

1 **Measuring Atmospheric CO₂ Enhancements from the 2017 British Columbia**
2 **Wildfires using a Lidar**

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11 **Abstract.** During the summer 2017 ASCENDS/ABOVE airborne science campaign, the
12 NASA Goddard CO₂ Sounder lidar overflew smoke plumes from wildfires in the British
13 Columbia, Canada. In the flight path over Vancouver Island on 8 August 2017, the
14 column XCO₂ retrievals from the lidar measurements at flight altitudes around 9 km
15 showed an average enhancement of 4 ppm from the wildfires. A comparison of these
16 enhancements with those from the Goddard Global Chemistry Transport model suggested
17 that the modeled CO₂ emissions from wildfires were underestimated by more than a
18 factor of 2. A spiral-down validation performed at Moses Lake airport, Washington
19 showed a bias of 0.1 ppm relative to *in situ* measurements and a standard deviation of 1
20 ppm in lidar XCO₂ retrievals. The results show that future airborne campaigns and
21 spaceborne missions with this type of lidar can improve estimates of CO₂ emissions from
22 wildfires and estimates of carbon fluxes globally.

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24 **Plain Language Abstract.** Wildfires are a major source of greenhouse gases. However,
25 there are large uncertainties in the estimated CO₂ emissions from wildfires in global
26 emissions inventories. The estimates of column-averaged CO₂ (XCO₂) from satellite
27 measurements using passive remote sensing techniques are significantly degraded or
28 screened out by the scattering from smoke in the scene. NASA Goddard Space Flight
29 Center has developed an integrated-path, differential absorption lidar approach to
30 measure XCO₂ from space. Measurements of time-resolved laser backscatter profiles
31 from the atmosphere allow this technique to accurately estimate XCO₂ and range to
32 terrain and water surfaces even in the presence of wildfire smoke. We demonstrate this
33 capability over Vancouver Island through the dense smoke plumes from wildfires in the
34 Canadian Rockies during the summer 2017 ASCENDS/ABOVE airborne science
35 campaign. To our knowledge this is the first use of lidar to remotely sense CO₂
36 enhancements from large wildfires. Future airborne campaigns and spaceborne missions
37 with this capability can improve estimates of CO₂ emissions from wildfires and help
38 estimates of carbon fluxes globally.

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40 **Key Points:**

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- 43 • NASA Goddard CO₂ Sounder Lidar can accurately measure CO₂ enhancements
44 from wildfires through dense smoke plumes
 - 45 • This is the first use of lidar to remotely sense CO₂ enhancements from large
wildfires

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- These types of lidar measurements can be used to validate estimates of CO₂ emissions from wildfires and improve estimates of carbon fluxes

49 **1. Introduction**

50

51 Wildfires are a major source of greenhouse gases. Fires were responsible for as much as a
52 fifth of the carbon released in 2019 from burning fossil fuels, down from about a quarter
53 at the beginning of the century (Ciais et al., 2013; Tian et al., 2016; Le Quere et al.,
54 2018). While this long-term decrease in fire emissions was driven by a decline in savanna
55 and grassland fires (Andela et al., 2017), a recent increase in forest fires has resulted in
56 concerns about the future role of fire in the global carbon cycle. Total carbon emissions
57 from forest fires in 2019 were 26% higher than in 2018, to 7.8 billion metric tons, the
58 highest since 2002, according to the Global Fire Emissions Database (GFED; van der
59 Werf et al., 2017). The unprecedented bushfires in Australia in 2019 emitted a combined
60 306 million metric tons of carbon dioxide (CO₂) in the August-December 2019 period,
61 which is more than half of Australia's total carbon footprint in the year. Brazilian
62 Amazon fires emitted 392 million metric tons of CO₂ in 2019 which was equivalent to
63 more than 80% of Brazil's 2018 greenhouse gas emissions (Lombrana et al., 2020).

64

65 During 2017 Canada had a record-breaking wildfire season in the province of British
66 Columbia (BC). A total of 1.2 million hectares had burned by the end of the 2017 fire
67 season, the largest ever in the province (Duran, 2017) and massive smoke plumes were
68 lofted into the stratosphere in the mid-August (Torres et al., 2020).

69

70 Generally, there are large uncertainties in the estimated CO₂ emissions from wildfires
71 with fire emissions inventories (Meyer et al., 2012; Andreae, 2019). Ground-based and
72 airborne measurements of fire emissions are few and are difficult to obtain. Atmospheric
73 column-averaged dry air mole fraction of CO₂ (XCO₂) retrievals using surface reflected
74 sunlight, e.g., the Orbiting Carbon Observatory-2 (OCO-2; Crisp et al., 2004) and the
75 Greenhouse gases Observation SATellite (GOSAT; Kuze et al., 2016) are significantly
76 degraded by scattering effects of fire smoke in the scene (Mao and Kawa, 2004;
77 Houweling et al., 2005; Aben et al., 2007; Butz et al., 2009; Uchino et al., 2012; Guerlet
78 et al., 2013).

