



Examining the implications of photochemical indicators on the O_3 - NO_x -VOC sensitivity and control strategies: A case study in the Yangtze River Delta (YRD), China

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Abstract. Ozone (O_3) has become a significant air pollutant in China in recent years. O_3 abatement is challenging due to the nonlinear response of O_3 to precursors nitrogen oxides (NO_x) and volatile organic compounds (VOCs). Photochemical indicators are widely used to estimate the O_3 -NO_x-VOC sensitivity, and this has important policy implications. However, the effectiveness of the indicators has seldom been evaluated. This study examines the applications of four indicators that include the ratio of the production rates of H_2O_2 and HNO_3 (P_{H2O2}/P_{HNO3}) , $HCHO/NO_2$, $HCHO/NO_y$, and reactive nitrogen (NO_y) in the Yangtze River Delta (YRD) with localized thresholds. The overall accuracy was high (> 92%) for all indicators and not significantly reduced with different simulation periods or in different locations of the region. By comparing with the O_3 isopleths, it was found that $HCHO/NO_2$ and $HCHO/NO_y$ showed the most consistency, whereas P_{H2O2}/P_{HNO3} (NO_y) tended to underestimate (overestimate) the positive response of O_3 to NO_x . Additionally, P_{H2O2}/P_{HNO3} was less likely to attribute the O_3 formation to mixed sensitivity than the other indicators, and this demonstrated a preference for a single-pollutant control strategy. This study also revealed that the details in the methodology used to derive the threshold values impacted the results, and this may produce uncertainties in the application of photochemical indicators.

1 Introduction

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Ozone (O₃) is one of the most important pollutants in the atmosphere, high concentrations of which pose a serious threat to human health, ecosystems, and global climate change (Wang et al., 2020b; Liang et al., 2019; De Marco et al., 2022; Feng et al., 2022; Skeie et al., 2020). Many metropolitan areas in China, such as Beijing-Tianjin-Hebei (BTH), the Yangtze River Delta (YRD), and Pearl River Delta (PRD), have been experiencing severe O₃ pollution in recent years (Gao et al., 2017; Lu et al., 2018). Moreover, unlike the decline in fine particulate matter (PM_{2.5}), the O₃ concentrations in urban areas have



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increased (Wang et al., 2019; Lu et al., 2020; Li et al., 2020).

O₃ is produced by photochemical reactions involving nitrogen oxides (NO_x) and volatile organic compounds (VOCs). However, the response of O₃ to either precursor is nonlinear due to the changing roles of NO_x and VOCs with different VOC/NO_x ratios in O₃ chemistry, making O₃ pollution abatement more challenging (Wang et al., 2011; Liu et al., 2013). Briefly, when O₃ increases with a reduction (increase) in NO_x (VOCs), O₃ formation is in a NO_x-saturated or VOC-limited regime that tends to occur in urban areas or cold seasons with relatively lower VOC/NO_x ratios (Tan et al., 2018; Sillman, 1995). Conversely, O₃ decreases with reduced NO_x and is typically insensitive to VOC levels in a NO_x-limited regime. This is often the case in rural and remote regions with lower NO_x emissions (Murphy et al., 2007; Sillman, 1999). Importantly, a reduction in either NO_x or VOCs can lead to less O₃ production when the O₃ formation regime is shifted from one to the other, and this is referred to as the transitional regime (Chen et al., 2021; Jin and Holloway, 2015).

Many approaches, including photochemical indicators, the observation-based modeling method (OBM), and the emission-based method (EBM) have been developed to predict the O₃-NO_x-VOC sensitivity (Sillman, 2002; Wang et al., 2020a; Shen et al., 2021). Among these approaches, the values of the photochemical indicators (such as VOC/NO_x, the ratios of hydrogen peroxide to nitric acid (H₂O₂/HNO₃) or production rates of H₂O₂ and HNO₃ (P_{H2O2}/P_{HNO3}), the ratios of formaldehyde to nitrogen dioxide or reactive nitrogen (HCHO/NO₂ or HCHO/NO_y), and the O₃ to nitric acid ratios (O₃/HNO₃), NO_y) can be derived directly from ground-based measurements, chemical transport models, or even satellite measurements if available, and thus have been widely used in previous studies (Martin et al., 2004; Jin and Holloway, 2015; Jiménez and Baldasano, 2004; Sillman, 2002; Liu et al., 2010). For instance, Ye et al. (2021) presented hourly H₂O₂/NO_z ratios based on the measurements at Mount Tai and found that O₃ formation is VOC-limited in the morning (H₂O₂/NO_z<0.15) and switched to NO_x-limited (H₂O₂/NO_z>0.2) in the afternoon. The ratio H₂O₂/HNO₃ is considered one of the most robust indicators to determine O₃ sensitivity, while the thresholds are highly variable for different regions (Peng et al., 2011; Xie et al., 2014; Ye et al., 2016). P_{H2O2}/P_{HNO3} is primarily used in modeling work, such as the spatial distribution of O₃ formation regimes over the modeling domain that was exhibited with the indicator (Liu et al., 2010; Du et al., 2022; Zhang et al., 2020). Based on long-term measurements of HCHO and NO₂ columns from satellites, Jin and Holloway (2015) found that near-surface O₃ formation in several megacities in China had changed from VOC-limited in 2005 to transitional in 2013 with reduced NO₂.

