

TITLE

Observing system choice can minimize interference of the biosphere in studies of urban CO₂ emissions

AUTHORS

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KEYWORDS

Urban CO₂, Biosphere, Fossil Fuel Emissions, Greenhouse Gas Observations

PLAIN LANGUAGE SUMMARY

11 Cities around the world have announced plans to reduce CO₂ emissions. Atmospheric CO₂
12 observations provide a potential pathway toward independent assessment of implemented policies.
13 However, these measurements can be strongly influenced by the urban biosphere, which can act as
14 both a source (respiration) and sink (photosynthesis) of CO₂. If using an observing approach that
15 introduces a local, urban background – e.g., observations via a downwind airborne transect that
16 captures an entire urban outflow – the relative role of the biosphere may be minimized. Here, we
17 combine back trajectory modeling with high-resolution surface fossil fuel and biosphere CO₂ fluxes
18 across six cities and one powerplant in the NE US to demonstrate that observing strategies using this
19 approach can greatly reduce biosphere interferences in studies of urban CO₂ (<10% biosphere
20 interference outside of summer months, on average) and pave the way to conduct robust studies of
21 urban fossil fuel CO₂ emissions.

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36 **ABSTRACT**

37 Cities around the world have introduced initiatives to reduce CO₂ emissions. Atmospheric
38 observations can provide evaluation and assessment of these initiatives by quantifying emissions,
39 considering local sources and sinks. The relative importance of the urban biosphere, which can act as
40 both a source (respiration) and sink (photosynthesis) of CO₂, has previously been suggested to strongly
41 impact urban CO₂ measurements, confounding the ability to use observations to study fossil emissions.
42 However, if using an observing framework that measures a local urban background and the direct urban
43 core outflow, e.g., along a downwind airborne transect, the biosphere's role may be minimized. Here,
44 we combine real, airborne observations of CO₂ downwind of select cities in the Northeast US with high-
45 resolution, back-trajectory modeling and spatially-and temporally-resolved surface biosphere and fossil
46 fuel fluxes to characterize the relative biosphere importance to urban CO₂ profiles. We show the
47 biosphere influence using this urban observing system to be small, averaging only 15% of the local CO₂
48 enhancement annually, <10% outside of summer, with a maximum influence of 29% in summer when
49 the biosphere drawdown is most pronounced. Furthermore, when considering two biosphere models
50 that differ by >80% , the impact on observed urban CO₂ signals is reduced to only 12% on average.
51 Urban observing frameworks that utilize this local background approach – including those via aircraft or
52 satellite observations – can minimize the biosphere's influence and thus help facilitate robust
53 assessments of urban fossil fuel CO₂ emissions.

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60 **INTRODUCTION**

61 Cities across the world have announced initiatives to reduce anthropogenic greenhouse gas
62 (GHG) emissions to combat climate change [1, 2]. CO₂ has been identified as the main GHG of concern as
63 it is the primary GHG emitted through human activities and is the single largest driver of our changing
64 climate. In the United States, urban areas contribute ~40% of nationwide CO₂ emissions [3, 4], so
65 studying and assessing CO₂ profiles in urban areas, particularly their sources and sinks, can be critical to
66 address climate concerns. Specifically, it can: 1) help inform climate policy by providing direct
67 information on emissions and assessing whether implemented policies are having desired effects and 2)
68 support fundamental carbon cycle research by constraining human inputs into the atmosphere to
69 disentangle changes driven by anthropogenic emissions or biosphere dynamics.

70 The biosphere acts both as a source (via respiration) and a sink (via photosynthesis) of CO₂, and
71 because of its potential significance to understanding urban CO₂, a number of previous studies have
72 emphasized the biosphere and the role it plays. Previous representations of urban domains, including
73 via isotopic measurements [5] and tower observations coupled with emission inventories [6] have
74 suggested the biosphere is a substantial contributor to observed CO₂ in urban domains [7-9]. These
75 results have particular pertinence for the expanding networks of urban tower observations in the US,
76 including Los Angeles [10, 11], the Northeast Corridor [12], Indianapolis [13], Boston [6], and Salt Lake
77 City [14]. However, if an urban area is characterized with an observing strategy where, at the city-scale,
78 the urban signal can be isolated using measurements of the rural-suburban-urban boundary to define a
79 local background (e.g., via an airborne platform that transects the full rural-suburban-urban outflow, or
80 a similar observing strategy based on satellite-based remote sensing), the influence of the biosphere and
81 the importance of how well it needs to be represented may not be as significant.

82 Here, we present a framework that utilizes real, airborne observations in model space to
83 characterize the relative biosphere contribution to urban CO₂ enhancements. We apply this framework

84 to select cities and a large point source in the Northeast U.S. over the course of a year to understand
85 seasonal variation in biosphere contributions to CO₂ enhancements when assessed with a local
86 background observing strategy for both whole city and large point source emission profiles. We use
87 multiple biosphere representations to evaluate how different methods to estimate biosphere fluxes
88 influences its relative significance to urban CO₂ observed with a local background framework. In
89 addition, we compare the biosphere impacts estimated from this approach to what the biosphere
90 contribution would be from an idealized-located tower downwind of one of the cities analyzed here.
91 Lastly, we assess surface fluxes of fossil fuel and biosphere models/inventories within each of the urban
92 cores studied here as a first-order, bottom-up comparison of the biosphere influence to urban CO₂
93 profiles.

94 **METHODS**

95 To ultimately understand the relative influence of the biosphere to an urban CO₂ enhancement
96 observed with a local background observing framework approach (i.e., with an observing strategy
97 where, at the city-scale, the urban signal can be isolated using measurements of the rural-suburban-
98 urban-suburban-rural boundary to define a local background; e.g., via downwind airborne transects that
99 capture the urban outflow), we use previously identified CO₂ outflows downwind of urban areas that
100 were part of the East Coast Outflow 2018 (ECO-2018) campaign (New York City, NY; Boston, MA.;
101 Philadelphia, PA.; Washington DC; Baltimore, MD.; Providence, RI.) [15]. The focus of this work is on
102 urban CO₂ profiles, but we additionally apply this framework to one, large point source (Lake Road
103 Generating Power Plant, VT.) to evaluate this approach's viability for point source assessments. These
104 airborne transects (Table 1; SI Fig. 1), which were designed specifically to target urban emissions, were
105 used as a basis for the modeling effort described next.

Table 1: The locations, dates, and times of CO₂ outflows identified as part of the ECO-2018 flight campaign. Observations along each of these flight paths were used in model space to assess the relative biosphere influence to urban/large point source CO₂ enhancements.

