Analyzing Tehran’s Air Pollution Using System Dynamics Approach

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Abstract: Air pollution is one of the major issues in urban management. City managers and planners pay a great deal of attention to this problem given its harmful effect on people’s health and the environment. This paper proposes a simulated model of the pollution level of the capital city of Tehran along with its source and outcomes based on system dynamics. First, it provides a comprehensive review of the sources of greenhouse emission along with the impact of natural factors on its growth. The data is collected from 2011 to 2021. Then a dynamic model is applied to simulate Tehran’s air pollution for twenty years (from 2011 to 2031). The statistical method of design of experiments (DoE) is considered to set the sensitive and controllable variables of the city’s pollution. The information is used to set plans to reduce the air pollution considering several scenarios. The results of the research show that the best strategy to reduce air pollution that can resolve other issues as well is to manufacture environmentally friendly cars that can rely on renewable energy resources as their fuels. The other highly important strategy is to preserve forests.

Keywords: air pollution; sustainable city; system dynamics; design of experiments (DoE); greenhouse gases; urban management

1. Introduction

Air is vital for the maintenance of human life and also necessary for other creatures on planet Earth. Without air, no one could ever keep on living for more than a few minutes. Air is made up of gases such as nitrogen, oxygen, carbon dioxide, and other gases [1,2]. A slight change in its composition may result in unpleasant atmospheric conditions for humans and the environment around them. This is called air pollution. Air pollution is a serious environmental issue, particularly in developing countries.

The path of air pollution in the city involves emission and expansion of air pollutants resulting from various factors. Throughout the path, air pollutants are dispersed via various dispersion pathways (air, water, soil, living organisms) thinning and undergoing chemical reactions [3]. In other words, the phenomenon of air pollution is a series including various events such as generation and dispersion of the pollutants from their source, their conversion, transmission, and emission into the atmosphere, and their impact on humans, materials, and the ecosystem [4]. Additionally, air pollution is a major health issue in developing countries. Despite better air quality in the last 50 years, air pollution still reduces the life span of humans and causes a large number of people to go to hospitals. It also causes much harm to the natural environment such that frequent studies indicate that people notice air pollution constantly, especially in urban and industrial regions. Furthermore, epidemiological studies in the last two decades have shown that air pollution
is one of the leading causes of death [5]. On the other hand, industrialization and population growth play the main role in air pollution. Air pollution is attributed to an increase in the concentration of greenhouse gases (air pollutants). Air pollutants are classified into several categories.

1.1. Primary and Secondary Pollutants

Primary pollutants enter the atmosphere directly from the source and include: (1) carbon compounds, such as CO, CO$_2$, CH$_4$, and VOCs; (2) nitrogen compounds, such as NO, N$_2$O, and NH$_3$; (3) sulfur compounds, such as H$_2$S and SO$_2$; (4) halogenated compounds, such as chloride, fluoride, and bromide; and (5) particulate matter (PM) or “aerosols”.

Secondary pollutants are those which do not get into the atmosphere right from the source but are formed by the primary pollutant compounds. They involve (1) NO$_2$ and HNO$_3$ composed of NO; (2) ozone (O$_3$) which is formed as a result of a photochemical reaction of nitrogen oxides and VOCs; (3) sulfuric acid made up of SO$_2$ and nitric acid made up of NO$_2$; (4) sulfate and nitrate aerosols (such as ammonium sulfate and nitrate ammonium) formed by the reaction of sulfuric acid and nitric acid with NH$_3$ interactively; and (5) organic aerosols formed from VOCs in gases-to-particle reaction. A study by Okedere estimates the contribution from the open burning of municipal wastes in some of Nigeria’s major cities to atmospheric levels of air pollutants [6].

1.2. Criteria Pollutants Involve

(1) Particular matter (PM), sulfur dioxide (SO$_2$), nitrogen oxides (NO and NO$_2$), carbon monoxide (CO), ozone (O$_3$), and lead (Pb); (2) toxic pollutants; (3) radioactive pollutants; (4) indoor pollutants [7].

Air pollution is a global issue that has attracted considerable attention from governments and researchers alike, and various research studies are carried out annually by researchers from different countries, which are published in journals. Table 1 refers to some of the most recent publications.