The thresholds of photochemical indicators that split VOC-limited, transitional, and NO_x-limited regimes vary from region to region, largely depending on local emissions and possibly related to meteorological conditions as well. It was found that the primary emissions of HCHO increased the value of HCHO/NO₂, and therefore, the O₃ sensitivity determined by the ratio could have been biased with given thresholds. For example, some VOC-limited regimes are misclassified as transitional (Liu



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et al., 2021). Du et al. (2022) compared the spatial distribution of O₃ sensitivity regimes diagnosed using different indicators and pointed out that P_{H2O2}/P_{HNO3} could be more affected by local emissions than HCHO/NO₂. As H₂O₂ and HNO₃ are very soluble and susceptible to deposition and aerosol formation, the indicator values that involve the two species may change with the meteorological conditions (Castell et al., 2009). However, certain thresholds were often used without considering the variability, such as the range of one to two for HCHO/NO₂ (Jin and Holloway, 2015; Tang et al., 2012; Ma et al., 2021). In particular, the integrated source apportionment method (ISAM) implemented in the community multiscale air quality model (CMAQ) attributes O₃ production to either VOC or NO_x tracers based on a comparison of the instantaneous P_{H2O2}/P_{HNO3} with 0.35, and this ultimately affects the source apportionment of O₃ using ISAM (Kwok et al., 2015). Whether photochemical indicators can accurately predict O₃ sensitivity to precursors with given threshold values has seldom been examined.

In light of the above, this work aims to revisit the effectiveness of photochemical indicators in the prediction of the O₃-precursor response based on a case study in the YRD region in Eastern China. Specifically, the work deployed the mostly-used method to derive localized thresholds of the four photochemical indicators (i.e., P_{H2O2}/P_{HNO3}, HCHO/NO₂, HCHO/NO₃, and NO₃) for Jiangsu Province in the YRD that has been experiencing severe photochemical pollution (Qin et al., 2021; Xu et al., 2021; Lu et al., 2018). The assessment was conducted to examine three aspects: (1) whether the localized threshold values are appropriate for a different location or a different year; (2) the consistency of the photochemical indicator approach with O₃ isopleths plots for predicting the O₃-NO_x-VOC response; and (3) possible uncertainties induced by the methodology with which the threshold values are determined. This work can provide insight into the use of photochemical indicators in understanding O₃ formation and making effective strategies for emission control.

2 Methods

2.1 Model configurations

The CMAQ model version 5.2 was applied to simulate the photochemical pollution over the YRD and its surrounding areas (Fig. 1) in July 2017. The simulation was performed at a 4-km resolution, using the chemical mechanism of saprc07tic. A regional O₃ pollution event that occurred from July 22 to 31 with the mean level of the observed daily maximum 8-hr average (MDA8) O₃ exceeding 85 ppb over the YRD was selected to study the relationship between the O₃-NO_x-VOC sensitivity and the values of the photochemical indicators. The CMAQ model was driven by a mesoscale meteorological model, the Weather Research and Forecasting (WRF) version 4.2.2. The 2017 emission inventory of anthropogenic air pollutants in the YRD compiled by the Shanghai Academy of Environmental Sciences (An et al., 2020) was adopted, with the details of the other emissions (i.e., biogenic emissions and open burning) found in Hu et al. (2016).







Figure 1: The simulation domain. The blue area indicates the YRD including Jiangsu, Anhui, Shanghai, and Zhejiang. Three representative sites of Caochangmen (CCM), Zhonghuamen (ZHM), and Laoshan (LS) are marked.

2.2 Model performance evaluation

The model performance on O₃ was examined by comparing the simulation with observational data at the monitoring sites supervised by the China National Environmental Monitoring Center (CNEMC) in 13 cities in Jiangsu Province. Three metrics (the normalized mean bias (NMB), the normalized mean error (NME), and the correlation coefficient (r)) as in Sheng et al. (2022) were calculated to evaluate the agreement between the simulation and observations. The criteria of the model performance were recommended by Emery et al. (2017). Table S1 shows that the NMB (NME) values in 12 (10) out of 13 cities met the criteria, indicating that the overall performance was good. In general, the model overestimated O₃, with the NMB in the range of -2%–17%. The correlations of the simulation with the observations (all r values above 0.5) indicated that hourly variations in the simulation agree well with the observations. Additionally, three sites in Nanjing (a megacity in Jiangsu), including Caochangmen (CCM), Zhonghuamen (ZHM), and Laoshan (LS), representing downtown, upwind, and downwind areas of the city (see Fig. 1), respectively, were selected to examine the consistency between the photochemical indicators and the O₃ isopleths in the implication of O₃-precursor sensitivities. Figure S1 shows that the model well reproduced the O₃ pollution events at the three sites during the simulation period.

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Other species involved in the photochemical indicators, such as HCHO, HNO₃, NO₂, and NO_y (NO_y (g)=NO+NO₂+NO₃+2×N₂O₅+HONO+HNO₃+HNO₄+peroxyacetyl nitrate and its homologs (PANs)+alkyl nitrates (ANs) and nitrogen oxychlorides (ClNO_x) in the model), were examined and compared with observations during the EXPLORE-YRD campaign (Li et al., 2021). Figure S2 showed that HCHO was underestimated by 51%, while NO₂ and NO_y were slightly overestimated, with NMBs of 16% and 7%, respectively. The anthropogenic emission inventory and model were set up as in the simulation for the EXPLORE-YRD campaign, and the simulated ratios of HCHO/NO₂ and HCHO/NO_y in this study could have been underestimated. In addition, the simulation could represent an environment with less abundant VOCs than in the real atmosphere over the YRD.

2.3 Determination of the indicator thresholds

The use of photochemical indicators requires threshold values that differentiate O_3 formation regimes. In this study, the thresholds of the indicators were derived based on the association of the O_3 reduction relative to the baseline emission scenario resulting from NO_x or VOC emission reductions (i.e., $\Delta O_{3 NOx}$ or $\Delta O_{3 VOC}$), with concurrent indicator values at all grid cells in Jiangsu Province. Therefore, two additional runs with either a 35% decrease in NO_x or a 35% decrease in VOC emissions were performed following Sillman et al. (1998) and other studies (Xie et al., 2014; Peng et al., 2011). The resulting O_3 changes ($\Delta O_{3 NOx}$ and $\Delta O_{3 VOC}$) were obtained to identify the O_3 formation regimes at each grid cell, with the criteria given in Table 1. It should be noted that the average changes in O_3 during 1–4 p.m. during the simulation period were used when O_3 peaked and was of most concern.