Model-Space Domain	Flight Date	Local Flight Time	Shorthand ID
Baltimore, MD	20-Apr-18	12:23-12:44	BALT-0420
Baltimore, MD	23-Apr-18	15:20-15:48	BALT-0423
Baltimore, MD	26-Apr-18	12:07-12:36	BALT-0426
Boston, MA	9-Apr-18	13:35-14:01	BOS-0409
Boston, MA	11-Apr-18	15:22-15:47	BOS-0411
Washington DC	14-Apr-18	15:39-16:08	DC-0414
Washington DC	20-Apr-18	12:07-12:25	DC-0420
New York City, NY	13-Apr-18	16:16-16:50	NYC-0413
New York City, NY	22-Apr-18	14:11-15:09	NYC-0422
Philadelphia, PA	20-Apr-18	13:37-14:17	PHIL-0420
Philadelphia, PA	23-Apr-18	14:00-14:33	PHIL-0423
Philadelphia, PA	1-May-18	12:09-12:37	PHIL-0501
Providence, RI	9-Apr-18	13:55-14:15	PVD-0409
Providence, RI	21-Apr-18	14:24-14:36	PVD-0421
Lake Road Generating Power Plant, VT	11-Apr-18	16:18-16:22	LRGPP-0411

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108 Stochastic Time-Inverted Lagrangian Transport (STILT) modeling framework and surface flux
 109 models/inventories for CO₂ source impact estimation

110 To translate the signals observed by the aircraft to fluxes on the surface, we use observation
 111 locations that were 10 seconds apart along the ECO-2018 transects as receptors into the Stochastic
 112 Time-Inverted Lagrangian Transport Model's R interface (STILT-R version 2 [16, 17]; herewith STILT).
 113 Each STILT footprint was calculated at a 0.02° x 0.02° surface resolution, using meteorological data from
 114 the High-Resolution Rapid Refresh (HRRR) model (hourly, 3km resolution), releasing 400 particle
 115 ensembles, and simulations were extended six-hours back in time (see SI Section 1 for justification of
 116 using six hours to represent a near-field urban domain and a case-study demonstrating its validity).

117 STILT outputs hourly surface influence footprints ($\frac{ppm}{\mu mol/m^2-sec}$) that when used in conjunction
 118 with hourly CO₂ surface fluxes ($\frac{\mu mol}{m^2-sec}$), e.g., from fossil fuel combustion or the biosphere, can estimate
 119 fossil fuel and biosphere CO₂ source impacts (*ppm*), respectively. Here, we convolve the footprints with

120 one fossil fuel inventory (Anthropogenic Carbon Emissions Systems (ACES) [18]) and two biosphere
 121 models (Vegetation Photosynthesis Respiration Model (VPRM) [8, 19] and Solar-Induced Fluorescence
 122 for Modeling Urban biogenic Fluxes (SMUrF) [20]). Briefly, the ACES inventory provides fine-scale (1km x
 123 1km), hourly fossil fuel CO₂ emissions for all major emitting sectors (e.g., power plants, on-road,
 124 commercial, residential, etc.) for 13 Northeast US states, and it has been demonstrated to be well-
 125 constrained in this region [21]. VPRM estimates ecosystem respiration and gross ecosystem exchange
 126 (GEE) at a 0.02°x0.02°, hourly resolution using a combination of Moderate Resolution Imaging
 127 Spectroradiometer (MODIS) inputs, meteorological variables (e.g., photosynthetically active radiation
 128 and temperature), and various optimized parameters. SMUrF utilizes Solar-Induced Fluorescence (SIF)
 129 as a proxy for photosynthesis and air, soil temperatures, and SIF-driven Gross Primary Production (GPP)
 130 to estimate ecosystem respiration (R_{eco}). ACES (1 km x 1 km) and SMUrF (0.05°x0.05°) were available at
 131 different resolutions than our STILT-generated footprints (0.02°x0.02°); for convolutions, we re-gridded
 132 ACES to be the same resolution and picked the nearest grid for convolutions with SMUrF.

133 We used two biosphere representations (VPRM and SMUrF), which use two fundamentally
 134 different approaches to represent the biosphere, to investigate the importance of its representation to
 135 the biosphere influence in studies of urban CO₂ outflows that employ a local background observing
 136 framework. The percent difference between the biosphere representations averaged within each urban
 137 core and aligned with the 6-hour back-trajectory from STILT was determined as:

$$\left\{ \begin{array}{l}
 \text{Percent Difference (\%)} = \frac{|VPRM - SMUrF|}{(VPRM + SMUrF)/2} * 100\%, \quad VPRM \text{ and } SMUrF \text{ same direction} \\
 \text{Percent Difference (\%)} = \frac{|VPRM - SMUrF|}{\max(|VPRM|, |SMUrF|)} * 100\%, \quad VPRM \text{ and } SMUrF \text{ opposing direction}
 \end{array} \right. \quad (\text{Eqn. 1})$$

139 We use a different approach to quantify percent error if the biosphere representations are
 140 showing opposite directions (e.g., VPRM net uptake and SMUrF net respiration) to best minimize
 141 variability/error when crossing the zero flux threshold (see SI Table 1 for a comparison of percent

142 differences between the two approaches across all urban study areas for all months). Since these two
143 biosphere representations are drastically different, they provide an excellent proxy for evaluating how
144 impactful the biosphere may be on urban fossil CO₂ studies within this framework – if these two
145 representations have minimal impact on the fossil evaluation, it indicates very low sensitivity of this
146 observing framework to the biosphere.

147 Quantifying daily fossil fuel and biosphere CO₂ contributions to urban outflows observed with a local
148 background framework

149 For each of the model-space study domains, we perform the convolution method described
150 above for modeled fluxes for each day of an entire calendar year. This enables us to evaluate how a
151 flight with winds as observed during ECO-2018 that capture an urban CO₂ enhancement would sample
152 fossil and biosphere fluxes from each urban and one point source domain over all seasons. The
153 footprints from STILT were the same for each day to again ensure model-space would represent a real,
154 observed urban CO₂ enhancement; we change the fossil fuel and biosphere fluxes aligned to each day.
155 To then quantify the relative influence of the fossil and biosphere activity to a simulated urban CO₂
156 outflow, we used an integrated sum approach that estimates the fossil fuel and combined (i.e., fossil
157 fuel plus biosphere) enhancement relative to simulated CO₂ levels outside of the urban core along the
158 tails of the enhancement (Fig. 1). This approach allows us to define a local background relative of an
159 urban enhancement, broadly consistent with the approach of defining urban outlined by Wu et al. [20,
160 22]. This stands in contrast to many urban study approaches using upwind tower observations, which
161 often do not isolate the local city in the same manner, leading to the urban tower having significant
162 sensitivity to fluxes outside of the local urban domain.