Table 1. Some articles on urban air pollution.

| Title | Authors |
|-------|---------|
| INSPiRE: an integrated approach to tackling household air pollution and improving health in rural Cambodia | Chandna and Honney, 2017 [8] |
| Long-term exposure to urban air pollution and the relationship with life expectancy in cohort of 3.5 million people in Silesia | Dziubanek et al., 2017 [9] |
| Highway toll and air pollution: evidence from Chinese cities | Fu and Gu, 2014 [10] |
| Estimation of spatial patterns of urban air pollution over a 4-week period from repeated 5-min measurements | Gillespie et al., 2016 [11] |
| Chemical characteristics of atmospheric PM2.5 loads during air pollution episodes in Giza, Egypt | Hassan and Khoder, 2017 [12] |
| Urban air pollution from the open burning of municipal solid waste | Okedere et al., 2019 [6] |
| Optimized stochastic methods for sensitivity analysis for large-scale air pollution model | Todorov, V. et al., 2021 [13] |
| Advanced stochastic approaches for Sobol’ sensitivity indices evaluation | Todorov, V. et al., 2021 [14] |

Hence, it is imperative to constantly monitor the level of pollutants in the air and ensure that air quality meets the standards [15]. Air quality in cities results from a complicated interaction between environmental conditions and human factors [3]. Additionally, different factors result in the unreliability of calculation risks which can lead to changes in
weather conditions and fluctuations in emission sources and activities in the long run as well [15]. Furthermore, despite our progress in the mechanical understanding of ecological processes, the complication of ecology in the real world still limits our ability to predict ecological dynamics, particularly once ongoing environmental pressures are imposed. Hence, developing models of prediction is a challenging problem due to the structural unreliability, unknown parameters, and need for unlimited data sets for sample predictions [16]. On the other hand, system dynamics is a method to learn complicated systems which deals with process modeling over time [17] and can help us understand ecological dynamics.

System dynamics is a means which helps us to focus on complex problems such as delays, feedback, and nonlinearity. System dynamics with its own development is a field of study which can include science, education, engineering, or medicine. Dynamic modeling consists of four stages, briefly as: (A) conceptualization: defines the purpose of the model and the boundaries, identifies the key variables of the model, describes the behavior or draws reference modes of the key variables, and graphs the basic mechanisms, feedback loops, and system; (B) formulation: involves converting feedback diagrams into level and rate equations, and estimating and selecting the values of the parameters; (C) testing: includes simulation of the model and testing the dynamic hypothesis, testing the model’s assumptions, and testing model behavior and sensitivity to perturbations; and (D) implementation: includes testing the model’s response to different policies, and translating the study insights into one accessible form [18].

The use of system dynamics to study environmental systems has gained interest in recent years. In 2010, a study was conducted on Suzhou water resources in China which addressed modeling of water carrying capacity of water resources from 2001 to 2030, in which optimization plans for three water resources were examined [19]. In 2012, another survey addressed the interactions between water resources and environmental flow, as well as economic and social status, which was aimed at the evaluation of socio-economic effects of various levels of environmental flows allocation in the Vihe River Delta in China [20]. Another study was carried out in China in the same year in which a hybrid approach made of genetic algorithm and system dynamics was considered to measure the pollution burden of coal production on the environment. The genetic algorithm was applied to optimize the system dynamics parameters [21]. Moreover, research was conducted in Spain in 2013 which used system dynamics to design a model to study the fate of persistent organic pollutants in soil. The simulation results indicated that the amount of linden in soil had decreased over a 10-year simulating process [22]. Bastan et al. showed how a system dynamics simulation approach can provide deep insights into the field of sustainable development and present efficient policies for agricultural sustainability [23].

The major part of learning and understanding of how systems and processes work is to observe them while they are in progress. To truly understand the causal loop relations of a system, one should purposefully change the system’s inputs and check the results in the output. In other words, experiments should be done in a way in which each experiment is regarded as a test. Design of experiments is a means in the scientific and engineering world in order to improve the product realization process which is used for evaluation and comparison of basic design configuration, evaluation of material alternatives, and determination of key product design parameters, which affect product performance. Its stages include (1). recognition of and statement of the problem; (2). selection of the response variable; (3). choice of factors, levels, and ranges; (4). choice of experimental design; (5). performing the experiment; (6). statistical analysis of the data; and (7). conclusions and recommendations [24].

