

Article

Mathematical Modeling of Photochemical and Chemical Interactions in Photochemical Smog Formation

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Abstract: Atmospheric pollution results from toxic gases in low concentrations, originating from natural processes and human activities. These gases interact with each other in the presence of solar radiation, forming much more complex compounds that contribute to the formation of photochemical smog. This study presents a mathematical model to estimate the daily concentrations of primary and secondary pollutants, assuming that spatial variation is not considered within a control volume. The model includes nitrogen oxides, ozone, hydrocarbons, aldehydes, alcohols, and other gases, which are related through 52 chemical and photochemical reactions with rate constants that depend on factors such as the time of day and temperature. The model formulation results in 31 ordinary differential equations that are solved using a variable-step algorithm in MATLAB R2019a. Two scenarios are simulated: the “closed-box” model (CBM), where there are no inflows or outflows of gaseous flux, and the “open-box” model (OBM), which includes inflows and outflows within the control volume. The OBM is particularly useful for predicting concentrations during thermal inversion episodes. The results show that several pollutants reach their maximum concentrations at midday, suggesting an increase in the formation of secondary pollutants under high solar radiation, especially in the closed-box model. In the open-box model, concentration peaks shift toward the afternoon. To compare both models, the closed-box system conditions are considered, incorporating airflow into the open-box model without accounting for pollutants transported by this flow. The complex nonlinear dynamics observed in the pollutants highlight the combined influence of solar radiation, temperature, and emission rates on air quality. This study underscores the usefulness of mathematical models in developing effective mitigation strategies and assessing environmental and public health impacts.

Keywords: modeling; troposphere; pollutants; air; photochemical

1. Introduction

The lower atmosphere, known as the troposphere, is a continuous source of pollution due to the emission of primary pollutants such as nitrogen oxides (NO_x), carbon monoxide (CO), and volatile organic compounds (VOCs). These substances originate naturally and from human activities, including vehicular traffic, industry, and the use of fossil fuels; a study by the National University of San Marcos, Lima, Peru, indicates that transportation contributes 70% of carbon monoxide emissions and 40% of nitrogen oxides emissions in the city. Studies conducted by the Technological University of Peru and the Private University of the North report that the paint industry and fuel marketing centers are sources of volatile organic compounds. Under favorable environmental conditions and solar radiation, these compounds can transform into a series of secondary pollutants, among which tropospheric ozone (O_3) stands out due to its adverse effects on human health and ecosystems [1].

Air pollution is one of the most pressing environmental issues globally. It has significant impacts on climate change and public health, increasing both morbidity and mortality rates [2]. For this reason, the reaction mechanisms proposed in the literature have focused their analyses on processes primarily involving nitrogen oxides, VOCs, and ozone, among others. For the determination of some primary pollutants such as hydrocarbons and soot particles, Raman spectroscopy is used, standing out for its precision and simultaneous analysis capability [3–5]. Using adsorption and oxidation techniques, a high percentage of NO_x and SO_2 can be removed from a gas mixture [6,7].

Tropospheric ozone, a secondary pollutant, is primarily formed through photochemical reactions in the presence of NO_x and VOCs. Its concentration is strongly correlated with these precursors and other meteorological factors such as solar radiation, temperature, and wind speed [8]. This pollutant is responsible for an increased incidence of cardiovascular and respiratory diseases and negatively impacts agricultural crops and natural ecosystems [9]. Tropospheric ozone responds almost directly to changes in atmospheric conditions and the concentrations of its precursors. Forecasting models have been developed to predict its concentrations within 24 h, enabling more precise planning for mitigating its effects [10]. A study on climate change and future tropospheric ozone concentrations in the Northern Hemisphere highlighted an increase in ozone exchange between the stratosphere and troposphere (STE) in Europe and eastern China, reinforcing the need for appropriate environmental policies [11]. Another study identified temperature as the dominant factor, explaining approximately 40% of annual ozone concentrations, and emphasized the interaction between multiple meteorological factors that significantly impact ozone levels, especially during summer [12].

Between 2020 and 2022, the impact of tropospheric ozone on 20 rice varieties in Bangladesh was evaluated. The use of ethylenedurea (EDU) increased grain yield by an average of 10.4%, highlighting the negative effect of ozone on agricultural production [13]. During the COVID-19 pandemic, an improvement in air quality was observed, along with a positive correlation between NO_2 levels and COVID-19 cases, suggesting that pollution could increase the risk of infection and mortality [14]. The reduction in ozone-depleting substances has decreased ultraviolet (UV) radiation at the surface, improving air quality; however, climate change may alter these effects in the future [15,16].

Urban air quality has been studied with a focus on ozone (O_3) concentration and its formation, influenced by both photochemical processes and the transport of precursors such as NO_x from industrial areas. Despite reductions in NO_x and volatile organic compounds, O_3 levels have remained high, indicating a nonlinear relationship with its precursors and the need for more effective mitigation strategies [17].

In northern China and the Pearl River Delta, under specific meteorological conditions such as static stability and the diurnal cycle of the mixing layer, an explosive increase

in O_3 was observed, highlighting the complex interaction between meteorology and air quality [18]. Globally, tropospheric ozone has increased since the 20th century, driven by anthropogenic factors and meteorological conditions such as anticyclones and sea breezes, which affect its transport and dispersion. This suggests that reducing NO_x does not always guarantee a decrease in O_3 , emphasizing the need for targeted policies for its control [19,20]. Furthermore, photochemical smog, composed of ozone and other secondary pollutants such as acids and peroxides, has been identified as primarily originating from nitrogen oxides and volatile organic compounds [21].

Other pollutants hazardous to human health and the environment include NO_2 and SO_2 , whose increase is attributed to fires, industrial activity, and other causes [22]. In areas with low industrial activity, the concentration of these two pollutants is significantly lower [23].

Mathematical modeling is widely applied to assess the impact of air pollutants on human and ecological health, considering factors such as emission sources (both natural and anthropogenic), meteorology (topography, temperature, wind currents), and chemical processes (reactions, aerosol formation, sinks) [2,24–26]. Various studies have compared numerical schemes for predicting air quality in urban areas. For example, a study on the cities of Los Angeles and Philadelphia found that the Zalesak method predicts ozone peaks more accurately than the SHASTA method in Los Angeles, while in Philadelphia, the differences are minimal [27]. Additionally, the modular SSH-aerosol model is presented, which simulates the evolution of gases and aerosols, improving prediction accuracy in urban areas through the use of computational fluid dynamics (CFD) software such as OpenFOAM [28].

Several mathematical modeling strategies were reviewed, including Gaussian, Lagrangian, and Eulerian models, together with CFD, highlighting their advantages and disadvantages along with advancements in parallel computing and adaptive mesh refinement [24,29–31]. Furthermore, an analytical solution for atmospheric pollutant dispersion was introduced using a fractional approach, which outperforms traditional solutions by considering memory effects in turbulence [32]. A study developed a mathematical model for industrial emission dispersion, incorporating factors such as the deposition velocity of fine particles and utilizing a second-order implicit finite difference scheme, demonstrating high accuracy when compared with field data [33]. The main limitation of these models is their reliance on Gaussian dispersion schemes, which omit chemical and photochemical reactions. Including such reactions would result in an immense number of differential equations, leading to implementation and convergence challenges.

One of the challenges before simulating pollutant dispersion is standardizing data for second-order reaction rate constants, as they are reported in various units. Atkinson et al. proposed a conversion table to address this issue and presented kinetic data on second- and third-order chemical and photochemical reactions [34]. Key mechanisms involved in these reactions include nitrogen oxides, ozone, atomic oxygen, hydrogen peroxide, and radicals such as hydroperoxide and hydroxyl [35,36].

A study conducted in India analyzed the temporal concentrations of gases such as NO_2 , $HCHO$, and $CHOCHO$ from January 2017 to December 2020, showing diurnal peaks for NO_2 and $HCHO$, while $CHOCHO$ reached its maximum at midday [37]. Additionally, an analytical solution for pollutant dispersion was presented, based on a three-dimensional, fractional-order advection-diffusion equation, which demonstrated better parameterization compared to Gaussian models, emphasizing the superiority of fractional-order derivatives in predictions [32].

A study on pollution in the Beijing–Tianjin–Hebei region analyzed haze control using a NO_2 diffusion model based on the Gaussian plume. It concluded that industrial emissions were the primary source of NO_2 , with higher concentrations near emission sources, and that atmospheric stability was greater in autumn and winter, which is crucial for formu-

lating effective environmental regulations [38]. In Tulsa, ozone modeling was conducted using neural networks, improving accuracy by incorporating past meteorological ozone data, achieving a correlation of 0.88 with atmospheric data, making it suitable for use in automated forecasting systems [39].

The mechanism and transformation of primary pollutants such as nitrogen oxides (NO_x), volatile organic compounds (VOCs), and sulfur dioxide (SO_2) into secondary pollutants like ozone (O_3) and aerosols, as well as the formation of photochemical smog and tropospheric chemistry, are explained through chemical and photochemical reactions in the troposphere [40]. The transformation of primary pollutants into secondary pollutants such as ozone and aerosols through sunlight-driven reactions is further detailed in studies [41,42].

