



Measurement report: Altitudinal Shift of Ozone Regimes in a Mountainous Background Region

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Abstract: Elevated background ozone (O_3) poses significant challenges for regional air quality management. Understanding the vertical distribution of O_3 and its precursors is critical yet underexplored in Southwest China. This study presents the first comprehensive altitudinal gradient analysis (550 m, 1,774 m, 2,119 m a.s.l.) in the Fanjingshan National Nature Reserve, a remote high-altitude site on the Yunnan-Guizhou Plateau. Continuous measurements (March–August 2024) revealed a marked positive gradient in O_3 (14.8 ± 15.2 ppb at mountain foot to 40.2 ± 14.7 ppb at mountaintop), contrasting with declining precursor concentrations. Random Forest–SHAP analysis identified relative humidity and NO_x as dominant controls at the mountain foot, whereas temperature and reactive VOCs governed O_3 variability aloft. Chemical box modeling (OBM-MCM v3.3.1) demonstrated net O_3 destruction at mountain foot (-1.93 ppb hr^{-1}) due to NO titration, shifting to net production at mountainside (0.35 ppb hr^{-1}) and mountaintop (0.29 ppb hr^{-1}). While O_3 formation remained NO_x -limited across all sites, sensitivity to anthropogenic hydrocarbons increased with altitude (RIR: -0.12 mountain foot to 0.51 mountaintop). Transport analysis indicated O_3 accumulation at mountain foot via regional transport, contrasting with mountainside and mountaintop, which function as net source regions. These findings necessitate altitude-specific O_3 control: prioritizing NO_x reduction at lower elevations while coordinating NO_x and VOC controls at higher altitudes. Expanding high-altitude monitoring, especially in under-monitored areas like Southwest China, is crucial for characterizing regional background pollution. Future studies require vertical monitoring with improved models to assess transboundary impacts and changing emissions.

Key words: Vertical Ozone Profiling; Ozone precursors; Photochemical Mechanisms; Regional Transport; Mount Fanjing

1 Introduction

Ozone (O_3) within the atmospheric boundary layer results from the disruption of fundamental NO - NO_2 - O_3 photochemical cycles through radical generation and transport processes under combined thermodynamic and dynamic factors (Atkinson and Arey, 2003; Carter, 2012). It exerts significant influences on biogeochemical cycles, vegetation net primary productivity, climate forcing, air quality, and human health (Feng et al., 2022; Tong et al., 2023; Feng et al., 2015; Wang et al., 2020; Wang et al., 2014). Although previous studies have demonstrated that regulating precursor emissions following the EKMA curve can mitigate photochemical pollution, control measures may fail due to neglecting the influence of horizontal regional transport, vertical exchange on O_3 and its precursor concentrations. However, conventional ambient air quality monitoring stations for O_3 are typically ground-based, and their fluctuating measurements cannot fully capture the spatial distribution of O_3 in both horizontal and vertical dimensions, thereby limiting their utility for air quality forecasting and the formulation of pollution control strategies. In key regions and megacities, rare vertical profiles of O_3 and its precursors have been obtained using platforms such as tethered balloons, manned/unmanned aircraft, and meteorological gradient towers (Dieu Hien et al., 2019; Li et al., 2022; Mo et al., 2022; Xue et al., 2011a). Undoubtedly, vertical profiles from tethered balloons, aircraft campaigns, and meteorological towers have significantly advanced our understanding of anthropogenic influences on large-



scale atmospheric processes and enriched theories of tropospheric chemistry and transport (Wang et al., 2025; Andreae et al., 2015; He et al., 2023; Shtabkin et al., 2020; Shi et al., 2019). However, none of these platforms can fully resolve the three-dimensional distribution of O₃ pollution or reliably reflect regional pollution levels and dynamics.

High-altitude mountain atmospheres are less influenced by surrounding ground-level pollution sources, making them suitable for studying pollutant transport and the exchange between the atmospheric boundary layer and the free troposphere. Observations from these sites better reflect regional pollution levels and their variations (Kanaya et al., 2013), while also serving as ideal locations for diagnosing and quantifying the impact of stratosphere-troposphere exchange on lower tropospheric O₃. To date, intensive observational studies have been carried out at several high-altitude mountain stations, such as Mt. Kenya, Mt. Waliguan, Mt. Abu, Mt. Happon, Mt. Bachelor Observatory, Sonnblick, Tanah Rata, Kislovodsk, Krvavec, Arosa, and Pico Mountain Observatory (Okamoto and Tanimoto, 2016). These studies focus on background concentrations of O₃ and its precursors, as well as temporal variations in long-range transport, and meteorological mechanisms. Regarding O₃ background concentrations, measurements from dozens of global observation stations indicate that near-surface O₃ levels in the Northern Hemisphere range between 20–60 ppbv, with specific concentrations varying by geographic location, altitude, and anthropogenic influence (Gaudel et al., 2020). Temporally, regional background and high-altitude stations exhibit increasing trends in mean O₃ concentrations, peak values, exceedance frequencies, and durations of high-O₃ episodes (Cooper et al., 2020). In summary, high-altitude observational campaigns employing multi-platform measurements, coupled with comprehensive analysis and modeling, have advanced our systematic understanding of upper lower-tropospheric air pollution and atmospheric chemistry. These efforts have enhanced knowledge of O₃ and precursors background concentrations, trends, and long-range transport dynamics. However, existing high-altitude stations are unevenly distributed globally, with most located in mid-latitudes. Expanding the global observational network through additional high-altitude stations will be crucial to improving O₃ and precursors monitoring and modeling capabilities. Since the 1990s, China has successively established regional atmospheric background monitoring stations at Longfengshan, Lin'an, Mt. Waliguan, Mt. Tai, Mt. Huang, Mt. Hua, and Mt. Heng, obtaining systematic observational datasets with regional characteristics (Kanaya et al., 2013; Xue et al., 2013; Xue et al., 2011b; Yang et al., 2017; Li et al., 2020). During the summer of 2006, Japanese and domestic researchers collaborated on a comprehensive atmospheric chemistry campaign (MTX2006) at Mt. Tai, providing in-depth insights into photochemical pollution, aerosol physicochemical properties, and the contributions of biomass burning and biogenic emissions to regional pollution (Kanaya et al., 2013). Additionally, Chinese and Japanese researchers conducted continuous measurements of O₃ and CO at Mt. Huang and Mt. Hua, employing observational and modeling approaches to study photochemical pollution in eastern China (Wang et al., 2006). Despite recent advances, critical observational gaps persist, including a sparse high-altitude monitoring network—particularly sparse over southwestern China compared to global standards, a predominance of near-surface measurements of ozone and its precursors with scarce data in the crucial 1000–2000 m layer, and a general lack of vertically resolved analyses of these species. Therefore, it is helpful to establish gradient observation stations at high altitudes of more than 2000 m on the plateau to carry out gradient observation research on O₃ and its precursors, which can help to accurately formulate regional background photochemical pollution control strategies.

Mt. Fanjing is the highest peak in the transitional slope between the Yunnan-Guizhou Plateau and the western Hunan hills, located in northeastern Guizhou Province, China. The summit reaches an elevation of 2570 m a.s.l., while the lowest point is at 500 m a.s.l., resulting in a topographic relief of over 2000 m. This area is influenced by both the Pacific Southeast Monsoon and the Indian Ocean Southwest Monsoon, creating a climatic convergence zone. The fluctuating temperature-humidity regimes provide critical conditions for the overlapping distribution of flora from different biogeographic regions. To investigate the influence of latitudinal zonality, vertical zonality, and slope aspect on meteorological conditions, as well as to comprehensively monitor meteorological dynamics under varying geographic conditions, the Mt. Fanjing Administration established and upgraded a multi-parameter observation network between 2012 and 2013. Due to its remoteness from anthropogenic pollution sources, Mt. Fanjing excellent environmental quality and serves as a relatively “clean area”. The protection and study of this pristine environment will provide valuable insights into atmospheric chemistry in background regions and support air pollution control strategies. However, to date, no project has investigated the vertical gradient



characteristics of O_3 and its precursors. In this study, leveraging the three-dimensional meteorological monitoring system of Mt. Fanjing, we established a gradient observation platform for atmospheric pollutants at the mountain foot, mountainside, and mountaintop, enabling the observation and analysis of vertical profiles of O_3 and its precursors. The main aim of this study is to better understanding the altitudinal gradient variations of O_3 production rates and sensitives. The findings will provide scientific support for the management of regional photochemical pollution and the conservation of ecological integrity in protected areas at the national level.

2 Materials and methods

2.1 Sampling location description

In March 2024, leveraging automated weather stations in the core zone of the Fanjingshan National Nature Reserve, we established observation sites for O_3 and its precursors at Heiwanhe (mountain-foot observatory: $27^{\circ}50'45''N$, $108^{\circ}46'3.87''E$; 550 m a.s.l.), Huixiangping (mountainside observatory: $27^{\circ}54'6.46''N$, $108^{\circ}42'14.33''E$; 1,774 m a.s.l.), and the upper cable car station (mountaintop observatory: $27^{\circ}54'36.59''N$, $108^{\circ}41'44.50''E$; 2,119 m a.s.l.) (*Figure 1*). The mountain-foot observatory is adjacent to the eastern gate of the Fanjingshan Scenic Area, with surrounding homestays and dining facilities to the south. Approximately 150 m east of the station lies an eco-friendly sightseeing vehicle route, while the Heiwan River. The mountainside observations were conducted at the Huixiangping Forest Meteorological Station. This site serves as a resting platform for hikers and is situated in a north-subtropical climate. The mountaintop observatory is positioned at the Wanbaoyan viewing platform near the upper cable car station. Within a 200 m radius, minor service facilities, including small restaurants and shops, exist. Collectively, the vertical gradient monitoring stations for O_3 and its precursors on Mt. Fanjing are situated far from anthropogenic sources. The observational data thus reliably represent the general altitudinal distribution patterns of O_3 and its precursors in the background region of the Yunnan-Guizhou Plateau. From March 1st to August 31th 2024, continuous measurements were conducted for O_3 , NO_x , CO, 57 VOCs species, and meteorological parameters.