The capital city of Tehran, with a population of over 8,000,000, has been suffering from severe air pollution due to fast environmental developments in the last couple of years [25]. Hence, it seems quite necessary to create a dynamic model aiming at examining the impact of various factors in the city’s air pollution. This article is aimed at building such a model. Then, with the statistical method of design of experiments (DoE), the sensitive and controllable variables are set for the city’s pollution and eventually, plans are offered to
reduce the air pollution and several scenarios are proposed to reduce Tehran’s air pollution for a better life. The main contributions of this paper are as follows:

It proposes a simulated model of the pollution level of the capital city of Tehran along with its source and outcomes based on system dynamics.

It considers multiple variables affecting air pollution by applying the design of experiments (DoE) concept to identify and regulate key factors affecting air pollution. Note that the use of DoE is a novel concept compared to existing works.

2. Proposed Method

This paper uses system dynamics to synthesize a simulated model of the polluted capital city of Tehran. It’s different implementing stages are as follows:

- **Stage 1.** Defining the dynamic hypotheses.
- **Stage 2.** Drawing a causal loop diagram for the system of greenhouse gases emissions in Tehran.
- **Stage 3.** Drawing a stock-flow diagram of the system of greenhouse gases emissions in Tehran.
- **Stage 4.** Compiling and formulating a model of the system of greenhouse gas emissions in Tehran in mathematical form, which shows interactions inside the system.
- **Stage 5.** Identifying critical variables through the statistical method of design of experiments.
- **Stage 6.** Demonstrating the output of the model based on optimal values.
- **Stage 7.** Defining scenarios for reducing greenhouse gases based on optimized values.

Data for this model were collected from government sites and the model was constructed using causal loops designed based on dynamic hypotheses defined by field experts. Further details pertaining to the data can be found in the Appendix A.

2.1. Causal Loop Diagram

The causal loop diagram is a simple map which shows the relations between the variables by means of arrows relating the cause to the effect. Additionally, the causal loop diagram is a system theory loop which is composed of two balance and reinforcement circles. In the causal loop diagram, the arrows with a mark of “+” are used for the direct relationship between causal loop and those with a minus mark “−” are applied for an inverse relation between causal loop [18].

This article represents Tehran’s air pollution in three loops which will be discussed below. It is necessary to first explain the micro-loops to demonstrate the causal loop diagram. That is why micro-loops are brought forward initially and then the causal loop diagram of the entire system will be illustrated. The diagram in Figure 1 represents the causal diagram of the cars which increases once the number of cars grows. Once the cars are totally gone out of the order (junk or scrapped cars) annually, the number of cars reduces.

![Figure 1. Cars loop.](image-url)
Figure 2 demonstrates the circle of a number of non-industrial units in which the units grow once the constructions get a license to expand. After their lifetime, a period of 20 years, they suffer total damage and get out of the circle.

![Figure 2. Non-industrial units loop.]

Figure 3 illustrates the circle of population variable; when the birth rate increases the population increases. It reduces by natural death and mortality due to pollution.

![Figure 3. Population loops.]

Figure 4 shows a causal loop diagram for the whole system. As demonstrated, pollution grows when the pollutants are dispersed. In fact, some of them are destroyed as a result of natural factors and the remaining causes’ deaths.

2.2. Stock-Flow Diagram

Although the causal loop diagrams involve many benefits, in some situations, they also have weaknesses such as the inability to show the stock-flows of the system. This is why stock-flows along with feedback are two central concepts in system dynamics theory.

The stock-flow diagram includes three basic elements: stock, flow, and converters [18]. Stocks suggest accumulations inside the system and its stagnation. Flows indicate stocks deviations for input and output. Finally, converters are auxiliary variables to better visualize variables and adjust flows behaviors [18].

The boxes in the diagram represent stock variables and faucets show input and output flows. Auxiliary variables are marked with no sign in the graph and are simply represented with their own names. To draw the stock-flow diagram, its variables are first presented in Table 2. Then, in Figure 5, the diagram is presented.
Figure 4. Causal loop diagram for the climate pollution in Tehran.