The WRF-Chem (Weather Research and Forecasting with Chemistry) model integrates meteorological components with chemical and photochemical processes in the atmosphere. It is used to simulate the formation and dispersion of secondary pollutants at regional and global scales. The model is widely applied in research on tropospheric ozone, aerosols, and other secondary pollutants [43]. The EMEP (European Monitoring and Evaluation Programme) employs atmospheric chemistry models to study primary and secondary pollutants at the regional level in Europe. EMEP models simulate chemical and photochemical processes, particularly those affecting the formation of ozone and aerosols. They focus on the interaction between pollutants emitted from human sources and their transformations in the atmosphere [44].

The GEOS-Chem model is a global atmospheric chemistry model that simulates the formation and distribution of primary and secondary pollutants. It is widely used to study the effects of photochemical chemistry on ozone and aerosol formation, including the interaction between NO_x and VOCs in the production of tropospheric ozone and fine particulate matter (PM2.5) [45].

The CAMx (Comprehensive Air Quality Model with Extensions) is an air quality model that incorporates both chemical and photochemical processes. It predicts the concentrations of primary and secondary pollutants, such as ozone and aerosols, and is used for regional-scale studies to address air pollution issues [46].

Kinetic data on the reaction rate constants for each component involved in photochemical smog formation are presented in some cases as a function of temperature, while in others, they are provided independently of temperature for first-, second-, and third-order kinetics [34].

A simplified resolution technique proposes a system of 15 differential equations corresponding to NO , NO_2 , O_3 , $HCHO$, $RCHO$, OLE , ALK , ARO , C_2H_4 , CO , H_2O_2 , PAN , HNO_2 , $RONO$, and RO_2NO_2 . Assuming a steady-state process, it also includes nine nonlinear algebraic equations for O , RO^* , HO^* , NO_3 , RO_2^* , HO_2^* , RCO_3^* , N_2O_5 , and HO_2NO_2 . Other components such as HCO^* , $RONO_2$, $HONO_2$, H_2 , CO_2 , H_2O , and O_2 are not considered in the mathematical development [47].

2. Materials and Methods

Modeling involves identifying, from the relevant literature, the reaction mechanisms of the most common pollutants and their corresponding kinetic constants; one of these mechanisms considers the interaction of 32 chemical species that generate 52 chemical and photochemical reactions. It includes formulating unsteady-state mass balance equations for each component, selecting a transport model coupled with meteorological conditions, and making reasonable assumptions regarding initial conditions and possible emission rates of primary pollutants. While most traditional models are based on the representation of dispersion using Gaussian profiles, the present work adopts the concept of a control volume under an Eulerian approach. This formulation allows a coupled description of

transport processes and chemical and photochemical reactions, facilitating their integration into local mass balances within the fluid domain. These elements are fundamental to the present study.

2.1. Mathematical Model

The mathematical modeling framework for the chemical and photochemical processes in the “closed-box” model is presented, focusing on a period of thermal inversion. In this scenario, kinetic parameters are required, including indirect solar radiation for photochemical reactions, along with initial concentrations and emission rates.

For the “open-box” model, under the same conditions as the closed-box model, additional parameters must be considered, such as altitude above sea level and thermal inversion height (stable atmosphere), which allow for the calculation of density and pressure. Using knowledge of the geometric dimensions of the control volume, the air inflow and outflow can be determined, along with the respective pollutant concentrations.

Figure 1 illustrates the corresponding modeling scheme for the chemical and photochemical interactions of tropospheric pollutants, considering mixing within the control volume. This volume is bounded at the top by the thermal inversion layer, while its width and length are determined by geographical conditions or arbitrarily set limits.

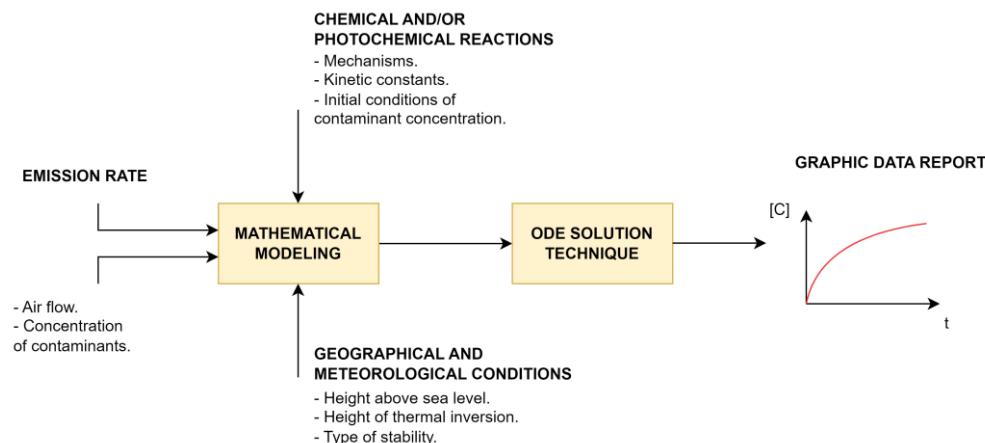


Figure 1. Modeling the process of tropospheric pollutant dynamics in an open-box model.

2.2. Reaction Mechanism

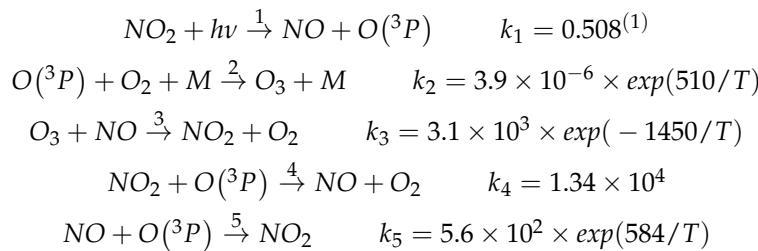
In the literature, various reaction mechanisms are found, ranging from those composed of a few chemical and photochemical reactions to highly complex mechanisms involving a series of chemical substances. Bazzel reports a reaction mechanism consisting of 7 photochemical reactions involving nitrogen oxides, ozone, nitrous acid, and formaldehyde, among others, whose constants depend on the solar zenith angle [48]; mechanisms for the formation and chemical and photochemical decomposition of bromine compounds have been proposed to understand the distribution of $CHBr_3$ as a function of the atmospheric column height [49]. A mechanism consisting of 65 chemical and photochemical reactions is described by Hampson and Garvin, 1977 [50], which involves the chemical and photochemical interactions of primary and secondary pollutants. A reaction mechanism consisting of 217 chemical equations is presented by Leone and Seinfeld, 1985 [51]. Another simplified mechanism consisting of 14 chemical and photochemical equations involves air as an energy dissipator and is present in the formation of ozone from monoatomic oxygen and diatomic oxygen [52].

McRae presents kinetic data of a complex mechanism of chemical and photochemical interactions in the atmosphere, composed of a system of 52 chemical equations, whose resolution applies a hybrid method of 15 ordinary differential equations and 9 nonlinear

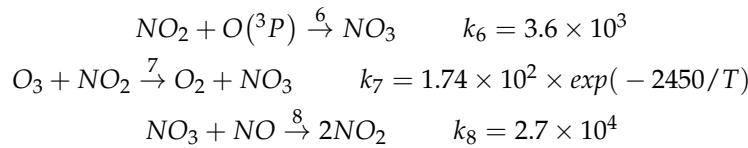
algebraic equations; it does not take into account 7 components, including water, oxygen, carbon dioxide, hydrogen, monoatomic oxygen in its various forms, the formyl radical, and M (third body) [47,53].

The kinetic model from the literature [47,53,54] is presented below, along with the rate constants of the chemical and photochemical reactions.

1. Photolysis of NO_2 and basic NO, NO_2, O_3 photolytic cycle



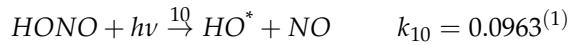
2. Chemistry of nitrogen trioxide: NO_3



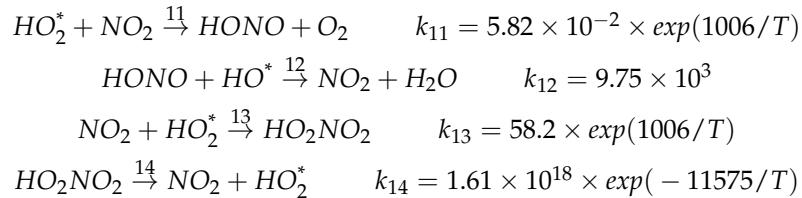
3. Nitrous acid and peroxy nitrous acid chemistry