2.2 In-situ synchronous observations of O_3 and its precursors

O_3 concentrations were quantified employing a ultraviolet (UV) absorption spectrometer (EAQM-3000; Jiangsu Skyray Instrument Co., Ltd., Jiangsu, China), a gold-standard technique recognized for its exceptional precision in tropospheric ozone monitoring. NO and NO_2 were detected using a high-resolution chemiluminescence analyzer (EAQM-2000; Jiangsu Skyray Instrument Co., Ltd., Jiangsu, China), specifically optimized for precise measurement of trace atmospheric constituents. CO levels were determined via a non-dispersive infrared photometer (EAQM-4000; Jiangsu Skyray Instrument Co., Ltd., Jiangsu, China) incorporating gas filter correlation technology, significantly improving molecular specificity and minimizing cross-interference from other atmospheric gases. All analytical instruments were subjected to rigorous calibration procedures. The experimental design and analytical methodology strictly adhered to China's National Environmental Protection Standards, with all QA/QC protocols implemented in accordance with Class I monitoring station requirements as stipulated by the Ministry of Ecology and Environment of China. Meteorological parameters, including ambient temperature, relative humidity, atmospheric pressure, wind speed and wind direction, were continuously monitored and recorded using an automated weather station (WZZ-M6; Jiangsu Weizhihuan Energy Technology Co., Ltd., Jiangsu, China).

VOCs were measured in real time using an Ntrace VsCMS6500 VOCs automated monitoring system (Ntrace VsCMS6500; Nutech (China) Co.,Ltd., Changsha, China). The instrument employs a unique “dual-column separation + center cutting + single FID” design, enabling comprehensive separation and analysis of VOCs compounds. Ambient air or calibration gas is introduced at a constant flow rate into the sampling system, where interfering substances are removed via cryogenic trapping, followed by enrichment of target analytes. Subsequently, the enriched compounds are thermally desorbed, separated by gas chromatography (GC), and quantified using a hydrogen flame ionization detector (FID). External standard calibration was applied for quantification. A series of standard gases (0.5, 2, 4, 6, 8, and 10 nmol/mol) were analyzed in triplicate from low to high concentrations. Calibration curves were constructed using least-squares regression, with target compound concentrations as the abscissa and peak areas as the ordinate. The correlation coefficients (R^2) of all calibration curves were ≥ 0.98 . The relative error between the measured and reference values at the lowest calibration point was $\leq 15\%$. Method



detection limit (MDL): ≤ 0.1 nmol/mol for 90% of the species. In summary, the system enables continuous sampling, processing, and detection of ambient VOCs, providing reliable quantification of 57 photochemical precursors.

2.3 Random forest modeling and SHAP-based interpretation

Random Forest (RF), an ensemble learning algorithm, improves predictive accuracy and robustness by aggregating predictions from multiple decision trees (Breiman, 2001; Zhang et al., 2025). The method employs bootstrap sampling and random feature selection to generate diverse trees, reducing overfitting while remaining applicable to both classification and regression tasks. Feature importance is quantified via two primary metrics: (1) the mean decrease in impurity (e.g., Gini index or entropy) at nodes where a feature splits data, or (2) the reduction in out-of-bag accuracy upon feature permutation. We implemented RF to evaluate meteorological and precursor drivers of O_3 variability using the R packages “randomForest”, “rfPermute”, and “A3”. A fixed random seed (123) ensured reproducibility. Initial parameter optimization used a 7:3 training-validation split, with the final model trained on the full dataset (parameters: importance = TRUE, ntree = 500, nrep = 1000). Model significance and cross-validated R^2 were assessed via 1,000 permutation tests. To elucidate the sensitivity and response dynamics between predictor variables and outcomes within the RF model, we applied SHAP (Shapley Additive Explanations), which provide both local and global insights, elucidating nonlinear predictor-response relationships in complex models like RF (Lundberg et al., 2020). In our analysis, positive SHAP values indicate O_3 -enhancing effects, while negative values denote suppression, with magnitudes reflecting relative contribution strength.

2.4 Concentration-Weighted Trajectory CWT model

The Concentration-Weighted Trajectory (CWT) model has been widely employed to evaluate the regional transport of air pollutants by integrating backward trajectory analysis with observed chemical concentration data (Xing et al., 2024; Wu et al., 2025; Zhu et al., 2023; Jain et al., 2023; Seibert et al., 1994). Prior to conducting CWT analyses, backward trajectories were computed for each sampling site. The requisite meteorological input datasets for the HYSPLIT model (v4.9 Hybrid Single-Particle Lagrangian Integrated Trajectory) were obtained from the National Oceanic and Atmospheric Administration (NOAA) Global Data Assimilation System (GDAS) archive (available at <ftp://arlftp.arl.hq.noaa.gov/pub/archives/gdas1>; last accessed: 25 May 2025). For the backward trajectory analysis, 36-h air masses arriving at each monitoring site during the sampling period were simulated, with 24 hourly initiations per day (from 00:00 to 23:00 UTC) at a temporal resolution of 1 hour. For CWT analysis, we employed the widely-used “openair” R package, which provides a user-friendly interface for comprehensive air quality data analysis. Additional technical specifications pertaining to the CWT methodology are available in previously published works (Zheng et al., 2019; Zheng et al., 2018).

2.5 Observation-based chemical box model

Comprehensive analysis of in situ O_3 production and loss pathways, and O_3 formation regime was conducted through an Observation-Based Model (OBM) framework integrated with the Master Chemical Mechanism (MCM v3.3.1), hereafter referred to as OBM-MCM. The detailed chemical mechanism, publicly available through the MCM repository (<http://mcm.york.ac.uk>, accessed 19 May 2025), encompasses an extensive reaction network comprising >5,800 chemical species participating in approximately 17,000 elementary reactions. Photolysis rate coefficients (J-values) were parameterized as a function of solar zenith angle and altitude using high-resolution lookup tables precomputed by the Tropospheric Ultraviolet and Visible radiative transfer model (Wolfe et al., 2016). Although the model explicitly accounted for boundary layer dilution processes, the zero-dimensional (0-D) formulation inherently excluded both vertical and horizontal advective transport mechanisms. Model constraints were rigorously defined by empirically measured meteorological parameters (temperature, relative humidity, and pressure) and precisely quantified trace gas concentrations (NO , NO_2 , CO , and speciated VOCs). To minimize initialization artifacts and achieve robust chemical steady-state conditions, all simulations incorporated a 72-hour spin-up period prior to formal analysis.

High-temporal-resolution observational data encompassing O_3 , NO_x , CO , speciated VOCs, along with meteorological parameters (temperature, relative humidity, pressure) and photochemically derived J_{NO_2} values were employed as inputs for



the OBM-MCM. This framework enabled precise quantification of in situ O_3 formation and consumption kinetics. To address the confounding influence of NO-mediated O_3 titration, we computed the net formation rate of odd oxygen ($O_3 + NO_2$) rather than considering O_3 in isolation. Each simulation incorporated a four-day model spin-up period to achieve thermodynamic equilibrium for unmeasured reactive intermediates, particularly hydroxyl radicals (OH), thereby ensuring robust chemical initialization. The relative contributions of local photochemical production versus regional transport were analytically decoupled through the relation: $R_{trans} = R_{meas} - R_{chem}$, where R_{trans} represents transport-derived flux, R_{meas} denotes observed O_3 variation, and R_{chem} signifies chemically driven formation rates. Comprehensive details regarding the computational chemistry methodology are available in the literature (Zhou et al., 2024; Zhong et al., 2025; Liu et al., 2022; Chen et al., 2020; Xue et al., 2014). Mechanistic analysis revealed that O_3 production primarily occurred via NO_x oxidation by peroxy radicals ($HO_2 + RO_2$), while dominant loss pathways involved photolytic decomposition coupled with reactive depletion through NO_2 , OH, HO_2 and unsaturated VOCs.

The relationship between O_3 and its precursors was rigorously evaluated using two complementary approaches: 1) The relative incremental reactivity (RIR), a diagnostic metric for assessing the sensitivity of O_3 formation to precursor emissions, is defined as the ratio of the change in O_3 production rate to the corresponding change in precursor concentration (Liu et al., 2020; Lu et al., 2019; Guo et al., 2017; Wang et al., 2017). 2) The Empirical Kinetics Modeling Approach (EKMA), which facilitated systematic mapping of O_3 - NO_x -VOC response surfaces to identify optimal emission control strategies. For each pollution episode, baseline conditions were established through temporal averaging of diurnal measurement profiles, forming the foundation for subsequent perturbation modeling. Our comprehensive sensitivity analysis employed a factorial design with 400 discrete emission scenarios, methodically varying anthropogenic hydrocarbon (AHC) and NO_x concentrations from ambient levels to 200% in 10% increments. This ensemble approach enabled high-resolution mapping of maximum O_3 formation isopleths, revealing the nonlinear coupling between precursor emissions and O_3 production.

