



1 **Evaluating vertically resolved Pandora formaldehyde**
2 **retrievals using airborne and ground-based in situ**
3 **measurements**

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Abstract. Pandora formaldehyde (HCHO) retrievals in the boundary layer provide continuous observations of HCHO vertical distribution, yet validation of these retrievals remains limited. This work evaluates Pandora retrievals of the vertical distribution of formaldehyde (HCHO) against aircraft in situ measurements collected at four U.S. East Coast sites during the summers of 2024 and 2025. Tropospheric columns and vertical profiles are consistent with aircraft-based observations within measurement uncertainty, with agreement that depends critically on the alignment between the Pandora viewing geometry and local horizontal concentration gradients. We further demonstrate a revised interpretation of vertical profile layer heights that improves agreement with aircraft profiles. These results support the use of Pandora HCHO observations for model evaluation and tropospheric chemistry applications.

1 Introduction

Formaldehyde (HCHO) is a key intermediate in the oxidation of volatile organic compounds (VOCs). Because HCHO is produced efficiently during VOC oxidation, its abundance provides insight into the spatial distribution and strength of hydrocarbon emissions as well as photochemical activity in the troposphere. Satellite-derived total column HCHO observations constitute an important tool for constraining VOC emissions (Sfendla et al., 2026; Shim et al., 2005), estimating atmospheric oxidizing capacity (Liao et al., 2019; Travis et al., 2022; Wolfe et al., 2019), and evaluating atmospheric chemistry models (Harkey et al., 2021).

Ground-based observations support assessments and interpretations of satellite HCHO retrievals (De Smedt et al., 2021; Vigouroux et al., 2020). Instruments in the Pandonia Global Network (PGN)—Pandora spectrometers—provide column HCHO measurements (Cede, 2024; Herman et al., 2009; Spinei et al., 2018) that are widely used for validation of satellite products from missions such as Tropospheric Emissions: Monitoring of Pollution (TEMPO) (Szykman and Liu, 2023) and Sentinel-5 Precursor / TROPOspheric Monitoring Instrument (TROPOMI) (Veefkind et al., 2012). In addition to total columns, Pandora instruments also retrieve lower tropospheric columns, near-surface mixing ratios, and vertically resolved profiles of NO₂ and HCHO using sky-scattered measurements in Multi-Axis Differential Optical Absorption Spectroscopy (MAX-DOAS) mode (Honniger, 2004). These boundary layer observations help bridge the gap between total column measurements and near-surface concentrations that are most relevant for air-quality applications. Vertically resolved retrievals also help constrain satellite retrieval errors arising from model based prior profile shape factors.

The profile retrieval technique implemented in the PGN differs from conventional MAX-DOAS inversion approaches. Typically, MAX-DOAS retrievals employ the optimal estimation approach (Frieß et al., 2019; Tirpitz et al., 2021), which involves full radiative transfer modeling to estimate photon path lengths, often requiring assumed a priori trace-gas profiles similar to satellite retrieval frameworks. In contrast, the PGN retrieval algorithm (Cede, 2024; Nowak, 2022) uses a computationally efficient formulation that derives vertical sensitivity information directly from measured differential slant column densities. Comparisons with more computationally intensive MAX-DOAS inversion methods using both synthetic datasets (Frieß et al., 2019) and field observations (Tirpitz et al., 2021) have shown that PGN retrievals achieve comparable performance while requiring substantially less computational effort. The few existing comparisons with aircraft observations—including column-level evaluations (Nowlan et al., 2018) and a recent case study using airborne in situ HCHO and NO₂ profiles (Sebol et al., 2025)—indicate that the retrievals perform well with respect to tropospheric columns, but vertical profile shapes may be biased. A systematic multi-site evaluation of Pandora MAX-DOAS HCHO profile retrievals against independent in situ measurements has not been carried out. In particular, the vertical attribution of measured slant columns and the near-surface representativeness of retrievals remain poorly characterized.



61 In this study we evaluate vertically resolved HCHO retrievals from Pandora MAX-DOAS observations
62 using in situ measurements obtained during recent airborne and ground-based field campaigns. The
63 analysis uses data collected during the Student Airborne Research Program (SARP) field campaigns in
64 2024 and 2025 and the Associating Local Emissions of trace Gases with Regional Observations from
65 Satellites (ALEGROS) campaign in 2024. These campaigns conducted repeat aircraft spiral profiles within
66 the Pandora viewing sectors at multiple sites, providing collocated in situ vertical distributions of HCHO in
67 the lower troposphere across a range of boundary layer conditions. The resulting dataset, spanning four
68 sites along the U.S. East Coast, enables a systematic assessment of Pandora MAX-DOAS HCHO profile
69 retrievals that has not previously been possible.

70 Using this dataset, we assess the consistency between Pandora-derived tropospheric columns, near-surface
71 concentrations, and vertically resolved mixing ratios and the corresponding in situ measurements. A key
72 component of this analysis is a revision to the vertical layer mapping that improves estimation of layer-
73 resolved mixing ratios within the boundary layer, resulting in better consistency between retrieved mixing
74 ratios and in situ observations.