79

80 NASA Goddard Space Flight Center has developed an integrated-path, differential
81 absorption (IPDA) lidar approach to measure global XCO₂ from space as a candidate for
82 NASA's planned Active Sensing of CO₂ Emissions over Nights, Days, and Seasons
83 (ASCENDS) mission (Kawa et al., 2018). This pulsed laser approach uses a step-locked
84 laser source and a high-efficiency detector to measure atmospheric absorption at multiple
85 wavelengths across the CO₂ line centered at 1572.335 nm. It has a high spectral
86 resolution and sub-ppm sensitivity to changes in XCO₂ (Abshire et al., 2018).
87 Measurements of time-resolved atmospheric backscatter profiles allow this technique to
88 estimate XCO₂ and range to any significant reflective surfaces with precise knowledge of
89 the photon path-length even in the presence of atmospheric scattering (Ramanathan et al.,
90 2015; Mao et al., 2018).

91

92 During July and August 2017, NASA conducted a joint ASCENDS/ABoVE (Arctic
93 Boreal Vulnerability Experiment) airborne science campaign using the NASA DC-8
94 aircraft based in Fairbanks, Alaska (Mao et al., 2019). The CO₂ Sounder lidar measured

95 XCO₂ from aircraft altitudes to ground, along with height-resolved backscatter profiles.
96 Other instruments on the DC-8 aircraft included the NASA Langley Research Center
97 ACES CO₂ lidar (Obland et al., 2018) along with a suite of *in situ* instruments including
98 AVOCET for CO₂ (Halliday et al., 2019), Picarro for CO₂, CH₄, and H₂O, and an
99 engineering test version of DLH for H₂O, CO, CH₄, and N₂O (Diskin et al., 2002). The
100 DC-8 aircraft's housekeeping data provided temperature, pressure, geolocation, and
101 positioning such as altitude and pitch/roll angles at flight altitude. Its radar altimeter also
102 provides a reference for ground elevation under all conditions since the radar
103 measurement penetrates clouds and smoke.

104
105 During the return flight from Alaska to California on August 8 the aircraft overflew dense
106 smoke plumes from fires in the Canadian Rockies on the segment from Vancouver Island
107 to central Washington State. Here we present the XCO₂ and backscatter measurements
108 over this region along with the validation spiral maneuver at Moses Lake airport in
109 central Washington, performed shortly after the region of XCO₂ enhancement. We then
110 compare the measured XCO₂ enhancements with those from the Goddard Parameterized
111 Chemistry Transport Model (PCTM) using GFED. This case study demonstrates the
112 capability of the CO₂ Sounder lidar approach to measure enhanced XCO₂ through dense
113 smoke plumes, which allows improving the estimates of CO₂ emissions from wildfires.

114 115 **2. Lidar Measurements from 2017 ASCENDS/ABoVE Airborne Science Campaign**

116 117 **2.1 2017 ASCENDS/ABoVE Airborne Campaign to Alaska**

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119 The CO₂ Sounder lidar has flown on DC-8 five times since 2010 over a variety of sites in
120 the US, along with other ASCENDS airborne lidar candidates and *in situ* CO₂ sensors
121 (Abshire et al., 2013, 2016, and 2018). The ASCENDS/ABoVE airborne science
122 campaign to Alaska was the first to extend these lidar measurement to the arctic region.
123 The 2017 campaign also allowed determining the horizontal gradients in XCO₂ during
124 the long transit flights between California and Alaska. In all, eight flights were conducted
125 over the Central Valley of California, the Northwest Territory Canada, and south and
126 central Alaska between July 20 and August 8, 2017. Forty-seven vertical spiral
127 maneuvers were conducted over a variety of atmospheres and surface types like desert,
128 vegetation, permafrost, and both the Pacific and Arctic Oceans. The XCO₂ retrievals
129 from the lidar measurements were validated against those computed from *in situ*
130 measurements of CO₂ vertical profiles made during the spiral maneuvers.

131
132 The final flight of the campaign was conducted on August 8, 2017, based out of
133 Fairbanks, AK and transited south back to Palmdale, CA (Figure 1, top panel). The flight
134 had six spiral-down maneuvers when over land including ones at Northway airport in
135 Alaska, Whitehorse airport in Yukon, Canada, Moses Lake airport in Washington,
136 Wildhorse airport in Oregon, Winnemucca airport in Nevada, and Edwards Air Force
137 Base in California. The flight also conducted two in-line descent-ascent maneuvers above
138 the Pacific Ocean just off the southern tip of Alaska before flying to Vancouver Island.
139 Other than the spirals, almost all the flight was at 8-9 km altitude, except for the final
140 segment between Reno NV and Edwards CA, which was flown at 12 km to allow

141 sampling upper tropospheric air.