3 Results and discussions

3.1 Observational overview

Mt. Fanjing exhibits a steep vertical meteorological gradient, manifesting as distinct conditions across altitudes: typically sunny conditions mountain foot, fog enveloping mountainside, and frequent rainfall at mountaintop (**Figures S1–S3**). Observational data revealed inverse altitudinal gradients for both wind speed and atmospheric pressure relative to temperature. Measured temperatures were 22.4 ± 5.1 , 16.8 ± 4.2 , and $15.0 \pm 3.9^\circ\text{C}$ at mountain foot, mountainside, and mountaintop, respectively (**Figure 2A**). Corresponding atmospheric pressures were 945 ± 4.5 , 819 ± 2.8 , and 783 ± 2.6 hPa (**Figure 2D**), while corresponding wind speeds were 0.6 ± 0.4 , 0.7 ± 0.6 , and 1.4 ± 1.4 m s⁻¹ (**Figure 2B**). Relative humidity peaked at mountainside ($91.7 \pm 9.7\%$), followed by the mountaintop ($90.1 \pm 9.7\%$), and was lowest at mountain foot ($86.5 \pm 12.2\%$) (**Figure 2C**). These observed vertical gradients in meteorological parameters may govern O_3 dynamics through modulation of both in-situ photochemical processes and advective transport mechanisms.

The campaign-averaged O_3 concentrations at mountain foot, mountainside, and mountaintop were 14.8 ± 15.2 , 30.4 ± 13.4 , and 40.2 ± 14.7 ppb, respectively, demonstrating a pronounced positive altitudinal gradient (**Figure 2E**). The measurements revealed significantly lower O_3 concentrations at mountain foot and mountainside relative to corresponding elevations on Mt. Tai (Xue et al., 2022; Wu et al., 2024) and Mt. Hua (Wu et al., 2025), underscoring the differential influence of regional pollution background and local photochemical processes on mountain atmospheric composition. The O_3 concentration measured at mountaintop was closely comparable to the national average background O_3 concentration (40.3 ± 11.9 ppb) (Chen et al., 2025). This alignment indicates that the high-altitude atmospheric environment at this site functions as an effective regional receptor for well-mixed tropospheric air masses, minimally influenced by proximal emission sources. As shown in **Figure 3**, the O_3 concentrations at mountaintop significantly exceeded those at Tanah Rata, La Quiaca Observatorio, Mt. Kenya and Lulin (Okamoto and Tanimoto, 2016), approached levels observed at Assekrem, Mt. Abu and Mt. Huang (Okamoto and Tanimoto, 2016; Gao et al., 2017; Li et al., 2007), but remained substantially lower than those recorded at Mt. Bachelor Observatory, Mt. Hua, Mt. Waliguan, Mt. Haplo, Jungfrauoch, Monte Cimone, Nanling, Mt. Tai, and Shennongjia (Gong et



al., 2018; Lyu et al., 2021; Okamoto and Tanimoto, 2016; Sun et al., 2016; Xu et al., 2016). In contrast to O₃, concentrations of precursors demonstrated a pronounced negative altitudinal gradient, declining progressively from the mountain foot to the mountainside, and mountaintop. Measured concentrations were as follows: VOCs (12.8±8.3, 5.1±1.9, 9.4±25.4 ppb) (**Figure 2F**), NO_x (8.0±11.8, 1.2±3.8, 1.3±1.7 ppb) (**Figure 2G**), and CO (0.4±0.2, 0.3±0.5, 0.3±0.2 ppm) (**Figure 2H**) at mountain foot, mountainside, and mountaintop, respectively. The CO concentrations at mountaintop were significantly higher than those recorded at Mt. Happon, Izaña, Zugspitze-Gipfel (Okamoto and Tanimoto, 2016), but marginally lower than levels observed at Mt. Tai (Wu et al., 2024; Kanaya et al., 2013). This pattern underscores the interplay between localized human activities, geographic context, and chemical processes in mountain environments. The atmospheric reactivity of VOCs reflects each species's ability to participate in chemical reactions. The total calculated OH reactivity were 4.4, 1.5, and 2.0 s⁻¹ at mountain foot, mountainside, and mountaintop, respectively (**Figure S4A**). The OH reactivity in this study was significantly lower than values observed at comparable altitudes of Mt. Tai (Wu et al., 2024) and in major Chinese cities including Beijing, Shanghai, Guangzhou, Chongqing, and Xianghe (Yang et al., 2020; Tan et al., 2019). While alkanes constituted the major fraction of VOCs concentrations-accounting for 43%, 57%, and 51% at mountain foot, mountainside, and mountaintop, respectively (**Figure S5A-C**)-they drove only 8%, 13%, and 16% of the total OH reactivity at these elevations (**Figure S4B and Figure S5D-F**). In contrast, isoprene contributed minimally to VOCs concentrations (3%, 3%, and 1% at mountain foot, mountainside, and mountaintop), yet dominated OH reactivity, accounting for 26%, 46%, and 31% respectively (**Figure S4D and Figure S5D-F**), revealing a striking decoupling between concentration abundance and chemical reactivity. Alkenes and aromatics contributed 41% and 25%, respectively, to the OH reactivity at the mountain foot; 13% and 28% at mountainside; and 28% and 25% at mountaintop (**Figure S4B, E and Figure S5D-F**). Altitudinal variations in reactive VOCs species may drive corresponding differences in photochemical O₃ production.

Meteorological parameters at both mountainside and mountaintop exhibited similar diurnal variations without statistically significant magnitudes. Slight increases in temperature were observed during the daytime, accompanied by slight decreases in wind speed and relative humidity. The diurnal temperature ranges (difference between daytime and nighttime) were 6.8% and 10.8% for the mountainside and mountaintop, respectively (**Figure 4A**). Corresponding diurnal changes in wind speed were -2.0% and -0.4% (**Figure 4B**), and for relative humidity, -1.3% and -3.2% (**Figure 4C**). Contrastively, significant diurnal variations were observed in temperature, wind speed and relative humidity at mountain foot, with amplitudes of 16%, 47% and -11%, respectively (**Figure 4A-C**).

O₃ concentrations revealed pronounced diurnal heterogeneity across altitudinal gradients, characterized by elevated daytime levels at mountain foot contrasting with nocturnal maxima at mountainside and mountaintop (**Figure 4D**). At mountain foot, O₃ concentrations exhibited a unimodal diurnal cycle with concentrations rising steadily from 08:00 local time, peaking at 13:00, maintaining stability until 15:00, then declining to morning baseline levels. In contrast, mountainside and mountaintop O₃ concentrations remained relatively stable with synchronous diurnal cycles: gradual depletion commencing at 08:00, reaching minima near 15:00, followed by progressive recovery to morning concentrations. Notably, during the 12:00-16:00 interval, mountainside minima approached peak concentrations observed at mountain foot, while mountaintop levels significantly exceeded those at both lower elevations. This elevational divergence underscores complex interplay among boundary layer evolution, precursor advection, altitudinal transport pathways, mountain-valley circulations, and synoptic influences in modulating O₃ behavior. O₃ precursor species exhibited a pronounced bimodal diurnal pattern at mountain foot (**Figure 4E-G**). A primary concentration peak occurs during 08:00-09:00, followed by a secondary peak at 19:00-20:00, closely synchronized with tourist influx. Notably, the morning peak in VOCs concentrations at mountaintop lags the mountain foot peak by approximately 1-2 hours. This temporal offset suggests a delay in pollutant accumulation at higher altitudes due to boundary layer transport processes. In contrast, the mountainside, experiencing minimal anthropogenic disturbance (limited to occasional hikers), displayed significantly attenuated diurnal concentration variations in O₃ precursors. The observed non-uniform diurnal variations in O₃ and its precursors across elevational gradients reflect emergent phenomena driven by elevation-dependent shifts in precursor sources, spatially heterogeneous photochemical modulation by solar radiation and temperature, and complex mountain boundary-layer dynamics-including vertical transport, mixing, and altitude-varying sink



strengths through deposition.

3.2 Feature Importance Affecting O₃

The feature importance ranking for O₃ as the dependent variable is presented in *Figure 5A-C*. The top five features were: relative humidity, NO_x, wind speed, temperature, and pressure at mountain foot (*Figure 5A*); temperature, relative humidity, VOCs, CO, and NO_x at mountainside (*Figure 5B*); and temperature, relative humidity, wind speed, VOCs, and wind direction at mountaintop (*Figure 5C*). Overall, meteorological factors exerted a consistently greater overall influence than chemical factors, underscoring the significance of meteorological factors in O₃ variability across the elevation gradient. *Figure 5D-F* displays SHAP summary plots for key features within the random forest model, corresponding to the mountain foot, mountainside, and mountaintop, respectively. These plots characterize the relative influence magnitude, feature prevalence, and effect direction of each variable on predicted O₃ concentrations. SHAP values farther from zero denote stronger feature importance; positive values indicate a positive association with O₃ levels, whereas negative values reflect an inverse relationship. For instance, elevated relative humidity corresponds to diminished SHAP values, indicating an inverse correlation between relative humidity and O₃ concentration. This likely arises from the complex interplay of physical and chemical processes, such as cloud–radiation interactions, air mass transport, and aqueous-phase chemistry (Liu et al., 2025; Luo et al., 2025).