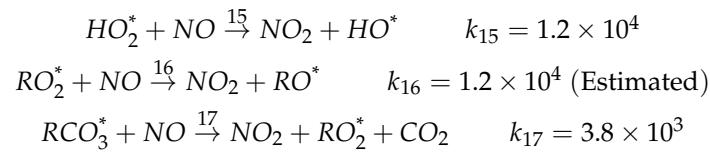
4. Photolysis of nitrous acid: $HONO$



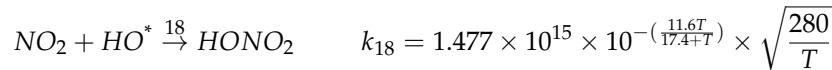
5. Nitrous acid chemistry



6. Conversion of NO to NO_2



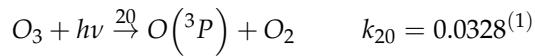
7. Nitric acid formation: $HONO_2$



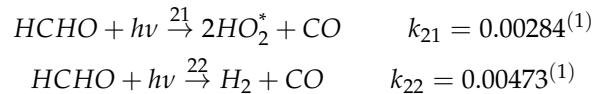
8. Hydroperoxyl radical formation



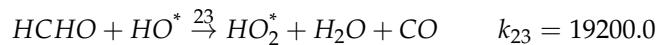
9. Photolysis of ozone



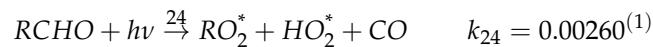
10. Photolysis of formaldehyde



11. Chemistry of formaldehyde



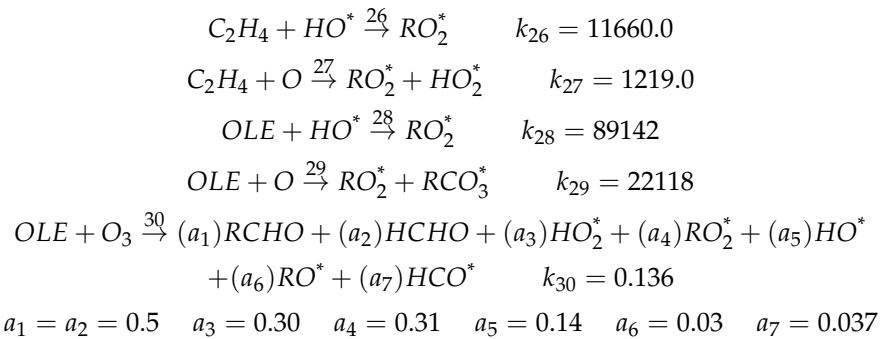
12. Photolysis of higher aldehydes



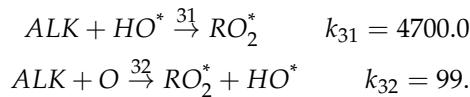
13. Higher aldehyde chemistry



14. Olefin chemistry: *OLE*



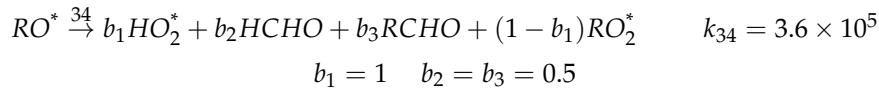
15. Alkane chemistry: *ALK*



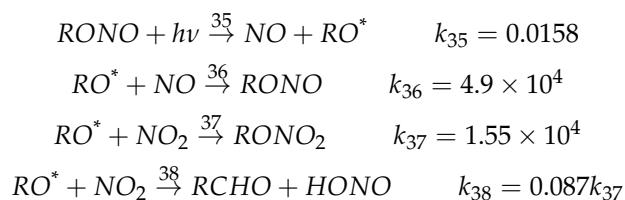
16. Aromatic chemistry



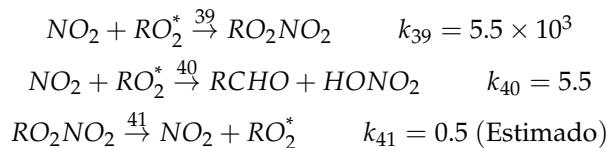
17. Alkoxy radical chemistry



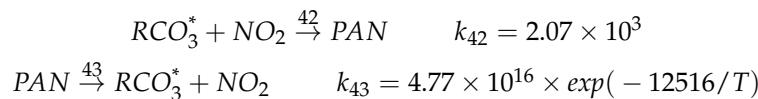
18. Photolysis and chemistry of RONO



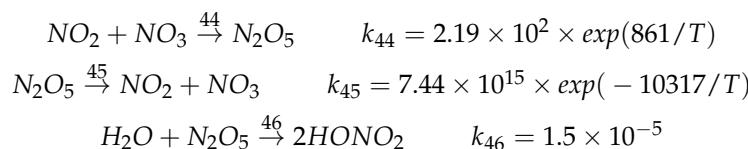
19. Peroxy nitrate chemistry



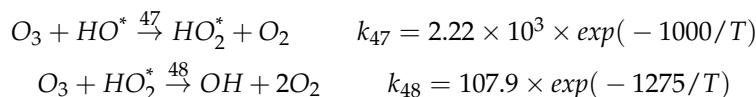
20. Peroxyacetyl nitrate chemistry: PAN



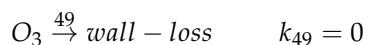
21. Dinitrogen pentoxide chemistry



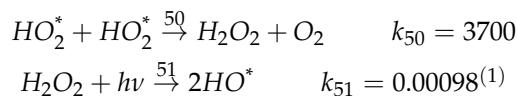
22. Ozone removal steps



23. Ozone wall loss for smog chamber experiments



24. Hydrogen peroxide production and photolysis



25. Recombination reaction for peroxyalkyl radicals



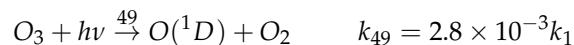
Most reactions are second order ($n = 2$). Reactions 1, 10, 14, 20, 21, 22, 24, 34, 35, 41, 43, 45, and 49 are first order ($n = 1$), while reaction 2 is third order ($n = 3$). In general, the units are expressed in $(\text{ppm}^{1-n} \text{ min}^{-1})$. Additionally, the reactions associated with Equations (1), (10), (20)–(22), (24), (35), and (51) correspond to photochemical processes,

while the remaining set of equations represents conventional chemical reactions that do not require activation by solar radiation.

$$k_{11} = 0.001k_{13}$$

$$\frac{k_{37}}{k_{37} + k_{38}} = 0.92 \text{ para } CH_3 \quad k_{38} = 0.087k_{37}$$

The O_3 value in smog chambers depends on experimental conditions. On the other hand, Leone and Seinfeld propose the following photochemical equation for ozone decomposition by solar radiation [51].



The rate constant for reactions 39 and 40 are based on the assumption that

$$\frac{k_{16}}{k_{39} + k_{40}} = 2.2$$

The photochemical reaction rate constants are obtained at the time of maximum solar radiation (12:00 hours). For the remaining hours between 06:00 and 18:00, the transformation equation shown in the work of Cutlip and Shacham [55] should be used.

If $k_i = if(sign(\sin(2\pi/24 \times (t - 6))) \leq 0)$, then $k_i = 0$; otherwise, it takes the following values: $k_i = (k_{max} \times 60) \times (\sin(2 \times \pi/24 \times (t - 6)))$, where sign is the sign function, which takes only three values +1, 0, -1, depending on the argument value. The net result is that the rate constant has values only between 06:00 and 18:00 hours and is zero for the other hours. The factor 60 converts the constant expressed in min^{-1} to h^{-1} .

The respective nomenclature of the molecules and radicals formed during the chemical and photochemical interactions is shown in Table 1.

Table 1. Nomenclature of chemical species.

Formula	Name	Formula	Name
NO	Nitric Oxide	NO ₂	Nitrogen Dioxide
NO ₃	Nitrogen Trioxide	O ₃	Ozone
HCHO	Formaldehyde	RCHO	Higher Aldehydes
OLE	Lumped Olefins	ALK	Lumped Alkanes
ARO	Lumped Aromatics	C ₂ H ₄	Ethylene
CO	Carbon Monoxide	H ₂ O ₂	Hydrogen Peroxide
PAN	Peroxyacetyl Nitrate	HONO	Nitrous Acid
RONO	Alkyl Nitrite	RO ₂ NO ₂	Peroxyalkyl Nitrate
N ₂ O ₅	Pentoxide Dinitrogen	HO ₂ NO ₂	Peroxynitric Acid
RONO ₂	Alkyl Nitrate	HONO ₂	Nitric Acid
H ₂	Hydrogen	CO ₂	Carbon Dioxide
H ₂ O	Water	O ₂	Oxygen
(O, O(^1D), O(^3P)) ⁽¹⁾	Monatomic oxygen, singlet and triplet	M	Third Body
RCO ₃ [*]	Peroxyacetyl Radical	HO ₂ [*]	Hydroperoxy Radical
RO ₂ [*]	Alkylperoxy Radical	HO [*]	Hydroxyl Radical
RO [*]	Alcoxyl Radical	HCO [*]	Formyl Radical

⁽¹⁾ In the system, the formulation of the ordinary differential equation system considers the three forms of oxygen as a single entity.

Figure 2 schematically illustrates the different relationships between reactants and reaction products.