Table 2. Variable units of air pollution stock-flow.

| Variable                                      | Unit            |
|-----------------------------------------------|-----------------|
| MR: Marking rate                             | car/year        |
| M: Marking                                   | car/year        |
| C: Cars                                       | car/year        |
| S: Scraping                                   | car/year        |
| SR: Scraping rate                            | 1/year          |
| ERC: Emission rate per car                   | ton/car/year    |
| EMS: Emissions by mobile source              | ton/year        |
| NLR: Non-industrial licensing rate            | building/year   |
| NL: Non-industrial licensing                 | building/year   |
| NB: Non-industrial buildings                 | building        |
| DR: Destruction rate                         | 1/year          |
| D: Destruction                               | building/year   |
| ENB: Emission rate per non-industrial building | ton/building/year |
| ILR: Industrial licensing rate               | building/year   |
| IL: Industrial licensing                     | building        |
| IB: Industrial buildings                     |                 |
| EIB: Emission rate per industrial building   | ton/building/year |
| ESS: Emission by stationary sources          | ton/year        |
| VOCs                                         | ton/year        |
| CO                                           | ton/year        |
| NOx                                          | ton/year        |
| Sox                                          | ton/year        |
| PM                                           | ton/year        |
| GG: Greenhouse gas                           | ton/year        |
| AR: Afforestation rate                       | hectare/year    |
| A: Afforestation                             | hectare/year    |
| F: Forests                                   | hectare         |
| FFR: Forest fire rate                        | hectare/year    |
| FF: Forest fire                              | hectare/year    |
| CCF: Cleaning capacity of forests            | ton/hectare/year |
Table 2. Cont.

| Variable                  | Unit              |
|---------------------------|-------------------|
| R: Rain                   | day/year          |
| CCR: Cleaning capacity of rain | ton/day/year      |
| BR: Birth                 | 1/year            |
| B: Birth                  | person/year       |
| POP: population           | person            |
| DR: Death rate            | person/year       |
| ND: Natural death         | person/year       |
| DP: Dying from pollution  | person/year       |
| DPR: Dying from air pollution rate | person/ton |

Figure 5. Stock-flow diagram of Tehran’s air pollution.

2.3. Compiling and Formulating the Model and Developing a Mathematical Method

Data were collected from a variety of sources to compile and formulate the variables; the references are provided in Table A1 in Appendix A.

The model contains five stock variables defined using the following Equations:

\begin{align*}
C &= \text{INTEG} (M-S, 6,534,000) \quad (1) \\
M &= MR \quad (2) \\
MR &= 398,399 \quad (3) \\
S &= C \times SR \quad (4) \\
SR &= 1/20 \quad (5) \\
EMS &= C \times ERC \quad (6) \\
ERC &= 0.0919765 \quad (7) \\
NL &= BR \quad (8) \\
NL &= BR \quad (9) \\
NLR &= 31,091.5 \quad (10)
\end{align*}
\[ D = DR \times NB \tag{11} \]
\[ DR = 1/25 \tag{12} \]
\[ IB = \text{INTEG}(IL, 4075) \tag{13} \]
\[ IL = ILR \tag{14} \]
\[ ILR = 48 \tag{15} \]
\[ ESS = (EIB \times IB) + (ENB \times NB) \tag{16} \]
\[ EIB = 19.341 \tag{17} \]
\[ ENB = 0.046036 \tag{18} \]
\[ \text{VOCs} = \left( \frac{71,944}{613,834} \times \text{EMS} \right) + \left( \frac{11,696}{107,927} \times \text{ESS} \right) \tag{19} \]
\[ \text{Co} = \left( \frac{494,201}{613,834} \times \text{EMS} \right) + \left( \frac{12,489}{107,927} \times \text{ESS} \right) \tag{20} \]
\[ \text{NOx} = \left( \frac{39,430}{613,834} \times \text{EMS} \right) + \left( \frac{46,094}{107,927} \times \text{ESS} \right) \tag{21} \]
\[ \text{Sox} = \left( \frac{2326}{613,834} \times \text{EMS} \right) + \left( \frac{35,085}{107,927} \times \text{ESS} \right) \tag{22} \]
\[ \text{PM} = \left( \frac{5933}{613,834} \times \text{EMS} \right) + \left( \frac{2563}{107,927} \times \text{ESS} \right) \tag{23} \]
\[ \text{GG} = (\text{CO} + \text{NOx} + \text{PM} + \text{Sox} + \text{VOCs}) - (\text{CCF} \times F + (\text{CCR} \times R)) \tag{24} \]
\[ F = \text{INTEG}(A - FF, 6452) \tag{25} \]
\[ A = \text{AR} \tag{26} \]
\[ \text{AR} = 93.583 \tag{27} \]
\[ \text{FF} = \text{FFR} \tag{28} \]
\[ \text{FFR} = 10.25 \tag{29} \]
\[ \text{CCF} = 68 \tag{30} \]
\[ \text{R} = 23 \tag{31} \]
\[ \text{CCR} = 1989 \tag{32} \]
\[ \text{POP} = \text{INTEG}(B - \text{ND} - \text{DP}, 8.29314 \times 10^6) \tag{33} \]
\[ B = \text{BR} \times \text{POP} \tag{34} \]
\[ \text{BR} = 0.014616 \tag{35} \]
\[ \text{ND} = \text{DR} \tag{36} \]
\[ \text{DR} = 39,317 \tag{37} \]
\[ \text{DP} = \text{PO} \times \text{DPR} \tag{38} \]
\[ \text{DPR} = 1/51.318 \tag{39} \]