75 2 Methods

76 2.1 Pandora instrumentation and retrieval approach

77 Each Pandora unit records direct sun and sky scattered spectra between roughly 290–540 nm using a
78 Czerny–Turner spectrometer (≈ 0.6 nm FWHM, 0.13 nm pixel). In MAX-DOAS mode, profiling sequences
79 of up to 15 elevation angles occur within ~ 10 min. Low elevation or high pointing zenith angles (PZAs:
80 typically, 80° - 89°) are scanned at 1° intervals. Certain instrument sites have dark and/or reflective
81 obstructions at the 89° viewing zenith angle, in which case the largest zenith angle in the sequence may be
82 capped at 88° or 87° . For instruments included in this analysis, the largest zenith angle scanned was always
83 89° . Scans at higher elevations, namely 75° and 60° , constrain the column abundance in the lower
84 tropospheric. While the column amounts are reported as ‘tropospheric columns’ in PGN Level 2 data files,
85 the upper bound of the vertical sensitivities of MAX-DOAS observations is typically 3–4 km. Zenith
86 observations serve as references for differential slant-column retrievals. Operational measurement
87 schedules for Pandoras also include short MAX-DOAS scans, which involve observations at 0° , 60° , 75°
88 and near-horizon angles, and provide lower tropospheric column loading and near-surface mixing ratios.
89 For the campaign period, all instruments operated on customized schedules that focused on profile scans,
90 interspersed with short MAX-DOAS scans and direct sun measurements.

91 Slant-column densities (SCDs) for HCHO and the air-mass indicator $\text{O}_2\text{--O}_2$ (oxygen collision-induced
92 absorption) are obtained via DOAS fitting in the 328.5–359 nm window. For this work, a modified fitting
93 window (328.5–365 nm) was used to capture the stronger $\text{O}_2\text{--O}_2$ absorption band in the UV. This translated
94 into lower noise and higher precision in $\text{O}_2\text{--O}_2$ slant columns, thus reducing the uncertainty in retrieved
95 tropospheric columns and surface concentrations (see Fig. S1). The zenith observations serve as a
96 reference: measured zenith SCD is subtracted from all other angles to obtain differential slant column
97 densities (dSCDs). This referencing makes Pandora MAX-DOAS retrievals largely insensitive to drifts in
98 wavelength registration and slit resolution. Vertical column densities (VCDs), near-surface mixing ratios
99 and vertical profiles are obtained from dSCDs using the NASA real time algorithm.

100 A schematic of the PGN MAX-DOAS profile retrieval approach is shown in Fig. 1. Full equations
101 corresponding to each retrieval step are available in Cede (2024) and are briefly explained in Appendix A.
102 Rather than inverting measured dSCDs through a full radiative transfer model, the algorithm uses $\text{O}_2\text{--O}_2$ as
103 a direct observational proxy for effective photon path length (Sinreich et al., 2013; Wagner et al., 2004).
104 This is feasible because the vertical distribution of $\text{O}_2\text{--O}_2$ is well-constrained by air density and measured



O₂–O₂ dSCDs encode information about where in the atmosphere photons are scattered and how far they travel. Normalizing trace gas dSCDs by the O₂–O₂ dSCD then partially cancels unknown radiative transfer effects, enabling column and profile estimates without an explicit forward model. The structural framework of this algorithm is physically motivated by single-scattering geometry. It also utilizes empirical formulations of aerosol (Appendix A, Eq. A8) and cloud corrections (Appendix A, Eqs. A11 and A12) that were evaluated against radiative transfer simulations during algorithm development and are treated here as fixed.

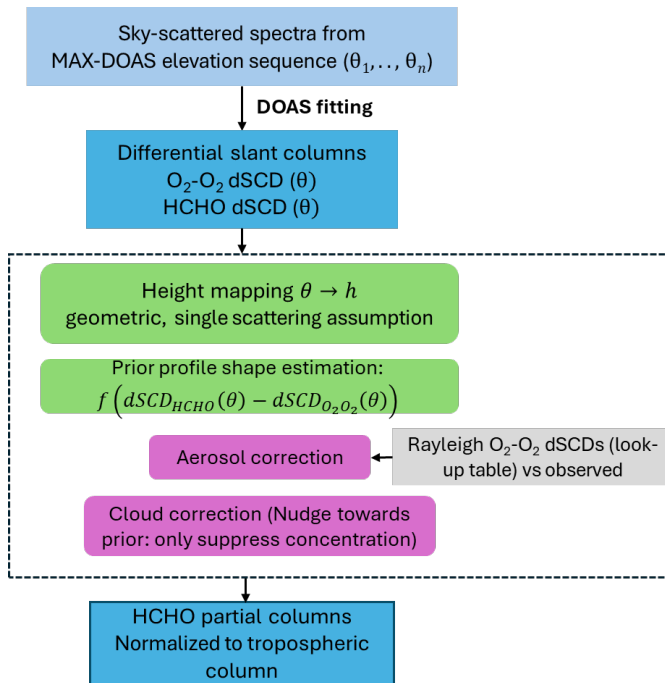


Figure 1. Schematic overview of the PGN MAX-DOAS profile retrieval algorithm. Retrieval steps are color-coded: green indicates analytical derivation grounded in single-scattering geometry; purple indicates empirical correction functions.

