

Retrievals of Ozone in the Troposphere and Lower Stratosphere Using FTIR Observations Over Greenland

Shima Bahramvash Shams^{ID}, Von P. Walden, James W. Hannigan^{ID}, and David D. Turner^{ID}

Abstract—When retrieving geophysical parameters, it is advantageous to have an estimate of prior information that is based on observations with associated uncertainties, but this is often not possible. Long-term ground-based remote sensing measurements and the ozonesonde program at Summit Station, Greenland, provide an opportunity to create a unique framework to retrieve atmospheric ozone using observationally based prior information in the Arctic. This study investigates the potential of using the ground-based polar atmospheric emitted radiance interferometer (P-AERI) to estimate ozone below 10 km. Downlooking or limb-viewing sensors, such as those from satellites, have limited sensitivity to the lower atmosphere; however, uplooking, ground-based instruments provide complementary information to satellite observations to improve trace gas estimates at lower atmospheric levels. Modern-Era Retrospective Analysis for Research and Application, version 2 (MERRA-2) is a reanalysis product that integrates satellite information but also inherits their higher uncertainties at lower atmospheric levels. An observation-based climatology of the uncertainty in the MERRA-2 ozone dataset is estimated using ozonesondes launched at Summit Station. MERRA-2 shows high accuracy in the middle stratosphere but larger uncertainties below 10 km. Retrievals that use spectral radiance measurements from the P-AERI improve the estimates of ozone concentrations in the troposphere and lower stratosphere by using prior information from MERRA-2 and our climatology of MERRA-2 uncertainties as the covariance of the prior. Using ozonesonde observations from 2012 to 2017 at Summit Station, Greenland, the quality of the retrieved results is assessed. Comparisons show that retrieved partial columns reduce the bias of MERRA-2 ozone estimation below 10 km, and the average tropospheric ozone concentration is improved significantly.

Index Terms—Atmospheric measurements, infrared radiometry, remote sensing.

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I. INTRODUCTION

GROUND-BASED remote sensing instruments, such as Fourier transform infrared (FTIR) spectrometers, can provide important information on atmospheric structure. One of the key advantages of ground-based infrared spectrometry compared to measurement from space or aircraft is the small contribution of reflectance and emission from the surface in the measurements. Using atmospheric radiance data over a broad spectral range from FTIR, a variety of atmospheric trace gases and meteorological parameters can be retrieved. In particular, ground-based atmospheric emitted radiance interferometer (AERI) instruments have been used in the analysis of cloud properties [1]–[6], trace gas retrievals such as CO [7], water vapor [8], [9], spectroscopic parameters of H₂O [10], [11], thermodynamic profiling [12]–[14], and radiative forcing associated with CO₂ [15] and CH₄ [16]. AERI instruments are sensitive to the middle to far infrared (3–21 μ m), and as atmospheric emission instruments, they have an advantage over solar FTIR because they can make measurements in the absence of sunlight. Thus, they are well suited for operation at high latitudes where there is no sunlight for a significant portion of the year.

An AERI instrument was deployed in June 2010 at Summit Station, Greenland, as part of the Integrated Characterization of Energy, Clouds, Atmospheric State, and Precipitation at Summit (ICECAPS) project [17] and is called the polar AERI (P-AERI). In this study, we introduce a retrieval framework that takes advantage of the sensitivity of the P-AERI to atmospheric ozone in the lower atmosphere and the year-round operation of the P-AERI. Ozone retrievals from the P-AERI are then compared to an existing reanalysis dataset as well as ozonesonde profiles.

The accurate measurement of tropospheric and lower stratospheric ozone at high latitudes is important for a thorough understanding of stratospheric–tropospheric exchange [18]–[20], ozone variability, long-term trends in ozone, and the parameters and mechanisms that modulate lower level ozone fluctuations [21]–[23]. Ozone is an important component of global climate [24]–[28]. Arctic ozone influences global circulation, surface temperature, and tropospheric weather regimes [29]–[33]. Tropospheric ozone over the Arctic is impacted by anthropogenic pollutants from lower latitudes and stratospheric intrusions [34], [35]. Lower stratospheric ozone is mostly modulated by dynamical

mechanisms and also impacted by ozone-depleting substances [36], [37]. Despite the consensus on the recovery of ozone in the middle and upper stratosphere [38]–[40], trends of decreasing ozone have been reported within lower stratospheric layers [39], [41]–[43]. However, some model simulation studies contradict the decreasing trend because of the high interannual variability at lower stratospheric levels [44]. Many of the studies mentioned above [39]–[43] focused on the lower latitudes because of the complexity of dynamical mechanisms, the uncertainties, and the sparsity of ozone observations in lower stratospheric/tropospheric layers over high latitudes. Thus, improving the estimation of tropospheric and lower stratospheric ozone at high latitudes is important for a complete understanding of ozone variations.

Reanalysis models assimilate observations, including satellite retrievals and meteorological measurements, to provide global coverage of atmospheric properties at fine temporal resolution. However, various uncertainties in the observations and assimilation methods, as well as uncertainties in the parameterizations of the physical processes within the modeling system, contribute to uncertainties in the final reanalysis products [45], [46]. Modern-Era Retrospective Analysis for Research and Application, version 2 (MERRA-2) reanalysis products have fine vertical resolution, global coverage, and agree well with observations at midstratospheric levels [47], [48]. Because of this, many studies have used MERRA-2 to study ozone in the middle stratosphere [48]–[50]. However, because of low sensitivity of spaceborne ozone estimations, MERRA-2 has larger uncertainties in the lower stratosphere and troposphere [47], [48]. The long record of P-AERI measurements (July 2010–April 2021) and ozonesonde profiles (2005–2017) at Summit Station motivates an exploration of the capabilities of the P-AERI to retrieve ozone that can be compared with ozonesonde measurements in the lower atmosphere at high northern latitudes.

Both statistical and physical approaches can be used in the retrieval process [51], [52]. A statistical retrieval uses different techniques from linear regression to more advanced machine learning techniques to identify an empirical relation between measured and unknown variables. Early attempts to retrieve ozone from emission FTIR used lookup tables and regression analysis to estimate ozone concentrations [53]. Although a statistical retrieval could be computationally time effective, they typically do not consider the uncertainties in measurements and do not provide detail description of information content and uncertainties in outputs. On the other hand, a physical retrieval incorporates the uncertainty in the observation along with physical simulations in an iterative manner to converge to an optimal solution. Previously, a combination of regression and physical retrieval was used to retrieve tropospheric column ozone from the surface to 300 hPa from an AERI using a case study from the midlatitudes [54].

In this study, tropospheric and lower stratospheric ozone is retrieved from emission FTIR spectra using the optimal estimation method (OEM) [52]. MERRA-2 ozone profiles and the observation-based climatology of MERRA-2 uncertainties are used as the retrieval's prior information and its uncertainty, respectively. The climatology of MERRA-2 uncertainties is

estimated using 12 years of ozonesonde measurements at Summit Station. The climatological differences show good agreement between MERRA-2 and ozonesondes in the middle stratosphere, but larger uncertainties below 10 km. The larger magnitude of the prior uncertainty (percentage variability) in the lower atmospheric layers puts more weight on P-AERI spectra in the OEM retrieval process, while the smaller uncertainties in the middle stratosphere put more weight on using the prior ozone information from MERRA-2. Thus, the retrieval framework focuses on the portion of the MERRA-2 ozone profile that has higher uncertainties. Moreover, the covariance among layers, as provided by the prior, provides additional constraints to limit the optimal solution and enable the retrieval to achieve a realistic result. To quantify the improvements of the retrieval results, coincident measurements of clear-sky P-AERI spectra and ozonesonde launches at Summit Station, Greenland, are used.

