Mid-Atlantic U.S. observations of radiocarbon in CO2: fossil and biogenic source partitioning and model evaluation

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Abstract. Accurately quantifying regional anthropogenic CO_2 fluxes is fundamental to improving our understanding of the
15 carbon cycle and for creating effective carbon mitigation policies, and the radiocarbon to total carbon ratio in atmospheric CO_2
($\Delta^{14}CO_2$) is a robust tracer of fossil fuel CO_2 that can discriminate between biogenic and fossil fuel CO_2 sources. NASA's
ACT-America airborne mission between 2016 and 2019 aimed to improve the accuracy of regional greenhouse gas flux
estimates, through refining understanding and characterization of fluxes and flux uncertainties in models. $\Delta^{14}CO_2$ observations
from 26 flights are presented for examining seasonal CO_2 source partitioning in the Mid-Atlantic U.S. Observed variability in
20 boundary layer CO_2 at time scales ranging from intra-day to seasonal was largely driven by biogenic CO_2 (CO_{2bio}) variability
that ranged from -19.7 ppm in summer to 16.2 ppm in fall, while fossil fuel CO_2 (CO_{2ff}) variability remained at 3.3 ± 2.0 ppm.
Carbonyl sulfide uptake was well-correlated with CO_{2bio} uptake, and examining this relationship, and that between CO_2 and
CO_{2bio} variability reinforces the seasonal extent of gross primary productivity response throughout ACT-America. We use
airborne $\Delta^{14}CO_2$ flask sampling alongside in situ carbon monoxide measurements to calculate high-frequency CO_{2ff} and
25 evaluate the magnitude and diurnal variability of modeled CO_{2ff}, deducing likely transport errors in an example flight. Although
ACT-America CO_{2ff} signals were attenuated due to broad source regions sampled, results illustrate the value of $\Delta^{14}CO_2$
sampling and observation-based methodologies for regional CO_2 flux attribution and evaluation and improvement of modeled
CO_2.

1 Introduction

30 Anthropogenic carbon dioxide (CO_2) emissions have driven more than a 50% increase in atmospheric CO_2 abundances globally
since pre-industrial levels, despite almost half of these emissions being removed from the atmosphere by terrestrial and oceanic
reservoirs (i.e., "sinks"; Ballantyne et al., 2012; Friedlingstein et al., 2022). Because total CO_2 fluxes encompass both biogenic

and anthropogenic processes, and regional spatial scales are ones at which carbon mitigation strategies are generally developed
and implemented, it is important that these CO_2 component processes be accurately quantified. Fossil fuel emissions represent
35   the bulk of the net atmospheric CO_2 flux annually on regional scales for North America (King et al., 2007; USGCRP, 2018).
However, while national-scale fossil fuel CO_2 fluxes for many countries are likely known to within 10% (Gregg et al., 2009;
Andres et al., 2012), fluxes and their variability are less certain on smaller spatial and temporal scales (USGCRP, 2018; Gurney
et al., 2021). Even so, fossil fuel fluxes are currently reported with uncertainties up to five times lower than biospheric CO_2
fluxes on regional scales (Hayes et al., 2012; King et al., 2015; USGCRP, 2018).

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Atmospheric CO_2 and its long-term global growth rate can be directly determined from in situ observations because observed
CO_2 over large land areas is a function of both fossil fuel emissions and net ecosystem exchange (NEE) with the terrestrial
biosphere. Distinguishing fossil fuel from biogenic CO_2 fluxes cannot be accomplished from CO_2 observations alone, however
(Shiga et al, 2014; Basu et al, 2016). The radiocarbon to total carbon ratio in atmospheric CO_2 (^{14}C:C, expressed as $\Delta^{14}CO_2$)
45   has been demonstrated to be a robust and largely unbiased tracer for accurately constraining recently-emitted fossil fuel CO_2
into the atmosphere because fossil fuels do not contain ^{14}C (Levin and Karstens, 2007; Graven et al., 2009; Vogel et al., 2010;
Miller et al., 2012; Turnbull et al., 2006, 2011a, 2011b, 2015). By isolating this fossil fuel component and its variability
relative to background levels, any remaining variability in CO_2 can be attributed to biogenic processes.

50   Inverse models using only CO_2 measurements can theoretically be used to disaggregate anthropogenic and biogenic CO_2 fluxes
when these emissions can be adequately separated in time and space, but given the low density of most regional measurement
networks, fossil and biogenic fluxes are co-located, so this disaggregation cannot be achieved in practice (Shiga et al., 2014;
Basu et al., 2016). In many situations from regional to global scales, prior fossil CO_2 emissions are traditionally pre-specified
and not optimized in CO_2 measurement-based inversions (Gurney et al., 2003; Ciais et al., 2010; Schuh et al., 2010; Lauvaux
55   et al., 2011) on the basis that fossil fluxes are much better known than terrestrial biospheric fluxes. Much work has been done
to characterize model uncertainty and improve estimates of CO_2. At the continental scale and for annual time frames, Feng et
al., (2019a) used a forward calibrated model ensemble (Feng et al., 2019b) to show that the uncertainty in simulated North
American atmospheric CO_2 mole fractions comes primarily from two approximately equivalent sources: fossil fuel emissions
and biogenic CO_2 fluxes. At these time scales, these two sources greatly outweighed the uncertainty contributions from
60   atmospheric transport and from continental boundary conditions. Brophy et al. (2019) emphasized the need for varying prior
fossil fuel emissions estimates in time, alongside needed improvements in the representation of atmospheric transport to
accurately represent CO_2 fluxes at ~sub-annual and regional scales. Modelers have also relaxed the assumption of perfectly
known fossil CO_2 emissions – and have lowered fossil fuel emissions uncertainties – by assimilating $\Delta^{14}CO_2$ measurements
(or CO_{2ff}) alongside CO_2 and carefully controlling systematic model errors and errors in prior flux estimates (Basu et al., 2016,
65   2020; Fischer et al., 2017).

Using aircraft $\Delta^{14}CO_2$ observations, a strong effort has been made to evaluate modeled CO_{2ff} for verifying regional emissions inventories (e.g. Graven et al., 2009; Graven et al., 2018). At the urban scale, studies have also successfully used $\Delta^{14}CO_2$ measurements alongside in situ carbon monoxide (CO) observations from surface- and aircraft-based platforms, to calculate

70 high-resolution proxy estimates of recently-added fossil fuel CO_2 to the atmosphere, given well-characterized fossil fuel CO_2:CO ratios (Vogel et al., 2010; Turnbull et al., 2011a, 2019; Lauvaux et al., 2020; Wu et al., 2022). NASA's multi-year Atmospheric Carbon and Transport (ACT)-America Earth Venture Suborbital mission was directed toward improving the accuracy of regional greenhouse gas flux estimates, specifically through refining our understanding and implementation of CO_2 and methane (CH_4) transport, fluxes and flux uncertainties and background levels in inverse models (Davis et al., 2021).

75 During this mission, five, six-week seasonal research flight campaigns were conducted between 2016 and 2019 for evaluation and subsequent improvement of terrestrial carbon cycle models and filling gaps in the eastern U.S. carbon monitoring network. With a strong focus on transport of carbon via weather systems, atmospheric trace gases and meteorological variables were also observed across multiple synoptic cycles, across cold and warm air sectors and from the boundary layer to the upper troposphere. While the ACT mission was not targeting large carbon emissions signals in urban areas as in the abovementioned

80 studies, we focus on carbon flux attribution and model evaluation here. We analyze a spatially dense set of airborne whole-air flask $\Delta^{14}CO_2$ measurements collected during ACT in the Mid-Atlantic U.S. alongside those of other gas-phase species such as carbonyl sulfide (OCS) and carbon monoxide (CO) to first distinguish between seasonal fossil fuel- and biogenically-driven atmospheric CO_2 variability with respect to background levels. Second, we apply the methodology of previous authors (Levin and Karstens, 2007; Vogel et al., 2010; Turnbull et al., 2011a; Maier et al., 2023) to the broad, Mid-Atlantic U.S. region,

85 calculating high-frequency fossil fuel CO_2 enhancements using in situ CO observations, and investigate the utility of this technique for evaluating modeled transport of atmospheric fossil fuel-CO_2 enhancements above background levels for an example ACT research flight along the northeastern corridor.