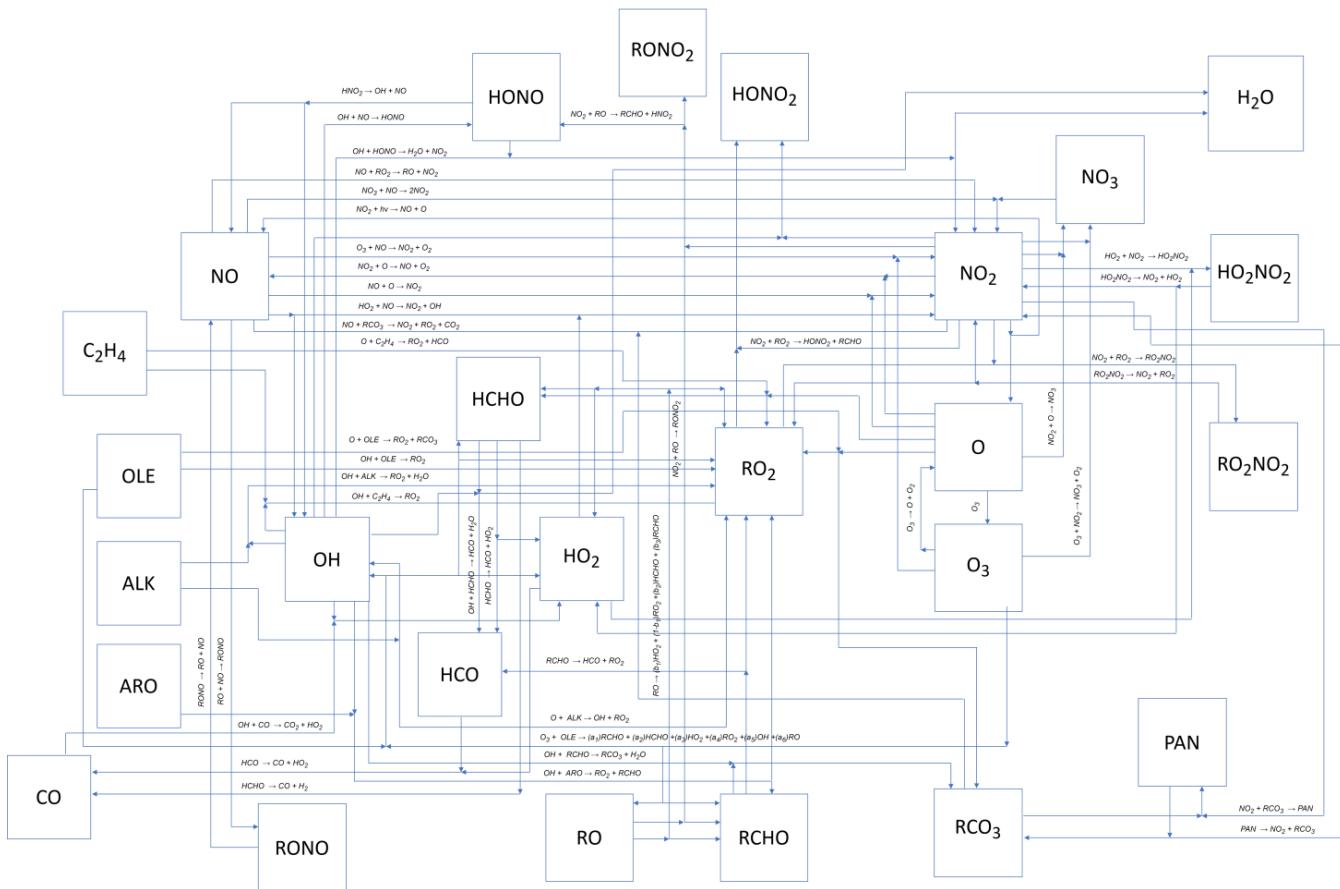


Figure 2. Photochemical smog formation mechanism consisting of 52 chemical and photochemical [47].

2.3. Mass Transfer Equations Derived from the Reaction Mechanism

The 52 chemical and photochemical reactions described generate 31 differential equations, which allow for calculating the temporal concentrations of the most stable pollutants and short-lived species such as free radicals.

The general form of the global mass balance is given by:

$$\frac{\partial}{\partial t} \iiint_v C_i dv = - \iint_s C_i (V \cdot n) ds + \iiint_v r_i dv + E_i \quad (1)$$

whose integration, after applying Gauss's Theorem, is:

$$\frac{dC_i}{dt} = \frac{(C_{i,E} - C_i)}{\tau} + r_i + E_i \quad (2)$$

whose simplification leads to:

$$\frac{dC_i}{dt} = F_i + r_i + E_i \quad (3)$$

Equation (3) is used in the closed-box model by setting $F_i = 0$ and either considering or ignoring the emission rate.

$C_{i,E}$: Incoming concentration of pollutant i in the control volume, ppm

C_i : Temporal concentration of pollutant i , ppmv

r_i : Reaction rate of component i , ppmv/h

E_i : Emission rate of pollutant i , ppmv/h

F_i : Net volumetric flow of i , ppmv/h

τ : Residence time, h

The inflow and/or emission of pollutants is considered only for the most important contaminants, such as nitric oxide, alkanes, alcohols, aldehydes, acetylene, etc., while the outflow term is applied to all components. Since oxygen, water vapor, and carbon dioxide are the major constituents, their inflow and outflow terms are nearly equal, making $F_i = 0$ of these components.

$$\frac{dNO}{dt} = F_{NO} + R_1 - R_3 + R_4 - R_5 - R_8 - R_9 + R_{10} - R_{15} - R_{16} - R_{17} + R_{35} - R_{36} + E_{NO} \quad (4)$$

$$\begin{aligned} \frac{dNO_2}{dt} = & F_{NO_2} - R_1 + R_3 - R_4 + R_5 - R_6 - R_7 + 2 \times R_8 - R_{11} + R_{12} - R_{13} + R_{14} + R_{15} \\ & + R_{16} + R_{17} - R_{18} - R_{37} - R_{38} - R_{39} - R_{40} + R_{41} - R_{42} + R_{43} - R_{44} + R_{45} + E_{NO_2} \end{aligned} \quad (5)$$

$$\frac{dNO_3}{dt} = F_{NO_3} + R_6 + R_7 - R_8 - R_{44} + R_{45} + E_{NO_3} \quad (6)$$

$$\frac{dN_2O_5}{dt} = F_{N_2O_5} + R_{44} - R_{45} - R_{46} + E_{N_2O_5} \quad (7)$$

$$\frac{dHO_2NO_2}{dt} = F_{HO_2NO_2} + R_{13} - R_{14} + E_{HO_2NO_2} \quad (8)$$

$$\frac{dHONO_2}{dt} = F_{HONO_2} + R_{18} + R_{40} + R_{46} + E_{HONO_2} \quad (9)$$

$$\frac{dO_3}{dt} = F_{O_3} + R_2 - R_3 - R_7 - R_{20} - R_{30} - R_{47} - R_{48} - R_{49} + E_{O_3} \quad (10)$$

$$\frac{dHCHO}{dt} = F_{HCHO} + R_{21} - R_{22} - R_{23} + a_2 \times R_{30} + b_2 \times R_{34} + E_{HCHO} \quad (11)$$

$$\frac{dRCHO}{dt} = F_{RCHO} - R_{24} - R_{25} + a_1 \times R_{30} + R_{33} + a_3 \times R_{34} + R_{38} + R_{40} + E_{RCHO} \quad (12)$$

$$\frac{dOLE}{dt} = F_{OLE} - R_{28} - R_{29} - R_{30} + E_{OLE} \quad (13)$$

$$\frac{dALK}{dt} = F_{ALK} - R_{31} - R_{32} + E_{ALK} \quad (14)$$

$$\frac{dARO}{dt} = F_{ARO} - R_{33} + E_{ARO} \quad (15)$$

$$\frac{dC_2H_4}{dt} = F_{C_2H_4} - R_{26} - R_{27} + E_{C_2H_4} \quad (16)$$

$$\frac{dCO}{dt} = F_{CO} - R_{19} + R_{21} + R_{22} + R_{23} + R_{24} + E_{CO} \quad (17)$$

$$\frac{dH_2O_2}{dt} = F_{H_2O_2} + R_{50} - R_{51} + E_{H_2O_2} \quad (18)$$

$$\frac{dPAN}{dt} = F_{PAN} + R_{42} - R_{43} + E_{PAN} \quad (19)$$

$$\frac{dHONO}{dt} = F_{HONO} + R_9 - R_{10} + R_{11} - R_{12} + R_{38} + E_{HONO} \quad (20)$$

$$\frac{dRONO}{dt} = F_{RONO} - R_{35} + R_{36} + E_{RONO} \quad (21)$$

$$\frac{dRO_2NO_2}{dt} = F_{RO_2NO_2} + R_{39} - R_{41} + E_{O_2NO_2} \quad (22)$$

$$\frac{dRONO_2}{dt} = F_{RONO_2} + R_{37} + E_{RONO_2} \quad (23)$$

$$\frac{dH_2}{dt} = F_{H_2} + R_{22} + E_{H_2} \quad (24)$$

$$\frac{dO}{dt} = F_O + R_1 - R_2 - R_4 - R_5 - R_6 + R_{20} - R_{27} - R_{29} - R_{32} + R_{49} + E_o \quad (25)$$

In (25), all forms of oxygen are included: monoatomic oxygen, triplet oxygen, and singlet oxygen.

$$\frac{dRO^*}{dt} = F_{O^*} + R_{16} + a_6 \times R_{30} - R_{34} + R_{35} - R_{36} - R_{37} - R_{38} + 2R_{52} \quad (26)$$

$$\begin{aligned} \frac{dHO^*}{dt} = & F_{HO^*} - R_9 + R_{10} - R_{12} + R_{15} - R_{18} - R_{19} - R_{23} - R_{25} - R_{28} + a_5 \times R_{30} \\ & - R_{31} + R_{32} - R_{33} - R_{47} + R_{48} + 2R_{51} \end{aligned} \quad (27)$$

$$\begin{aligned} \frac{dRO_2^*}{dt} = & F_{RO_2^*} - R_{16} + R_{17} + R_{24} + R_{26} + R_{27} + R_{28} + R_{29} + a_4 \times R_{30} + R_{31} + R_{32} + R_{33} \\ & + (1 - b_1) \times MM \times R_{34} - R_{39} - R_{40} + R_{41} - 2R_{52} \end{aligned} \quad (28)$$

$$\begin{aligned} \frac{dHO_2^*}{dt} = & F_{HO_2^*} - R_{11} - R_{13} + R_{14} - R_{15} + R_{19} + 2R_{21} + R_{23} + R_{24} + R_{27} + a_3 \times R_{30} \\ & + b_1 \times R_{34} + R_{47} - R_{48} - 2R_{50} \end{aligned} \quad (29)$$

$$\frac{dRCO_3^*}{dt} = F_{RCO_3^*} - R_{17} + R_{25} + R_{29} - R_{42} + R_{43} \quad (30)$$