Our study uses the OEM to retrieve ozone from the P-AERI while being constrained by the climatology from the MERRA-2 product. The retrieved ozone columns are compared to ozonesondes and MERRA-2 from the ground to 10 km at Summit Station, Greenland. Section II describes the datasets used in this study, including the P-AERI, the MERRA-2 reanalysis, and ozonesondes. In Section III, the retrieval methodology and procedures are presented. In Section IV, the retrieved results and the details of the information content are discussed. The comparisons to ozonesondes are discussed in Section V. Section VI presents the conclusions of this research study.

II. DATA

This section provides a description of the datasets used in this study (P-AERI, MERRA-2, and ozonesondes), including their uncertainties.

A. Polar AERI

Infrared spectra acquired using the P-AERI at Summit Station, Greenland (72°N, 39°W), are used to retrieve ozone. The P-AERI was designed and built by the Space Science and Engineering Center at the University of Wisconsin–Madison [55]–[57]. The P-AERI was deployed as part of the ICECAPS project in June 2010 [17]. The P-AERI measured downwelling infrared spectral radiance continuously at Summit Station until May 2021. The effective laser wavenumber during ICECAPS was determined to be 15799.33 cm^{-1} , which corresponds to a maximum optical path length of about 1.037 cm. The spectral range is $480\text{--}3000\text{ cm}^{-1}$ ($3\text{--}21\text{ }\mu\text{m}$) with moderate unapodized spectral resolution of 0.48 cm^{-1} [55], [56], [58]. A representative example of a clear-sky measurement of downwelling infrared radiance by the P-AERI is shown in Fig. 1. The radiance at $500\text{--}800\text{ cm}^{-1}$ corresponds to CO_2 and H_2O , $995\text{--}1065\text{ cm}^{-1}$ to the ozone bands, and $1200\text{--}1400\text{ cm}^{-1}$ to CH_4 , N_2O , and H_2O . The radiometric calibration is determined using two well-characterized infrared sources [55]. The radiometric uncertainty of the P-AERI (3σ value) is less than 1% at ambient radiance [55]. The spectral calibration is approximately 1 ppm (standard deviation of

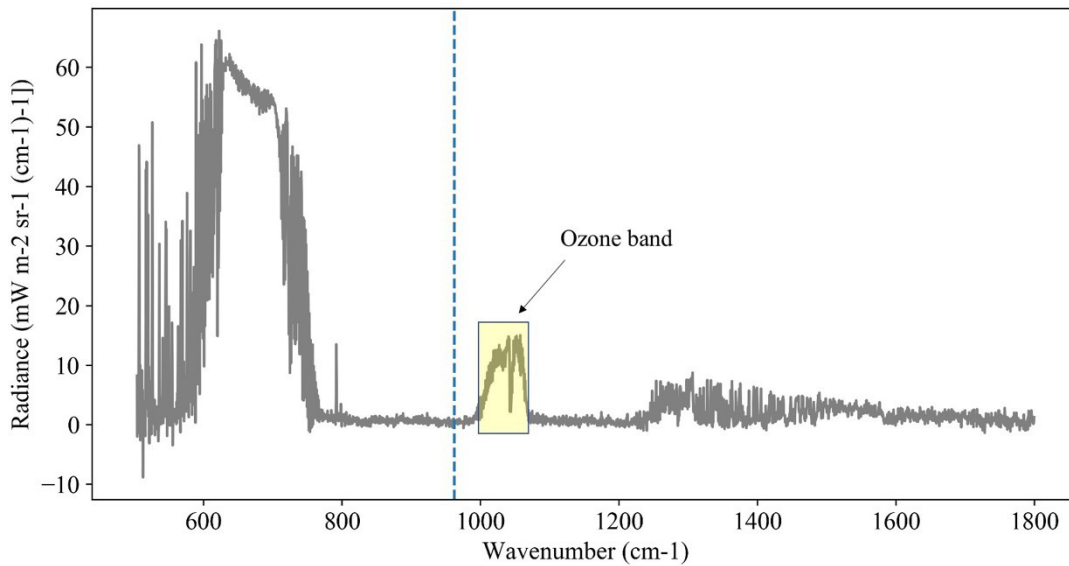


Fig. 1. Spectral range and radiance of P-AERI channel 1. The spectral range of 995–1065 cm^{-1} is the ozone band and it is used for the retrieval. The radiance at 962.36 cm^{-1} is shown by the blue dashed line and used to determine clear-sky conditions.

$\pm 0.014 \text{ cm}^{-1}$). The AERI's noise level increases markedly near the end of the spectral range of its detector due to very small radiance signal (by being in a window channel in clear-sky, dry, and cold conditions) and large random error, which explains the negative radiance around 540 cm^{-1} in Fig. 1.

The P-AERI data from January 2012 to May 2021 have the best calibration (due to a temperature correction that was necessary for the infrared sources from July 2010 through December 2011). In this study, ozone retrievals are performed for 5.5 years from January 2012 to July 2017. The end of the study period corresponds to the termination of the ozonesondes program at Summit Station, which provides the comparison data to assess the quality of retrievals. The P-AERI measurements are taken approximately every 30 s year-round for a total of about 3300 spectra each day. Continuous radiance measurement is an advantage of using P-AERI observations for ozone retrievals. When clouds are present, they contribute radiance in the 9.6- μm ozone band, so it is important to determine the periods of clear-sky conditions to achieve high-quality ozone retrievals. The mean and standard deviation of a 30-min rolling window of the radiance at 962.36 cm^{-1} are used to determine clear-sky conditions because of the low gaseous optical depth at this wavenumber, which make it very sensitive to the emission from overlying clouds. Even very small, condensed water amounts can greatly increase the radiance observed in the 11- μm window [59], and because clouds are seldom spatially uniform, they introduce a temporally varying signal that can be easily identified using thresholds on the mean and standard deviation of the rolling window. Because the infrared emission from atmospheric gases varies slowly with seasons [due to changes in atmospheric temperature and gas concentrations (e.g., water vapor)], thresholds are set for the mean and standard deviation of clear-sky conditions for each month. To reduce the uncertainty in the spectral radiances (and to increase the signal-to-noise ratio), the clear-sky infrared spectra are averaged over 30-min time periods. The statistics of clear-sky spectra are shown in Fig. 2. Most days with clear

sky only have a single period that is cloud-free for more than 30 min, but some days have more clear periods. Using the monthly thresholds, we have identified more than 200 days with at least one 30-min period of clear-sky conditions per year. The retrievals performed in this study are also limited to times that have coincident ozonesonde profiles that are used to assess the uncertainties of the retrieval. For days with multiple 30-min clear-sky periods, the averaged spectra of the nearest clear period to the coincident ozonesonde are selected for this study. These criteria resulted in a total of 201 cases, with more than 30 cases for each year (except in 2017, which only had six months of ozonesonde data).