Dependence analysis further quantifies the effects of varying input features on O₃ levels (*Figure S6-S8*). Relative humidity was determined to be the key feature at mountain foot. Specifically, O₃ concentrations exhibited a nonlinear inverse relationship with relative humidity, with notably enhanced O₃ levels under low- relative humidity conditions. Substantially elevated SHAP values further confirmed this relationship when relative humidity fell below approximately 80% at mountain foot (*Figure S6A*), and below 90% at both mountainside (*Figure S7B*) and mountaintop (*Figure S8B*). The relative humidity threshold for O₃ perturbation identified in this study was substantially higher than values reported for Hangzhou (~70%) (Zhang et al., 2024b) and Nanjing (~75%) (Zhang et al., 2024a). As relative humidity increased from 60% to 80%, the corresponding SHAP value diminished from approximately 15 ppb to near zero, indicating a reduction in the influence of relative humidity on O₃ concentration of ~15 ppb at mountain foot, which was approximately twofold higher than that reported in Haikou (~7.5 ppb) (Liu et al., 2025). At higher elevations (mountainside and mountaintop), the same relative humidity increase resulted in a smaller SHAP value decrease, from ~10 ppb to ~5 ppb, corresponding to an influence reduction of ~5 ppb, which remained lower than those recorded. This negative correlation was attributable primarily to the suppression of O₃ formation under elevated RH, which reduced air temperature, shortened peroxy radical chain lengths, diminished NO₂ cycling via aerosol liquid water enhancement, and catalytically destroyed existing O₃ through vapor-phase hydrolysis (Qiu et al., 2025; Yu, 2019; Zhang et al., 2024b; Zhang et al., 2024a). Temperature emerged as the predominant feature at both mountainside and mountaintop, with a notable nonlinear negative correlation (*Figure S7A and S8A*). Temperature exhibited threshold-and elevation-dependent effects on O₃ concentration variations, with critical thresholds of 20 °C (*Figure S6D*), 18 °C (*Figure S7A*), and 15 °C (*Figure S8A*) at mountain foot, mountainside, and mountaintop, respectively. This contrasted with positive temperature-O₃ correlations in urban plains, underscoring the critical role of mountain-specific processes (e.g., valley winds and cloud shading) in modulating O₃ dynamics. Wind speed and direction, collectively explained 21.3% of the O₃ concentration variability at mountaintop, exceeding their explanatory power at mountain foot and mountainside. Particularly, prevailing westerly winds (270° ± 30°) consistently elevated O₃ levels (positive SHAP values), whereas easterly flows (90° ± 30°) suppressed concentrations. This distinct pattern stems from divergent air mass origins: westerlies transported dry, warm air masses enriched with anthropogenic precursors from the industrialized regions of Guizhou, while easterlies originated from the cleaner Wuling Mountains, thereby inhibiting photochemical O₃ production.

Regarding precursor species, O₃ variability was governed predominantly by NO_x at mountain foot, whereas at mountainside and mountaintop, O₃ levels were regulated primarily by VOCs species. Situated within the direct influence zone of near-ground pollution sources, the mountain foot site exhibits elevated NO_x emissions and consequently high atmospheric NO_x concentrations. These increased NO_x levels directly deplete O₃ through titration. Concurrently, high NO_x concentrations



promote the conversion of peroxy radicals to nitrate, thereby suppressing O_3 production. Consequently, variations in NO_x concentration-particularly the titration effect of NO-play a major role in driving the observed O_3 concentration variability. However, with increasing elevation, anthropogenic NO_x emissions intensity diminishes markedly. Concurrently, specific VOCs-reactive precursors and their oxidation products-influenced by enhanced biogenic emissions and transport-can exhibit greater efficiency for O_3 production under low- NO_x conditions. In brief, this systematic shift in precursor contribution with increasing elevation fundamentally reflects the dynamics shaped by photochemical regimes, source-sink distribution patterns, boundary layer physics, and key meteorological drivers within complex mountainous terrain.

3.3 Unveiling the O_3 formation mechanism

Reaction rates substantially influencing O_3 production or destruction are summarized in **Table 1** on a daily average basis. O_3 production rates peaked at mountain foot ($12.83 \text{ ppb hr}^{-1}$), significantly exceeding values observed at mountainside (2.29 ppb hr^{-1}) and mountaintop (2.92 ppb hr^{-1}). This vertical gradient arises primarily from variations in NO concentrations, with significantly higher NO emission intensities at mountain foot-driven by anthropogenic activities-compared to mountainside and mountaintop. Elevated NO concentrations accelerate HO_2 and RO_2 reactions with NO, thereby promoting radical cycling rates and consequently amplifying O_3 production (Wu et al., 2024; Ma et al., 2002). The $HO_2 + NO$ reaction dominated O_3 production pathways across all sites, with mean rates of 6.09 ppb hr^{-1} (47.5%), 1.23 ppb hr^{-1} (53.8%), and 1.48 ppb hr^{-1} (50.8%) at mountain foot, mountainside and mountaintop, respectively. The $RO_2 + NO$ and $CH_3O_2 + NO$ reactions constituted the secondary pathways. Corresponding mean rates were 3.80 and 2.94 ppb hr^{-1} (29.6% and 22.9%) at mountain foot, 0.71 and 0.35 ppb hr^{-1} (31.1% and 15.1%) at mountainside, and 0.98 and 0.46 ppb hr^{-1} (33.4% and 15.8%) at mountaintop. The O_3 removal rate reached 14.77 ppb h^{-1} at mountain foot, significantly higher than rates observed at mountainside (1.94 ppb h^{-1}) and mountaintop (2.64 ppb h^{-1}) (**Table 1**). The dominant removal pathway at mountain foot was the reaction of O_3 with NO_2 (7.14 ppb h^{-1}), accounting for 48.3% of the total removal. This was followed by reactions with 2-Butene (2.79 ppb h^{-1} , 18.9%) and photolysis ($O_3 + hv$, 1.36 ppb h^{-1} , 9.2%). In marked contrast, $O_3 + hv$ was the primary removal mechanism at both mountainside and mountaintop, with mean daily rates of 0.43 ppb h^{-1} (22.3%) and 0.54 ppb h^{-1} (20.4%), respectively. Secondary pathways included $O_3 + NO_2$ (0.36 ppb h^{-1} , 18.4% at mountainside; 0.43 ppb h^{-1} , 16.3% at mountaintop) and $O_3 + 1,2\text{-Dimethylethylene}$ (0.32 ppb h^{-1} , 16.4% at mountainside; 0.38 ppb h^{-1} , 14.5% at mountaintop). This vertical variation is primarily attributed to altitudinal gradients in meteorological parameters (temperature, and solar irradiation) and precursor concentrations. Figure 8 depicts the campaign-mean diurnal cycle of O_3 production, destruction, and net rates at mountain foot, mountainside, and mountaintop. One can see that both O_3 production and destruction occur concurrently during the daytime, peaking around solar noon. At mountain foot, O_3 destruction exceeded production, resulting in net O_3 destruction with a rate of 1.93 ppb hr^{-1} (**Figure 6A and Table 1**), contrasting with net O_3 production at mountainside (0.35 ppb hr^{-1}) (**Figure 6B and Table 1**) and mountaintop (0.29 ppb hr^{-1}) (**Figure 6C and Table 1**). These values were significantly lower than those recorded at Mount Tai (7.4 ppb hr^{-1} at mountain foot, 5.6 ppb hr^{-1} at mountainside and 6.4 ppb hr^{-1} at mountaintop)(Kanaya et al., 2009; Wu et al., 2024), comparable to measurements at Jungfraujoch for midday in winter, spring, and summer (from -0.05 to $+0.4 \text{ ppb hr}^{-1}$)(Parker et al., 2009; Zanis et al., 2003), but substantially higher than the data obtained from Mt. Cimone in June ($2\text{-}3 \text{ ppb d}^{-1}$)(Fischer et al., 2003) and Mauna Loa for four seasons (from -0.8 to -0.4 ppb d^{-1})(Cantrell et al., 1996). The peak net O_3 production at mountain foot occurred around 08:00 (4.4 ppb hr^{-1}), approximately 2-3 hours earlier than that observed at mountainside (3.4 ppb hr^{-1}) and mountaintop (7.3 ppb hr^{-1}), respectively. This temporal offset may be linked to differences in the timing of precursor accumulation driven by boundary-layer development and tourist activity.

To evaluate O_3 control strategies, we conducted a suite of sensitivity simulations quantifying responses of three O_3 metrics-daily maximum hourly concentration (MConc), daily maximum net production rate (ROM), and daily average net production rate (ROD)-to precursor emissions. Analysis of Empirical Kinetic Modeling Approach (EKMA) curves and relative incremental reactivity (RIR) revealed that O_3 formation at mountain foot, mountainside, and mountaintop was predominantly NO_x -limited (**Figure 7**), with corresponding RIR values of 0.57, 0.83, and 0.78 (**Figure 8A**). However, a progressive shift towards increasing VOC sensitivity was observed with increasing altitude. RIR values for VOCs were -0.2,



0.42, and 0.55 at mountain foot, mountainside, and mountaintop, respectively, especially anthropogenic hydrocarbons-sensitive (RIR: -0.12–0.51), followed by reactive aromatics (-0.03–0.06) and alkenes (-0.04–0.04) (**Figure 8B**). This indicates that the availability and reactivity of anthropogenic hydrocarbons, reactive aromatics and alkenes become increasingly critical rate-limiting factors for O₃ production at higher elevations. Negative RIR values for VOCs at mountain foot indicated that O₃ formation is under a strong NO_x-limited regime, implying that VOCs reduction could paradoxically exacerbate O₃ pollution in this region. With increasing elevation, NO_x concentrations decreased more rapidly due to greater distance from emission sources and enhanced dilution. On the sensitivity curve, as NO_x concentrations decline to a certain level, variations in VOCs concentrations exert a proportionally greater influence on the O₃ formation, even when the system remains predominantly within the NO_x-limited regime. This finding underscores the necessity of accounting for spatial heterogeneity in atmospheric chemical processes induced by elevation when evaluating O₃ pollution control strategies in complex terrains, such as high mountains. While lower elevations remain primarily NO_x-controlled, higher elevations exhibit transitional characteristics marked by enhanced VOCs sensitivity. Consequently, implementing altitude-specific precursor reduction strategies—either differentiated or coordinated controls targeting both NO_x and specific VOCs—may prove more effective than uniform regional mitigation measures.