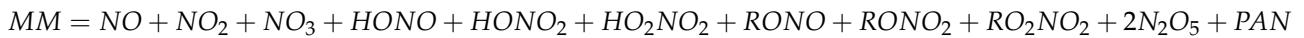
$$\frac{dHCO^*}{dt} = F_{HCO^*} + R_{30} \quad (31)$$

$$\frac{dO_2}{dt} = -R_2 + R_3 + R_4 + R_7 + R_{11} + R_{20} + R_{47} + 2R_{48} + R_{50} \quad (32)$$

$$\frac{dCO_2}{dt} = R_{17} + R_{19} \quad (33)$$

$$\frac{dH_2O}{dt} = R_{12} + R_{23} - R_{46} \quad (34)$$

Also considering:



The kinetic components of the convective mass transport equation are represented as a function of the kinetic constants and the respective concentrations, as shown in Table 2.

Table 2. Kinetic component of the matter conservation equation for each component.

Reaction Rate		
$R_1 = k_1 \times NO_2$	$R_2 = k_2 \times O_2 \times O \times M$	$R_3 = k_3 \times NO \times O_3$
$R_4 = k_4 \times NO_2 \times O$	$R_5 = k_5 \times NO \times O$	$R_6 = k_6 \times NO_2 \times O$
$R_7 = k_7 \times NO_2 \times O_3$	$R_8 = k_8 \times NO \times NO_3$	$R_9 = k_9 \times NO \times HO^*$
$R_{10} = k_{10} \times HONO$	$R_{11} = k_{11} \times NO_2 \times HO_2^*$	$R_{12} = k_{12} \times HONO \times HO^*$
$R_{13} = k_{13} \times NO_2 \times HO_2^*$	$R_{14} = k_{14} \times HO_2NO_2$	$R_{15} = k_{15} \times NO \times HO_2^*$
$R_{16} = k_{16} \times NO \times RO_2^*$	$R_{17} = k_{17} \times NO \times RCO_3^*$	$R_{18} = k_{18} \times NO_2 \times HO^*$
$R_{19} = k_{19} \times HO^* \times CO$	$R_{20} = k_{20} \times O_3$	$R_{21} = k_{21} \times HCHO$
$R_{22} = k_{22} \times HCHO$	$R_{23} = k_{23} \times HCHO \times HO^*$	$R_{24} = k_{24} \times RCHO$
$R_{25} = k_{25} \times RCHO \times HO^*$	$R_{26} = k_{26} \times C_2H_4 \times HO^*$	$R_{27} = k_{27} \times C_2H_4 \times O$
$R_{28} = k_{28} \times OLE \times HO^*$	$R_{29} = k_{29} \times OLE \times O$	$R_{30} = k_{30} \times O_3 \times OLE$
$R_{31} = k_{31} \times ALK \times HO^*$	$R_{32} = k_{32} \times ALK \times O$	$R_{33} = k_{32} \times ARO \times HO^*$
$R_{34} = k_{34} \times RO^*$	$R_{35} = k_{35} \times RONO$	$R_{36} = k_{36} \times NO \times RO^*$

Table 2. Cont.

Reaction Rate		
$R_{37} = k_{37} \times NO_2 \times RO^*$	$R_{38} = k_{38} \times NO_2 \times RO^*$	$R_{39} = k_{39} \times NO_2 \times RO_2^*$
$R_{40} = k_{40} \times NO_2 \times RO_2^*$	$R_{41} = k_{41} \times RO_2NO_2$	$R_{42} = k_{42} \times NO_2 \times RC\bar{O}_3^*$
$R_{43} = k_{43} \times PAN$	$R_{44} = k_{44} \times NO_2 \times NO_3$	$R_{45} = k_{45} \times N_2O_5$
$R_{46} = k_{46} \times N_2O_5 \times H_2O$	$R_{47} = k_{47} \times O_3 \times HO^*$	$R_{48} = k_{48} \times O_3 \times HO_2^*$
$R_{49} = k_{49} \times O_3$	$R_{50} = k_{50} \times HO_2^* \times HO_2^*$	$R_{51} = k_{51} \times H_2O_2$
$R_{52} = k_{52} \times RO_2^* \times RO_2^*$		

2.4. Atmospheric Parameters

The barometric pressure as a function of altitude above sea level is given by:

$$p(h) = p_0 \left[1 - \frac{Lh}{T_0} \right]^{\frac{gM_a}{RL}} \quad (35)$$

p_0 : Atmospheric pressure at sea level, 101,325 Pa

$p(h)$: Atmospheric pressure at any altitude

L : Adiabatic temperature gradient, 0.0065 K/m

T_0 : Temperature at sea level, fluctuating between 273 K and 298 K

g : Gravitational acceleration, 9.81 m/s²

M_a : Molecular mass of air, 0.02896 kg/mol

R : Universal gas constant, 8.314 J/mol K

h : Altitude above sea level, m

H : Scale height, m

z : Air column height, m

The air pressure as a function of the air column is calculated using:

$$\frac{dp}{dz} = -\rho_a g \quad (36)$$

If we assume that air behaves similarly to an ideal gas:

$$\rho_a = \frac{PM_a}{RT_0} \quad (37)$$

Replacing (37) in (36):

$$\frac{dp}{p} = -\frac{gM_a}{RT_0} dz \quad (38)$$

The integration leads to:

$$p(z) = p(h) \exp(-\frac{gM_a}{RT_0} z) \quad (39)$$

Equation (40) is known as the barometric law. Now, it is necessary to define the parameter H , given by:

$$H = \frac{RT_0}{gM_a} \quad (40)$$

Substituting (41) into (40), we obtain the equations for calculating pressure and density in an air column:

$$p(z) = p(h) \times \exp(-z/H) \quad (41)$$

Similarly, in the case of air density at a given altitude:

$$\rho(z) = \rho(h) \times \exp(-z/H) \quad (42)$$

These equations are also used to obtain average values below the thermal inversion zone. With these parameters, it is possible to calculate the residence time of pollutants in a control volume, which is given by:

$$\tau = \frac{v}{Q} = \frac{L\rho}{G} \quad (43)$$

The volume is calculated based on the geometric dimensions of the control volume:

$$v = L \times W \times z \quad (44)$$

The flow rate is calculated using

$$Q = V \times W \times z \quad (45)$$

The velocity is obtained from the mass flow velocity

$$V = G/\rho \quad (46)$$

G : Mass flow velocity, $\text{kg}/\text{m}^2\text{h}$

ρ : Gas column density, kg/m^3

V : Horizontal air velocity in the horizontal direction, m/s

L : Length of the control volume, m

W : Width of the control volume, m

v : Control volume, m^3

Q : Volumetric flow rate, m^3/h

2.5. Flow Parameters

A control volume is first established, with its boundaries determined based on the area of interest in the study. The height h is defined by the inversion layer, complemented by the length L and width W . This allows for the determination of the volumetric flow rate Q using the wind velocity V , the cross-sectional area $h \times W$, and the density of the gas mixture ρ entering the control volume. Additionally, it is possible to incorporate residence time and/or spatial velocity into the inflow and outflow terms of the gas mixture within the model.

2.6. Atmospheric Stability

Modeling requires incorporating the type of atmospheric stability. Atmospheric stability is crucial in understanding how pollutants disperse in the atmosphere. Depending on the type (stable, unstable, or neutral), the air can limit or favor the ascent and mixing of pollutants. Under stable conditions, pollutants remain trapped near the ground, affecting air quality; on the other hand, under unstable conditions, the air mixes more easily, which helps disperse pollutants. Therefore, understanding the type of stability is essential for predicting and controlling air pollution. A stable atmosphere occurs when the temperature gradient is greater than the dry adiabatic lapse rate of $9.8 \text{ }^\circ\text{C}/\text{km}$, leading to the potential occurrence of thermal inversion.

Stability A (absolute stability): No thermal inversion occurs, as the temperature decreases with altitude.

Stability B (neutral stability): No thermal inversion occurs, since the environmental temperature gradient is very close to the dry adiabatic lapse rate.

Stability C (conditional instability): Thermal inversion may occur at very low levels if the air cools rapidly, but the atmosphere remains generally prone to upward movement.

Stability D (high instability): No inversion occurs, as the atmosphere favors air ascent.

Stability E (conditional stability): Inversion may occur at higher levels if the air stabilizes, preventing further ascent.

Stability F (positive thermal slope or high-gradient stability): Thermal inversion occurs as warmer air in the upper layers prevents colder air from rising from the lower layers. The temperature gradient ranges between 1 and 3 °C per 100 m. Thermal inversion occurs below the troposphere, varying in altitude from 8 to 10 km at the poles and 16 to 18 km at the equator [56].