B. MERRA-2

Using the Goddard Earth Observing System (GEOS) version 5 general circulation model and the Gridpoint Statistical Interpolation (GSI) [60], the MERRA-2 reanalysis products are created by NASA's Global Monitoring and Assimilation Office (GMAO) [46], [61] (available at GES DISC; <http://disc.sci.gsfc.nasa.gov>) at a spatial resolution of 0.5×0.625 . Various satellite data products, including the Solar Backscatter Ultraviolet Radiometer (SBUV), ozone monitoring instrument (OMI), and the Aura microwave limb sounder (MLS) are integrated into the MERRA-2 ozone estimation [46], [47].

When using any reanalysis dataset such as MERRA-2, uncertainties in the model physics, data assimilation methods, and observations can cause uncertainties in the products [45]. The global average of MERRA-2 evaluations has a standard deviation of 5% in the ozone concentrations in the middle to upper stratosphere (10–30 km) [47]. However, the uncertainties increase at lower altitudes with standard deviations of 20%–24% in the global estimation in the lower stratosphere and troposphere [47]. Previous MERRA-2 validation using ozonesondes in the middle to upper stratospheric layers over the high latitudes during highly altered circulation of sudden

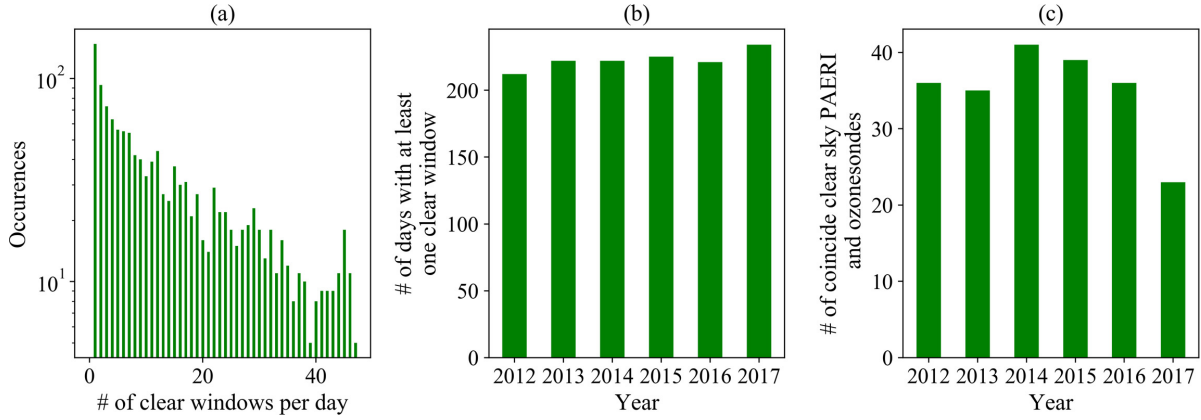


Fig. 2. Statistics of clear-sky data. (a) Occurrences of clear-sky windows (30 min) per day in all five and half years. (b) Number of days with at least one clear-sky window for each year. (c) Number of coincident days of at least one clear-sky window and an ozonesonde launch.

stratospheric warmings showed good agreement with observations with the standard deviation of 4%–7% [48]. However, the tropospheric and lower stratospheric layers show a mean difference of 5%–18% with a standard deviation of 15%–26% [48].

C. Ozonesondes

Ozonesondes provide an *in situ* measurement of the vertical profile of ozone using electrochemical concentration cells (ECCs) [62]. The response time and ascent rate of the ozonesonde balloons yield a vertical resolution of around 100 m. The measured ozone is expected to have an uncertainty of 3%–5% [62]–[65]. The fine vertical resolution and quality of measurements make ozonesondes an independent source for ozone validation [34], [48], [66], [67] as well as a direct source for monitoring ozone [22], [68]–[71].

In this study, we use ozonesonde profiles from Summit Station, Greenland (72°N, 39°W) to create a climatology of MERRA-2 uncertainty and as a source of validation. The ozonesonde program at Summit Station was started by the Global Monitoring Laboratory (GML) of the National Oceanic and Atmospheric Administration (NOAA) in February 2005 and continued until the summer of 2017. The ozonesonde data are available from NOAA's Earth System Research Laboratory. More details of the Summit Station ozonesondes and analysis can be found in previous studies [22], [72].

III. METHODS

Retrieving trace gas information from observed infrared spectra is an ill-posed problem. Different approaches can be used to solve this problem. The OEM [73], [74] is based on Bayes theorem to retrieve a rigorous and physically based quantity from measured spectra. The basic equation of $\mathbf{y} = \mathbf{K}\mathbf{x} + \mathbf{e}$ relates the state vector, $\mathbf{x}$ (unknowns), to the measurement vector, $\mathbf{y}$. In this case, the downwelling spectral radiance measurement is $\mathbf{y}$, the state vector ($\mathbf{x}$) is the ozone concentration profile in mixing ratio, and $\mathbf{e}$ is an error term. $\mathbf{K}$ is the derivative of a forward model ($\mathbf{F}$) with respect to the state vector, $(d\mathbf{F}(\mathbf{x})/d\mathbf{x})$, and is called the Jacobian matrix or kernel. In passive remote sensing retrievals, the forward

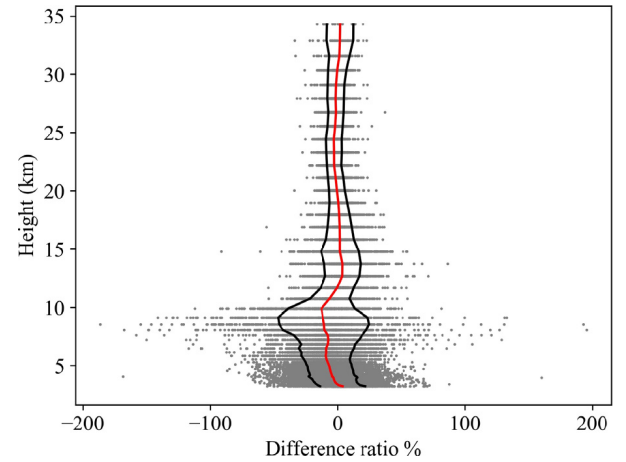


Fig. 3. Profile of relative differences of MERRA-2 ozone mixing ratio [computed as (MERRA-2 minus ozonesonde)/ozonesonde]. The mean of differences is shown in the red line. The mean $\pm$ standard deviation relative difference is shown as the black line.

