefficient of determination  $(R^2)$  is a statistical measure that gives information about the 'goodness of fit' of the model. In other words, it shows how well the regression line approximates the observed data points. If  $R^2 = 1$ , the model is exactly predicting the test data while  $R^2 = 0$  indicates there is no correlation between the model results and the field measurements. However; a value of 0.75 or higher is recognized as good model (Xiong et al., 2014).

## 3.2.4. Model Validation

Validation of a simulation model is the next stage after ensuring that the model is well-calibrated. Validation is defined as the process of matching the output of the calibrated simulation model with some reallife measurements that were not used in the calibration. The evaluation measures used in this process should be different from the measures used in the calibration process. Alternatively, the measurements of the same variable can be utilized but for different locations, time periods, and/or traffic conditions (Park and Schneeberger, 2003). The model is deemed "valid" if the chosen MOE from the unused real-life dataset is close enough to the simulation values. Otherwise, the calibration process has to be re-executed until comparable results are achieved.

Using travel time as MOE has been extensively reported in various studies. In this study, average travel times were selected as MOE for model validation. Average travel time was determined for all different vehicle classes for both the observed and the simulated travel times. The observed average travel times were computed using the collected data as described before. In this study, the evaluation criteria used in the validation process include:

#### - Mean Absolute Percent Error (MAPE)

MAPE is a measure that used to calculate the average difference between simulated and observed data. The formula for the "MAPE" is (Eq. (2)):

$$MAPE = \frac{1}{N} \sum_{i=1}^{N} \left| \frac{O - M}{O} \right|$$
(2)

#### - Root Mean Square Error (RMSE)

RMSE is a measure of average variation between observed and simulated data. It is often used to reflect the absolute deviation of data. The equation for calculating RMSE is shown below (Eq. (3)):

$$RMSE = \frac{\sqrt{\frac{\sum_{i=1}^{N} (O - M)^2}{N - 1}}}{\left(\frac{\sum_{i=1}^{N} O}{N}\right)}$$
(3)

#### - Normalized Root Mean Square Error (RMSN)

RMSN is a measure of average variation between observed and simulated data. It is often used to indicate the relative deviation of data.

The Normalized Root Mean Square Error (RMSN), can be calculated according to Balakrishna et al. (2007), as follows (Eq. (4)):

$$RMSN = \frac{\sqrt{N \cdot \sum_{i=1}^{N} (O - M)^2}}{\sum_{i=1}^{N} O}$$
(4)

Where: O = Observed values; M = Simulated values; and N = Number of samples or observations.

The FHWA guide (FHWA, 2004) presented the evaluation criteria of the Wisconsin Department of Transportation (DOT) which generally suggest that a 15% error margin can be acceptable in similar exercises.

# 3.3. Traffic Management Measures

Various traffic management measures can be implemented to reduce air emissions due to road traffic. The following subsections describe the investigated measures in this study.

# 3.3.1. Improvement of Road Traffic Flow

Improvement of road traffic flow could reduce the number of traffic jams and thus reduce air emissions due to road traffic, because these emissions tend to increase at lower speeds and particularly with stop-startdriving (Cerezo, 1996). To study the effect of improvement of road traffic flow on air emissions, a new scenario was created using micro-simulation VISSIM software V. 7.00. The new scenario (Scenario 1) was created in the same network by increasing allowed road speed by 10 km/h hypothetically. This will be achieved by measures such as improving of road pavement condition and reducing number of side accesses.

# 3.3.2. Promotion of Public Transportation

Promotion of public transportation could decrease the use of private cars,

and thus reduce the congestion and the air emissions due to road traffic (Cerezo, 1996). To study the effect of promotion of public transportation on air emissions, a new scenario was created using microsimulation VISSIM software V. 7.00. The new scenario (Scenario 2) was created in the same network with decreasing of traffic volumes by 10% hypothetically. It can be achieved by measures such as enhancing public transport services by increasing frequency, convenience and travel speed of public transport and setting restrictions for access to city centers for private and heavy traffic.

# 4. Results

# 4.1. Model Calibration

Initially, simulation runs were executed with the default parameters of VISSIM (i.e. with no calibration) for the selected network. The traffic volumes resulted from simulation were determined and compared with the observed volumes by values of GEH. The values of GEH for the initial simulation runs are presented in Table 4. The results indicated that, running the model with default parameters would not be appropriate and hence, the speed distributions of all vehicle classes had to be edited in the simulation model to be as close as possible to the observed average travel speeds.

The model was re-run again and the simulated traffic volumes of the calibrated model were determined. Again, GEH statistics were computed for the calibrated model and the results are presented in Table 4. As shown in Table 4, the statistics of the calibrated model indicate a significant improvement over the results of the default model. The hypotheses of similarity were accepted and model calibration was considered to be successfully completed.

Also, coefficient of determination  $(R^2)$  is calculated to evaluate the accuracy of the model by indicating the correlation between field counts and simulated counts. The final comparison of the observed and simulated traffic volumes is shown in Fig. 2, in which most of the comparison points conform to the diagonal line. Validation of the simulation model will be the final step to make sure that the model replicates field conditions for another measure.