Table 1: Identification of the O₃ formation regimes at each grid cell based on the O₃ changes resulting from perturbed emissions as in Sillman et al. (1998) or the sensitivity coefficients using the high-order decoupled direct method (HDDM) as in Wang et al. (2011).

Method O ₃ formation regime		Definition		
Perturbed simulations	VOC-limited	$\Delta O_{3 \text{ VOC}} \ge 5 \text{ ppb}$ and $(\Delta O_{3 \text{ VOC}} - \Delta O_{3 \text{ NOx}}) \ge 5 \text{ ppb}$		
	NO _x -limited	$\Delta O_{3 \text{ NOx}} \ge 5 \text{ ppb}$ and $(\Delta O_{3 \text{ NOx}} - \Delta O_{3 \text{ VOC}}) \ge 5 \text{ ppb}$		
High-order decoupled	VOC-limited	$S_{\text{VOCs}}^1 \ge 5 \text{ ppb and } (S_{\text{VOCs}}^1 - S_{\text{NOx}}^1) \ge 5 \text{ ppb}$		
direct method (HDDM)	NO _x -limited	$S_{Nox}^1 \ge 5$ ppb and $(S_{Nox}^1 - S_{Vocs}^1) \ge 5$ ppb		

This study focused on four photochemical indicators that are frequently used, including P_{H2O2}/P_{HNO3}, HCHO/NO₂, HCHO/NO_y, and NO_y, with the underlying mechanisms extensively described in the literature (Xie et al., 2014; Kwok et al., 2015; Sillman, 1995; Duncan et al., 2010). The indicator values for each grid cell were calculated based on the simulated concentrations of related species, except for P_{H2O2}/P_{HNO3}, which were obtained using the integrated reaction rate (IRR) of the CMAQ process analysis tool (Gipson and Young, 1999). The percentile distribution of the indicator values for the VOC-



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limited grid cells and the NO_x-limited grid cells were examined individually. The photochemical indicators typically showed higher values at NO_x-limited grid cells than in VOC-limited locations. Finally, the thresholds of a certain indicator were determined using the 95th percentile value of the indicator for the VOC-limited grid cells and the 5th percentile value for the NO_x-limited grid cells as the lower and upper limits of the transition intervals, respectively. Indicator values lower than the transition intervals suggested O₃ formation was VOC-limited, while higher values than the transition intervals were associated with NO_x-limited regimes. In some cases, the 95th percentile VOC-limited value was higher than the 5th percentile NO_x-limited value, indicating the indicator was invalid and could not be used to determine the O₃ sensitivity properly.

2.4 Evaluation of the indicators

2.4.1 Evaluation metrics

Some metrics have been developed to estimate the uncertainties in the determination of O₃ regimes using photochemical indicators (Wang et al., 2011; Ye et al., 2016). This work applied Error A, Error B, and the overall accuracy (OA) in Ye et al. (2016) for the evaluation, with the equations given in Table 2. For example, the ErrA_NO_x describes the situation where among all grid cells (colored areas in Fig. 2d) with indicator values higher than the upper limit of the transition interval (i.e., assigned to NO_x-limited regimes with the indicator), there existed VOC-limited grid cells that were incorrectly assigned to the NO_x-limited regime. ErrB_NO_x estimates the fraction of the NO_x-limited grid cells (blue areas in Fig. 2d) that are not classified as a NO_x-limited regime since the corresponding indicator values are lower than the upper threshold. ErrA_VOC and ErrB_VOC were defined analogously. OA describes the fraction of all correctly classified grid cells using the indicator at all VOC- or NO_x-limited grid cells.

Table 2: Metrics of the indicator performance evaluation provided in Ye et al. (2016).

Metric	Equation
Error	Number of VOC-limited grid cells with indicator values above the upper threshold
A_NO _x	Number of all NO_x -limited and VOC-limited grid cells with indicator values above the upper threshold $\times 100\%$
Error	Number of NO _x -limited grid cells with indicator values below the upper threshold ×100%
B_NO _x	Number of all NO _x -limited grid cells
Error	Number of NO _x -limited grid cells with indicator values below the lower threshold
A_VOC	$\overline{\text{Number of all NO}_{x}\text{-limited and VOC-limited grid cells with indicator values below the lower threshold}}^{\times 100\%}$
Error	Number of VOC-limited grid cells with indicator values above the lower threshold ×100%
B_VOC	Number of all VOC-limited grid cells
OA	OA=(Number of VOC-limited grid cells with indicator values below the lower threshold + Number of NO _x -limited grid cells with indicator values above the upper the Number of all NO _x -limited and VOC-limited grid cells



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The indicator evaluation was conducted for the simulation in 2017 (see section 2.1). Additionally, a simulation for the same period (over July 22–31) in 2018, which has an identical model configuration and inputs in all aspects except for the meteorology, was performed for the evaluation as well. Thus, applications of the photochemical indicators with derived thresholds in a different location within the region, in a different year, or with changes in both, can be examined separately. This provides insight into the uncertainties associated with emissions or meteorology.