The variations in O₃ mixing ratios are governed predominantly by chemical and physical processes. As depicted in **Figure 9**, a distinct altitudinal gradient in O₃ transport dynamics occurred. During daytime, net O₃ transport at mountain foot was characterized by predominantly inward (convergent) fluxes, which contributed to the accumulation of observed daytime O₃ concentration (**Figure 9A**). This suggests that the foothill area acts as a receptor for O₃ or its precursors, likely influenced by upslope flows advecting air masses from the surrounding lower terrain or valley regions. CWT analysis revealed that O₃ influx predominantly occurs via northeasterly transport pathways (**Figure S9A**), while precursor species are primarily influenced by southwesterly advective transport (**Figure S9B-D**). Conversely, at mountainside and mountaintop, the dominant pattern shifts to outward (divergent) fluxes of O₃, which resulted in slight daytime O₃ decreases at these elevations (**Figure 9B-C**). This signifies that these higher elevation sites function as net sources of O₃ within the local atmospheric column during the day. The efflux is potentially driven by several factors: 1) intensified turbulent mixing and venting within the mountain boundary layer facilitating vertical export, 2) horizontal advection away from the mountain peak by prevailing winds aloft, and/or 3) local photochemical production exceeding titration or deposition losses at these altitudes under strong solar insolation. This vertical divergence in O₃ transport regimes—convergence at the base and divergence aloft—highlights the complex interplay between mountain-induced meteorology (e.g., thermally driven circulations, boundary layer development), atmospheric mixing processes, and photochemistry in governing the spatial distribution and fate of O₃ in mountainous terrain.

4 Conclusions and implications

Based on comprehensive vertical gradient observations and modeling analyses at Mount Fanjing, this study elucidates the characteristics and governing mechanisms of O₃ and its precursors in the high-altitude background atmosphere of the Yunnan–Guizhou Plateau. The results reveal a pronounced positive altitudinal gradient in O₃ concentrations, with values of 14.8 ± 15.2 ppb at 550 m, 30.4 ± 13.4 ppb at 1774 m, and 40.2 ± 14.7 ppb at 2119 m, contrasting with the decreasing trends of precursors such as VOCs, NO_x, and CO. A clear shift in the dominant factors controlling O₃ variability was observed along the elevation gradient. At the mountain foot, O₃ levels were governed mainly by NO_x and RH, with high RH (>80%) suppressing O₃ through lowered temperatures, shortened radical chain lengths, and enhanced aqueous-phase loss. In contrast, temperature and VOCs species became the primary drivers at the mountainside and summit.

O₃ formation regimes also exhibited elevation dependence. Net O₃ destruction prevailed at the mountain foot (-1.93 ppb hr⁻¹), dominated by O₃+NO₂ reactions (48.3% of loss). Conversely, the mountainside and mountaintop showed net production ($+0.35$ and $+0.29$ ppb hr⁻¹), primarily through HO₂/RO₂+NO reactions (>80% of production), with photolysis as the major loss pathway. Chemically, all elevations fell under NO_x-limited conditions, but VOCs sensitivity increased with altitude, with anthropogenic hydrocarbons dominating VOCs reactivity aloft, underscoring their growing importance under low-NO_x conditions. Transport mechanisms further shaped O₃ distribution: daytime convergent flow at lower elevations imported O₃



and precursors via northeasterly winds, while divergent fluxes at higher levels exported O₃ from the mountainside and summit due to turbulent venting and advection. These patterns highlight the critical role of mountain-induced meteorology in regional O₃ dynamics. Our findings underscore the necessity of altitude-specific O₃ mitigation strategies and advocate for expanded high-altitude monitoring to better represent regional background pollution in understudied areas.

Data availability

The datasets generated and/or analyzed during this study, including those underpinning the main figures and supplementary information, are available at <https://doi.org/10.5281/zenodo.17239121>, Yang et al., (2025).

CRediT authorship contribution statement

Y.Y., H.Z. and Y.H.W. conceived and designed the study. Y.Y., H.H., Y.H.W., H.B.L., and X.W.H. conducted the experimental measurements. Y.Y. and H.Z. analyzed the data and drafted the manuscript. H.Z., Z.Y., and Y.Y. performed the box model simulations. All authors contributed to the interpretation of the findings and provided critical feedback on the manuscript.

Competing interests

The authors declare no competing interests, whether financial or non-financial, that could be perceived as influencing the work reported in this study.

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Figures

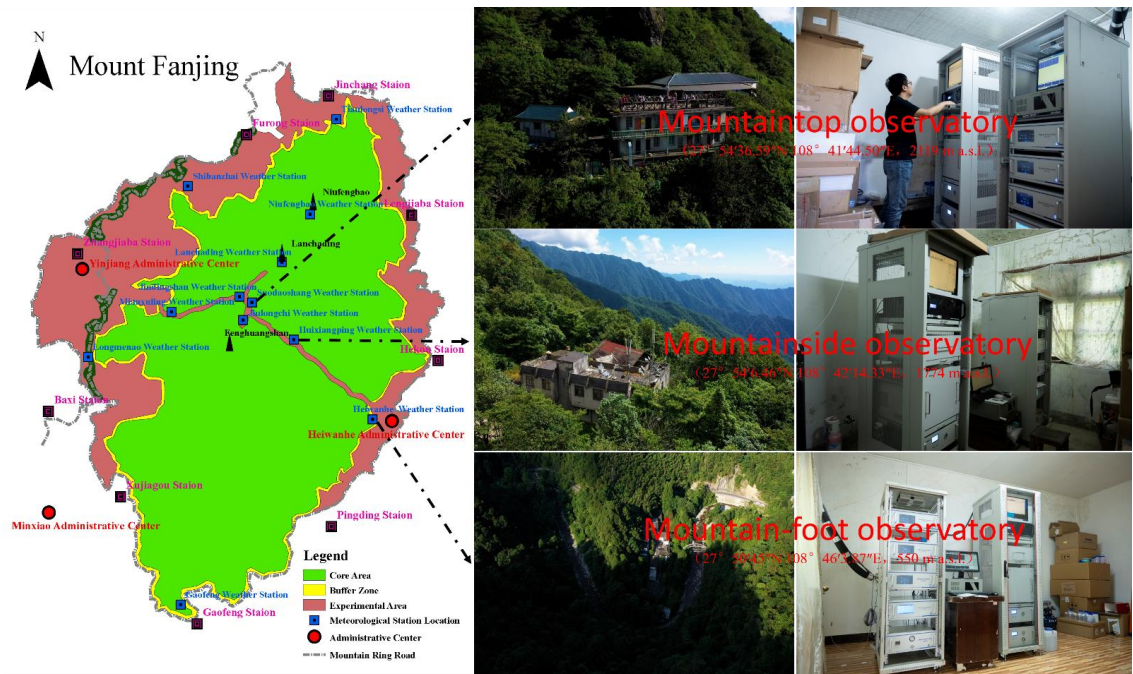
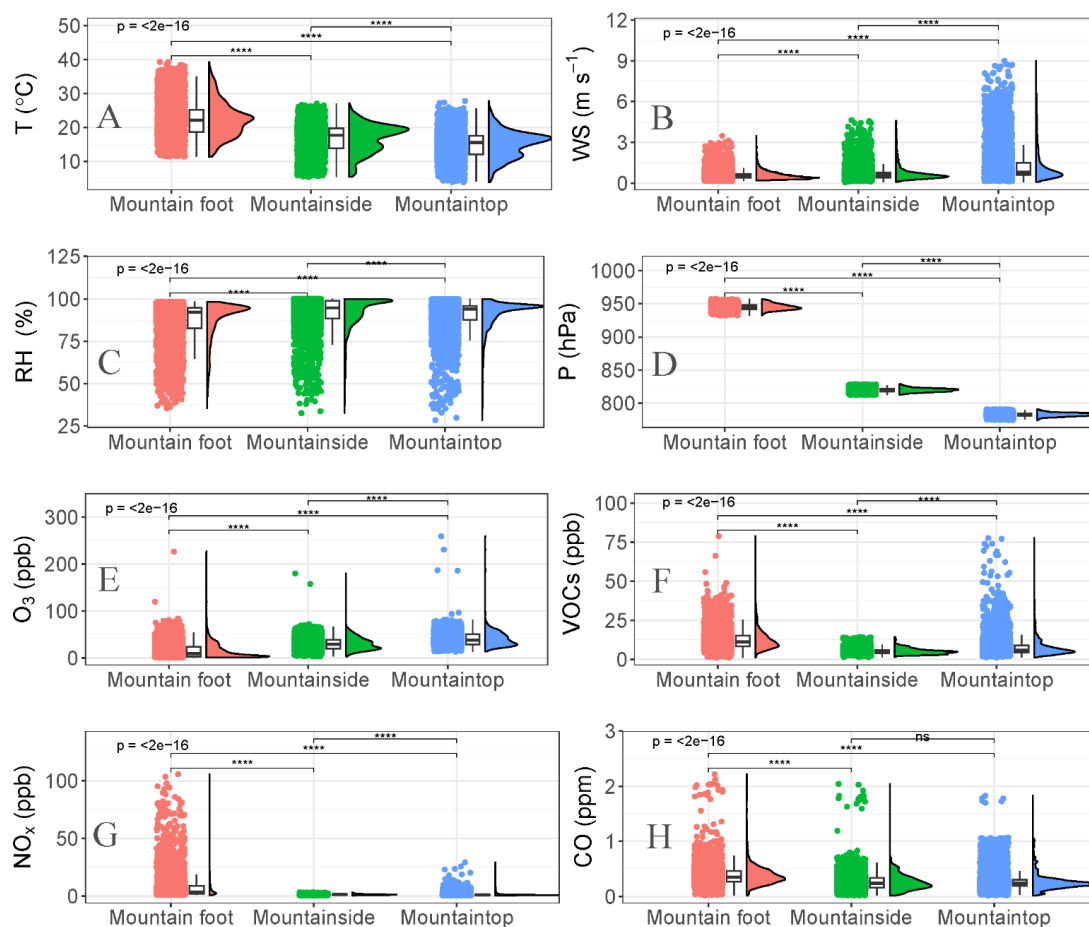
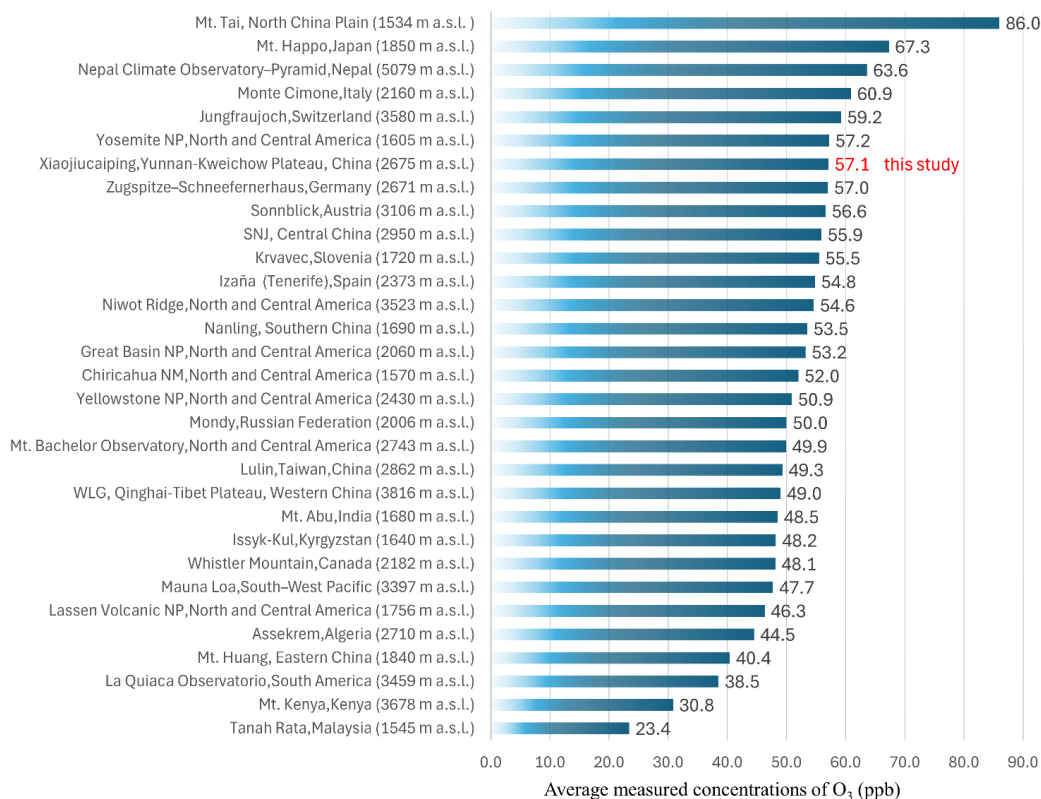