model is a radiative transfer model (RTM). The forward model in ground-based remote sensing retrieval uses spectroscopic information, along with the concentration of atmospheric trace gases, temperature, and pressure on a vertical grid, to estimate the downwelling radiation at the surface. Considering all of the uncertainties in the measurements and assuming Gaussian distributions for the variables, OEM simplifies the Bayes theorem to achieve an optimal estimation of the state vector from

$$\hat{\mathbf{x}}_{n+1} = \mathbf{x}_a + (\mathbf{K}^T \mathbf{S}_e^{-1} \mathbf{K} + \mathbf{S}_a^{-1})^{-1} \times \{\mathbf{K}^T \mathbf{S}_e^{-1} [\mathbf{y} - \mathbf{F}(\hat{\mathbf{x}}_n)] + \mathbf{K}(\hat{\mathbf{x}}_n - \mathbf{x}_a)\} \quad (1)$$

$$\mathbf{G} = (\mathbf{K}^T \mathbf{S}_e^{-1} \mathbf{K} + \mathbf{S}_a^{-1})^{-1} \mathbf{K}^T \mathbf{S}_e^{-1} \quad (2)$$

where n is the number of iterations, $\hat{\mathbf{x}}$ is the expected state vector, $\mathbf{x}_a$ is the prior, $\mathbf{S}_e$ is the covariance of measurement uncertainties, and $\mathbf{S}_a$ is the covariance of prior information. OEM addresses the inverse problem by iteratively updating the posterior probability density function (pdf) of the state vector. A characteristic metric of the retrieval is the averaging kernel ($\mathbf{A}$), also known as the resolution matrix, which is defined as $\mathbf{A} = \mathbf{G}\mathbf{K}$. Important details of the information content associated with a retrieval can be derived from the averaging kernel, including the vertical resolution of the retrieval and the degrees

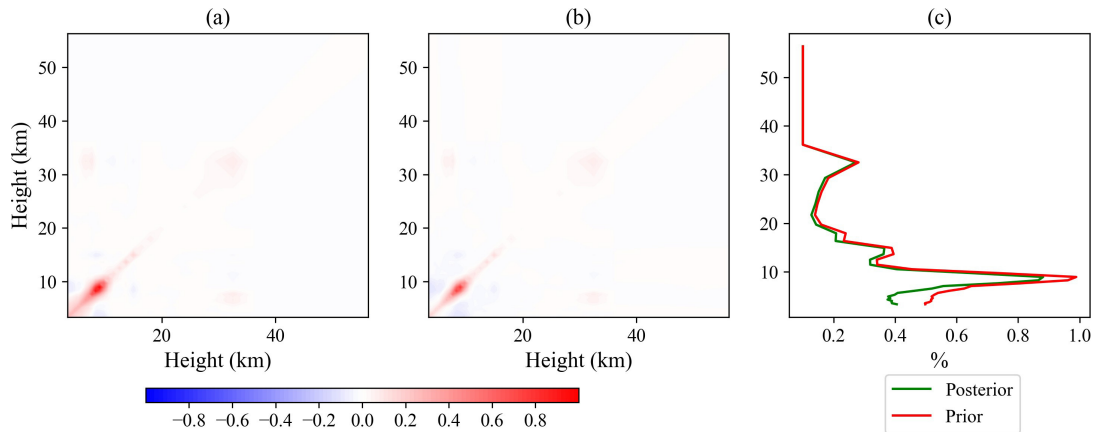


Fig. 4. Covariance of *a priori* and posterior. (a) Covariance of S_a , which is based on three standard deviations of MERRA-2 ozone covariance of uncertainties in comparison to ozonesondes. (b) Posterior covariance of retrieval. (c) Root square of diagonal element of covariance matrix for both prior and posterior by height.

of freedom of the signal (DOFS). DOFS for the retrieval is estimated from the trace of the matrix A and indicates the level of detail of the information that can be retrieved from the spectra.

In this study, the Spectra FITting Algorithm, version 4 (SFIT4), is used to retrieve the ozone profile. SFIT4 is one of the algorithms that apply the OEM to infrared spectra [75]. SFIT has been used to retrieve ozone and other trace gases from solar FTIR sensors [65], [76], [77]. The embedded RTM in SFIT4 has shown good agreement when compared to the line by line radiation transfer model (LBLRTM) [75]. The spectroscopic database from HITRAN 2008 is used in the retrieval [78]. The convolution of spectra calculated by SFIT4 with the P-AERI instrument line shape (ILS) creates synthetic spectra that have similar spectral characteristics to those measured by the P-AERI. Three-hourly temperature and ozone profiles from MERRA-2 and monthly profiles of water vapor and CO₂ from the Whole Atmosphere Community Climate Model WACCM4 [79] are used as the prior information for retrievals in this study.

The P, Q, and R branches of ν_3 vibrational-rotational modes of ozone molecules exhibit emission (and absorption) near $9.6 \mu\text{m}$ [80], creating a broad spectral feature at midspectral resolution spectra from about 995 to 1070 cm^{-1} . This spectral range is used for the ozone retrievals in this study and is shown in Fig. 1. To account for other gases in the retrieval spectral range, the column amount of CO₂ (with 0.01% S_a , correspond to 4 ppmv) and water vapor (with 0.25% S_a , correspond to 20 ppmv at surface level) are considered in the retrieval. MERRA-2 ozone profiles are used as the ozone prior information.

The vertical layering of retrieval matrices should be adequate to capture the profile shape. In this study, 34 layers with a consistent growth rate between 3.25 and 55 km are used. To estimate the uncertainty of the prior ozone information (S_a), the long record of ozonesondes program at Summit Station is compared to MERRA-2 ozone estimations. These comparisons, as shown in Fig. 3, include the relative difference [(MERRA-2 minus ozonesonde)/ozonesondes], mean, and standard deviation of differences. The uncertainties are

estimated using the entire dataset of ozonesondes from Summit from 2005 to 2017. The climatology of MERRA-2 ozone uncertainty is estimated as the covariance of differences in comparison to ozonesonde and is used as the covariance of the prior, as shown in Fig. 4(a). By considering the level-to-level interdependence of uncertainties in MERRA-2 (nondiagonal elements of S_a), this framework adds further constraints based on the observations, so that the retrieval achieves a more realistic result.

To allow the retrieval to consider the full range of uncertainties in the OEM process, three times the standard deviations are used as S_a (hereafter 3σ) at each level [see Fig. 4(a)]. The climatology of MERRA-2 uncertainties (S_a) can be summarized as: 1) from ground to 7 km, 3σ allows 50% change of ozone at each level, and the dependence among layers has the highest spatial extent in these layers; 2) from 7 to 10 km, 3σ allows $\sim 90\%$ ozone variations at these levels with a high degree of dependence to nearby layers in the retrieval process; and 3) above 10 km, the dependence of uncertainties among layers is diminished, and the diagonal values show that 3σ occurs from $\sim 20\%$ and quickly decreases with altitude down to about 5% above 15 km. The variability in the vertical uncertainties reflects the high quality of midstratospheric ozone in MERRA-2, and the large uncertainties of MERRA-2 below 10 km are consistent with previous studies [47], [48]. The observation-based S_a allows the retrieval to weight the P-AERI spectral information more heavily to determine the optimal ozone estimate below 10 km, while above 15 km, the retrieval mostly relies on the prior (MERRA-2). Between 10 and 15 km, the retrieval combines information from both sources.