2 Data and Methods

2.1 ACT-America flight campaigns

90 During the ACT mission, the eastern half of the coterminous U.S. was broadly surveyed by two research aircraft: the NASA Langley Beechcraft B-200 King Air and the NASA Wallops C-130 Hercules. Of the Mid-Atlantic, Midwestern and Southern U.S.-focused measurement regions, air samples for $\Delta^{14}CO_2$ measurement were collected only in the Mid-Atlantic region. Sampling in the Mid-Atlantic region focused heavily on the extensive forests and croplands with flights capturing distant urban influence, and a selected few flights specifically targeted the urban northeastern corridor. The first ACT campaign occurred

95 during summer 2016 (July 18th -August 28th) and is omitted from this analysis due to the absence of $\Delta^{14}CO_2$ sampling. Four subsequent ACT campaigns described here occurred during winter 2017 (March 1st-10th), fall 2017 (October 1st-15th), spring 2018 (May 4th-20th), and summer 2019 (July 7th-27th) for a total of 26 flights with atmospheric sampling for $\Delta^{14}CO_2$. Research

flights were primarily conducted between 11:00 and 18:00 local time to sample a well-developed atmospheric boundary layer (ABL), through long, level-leg flight transects. Airborne atmospheric sampling during individual research flights was focused at altitudes within the ABL where greenhouse gas abundances are strongly influenced by surface fluxes with occasional sampling in the free troposphere (FT, ~2400 to 9000 m above mean sea level (ASL)) to determine chemistry and isotopic composition of background air against which observed ABL enhancements or depletions are assumed to occur. Because a scientific goal of ACT was to improve the simulation of atmospheric transport of greenhouse gases in regional-scale inversions under real-world conditions, research flights were organized to sample across synoptic weather patterns (Pal et al. 2020). Wei et al., (2021) describe the ACT observational and numerical data products.

2.2 Whole-air sample collection and flask radiocarbon measurements

NOAA Programmable Flask Packages (PFPs) were installed on each aircraft for the collection of whole-air samples. Each PFP contains 12, 0.7 L borosilicate glass flasks and for ACT campaigns between 2017 and 2019, these automated systems were connected to a Peltier gas chiller to efficiently dry air samples to less than 1% atmospheric water vapor prior to air sample collection. A Programmable Compressor Package (PCP) pressurized each flask to 275 kPa yielding ~ 2 standard liters (L) of air collected per flask. Technical details of these flask sampling systems used during ACT are described in Baier et al. (2020).

For each flight day, six to twelve total flasks were filled on each aircraft platform between the altitudes of 300-9000 m ASL. The ratio of well-mixed ABL to FT flasks sampled during ACT was generally 5:1. Because approximately 2 L of sample air is required for high precision $\Delta^{14}CO_2$, in addition to the long-lived greenhouse gases (CO_2, CH_4) and other low concentration gases such as OCS and CO, two PFP flasks were filled in parallel (i.e. 'paired sampling') when sampling for $\Delta^{14}CO_2$ measurements. Post-flight, PFPs were transferred to the NOAA Global Monitoring Laboratory and the University of Colorado/INSTAAR Stable Isotope Laboratory for measurement. The first flask of each paired sample is analyzed for greenhouse gases, OCS, and a suite of hydrocarbons, halocarbons, and stable isotope ratios ($^{13}C:^{12}C$) of CO_2 and methane using methods described in Baier et al. (2020), online (https://gml.noaa.gov/ccgg/aircraft/analysis.html) and in Vaughn et al. (2004). Remaining sample air in the first flask after these measurements is combined with the entire second, paired flask for a single $\Delta^{14}CO_2$ measurement. The University of Colorado INSTAAR Laboratory for AMS Radiocarbon Preparation and Research (NSRL) conducts the CO_2 extraction from flask sample air. Pure CO_2 is archived until near the time of $\Delta^{14}CO_2$ measurement, at which time the pure CO_2 is graphitized to pure carbon "targets" and these graphitized samples are transferred to the University of California at Irvine for high-count accelerator mass spectrometry (AMS) ^{14}C measurement (Lehman et al., 2013).

2.3 Radiocarbon-based CO₂ partitioning

Radiocarbon (^{14}C) is produced naturally in the troposphere and stratosphere via cosmic ray-induced reactions between neutrons and atmospheric nitrogen (N_2) (Libby, 1946). Measurements are presented as $\Delta^{14}CO_2$ in units of per mil (‰), or the part per thousand deviation of the measured ^{14}C:C ratio from that of the international measurement standard material, after correction for mass dependent fractionation and radioactive decay since the date of collection (Stuiver and Polach, 1977). Given that the ^{14}C half-life is approximately 5700 years (Godwin, 1962), and that fossil fuel carbon has been isolated from the atmosphere for millions of years, the radiocarbon content of fossil fuels is zero and therefore $\Delta^{14}CO_{2ff}$ is -1000 ‰ on the delta (Δ) scale. Therefore, addition of fossil fuel CO_2 to the atmosphere produces a quantitative reduction in atmospheric $\Delta^{14}CO_2$ over and downwind of emissions sources. Paired observations of $\Delta^{14}CO_2$ and CO_2 in flasks sampled in the Mid-Atlantic region can thereby provide a robust method for distinguishing between terrestrial biogenic (CO_{2bio}) and fossil fuel (CO_{2ff}) CO_2 sources given a known background. The CO_2 mass balance over land is:

$$CO_{2obs} = CO_{2bg} + CO_{2ff} + CO_{2bio}. \qquad (1)$$

Expanding CO_{2bio}, which results from Net Ecosystem Exchange, into a sum of photosynthetic (*photo*) and respiration (*resp*) fluxes ($CO_{2resp} + CO_{2photo}$) and adding the respective isotopic signatures ('Δ', for $\Delta^{14}CO_2$) for all terms gives Equation (2):

$$\Delta_{obs}CO_{2obs} = \Delta_{bg}CO_{2bg} + \Delta_{ff}CO_{2ff} + \Delta_{resp}CO_{2resp} + \Delta_{photo}CO_{2photo} + N^{14}C_{nuc}/R_{std}, \qquad (2)$$

where the individual budget terms are the product of the isotopic ratio and CO_2 mole fraction, which is a conserved quantity (Tans et al., 1993). Here, *obs* is the observed atmospheric mole fraction of CO_2 in the ABL and is a function of variations in fossil fuel (*ff*) and biogenic (*bio*) CO_2 differences from a varying CO_2 background (*bg*) level. Photosynthesis and respiration contributions are stated separately in the isotopic mass balance (Eq. 2), because they carry different isotopic signatures. This isotopic "disequilibrium" between photosynthetic and respiration fluxes arises from the rapidly decreasing ($\sim$ -5 ‰ yr^{-1}) background $\Delta^{14}CO_2$ (e.g., Lehman, 2013; Graven et al., 2020). Δ_{resp} is more positive than Δ_{photo} because Δ_{resp} is associated with biospheric CO_2 fixed by photosynthesis when the ambient atmospheric $\Delta^{14}CO_2$ (i.e. Δ_{bg}) was higher. Δ_{photo} is assumed equal to Δ_{bg} because the 'Δ' notation accounts for mass-dependent fractionation processes such as those occurring during photosynthesis. There is also a small impact on Δ_{obs} from pure ^{14}C emitted from certain nuclear reactors, which is added as the last term in Equation 2 (Graven and Gruber, 2011). Here N is the mass dependent Δ normalization factor, which is close to 1 (Basu et al., 2016) and $^{14}C_{nuc}$ is the mole fraction of $\Delta^{14}CO_2$ resulting from the ^{14}C flux from reactors and R_{std} is the standard material ^{14}C/C ratio.

Combining Equations (1) and (2) results in:

$$CO_{2ff} = \frac{CO_{2obs}(\Delta_{obs}-\Delta_{bg})}{(\Delta_{ff}-\Delta_{bg})} - \frac{CO_{2resp}(\Delta_{resp}-\Delta_{bg})}{(\Delta_{ff}-\Delta_{bg})} - N^{14}C_{nuc}/R_{std}(\Delta_{ff}-\Delta_{bg}). \tag{3}$$

The variables in Eq. (3) are either known or can be measured directly. Historically, CO_{2ff} has been approximated by only the first term on the right hand side of Equation (3) (Levin et al., 2003). Disequilibrium and nuclear fluxes both add ^{14}C to the atmosphere. A correction (CO_{2corr}) is made to CO_{2ff} (Turnbull et al., 2009) to unmask these effects and reveal the full magnitude of the fossil fuel-CO_2 emissions signal, making it somewhat convenient to rewrite Equation (3) as:

$$CO_{2ff} = \frac{CO_{2obs}(\Delta_{obs}-\Delta_{bg})}{(\Delta_{ff}-\Delta_{bg})} - CO_{2corr}. \tag{4}$$

Flask ABL measurements for CO_{2obs} are defined as those samples collected at altitudes less than 1500 m ASL for all ACT flight campaigns. ABL samples typically contain the strongest regional flux signatures, and thus are likely to yield the largest $\Delta^{14}C$ signals. Conversely, we determine daily Δ_{bg} and CO_{2bg} values from measurements of $^{14}CO_2$ and CO_2 in flasks sampled at altitudes greater than 4000 m ASL for all flight campaigns to capture air that is within the well-mixed FT in all seasons. The FT data, while perhaps not always representative of the true, local-scale background signature (e.g. Turnbull et al., 2015), is expected to provide a reliable estimate of the regional- to continental-scale background (Baier et al., 2020). Here, Δ_{bg} in Eq. 2 is derived as the daily mean FT $\Delta^{14}CO_2$, typically between 4000 and 9000 m ASL. All flask FT measurements are filtered to remove samples with high CO (above three standard deviations from a smooth curve fit), which can indicate the potential influence of local pollution on continental background values. Similarly, CO_{2bg} and other trace gas background mole fractions were derived as mean mole fractions of FT values for these species. For the ACT study, CO_{2corr} is derived for each flask sample by convolving estimated gridded monthly disequilibrium fluxes of $\Delta^{14}CO_2$ from the terrestrial biosphere and fluxes from nuclear reactors with surface influence functions for each flask sample location (i.e., for each flask "receptor", per below).