2.1.1 Height mapping

As shown in Fig. 1, each elevation angle (θ) in the scan sequence is mapped to a vertical atmospheric layer. Layer centers are defined as half the inferred geometric vertical extent of the mean photon path.

$$h_{mid}(\theta) = \frac{1}{2} \frac{dSCD_{O_2O_2}(\theta) + VCD_{O_2O_2}}{n_{O_2O_2}^{surface}} \cos(\theta) \quad (1)$$

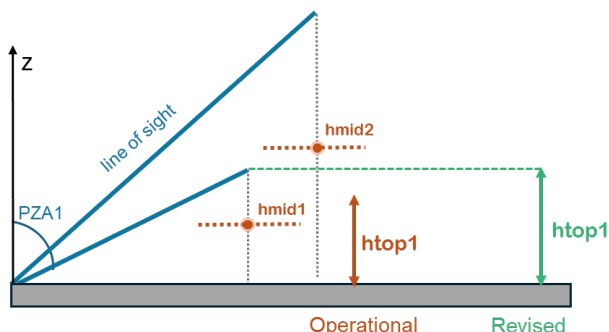
Layer boundaries are then placed midway between adjacent layer centers (Fig. 2):

$$h_{top,i} = \frac{h_{mid,i} + h_{mid,i+1}}{2} \quad (2)$$



123 The topmost layer top is simply set to the highest layer center. The number density per layer multiplied by
124 the layer width provides partial columns per layer, which are then scaled by the independently retrieved
125 tropospheric vertical column density.

126 This mapping implicitly assumes that the vertical sensitivity kernel for a given elevation angle is symmetric
127 about its midpoint and that contributions above and below this height are approximately balanced.
128 However, MAX-DOAS sensitivity kernels are strongly asymmetric and bottom-weighted, with most
129 sensitivity concentrated near the surface and a long, weak tail extending to higher altitudes. Interpreting the
130 geometric proxy as a center height effectively creates strongly overlapping layers associated with low
131 elevation angles. As demonstrated in Section 3.3, this leads to overweighting the contribution of
132 atmospheric layers between 100 m and 1 km to the boundary layer column, resulting in systematic
133 accumulation biases in trace gas loading when column normalization is enforced.



134

135 **Figure 2. A schematic representation of the mapping between pointing zenith angles and atmospheric layer**
136 **heights as calculated in the operational and the revised implementations.**

137 To address this, an alternative height mapping scheme is presented here, which treats the inferred vertical
138 path length as the cumulative altitude range sampled by photons contributing to the measurement. In this
139 formulation, each elevation angle contains the integrated contribution from the surface up to h_{top} , and the
140 vertical profile is constructed from the marginal information between successive cumulative sensitivity
141 envelopes. The layer boundary corresponds to an effective upper sensitivity bound:

$$142 \quad h_{top}(\theta) = \frac{dSCD_{O_2O_2}(\theta) + VCD_{O_2O_2}}{n_{O_2O_2}^{surface}} \cos(\theta) \quad (3)$$

143 Empirical comparison with collocated in situ profiles (Section 3.3) demonstrates that this reinterpretation
144 reduces systematic profile shape mismatch in near-surface layers observed under the h_{mid} formulation.
145 Further, the column closure enforced in the algorithm (final step in Fig. 1) ensures that once the
146 measurement sensitivity is appropriately distributed within the lower layers — where most of the
147 retrievable information resides — the remaining column can be allocated to higher altitudes without
148 introducing significant errors in the overall profile shape. It should be noted that the number of scan angles
149 (and therefore, output layers) does not reflect the true degrees of freedom in the retrieved profile. A full
150 treatment of the information content in the observations used here is beyond the scope of this work, but we
151 emphasize that profiles should be interpreted as representing coarse boundary-layer structure.



2.1.2 Data quality filtering

All instruments used in this study were intensively monitored prior to and during the intercomparison period to ensure good performance (e.g., thermal stability, clean field of view, low humidity). The goal of quality filtering then was to eliminate cloud-contaminated observations. The following thresholds were used to obtain a filtered dataset: (1) root mean square of the fitting residual < 0.0014 . This threshold was the median fitting residual across the instruments in this study, only during the campaign period. (2) Relative uncertainty in the tropospheric column $< 30\%$. (3) Relative uncertainty in the surface mixing ratio $< 30\%$. The third criterion only applies to full profiles, while valid tropospheric column data only need to satisfy the first two criteria.

2.2 Campaign overview

Data from four PGN sites—with multiple aircraft spirals at each location—were used in this study. Site names are listed in Table 1 and hereafter, the short names specified in the table will be used to refer to the Pandora sites. During ALEGROS (2024) and SARP-East (2024, 2025) field campaigns, the Dynamic Aviation King Air B200 aircraft performed spiral ascents/descents over each site, with an in situ payload measuring HCHO, NO₂, O₃, CH₄, CO₂ and geolocation and meteorological parameters. HCHO mixing ratios were measured at 1 Hz using the Compact Airborne Formaldehyde Experiment (CAFE) instrument (St. Clair et al., 2019) which has a precision of 100 ppt/s, a calibration uncertainty of 20% and an offset uncertainty of 100 pptv. Navigation data and static temperature and pressure were measured by the NASA Turbulent Air Motion Measurement System (TAMMS) instrument (Barrick et al., 1996). At the RRC site, an Aeris Ultra was collocated with Pandora 31 to measure the local HCHO concentration at the surface.

Spiral locations were chosen to be in the Pandora line of sight whenever possible. Pandora MAX-DOAS scans within ± 10 min of each aircraft spiral were retained for comparison for all sites other than Essex, MD, where the time coincidence criterion was relaxed to ± 30 min. After filtering Pandora data to meet the quality thresholds defined above, a total of 30 aircraft spirals were used for column comparisons and a subset of 26 were used in profile comparisons.