Another essential component of OEM is the requirement of a measurement uncertainty estimate S_e . S_e has two main components [53], [81]: S_y (the covariance of measured radiance) and S_b (the covariance of the forward model and associated parameters). The random noise of $0.4 \text{ mW}/(\text{m}^2 \text{ sr cm}^{-1})$ is the estimated uncertainty of the P-AERI radiance for a 30-s sky view in the 995 – 1070-cm^{-1} region. Because the P-AERI spectra are averaged for 30 min during clear skies, the spectral noise reduces by the square root of the number of averaged spectra. The covariance of the parameters in the forward model

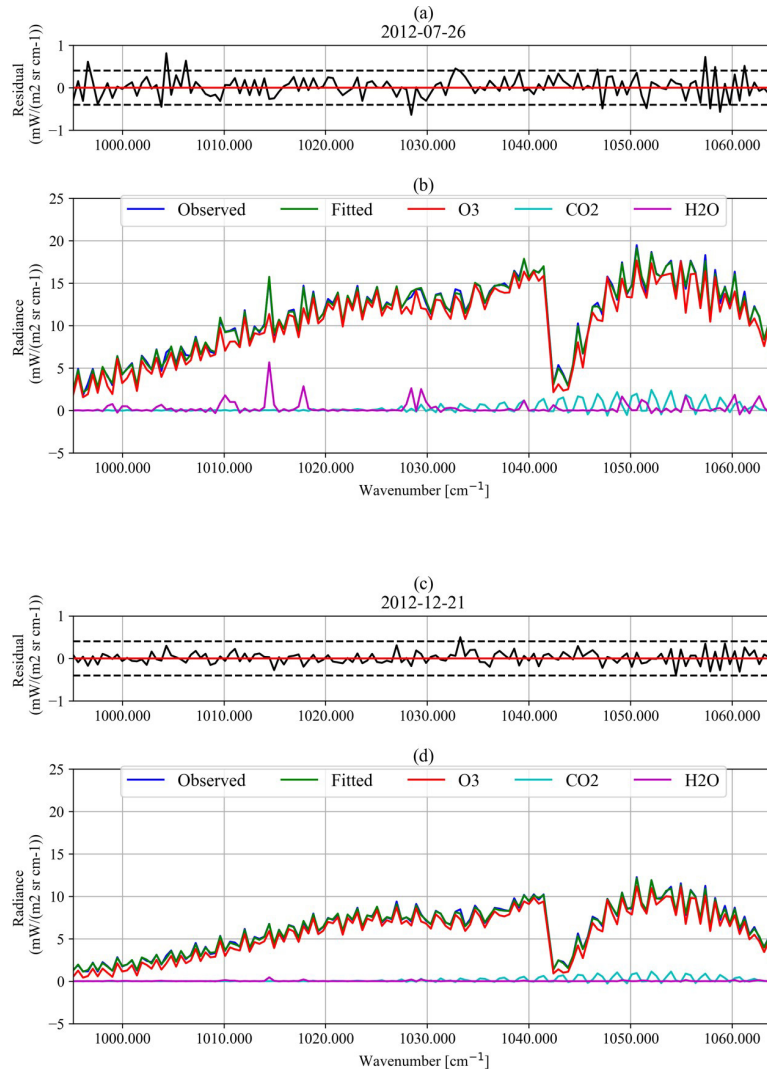


Fig. 5. Observed radiance (blue), fitted synthetic radiance (green), the radiance contribution of ozone (red), CO_2 (cyan), and H_2O (pink), and the residual in the spectral range of radiances in $995\text{--}1065\text{ cm}^{-1}$ after the retrieval. The observed and simulated spectral radiances for a humid and a day are shown in (b) and (d). The residual of the simulated spectral and the observed spectra for these days are shown in (a) and (c). The upper panels (b) show the humid summer day July 26, 2012, with a significant contribution of water signature in the spectra and the lower panel (d) the dry winter spectra from December 21, 2012. Black dashed lines in the residual panel (a) and (c) indicate plus and minus the spectral noise level $[\pm 0.4\text{ mW}/(\text{m}^2\text{ sr cm}^{-1})]$.

and the spectroscopic dataset has a significant contribution to the covariance of errors (S_e) [13], [81]. However, quantifying S_b , which translates to quantify the uncertainties of all of the absorption line parameters, would require extensive work and has not been done. Thus, a simplified estimation has been applied [13]. In this study to prevent unrealistic fluctuations of retrieved data by introducing exaggerated low error covariance [13], the final estimation of error covariance (S_e) for ozone retrieval is estimated to be $0.16\text{ mW}^2/(\text{m}^4\text{ sr}^2\text{ cm}^{-2})$ [whereas if we assumed $S_b = 0$, the estimate of the error covariance used in S_e would be $0.0025\text{ mW}^2/(\text{m}^4\text{ sr}^2\text{ cm}^{-2})$].

IV. RETRIEVAL INFORMATION CONTENT

To assess the retrieval performance, retrievals in clear-sky situations were first evaluated using information provided by the retrieval algorithm itself. To understand the information content of the retrievals, different aspects of the retrieval will be discussed including: the spectral fit and contribution of each trace gas, the Jacobian matrix, the average kernel profiles,

DOFS, and the posterior covariance matrix. Fig. 5(b) and (d) shows the observed spectral radiance, simulated radiances (fitted), the residual of fitted radiances, and the contribution of ozone, CO_2 , and water vapor to the fit spectra in $995\text{--}1070\text{ cm}^{-1}$ spectral region for cases on July 26, 2012, and December 12, 2012. Comparison of the spectral signatures on summer and winter days shows the importance (and the interference) of water vapor on the observed radiance during a summer day at Summit Station. The largest interfering signatures of water vapor are between 1010 and 1020 cm^{-1} and 1028 and 1030 cm^{-1} during humid spring/summer days [see Fig. 5(b)]. From 1030 to 1060 cm^{-1} , both CO_2 and water vapor contribute to the fitted spectra. The residual of the fitted spectra and the observed radiance has mostly random structure and is below the radiance random noise limit of $0.4\text{ mW}/(\text{m}^2\text{ sr cm}^{-1})$, as shown in Fig. 5(a) and (c).

The averaged Jacobian matrix of all cases indicates the vertical sensitivity of the retrieval at each wavenumber and is shown in Fig. 6. The Jacobian matrix has large values

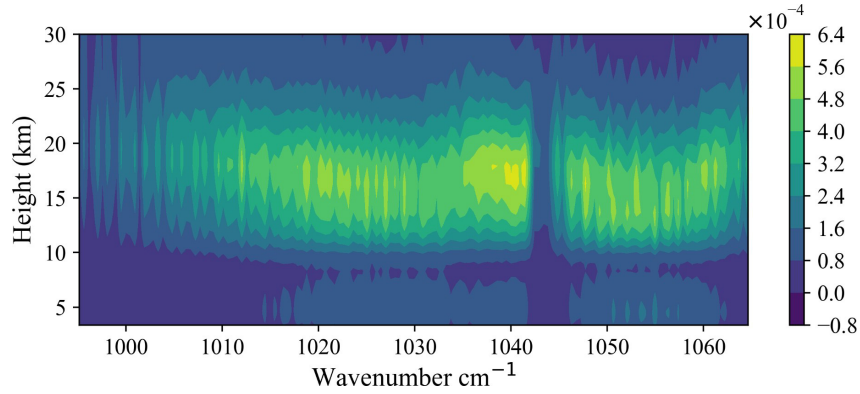


Fig. 6. Average Jacobian matrix in unit of radiance divided by O_3 mixing ratio for all retrieved cases.