For modeling purposes, stability type F is considered, with an inversion height between 200 and 500 m.

2.7. Characteristics of the Developed Program

A software program has been developed in MATLAB that allows for determining the concentration of the 32 components from the 52 chemical equations of the reaction mechanism, expressed in 31 differential equations for the closed-box and open-box models.

The closed-box model resembles a batch reactor, but in addition to including the accumulation and reaction terms, it incorporates the emission rate term for some primary pollutants. This model depends on the intensity of radiation, which is expressed in terms of the kinetic constants of photochemical reactions, which in turn depend on the time of day. It also depends on the average temperature of the medium, the initial concentration of primary pollutants, and the possible consideration of the emission rate of a primary pollutant. The emission rate is estimated by the product of the emission factor (amount of pollutant emitted per unit of activity), the amount of input consumed by the activity, and the factor associated with the control systems.

The open-box model is similar to a semicontinuous reactor regarding pollutant flow but continuous regarding global air flow. The global air stream is modeled as a continuous flow under steady-state conditions, implying the existence of inflow and outflow without net accumulation within the control volume. However, the components studied here, being trace species in the gas mixture, are considered under a semicontinuous regime, thereby allowing for an accurate representation of their temporal concentration variations within the analyzed domain. This model depends on the intensity of solar radiation, expressed in terms of the kinetic constants of photochemical reactions, which in turn depend on the time of day, the temperature that influences the reaction rate of some chemical reactions in the kinetic process, and meteorology. In this open-box model, it is necessary to establish a control volume considering the height of thermal inversion, as well as the mass flow velocity. Additionally, it is possible to calculate the barometric pressure and air density at a given altitude above sea level and, from it, obtain the pressure and density of the air column from ground level to the thermal inversion height.

All these parameters allow for obtaining the residence time of pollutants in the control volume, which is incorporated into the mass transfer model given in Equation (2). Additionally, inflows of primary pollutants into the control volume, along with their emission rate, can be considered.

The algorithm used for solving the differential equations is the implicit stiff method, characterized by a variable step size due to the highly negative or very small eigenvalues of the Jacobian matrix of the equations. These eigenvalues can cause the equation system

to be solved unstably if inadequate methods are used or if the integration step is not sufficiently small [57,58].

Required data for the closed-box model:

- Reaction rate constants given in Section 2.2. Reaction Mechanism;
- Initial concentration of all pollutants in this kinetic model;
- Average temperature of the troposphere;
- Start of the simulation process;
- End of the simulation process;
- Numerical resolution algorithm for ordinary differential equations: stiff method.

Required data for the open-box model; the same data used for the closed-box model are required, in addition to the following:

- Control volume;
- Mass flow velocity;
- Altitude above sea level;
- Height of the air column (thermal inversion height);
- Density and pressure of the air column.

2.8. Typical Initial Concentration Values

NO concentration: In a city with high pollution levels, during the day, the NO concentration can range from 1 to 10 ppmv in areas near emission sources such as roads or industries, while in rural areas, it ranges between 0.01 and 0.1 ppmv [40,59,60].

NO₂ concentration: The concentration of nitrogen dioxide varies from 0.01 to 0.1 ppmv in rural areas, from 0.5 to 3 ppmv in urban areas, and in industrial or high-emission areas, the concentration can fluctuate between 5 and 20 ppmv [40,59,60].

HNO₂ concentration: The concentration of nitrous acid in the troposphere fluctuates between 0.01 ppbv in rural areas and 0.1 ppbv in high-pollution zones [40,41,61,62].

HCHO concentration: The concentration of formaldehyde varies between 0.05 and 0.1 ppmv in rural areas, 0.1 to 1 ppmv in urban areas, and during high-concentration episodes, it can reach 2 to 3 ppmv [40,41,63].

RCHO concentration: The concentration of higher aldehydes is around 0.01 ppmv in rural areas, varies between 0.01 and 0.1 ppmv in urban areas, and reaches approximately 0.1 ppmv during high-concentration episodes [40,41,63,64].

OLE concentration: The primary alcohols found in the atmosphere include methanol, ethanol, propanol, and others, with a global concentration ranging between 0.1 and 5 ppmv, depending on whether the area is rural, urban, or experiencing high-concentration episodes [40,65–68].

ALK concentration: The main alkanes present in the atmosphere include methane, whose tropospheric concentration fluctuates between 1.8 and 2 ppmv; ethane, which varies between 0.2 and 0.5 ppmv; propane, ranging from 0.05 to 0.2 ppmv; and butane, between 0.01 and 0.1 ppmv [40,69–71].

C₂H₄ concentration: The concentration of ethylene varies between 0.0001 and 0.0005 ppmv in rural areas and reaches 0.001 ppmv in urban areas with agricultural emissions [40,41,45].

CO concentration: The typical concentration of carbon monoxide in the troposphere ranges from 0.05 to 0.1 ppmv in rural areas and up to 0.1 to 1 ppmv in urban or polluted areas. During forest fire events, CO concentration can be significantly higher, reaching 10 ppmv or more [40].

H₂O₂ concentration: In rural and low-pollution areas, the typical concentration of hydrogen peroxide is generally between 0.00001 and 0.00005 ppmv. In urban or industrial areas, the concentration can be slightly higher, reaching approximately 0.0001 ppmv [40,66,67].

PAN concentration: The concentration of peroxyacetyl nitrate in rural or low-pollution areas is typically around 0.00001 to 0.00005 ppmv. In urban or industrial areas, concentrations can reach up to 0.0001 ppmv [40,67].

ARO concentration: The concentration of aromatic organic compounds mainly consists of benzene, toluene, xylene, and others. In rural or low-pollution areas, benzene concentrations can range from approximately 0.0001 to 0.0005 ppmv. In urban or highly polluted areas, such as cities or industrial zones, benzene concentrations can exceed 0.001 ppmv, with toluene and xylene levels being similar [40,72,73].

H₂ concentration: The hydrogen concentration in rural or nonpolluted areas typically ranges from 0.5 to 1 ppmv. In urban or industrial areas, hydrogen concentrations can increase to 1 to 1.5 ppmv due to emissions from human activities.

Table 3 presents the data used for the simulation. The initial concentration of each primary pollutant is based on typical data.

Table 3. Data required for the execution of the CBM and OBM models.

Component	Initial Concentration (ppm)	Emission Rate (ppm/h)		Inlet Flow (ppm/h) OBM
		CBM	OBM	
<i>NO</i>	0.01–10 (1)	0.02	0.02	0.001
<i>NO₂</i>	0.01–5 (0.05)	0.01	0.01	0.003
<i>NO₃</i>	Traces (0)	0	0	0.00001
<i>O₃</i>	Traces (0)	0	0	0
<i>HNO₂</i>	0.00001–0.0001 (0.00001)	0	0	0
<i>HO₂NO₂</i>	Traces (0)	0	0	0
<i>N₂O₅</i>	Traces (0)	0	0	0
<i>HCHO</i>	0.05–3 (0.1)	0.03	0.03	0.0001
<i>RCHO</i>	0.01–0.1 (0.02)	0	0	0
<i>OLE</i>	0.1–5 (0.2)	0	0	0.002
<i>ALK</i>	0.01–2 (1)	0.1	0.1	0
<i>C₂H₄</i>	0.0001–0.001 (0.0005)	0.0002	0.0002	0.00001
<i>CO</i>	0.05–10 (0.1)	0.02	0.02	0.00001
<i>H₂O₂</i>	0.00001–0.0001	0	0	0
<i>PAN</i>	0.00001–0.0001 (0.00005)	0	0	0
<i>ARO</i>	0.0001–0.001 (0.0005)	0.000008	0.000008	0.000004
<i>RO₂NO₂</i>	Traces (0)	0	0	0
<i>RONO₂</i>	Traces (0)	0	0	0
<i>HONO₂</i>	Traces (0)	0	0	0
<i>H₂</i>	0.5–1.5 (1)	0	0	0
<i>CO₂</i>	(420)	0	0	420
<i>H₂O</i>	10,000–20,000 (15,000)	0	0	15,000
<i>O₂</i>	(210,000)	0	0	210,000
<i>M</i>	(1,000,000)	0	0	1,000,000

The emission rate for both the closed-box model (CBM) and the open-box model (OBM) is the same, with values similar to the initial concentrations. The inlet and outlet concentrations of *CO₂*, *H₂O*, *O₂*, and *M* (air) remain unchanged and do not vary significantly due to their high concentration, which causes the first term of Equation (3) to be canceled for these compounds. For the remaining components, only the output and generation rate are considered.

On the other hand, for free radicals not included in Table 3, the emission and input terms are not considered; it is assumed that only generation and output from the control volume occur over time.

3. Results

Most of the analyzed studies focus on examining the formation and removal of tropospheric ozone through multiple chemical and photochemical reactions and its relationship with its main precursors, such as carbon monoxide, nitrogen dioxide, and other components forming what is known as photochemical smog. This study also conducts this analysis using two mass transfer models: the closed-box model and the open-box model.

3.1. Closed-Box Model

A computational program called CBM has been developed using MATLAB to solve a system of 31 ordinary differential equations. The solutions are obtained from the data presented in Table 2. The main variables involved in this process include the initial concentrations of each component in the differential equations, and the emission rates of some primary pollutants. Another relevant variable is the ambient temperature, which influences the reaction rate constants, along with solar radiation, which indirectly affects the rate constants in photochemical reactions.