Table 1: Summary of PGN sites and aircraft spiral comparisons used in this study. “No. of spirals” gives the total collocated spiral count retained after quality filtering; values in parentheses indicate the subset with valid profile retrievals used in Sect. 3.3.

Site name	Short name	Pandora #	No. of spirals (valid profiles)
Rice River Center, Charles City, VA	RRC	31	8 (8)
Wallops Island, VA	Wallops	40	4 (4)
Essex, MD	Essex	75	16 (11)
Chesapeake Bay Bridge Tunnel, Virginia Beach VA	CBBT	255	3 (3)

2.3 Data analysis

For profile comparisons, aircraft-measured pressure and temperature were used to convert Pandora partial columns (mol m^{-2}) to volumetric mixing ratios (ppb). For each comparison, aircraft profiles were vertically smoothed into 200 m bins. Means and standard deviations were calculated on a regular 200 m height grid. If a Pandora instrument performed time-coincident sky scans at multiple azimuths, only scans directed toward the spiral location were used. Each profile was resampled to a regular height grid with 200 m spacing using linear interpolation.



Volumetric HCHO mixing ratios from CAFE were converted to molecules m^{-3} units for calculating integrated columns. Profiles were extended from the lowest altitude sampled by the aircraft to the Pandora station altitude by linear interpolation between the lowest aircraft observation and a surface concentration observation. For the RRC site, the surface concentration was measured by the in situ instrument. For all other sites, Pandora near-surface observations were used.

3 Results and Discussion

3.1 HCHO column and its diurnal evolution

The availability of surface measurements at RRC, along with aircraft spirals aligned with the Pandora line of sight, enables an assessment of all Pandora MAX-DOAS data products at this site: tropospheric columns and column top heights (based on $\text{O}_2\text{-O}_2$ geometric path), near-surface concentrations and vertical profiles. Figure 3a shows agreement between Pandora tropospheric columns and integrated in situ columns ($n = 6$ spirals), when the Pandora viewing direction aligns with the spiral location (shown in Fig. 3b). These observations highlight the importance of sampling geometry when comparing Pandora observations with in situ measurements (Fig. 3 and *Supporting Information*, Fig. S2). If the default MAX-DOAS azimuth for the Pandora at this site was towards southwest, discrepancies between Pandora and in situ columns would be up to a factor of two.

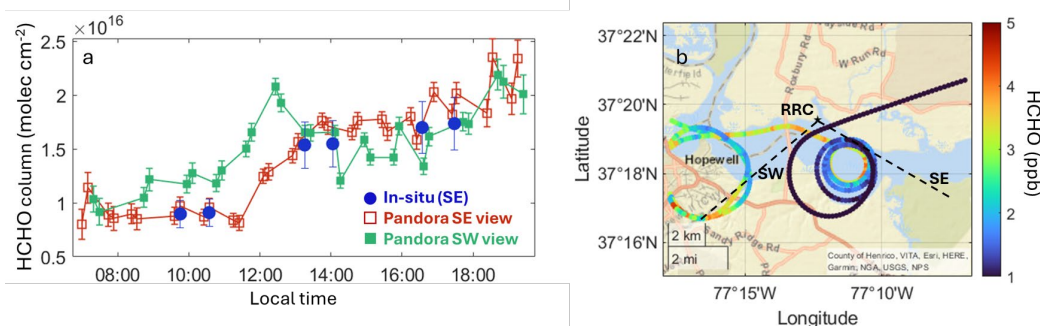


Figure 3. (a) Diurnal variation in HCHO boundary layer column on 06/25/2024 as captured by integrated columns from in situ aircraft observations over RRC and Pandora 31 tropospheric columns observed in southeast and southwest viewing directions. (b) A map depicting the Pandora location and its line-of-sight along the two viewing directions (dashed line). Overlaid are airborne formaldehyde measurements over the area, including the first spiral of the day.

Across all sites, Pandora tropospheric HCHO columns agree with aircraft-derived columns, with most of the data in Fig. 4 falling close to the 1:1 line with a Pearson correlation coefficient of 0.89. Uncertainties due to extending the in situ profiles to the surface are as high as 50% when the lowest altitude of the spiral is over ~ 500 m. The lowest spiral altitude explains a significant portion of the divergence between Pandora and in situ column but as shown in Fig. 3, there can be significant horizontal heterogeneity in HCHO. Therefore, without knowledge of the three-dimensional HCHO fields, discrepancies between in situ and Pandora observations cannot be definitively classified as biases.

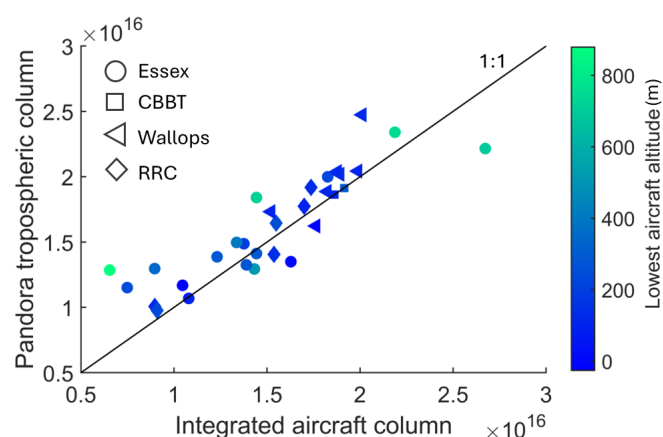


Figure 4. Pandora tropospheric columns versus integrated columns from in situ aircraft observations for all aircraft spirals ($n=30$) valid for comparison. The four comparison sites are denoted by different shapes. Datapoints are shaded by the lowest altitude sampled by the aircraft during a given spiral. The solid line is 1:1.