Note that the first stock variable denotes the number of cars and Equation (1) represents this number in 2011. Equation (2) represents the input flow to the car variable, which is equal to the rate of cars per year. The corresponding number in 2011 is represented by Equation (3). Equation (4) represents the car stock which equals the number of cars multiplied by the rate of scrapped (or junk) cars. The corresponding value in 2011 is represented by Equation (5). Equation (6) depicts the amount of pollutants generated by cars which is the number of cars multiplied by the pollution rate of each car, the value in 2011 of which is given by Equation (7). Equally, Equation (8) relates to the stock variable of non-industrial buildings along with the corresponding value in 2011. Equation (9) shows the input variable of non-industrial buildings which equals the number of issued licenses and permits per year. The corresponding value in 2011 is given in Equation (10). Equation (11) shows
the output variable of non-industrial buildings, i.e., the number of buildings multiplied by the rate of their destruction. The corresponding value for our case study is given in Equation (12). Equation (13) indicates the number of industrial buildings. Equation (14) indicates the number of industrial construction permits per year. The corresponding value for our case study is given in Equation (15). Equation (16) represents the amount of greenhouse gas emissions by stationary sources which is equal to the amount of greenhouse gases emitted by industrial and non-industrial buildings. Equations (17) and (18) depict the rate of greenhouse gas emissions by industrial and non-industrial constructions, respectively. Equation (19) through Equation (23) provide the amount of greenhouse gases from mobile and stationary emission sources each (cars and buildings). Equation (24) provides the sum of all greenhouse gases. Equation (25) represents the stock variable of forests along with its corresponding value in 2011. Equation (26) represents forestry rate per year, whose value in 2011 is depicted in Equation (27). Equation (28) represents the forests fires per year. The corresponding value in 2011 is given in Equation (29). Equation (30) represents the rate of greenhouse gases cleaned by one hectare of forest. Equations (31) and (32) respectively show the number of rainy days per year as well as the rate of cleaning greenhouse gases on one rainy day. Equation (33) represents the city’s population along with its value in 2011. Equation (34) illustrates the population growth computed by multiplying the population rate by birth rate. Its corresponding value in 2011 is given by Equation (35). Equation (36) represents the number of deaths (death toll) of the first output flow variable of the population. This rate for our study is given by Equation (37). Equation (38) shows the second output variable of the population (death rate as a result of pollutants), indicating the number of deaths due to pollution, whose amount equals the amount of pollutants by the pollutant rate that can result in one death. Its value for our case study is given by Equation (39).

2.4. Identifying Major Variables and Improving Them by Design of Experiments (DoE)

The statistical method of design of experiments was carried out using the Minitab software version 17.3.1. The main variables were identified as follows:
1. Eight variables out of eleven were identified using the factorial design of Plackett–Burman.
2. The eight variables were entirely tested. The corresponding uppermost and lowermost rates for these variables are illustrated in Table 3.