Surface influence functions (referred to as "footprints"), representing the sensitivity of air parcels at a particular receptor location to upwind surface fluxes (in units of ppm m^2 s µmol^{-1}), are derived by first using the Hybrid Single-Particle Lagrangian Integrated Trajectory (HYSPLIT) model (Draxler and Hess, 1997) running 500 randomly perturbed, 10-day back trajectories for all ABL flask receptors. Back trajectories are driven by 27-km horizontal resolution Weather Research and Forecasting (WRF) meteorological fields, nudged to the ERA-5 reanalysis meteorology within the ACT-America North American model domain (Feng at al., 2021b, see Section 2.4). Individual footprints are calculated for all 500-particle trajectory ensemble members based on their residence time over a given gridded area (Lin et al, 2003) within the surface boundary, defined as a vertical column of air in each 1º x 1º grid cell between 125 m AGL and 50% of the WRF-calculated atmospheric boundary layer height (Fig. 1). Disequilibrium fluxes from the terrestrial biosphere are derived by convolving impulse-response functions

from the CASA biogeochemical model (Thompson and Randerson, 1999) with the atmospheric history of ^{14}C (Miller et al. 2012; LaFranchi et al., 2016, with updates in Zhou et al., 2020) while nuclear reactor fluxes are obtained from Graven et al. (2011), effectively assuming regional nuclear $^{14}CO_2$ has not changed significantly over time through 2019. These gridded fluxes (monthly, time-dependent in the case of disequilibrium flux) are converted to the two components of CO_{2corr} via

195 convolution with footprints derived from HYSPLIT and are calculated individually for each ABL flask sample pair for which CO_{2ff} is derived using Equation (4). The magnitude of CO_{2corr} attributable to the terrestrial biosphere in the Mid-Atlantic U.S. is approximately 0.10 ± 0.04 ppm CO_{2ff} (1σ) during the winter deployment and exhibits a seasonal cycle with a maximum during the summer of 0.70 ± 0.28 ppm CO_2 (1σ), indicating similar seasonal behavior as that calculated by Miller et al. (2012) for the Mid-Atlantic U.S. The magnitude of CO_{2corr} attributable to nuclear reactor emissions in the Mid-Atlantic U.S. is 0.25

200 ± 0.44 ppm CO_2 and does not exhibit a clear seasonal pattern.

2.4 Forward model CO_{2ff} using the PSUWRF model

A separate, Eulerian 27 km horizontal resolution implementation of WRF-Chem version 3.6.1 run (Feng et al., 2021a,b) was developed and implemented during the ACT time period (2016-2019) to simulate atmospheric transport of CO_2 component fluxes along all flight tracks for each seasonal campaign. Note that, while the Lagrangian HYSPLIT model mentioned above

205 to calculate CO_{2corr} in Equation 4 for each flask measurement also used WRF for input meteorological fields, this was not identical to PSUWRF used for the Eulerian simulations. PSUWRF model runs were implemented prior to this manuscript development and could not be re-run, but both WRF versions used similar setups and thus we expect there to be no major inconsistencies between the use of these two different versions for calculating CO_{2ff} using flask or model output.

210 PSUWRF separately simulates the contributions of CO_2 boundary conditions, and biogenic, ocean, fossil fuel and biomass burning CO_2 fluxes to total CO_2 at any time and location in the model domain (Feng et al., 2019a,b). The microphysics, PBL and cumulus parameterization schemes used in this run are Thompson microphysics, the Mellor-Yamada-Nakanishi-Nino version 2 (MYNN2), and the Kain-Fritsch schemes, respectively (Feng et al., 2019a). As the ACT campaigns spanned multiple years, CO_2 oceanic, fossil fuel and biomass burning fluxes, and total CO_2 boundary conditions come from CarbonTracker

215 version 2017 (CT2017) through 2017 and CT-NRTv2019-2 (Jacobson et al., 2020) afterwards. CT2017 and CT-NRTv2019-2 together are called 'CT' for simplicity. The CT fossil fuel tracer used here is derived from an average of the ODIAC and Miller emissions datasets (Jacobson et al., 2020). These emissions datasets when run separately in PSUWRF between 2016-2018, indicate agreement to within ~2 ppm. The two datasets are thereafter averaged to create a single flux product as in CT for model runs in 2019. Here, simulated fossil fuel CO_2 (CO_{2ffmod}) is calculated using the same approach as flask CO_{2ff}, where

220 average FT values of this tracer are subtracted from ABL values. The PSUWRF model transport is nudged to ERA5 reanalysis data to improve the depiction of atmospheric transport relative to ACT in-flight meteorological observations (Gerken et al., 2021). PSUWRF model output was extracted to ACT research flight track locations at hourly resolution for comparison and evaluation relative to high-frequency, measurement-based CO_{2ff} described in Section 2.5.

2.5 Calculation of "pseudo" CO_{2ff}

225   As in previous studies (Levin and Karstens, 2007; Vogel et al., 2010; Turnbull et al., 2011a, Maier et al., 2023), we employ in situ CO measurements for estimating CO_{2ff} (referred to as "pseudo-CO_{2ff}" and denoted as CO_{2ff}') at high temporal resolution. ACT in situ CO measurements were made by cavity ring-down spectroscopy at 0.4 Hz, drift-corrected during flight and calibrated once weekly using gas standards on the World Meteorological Organization (WMO) X2014A CO scale, just as NOAA flask-based measurements (Wei et al., 2020; DiGangi et al., 2021). After calibration, CO data were averaged to 0.2

230   Hz, the maximum frequency of CO_{2ff}'. We calculate CO_{2ff}' using Equation (5) by applying a median ratio of CO to CO_{2ff} (R_{CO}) calculated from flask samples collected each day (typically from 5-12 flask samples), and apply this value for the in situ CO time series collected on that day's ACT research flight:

$$CO_{2ff}' = \frac{CO_{obs}' - \overline{CO'_{bg}}}{R_{CO}} .$$ (5)

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Here, CO_{obs}' are 0.2 Hz ABL (below 1500 m ASL) in situ CO dry mole fraction observations and $\overline{CO'_{bg}}$ is the mean daily CO mole fraction observed in the FT (above 4000 m ASL) for each flight day. In situ FT measurements are filtered to remove high CO or local emissions influence. Similar to above, data with CO greater than three standard deviations from a smooth curve fit to background values are filtered. Each of these terms is derived exactly as the analysis of flask CO_{2ff} above. Due to the

240   limited number of flasks sampled on each flight, R_{CO} is calculated as a daily median CO enhancement ($CO_{obs}' - \overline{CO'_{bg}}$, or ΔCO) ratio to CO_{2ff} and we assume that this daily median value is representative of the entire flight region each day. In the analysis that follows, we assess the utility of this "pseudo-CO_{2ff}" method for determining fossil fuel CO_2 in evaluating modeled CO_{2ff} over larger regions where routine, high-density discrete flask sampling is not available. In this work, CO_{2ff}' is calculated only for flights when flask measurements were sampled and analyzed for $\Delta^{14}CO_2$ in order to reduce errors associated with

245   assumptions in R_{CO} spatial and temporal variability.

2.6 CO_{2ff} uncertainty derivation

The total estimated uncertainty in CO_{2ff} is determined by a propagation of estimated uncertainties for individual budget terms through Equation (4). Flask-based CO_{2obs} and Δ_{obs} measurement uncertainties are 0.1 ppm and 1.8‰, respectively. Here Δ_{ff} is, by definition -1000‰, and carries no uncertainty. The uncertainty in Δ_{bg} is calculated through propagation of error in a) Δ_{obs}

250   and b) the mean diurnal standard deviation of FT $\Delta^{14}CO_2$ observations (Δ_{bg}) for each ACT flight. The relative uncertainty in CO_{2corr} associated with biospheric disequilibrium and nuclear reactor fluxes is assumed to be 50% for each term. Daily aggregated CO_{2ff} sample uncertainties are determined by using a 1,000-member Monte Carlo simulation that propagates these uncertainties through Equation (4) to provide an estimate of the flask-derived CO_{2ff} uncertainty per research flight. The average

campaign-wide CO_{2ff} uncertainty (from the inner 68% confidence interval of the 1000-member CO_{2ff} normal distribution per research flight) is 1.22 ppm.