3.2 Near-surface HCHO concentrations

Figure 5 compares near-surface HCHO mixing ratios retrieved by Pandora 31 at the RRC site with the collocated Aeris measurements for three days each during the summers of 2024 and 2025. Pandora observations are shown for two viewing directions, while the in situ instrument provides a point measurement at the Pandora location. As with tropospheric column comparisons, spatial variability influences agreement between the two datasets. On three of the six days (Fig. 5a, b, and e), the in situ measurements closely track the Pandora near-surface concentrations retrieved in the southeast viewing direction. On two additional days (Fig. 5d and f), the in situ observations instead show stronger agreement with the Pandora southwest retrievals. Figure 5c shows one example where the Pandora and in situ measurements exhibit somewhat different diurnal behavior, suggesting that the Pandora observations were influenced by air masses that differed from those sampled at the instrument location.

Taken together, these comparisons indicate that when the Pandora viewing direction samples air masses similar to those present at the instrument location, the retrieved near-surface HCHO mixing ratios agree well with the in situ measurements. Several aspects of the RRC site likely contribute to this strong agreement. The in situ sampling inlet is located at approximately the same height as the Pandora instrument and in an open environment with minimal nearby obstructions. In addition, the dominant horizontal gradients during the observed periods appear to occur along an east–west axis, which can be effectively sampled by the two Pandora viewing directions given the unobstructed lines of sight. Conversely, this highlights an important limitation when comparing Pandora retrievals with surface in situ monitors: Pandora measurements integrate scattered sunlight along horizontal paths extending several kilometers from the instrument. As a result, agreement with a single surface monitor will depend on the spatial structure of the trace gas field and the alignment of the viewing geometry with local gradients. This consideration becomes even more important for species such as NO_2 , which often exhibit stronger spatial heterogeneity near emission sources.

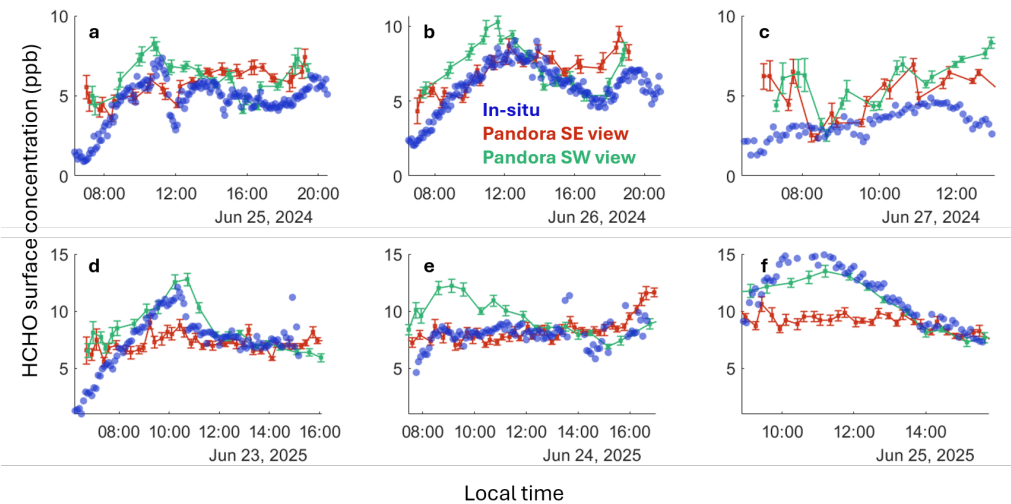
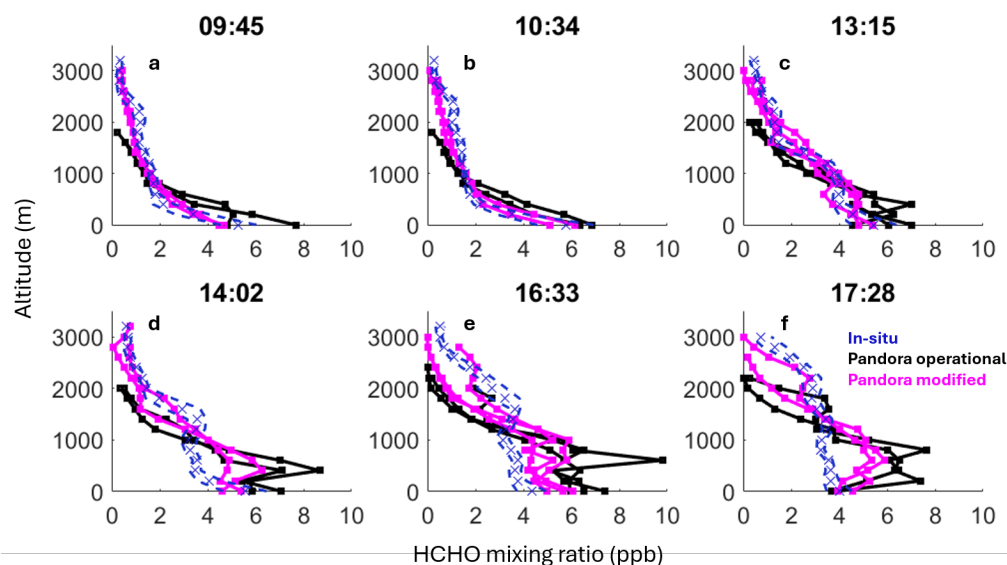