163 Here, the urban-defined region for each city was characterized by combining: (1) satellite
164 imagery, (2) STILT-derived back trajectories for each model-space domain (i.e., along the simulated
165 airborne transect, a CO₂ enhancement downwind of the urban domain could be identified against an

166 urban-background CO₂ level on the tails, outside of the primary urban domain), and (3) local
167 government/commission-defined urban boundaries (SI Fig. 1 for urban area designation in each of the
168 study domains). The urban fossil fuel CO₂ enhancement is estimated as the sum of the fossil fuel
169 enhancement relative to the simulated fossil fuel CO₂ at the tails of the enhancement (black-shaded
170 area in Fig. 1; SI Fig. 2 for simulated CO₂ source contributions for each flight-path on the day of the
171 flight). To account for uneven enhancements on either side of the urban-defined domain – which can be
172 attributed to unequal fluxes due to land-use variation – we use a straight line between the tails as the
173 local background concentration (the dashed, angled lines in Fig. 1). This value relative to the sum of the
174 combined (fossil fuel + biosphere) enhancement (red-shaded area in Fig. 1) within the urban-defined
175 domain can then be used to quantify the relative biosphere and fossil fuel influence to an urban outflow
176 when observed via airborne sampling as:

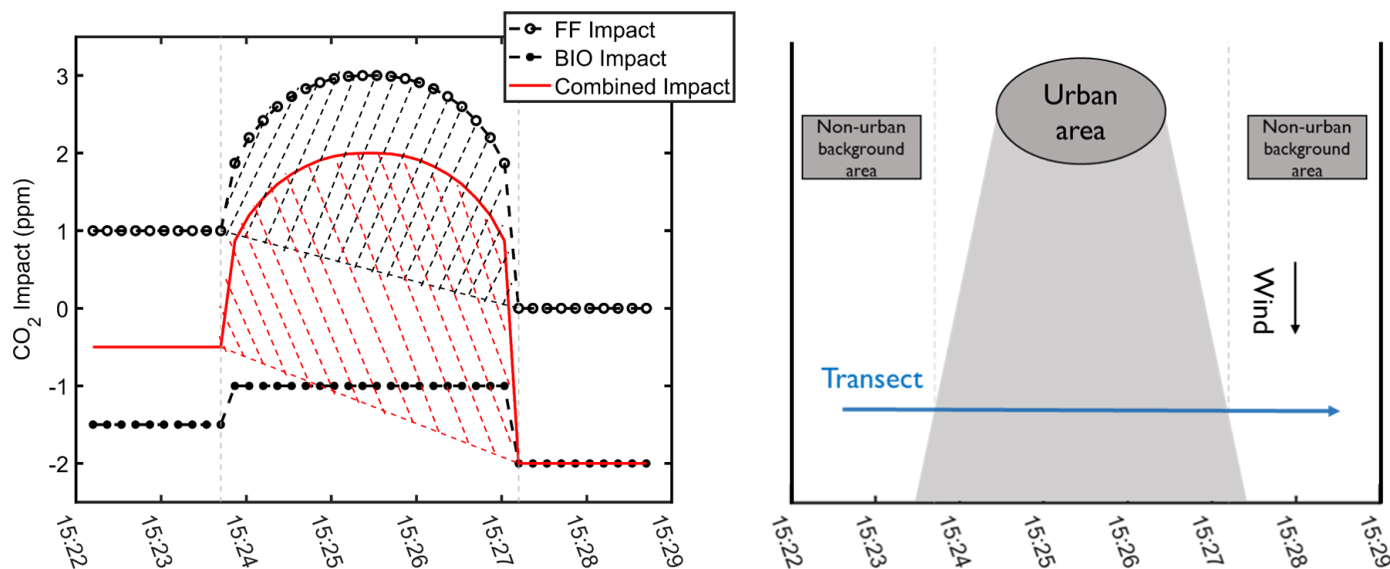
$$177 \quad \text{Biosphere contribution (\%)} = 1 - \frac{\text{urban fossil fuel CO}_2 \text{ contribution}}{\text{combined (fossil fuel+biosphere) urban CO}_2 \text{ contribution}} \quad (\text{Eqn. 2})$$

178 where the fossil fuel contribution can then be determined as:

$$179 \quad \text{Fossil Fuel contribution (\%)} = 1 - \text{Biosphere contribution (\%)} \quad (\text{Eqn. 3})$$

180 These straightforward calculations enable us to discern the relative impact of the biosphere and
181 fossil emissions for each simulated airborne sample. These percentages directly translate into the
182 relative importance in any follow-on analysis focused on quantifying fossil emissions. If the biosphere
183 contributes to 10% of the total observed signal (urban-integrated combined sum), then ignoring the
184 biosphere would only contributed to a 10% error in a fossil analysis. In contrast, if the biosphere
185 represented 50% of the signal, neglecting it would directly translate to a 50% error.

186 Fig. 1: Example urban CO₂ enhancement along a downwind, airborne transect showing simplified,
 187 sample fossil fuel and biosphere contributions. (Left) The bold, red line is the combined (fossil fuel +
 188 biosphere) CO₂ enhancement. Vertical grey lines indicate the location of the urban-defined region. The
 189 shaded black area is the urban-integrated fossil fuel sum while the shaded red area is the urban-
 190 integrated combined sum. (Right) A sample visual of an airborne transect that could produce this CO₂
 191 enhancement



192
 193 Comparing airborne-observation simulated and idealized-located tower biosphere CO₂ contributions

194 Using the NYC-0413 transect (the only transect assessed here whose STILT-derived, 72-hour,
 195 average back-trajectory stayed in the Continental U.S.; SI Fig. 3), we treat the receptor with the highest
 196 simulated fossil fuel contribution as the idealized location of a CO₂ observing tower downwind of the
 197 urban core. This is considered the idealized location to assess an urban fossil fuel CO₂ impact as this
 198 location captures the maximum CO₂ enhancement (and presumably fossil emission signature) from the
 199 urban core (of receptors along the transect). We follow the same source-impact estimation approach
 200 using STILT described above, but instead run STILT at 0.2°x0.2° resolution, 72-hours back in time, and
 201 use VULCAN Version 3.0 fossil fuel emission inventory [23] outside of the near-field NYC domain (the
 202 ACES inventory does not include emission estimates outside of the NE US; see SI Fig. 3 for spatial
 203 domains). In the near-field NYC domain, we utilize the same CO₂ source impact results when STILT was
 204 run at a 0.02°x0.02° resolution and used ACES as described above. Conserving source contributions in

205 the near-field domain allows us to make a more direct comparison between an idealized-located tower-
206 observed CO₂ source contribution and one generated from an observed outflow with a local background
207 (i.e., via a downwind airborne transect). The fossil fuel and biosphere CO₂ impacts as assessed with the
208 idealized-located tower were compared against a 6-hour (the same time duration as the back trajectory
209 model runs) CO₂ source impact at the same receptor and with the source contributions integrated along
210 the entire urban outflow.