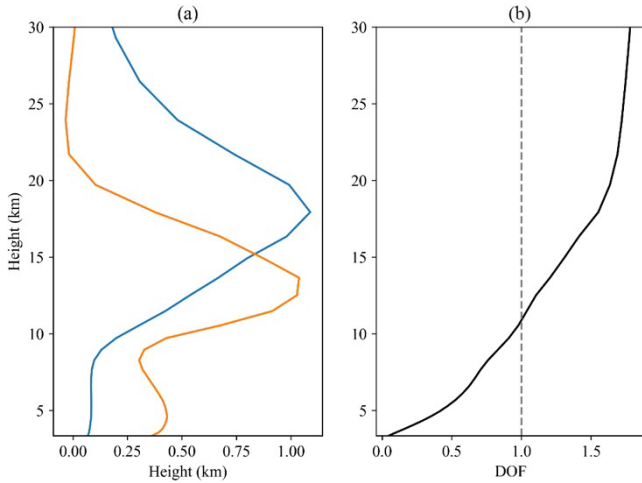


Fig. 7. Mean of averaging kernel profiles and DOFS of all retrieved cases. (a) Average kernel profiles, orange line shows the sensitivity to the lower atmosphere and the blue line corresponds to the middle stratospheric layers. (b) Cumulative profile of DOFS.

around 17 km and has maximum values between 1010 and 1040 cm^{-1} and 1045 and 1060 cm^{-1} . The vertical sensitivity of the Jacobian matrix is impacted primarily by the vertical structure of the ozone profile up to 17 km as the ozone profile is fairly uniform from the ground to ~ 7 km (the climatological tropopause at Summit Station) but then rapidly increases. Although the maximum peak of ozone concentration is higher in the middle stratosphere, the sensor sensitivity decreases above 17 km, which reduces the ability of the P-AERI to detect ozone above this height. Because MERRA-2 provides 3-h ozone profiles that agree well with ozonesondes in the middle stratosphere, the focus of this study is to retrieve accurate ozone concentrations below 10 km, where the Jacobian matrix shows sensitivity between 1015 and 1042 cm^{-1} and 1045 and 1062 cm^{-1} .

The posterior covariance matrix defined as $(K^T S_e^{-1} K + S_a^{-1})^{-1}$ is reduced compared to the covariance of the prior, as shown in Fig. 4(b). The root square of diagonal elements of the posterior covariance, representing the uncertainty in the expected state vector, is decreased compared to S_a . This reduction is more pronounced from surface to 7 km, as shown in Fig. 4(c). The level-to-level covariance from the surface (3.2 km) to 7 km is dramatically reduced, and some reduction

is evident in the 7–10-km layer. The mean averaging kernel profiles and DOFS are shown in Fig. 7. The averaging kernel indicates the smoothing of the true state vector based on the sensitivity of the sensor. The summation of the rows of the averaging kernel (until the cumulative DOF reaches to a full degree) shows the averaging kernel profiles [see Fig. 7(a)], which translate to the possible retrieved partial columns based on the sensitivity of the sensor [73]. As shown in Fig. 7(a), the averaging kernels show two profiles, one sensitive to the middle stratosphere and troposphere. The cumulative DOFS reaches 1 by around 10 km; thus, one partial column of ozone can be retrieved below 10 km. The time series of DOFS for two partial column ozone (PCO) values from the surface to 10 km and from 10 to 30 km is shown in Fig. 8. The seasonal variability in the DOFS for the two levels is evident in Fig. 8. Below 30 km, the sensitive altitude range of the sensor, a higher molecular density of ozone creates a stronger radiance signature in the measured P-AERI spectra and, consequently, increases the information content. The seasonal variability of the DOFS for PCO values exhibits the same cycle as ozone; the middle stratospheric ozone reaches a maximum during spring caused by the residual circulation, while the tropospheric and lower stratospheric ozone are impacted by photolysis and, therefore, peak during summer. More discussion on the seasonal cycle of ozone and the associated drivers at northern high latitudes (including Summit Station) is found in a previous study [22].

V. COMPARISON WITH OZONESONDES

In this section, the retrievals are compared with ozonesonde data that were obtained within 24 h of clear-sky P-AERI radiance measurements. Summit Station is a high altitude (3250 m), remote site that is far from sources of tropospheric ozone precursors at high northern latitudes. Due to the consistent presence of sunlight during polar days and the lack of sunlight during polar nights, diurnal variations are small in tropospheric ozone concentrations. Thus, to have more robust statistics, the ozonesondes within a day of P-AERI retrieval provide an adequate set of validation measurements. However, more than 63% of the ozonesondes and clear-sky P-AERI measurements are less than 5 h apart, while less than 10% of the cases have a time difference greater than 15 h. Moreover, there is no significant correlation between the time difference

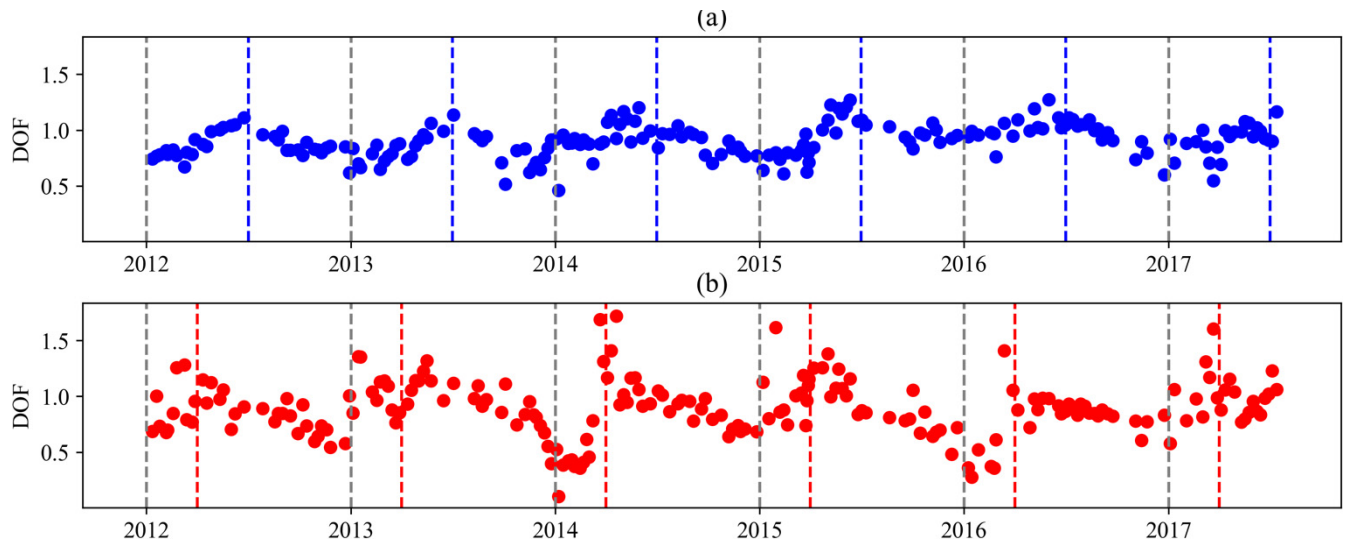


Fig. 8. Time series of DOFS and the impact of seasonal cycle on DOFS. Ground to (a) 10 km and (b) 10–30 km. Blue and red dashed lines are plotted on July 1st and April 1st of each year, respectively.

