

Improvements of a low-cost CO₂ commercial nondispersive near-infrared (NDIR) sensor for unmanned aerial vehicle (UAV) atmospheric mapping applications

Yunsong Liu^{1,2}, Jean-Daniel Paris^{1,2}, Mihalis Vrekoussis^{2,3}, Panayiota Antoniou², Christos Constantinides², Maximilien Desservettaz², Christos Keleshis², Olivier Laurent¹, Andreas Leonidou², Carole Philippon¹, Panagiotis Vouterakos², Pierre-Yves Quéhé², Philippe Bousquet¹, and Jean Sciare²

¹Laboratoire des Sciences du Climat et de l'Environnement, CEA-CNRS-UVSQ, UMR8212,

IPSL, 91191 Gif sur Yvette, France

²Climate and Atmosphere Research Center (CARE-C), the Cyprus Institute, 2113 Nicosia, Cyprus

³Institute of Environmental Physics and Remote Sensing (IUP) & Center of Marine Environmental Sciences (MARUM), University of Bremen, 28359 Bremen, Germany

Correspondence: Yunsong Liu (yunsong.liu@lsce.ipsl.fr)

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Abstract. Unmanned aerial vehicles (UAVs) provide a costeffective way to fill in gaps between surface in situ observations and remotely sensed data from space. In this study, a novel portable CO₂ measuring system suitable for operations on board small-sized UAVs has been developed and validated. It is based on a low-cost commercial nondispersive near-infrared (NDIR) CO₂ sensor (Senseair AB, Sweden), with a total weight of 1058 g, including batteries. The system performs in situ measurements autonomously, allowing for its integration into various platforms. Accuracy and linearity tests in the lab showed that the precision remains within ± 1 ppm (1 σ) at 1 Hz. Corrections due to temperature and pressure changes were applied following environmental chamber experiments. The accuracy of the system in the field was validated against a reference instrument (Picarro, USA) on board a piloted aircraft and it was found to be ± 2 ppm (1σ) at 1 Hz and ± 1 ppm (1σ) at 1 min. Due to its fast response, the system has the capacity to measure CO₂ mole fraction changes at 1 Hz, thus allowing the monitoring of CO₂ emission plumes and of the characteristics of their spatial and temporal distribution. Details of the measurement system and field implementations are described to support future UAV platform applications for atmospheric trace gas measurements.

1 Introduction

According to the IPCC (2022), the global mean temperature will increase by at least 1.5 °C in the next 20 years relative to the pre-industrial period for all scenarios. This warming, attributed to human activities, is driven by the increased emissions of heat-trapping greenhouse gases (GHGs) in the atmosphere. Impacts of global warming, such as heat waves, extreme precipitation events, sea-level rise, and biodiversity loss, are already visible, affecting human societies and natural ecosystems (IPCC, 2018; Khangaonkat et al., 2019). Because of its importance, global warming has become one of the most critical challenges of the 21st century from both a scientific and societal perspective. To tackle global warming, almost all members of the United Nations agreed to join forces to keep the warming below 2 °C (ideally below 1.5 °C) under the Paris Agreement of 2015. This agreement intensifies the need to strengthen our capacity of having highquality and accurate observations of atmospheric GHG at all scales including local, regional, and global measurements both at the surface and vertically resolved. Atmospheric concentration measurements from various platforms can therefore be used to estimate emissions at different scales.

Carbon dioxide (CO_2) is the most abundant, humanreleased GHG in the atmosphere. Notably, the CO_2 mole fraction recently reached a new high in 2020 of 413.2 \pm 0.2 µmol mol⁻¹ (ppm), which is 49 % over its preindustrial level (WMO, 2021). About 90 % of total CO₂ emissions emanate from fossil fuel combustion, with around 26 % of it being taken up by the oceans and 30 % by land surfaces (Friedlingstein et al., 2022).