Figure 5. Diurnal variation in HCHO near-surface concentrations observed in 2024: (a) 06/25, (b) 06/26, (c) 06/27 and 2025: (d) 06/23, (e) 06/24, (f) 06/25. Pandora #31 near-surface observations in southeast and southwest viewing directions are shown along with in situ measurements at the Pandora site.

3.3 Vertical structure

To evaluate the ability of Pandora MAX-DOAS retrievals to capture the vertical distribution of HCHO, retrieved profiles are compared with aircraft in situ measurements obtained during spiral ascents and descents conducted near each Pandora site. Figure 6 shows profile comparisons at the RRC site from the 2024 flights. On 06/25, three flights included two spirals each within the Pandora southeast line of sight. The aircraft observations were extended downward from the lowest aircraft altitude using the in situ surface mixing ratio from the ground site. For each spiral, two Pandora profiles were within the comparison window. Each panel shows the aircraft in situ profile alongside retrievals from both the operational Pandora layer mapping and the revised height-mapping approach described in Sect. 2.1. The revised mapping typically captures the shape of the observed vertical distribution, including the near-surface enhancement and the gradient through the boundary layer. The operational mapping, by contrast, concentrates retrieved mixing ratios into the lower layers — a consequence of the way layer boundaries cluster near the surface when defined as midpoints between sensitivity centroids. This produces a vertical shape that bulges out in the first few 100 m above the surface and drops to very low values above ~1 km. This provides the motivation for the revised formulation adopted here. One profile comparison in Fig. 6 (panel f) shows a marked difference between even the modified Pandora profile and the in situ measurements. This seems to be largely a consequence of differences in the integrated column, rather than of profile shape attribution. Figure 3 shows that Pandora column observations in the southeast direction exhibit a fluctuating behavior starting ~16:30 local time, with some data points higher than the aircraft-based column. Pandora near-surface concentrations (Fig. 5) at this time were also larger than the local in situ measurement. This may be due to spatial gradients in HCHO along the Pandora line of sight.



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271 **Figure 6. HCHO vertical profiles at RRC (Charles City, VA) on 06/25/2024 for morning (panels a and b), near-**
 272 **noon (panels c and d), and afternoon spiral (panels e and f) pairs. Each panel is labeled with the midpoint of the**
 273 **spiral time (local). Aircraft in situ measurements are shown as means (blue crosses) bounded by two standard**
 274 **deviations (dashed blue lines). Aircraft profiles are extended to the surface using the in situ near-surface mixing**
 275 **ratio from this site. Pandora retrievals using the operational layer mapping (black) and the revised height**
 276 **mapping (magenta) are shown.**

277 The consistency between retrieved and observed vertical shapes is broadly reproduced across the full set of
 278 campaign sites. Given the largest number of collocated profile comparisons at Essex ($n = 11$), this site
 279 provides sufficient sampling to characterize the median profile shapes and their variability. Figure 7 shows
 280 a summary of these comparisons; individual profiles are included in the Supplement (Fig. S3). Detailed
 281 profile comparisons at Wallops (P40), CBBT (P255), and additional RRC flights are also provided in the
 282 Supplement (Figs. S4–S7). Across these comparisons, the revised mapping generally captures the shape of
 283 the observed boundary-layer profiles. The degree of agreement reflects both the vertical sensitivity of the
 284 retrieval and the degree to which the Pandora viewing direction samples air masses representative of the
 285 spiral location. The wider range in Pandora near-surface retrievals relative to in situ at Essex likely reflects
 286 the limited number of aircraft profiles extending to low altitudes at this site, rather than excess retrieval
 287 variability.

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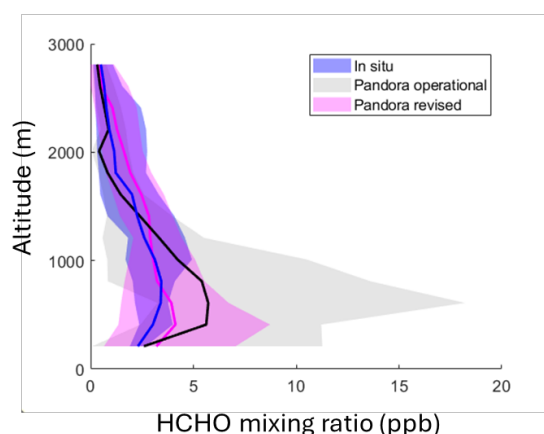


Figure 7. Medians (solid lines) and minimum-maximum ranges (shaded areas) of HCHO vertical profiles at Essex, MD from $n=11$ profile comparisons between 06/17/2024 and 06/22/2024. Pandora observations below the lowest flight altitude are not shown here. Single profiles are shown in Figure S3.