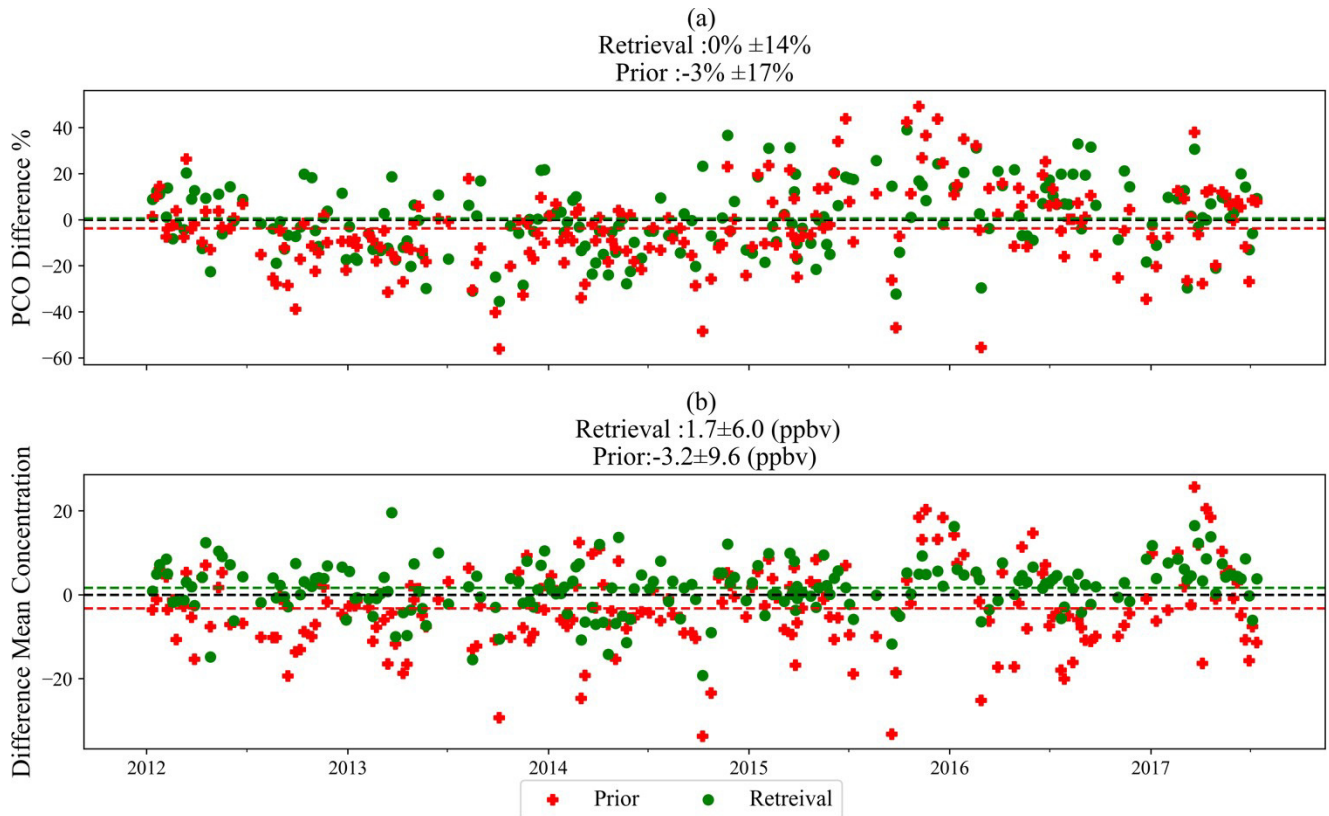


Fig. 9. Comparison of the retrieval and *a priori* to the independent ozonesonde observation. (a) PCO comparison from ground to 10 km, and the relative differences are $(x\text{-PCO} - \text{ozonesondes-PCO})/\text{ozonesondes-PCO} \times 100$, x be retrieved or the prior. (b) Comparison of mean ozone concentration up to 7 km (where ozone profile is fairly uniform), and the differences are $x\text{-ozonesonde mean concentrations}$. The average relative difference and difference concentrations are shown by dashed lines. The green color is assigned to the comparisons corresponding to retrievals and red to the prior. The black dashed line shows the zero line, and the mean and standard deviations are reported on top of each panel.

between the P-AERI and ozonesonde observations and the differences in concentrations between the P-AERI retrievals and ozonesondes.

Fig. 9 shows the PCO difference between the retrievals (green) and prior from MERRA-2 (red) relative to

the ozonesondes. Fig. 10 shows 16 examples of retrieved ozone to illustrate how the retrieved ozone compares to the MERRA-2 prior and ozonesondes. Fig. 9(a) shows the relative difference of the PCO from surface to 10 km from ozonesondes subtracted from retrieved PCO and divided by