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143 The bottom panel of Figure 1 shows *in situ* CO₂ concentrations at aircraft altitudes
144 measured by AVOCET for the flight. AVOCET has a stated precision of ± 0.1 ppm (1-
145 sigma) and accuracy of ± 0.25 ppm (Halliday et al., 2019). It shows significant horizontal
146 and vertical gradients of CO₂ at the aircraft altitude, which is a typical seasonal pattern in
147 the area. The CO₂ concentrations were higher near the surface at Fairbanks, Northway,
148 and Whitehorse airports during the morning time of the flight due to the overnight
149 accumulation of respiration and local emissions. Meanwhile, the higher CO₂ over
150 Winnemucca, NV and Edwards Air Force Base, CA were presumably due to regional
151 emissions, as there is little surface uptake over the deserts. The *in situ* measurements
152 show high CO₂ in the free troposphere during the flight segment over Pacific Ocean and
153 lower CO₂ in the following segment onto land before the spiral at Moses Lake, WA. It
154 was notable that no outstanding CO₂ enhancements were seen at the aircraft altitude
155 between the spirals at Pacific 2 and Moses Lake compared to CO₂ values at the same
156 flight altitude before and after the segment.

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158 **2.2 XCO₂ Measurements from the CO₂ Sounder Lidar**

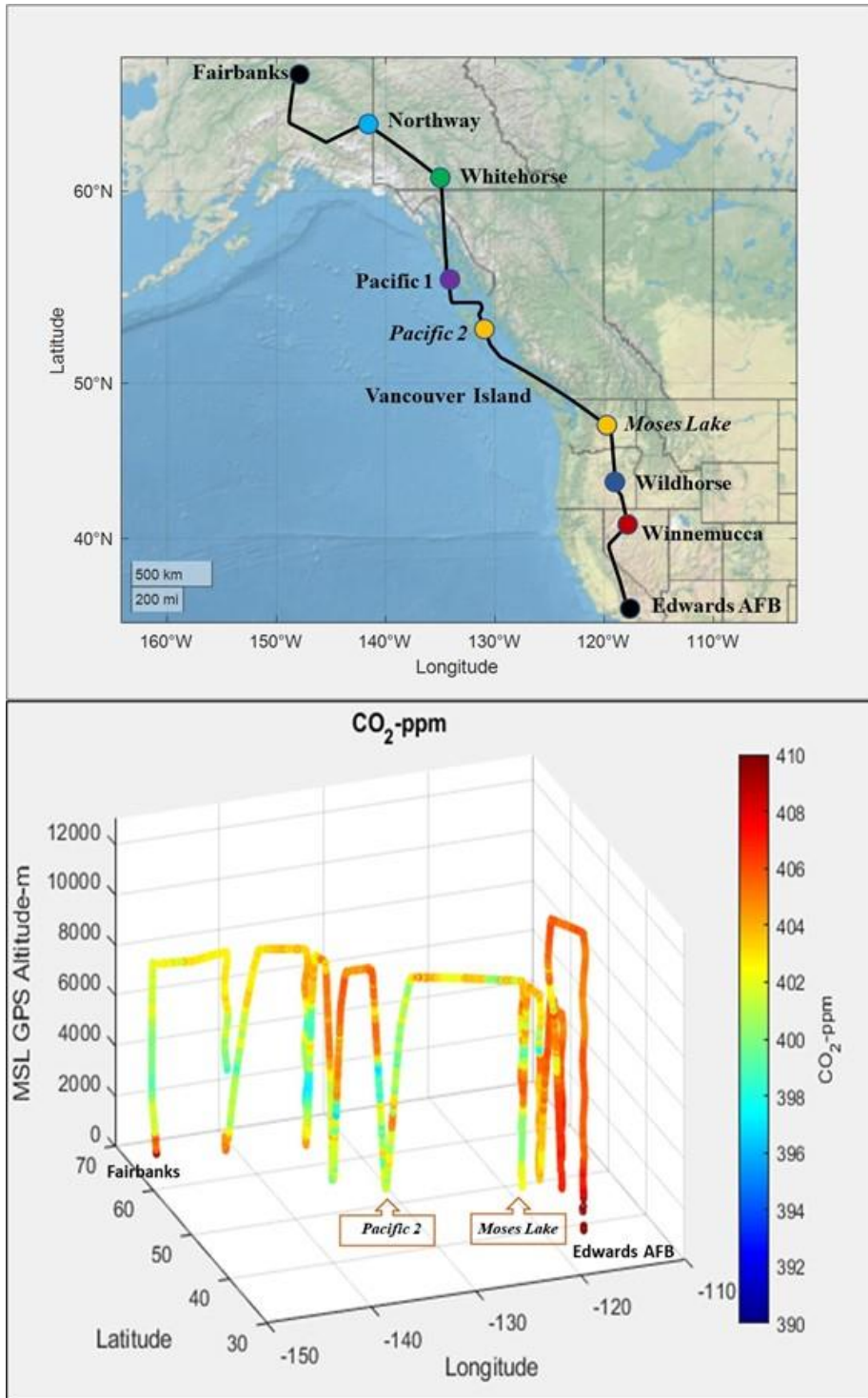
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160 The airborne CO₂ Sounder lidar uses a tunable laser to measure absorption across the
161 vibration–rotational line of CO₂ centered at 1572.335 nm (Abshire et al., 2018). The lidar
162 transmits 1- μ s wide laser pulses at a rate of 10 kHz and the laser is stepped in 30
163 wavelengths across the CO₂ line at a rate of 300 Hz. The wavelength separation of each
164 laser pulse was 250 MHz near line center and increased to 2 GHz at line wings to allow
165 for more online samples. The laser line width is narrower than 30 MHz. The spectral
166 resolution of the laser is over two hundred times higher than that of GOSAT, over three
167 hundred times higher than that of OCO-2, and over twenty times higher than that of the
168 ground-based Fourier Transform Spectrometers for the Total Carbon Column Observing
169 Network (Wunch et al., 2011). The high spectral resolution allows sampling the fully-
170 resolved CO₂ line shape, including line width and line center position (Ramanathan et al.,
171 2013), resulting in high sensitivity to CO₂ changes in the atmospheric column (Mao and
172 Kawa, 2004).