211 Comparing fossil fuel and biosphere CO₂ contributions from the local background observing framework
212 approach with surface fluxes within each city

213 We assess surface fluxes for fossil fuel inventories (ACES and VULCAN) and biosphere models
214 (VPRM and SMUrF) for an entire year, 2018, within each of the six urban cores in this study. For each
215 city, we determine the sum of the flux fields within each city's urban core (SI Fig. 1) and calculate daily
216 and 6-hour summations aligned with the same hour of day as the airborne observations (which
217 generally represented the maximum biosphere drawdown given that most flights occurred between
218 noon and 4:00 pm local time; Table 1). We then compare a "bottom-up" emission inventory fraction of
219 urban CO₂ from fossil fuels following the approach outlined in Eqns. 2 and 3 (only using ACES and the 6-
220 hour period aligned with the back trajectories of the various airborne transect times) with the findings
221 from the local background observing framework approach (defined as "top-down") within each urban
222 domain.

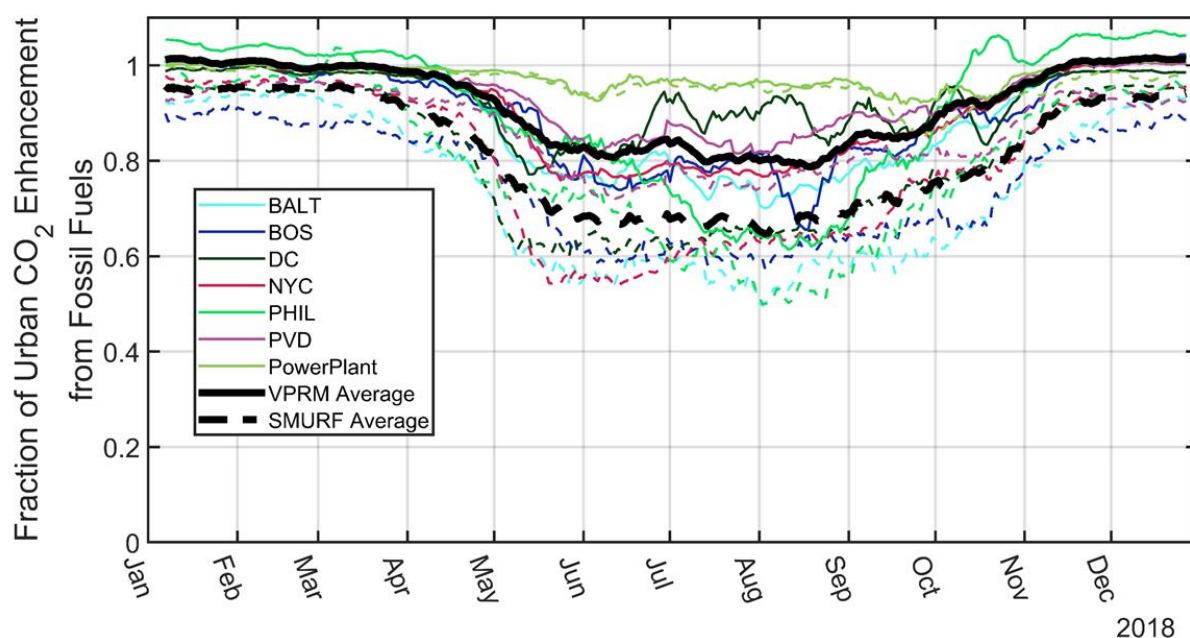
223 **RESULTS**

224 Annual simulated fossil fuel and biosphere CO₂ source contributions to urban outflows observed with a
225 local background approach

226 Across the seven study sites (6 cities and 1 point source) and 15 unique flight paths assessed
227 here, we find the fossil fuel CO₂ contribution always exceeds the biosphere contribution – independent
228 of biosphere model used – when the observing technique employs a local background observing
229 approach, i.e., via a downwind airborne transect. Outside of the summer months when a photosynthesis

230 signal is present, the biosphere contribution is generally negligible in the modeling study presented here
 231 (Fig. 2), suggesting observing frameworks that use this local background approach under these
 232 conditions can generally ignore the biosphere and the role it plays to urban CO₂ profiles. During the
 233 summer months, the average biosphere contribution was more pronounced, with the largest monthly
 234 contribution across cities occurring in August when the average modeled biosphere contribution was
 235 26% (range across sites: 15-36) using VPRM and 39% (23-47) using SMURF, respectively (Fig. 2 & SI Table
 236 2). SMURF showed consistently higher biosphere interference compared to VPRM throughout the year,
 237 consistent with it having a stronger urban signal in each of the study cities (Table 2). The local
 238 background observing framework approach assessed here directly compares influences within an urban
 239 core to its surrounding suburban/rural-background, and SMURF showed a larger gradient between the
 240 urban core and surrounding background throughout the year across cities (SI Fig. 4). With this observing
 241 framework, it is the magnitude of the urban-suburban-rural gradient in fluxes that drives the measured

Fig. 2: Simulated fraction of urban CO₂ from fossil fuels for each study domain (6 cities and 1 powerplant) over the course of a year using a local background observing framework (i.e., downwind airborne transect). All data is plotted as a rolling 14-day average. The solid lines represent simulations with VPRM as the biosphere representation while the dotted lines represent simulations with SMURF. City-wide data shown here in the average of all transects in each city; see SI Fig. 5 for transect specific results within each of the study cities.



242 signal; even if there is a reasonably strong biosphere flux, it only impacts this observing strategy if there
243 is a gradient in that flux from the urban core to the suburban-rural zone. The presence of a very large
244 gradient in fossil emissions over this spatial domain creates the large urban signal, and the absence of a
245 gradient of similar magnitude in the biosphere minimizes its impact. The 7-day rolling-average of all
246 sites' fossil fuel contribution was slightly above one at the beginning of January and the end of
247 December, which can be explained by a stronger relative respiration enhancement in urban-background
248 CO₂ relative to urban levels, enhancing the observed urban CO₂ signal.