The program was executed while keeping the initial concentrations and emission rates constant and considering three ambient temperature values: 278, 288, and 298 K. The results of the concentrations as a function of the time of day (from 05:00 to 19:00) present the following trend:

- Photochemical pollutants: Compounds such as O_3 , O , HO^* , HO_2^* , RCO_3^* , NO_3 , $HONO$, HO_2NO_2 , N_2O_5 , RO_2^* , RO^* , RCO_3^* , PAN , $RONO$, and $RO_2RO_2^*$ show concentration peaks between 07:30 and 14:00, followed by a decrease as the intensity of solar radiation diminishes, reaching a minimum at 18:00. After this point, chemical reactions continue without the intervention of solar radiation.
- Continuously decreasing pollutants: Other pollutants such as ALK , OLE , ARO , $HCHO$, $RCHO$, and H_2O_2 , tend to show a decrease in their concentrations because the generation rate is lower than the consumption rate. However, this behavior is nonlinear, as the concentration of each component depends on variations in the concentration of other components due to system dynamics.
- NO and NO_2 : The concentrations of NO and NO_2 exhibit inverse behavior. The NO concentration decreases while that of NO_2 increases. Both compounds directly influence the formation of tropospheric ozone, and this pattern is observed at the simulated temperatures, being the direct result of the reaction mechanism that regulates the dynamics of the relationship between these two compounds.
- Effect of temperature: The analysis of pollutants as a function of temperature was grouped into two categories:
 - Compounds with concentration peaks: For compounds such as HO_2^* , $HONO$, NO_3 , O , HO^* , RCO_3^* , RO^* , RO_2^* , $RONO$, and RO_2NO_2 , a direct relationship with temperature is observed, as their maximum concentrations increase with rising temperature. In contrast, HO_2NO_2 , N_2O_5 , O_3 , and PAN show an inverse relationship with temperature.
 - Compounds without concentration peaks: In the case of compounds such as ALK , ARO , C_2H_4 , HCO^* , and OLE , an inverse relationship with temperature is observed. For CO , H_2 , and H_2O_2 , the relationship is also inverse, while for NO , NO_2 , $RCHO$, and $RONO_2$, concentration changes with temperature are minimal.

3.2. Open-Box Model

A computational program called OBM has been developed, which uses the same initial conditions and emission rates as the CBM dispersion model. Additionally, the OBM calculation program considers the input of altitude above sea level, the height of the air

column below the inversion layer, the mass flow rate of air, and the limits of the control volume. These data are used to estimate the residence time incorporated into the kinetic model. The additional variables are shown in Table 4.

Table 4. Additional data required for the OBM model.

Variables	Data
Temperature (K)	278, 288, 298
Height above sea level (m)	0–4000 (2700)
Inversion layer height (m)	100–800 (500)
Mass air flow rate ($\text{kg}/\text{m}^2\text{h}$)	100–400 (250)
Control volume length (m)	(1000)
Control volume width (m)	(200)

The data in parentheses have been used in the execution of the program.

The execution of the OBM calculation program shows a faster tendency in the dispersion of primary pollutants. In general, a decrease in the concentrations of all chemical substances present in the model is observed, attributed to advective or convective transport, which disperses the pollutants more efficiently.

The maximum concentrations obtained in the OBM model tend to shift, with peak values occurring later in the afternoon. This phenomenon is observed, for example, in compounds such as O_3 , HO , HO_2 , RO_2^* , HO_2^* , RO_3^* , PAN , and $RO_2NO_2^*$, whose peaks are reached after 14:00. This behavior is mainly attributed to convective flow.

On the other hand, the concentrations of other components, such as ALK , OLE , ARO , among others, tend to decrease more rapidly in the OBM model compared to the CBM model due to photochemical processes and, above all, advective flow.

Regarding the effect of temperature, the concentrations of the various components show behavior similar to that of the CBM model. That is, in some cases, there is a direct relationship between concentration and temperature, while in others, the relationship is inverse. However, it is observed that the concentration levels vary between the two models, generally being lower in the OBM model.

Figure 3 presents comparative concentration graphs of the most relevant pollutants at a temperature of 288 K for the CBM and OBM models.

In the CBM, higher concentrations of primary and secondary pollutants are observed compared to the OBM. However, in some cases, concentrations are lower in the CBM, which can be attributed to the complex system dynamics and the various physical, chemical, and photochemical factors involved.

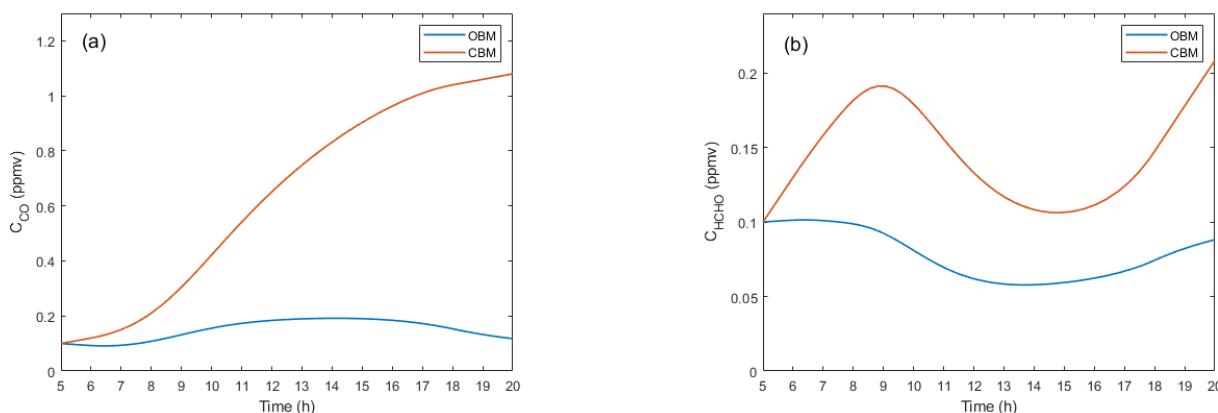
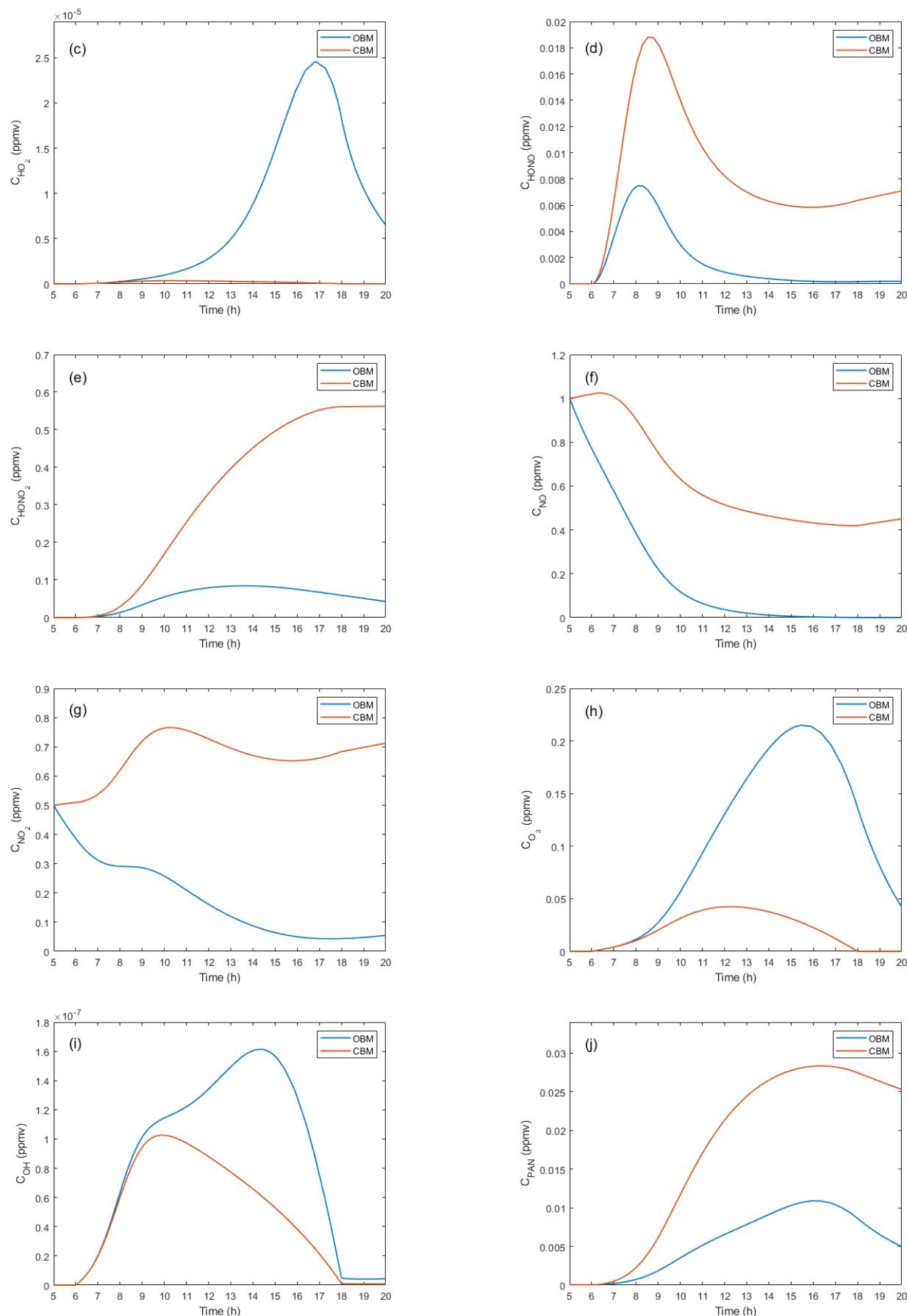


Figure 3. Cont.