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174 The lidar retrieval algorithm uses a least-squares fit between the 30 wavelengths of the
175 lidar measurements and the calculated CO₂ absorption line shape to retrieve XCO₂
176 (Ramanathan et al., 2018; Sun et al., 2021). The approach allows use of a standard linear
177 least squares method to simultaneously solve for Doppler frequency shift, surface
178 reflectance at off-line wavelengths, and a linear non-uniformity (slope) in the receiver
179 spectral response as well.

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Figure 1. *Top*: map of the ground track for the return flight from Fairbanks, AK to Palmdale, CA on August 8, 2017. Fairbanks and the locations of eight spiral down flight segments are marked in circles, including two in-line descent-ascent maneuvers over Pacific Ocean labeled as Pacific 1 & Pacific 2. *Bottom*: 3-D (latitude, longitude, and flight altitude) sideview of *in situ* CO₂ mixing ratio measurements from onboard AVOCET for this flight. The data were sampled at 1-s intervals.

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190 In the retrieval forward calculations, the spectroscopy database HITRAN 2008 (Rothman
191 et al., 2009) and the Line-By-Line Radiative Transfer Model (Clough et al., 1992; Clough
192 and Iacono, 1995) V12.1 were used to calculate CO₂ optical depth and create look-up
193 tables (LUTs) for a prior with a vertically uniform CO₂ concentration of 400 ppm. We
194 then used these LUTs to retrieve the best-fit XCO₂ by comparing the lidar sampled line
195 shapes with the calculated absorption line shapes and then scaling the prior without any
196 inversion constraints. The retrievals used the atmosphere state (pressure, temperature, and
197 water vapor profiles) from the near real time forward processing data of the Goddard
198 Earth Observing System Model, Version 5 (GEOS-5; Rienecker et al., 2011). Data on the
199 full model grid (0.25° latitude×0.3125° longitude×72 vertical layers, every 3h) were
200 interpolated to flight ground track position and time for the atmospheric CO₂ and H₂O
201 absorption calculations.

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203 3. Data Analysis Results

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205 3.1 CO₂ Enhancements from BC Wildfires

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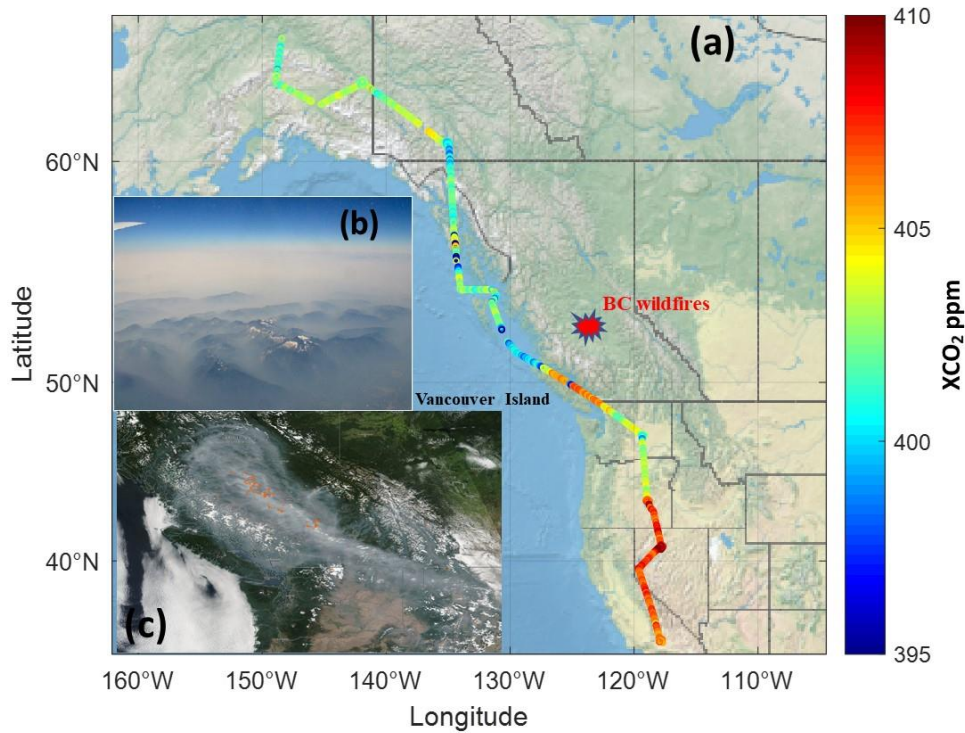
207 Figure 2(a) shows the cloud-free XCO₂ retrievals from the lidar for the entire flight on
208 August 8, 2017. Significant XCO₂ enhancements were clearly seen in the segment over
209 Vancouver Island and across the Strait of Juan de Fuca into Washington State. Such CO₂
210 enhancements were not evident in the *in situ* measurements at flight altitudes (Fig. 1).
211 Compared to single-point *in situ* measurements, this shows a benefit of the lidar's XCO₂
212 measurements to capture CO₂ variations in the full atmospheric column below the
213 aircraft.