**Fig. 2.** *Comparison of Observed and Simulated Counts for the Selected Network* 

#### Table 4

Results of GEH for Model Calibration

Road		Direction	No Calibration GEH	Calibration Criteria (GEH < 5) met?	After Calibration GEH	Improvement (%)
Stad	Link 1	Eastbound	0.66	Yes	0.54	18%
		Westbound	5.92	No	0.94	84%
Stad	Link 2	Eastbound	0.66	Yes	0.54	18%
		Westbound	5.52	No	1.07	81%
	Link 3	Eastbound	1.57	Yes	1.56	1%
El-Khalifa		Westbound	5.60	No	4.82	14%
El-Khalifa	Link 4	Eastbound	0.60	Yes	0.60	0%
		Westbound	4.53	Yes	4.37	4%
<b>F1 M</b>	Link 5	Northbound	3.79	Yes	1.87	51%
El- Masnaa		Southbound	6.32	No	0.73	88%
Ibrahim El-	Link 6	Northbound	3.45	Yes	3.45	0%
Moghazy		Southbound	2.66	Yes	2.42	9%

# 4.2. Model Validation

The simulated average travel times of the calibrated model were compared to the observed travel times. Different error measurements were computed for all vehicle classes and the results are presented in Fig. 3 to Fig. 5. From these figures, it can be

noticed that, all the error measurements were below 7% indicating a reasonable matching between the simulated and the observed travel times. Overall, all validation results were satisfactory with minimal errors. Therefore, it can be concluded that, the model is successfully calibrated and validated.



#### Fig. 3.

Mean Absolute Percent Error Measurements for Model Validation of all Vehicle Classes for the Selected Network



#### Fig. 4.

Root Mean Square Error Measurements for Model Validation of all Vehicle Classes for the Selected Network



#### Fig. 5.

Normalized Root Mean Square Error Measurements for Model Validation of all Vehicle Classes for the Selected Network

#### 4.3. Simulation Analysis

Traffic simulation model is used to evaluate different scenarios, in terms of travel times, air emissions of CO,  $NO_x$  and HC, and fuel consumption. Such measures were collected as outputs from model simulation, in order to evaluate the effectiveness of each investigated traffic management scenario.

The results of the 3600 sec (60 minutes) simulation period were used for the analysis. The results of total air emissions, fuel consumption and travel time aggregated on all network links for all scenarios are illustrated in Table 5. Also, the comparisons of different scenarios are shown in Fig. 6 to 10. These figures showed that;

• Increasing of road speed can reduce the emissions of CO, NO<sub>x</sub> and HC

and fuel consumption for all vehicle classes by about 3.35%. Also, there is a significant reduction in average travel time for all vehicle classes. The reduction rate of average travel time is about 7.21%.

• Decreasing of traffic volumes can reduce the emissions of CO,  $NO_x$  and HC and fuel consumption for all vehicle classes by about 9.61%. Also, there is no significant reduction in average travel time for all vehicle classes. The reduction rate of average travel time is about 1.10%.

It can be concluded that, the original scenario (Scenario 0) exhibited higher air emissions than other scenarios. Also, Scenario 1 exhibited lower travel time than other scenarios.

### Table 5

Total Air Emissions, Fuel Consumption and Travel Time Aggregated on all Network Links for Different Scenarios

Scenario	Original Scenario (Scenario 0)	Scenario 1	Scenario 2
CO (gm)	7069	6832	6389
NO <sub>x</sub> (gm)	1376	1330	1243
HC (gm)	1639	1584	1481
Fuel Consumption (gal)	102	98	92
Travel Time (sec)	721	669	713



### Fig. 6.

Total CO Emissions Aggregated on all Network Links for Different Scenarios



**Fig. 7.** Total NO<sub>x</sub> Emissions Aggregated on all Network Links for Different Scenarios







**Fig. 9.** *Total Fuel Consumptions Aggregated on all Network Links for Different Scenarios* 



**Fig. 10.** Total Travel Times Aggregated on all Network Links for Different Scenarios



# 5. Conclusions and Recommendations

This paper investigates the impact of traffic management measures on air emissions due to road traffic. The micro-simulation model was developed for the selected network. Then, it calibrated and validated using the collected data. The model was calibrated using traffic volume and travel speed parameters and it was validated using travel time parameter. The study investigated three scenarios investigated using the developed model. The analysis was carried out using micro-simulation VISSIM software. Three major types of vehicle emissions were estimated; Carbon Monoxide (CO), Nitrogen Oxides (NOX) and Hydrocarbons (HC), also travel time and vehicle fuel consumption were estimated and aggregated on all network links for every scenario. Analysis showed that,

- 1. The superiority of the calibrated model compared to the model with default parameters in terms of improved GEH statistics.
- 2. Different error measurements were computed to assess the validity of the model. In general, all the errors were below 7% showing a reasonable matching between the observed and the simulated travel times.
- 3. The improvement of road traffic flow in the study network can reduce the emissions of CO, NOX and HC and fuel consumption for all vehicle classes by about 3.35%. Also, there is a significant reduction in average travel time for all vehicle classes. The reduction rate of average travel time is about 7.21%.
- 4. The promotion of public transportation in the study network can reduce the emissions of CO, NOX and HC and fuel

consumption for all vehicle classes by about 9.61%. Also, there is no significant reduction in average travel time for all vehicle classes. The reduction rate of average travel time is about 1.10%.

In summary, the improvement of road traffic flow in the study network can reduce air emissions as well as travel time whereas; the promotion of public transportation in the study network can reduce air emissions but cannot reduce travel time. Therefore the recommendations for practitioners may include improving road pavement condition, reducing number of side accesses, driving at a medium speed, well design of intersections, enhancing public transport services by increasing frequency, convenience and travel speed of public transport, and setting restrictions for access to city centers for private and heavy traffic.

A future extension of this work will include obtaining emissions field data in order to verify the results obtained from VISSIM using emissions detection instrument.

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