4 Conclusions

This study evaluates vertically resolved HCHO retrievals from Pandora MAX-DOAS observations against aircraft in situ measurements collected during the SARP East 2024 and 2025, and ALEGROS campaigns at multiple U.S. East Coast sites. Across 30 collocated spirals, Pandora tropospheric columns are consistent with aircraft-integrated columns within measurement uncertainty. Spatial representativeness appears to be a key factor controlling the quality of individual comparisons. Careful attention to viewing geometry is essential when interpreting agreements or discrepancies between Pandora and in situ observations.

A key finding of this work is that the retrieved vertical profile shapes can be significantly improved using a revised PZA-to-height mapping formulation. By reinterpreting the O_2-O_2 derived geometric proxy as an upper sensitivity bound rather than a layer midpoint, the revised approach distributes retrieved mixing ratios more consistently with the asymmetric, surface-weighted sensitivity structure of MAX-DOAS observations. Comparison with aircraft profiles demonstrates that the revised Pandora profiles are in substantially better physical agreement with the in situ data than the operational mapping. We recommend adopting the revised height mapping in operational PGN retrievals, to avoid systematic concentration of the HCHO column into the lowest few retrieval layers.

These results have direct implications for satellite validation. Pandora direct-sun and MAX-DOAS HCHO columns, combined with the MAX-DOAS profile shape, provide complementary constraints for evaluating geostationary retrievals from TEMPO and other GEO instruments. While column observations anchor the total burden, the profile shape can help inform the vertical distribution used in satellite air-mass factor calculations. Recent work has shown that NASA's GEOS Composition Forecast (GEOS-CF) model, which provides a priori inputs to TEMPO retrievals, significantly overestimates tropospheric HCHO columns over the Southeast and Northeast Coastal US, driven largely by errors in the modeled vertical distribution in the lower boundary layer (Zhao et al., 2025). The consistency of revised Pandora MAX-DOAS HCHO retrievals with independent aircraft data supports their use in diagnosing and reducing such errors.

Looking ahead, extending this evaluation to NO_2 presents additional challenges: NO_2 exhibits substantially stronger spatial heterogeneity near emission sources, making the path-integrated character of MAX-DOAS observations a more significant complication than for HCHO. In this work, we highlight the challenges in



320 comparing path-integrated remote sensing measurements with point samples. A robust treatment of
321 representativeness issues will likely require experimental designs that explicitly map horizontal structure
322 rather than relying on point-to-path comparisons alone or limiting comparisons to regions that are more
323 homogeneous.

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327 **Appendix A: Pandora MAX-DOAS retrieval**

328 The following describes the mathematical formulation of the PGN MAX-DOAS profile retrieval. This
329 appendix is provided for completeness; a high-level description of the retrieval logic is given in Sect. 2.1.
330 The algorithm derives tropospheric columns, near-surface mixing ratios, and vertically resolved profiles
331 from sequences of sky-scattered spectra using O₂–O₂ as an observational proxy for photon path length,
332 without requiring explicit radiative transfer inversion.

333 **A1. Photon path length from O₂–O₂**

334 In the complete forward problem, each measured differential slant column (dSCD), at elevation angle θ , is
335 an integral of the absorber concentration (n) weighted by a complex photon path distribution:

$$336 \quad dSCD(\theta, \lambda) = \int_0^\infty n(z) W(z; \theta, \lambda, \text{atmosphere}) dz \quad (A1)$$

337 where the weighting function $W(z)$ depends on geometry, Rayleigh scattering, aerosols, surface
338 reflectance, and multiple scattering. Rather than inverting measured differential slant column densities
339 (dSCDs) through a full radiative transfer model, the algorithm uses O₂–O₂ (oxygen collision-induced
340 absorption) as a direct observational proxy for effective photon path length: because the vertical
341 distribution of O₂–O₂ is well-constrained by air density, measured O₂–O₂ dSCDs encode information about
342 where in the atmosphere photons are scattered and how far they travel. The measured O₂–O₂ differential
343 slant column can be approximated as:

$$344 \quad dSCD_{O_2O_2}(\theta) = \int_0^\infty n_{O_2O_2}(z) W(z; \theta) dz \approx n_{O_2O_2}^{\text{surface}} L_{\text{eff}}(\theta) \quad (A2)$$

346 where $L_{\text{eff}}(\theta)$ is an effective photon path length. This approximation replaces the full radiative transfer
347 kernel with a single scalar quantity per elevation angle.

348 **A2. Tropospheric column**

349 The tropospheric column is estimated from trace gas observations at 75° zenith angle, with an O₂–O₂-
350 derived effective AMF. If the 60° measurement is purely single scattering and artifact-free, its AMF from
351 geometry ($1/\cos(60^\circ)$) is 2—then the denominator in the equation below reduces to $dSCD_{O_2O_2}(75)$. The
352 AMF formulation in eq. A3 corrects the 75° observation by comparing the observation at 60° to a Rayleigh,
353 single-scattering atmosphere:

$$354 \quad q_{\text{gas}} = \frac{dSCD_{\text{gas}}(75)}{AMF(75)} = \frac{dSCD_{\text{gas}}(75)}{dSCD_{O_2O_2}(75) - dSCD_{O_2O_2}(60) + 2 \cdot q_{O_2O_2, \text{clim}}} q_{O_2O_2, \text{clim}} \quad (A3)$$