214

215 The smoke plumes from wildfires in the Canadian Rockies were clearly seen from DC-8
216 aircraft over Vancouver Island and the fire and thermal anomalies map from
217 Aqua/MODIS on the same day (Figure 2(b) and (c)). The smoke plumes were transported
218 by wind from the Canadian Rockies into eastern Washington State and further down into
219 Montana. Meanwhile, some smoke plumes and a large amount of CO₂ emissions from the
220 fires were also transported to Vancouver Island.

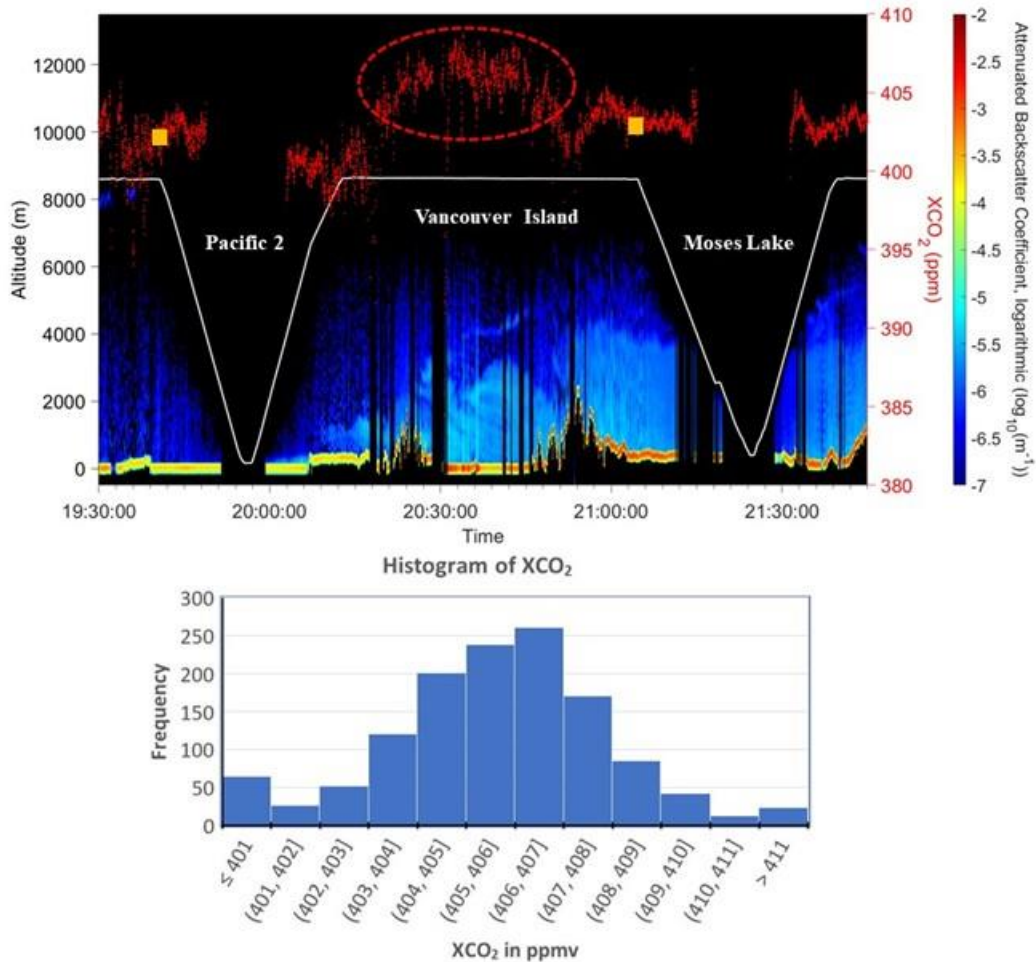
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222 Figure 3 shows the time series of the cloud-free XCO₂ retrievals together with the
223 attenuated backscatter profiles for the flight segment from Pacific Ocean to Washington
224 State. Dense smoke layers were seen in the lidar backscatter profiles near Vancouver
225 Island and peaked at the top of the boundary layer near 2 km. The lidar range was used to
226 distinguish ground returns from cloud returns after comparison to onboard radar
227 altimetry. The XCO₂ retrievals over Vancouver Island and western Washington State
228 have a median value of 406.5 ppm. The XCO₂ computed from the *in situ* vertical profiles
229 of CO₂ mixing ratio during the spiral maneuvers at Pacific 2 and Moses Lake were 401.9
230 and 402.6 ppm, respectively. Therefore, the averaged XCO₂ enhancement within the
231 segment from Vancouver Island to western Washington State was estimated to be 4 ppm.
232 The XCO₂ enhancement segment spanned about 30 minutes, which at DC-8 aircraft
233 speed of 200 m/s, corresponds to a ground-track length of 360 km.