2.4.2 O₃ isopleths

The O₃-NO_x-VOC sensitivity determined by photochemical indicators was compared with the O₃ isopleths that describe the nonlinear relationship between O₃ and the precursors and inform decision-makers of emission control strategies in a different manner. The response of O₃ to the domain-wide reductions in VOC and NO_x emissions at three representative sites (CCM, ZHM, and LS) during 1–4 p.m. during the simulation period was investigated. Specifically, a total of 36 emission scenarios with anthropogenic VOC and NO_x emissions reduced by 0, 20, 40, 60, 80, and 100%, both singly and in combination were simulated to construct the O₃ isopleth diagram. The O₃ formation regimes based on the thresholds derived in section 2.3 are indicated with color and overlapped with the isopleths.

2.5 Uncertainties in the thresholds

In section 2.3, the O₃ formation regimes were presumably identified according to changes in O₃ with the emission reductions, while the rules could be different in detail. For example, Liang et al. (2006) examined changes in the 8-hr O₃ with a 25% reduction in VOC or NO_x emissions instead of the 1-hr O₃ with 35% reductions, as in Sillman (1995). Accordingly, the criteria changed from 5 ppb to 2.5 ppb. Zhang et al. (2020) explored the correlation of indicator values with the relative O₃ changes resulting from a 50% reduction in the NO_x or VOC emissions. Du et al. (2022) applied the decoupled direct method (DDM) to investigate the sensitivity of O₃ to precursors, instead of perturbing emissions. The impacts on the derived thresholds of the indicators were explored by conducting a few individual tests:

- (1) The criteria for the O₃ reduction (see Table 1) were set at 2 ppb, 3 ppb, and 6 ppb.
 - (2) The relative changes in O₃ were examined using the criteria of 2%, 5%, and 8% instead of an absolute change of 5 ppb.
 - (3) The 35% emission reduction scenarios were replaced with a 20% and a 40% reduction in the VOC or NO_x emissions.
- (4) The high-order decoupled direct method (HDDM) (Cohan et al., 2005) was adopted to split the O₃ formation regimes in two manners referred to as "DDM1" and "DDM2". The HDDM estimated the first- and second-order sensitivity coefficients of O₃ to NO_x and VOCs (i.e., S¹_{NOx}, S¹_{VOCs}, S²_{NOx}, S²_{VOCs}). With the DDM1 approach, the NO_x-limited locations were defined as those grid cells where S¹_{NOx} was higher than 5 ppb and at least 5 ppb higher than S¹_{VOCs} (Table 1), following Wang et al. (2011). The VOC-limited grid cells were defined similarly. The DDM2 approach estimated O₃ changes with 35% reductions in



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emissions via the Taylor series expansions (Cohan et al., 2005) instead of the perturbed simulations.

3 Results and discussion

3.1 Threshold values of the photochemical indicators

The changes in the O_3 concentrations when reducing VOC or NO_x emissions were used to determine the O_3 formation regimes. Figure 2a shows that high O_3 concentrations (averaged from 1 to 4 p.m. during July 22–31) primarily occurred along the Yangtze River and southern Jiangsu, and O_3 in northern Jiangsu was relatively low. The NO_x emission reduction predominantly led to O_3 decreases in Jiangsu, while only a few locations, such as near the estuary of the Yangtze River, showed an increased O_3 relative to the base case (Fig. 2b). The VOCs emission reductions resulted in less significant O_3 decreases compared to the NO_x reduction scenario (Fig. 2c). The O_3 sensitivity regime for each grid cell in Jiangsu was identified with the method depicted in section 2.3 (Fig. 2d). The O_3 formation in central and southern Jiangsu, where O_3 is abundant, is typically NO_x -limited as a result of a relatively low NO_x concentration in the afternoon (Duncan et al., 2010; Jin et al., 2017; Jin and Holloway, 2015). Some areas along (and on) the Yangtze River were characterized to be VOC-limited regimes, with substantial emissions of NO_x from ship emissions and industry (Sheng et al., 2022). However, not all grid cells in Jiangsu were assigned to the VOC- or NO_x -limited regimes. The white areas in Jiangsu in Fig. 2d indicate that O_3 was insensitive to both NO_x and VOCs (i.e., ΔO_3 $_{VOC}$ < 5 ppb and ΔO_3 $_{NOx}$ < 5 ppb); alternatively, O_3 was sensitive to NO_x (VOC) emissions but showed small changes in O_3 with the perturbation of NO_x (VOC) relative to the VOC (NO_x) reduction scenario (i.e., $|\Delta O_3|_{VOC}$ $-\Delta O_3|_{NOx}$ < 5 ppb).



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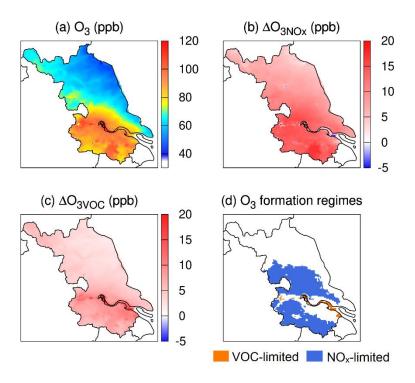


Figure 2: (a) Simulated spatial distribution of O_3 averaged from 1 to 4 p.m. during July 22-31 in Jiangsu Province, (b-c) O_3 reduction due to a 35% reduction in domain-wide NO_x and VOC emissions, respectively, (d) O_3 formation regimes determined with perturbed simulations.