Similar to above, we calculate an average campaign uncertainty in CO_{2ff}' of 4.96 ppm (from the inner 68% confidence interval in CO_{2ff}' of a 1,000-member Monte Carlo approach for each flight day) using Equation (5). We estimate daily uncertainties by conservatively incorporating the CO dry mole fraction measurement precision on 0.2 Hz measurements (5 ppb), the uncertainty in daily background CO from CO_{obs} and the daily standard deviation in FT CO, and the uncertainty in R_{CO} from the width of a normally-distributed 68% confidence interval about average R_{CO} values.

The uncertainty in both CO_{2ff} and CO_{2ff}' is higher for ACT research flights due to the uncertainty caused by using a single CO or $\Delta^{14}CO_2$ background value for each flight, when in truth, different sectors of a research flight (e.g. the cold and warm sectors during frontal crossing flights) could experience different background conditions or could be influenced by deep convective mixing, violating the assumption that the continental background value is described by the FT value. Elevating the altitude threshold at which background values are defined (i.e. 4000 m ASL) can better isolate continental background air from lower tropospheric air that has been mixed with local emissions sources. However, this can also lead to higher uncertainties in FT air that is considered as a background value for local ABL measurements due to the fact that this air at higher altitudes may have experienced longer-range transport.

3 Results and Discussion

3.1 Mid-Atlantic $\Delta^{14}CO_2$ background in general agreement with NOAA aircraft network

Figure 2a shows the vertical distribution of $\Delta^{14}CO_2$ measured for all seasonal campaigns between winter 2017 and summer 2019, highlighting this difference and indicating generally lower values and significantly greater variation relative to the FT throughout the bottommost troposphere due to surface emissions of ^{14}C-free fossil fuels.
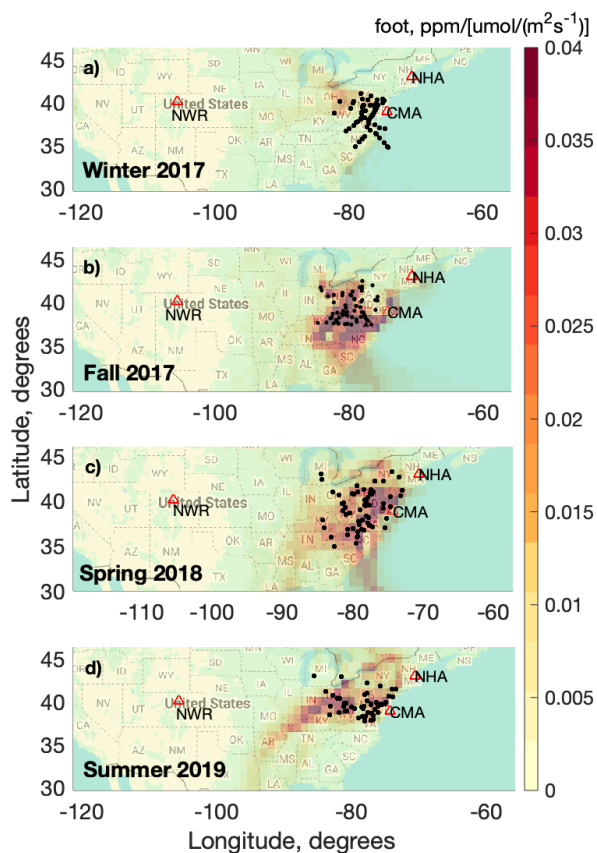
Figure 1. Surface influence for ABL $\Delta^{14}CO_2$ samples measured for ACT seasonal deployments in a) winter 2017, b) fall 2017, c) spring 2018, and d) summer 2019. ABL flask sample locations are denoted in white, with GGGRN Aircraft Program network sites indicated at Niwot Ridge, CO (NWR), Cape May, NJ (CMA), and Portsmouth, NH (NHA). Plot maps are produced using Google Maps API (©Google Maps).

Figure 2b shows the decrease in atmospheric $\Delta^{14}CO_2$ in the FT over time during ACT and within the broader NOAA Global Greenhouse Gas Reference Network (GGGRN) due to global-scale addition of fossil fuel CO_2 and justifies the use of ACT FT for a definition of background atmospheric conditions. As expected, values of Δ_{bg} throughout the upper FT are relatively

280   homogenous during each campaign and are in rough agreement with other FT flask $\Delta^{14}CO_2$ measurements within the GGGRN (Sweeney et al., 2015; Miller et al., 2012) such as Cape May, NJ (CMA) and Portsmouth, NH (NHA). For winter, fall and spring seasons, ACT FT $\Delta^{14}CO_2$ is slightly higher than that in the GGGRN sites, including the upwind, high-altitude (3000-4000 MASL) Niwot Ridge, CO (NWR) site. During ACT, FT samples are frequently obtained from higher altitudes than the GGGRN $\Delta^{14}CO_2$ paired flask samples with fixed collection altitudes, resulting in the potential for ACT FT air to originate

285   from different latitudes and deviations of ~ 2‰ around GGGRN $\Delta^{14}CO_2$ observations. Early in the ACT winter 2017 campaign, FT $\Delta^{14}CO_2$ is lower than the mean values in the GGGRN (Fig. 2b), and is accompanied by higher than average CO mole fractions. This result potentially indicates a small influence from local pollution sources that are depleted in FT $\Delta^{14}CO_2$. Nonetheless, these ACT FT data agree with (± 1σ standard deviation) GGGRN continental background or FT values and are included in this analysis.
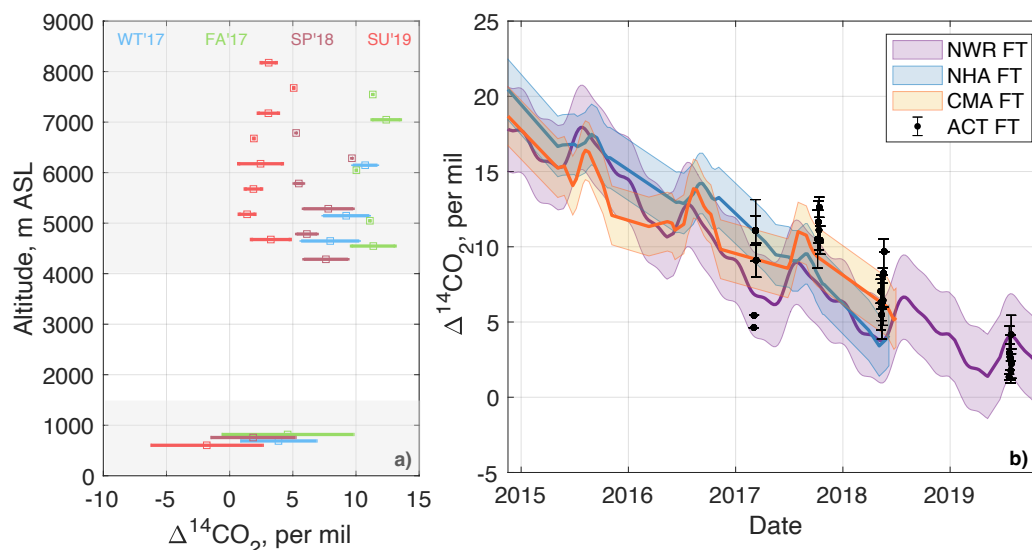
Figure 2. a): Vertical distribution of $\Delta^{14}CO_2$ during all ACT campaigns in the Mid-Atlantic U.S. ABL values shown are the average of all observations below 1500 m ASL (values are separated vertically for clarity) for each seasonal campaign while horizontal bars represent their 1σ standard deviation in time. FT values (altitudes greater than 4000 m ASL) shown are 500-m binned averages and their 1σ standard deviation in $\Delta^{14}CO_2$ for each 500 m altitude bin. Shaded regions represent ABL and FT definitions. b): Daily mean ACT FT $\Delta^{14}CO_2$ background measurements (± 1σ standard deviation) compared to NOAA GGGRN aircraft network $\Delta^{14}CO_2$ FT (~4000 m ASL) measurements. GGGRN data are seasonal (3-month averaged) aircraft FT $^{14}CO_2$ obtained offshore of Cape May, NJ (CMA) and offshore of Portsmouth, New Hampshire (NHA) and surface $\Delta^{14}CO_2$ for the high-altitude site, Niwot Ridge, CO (NWR) at ~3500 m ASL.