Figure 1. **Left panel:** Distribution of automatic weather stations within the core zone of the Fanjingshan National Nature Reserve. Numbers adjacent to stations denote elevation (image credit: Administration of Guizhou Fanjingshan National Nature Reserve). **Right panel:** External settings and instrumental configurations of O₃ and precursor monitoring stations at the mountain foot (Heiwanhe), mountainside (Huixiangping), and mountaintop (upper cable car station) (image credit: Colorful Guizhou, photographed on 24 July 2024).



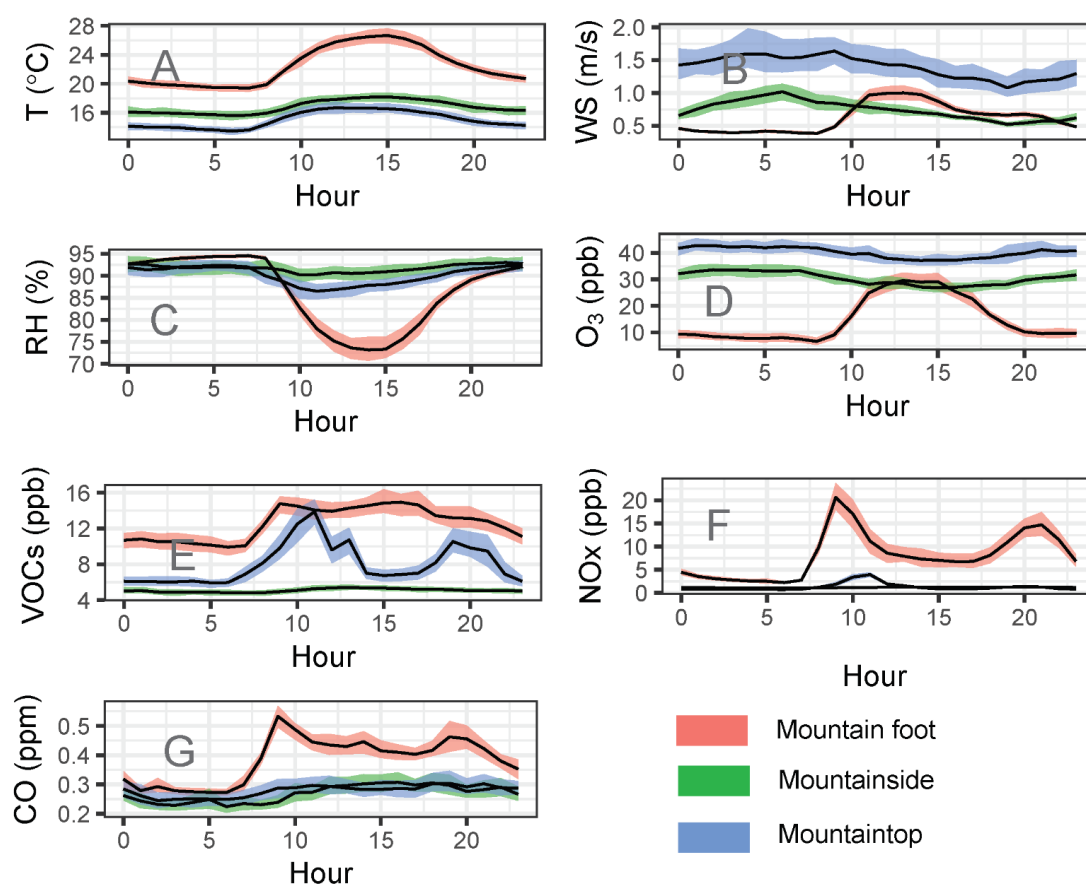
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770 Figure 2. The comparatively visualize the distribution of O_3 and its precursors (VOCs, NO_x and CO) and meteorological
 771 parameters (ambient temperature (T), wind speed (WS), relative humidity (RH) and atmospheric pressure (P)) at the mountain
 772 foot, mountainside and mountaintop. Each seasonal dataset is represented through four complementary graphical elements: 1.
 773 Right “cloud” component: A kernel density estimation illustrating the probability distribution of target species concentrations,
 774 highlighting seasonal variations in central tendency and dispersion; 2. Left “raindrop” component: Horizontally jittered data
 775 points depicting raw measurements, revealing underlying data density and distributional characteristics; 3. Central “boxplot”
 776 component: A Tukey-style boxplot summarizing key statistical descriptors, enabling robust inter-seasonal comparisons of
 777 distributional properties, with medians marked by black lines in the middle and the 25th and 75th percentiles as bounds.
 778 Whiskers extend to $1.5 \times$ the interquartile range; 4. Top “asterisk” component: The six-pointed star denoting the results of a
 779 kruskal.test assessing the statistical significance of differences between two interconnected datasets, ns: $p > 0.05$; *: $p \leq 0.05$;
 780 **: $p \leq 0.01$; ***: $p \leq 0.001$; ****: $p \leq 0.0001$.