Table 3. The high and low rates of each variable.

| Variable | Lowermost Rate | Uppermost Rate |
|----------|----------------|----------------|
| MR       | 358,600        | 438,240        |
| ERC      | 0.0827         | 0.1012         |
| ILR      | 43             | 52             |
| EIB      | 17.4           | 21.27          |
| ENB      | 0.041          | 0.051          |
| NL       | 27,982         | 34,200         |
| AR       | 84             | 103            |
| FFR      | 9              | 12             |
| CCF      | 61             | 75             |
| R        | 20             | 26             |
| CCR      | 1790           | 2187           |

To design the experiments of the first stage, which was of Plackett–Burman, the following path was followed in Minitab software.
Stat→DoE→Factorial→Create Factorial Design→Plackett–Burman design
Figure 6 depicts the results of 48 experiments carried out in the first stage.
As represented in Figures 6 and 7, eight variables were identified for marking cars, greenhouse gas emission rate of each car, greenhouse gas emission rate of each industrial building, greenhouse gas emission rate of each non-industrial unit, as well as forestry, cleaning rate of per hectare of forest, raining, and air cleaning rate of rain, which were the major pollution system’s variables.

Entering each amount, the existing greenhouse gas emission in the air reached 0.0005584 tons per year. Different scenarios were defined based on the amounts. To do this, the optimal amount of each variable was entered in the stock-flow diagram and then the effect of the same variable, and finally, the amount of greenhouse gas emissions...
was observed. Finally, the optimal amounts of all variables were entered into the model and their entire effect was examined on the amount of greenhouse gases.

3. Results and Discussion

In this research, the model was run for 20 years (from 2011 to 2031). First, data from 10 years ago were given to the model then the actual results were compared with the model results in the first 10 years (2011–2021). The results were almost similar. Therefore, the results of the model in the next 10 years (2021–2031) can be trusted.

For performance analysis, the optimal values from Figure 8 were compared to the current and optimal status of three important variables (marking rate, emission rate of car, and emission rate of industrial buildings). Additionally, depending on the variables’ status, their behavior reasons, scenarios, and strategies were discussed.

Figure 9 indicates the rate of car marking based on the current and optimal rates and Figure 10 shows the impact of current and optimal car rate on the amount of greenhouse gas emissions. In the figures, number 1 shows current status, number 2 shows the optimal status of the greenhouse gases amount, and time (0) refers to 2011. Cars are the major source of greenhouse gas emissions in cities. On the other hand, their excessive increase not only increases the air pollution in the cities but it also affects other problems of the cities such as traffic jams, accidents, and so on. Therefore, it is recommended that the car rate be reduced to reduce air pollution and other urban problems. If this is not possible, it is a good idea to scrap the worn-out cars at a higher rate.

![Figure 9. Comparing current and optimal marking rate.](image1)

![Figure 10. The effect of marking rate on the amount of greenhouse gases.](image2)
Figure 11 depicts the greenhouse gas emissions of each car based on current and optimal rates. Figure 12 shows the effect of greenhouse gas emission rates by each car on the current and optimal amount of greenhouse gas emissions.

![Graph of emission rate per cars](image)

**Figure 11.** Comparing current and optimal cars emission rate.

![Graph of greenhouse gases](image)

**Figure 12.** The effect of cars emission rate on the amount of greenhouse gases.

Based on the above results, we can conclude that the best strategy to reduce air pollution in cities is to reduce greenhouse gas emissions by cars since (1) cars are traveling on city roads (even on off working days), (2) the most fossil fuels in the cities are consumed by cars, and (3) they are inside the cities whereas factories are located outside. Additionally, manufacturing low emission and environmentally friendly cars can drastically reduce greenhouse gas emissions.

Figures 13 and 14 represent greenhouse gas emissions by industrial buildings and the effect of their optimal and current amount on greenhouse gases emissions, respectively.

As shown by the above figures, greenhouse gas emissions by industrial units are considered one of the sources which generate pollutants. Note, however, that although greenhouse gas emissions by an industrial building are much more than mobile sources, their effect is less, since the number of industrial buildings in cities is negligible compared to the number of cars (approximately 1/1000). On the other hand, industrial units stop generating pollutants when they are closed during holidays and weekends whereas cars never stop moving and generating greenhouse gases.
As shown by the above figures, greenhouse gas emissions by industrial units are considerably lower when all major and critical variables are at their optimal levels. Those results confirm the importance of optimizing all major and critical variables to drastically reduce greenhouse gas emissions. It is worth noting, however, that whereas certain variables such as EMS can be optimized using a set of strategies and proper planning, others, such as EIB and ENB, cannot in a well-established urban setting.