249 Large variability existed between VPRM and SMUrF within each of the six urban cores studied
250 here; however, when assessing CO₂ source contributions using a local background observing framework,
251 this variability/error was minimized (Table 2 and SI Fig. 4). The highest, monthly difference between
252 average urban, 6-hour fluxes (aligned with the back-trajectory modeling time period) between VPRM
253 and SMUrF averaged across the city study domains was 136% (102-180) in September where VPRM
254 predicted an average biosphere flux of 668 (-517-1,464) nmol m⁻² s⁻¹ across cities while SMUrF modeled
255 a flux of -524 (-1,837-106) nmol m⁻² s⁻¹ averaged across cities, where the reported fluxes are aligned with
256 the same 6-hour interval aligned with the back-trajectory modeling. The percent difference between
257 SMUrF and VPRM to the observed urban CO₂ enhancement when using the local background approach,
258 on the other hand, was 15% (9.2-19). Across the summer month (June-Aug.), when the average
259 biosphere contribution to urban CO₂ profiles is highest, and the average difference to the observed
260 urban CO₂ enhancement when using the local background approach between the two biosphere
261 representations was 16% (34% SMUrF and 18% VPRM averaged across cities), the percent error
262 between biosphere models was 60% (32-116).

263

264

265 Table 2: Monthly comparison of biosphere fluxes (VPRM and SMUrF) across study cities that
 266 demonstrates large errors between biosphere models are minimized when assessing urban CO₂
 267 emissions with a local background observing framework (LBOF; i.e., downwind airborne transect). The
 268 monthly average fluxes shown here correspond to the same 6-hour interval during the day that the back
 269 trajectory-modeling LBOF captures.

270

		Average Urban Surface Flux (nmol m ⁻² s ⁻¹)		Percent Difference Between VPRM and SMUrF Urban Surface Fluxes (%)	Percent Difference Between VPRM and SMUrF Contribution with LBOF (%)
		VPRM	SMUrF		
JAN	BALT	772	6	197	8
	BOS	627	102	144	11
	DC	699	301	80	4
	NYC	422	380	11	4
	PHIL	709	266	91	7
	PVD	363	326	11	6
	All City Average	599	230	89	7
FEB	BALT	980	237	122	7
	BOS	831	206	121	10
	DC	1058	583	58	4
	NYC	592	526	12	3
	PHIL	861	445	64	6
	PVD	516	486	2	4
	All City Average	806	414	63	6
MAR	BALT	817	-87	111	10
	BOS	547	170	105	12
	DC	783	186	123	4
	NYC	458	344	28	4
	PHIL	751	168	127	4
	PVD	402	303	28	4
	All City Average	626	181	87	6
APR	BALT	634	-366	158	15
	BOS	546	-254	146	10
	DC	606	-9	101	10
	NYC	444	240	60	7
	PHIL	530	-83	116	15
	PVD	401	168	82	5

	<i>All City Average</i>	527	-51	111	11
MAY	BALT	-1195	-2897	83	20
	BOS	-1984	-2669	29	16
	DC	-1089	-1660	42	20
	NYC	-443	-865	65	21
	PHIL	-653	-1442	75	11
	PVD	-1058	-463	78	9
	<i>All City Average</i>	-1070	-1666	62	16
JUNE	BALT	-2148	-4710	75	21
	BOS	-4259	-5853	32	15
	DC	-1646	-2814	52	21
	NYC	-1487	-1902	25	22
	PHIL	-1874	-3059	48	11
	PVD	-2640	-1283	69	9
	<i>All City Average</i>	-2342	-3270	50	17
JUL	BALT	-1864	-3427	59	19
	BOS	-3354	-4268	24	19
	DC	-1349	-1957	37	26
	NYC	-995	-1667	50	16
	PHIL	-1463	-2474	51	8
	PVD	-2338	-685	109	8
	<i>All City Average</i>	-1894	-2413	55	16
AUG	BALT	-1082	-2360	74	18
	BOS	-1552	-2691	54	13
	DC	-1122	-1048	7	24
	NYC	-504	-726	36	15
	PHIL	-342	-1293	116	12
	PVD	-993	-82	169	8
	<i>All City Average</i>	-933	-1367	76	15
SEP	BALT	1219	-772	163	19
	BOS	-517	-1837	112	18
	DC	1250	-21	102	15
	NYC	723	-255	135	14
	PHIL	1464	-364	125	13
	PVD	-133	106	180	9
	<i>All City Average</i>	668	-524	136	15
OCT	BALT	1173	-534	146	21

	BOS	593	-812	173	20
	DC	1122	-7	101	13
	NYC	674	40	178	13
	PHIL	766	-111	115	21
	PVD	381	224	52	9
	All City Average	785	-200	127	16
NOV	BALT	1049	-31	103	15
	BOS	918	128	151	14
	DC	973	279	111	5
	NYC	683	415	49	7
	PHIL	940	207	128	12
	PVD	678	349	64	6
	All City Average	874	225	101	10
DEC	BALT	956	549	54	8
	BOS	759	495	42	13
	DC	862	759	13	3
	NYC	577	692	18	5
	PHIL	846	709	18	11
	PVD	513	561	9	8
	All City Average	752	627	26	8

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273 Comparing fossil fuel and biosphere CO₂ source contributions between idealized-located tower and local
 274 background observing framework

275 CO₂ source contributions to the idealized-located tower showed a large relative biosphere
 276 contribution throughout the year (respiration signal in winter and drawdown in summer months)
 277 compared to fossil fuel contributions (Fig. 3a), consistent with the findings from previous tower-network
 278 studies of urban CO₂ that suggest the biosphere is an important parameter when assessing urban CO₂ [5,
 279 6]. However, when utilizing a local background observing framework, the biosphere impact, although
 280 still present, is considerably less pronounced (Fig. 3b for 6-hour back from the idealized-located tower
 281 and Fig. 3c for the source contribution along the entire NYC-0413 transect). During peak biosphere CO₂
 282 contributions (June 4-10) of the 72-hour back trajectories to the idealized-located tower along the NYC-
 283 0413 downwind flight path, we find the biosphere contribution from VPRM and SMUrF exceeds the

284 fossil contribution, with 7-day rolling average contributions of 59.1% and 73.2% to the urban CO₂
 285 enhancement, respectively. During this same period, using the local background observing framework
 286 approach, the biosphere represents just 9.8% and 28.6% of the simulated CO₂ profile, respectively (Fig.
 287 3c). These findings further suggest that while the biosphere is important to characterize in certain
 288 observing frameworks, when urban outflows are measured with a local background, as facilitated by
 289 airborne or satellite-based observations, the biosphere is less important to characterize and errors in its
 290 representation will not compound in studies of urban fossil fuel CO₂.