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Figure 2. (a) the cloud-free XCO₂ retrievals from the CO₂ Sounder lidar for the flight on August 8, 2017. Significant XCO₂ enhancements were seen in the flight segment crossing Vancouver Island. The British Columbia wildfires are marked to the north of these enhancements. (b) image of the smoke plumes from the wildfires in Canadian Rockies as seen from DC-8 over Vancouver Island (Photo by Graham Allan). (c) true color image from Aqua/MODIS showing the smoke and fires on the same day.



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245 Figure 3. *Top*: the time series of cloud-free XCO₂ retrievals from the 1-s averaged lidar
 246 data (right axis) and the range-corrected attenuated backscatter profiles sampled at a
 247 vertical resolution of 15-m. Ground returns are strong and colored in yellow and red, and
 248 the returns from aerosols are light blue. The red dots are the 1-s XCO₂ retrievals
 249 smoothed with 9-point running mean. Aircraft GPS flight altitudes are marked in a white
 250 line. For reference, orange squares are the *in situ* XCO₂ from AVOCET during the 2nd in-
 251 line descent-ascent maneuver over Pacific Ocean and the spiral down maneuver at Moses
 252 Lake airport, Washington. The XCO₂ enhancements near Vancouver Island are circled.
 253 *Bottom*: a histogram of the 1-s averaged XCO₂ retrievals in the enhancement segment.

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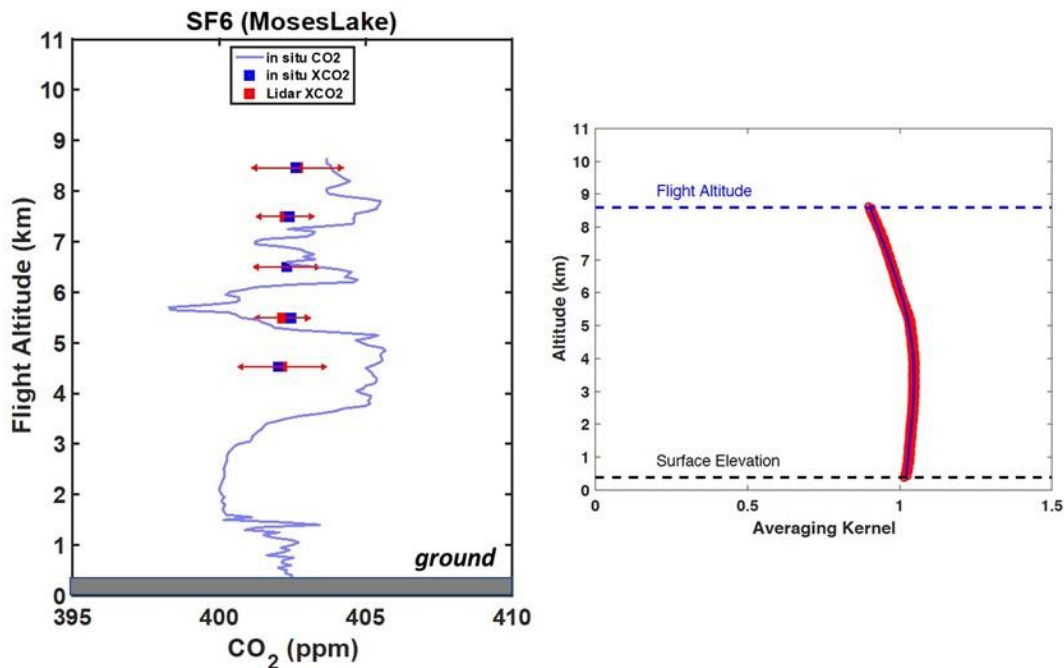
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256 3.2 Validation of the Lidar XCO₂ Measurements