The O_3 reductions (ΔO_3 Nox or ΔO_3 VoC) against photochemical indicator values at all grid cells over the domain were examined (Fig. 3). The mean values of the ΔO_3 Nox slightly increased with higher P_{H2O2}/P_{HNO3} , while the ΔO_3 VoC values showed a decreasing trend, indicating that a reduction in NO_x (VOC) emissions became more (less) effective for O_3 abatement in the locations with higher indicator values. Other indicators displayed a similar pattern except for NO_y , which displayed an opposite result. Due to the contrasting dependence of the ΔO_3 Nox and the ΔO_3 VoC on the indicator values, the thresholds of the photochemical indicators that separate the O_3 formation regimes were determined (Table 3 and vertical lines in Fig. 3). The localized thresholds for P_{H2O2}/P_{HNO3} (0.55, 0.75) were remarkably higher than that implemented in the model, i.e., 0.35, which would lead to misattribution of the VOC-limited or the transitional regime to the NO_x -limited regime in some locations if the single value was directly used. The thresholds differed from those proposed by researchers in other countries, and some regions in China had significantly distinct thresholds as well (Du et al., 2022; Zhang et al., 2020; Tonnesen and Dennis, 2000a; Liu et al., 2010).



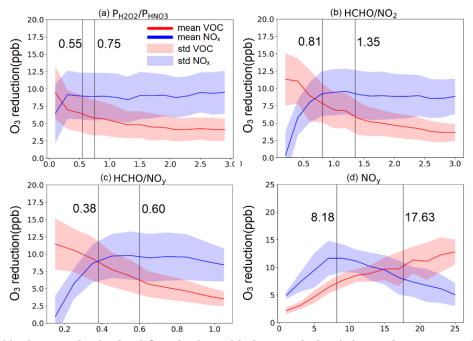


Figure 3: Relationships between the simulated O_3 reductions with the perturbed emissions and concurrent values of P_{H2O2}/P_{HNO3} , HCHO/NO₂, HCHO/NO₃, and NO₃ at all the grid cells in the domain. The red (blue) lines represent the average O_3 changes resulting from VOC (NO₃) emission reductions, with the shaded areas for the standard deviations. The thresholds derived in this study are indicated as the grey vertical lines.

Table 3: Thresholds of the indicators derived in this study and reported in previous work.

Indicators	This study	Other studies		
P _{H2O2} /P _{HNO3}	(0.55, 0.75)	US: 0.06 (Tonnesen and Dennis, 2000a)		
		China: 0.2 (Liu et al., 2010)		
		North China Plain: (0.08, 0.2) (Zhang et al., 2020)		
		East China: (0.30, 1.10) (Du et al., 2022)		
HCHO/NO ₂ (surface)	(0.81, 1.35)	US: (0.8, 1.8) (Tonnesen and Dennis, 2000b)		
		East Asia: (0.5, 0.8) (Jin et al., 2017)		
		East China: (0.36, 0.45) (Du et al., 2022)		
		Yangtze River Delta: (0.55, 1.0) (Liu et al., 2021)		
HCHO/NO _y	(0.38, 0.60)	US: 0.28 (Sillman, 1995)		
		California: (0.5, 0.9) (Lu and Chang, 1998)		
		Pearl River Delta: 0.41 (Ye et al., 2016)		



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NO _y	(8.18, 17.63)	US: 20 (Sillman, 1995)	
		California: 5 (Lu and Chang, 1998)	
		Germany: 7.78 (Vogel et al., 1999)	
		Mexico City: 8.75 (Torres-Jardon et al., 2009)	

This work revisited other photochemical indicators including the surface HCHO/NO₂, HCHO/NO₃, and NO₃. The transition interval of HCHO/NO₂ was determined as (0.81, 1.35), showing a higher and wider range than in other work (Jin et al., 2017; Liu et al., 2021; Du et al., 2022), but relatively consistent with Tonnesen and Dennis (2000b). It is worth noting that the values mentioned above are all based on measurements or simulations at the ground level, while column HCHO/NO₂ retrieved from satellite measurements are more widely used, with thresholds of (1, 2) (Jin and Holloway, 2015; Jin et al., 2017; Duncan et al., 2010; Wang et al., 2021). The transition range for the column HCHO/NO₂ is typically broader than that for the surface ratio, e.g., the former of (1.17, 2.42) versus the latter of (0.36, 0.45) in East China (Du et al., 2022). The thresholds of HCHO/NO₂ and NO₃ in this study were (0.38, 0.60) and (8.18, 17.63), respectively, comparable to the values proposed previously. The comparison shown in Table 3 suggests that the thresholds were mostly location-specific, possibly as they are dependent on atmospheric conditions and emission features. The feasibility of a given threshold for an indicator applied throughout a region, or over a short-term period (the order of 1–2 years) is evaluated in section 3.2. However, the methodology that is used to derive thresholds can also lead to distinct threshold values, and this is discussed in section 3.4.

The spatial distributions of the O_3 formation regimes identified by the photochemical indicators with localized thresholds are exhibited in Fig. 4. Most of the areas in Jiangsu (> 80% of the grid cells covering Jiangsu, Table S2) were attributed to NO_{x} -limited regimes in terms of the summertime O_3 formation in the afternoon using all indicators. A few urban centers and some locations near the Yangtze River were identified as VOC-limited or transitional regimes where the four indicators largely showed differences. The O_3 formations in these areas were primarily VOC-limited according to P_{H2O2}/P_{HNO3} and showed a transition from VOC-limited to NO_x -limited at the surrounding grid cells. However, other indicators tended to assign the areas along and to the south of the river to transitional regimes. The fraction of the VOC-limited regimes in the area in Jiangsu with P_{H2O2}/P_{HNO3} (~13%) was considerably higher than that using HCHO/ NO_2 and HCHO/ NO_y (4-5%). NO_y , which is the least impacted by VOC emissions, attributed the least areas to the VOC-limited regimes (~2%) concerning O_3 formation. Accordingly, transitional regimes were the least using P_{H2O2}/P_{HNO3} , ~5% compared to 13–15% using HCHO/ NO_2 , HCHO/ NO_y , and NO_y .