291

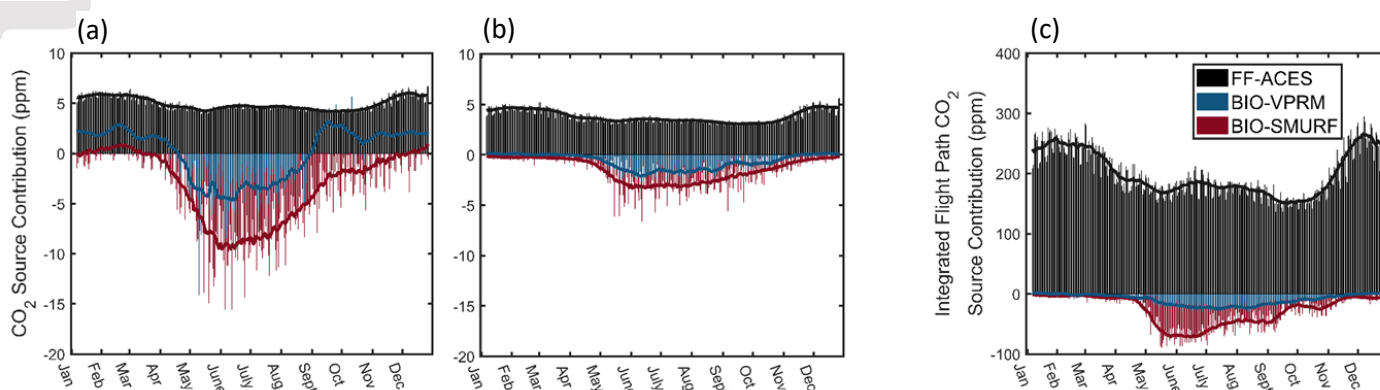


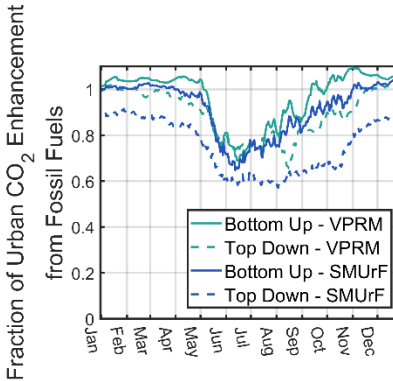
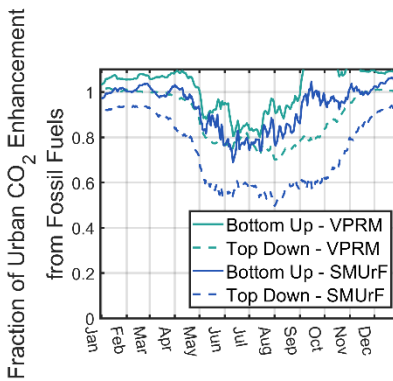
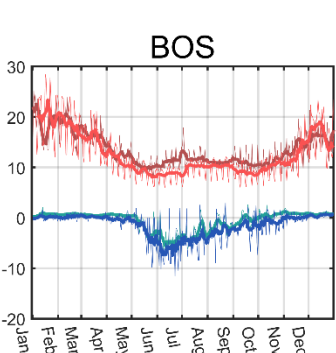
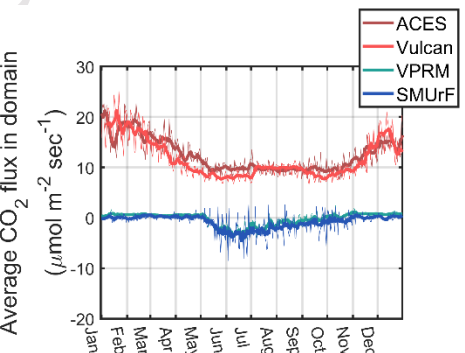
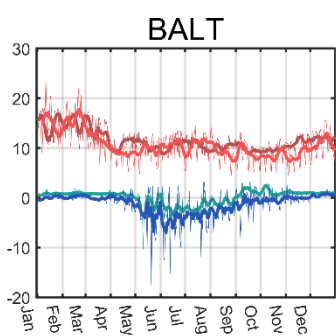
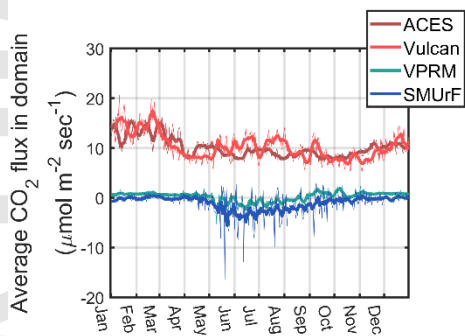
Fig. 3: Comparison of fossil fuel and biosphere CO₂ contributions to: (a) Idealized-located tower convolved 72 hours back in time; (b) The same idealized-located tower convolved six hours back in time (the same duration of the back trajectory modeling convolution); and (c) Along the entire urban outflow for the NYC-0413 transect. Seven-day rolling averages for both fossil fuel and biosphere source contributions are indicated by the bold lines.