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Figure 3a highlights the spatial extent of flask sampling during the ACT mission and density of flask sampling over the four seasonal campaigns with an average $\Delta^{14}CO_2$ ABL-FT difference of -5‰. Using Equation (4), CO_{2ff} is calculated for all flask samples, and general agreement between largely negative $\Delta^{14}CO_2$ ABL signals and higher CO_{2ff} signals are seen throughout ACT (Fig. 3b). ACT observed CO_{2ff} ranges from -0.8 to 15.5 ppm, noting that negative CO_{2ff} values are not physically realistic

295 but occur due to the measurement uncertainty of $\Delta^{14}CO_2$. Compared to other surface and aircraft observations of CO_{2ff} throughout the U.S., the range of ACT CO_{2ff} is comparable to CO_{2ff} measured in other studies in the Mid-Atlantic U.S. (Miller et al., 2012: -3-13 ppm CO_{2ff}), downwind of Sacramento, CA (Turnbull et al., 2011a: 2.4-8.6 ppm CO_{2ff}) and in Colorado urban regions (Graven et al., 2009: 0-20 ppm CO_{2ff}). Maximum values of CO_{2ff} measured during ACT were however much lower than those over the densely populated urban area of Los Angeles, CA (Miller et al., 2020: -1.3-48.4 ppm CO_{2ff}) due to

300   its more rural sampling focus. Few ACT flight legs were aimed at capturing local, urban-scale CO_{2ff} signals. As such, the
analysis below requires averaging of these signals as a more robust way to qualitatively analyze the ACT flask CO_{2ff} data.
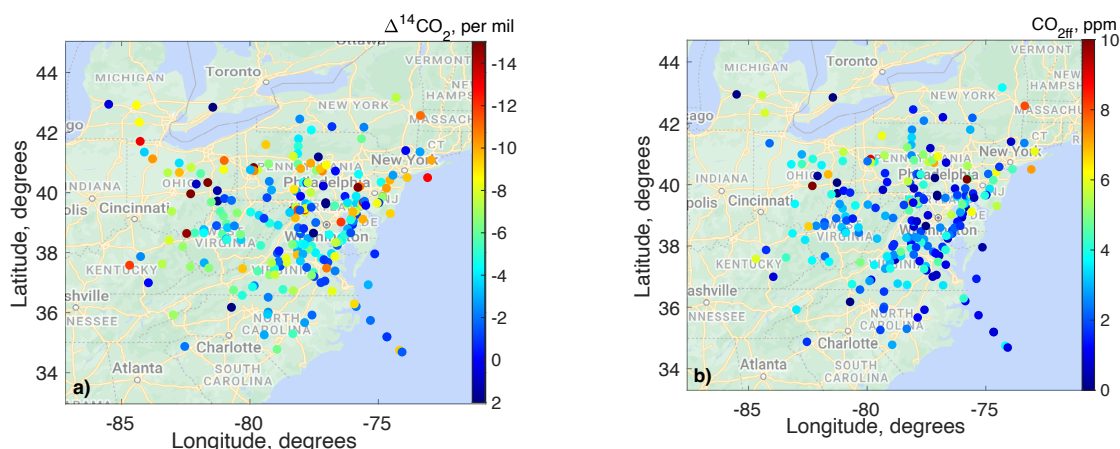
Figure 3: a) $\Delta^{14}CO_2$ ABL-FT differences for all seasons in the Mid-Atlantic U.S. Note that the color bar is reversed to indicate warmer colors for more negative $\Delta^{14}CO_2$. b) ABL CO_{2ff} calculated for the same flask samples shown in a) for all seasons in the Mid-Atlantic U.S. Plot maps are produced using Google Maps API (©Google Maps).

3.2 Partitioning of seasonal CO_{2tot} indicates biogenic-driven variability

For consistency throughout ACT seasonal campaigns, ABL values for all species in the subsequent analysis are reported as an
305   ABL-FT difference. Figure 4 shows CO_{2ff}, alongside biogenic CO_2 (CO_{2bio}) calculated from Equation (1), displayed by month
of the year for ACT deployments in winter (2017), spring (2018), summer (2019) and fall (2017) in the Mid-Atlantic United
States. Total (CO_{2tot}), fossil fuel and biogenic CO_2 are shown to indicate relative contributions to the observed CO_{2tot} variability
(Eq. 1). CO_{2tot} observed in the ABL varies seasonally; while most often positive for flights conducted in late fall and winter,
CO_2 depletion is seen during summer as expected due to net photosynthetic uptake by the terrestrial biosphere. Seasonal
310   changes in ABL CO_{2tot} were driven by CO_{2bio}, which varied from -19.7 ppm during summer flights to 16.2 ppm in late fall. In
early fall, negative CO_{2bio} was observed (Fig. 4) from strong surface influence and net uptake in southwestern Virginia (Fig.
1). For the remainder of the fall campaign, flights indicated generally positive CO_{2bio} with net respiration and weaker surface
influence. CO_{2ff} has a lower contribution to CO_{2tot} variability during ACT in the Mid-Atlantic region from one campaign to
the next, with a positive contribution of 3.3 ± 2.0 ppm. Given the small number of flights shown over four years, it should be
315   noted that the seasonality in CO_{2ff} may not represent true climatological CO_{2ff} variability and should not be interpreted in terms
of uniform fossil fuel emissions because there is likely seasonal variation in ABL depth, which can counteract seasonally-
varying emissions.
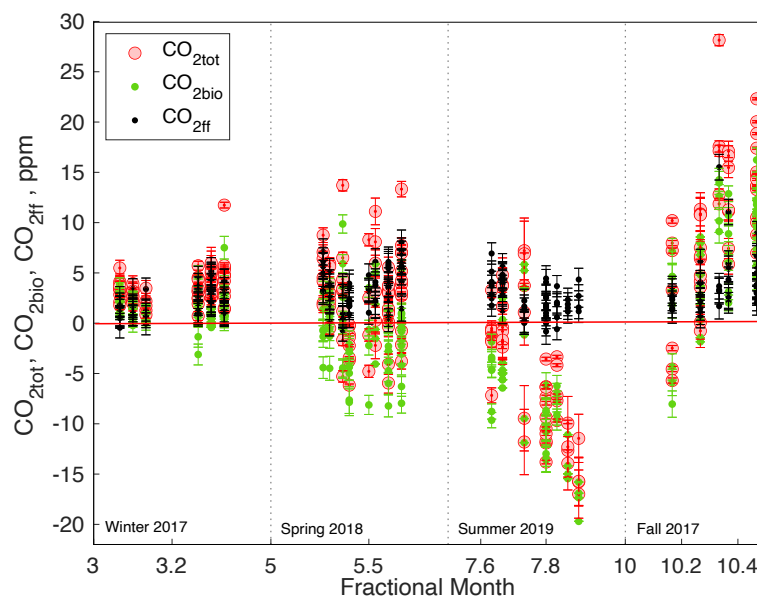
Figure 4. Calculated total CO₂ (CO$_{2tot}$), partitioned into biogenic CO₂ (CO$_{2bio}$) and fossil fuel CO₂ (CO$_{2ff}$) components using Eq. (4) in winter 2017; spring 2018; summer 2019; and fall 2017. As seasonal campaigns were not conducted consecutively within a single year, we examine variability according to a "climatological" year. Note that the x-axis is intentionally non-linear to show intra-day variability. One sigma error bars are shown for CO$_{2tot}$, CO$_{2bio}$, and CO$_{2ff}$.

The flask sampling region for this study encompasses rural to suburban regions, reducing signal to noise in $\Delta^{14}CO_2$ and
320   complicating the interpretation of CO$_{2tot}$ observed due to a wide mixture of surface sources (and sinks). As CO$_{2tot}$ is largely controlled by CO$_{2bio}$, we correlate (using Pearson's R^2 values) and regress CO$_{2tot}$ with CO$_{2ff}$ and with CO$_{2bio}$ (see Table S1 of supplementary material). Regressions between CO$_{2bio}$ and CO$_{2tot}$ result in slopes between 0.86 and 0.92 for all seasons. Strong correlations between CO$_{2bio}$ and CO$_{2tot}$ exist from spring through fall with R^2 values between 0.73 and 0.92. Somewhat strong relationships between CO$_{2tot}$ and CO$_{2bio}$ were found during the winter 2017 campaign with a regression slope close to 1 but
325   relatively weak correlation ($R^2 \sim 0.3$), indicating that observed CO₂ had a non-negligible influence from biogenic CO₂ exchange as was also found in previous studies (Potosnak et al., 1999; Turnbull et al., 2006; Miller et al., 2012; Baier et al., 2020). Correlations between CO$_{2tot}$ and CO$_{2ff}$ are weak, with associated regression slopes highest during winter and smaller slopes observed for the spring, summer, and fall deployments when ecosystem CO₂ fluxes are higher.