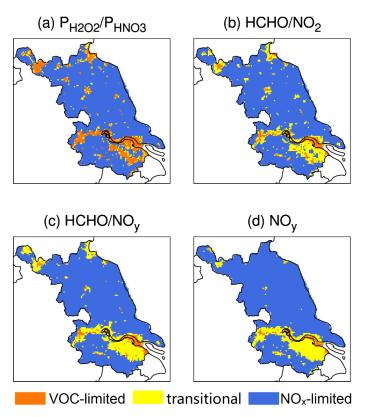


Figure 4: O_3 formation regimes based on the indicator values and thresholds in Table 3. The orange indicates a VOC-limited regime. The blue indicates a NO_x -limited regime. The yellow is associated with a transitional regime.

3.2 Evaluation of the photochemical indicators

Table 4 shows the accuracy of the photochemical indicators in splitting the O₃ formation regimes with thresholds derived in this study. As expected, the ErrB and OA for the application in Jiangsu in 2017 (see the case "Sim_2017+Jiangsu" in Table 4) were approximately 5% and 95%, respectively, as a result of the nature of the method (see section 2.3). The values of ErrA were lower than ErrB, suggesting that chemical regimes determined by the indicator values were largely consistent with that based on O₃ changes when perturbing emissions. However, it is more likely that sensitivities indicated by O₃ changes do not necessarily correspond to indicator values in the correct intervals. Additionally, the ErrA_VOC and ErrA_NO_x for P_{H2O2}/P_{HNO3} were higher than that of the other indicators, likely due to less area associated with mixed O₃ sensitivity using P_{H2O2}/P_{HNO3} (see Table S2), which could increase the chance of misclassification.



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Table 4: Evaluation of the photochemical indicators in the YRD (unit of %).

		ErrA_VOCs	ErrB_VOCs	ErrA_NO _x	ErrB_NO _x	OA
Sim_2017+Jiangsu	P _{H2O2} /P _{HNO3}	3.9	6.6	0.1	5.1	94.8
	HCHO/NO ₂	2.9	5.7	0.0	5.1	94.9
	HCHO/NO _y	0.0	6.6	0.0	4.7	95.2
	NO _y	0.0	5.7	0.0	4.9	95.1
Sim_2017+Other	P _{H2O2} /P _{HNO3}	7.4	0.0	0.0	2.0	98.0
	HCHO/NO ₂	2.2	0.0	0.0	1.7	98.3
	HCHO/NO _y	2.1	0.0	0.0	0.7	99.3
	NOy	0.0	12.8	0.0	1.3	98.6
Sim_2018+Jiangsu	P _{H2O2} /P _{HNO3}	2.4	0.0	0.0	4.3	96.9
	HCHO/NO ₂	0.0	0.8	0.0	7.0	94.8
	HCHO/NO _y	0.0	0.8	0.0	10.1	92.2
	NO _y	0.0	8.3	0.3	0.3	97.4
Sim_2018+Other	P _{H2O2} /P _{HNO3}	2.7	7.0	0.1	1.7	98.2
	HCHO/NO ₂	6.2	14.3	0.1	1.6	98.0
	HCHO/NO _y	0.5	12.2	0.1	1.2	98.5
	NO _y	0.0	51.3	0.1	1.3	97.3

The above evaluation also showed that emissions could have a minor effect on the performance of the indicators on a regional scale. The thresholds derived from the simulation in Jiangsu well identified O_3 sensitivities in other regions of the YRD (see the case "Sim_2017+Other"), where emissions were significantly different (e.g., abundant biogenic VOCs in the southern YRD vs. dominance of anthropogenic emissions in Jiangsu) and with OAs greater than 98% for all the indicators. This suggests that it is plausible to specify regionwide thresholds for photochemical indicators. In addition, the indicators can generally separate O_3 formation regimes in Jiangsu for the same period in 2018 (see the case "Sim_2018+Jiangsu") when the meteorology could differ from that in 2017. However, HCHO/NO_y and HCHO/NO₂ appeared to be more easily affected by meteorology, as indicated by the ErrB_NO_x that showed that many NO_x-limited grids had indicator values lower than the upper thresholds. The OAs were relatively high in the case when both the emissions and meteorology changed (see the case "Sim_2018+Other"), although the ErrB_VOC were remarkably higher. This was due to fewer VOC-limited grids in the YRD, half of which had indicator values within the transition intervals and thus were misclassified as the transitional regime. The evaluation showed that indicators could perform better in detecting O_3 formation regimes that were predominant. However, the drawback of the evaluation is that it does not account for grids that are neither VOC-limited nor NO_x-limited in terms of the O_3 formation based on the ΔO_3 Nox or the ΔO_3 Nox (i.e., the white area in Jiangsu in Fig. 2d), and



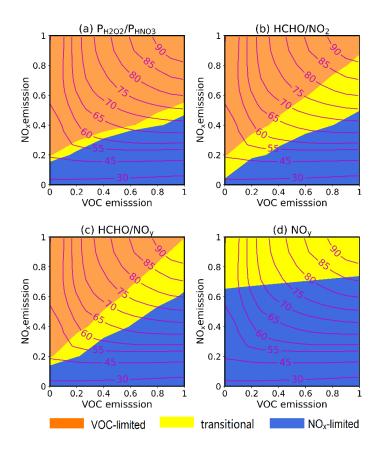
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the uncertainties of the photochemical indicators in these areas are unknown.