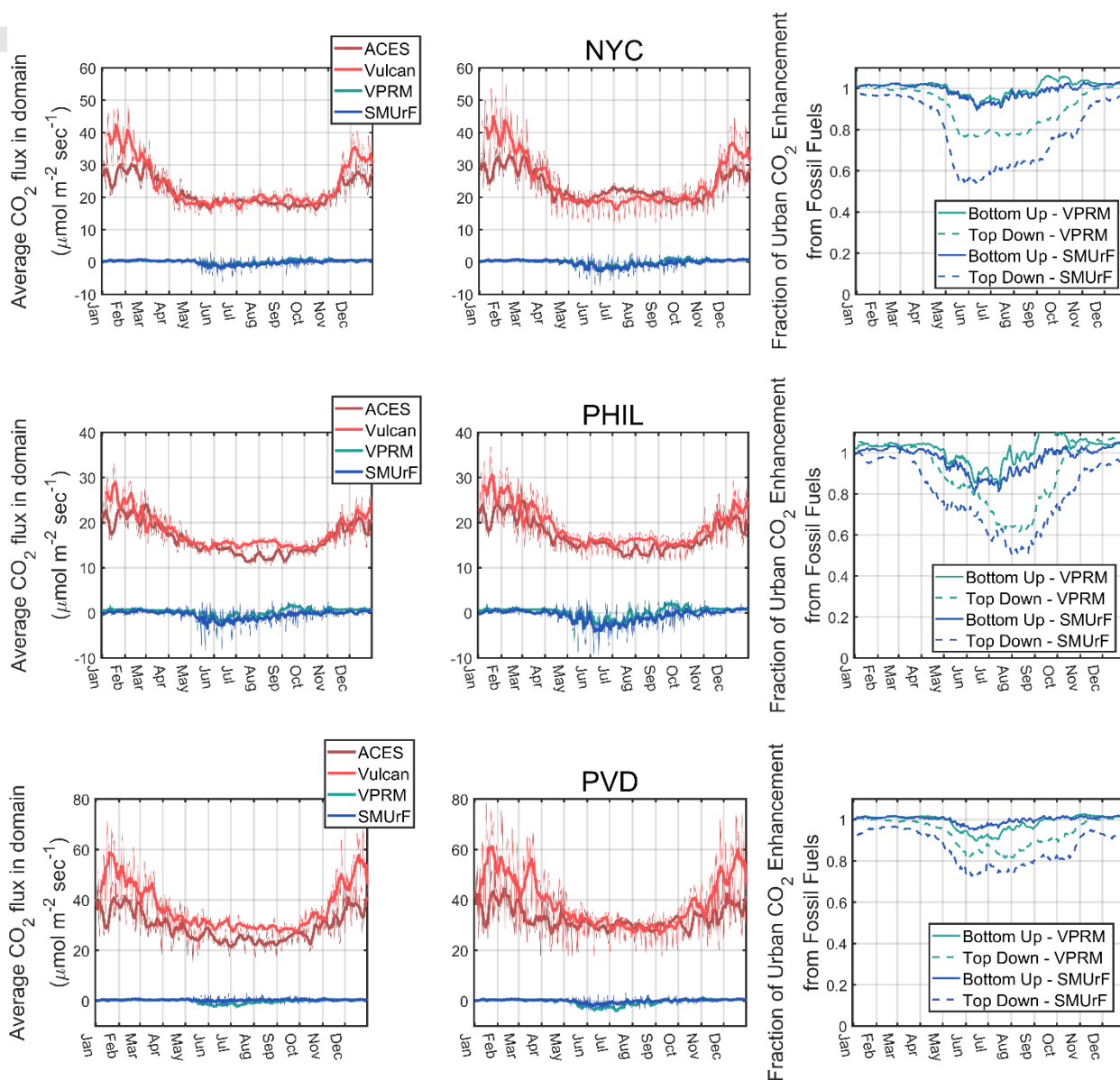
292

293 Bottom-up fossil fuel and biosphere CO₂ surface flux comparison with top-down local background
 294 approach to assess CO₂ source contributions

295 For each of the six cities assessed here, fossil fuel CO₂ emissions have higher magnitudes than
 296 CO₂ signals from the biosphere throughout the year (Fig. 4, left column), even during afternoon periods
 297 (6-hour period aligned with downwind airborne observation back-trajectory modeling time period)
 298 when the biosphere drawdown would be more pronounced (Fig. 4, middle column). When comparing

299 bottom-up fluxes against top-down local background CO₂ source contributions, we find that across
300 cities, the bottom-up approach generally showed a lower biosphere influence than the top-down
301 approach (Fig. 4, right column & Fig. 5) for both biosphere representations. This is attributed to the
302 difference in what each approach is directly assessing: the emission inventory approach only assesses
303 direct urban fluxes while the flight-path approach is geared toward assessing enhancements relative to a
304 local background – it's capturing the biosphere gradient along the urban-suburban-rural transect. The
305 biosphere gradient between the urban core and its surrounding (SI Fig. 4) results in the increased
306 relative contribution of the biosphere to a simulated urban CO₂ enhancement. We further find that
307 SMUrF generally resulted in a larger biosphere contribution compared to VPRM with both the bottom-
308 up and top-down approaches, consistent with it having a stronger signal within each of the urban cores
309 assessed here and a steeper urban-suburban-rural gradient, respectively (Table 2).

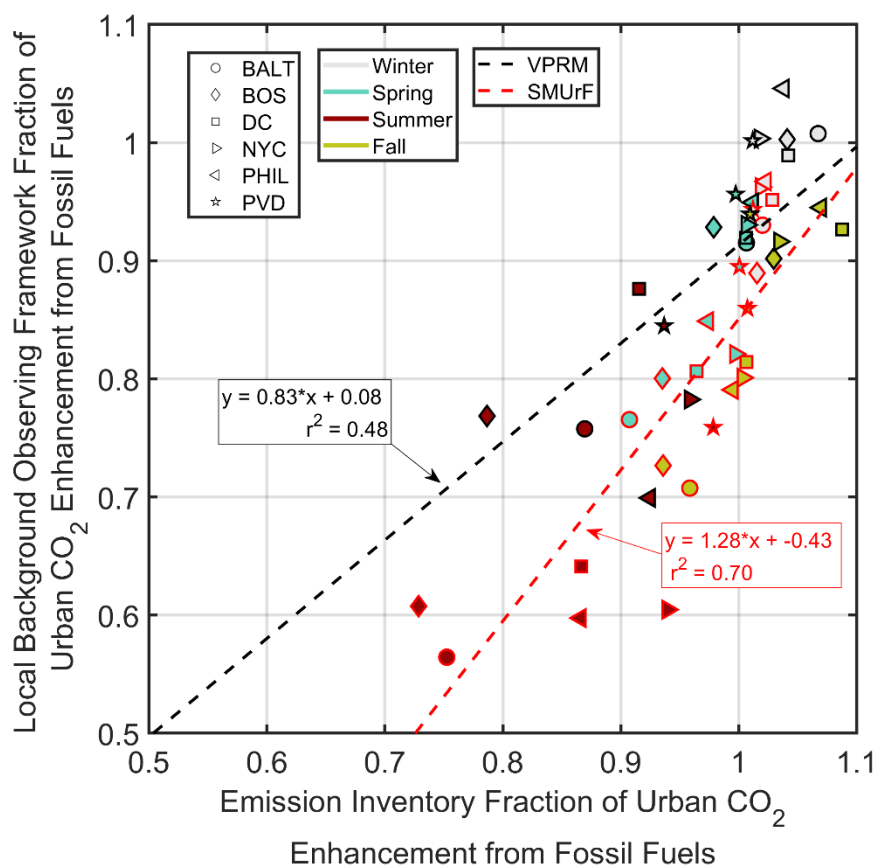




314 Fig. 4: Comparison of fossil fuel and biosphere surface fluxes ($\mu\text{mol m}^{-2} \text{s}^{-1}$) across the six study cities for
 315 (left column) daily-average within the domain and (middle column) 6-hour average (aligned with back-
 316 trajectory modeling time domain) within the domain. (Right column) Comparison of the fraction of
 317 urban CO_2 from fossil fuels between bottom-up, emission inventory and top-down, local background
 318 observing framework (i.e., downwind airborne transect).