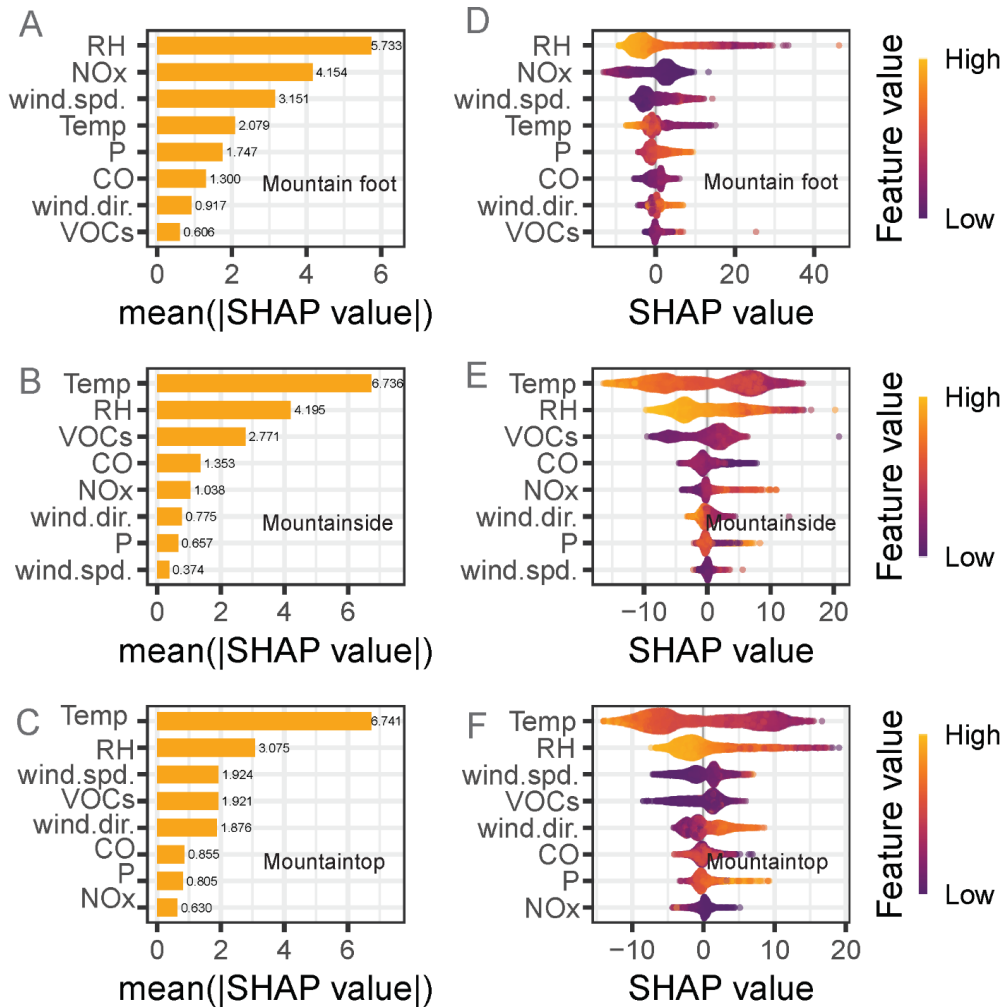
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782 Figure 3. Comparison of O₃ concentrations at Mt. Fanjing summit with those observed at mountain sites across the globe. All
 783 average values are reported in ppb. Data were compiled from published literature (Okamoto and Tanimoto, 2016; Li et al.,
 784 2007; Lyu et al., 2021; Xu et al., 2016; Gong et al., 2018; Sun et al., 2016).

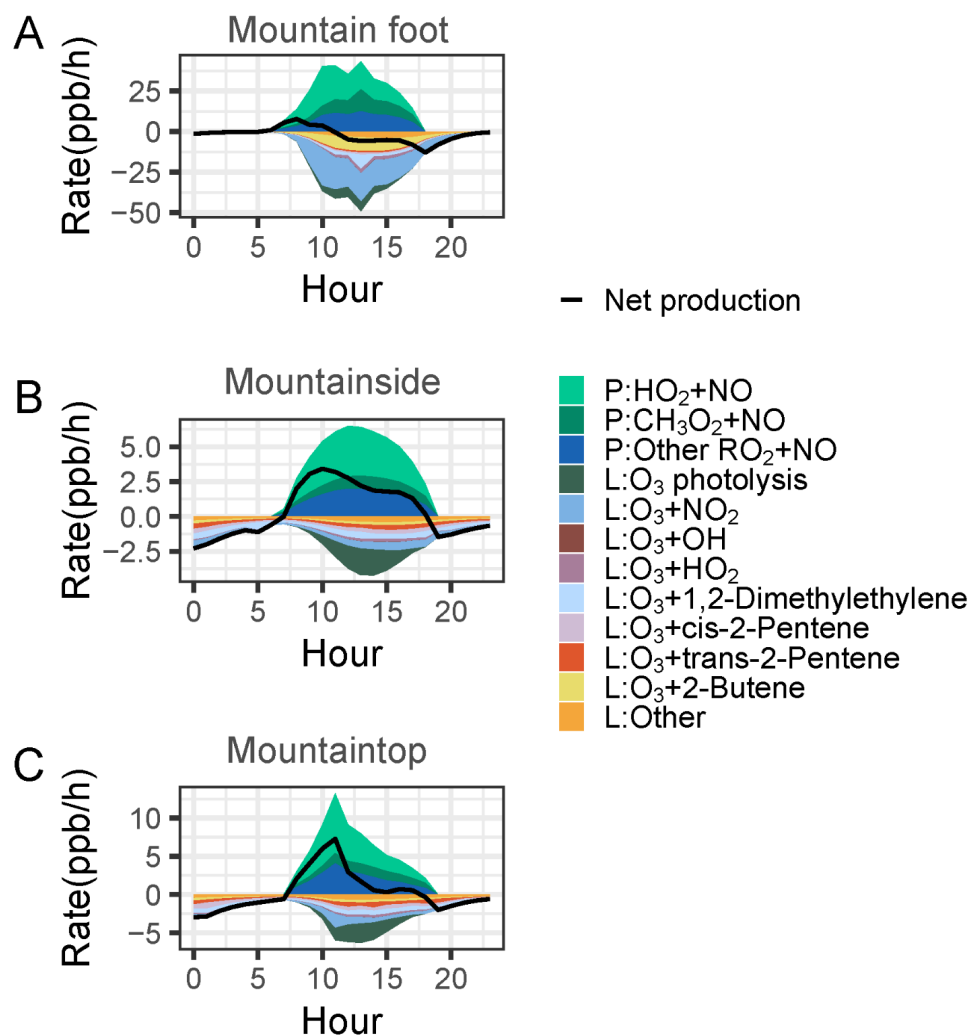


785

786 Figure 4. Diurnal variations in O₃ and its precursors, and meteorological parameters across the respective elevation sites
 787 throughout the observational period. Hourly means are represented by solid lines, with shaded areas denoting 95% confidence
 788 intervals.



789
790 **Figure 5.** **A-C** The mean absolute SHAP (|SHAP|) values of the variables, computed using the SHapley Additive exPlanation
791 (SHAP) method, quantify the relative importance of each feature in influencing O₃ concentrations within the random forest
792 regression model at the respective elevation sites. The relative importance of each feature is quantified by computing the mean
793 absolute SHAP value for each feature across the entire dataset. A higher mean absolute SHAP value indicates greater relative
794 contribution of the corresponding feature to the model's predictions. **D-E** Feature importance analysis of the random forest
795 regression model, with O₃ concentration as the dependent variable, was performed using the SHAP method across different
796 elevation sites. In the summary plot, the absolute magnitude of a feature's SHAP value deviation from zero provides a robust
797 metric for assessing its relative contribution to the model's predictive output. Positive SHAP values indicate features that
798 positively influence the target variable, whereas negative values reflect inhibitory or adverse effects.



799

800 Figure 6. Mean diurnal profiles delineating the simulated contributions of individual pathways to O₃ production and
 801 destruction rates are presented, accompanied by corresponding net O₃ production rates across altitudinal gradients.