330   The fraction of negative CO$_{2tot}$ during biogenically active months in spring (March-April), summer (July), and fall (October) are 28%, 62%, and 4%, respectively (Table S1). However, when examining the fraction of negative CO$_{2bio}$ (indicating net CO₂

uptake), these percentages are much larger at 57%, 82%, and 8%, respectively. The difference between negative CO_{2tot} and negative CO_{2bio} fractions of 20-30% during months when the ecosystems are more active highlights the importance of using $\Delta^{14}CO_2$ to enable partitioning of CO_{2tot} into fossil fuel and biogenic components; this would not be possible when considering
335   CO_{2tot} alone, as a portion of this flux component would be masked by fossil fuel emissions.
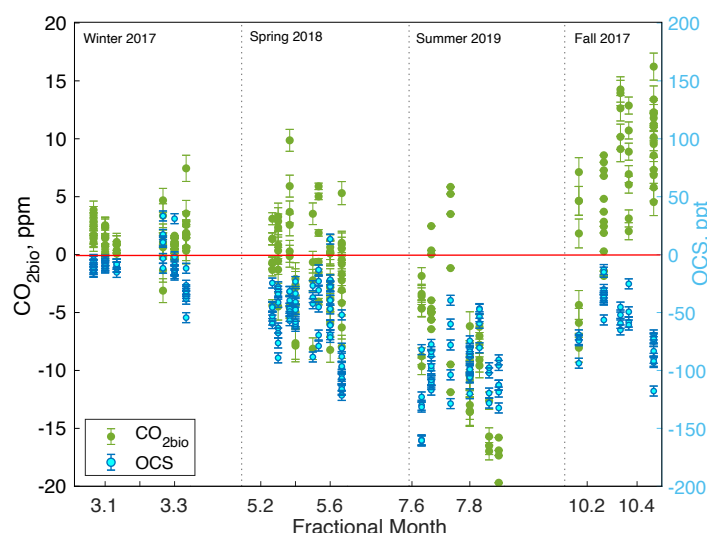
Figure 5. ABL biogenic CO_2 (CO_{2bio}) and OCS in winter 2017; spring 2018; summer 2019; and fall 2017. As seasonal campaigns were not conducted consecutively within a single year, we examine variability according to a "climatological" year. Note that the x-axis is intentionally non-linear to show intra-day variability. One sigma error bars are shown for CO_{2bio} and OCS.

We also examine carbonyl sulfide (OCS) ABL-FT differences alongside CO_{2bio} as an additional tracer for biogenic processing as OCS is taken up by plants similarly to CO_2 but not respired (Montzka et al., 2007; Campbell et al., 2008, Fig. 5). We observe a strong positive correlation with CO_{2bio} and OCS in Figure 5 and when combining data from winter, spring, and summer
340   seasons where CO_{2bio} is either negative or weakly positive (see Fig. S1a of supplementary material), consistent with photosynthesis explaining negative CO_{2bio} values. A modest positive correlation (R = 0.54) is seen when combining all seasonal ACT data. Including or excluding the winter data, given the insignificant slope between OCS and CO_{2bio} (Table S1), does not affect this correlation.

345   For fall 2017, the overall correlation between CO_{2bio} and OCS is low and not significant (Table S1). However, statistically significant correlations are seen between negative CO_{2bio} and OCS in early fall due to continued uptake of CO_2 by the terrestrial biosphere (Fig. 5). A moderate negative correlation is observed between OCS and CO_{2bio} in late fall after net respiration is seen (Fig. 5, Fig. S1b). During this time, CO_{2bio} of up to ~16 ppm occurs alongside OCS ABL-FT differences of almost -120 ppt. The net negative correlation during fall between CO_{2bio} and OCS indicates that there is ecosystem OCS uptake occurring
350   after net photosynthesis has stopped such as soil uptake (Whelan et al., 2022) or uptake from senescing or decaying fall

vegetation (Sun et al., 2016) that is decoupled from gross primary productivity. The fact that fall OCS ABL-FT differences are as large as those found in summer likely relates to ABL signals of trace gases being accentuated by weaker vertical mixing during this time of the year. While OCS uptake is seen throughout the ACT seasonal campaigns, CO_{2tot} changes sign to indicate net respiration in the fall and winter (Montzka et al., 2007; Kuai et al., 2022, Ma et al., 2023); correlations between negative

355 CO_{2tot} and OCS weaken in the early fall due to proportionally high fossil fuel emissions, providing insufficient information about gross primary productivity. Parazoo et al. (2021) found that models underestimate observed CO_{2tot} during ACT, and have decreased fidelity in reproducing carbon gross primary productivity inferred from observations throughout the U.S. Since stronger relationships emerge between OCS with CO_{2bio} than with CO_{2tot}, using OCS as an additional tracer could further inform and constrain these model processes.

360 **3.3 Example Case: "Pseudo-CO_{2ff}" as a product for evaluating model error**

We use CO_{2ff} as a model "transport tracer", to examine model errors as CO_{2ff} fluxes are assumed to be relatively well-known (Turnbull et al., 2009). Well-known emissions and well-measured mole fractions directly tracing those emissions allow us to evaluate the atmospheric tracer transport that connects the two. Here, we compare a "pseudo-CO_{2ff}" or high-frequency CO_{2ff}' (Equation 5) to CO_{2ffmod} simulated using the PSUWRF forward model, based on Miller fossil fuel emissions (Jacobson et al.,

365 2020). This CO_{2ff}' product being high-frequency can also capture more diurnal variability in fossil fuel CO_2, and therefore evaluate modeled fossil fuel CO_2 more thoroughly, than measurements made using discrete flask samples.

3.3.1 Variability and uncertainty of R_{CO} and CO_{2ff}

Our CO_{2ff}' estimation relies on the assumption that the emission ratio of CO to CO_{2ff} (R_{CO}) is constant for a given flight day. We note that the numerator of R_{CO} represents the enhancement in all CO sources (fossil, but also potentially biomass burning

370 and oxidation of volatile organic compounds (VOCs)) as well as a very small amount of CO oxidative loss. Because all of these sources are "calibrated" relative to CO_{2ff} (the denominator), it is not necessary to explicitly define non-fossil components of this ratio.
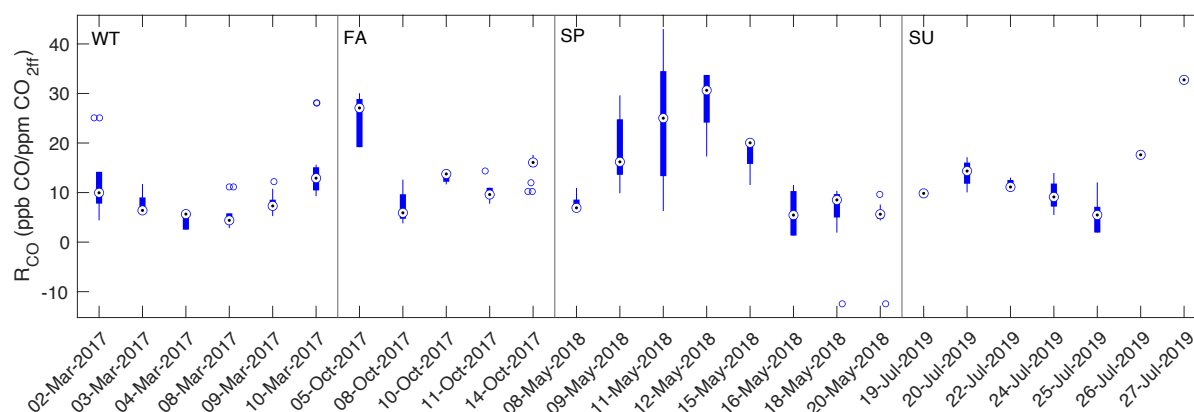
Figure 6. Daily R_{CO} calculated from discrete flask samples during winter (WT, 2017), fall (FA, 2017), spring (SP, 2018), and summer (SU, 2019). Median R_{CO} for each research flight is shown as a bullet point, with the interquartile (IQ) range shown as the thick bar line. Outliers greater than 1.5 times the IQ range are shown as separate open circles, and maximum extreme values as "whiskers" that do not qualify as outliers (thin lines) extending from the IQ ranges. Flask data shown are combined for B-200 and C-130 aircraft.

375 Figure 6 indicates that the median R_{CO} calculated for all ACT missions (10.03 ± 8.15 ΔCO per ppm CO_{2ff}; 68% CI) is slightly higher than in urban studies (Turnbull et al., 2011a; Miller et al., 2012; Turnbull et al., 2014), possibly due to CO additions from VOCs throughout the ACT Mid-Atlantic region. Day-to-day or even diurnal variability within flasks sampled during each seasonal deployment is largely variable, and while it is unclear which process causes this variability, variability in trace gas background levels (currently represented by a single daily value) or spatial differences in VOC oxidation could be

380 influencing factors. Despite the expectation that greater oxidation of VOCs will produce more CO during the summer, we see the highest average R_{CO} values during Spring.