319

320



321 Fig. 5: Seasonal comparison of bottom-up emission inventory and top-down local background observing
 322 framework approach (i.e., downwind airborne transect) approach to quantify the fraction of urban CO₂
 323 from fossil fuels across the six study cities and two biosphere representations (VPRM and SMUrF).

324

325 **DISCUSSION, LIMITATIONS, & CONCLUSIONS**

326 Cities in the United States have spearheaded nationwide efforts for deep de-carbonization
 327 efforts, including the introduction of ambitious CO₂ emission reduction pathways. Observation-based
 328 evaluations of fossil fuel CO₂ emissions can be influenced by the biosphere, which can act as both a
 329 source (respiration) and sink (photosynthesis) of CO₂; thus, understanding and characterizing the role of
 330 the biosphere to studies of urban CO₂ can help constrain fossil emissions in cities. A frequently used
 331 approach to characterize urban CO₂ is to analyze tower (or tower-network) observations, and although
 332 this approach can provide a long temporal record, the spatial sensitivity of these stationary sensing

333 platforms are dictated by meteorological conditions and inversion approaches that often capture air
334 masses that travel distances well upwind of a direct urban core, potentially inflating the role the
335 biosphere plays to urban CO₂ profiles. For example, a previous tower-based study in Boston, MA,
336 concluded that biosphere activity offset 100%, 58%, and 20% of the afternoon anthropogenic urban CO₂
337 enhancement in July, September, and October, respectively [6]. In addition, using a network of towers in
338 Indianapolis, IN. as part of the INFLUX campaign, Turnbull et al. [13] showed the biosphere contributes
339 >50% of urban CO₂ signals in summer months, while Miller et al. [5], also using multiple towers in Los
340 Angeles, CA., concluded that the urban biosphere is 33% of the annual mean fossil fuel contribution – a
341 surprisingly large contribution for a dry, metropolitan area.

342 In this study, we estimate the influence of the biosphere to cities' CO₂ profiles using a back-
343 trajectory modeling approach that employs a local background observing framework, consistent with
344 what would occur if a city's CO₂ outflow was measured via airborne or satellite-based observations. We
345 find that studies of urban CO₂ that employ such a local background approach substantially minimizes
346 potential biosphere interference to urban CO₂ profiles – across study sites here, the biosphere
347 contribution to the total CO₂ signal was 15% across the year (<10% in non-summer months), so in
348 studies conducted outside of summer months that use this observing framework, it can be concluded
349 that errors in fossil fuel emission estimates will be <10% due to biosphere interference, suggesting this
350 approach can be used to conduct robust assessments of urban fossil fuel CO₂ emissions. Similar space-
351 based observation strategies that capture local background will have similar success. For ground-based
352 observations, capturing the local background will have similar success. For ground-based observations,
353 capturing the local background across varied wind conditions becomes more challenging, and can
354 greatly increase the number of observations needed to minimize the biosphere influence. Dense
355 networks of surface-level, lower-cost monitors [24] could potentially be leveraged to assess urban fossil
356 fuel emissions with observations of local-background to attempt to minimize biosphere influences. This

357 does not imply sparse urban CO₂ observing networks do not provide additive information on CO₂ fossil
358 emissions, but merely that this observing strategy has greater sensitivity to biosphere fluxes.

359 In addition, the local background observing framework minimizes errors in how well the
360 biosphere needs to be represented in studies of urban CO₂. Here, we used two biosphere models that
361 had an average bias of >80%; but the difference of impact a simulated CO₂ enhancement, on average,
362 was ~12%, further demonstrating the efficacy of this observing approach to characterizing urban fossil
363 fuel emissions. Outside of summer months, the biosphere can generally be ignored in studies of urban
364 CO₂ that utilize such a local background observing framework, providing opportunities to perform robust
365 studies of fossil fuel emissions within cities.

366 Model simulations for each study domain were assessed in the afternoon, consistent with the
367 timing of most airborne campaigns, which are scheduled for times with developed boundary layer
368 heights to best capture urban emissions and be less subject to long-range transport. In addition, this
369 approach allows for comparison to satellite retrievals whose overpasses occur in the afternoon and
370 which can be used to sample the urban domain considering a local background much like the flight-
371 based sampling approach. Further, the time of the airborne transects as assessed here generally aligns
372 with the period when the biosphere drawdown will be more pronounced, suggesting the findings
373 presented here capture a high biosphere contribution to urban CO₂ from the local background
374 approach. An additional advantage of using this local background approach to isolating fossil fuel
375 emissions is the observations do not have to be directly along the edge of a city, but can have additional
376 biosphere interferences between the observing points and urban core as this approach is driven by
377 defining the urban-suburban-rural gradient, unlike in single-tower based studies where that additional
378 biosphere contribution would further promote the amplified role of the biosphere to urban CO₂.

379 Cities around the world, which are often associated with large carbon footprints, have
380 announced targets to reduce emissions and transition to net-zero economies across their supply chains.
381 Using top-down atmospheric observations has become a critical tool to evaluate such emissions;
382 however, for CO₂ specifically, these inversion approaches are subject to interferences from the
383 biosphere, which can act as both a source and sink of CO₂. Here, using high-resolution, back-trajectory
384 modeling approaches, we show that using observing frameworks that incorporate a local background
385 (e.g., via airborne or space-based observations) can minimize biosphere interference to urban CO₂ for
386 estimate whole city emissions, paving the way to conduct robust studies of urban fossil fuel
387 CO₂ emissions to ultimately reduce emissions and combat climate change.

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394 **OPEN RESEARCH**

395 Data Availability Statement:

396 Aircraft observation data as part of the ECO flight campaign is available at
397 ftp://aftp.cmdl.noaa.gov/data/campaign/ECO_2018/ECO_2018_insitu_10sec.nc. STILT back-trajectory
398 model source code is available for downloaded at <https://uataq.github.io/stilt/#/>. ACES emission
399 inventory is available from the ORNL DAAC at
400 https://daac.ornl.gov/CMS/guides/CMS_Carbon_Emissions_NE_US.html. VPRM hourly flux data is
401 available at <https://data.nist.gov/od/id/mds2-2382>. SMUrF hourly flux data is available from the ORNL
402 DAAC at https://daac.ornl.gov/cgi-bin/dsvviewer.pl?ds_id=1899. HRRR met data is available at
403 <https://rapidrefresh.noaa.gov/hrrr/>

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