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258 A vertical spiral-down maneuver was conducted shortly after the CO₂ enhancement
 259 segment shown in Fig. 3 from a flight altitude of 9 km to ground over Moses Lake in
 260 central Washington State. This allowed a comparison between the XCO₂ retrievals from
 261 the lidar and those constructed from the *in situ* vertical profile of CO₂. During the spiral
 262 hazy conditions were seen below 4.5 km in the lidar backscatter profiles (Fig. 3). The

263 AVOCET analyzer sampled every 1-s and the lidar XCO₂ retrievals were also based on
 264 1-s averaged laser signals returned from ground. For the best estimation of the
 265 atmosphere state during the spiral maneuver, these retrievals used vertical profiles
 266 simultaneously measured by onboard DC-8 *in situ* instruments at an interval of 1-s. The
 267 *in situ* XCO₂ was computed from the *in situ* vertical profile integrated using the lidar's
 268 retrieval averaging kernel as vertical weighting. Both lidar and *in situ* XCO₂ were then
 269 averaged in each 1-km atmosphere layer for comparison. Figure 4 shows an average
 270 difference of less than 0.1 ppm for flight altitudes above 5-km and an average standard
 271 deviation of approximately 1 ppm. Validation results from other profiles throughout the
 272 campaign were within +/- 0.5 ppm (1-sigma) of the *in situ* data. Therefore, the 4 ppm
 273 XCO₂ enhancement from the Canadian wildfires was highly significant in relative to the
 274 lidar measurement uncertainty.
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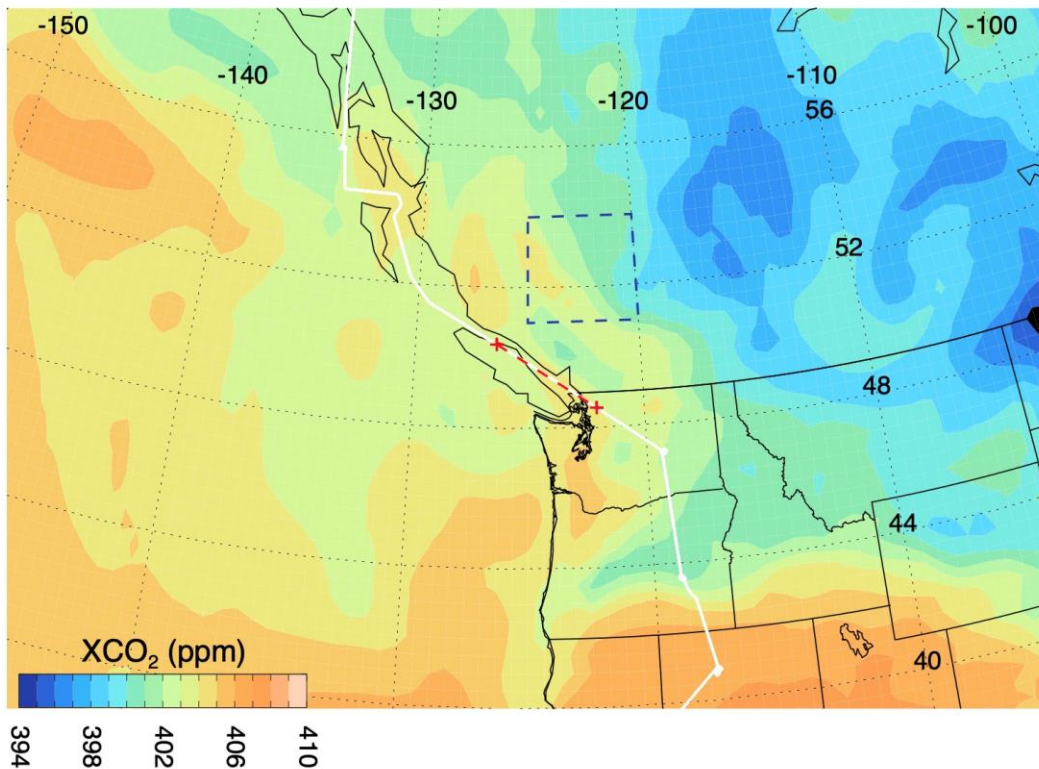


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 277 Figure 4. Comparison of cloud-free lidar XCO₂ retrievals with those from *in situ*
 278 measurements as a function of flight altitude during the spiral maneuver at Moses Lake,
 279 WA on August 8, 2017. The *in situ* XCO₂ values are marked in blue squares and the
 280 lidar's XCO₂ retrieval values are marked in red squares. The red error bars are ±1
 281 standard deviation of the lidar's XCO₂ retrievals. The XCO₂ vertical averaging kernel for
 282 this profile segment is shown at right.
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284 3.3. Improving Estimates of CO₂ Emissions from Wildfires

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 286 Figure 5 shows the integrated XCO₂ below 320 mb (~ 9 km) from CO₂ simulations by the
 287 Goddard PCTM (Kawa et al., 2004 and 2010) on the same day. Note that averaging
 288 kernels were not applied to the model XCO₂ for estimating relative changes due to fire
 289 emissions. The PCTM CO₂ simulation is driven by meteorological data from the Modern-