As mentioned above, we calculate CO_{2ff}' only for days when flask measurements were analyzed for $\Delta^{14}CO_2$. When comparing CO_{2ff}' with flask-based CO_{2ff}, the mean bias is 0.7±2.1 ppm (1σ) (Fig. 7). The relatively large standard deviation in Fig. 7

385 indicates that flask samples collected within a single research flight were not always representative of the entire domain surveyed on that day by continuous analyzers. Point sources of CO_2 from power plants that do not emit CO captured by continuous measurements could potentially skew this comparison alongside other non-fossil fuel CO sources, but the small bias between the two suggests that CO_{2ff}' is generally in good agreement with flask CO_{2ff}.
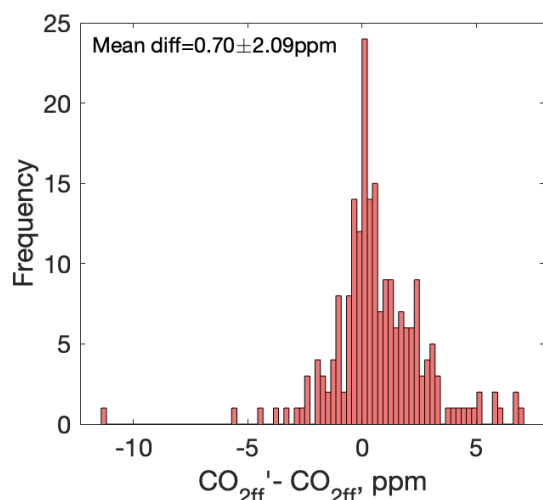
Figure 7. Histogram of in situ (CO$_{2ff}$') minus flask-derived (CO$_{2ff}$) fossil fuel CO$_2$ for total-campaign corresponding measurement times (winter 2017 through summer 2019) in the Mid-Atlantic region. Here, CO$_{2ff}$' is calculated for corresponding flask sample times. By definition of CO$_{2ff}$ and CO$_{2ff}$' above, this comparison indicates ABL sample comparisons only.
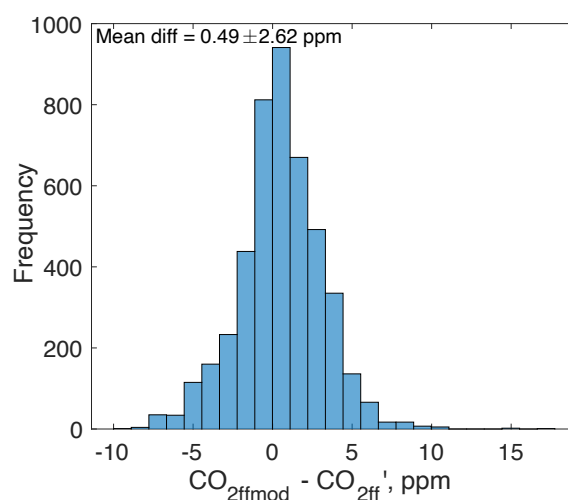
Figure 8. Histogram of PSUWRF CO$_{2ffmod}$ – CO$_{2ff}$' for all flask Δ^{14}CO$_2$ sample days between winter 2017 and summer 2019 in the Mid-Atlantic region.

3.3.2 Comparison of CO$_{2ff}$' and CO$_{2ffmod}$

Figure 8 shows the average difference between CO$_{2ffmod}$ computed along ACT flight tracks and ABL CO$_{2ff}$'. Several research flight days indicate rough agreement between CO$_{2ffmod}$ and CO$_{2ff}$' within a mean of ~0.49±2.6 ppm (1σ), but higher differences outside of average CO$_{2ff}$' uncertainty bounds are seen in Fig. 8 along with a slight skewness. The extent of this positive bias in CO$_{2ffmod}$ is visualized in Fig. 9 for fall 2017, spring 2018 and summer 2019 campaigns, while the winter 2017 campaign model bias in CO$_{2ffmod}$ is slightly negative.

While there are logistical advantages to using this "pseudo-CO$_{2ff}$" method from a Δ^{14}CO$_2$ measurement capacity standpoint, this product has a higher uncertainty (4.96 ppm) associated with large intra-day background variability and assumptions in the value of R$_{CO}$. For this reason, we caution the overgeneralization of the use of CO$_{2ff}$ as a transport tracer with this dataset and have chosen July 24, 2019 as an example of where positively biased CO$_{2ffmod}$ can be evaluated using CO$_{2ff}$'. Here, a series of B-200 flights was conducted downwind of three major eastern U.S. cities (New York City, NY; Philadelphia, PA; and

Washington, DC). ABL flask sampling on each flight leg informed the derivation of CO_{2ff}', and both estimates are in rough agreement as seen in Fig. 10. This individual flight day had higher fossil fuel CO_2 signals and lower variability in background CO measurements, with a lower CO_{2ff}' uncertainty of ~2 ppm relative to the campaign-wide average.
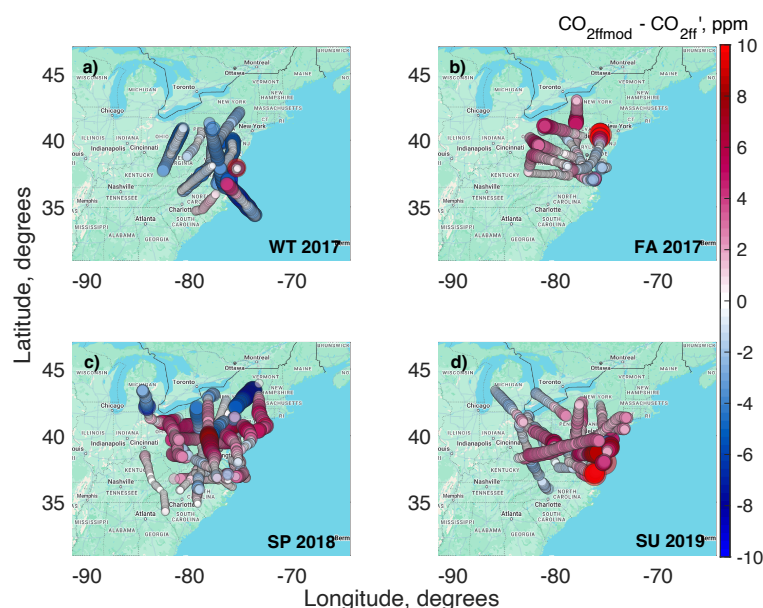
Figure 9. Mapped modeled versus CO_{2ff}' for flask $\Delta^{14}CO_2$ sample flight days for a) winter (WT) 2017, b) fall (FA) 2017, c) spring (SP) 2018 and d) summer (SU) 2019 campaigns. Points are sized by the model-data differences. Plot maps are produced using Google Maps API (©Google Maps).
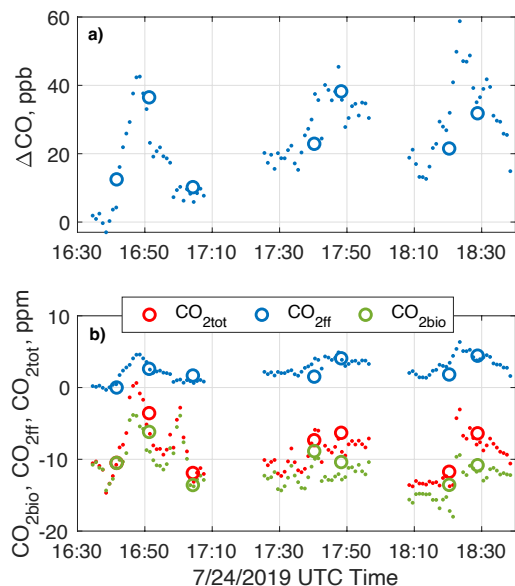
Figure 10. In situ (points) and flask (circles) observations for the B-200 aircraft on July 24, 2019. (a) CO ABL-FT differences (ΔCO) calculated from flasks and in situ continuous measurements. ΔCO from flasks is used to calculate R_{CO} and thus CO_{2ff}' using Eq. 5 using in situ measurements. The average difference in ΔCO (in situ – flask) is 0.6±8.7 ppb. (b) CO_{2bio}, CO_{2ff}', and CO_{2tot} calculated from both in situ high-frequency data (points) and from flask measurements (circles) on July 24, 2019 showing good agreement. Average differences in CO_{2bio}, CO_{2ff}', and CO_{2tot} (in situ – flask) are -1.6 ± 1.2 ppm, 1.2 ± 1.3 ppm and 0.4 ± 1.2ppm, respectively.
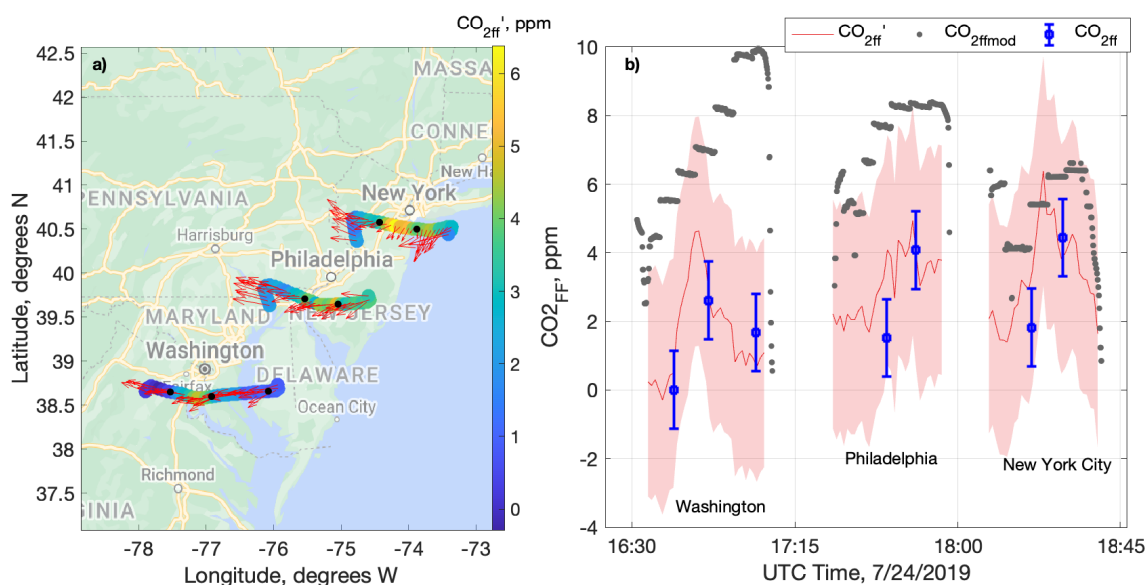
Figure 11. a) Psuedo-CO_{2ff} (CO_{2ff}') calculated along the ACT B-200 ABL flight track on July 24, 2019 (plot produced using Google Maps API, ©Google Maps). Flight tracks are colored by CO_{2ff}' which proceed northerly from Washington, DC to New York City, NY. Black points indicate flask sample locations. Red arrows indicate the wind direction as calculated from aircraft flight data. b) Fossil fuel CO_2 calculated from in situ measurements (CO_{2ff}'), from flask $\Delta^{14}CO_2$ (CO_{2ff}) and from PSUWRF using CT fossil fuel emissions (CO_{2ffmod}). Discretization in CO_{2ffmod} is due to the native 27-km resolution of PSUWRF. Error shading on