3.3 Consistency with the O₃ isopleths

O₃ formation regimes identified by the indicator values with varied combinations of VOC and NO_x emissions were plotted together with the O₃ isopleths to examine the indicators for implications in effective emission control strategies. Figure 5 and Figs. S3–S4 show the comparison at the CCM, ZHM, and LS sites, respectively. As no significant variations were found at the three sites, only the plot for CCM is shown in the main text. The P_{H2O2}/P_{HNO3} value with the base emission points to a VOC-limited regime at the CCM site, whereas the isopleths suggest that a reduction in the NO_x emissions should lead to reduced O₃. This also happens in some cases with lower VOC and NO_x emissions, indicating that P_{H2O2}/P_{HNO3} is likely to underestimate the positive sensitivity of O₃ to NO_x. The indicators were assumed to have constant thresholds that do not change with emissions; this could partially explain the inconsistency of the indicators with the O₃ isopleths. The P_{H2O2}/P_{HNO3} values at the LS and ZHM sites showed similar results, except that the O₃ formation at LS site shifted from VOC-limited to a transitional regime earlier with decreasing NO_x emissions than at the other two sites as indicated by the ratio.





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Figure 5: O_3 isopleths (red lines) overlap with the O_3 formation regimes (shading color) identified with P_{H2O2}/P_{HNO3} (a), $HCHO/NO_2$ (b), $HCHO/NO_y$ (c), and NO_y (d) at the CCM site. The orange indicates a VOC-limited regime. The yellow indicates a transitional regime, and the blue indicates a NO_x -limited regime.

It appears that HCHO/NO_y was the most consistent with the O₃ isopleths among the four indicators, followed by HCHO/NO₂. Both HCHO/NO_y and HCHO/NO₂ suggest that all three sites resided in transitional regimes under the base emission scenario, except for CCM using HCHO/NO₂, which showed that O₃ would respond to NO_x changes negatively. However, HCHO/NO_y and HCHO/NO₂ had more emission scenarios assigned to transitional regimes compared with P_{H2O2}/P_{HNO3}, which was in agreement with the spatial distributions shown in Fig. 4. This implies that the two indicators emphasize controlling both VOC and NO_x rather than a single pollutant. However, the HCHO levels could be affected by primary emissions (Liu et al., 2021) or highly variable due to the dependence of isoprene on temperature (isoprene dominating HCHO production with intensive biogenic emissions) (Duncan et al., 2010). Therefore, the link between the HCHO/NO_y (or HCHO/NO₂) values and the O₃-VOC-NO_x sensitivity and resulting emission control strategies of anthropogenic precursors (that are controllable) is more uncertain under certain conditions.

The O₃ formation at the CCM site was dominated by NO_x-limited regimes based on NO_y, which was predominantly affected by NO_x emissions, and did not show a sole sensitivity to VOCs (Fig. 5d). This was different from other indicators as well as the O₃ isopleths in terms of policy implications. The comparison between the O₃-VOC-NO_x sensitivity revealed by NO_y and the O₃ isopleths showed that the indicator tended to overestimate the response of O₃ to NO_x (i.e., the VOC-limited regimes were misclassified as transitional regimes in the upper left corner of Fig. 5d) or underestimate the O₃-VOC sensitivity (i.e., the transitional regimes were misclassified as NO_x-limited regimes in the middle-left portion of Fig. 5d).

3.4 Uncertainties associated with the methodology

Apart from local emissions and meteorology that could partially explain the gap between the indicator thresholds proposed earlier and in this work, we found that the methodology to derive the thresholds could be slightly different and the impacts are unknown. Therefore, a series of tests were conducted to examine how those details in the methodology alter the thresholds of P_{H2O2}/P_{HNO3} (Fig. 6).



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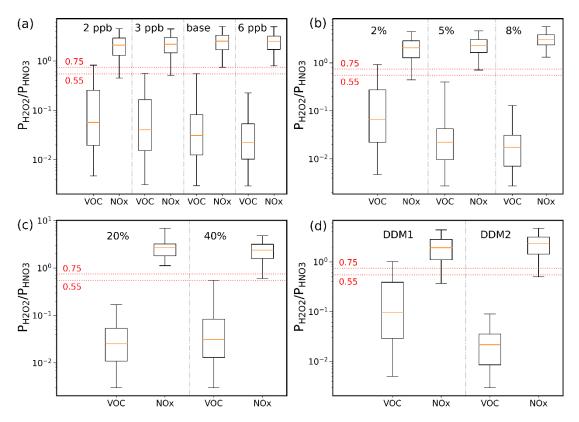


Figure 6: The percentile distributions of the $P_{\rm H2O2}/P_{\rm HNO3}$ values at the VOC- or NO_x-limited grid cells with different setups in methodology. The boxes indicate the 25th and 75th percentiles of the indicator values with the median values marked with red horizontal lines in the boxes, and the whiskers extend to the 5th and 95th percentiles. In each group, the upper whisker of the box in the VOCs column and the lower whisker of the box in the NO_x column reflect the lower and upper values of the threshold intervals, respectively. The threshold intervals in Table 3 are indicated with red dotted lines.

First, the VOC- (NO_x -) limited grid cells were defined as the locations where the reduction in O_3 as a response to the VOC (NO_x) reduction exceeded the O_3 reduction associated with the reduced NO_x (VOC) by more than 2 ppb, 3 ppb, and 6 ppb instead of the original 5 ppb (see Table 1). The threshold intervals are wider with higher criteria, e.g., the range of (0.3, 0.8) with sensitivity differences of > 6 ppb versus (0.55, 0.75) with > 5 ppb, leading to fewer VOC- or NO_x -limited grid cells based on the indicator and more transitional grid cells with less chance of misclassification (Fig. 6a). However, the sensitivity differences of 2–3 ppb are too small to differentiate sensitivity types of the grid cells using the indicators, as the 95th percentile of the indicator values at the VOC-limited grids is higher than the 5th percentile of the values at the NO_x -limited grids when determining the thresholds (see section 2.3). Hence, the indicators are invalid. Similarly, if the relative changes in O_3 were applied to quantify the O_3 -VOC- NO_x sensitivity, such as using the differences of 2%, 5%, or 8% as the



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criteria, the derived thresholds would be considerably different as well (Fig. 6b). The relative changes make more sense since O_3 changes of \sim 5 ppb due to a certain amount of VOC or NO_x reduction are less likely to take place in locations with low O_3 than in relatively polluted regions. As a result, the grid cells with low O_3 are seemingly insensitive to precursors and were excluded from the VOC- or NO_x -limited grids that were used to derive the thresholds. This can particularly affect the lower limits of the thresholds with much fewer VOC-limited grids, as in this study, and vice versa.