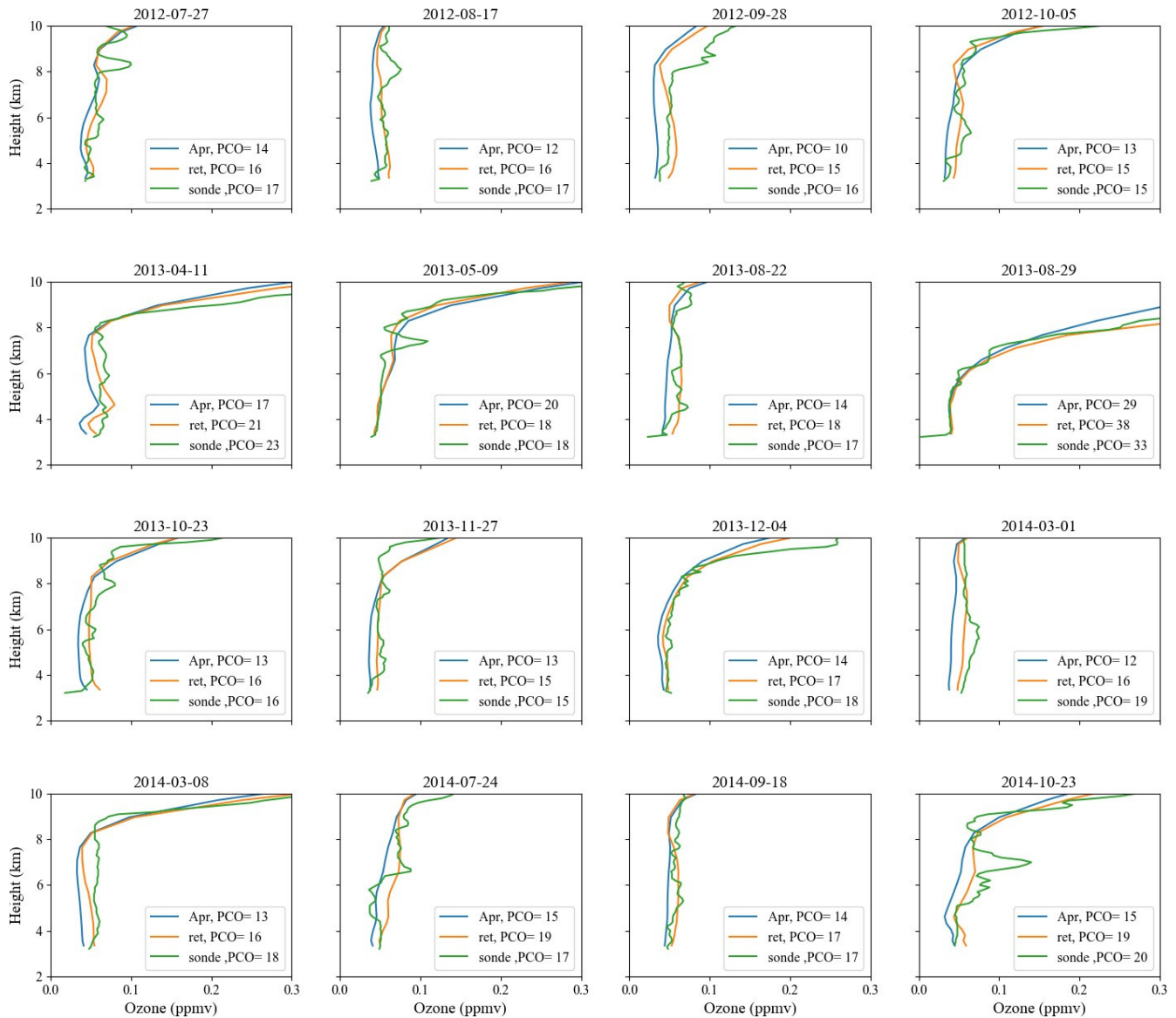


Fig. 10. Tropospheric/lower stratospheric ozone profile; 16 cases of retrieval elaborate the tropospheric/lower stratospheric ozone profile of the prior, final retrieval, and ozonesondes in blue, orange, and green lines, respectively. The associated PCO of ground to 10 km is reported in the lower right of each plot.

ozonesonde's PCO multiplied by 100; the same quantity is also shown using the prior information. The retrieved PCO improved the bias in the prior and reduced the standard deviation from 17% to 14%. Because the estimation of molecular density depends on temperature, we also looked at mean ozone concentrations for tropospheric layers to focus solely on ozone estimates. To consider the fairly uniform ozone profile concentration from surface to 7 km (climatological tropopause at Summit station; evident in Fig. 10), the vertical mean concentration of ozone (surface to 7 km) is compared to ozonesondes. The mean ozone concentration bias from the surface to 7 km is 1.7 ± 6 ppbv, which improves the original bias (-3.2 ± 9.6 ppbv) between the prior and ozonesondes, as shown in Fig. 9(b). It is important to mention that, because of the moderate spectral resolution of the P-AERI, this emission FTIR is unable to vertically resolve the detailed structures of ozone profile. Thus,

Fig. 10 illustrates that final retrieval profiles have reasonable structure and improvement is not accomplished through random fluctuations. To further illustrate the improvement from the retrieval, the pdfs of differences between the retrieval and MERRA-2 relative to the ozonesondes are shown in Fig. 11. The significant improvement of the average tropospheric ozone that improves the bias below 10 km is evident in Fig. 11.

The retrieval is most successful in improving the tropospheric ozone from surface to 7 km compared to surface to 10 km, which is due to the higher sensitivity of the FTIR in this region as seen in both the averaging kernel and the Jacobian matrix. The presented cases in Fig. 10 and the improvement in statistics shown in Fig. 11 emphasize on and clarify the improvements of posterior covariance matrix, as shown in Fig. 4(b), which show significant improvement in the diagonal and off-diagonal values of uncertainties. It should be noted

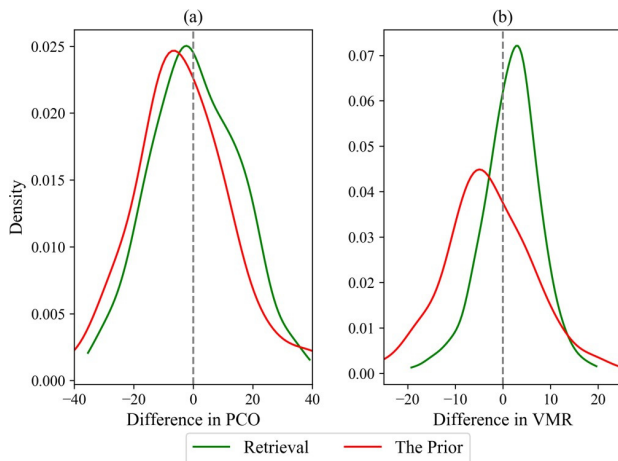


Fig. 11. PDFs of differences between retrieval and ozonesondes (green) versus MERRA-2 and ozonesondes (red) for (a) PCO from the ground to 10 km and (b) average mixing ratio from the ground to 7 km.

that even though the uncertainties are reduced, the remaining uncertainties in the posterior covariance matrix are important modulator of variability of the final retrieved values.

VI. CONCLUSION

The improved estimation of tropospheric and lower stratospheric ozone concentrations is critical to gain further knowledge of tropospheric–stratospheric ozone exchange, the impact of anthropogenic pollutants on ozone, ozone transport mechanisms, and the influence of climate change and ozone. Reanalysis models such as MERRA-2 integrate a variety of satellite measurements and meteorological data to provide ozone estimates with daily or subdaily temporal resolution around the globe. However, despite good agreement of MERRA-2 ozone estimates with observation in the middle stratosphere, larger uncertainties exist in the lower stratosphere and troposphere.

The long record of emission FTIR measurements obtained by the P-AERI since 2010 and the ozonesonde program between 2005 and 2017 at Summit Station, Greenland, provides a unique opportunity to investigate the capability of emission FTIR to improve the ozone estimates of the lower stratosphere and troposphere at high northern latitudes and to quantify the comparisons to MERRA-2.

MERRA-2 ozone profiles are used as the first guess of the retrieval process. The retrieval process utilizes the climatological MERRA-2 uncertainties to focus the ozone retrieval through the OEM on altitudes that are uncertain. Using ozonesondes, the climatological uncertainties of MERRA-2 at Summit Station are estimated and are used as an observational-based covariance matrix of the prior in the retrieval process. The retrieval weights the MERRA-2 product heavily in the middle stratosphere where uncertainties are low and the P-AERI spectra more heavily in lower atmosphere layers where MERRA-2 uncertainties are high. The off-diagonal elements of the covariance matrix, which show the level-to-level dependence of uncertainties in MERRA-2 ozone dataset, introduce extra constraints to the ill-posed problem and lead to more realistic results.

The retrieved ozone from the ground to 10 km reduces a 3% underestimation of the ozone partial column in the MERRA-2 compared to ozonesondes and decreases the standard deviation by 3%. Moreover, the mean concentration of tropospheric ozone, from the ground to 7 km, is significantly improved, while the standard deviation is decreased from 10 to 6 ppbv.

In conclusion, emission FTIR with moderate spectral resolution can be used to improve lower atmospheric ozone estimates in the Arctic. Considering the network of ozonesonde measurements globally, the observation-based estimate of *Sa* can be used in a variety of locations and latitudes. Having an extended FTIR network could expand the findings of this study to the global scale to improve the monitoring of ozone profiles below 10 km. Moreover, the integration of the emission and solar FTIR as assimilation data for reanalysis models such as MERRA-2 could improve ozone datasets globally.

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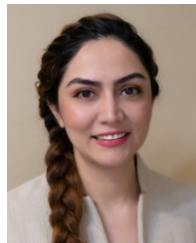
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