356 **A3. Profile shape prior**

357 For the profile retrieval, dSCDs for the trace gas and O₂–O₂ are normalized to their respective tropospheric
358 columns, to yield differential air mass factors (dAMFs):

$$359 \quad dAMF = \frac{dSCD}{q} \quad (A4)$$

361 The Rayleigh-equivalent dAMF is computed from a pre-calculated lookup table.



A profile shape prior is constructed from the relative deviation of trace gas and O₂–O₂ dAMFs. If HCHO had the same vertical distribution as O₂–O₂, (dAMF_{gas} – dAMF_{O₂O₂}) would vanish; a nonzero residual encodes vertical structure relative to the O₂–O₂ profile. The denominator includes an empirical normalizing term that prevents collapse under Rayleigh-dominated conditions, and the numerator includes an empirically calibrated offset:

$$ps(\theta) = \frac{0.5 + dAMF_{gas}(\theta) - dAMF_{O_2O_2}(\theta)}{dAMF_{O_2O_2}(\theta) + 1}; \quad (A5)$$

The profile shape prior used in subsequent steps is normalized:

$$psn(\theta) = \frac{ps(\theta)}{\max(ps(\theta))} \quad (A6)$$

A4. O₂–O₂ reference and aerosol correction

The number density at each elevation angle is estimated by combining three contributions:

$$n_{gas}(\theta) \propto 2 \times \text{Direct signal} + 0.5 \times \text{Aerosol correction} - \text{O}_2\text{O}_2 \text{ baseline} \quad (A7)$$

The first term is simply the trace gas dSCD observation. The second estimates how much aerosol scattering has reduced the gas dSCD, derived from the deviation of the observed O₂–O₂ AMF from its Rayleigh value (from a precomputed look up table), and scaled by the trace gas tropospheric column:

$$\text{Aerosol correction} = [dAMF_{O_2O_2-Rayleigh}(\theta) - dAMF_{O_2O_2}(\theta)] \times q_{gas} \quad (A8)$$

The third removes the contribution that would exist even if the trace gas had the same profile as O₂–O₂, isolating the residual vertical structure:

$$\text{O}_2\text{O}_2 \text{ baseline} = dAMF_{O_2O_2}(\theta) \times q_{gas} \quad (A9)$$

Combining the above terms with the profile shape prior, and normalizing by the absolute O₂–O₂ slant column:

$$n_{gas}(\theta) = \frac{2dSCD_{gas}(\theta) + 0.5 \times [dAMF_{O_2O_2-Rayleigh}(\theta) - 3 \times dAMF_{O_2O_2}(\theta)] q_{gas}}{dSCD_{O_2O_2}(\theta) + q_{O_2O_2}} \times n_{O_2O_2}^{surface} \times psn(\theta) \quad (A10)$$

A5. Profile shape correction

The profile shape implied by n_{gas} is compared against the O₂–O₂ geometric prior psn . A correction factor is calculated as:

$$cf(\theta) = psn(\theta) - \frac{n_{gas}(\theta)}{\max(n_{gas}(\theta))} \quad (A11)$$

When cf is negative— indicating that the retrieved n_{gas} implies a more elevated profile shape than the O₂–O₂ geometry predicts — a downward correction is applied:

$$n_{gas}(\theta) = n_{gas}(\theta) \times [1 + cf(\theta)], \text{ where } cf < 0 \quad (A12)$$



393 The intended purpose of this asymmetric correction is to suppress concentrations at upper layers that may
394 reflect cloud contamination: if clouds enhance scattering aloft, n_{gas} would be inflated relative to the
395 geometric prior.

396 **A6. Column normalization**

397 Negative n_{gas} values are set to zero. Partial columns are computed using the corresponding number density
398 and layer width. Partial columns are then scaled to enforce closure with the independently retrieved
399 tropospheric column q_{gas} .

400 **A7. Near-surface mixing ratios**

401 Near-surface trace gas mixing ratios are retrieved by extrapolating to a principal zenith angle of 90° — the
402 horizon — from the largest available elevation angle, which is typically 89° . To account for variability in
403 atmospheric mixing conditions, four concentration estimates are computed based on combinations of this
404 extrapolated horizon value and the largest measured elevation angle, under both fully mixed and
405 heterogeneous atmospheric assumptions. The largest value among these four estimates is selected as the
406 reported surface concentration, along with a heterogeneity flag that characterizes the degree of vertical
407 mixing near the surface. Details regarding heterogeneity flags are in Cede {(2024)}.

408



409 **Code and data availability**

410 All data are publicly available online at the following locations:

411 SARP and ALEGROS airborne and ground in situ measurements:

412 <https://www-air.larc.nasa.gov/cgi-bin/ArcView/sarp.2024>

413 <https://www-air.larc.nasa.gov/cgi-bin/ArcView/sarp.2025>

414 Pandora operational data: <http://data.pandonia-global-network.org>

415 Custom Pandora retrievals used in the intercomparison:

416 <https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/NNZ0HB>

417 Code available on request.

418 **Author contributions**

419 TH and AP designed Pandora observation strategies and collected ground in situ measurements, with
420 support from RL and PZ. GW, and ED designed flight plans; GW and JSC collected the CAFE data; and
421 KT collected the TAMMS data. AP conducted the intercomparison analysis and devised the modified
422 retrieval height mapping. BP, JP and ES contributed valuable suggestions on analysis. AP prepared the
423 manuscript with contributions from all co-authors.

424 **Competing interests**

425 At least one of the (co-)authors is a member of the editorial board of Atmospheric Measurement
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