In Fig. 6c, the VOC and NO_x emissions were individually reduced by 20% or 40% instead of 35%, but with a constant criterion of > 5 ppb for the sensitivity differences. The less reduction in emissions resulted in wider threshold intervals (0.18, 1.12) with a 20% reduction versus (0.55, 0.75) with a 35% reduction, and (0.55, 0.60) with a 40% reduction. This was similar to the case with higher criteria (Fig. 6a). In addition, if the VOC- or NO_x-limited grids were defined based on the S ¹_{VOCs} and S¹_{NOx} in the HDDM (see Table 1 and DDM1 in Fig. 6d), the indicator values for the VOC-limited grids and the NO_xlimited grids would overlap to a large extent, with the 95th percentile of the indicator values at the VOC-limited grids of 1.0, and the 5th percentile of the values at the NO_x-limited grids of 0.37. Therefore, the indicator could not be used. S_{VOCs}^1 and S¹_{NOx} approximate the O₃ changes with a 100% reduction in the VOC and NO_x emissions, respectively, without considering the nonlinearity (i.e., higher-order sensitivities) in O₃ chemistry. This finding is consistent with Fig. 6c which shows a quite narrow range of threshold intervals with more reduction in the precursor emissions. Additionally, the upper and lower thresholds derived with the DDM2 approach were 0.50 and 0.11, lower than 0.75 and 0.55, respectively. This was due to the lower O₃ changes estimated with the second-order Taylor expansions than with the perturbed simulations (Fig. S5), similar to the case with a 20% reduction in the VOC or NO_x emissions, and possibly reflecting the importance of higher-order terms in the HDDM as well. Utilizing P_{H2O2}/P_{HNO3} as an example indicated that the thresholds of the photochemical indicators are dependent on the methods or parameters in the methodology. This could be one of the major uncertainties in the application of indicators to determine O₃ chemical regimes.

4 Conclusions

This work revisited four photochemical indicators, including P_{H2O2}/P_{HNO3}, HCHO/NO₂, HCHO/NO_y, and NO_y, in implications of the O₃-NO_x-VOC sensitivity and emission control strategies using a case study in the YRD. The threshold intervals for Jiangsu (one of the provinces in the YRD) were derived, (0.55, 0.75) for P_{H2O2}/P_{HNO3}, (0.81, 1.35) for the surface ratio of HCHO/NO₂, (0.38, 0.60) for HCHO/NO_y, and (8.2, 17.6) for NO_y, based on the relationship between the simulated indicator values and O₃ changes resulting from a 35% reduction in the VOC or NO_x emissions. The indicators along with the localized thresholds were applied in other areas in the YRD, for a different simulation period, and with a combination of different locations and time to estimate the uncertainties related to the emissions and meteorology. The indicators displayed

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good performance in all cases, with OA values greater than 92%, while HCHO/NO_y and HCHO/NO_z might have been susceptible to meteorological factors. However, HCHO/NO_y and HCHO/NO_z were the most consistent with the O_z isopleths among the four indicators and could be informative for policymakers. The P_{H2O2}/P_{HNO3} ratio was less likely to attribute O_z formation to mixed sensitivity and tended to underestimate the positive sensitivity of O_z to NO_z compared to the isopleths. In contrast, NO_z, which was not constrained by the VOC abundance, overestimated the positive response of O_z to NO_z and

underestimated the O₃-VOC sensitivity in some conditions.

Importantly, the intrinsic characteristics of the indicators as well as the methods used to obtain the thresholds affected the effectiveness of the indicators. A series of sensitivity tests were conducted to investigate the impacts with respect to the criteria for the definition of VOC- or NO_x-limited grids, the amount of VOC and NO_x emission perturbations, and the method, i.e., the HDDM tool versus emission reduction, to estimate the O₃-NO_x-VOC sensitivity in the first place. The results showed that the more stringent the criteria were (i.e., with larger the O₃ sensitivity differences or lower emission reductions using a given criterion), the wider the transition intervals were. This led to more attribution to the mixed O₃ sensitivity and showed a preference for concurrent control of both precursors. However, the indicator values at the VOC- and NO_x- limited grids could overlap in a large part and could not split the O₃ sensitivity when the criteria were set too low. Finally, the first-order coefficients in the HDDM alone with a specified sensitivity difference could not be used to derive thresholds of indicators in this study, while Taylor expansions with first- and second-order sensitivity coefficients could result in lower and broader threshold intervals.

Photochemical indicators have the advantage of identifying O₃ formation regimes promptly and are useful in policy implications. According to the case study in this work, we found that it was inappropriate to directly use the threshold of 0.35 for P_{H2O2}/P_{HNO3} in the CTMs, as the value could be location-specific. More broadly, all of the indicators should be employed with specified thresholds that have been localized and thoroughly evaluated. In addition, the indicators could be different from each other. As such, it is necessary to understand the pros and cons of each indicator prior to their use.

Code and data availability

The CMAQ outputs are currently available upon request. All python codes used to create any of the figures are available upon request.

Author contributions





XL, MQ and JH conceived and designed the research. XL performed the simulations and analyzed the data. MQ, JH, HS and JL contributed to result discussion. XL and MQ wrote the manuscript with substantial contributions from all of the authors.

405 Competing interests

The authors declare that they have no conflict of interest.

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