290 Era Retrospective analysis for Research and Applications (Bosilovich, 2013), which is a
 291 NASA reanalysis using GEOS-5. The vertical mixing in PCTM is parameterized for both
 292 turbulent diffusion in the boundary layer and convection. PCTM is run at 0.625°
 293 longitude \times 0.5° latitude with 56 hybrid vertical levels and outputs hourly. PCTM uses
 294 GFED4s (including small fires) for the CO₂ emissions from wildfires. GFED includes an
 295 ecosystem model that uses satellite observations of burned area and ecosystem
 296 productivity to estimate fuel loads and combustion (van der Werf et al., 2017).
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 300 Figure 5. Map of XCO₂ (ppm) from ground to 320 mb simulated by Goddard PCTM at
 301 21 GMT on August 8, 2017. White line is the ground track of the airborne campaign
 302 flight and the two red pluses and red dashed line mark the flight segment where the XCO₂
 303 enhancements were seen in the lidar retrievals. The box delineated by the dashed blue
 304 line indicates the area over which the BC fire emissions were calculated.
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306 The modeled XCO₂ at 21 GMT shows enhancements up to ~2 ppm over the Canadian
 307 Rockies in response to a total release of 837 Gg C day⁻¹ from the BC fires within the area
 308 (51-54°N, 120-125°W) on August 8 estimated by GFED. The modeled XCO₂
 309 enhancements near Vancouver Island (estimated from the local maximum on the contour
 310 map near the flight track in Figure 5 as well as from the model-interpolated XCO₂ along
 311 the track similar to Figure 3) were about 1 ppm. Compared to 4 ppm averaged
 312 enhancement of lidar XCO₂ for the equivalent atmospheric columns, the modeled
 313 enhancements were low. The underestimate of XCO₂ in the model could be due in part to
 314 model diffusion and transport shortcomings. Given, however, the spatial scale of the

315 observed XCO₂ perturbation (~360 km) and multi-day duration of the fires, along with
316 past performance of PCTM using analyzed winds to simulate CO₂ gradients in frontal
317 systems and other relatively fine-scale features (Parazoo et al., 2008) as well as the parent
318 GEOS-5 model use for aerosol plume simulations, we expect that the XCO₂ perturbation
319 would be close to that observed if the emissions were correct. The daily CO₂ release
320 estimate from another dataset of fire emissions, the Quick Fire Emissions Dataset
321 (QFED; Darmenov and Silva, 2015), in the same area on the same day was 1122 Gg C
322 day⁻¹. The QFED estimate was 34% higher than that from GFED but proportionally still
323 underestimated at least by a factor of 2. QFED is based on the detection of fire radiative
324 power calibrated against observations of aerosol optical depth. Our findings in this case
325 study highlight the potential of airborne and spaceborne lidar XCO₂ measurements for
326 evaluating atmospheric models and global emissions inventories.

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328 **4. Conclusion and Discussion**

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330 Analysis of lidar measurements from the summer 2017 ASCENDS/ABoVE airborne
331 science campaign show the capability to measure XCO₂ enhancements through dense
332 smoke plumes from wildfires in British Columbia, Canada. On the overpass of
333 Vancouver Island on August 8, the retrievals from the lidar measurements showed an
334 average 4 ppm enhancement in XCO₂ beneath the aircraft. A spiral maneuver made after
335 the smoke plume showed the XCO₂ measurements had small bias and high precision, and
336 a high spatial resolution (~ 200-m). The modeled enhancements from the Goddard PCTM
337 which uses the GFED fire emission database were about 1 ppm near Vancouver Island.
338 The result suggests that the CO₂ emissions from GFED for the BC wildfires were
339 underestimated by a factor of 2 or more for that day.

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341 The results show that future airborne campaigns and spaceborne missions with this
342 capability should improve modeling of CO₂ emissions from wildfires. This will benefit
343 atmospheric transport process studies, carbon data assimilation, and global and regional
344 carbon flux estimates. Along with the expected increase in the net contribution of forest
345 fires to global carbon emissions, improved capabilities to constrain wildfire emissions is
346 greatly needed.

347

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354 downloaded at the LaRC airborne science data site, [https://www-air.larc.nasa.gov/cgi-](https://www-air.larc.nasa.gov/cgi-bin/ArcView/ascends.2017)
355 [bin/ArcView/ascends.2017](https://www-air.larc.nasa.gov/cgi-bin/ArcView/ascends.2017). This paper is dedicated to our friend and colleague Dr.
356 Graham R. Allen who had a key role in the 2017 ASCENDS/ABoVE airborne
357 campaign but who passed away in May 2020. Dr. Allen's work was essential in the
358 development of the Goddard CO₂ Sounder lidar, airborne campaigns, and data
359 analysis over the past decade and he is missed. We thank two anonymous reviewers
360 whose edits and comments/suggestions helped improve and clarify this manuscript.

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