**Figure 3. Cont.**

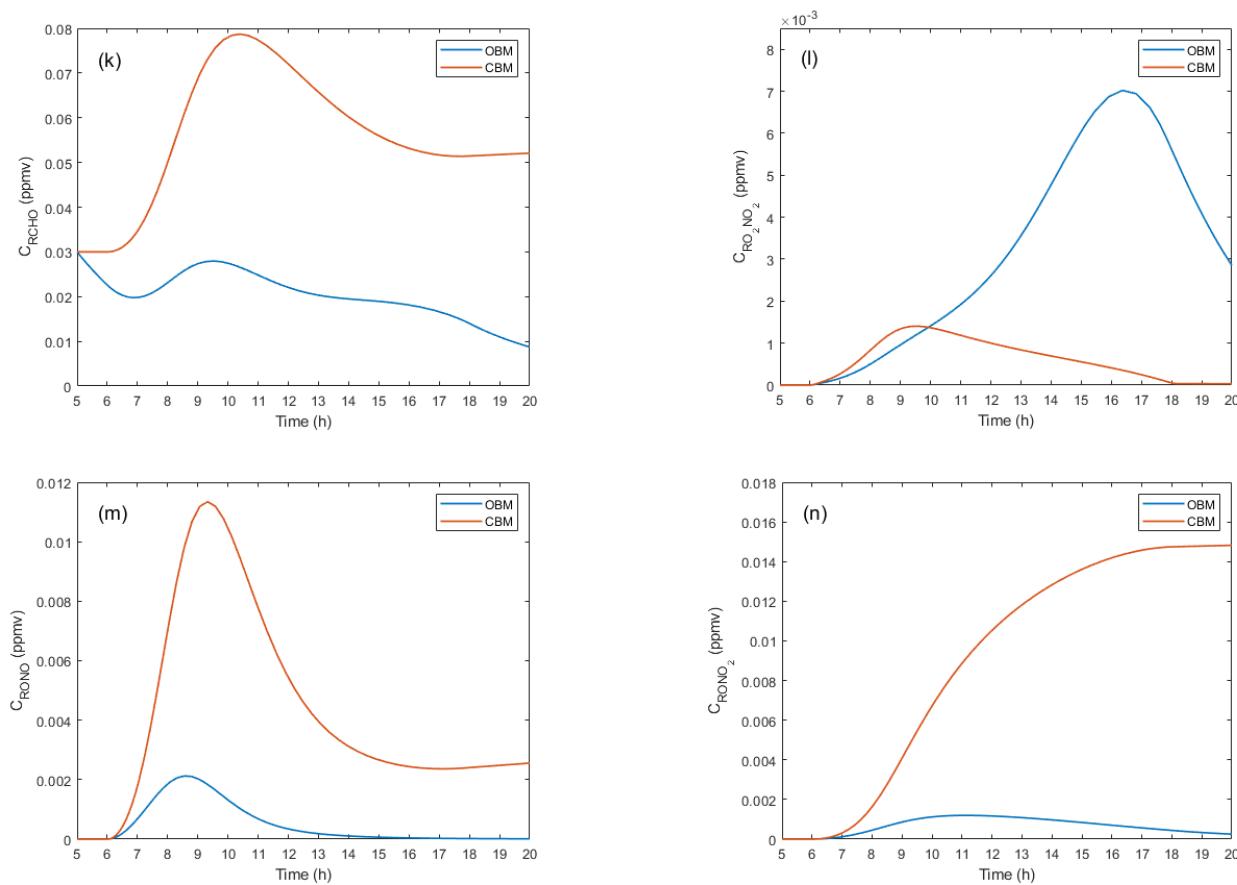


Figure 3. Concentration profiles of the main pollutants.

Figure 3 shows the concentrations of the main pollutants, from which it can be observed that for carbon monoxide (see Figure 3a), the concentration is significantly higher in the CBM than in the OBM. In the CBM, the increase in CO occurs both in the absence and presence of solar radiation, with the latter increase being much more pronounced. In the OBM, the increase in concentration is more moderate and decreases further in the afternoon hours due to the absence of solar radiation. For formaldehyde (see Figure 3b), it is found that in both models, it is generated even in the absence of solar radiation; however, its concentration is higher in the CBM. With solar radiation, concentration variations increase considerably in both cases.

On the other hand, the concentration of nitric acid (HNO_3), as shown in Figure 3e, is lower for the OBM due to the effect of the inflow, which facilitates the dispersion of the pollutant. The generation of HNO_3 depends on solar radiation, and its concentration stops increasing after 18:00 hours. Both models for nitrogen monoxide (NO) show that the concentration tends to decrease (see Figure 3f), but it is always higher in the CBM. This compound is closely related to ozone generation. For nitrogen dioxide (NO_2), the CBM maintains higher NO_2 concentrations (see Figure 3g), while in the OBM, the concentrations decrease due to the inflow of air.

In Figure 3h, the ozone (O_3) curve for the OBM shows lower concentrations with a behavior dependent on solar radiation. After 18:00 hours, no significant changes in concentrations are observed. However, for the OBM, a concentration peak is detected between 14:00 and 16:00 hours, attributable to the system dynamics. For the hydroxyl radical (OH) profile, the CBM records higher concentrations due to the conditions and dynamics of the process (see Figure 3i). In both models, a direct dependence on solar radiation is observed before 6:00 hours and after 18:00 hours. The PAN concentration is observed to be higher in the CBM (see Figure 3j), with its formation beginning with solar

radiation at 6:00 hours. In the afternoon, its concentration decreases even in the absence of solar radiation due to ongoing chemical reactions. This behavior is also observed in the OBM, although at lower levels.

The concentration of higher aldehydes for the CBM reaches a peak around 10:00 hours and subsequently decreases due to interactions with other chemical species. In the OBM, the trend is similar but with lower levels due to advective transport, as shown in Figure 3k. For peroxyacetyl nitrate, in the OBM, concentrations are higher and depend on solar radiation, especially in the morning hours (see Figure 3l). In the CBM, concentrations are lower but also dependent on solar radiation before 6:00 hours and after 18:00 hours.

Finally, the concentrations of alkyl nitrite are higher in the CBM, although both models show similar behavior (see Figure 3m). The generation of this compound depends on solar radiation, with a peak around 9:00 hours. After 15:00 hours, no significant changes are observed in either model. The concentration behavior of alkyl nitrate shows a constant increase for the CBM, which stops after 18:00 hours due to the absence of solar radiation. In the OBM, concentrations are lower due to the effect of advective transport. Both models depend on the onset of solar radiation for the formation of the compound.

4. Discussion

Kafle et al. and Tesche et al. use mathematical approaches to simulate pollutant dispersion using a modular model that integrates meteorology and chemical processes [24,27], whereas this study employs two models, CBM and OBM, with the latter being more detailed by considering factors such as altitude and airflow. The OBM model shows faster dispersion of pollutants due to the atmospheric dynamics included, compared to the CBM model. Additionally, peaks in photochemical pollutant concentrations (O_3 , HO , HO_2 , etc.) are observed during hours of maximum solar radiation. This study reports a shift in these peaks toward the afternoon due to convective flow, which is more pronounced in the OBM model, as it incorporates more detail in air dynamics [32]. This behavior is similar to observations in studies on solar radiation dynamics in the atmosphere.

The influence of temperature on pollutant concentrations follows a similar trend in the CBM and OBM models. Some compounds show a direct relationship with temperature, while others show an inverse relationship. However, the concentrations obtained in the OBM model tend to be lower due to more efficient dispersion [36,40]. The relationship between temperature and pollutant concentration has also been explored in previous studies, where concentration peaks tend to rise with temperature for certain compounds [74].

The OBM model presents faster dispersion and lower concentrations compared to the CBM model, highlighting the importance of considering dynamic factors such as altitude and airflow in air quality simulations [43]. This is consistent with other studies showing that convective dispersion plays a key role in pollutant dispersion [38].

5. Conclusions

The study of gaseous pollution modeling in the troposphere reveals that the interaction of primary pollutants, such as nitrogen oxides and volatile organic compounds, under high solar radiation conditions, leads to the formation of tropospheric ozone and other secondary pollutants. Through the mathematical model, which includes 31 ordinary differential equations, it was demonstrated that the concentrations of most secondary pollutants peak at midday, indicating a direct correlation between solar radiation and the formation of photochemical smog. Additionally, a complex dynamic of pollutants is observed, with nonlinear concentration behaviors influenced by temperature factors and emission rates. Moreover, the study highlights the importance in considering both chemical mechanisms and environmental conditions for assessing air quality and its impact

on public health and the environment. The implementation of “open-box” and “closed-box” models provides highly valuable tools for predicting pollution episodes and developing effective mitigation strategies.

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