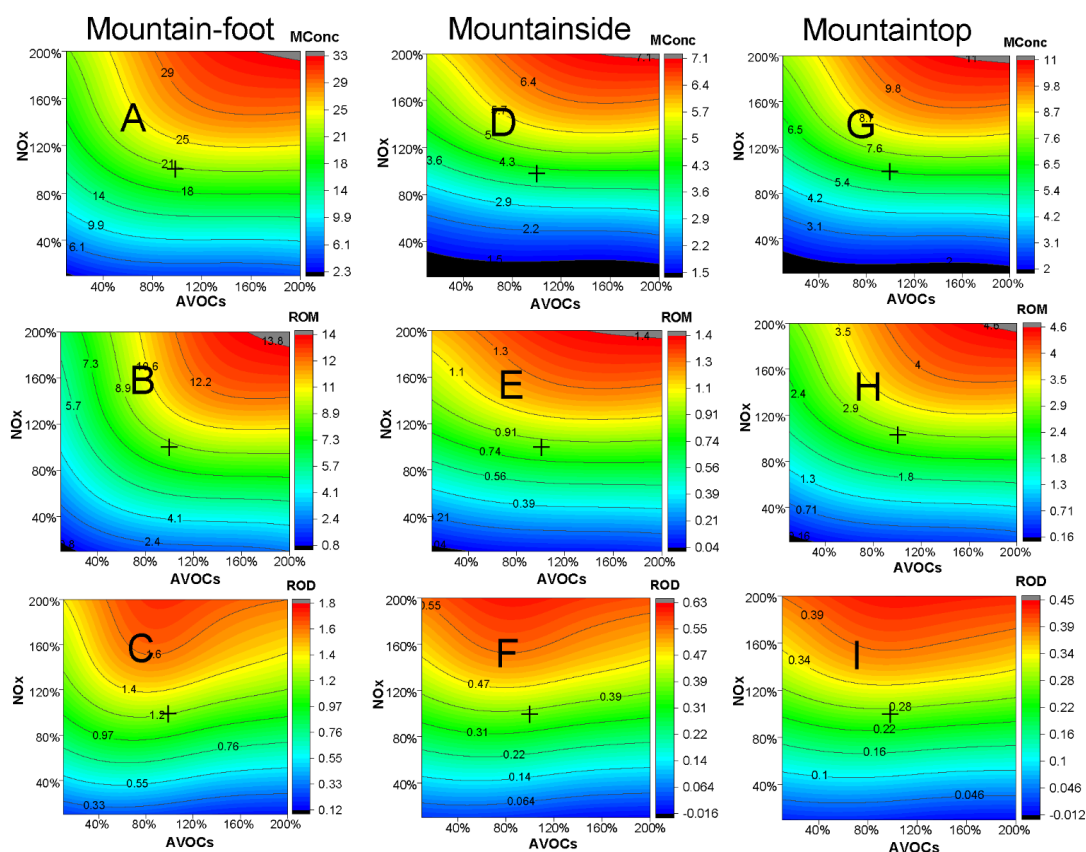
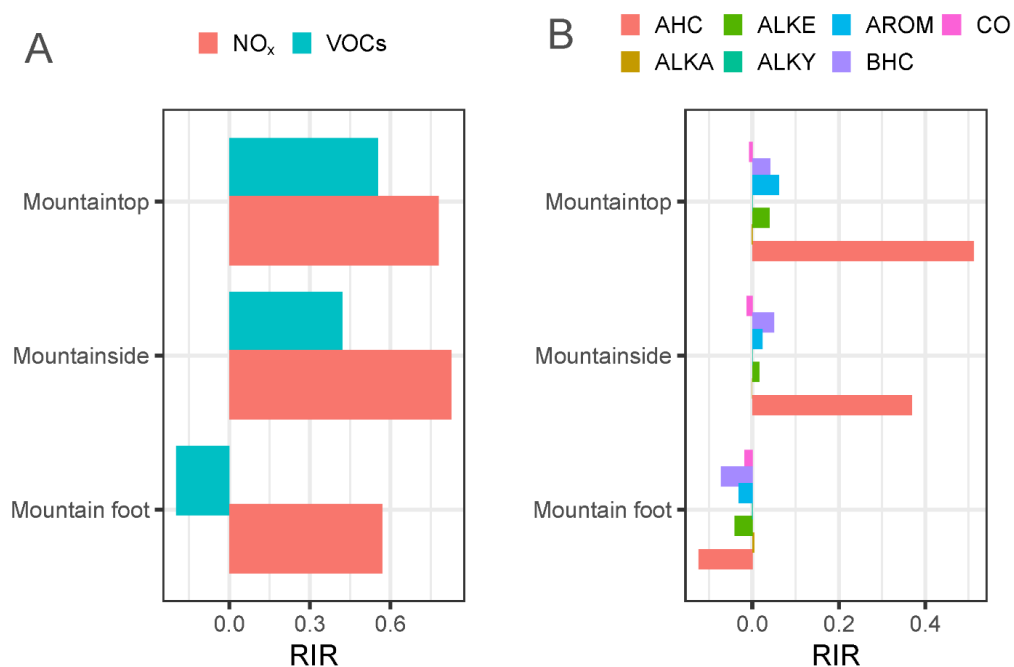


Figure 7. The isopleths of the daily maximum hourly O₃ concentration (MConc) (units: ppb), daily maximum net O₃ production rate (ROM) (units: ppb/h), and daily average net O₃ production rate (ROD) (units: ppb/h) as a function of the relative changes in anthropogenic VOCs (AVOCs) and NO_x across varying elevation sites during the entire observational period. The black cross represents at point 1,1 represents modelled Mconc, ROM and ROD at observed VOC and NO_x concentrations.



808
 809 Figure 8. Model-calculated relative incremental reactivity (RIR) distributions for (A) major O₃ precursor groups and (B)
 810 VOCs sub-groups across altitudinal gradients during the observational campaign. Key to abbreviations: AHC, anthropogenic
 811 hydrocarbons; ALKA, alkanes; ALKE, alkenes; ALKY, acetylene; AROM, reactive aromatics; BHC, biogenic hydrocarbons.

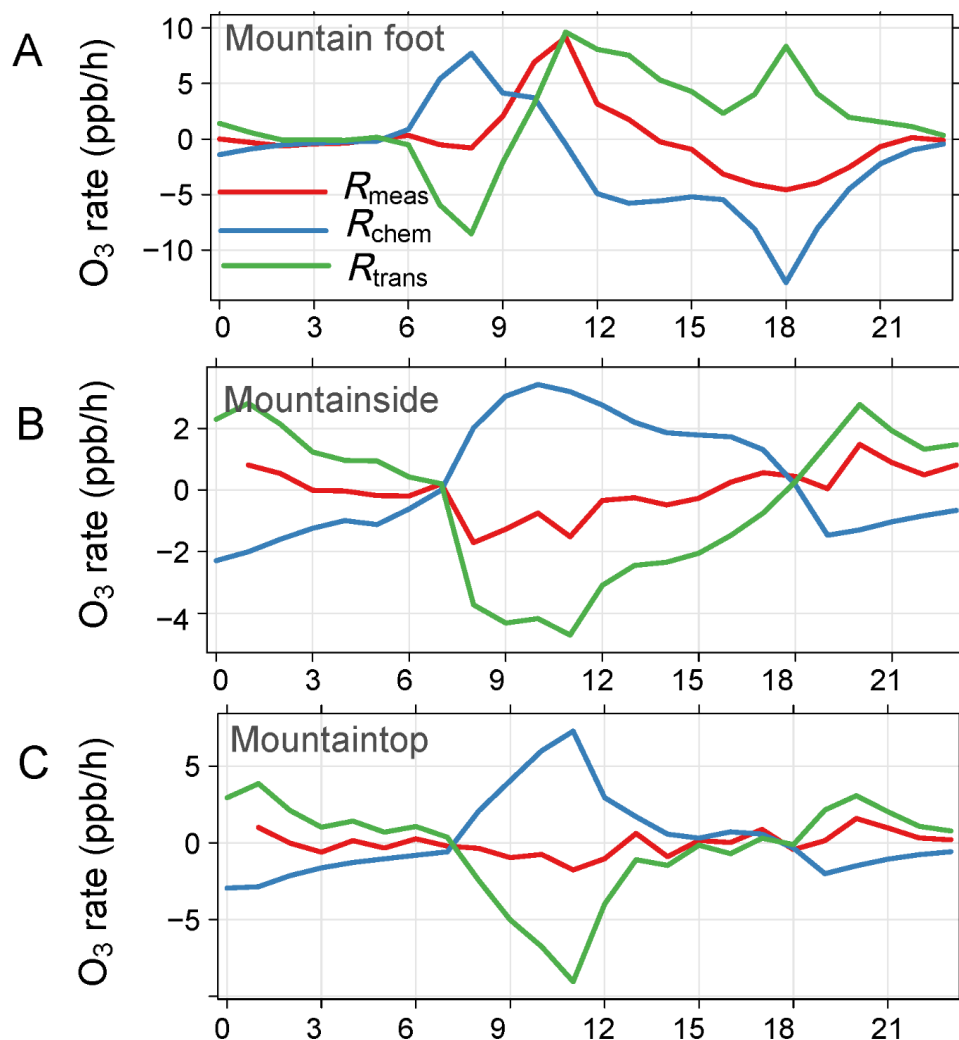


Figure 9. The accumulation of O₃ arises from both local photochemical production and regional transport processes across altitudinal gradients. In the legend, R_{chem} , R_{tran} , and R_{meas} denote the local O₃ photochemical production rate, regional transport contribution, and observed O₃ formation rate, respectively.



828

829 **Tables**

830 Table 1. Daily-averaged chemical reaction rates governing O₃ budget

<i>Reaction</i>	<i>Mountain foot</i>	<i>Mountainside</i>	<i>Mountaintop</i>
<i>Production</i>			
<i>HO₂ + NO</i>	6.09 (47.5%)	1.23 (53.8%)	1.48 (50.8%)
<i>CH₃O₂ + NO</i>	2.94 (22.9%)	0.35 (15.1%)	0.46 (15.8%)
<i>RO₂ + NO</i>	3.80 (29.6%)	0.71 (31.1%)	0.98 (33.4%)
<i>Total</i>	12.83	2.29	2.92
<i>Destruction</i>			
<i>O₃ + hv</i>	1.36 (9.2%)	0.43 (22.3%)	0.54 (20.4%)
<i>O₃ + NO₂</i>	7.14 (48.3%)	0.36 (18.4%)	0.43 (16.3%)
<i>O₃ + OH</i>	0.02 (0.2%)	0.00 (0.2%)	0.00 (0.2%)
<i>O₃ + HO₂</i>	0.58 (3.9%)	0.08 (4.2%)	0.12 (4.4%)
<i>O₃ + 1,2-Dimethylethylene</i>	0.93 (6.3%)	0.32 (16.4%)	0.38 (14.5%)
<i>O₃ + cis-2-Pentene</i>	0.36 (2.5%)	0.20 (10.4%)	0.34 (12.7%)
<i>O₃ + trans-2-Pentene</i>	0.36 (2.4%)	0.21 (11.0%)	0.32 (12.3%)
<i>O₃ + 2-Butene</i>	2.79 (18.9%)	0.13 (6.8%)	0.20 (7.5%)
<i>Other</i>	1.22 (8.3%)	0.20 (10.4%)	0.31 (11.7%)
<i>Total</i>	14.77	1.94	2.64

831 The values represent production rates in ppb/hr, with relative percentage contributions provided in parentheses.