CO_{2FF}' is the derived 1σ uncertainty on the CO_{2ff}' calculation for July 24, 2019 and error bars on flask CO_{2ff}
reflect a 1σ uncertainty.

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Figure 11 shows calculated CO_{2ff}' along three ABL flight legs with large positive fossil fuel CO_2 downwind of these urban
areas. Comparing modeled (CO_{2ffmod}), calculated (CO_{2ff}'), and flask-derived (CO_{2ff}) CO_{2ff}, we find that CO_{2ff}' plumes observed
downwind of cities do not always align with CO_{2ffmod} throughout the research flight. Significant differences are observed
between CO_{2ffmod} and CO_{2ff}' magnitudes, particularly during the Washington, D.C. flight leg of up to 60% beyond uncertainty

410 bounds, with the urban emissions plume misaligned (Fig. 11). CO_{2ffmod} is consistently higher than CO_{2ff}' in the Philadelphia
flight leg as well, with general agreement in the plume variability. Finally, we note that the PSUWRF model simulates CO_{2ff}
in the New York City flight leg relatively well with respect to flask CO_{2ff}.

Our current knowledge of CO_2 fossil fuel fluxes is expected to be accurate to within 20% at the regional scale (Gurney et al.,

415 2020). Thus, a 60% overestimation in simulated CO_{2ff} beyond CO_{2ff}' uncertainty bounds around Washington, D.C. cannot be
explained by wind speed or boundary layer depth errors at the time of the flights when compared to in situ meteorological and
trace gas observations: the PSUWRF modeled wind speeds and directions compare well to in situ wind speed and directions
measured on the B-200 to within 1 m s⁻¹ and 10 degrees, respectively. ABL depths derived from B-200 potential temperature
soundings flown at the endpoints of each urban transect in Fig. 11 compare well with ABL depth estimated from PSUWRF

420 potential temperature to within 300 m on average. However, it is possible that the model vertical transport parameterizations
are erroneous earlier in the day, which could produce errors later in the day in CO_{2ff} accumulation and venting relative to that
observed downwind of Washington, D.C., and even Philadelphia, PA. While fossil fuel emissions uncertainties within
PSUWRF can result in observed differences between CO_{2ff}' and CO_{2ffmod}, there is some evidence that model fluxes did not
play a significant role in these discrepancies. PSUWRF simulations between 2017-2018 indicate that independently-run fossil

425 fuel emissions fields with differing spatial and temporal resolution (i.e. ODIAC and Miller fields) result in small CO_{2ffmod}
differences that are within 1.5-2 ppm on average. While these fossil fuel emissions products were not run separately in the
PSUWRF model beyond 2018, a 2 ppm CO_{2ffmod} variability is substantially smaller than that shown in Fig. 11 downwind of
Washington and Philadelphia (3-6 ppm CO_{2ff}), suggesting that model transport, rather than fluxes are erroneous on this
particular flight. More intensive work and new PSUWRF simulations beyond the scope of this case study would be needed to

430 verify this.

While this example does not represent the average CO_{2ffmod} bias with respect to CO_{2ff}' during ACT, it illustrates the utility of
CO_{2ff} as a model transport tracer. Differentiating between regional model transport and flux errors can be improved upon if
derived CO_{2ff}' uncertainties can be reduced with more intentional $\Delta^{14}CO_2$ flask sample verification as has been done in similar

435 studies (Maier et al., 2023). In this example, the high-frequency CO_{2ff}' product is able to more easily highlight potential
modeled errors in the representation of diurnal CO_{2ff} plume variability. As discussed in Miller et al., (2020), deriving CO_{2ff}'

in this way, using information gleaned from a small subset of $\Delta^{14}CO_2$ flask sample measurements, can provide a means for determining the fossil fuel and biogenic components of CO_{2tot} for regional or city-scale studies. Despite higher average uncertainties than flask-based methods, comparisons to models might allow for examination and improvements upon gross
440 model errors for more accurate CO_2 mole fraction estimation and process-based studies.

4 Conclusions

The ACT-America mission enabled a dense, seasonally-diverse dataset of atmospheric $\Delta^{14}CO_2$ measurements from flasks sampled throughout the Mid-Atlantic U.S. and the capability to disaggregate total CO_2 into biogenic and fossil fuel components, which is critical for regional CO_2 source attribution. Seasonal campaigns occurring between 2017 and 2019
445 indicate that biogenic CO_2 exchange in the Mid-Atlantic is the primary driver of CO_2 boundary layer variability in this region as expected, while CO_{2ff} remains relatively constant during the ACT mission. Consistent OCS uptake in all seasons is observed and correlates with negative CO_{2bio} confirming CO_2 uptake by photosynthesis throughout the ACT campaigns. With the broad nature of flask sampling during ACT, the signal to noise ratio in $\Delta^{14}CO_2$ data was low and campaigns investigating strictly urban CO_2 signals (e.g. Plant et al., 2019) can better highlight the utility that routine observations of $\Delta^{14}CO_2$ can provide,
450 including critical information for stakeholders in assessing carbon reduction strategies on regional to sub-regional scales. Several studies have shown the value of incorporating $\Delta^{14}CO_2$ alongside CO_2 measurements as an improved model constraint on regional CO_2 fluxes and future work will require a continued effort to ensure routine $\Delta^{14}CO_2$ measurements throughout the NOAA GGGRN and other networks. Here, we have used a small number of flask $\Delta^{14}CO_2$ samples, taken alongside continuous in situ CO measurements, to provide a high-frequency "pseudo-CO_{2ff}" product using the relationship between in situ CO ABL-
455 FT differences and flask-derived CO_{2ff}. This product is used to evaluate modeled CO_{2ff} at a higher temporal resolution than discrete flasks can provide, and to deduce potential errors in diurnal model transport at the regional scale. Although the ACT dataset is limited to days where $\Delta^{14}CO_2$ signals are higher, this work highlights the value of such a product in future campaigns and measurement networks as a model evaluation tool.

460 **Data Availability**

Observational in situ and flask data from the NASA ACT-America mission used for the analysis in this study are publicly archived at the Oak Ridge National Laboratory repository at https://doi.org/10.3334/ORNLDAAC/1593 (Davis et al., 2018) in addition to PSUWRF output for the ACT-America campaign at https://doi.org/10.3334/ORNLDAAC/1884 (Feng et al., 2021).. Individual flask data with $^{14}CO_2$ measurements from ACT-America can be found at
465 (https://doi.org/10.3334/ORNLDAAC/1575). All data are synced to meteorological and location data onboard each aircraft and available for downloading and merging (https://doi.org/10.3334/ORNLDAAC/1574). GGGRN $\Delta^{14}CO_2$ and CO flask data

used to supplement this study from NWR, NHA, and CMA are available at https://doi.org/10.15138/87ny-6277. The NOAA HYSPLIT model is publicly available via https://www.ready.noaa.gov/HYSPLIT.php for registered and unregistered versions.

470 **Author Contributions**

BCB led the investigation and curated the data alongside JPD, YC, SL and CW. CS, KD, JM, JPD and SL assisted with acquiring funding for this work. BB, JM, and SL conceptualized the analysis and contributed to writing the original draft of this manuscript. SF and TL developed and contributed model analysis and output. All authors contributed to reviewing and editing the manuscript.

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Competing Interests

The authors declare that they have no conflict of interest.

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