Figure 13. Comparing the current and optimal industrial buildings emission rate.

Figure 14. The effect of industrial building emission rate on the amount of greenhouse gases.

Figure 15 depicts the amount of greenhouse gas emissions provided by the model when all major and critical variables are at their optimal levels.

Figure 15. The greenhouse gases status when all critical variables are optimized.

The above figure shows a huge drop in greenhouse gas emissions when the major and critical variables are at their optimum values. Those results confirm the importance of optimizing all major and critical variables to drastically reduce greenhouse gas emissions. It is worth noting, however, that whereas certain variables such as EMS can be optimized using a set of strategies and proper planning, others, such as EIB and ENB, cannot in a well-established urban setting.

4. Conclusions

This paper proposed a simulation model of the air pollution levels in the capital city of Tehran. The model considered air pollution sources and outcomes based on system
dynamics. Its aim was to devise strategies to mitigate air pollution for the city of Tehran. The dynamics analysis and obtained results showed that the leading cause of pollution in cities is greenhouse gas emissions. Thus, the best strategy to reduce air pollution is to reduce greenhouse gas emissions by cars since first, cars are traveling on city roads, second, the most fossil fuels in the cities are consumed by the cars, and third, they are inside the cities whereas factories are located outside. Hence, manufacturing environmentally friendly cars which can rely on renewable energy resources as their fuels can provide a viable solution to reduce air pollutants. The other highly important strategy is to preserve forests which are the most valuable resources after water and air, they are countries’ wealth for a better life. Additionally, strategies aiming at reducing pollution by industrial units should also be considered. It should be noted that avoiding environmental pollution is better and cheaper than cleaning up or mitigating the impact of pollution.

Future works can focus on developing similar models based on data collected from other countries and draw global solutions to the pollution problem.

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**Appendix A**

**Table A1. Sources of collected data.**

| Variable | Source | Website Address and References |
|----------|--------|---------------------------------|
| Cars     | Analytical website of Asr-e Iran, 1394 (2015) | www.asriran.com [26] (accessed on 20 October 2021). |
| Rate     | Tehran statistical yearbook (Statistics website) | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Non-industrial buildings | Statistics inference of Tehran municipality 2007–2011 mean 25 | www.tehran.ir [28] (accessed on 20 October 2021). |
| Non-industrial licensing rate | Tehran municipality website | www.tehran.ir [28] (accessed on 20 October 2021). |
| Destruction rate | Question from expert | - |
| Industrial buildings | Tehran Industrial, Mining, and Trade Organization | www.teh.mimt.gov.ir [29] (accessed on 20 October 2021). |
| Industrial licensing rate | Tehran municipal website | www.tehran.ir [28] (accessed on 20 October 2021). |
| GG + VOCs + CO + NOx + Sox + PM | Tehran pollutant emission inventory 2013 | [25] |
| Emission rate per cars | Based on emission inventory and the number of cars in 2013 | [25] |
| Emission rate per industrial buildings | Based on emission inventory and number of industrial buildings, 2013 | [25] |
### Table A1. Cont.

| Variable | Source | Website Address and References |
|----------|--------|--------------------------------|
| Emission rate per non-industrial buildings | Based on emission inventory and number of non-industrial buildings, 2013 | [25] |
| Forests | Tehran statistics yearbook | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Afforestation rate | Tehran statistics yearbook | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Forest fires | Tehran statistics yearbook | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Cleaning capacity of forests | Iran’s students news agency (ISNA) | www.gilan.isna.ir [30] (accessed on 20 October 2021). |
| Rain | Tehran statistics yearbook, geophysical station | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Cleaning capacity of rain | Based on the statistics of daily emission inventory | [25] |
| Population | Tehran statistics yearbook | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Birth rate | Tehran statistics yearbook | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Natural death | Tehran statistics yearbook | www.amar.org.ir [27] (accessed on 20 October 2021). |
| Dying from pollution | Islamic republic of Iran’s news agency (IRNA), compared to pollution in 2013 | www.irna.ir [31] |

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