Systematic in situ ground-based measurements of CO₂ started in 1958 in Mauna Loa in Hawaii (Pales and Keeling, 1965). Since then, in situ measurements at many locations but also from various mobile platforms (e.g., cars and ships) have significantly improved our knowledge of the CO₂ spatial and temporal distribution (Daube et al., 2002; Agustí-Panareda et al., 2014; Liu et al., 2018; Defratyka et al., 2021; Paris et al., 2021). Throughout time, in situ measurements have been complemented by remote sensing providing spacebased global observations of CO2 column-averaged mole fraction data and ground-based remote-sensing observations from various instruments (Bovensmann et al., 1999; Wunch et al., 2011, 2017; Turner et al., 2015; Jacob et al., 2016; Frey et al., 2019; Suto et al., 2021). Meanwhile, CO2 instrumentation on board airborne platforms has been developed in the past 20 years (e.g., Watai et al., 2006; Sweeney et al., 2015). These measurements are meant to fill the gap between ground-based observations and remote-sensing spacebased observations to better represent CO₂ spatial distribution at large scales. However, manned (piloted) aircraft which can carry standard analyzers are costly and complex to organize, requiring frequent maintenance (Berman et al., 2012; Bara et al., 2017). Furthermore, at smaller geographical scales (landscape, industrial assets, urban area), manned airborne platforms have strong limitations and cannot fly at low speed in all areas. Unmanned aerial vehicles (UAVs) have been demonstrated to be useful to detect and map emission plumes of other trace gases because of their ability to operate at very low speed/altitude and with slow cruising speeds (e.g., Barchyn et al., 2018). Additionally, UAVs, unlike piloted aircraft, can operate over hazardous areas such as volcanic eruptions and forest wildfires. Actually, high-precision calibrated CO2 instruments have been deployed in manned aircraft (e.g., Paris et al., 2008; Xueref-Remy, et al., 2011; O'Shea et al., 2014; Pitt et al., 2019; Barker et al., 2020), but they are too heavy, large, and expensive for UAV applications. However, until now very few calibrated CO₂ measurements have been reported in the literature (Kunz et al., 2018) due to the challenge of measuring this species with sufficient precision.

A large part of the anthropogenic CO_2 originates from point emission sources such as power plants burning fossil fuels (Pinty et al., 2017; Reuter et al., 2021). An appropriate sensor for UAV platforms would have the potential to provide independent CO_2 measurements across these source plumes to verify mitigation strategies. Often the CO_2 signals of strong emitters can be mixed with strong biospheric signals even at local scales. In addition, the planetary boundary layer (PBL) dynamics can strongly influence atmospheric concentrations. It is therefore important to separate the influence of exogenic factors and isolate the contribution from targeted emission plumes. Another potential application of a UAV CO₂ system is to document the spatial distribution of CO₂ around fixed observations. Watai et al. (2006) argued that UAVs have the potential to provide measurements close to the surface and inside the PBL complementary to data obtained from fixed observatories such as tall towers and make frequent and simultaneous measurements in multiple locations at low cost. In this case, UAV measurements help separate signal variability into a large-scale footprint of ground stations and variability due to local influences. Despite these challenges, there have been ongoing efforts to develop compact, lightweight, and low-powered GHG sensors, able to be integrated into UAVs to address these needs. Berman et al. (2012) developed a highly accurate UAV greenhouse gas system (but heavy: 19.5 kg) for measuring carbon dioxide (CO₂) and methane (CH₄) mole fraction. Malaver et al. (2015) integrated a non-dispersive near-infrared (NDIR) sensor (3285 g) for CO₂ measurement into a solar-powered UAV for effective 3D monitoring. Kunz et al. (2018) reported the development of high-accuracy (± 1.2 ppm) CO₂ instrumentation well-suited for UAVs. However, the commercial CO2 sensor used in the study was disassembled and redesigned, making it difficult to replicate widely. Allen et al. (2019) applied a UAV CO₂ sensor system to infer a landfill gas plume. Chiba et al. (2019) developed a UAV system (2.7 kg) to measure regional CO₂ mole fraction and obtain vertical distributions within 1.75 ppm standard deviation over a farmland area and deduced vegetation sink distribution from their results. More recently, Reuter et al. (2021) developed a lightweight (about 1.2 kg) UAV system to quantify CO₂ emissions of point sources with a precision of 3 ppm at 0.5 Hz. Moreover, very high-precision and commercial sensors (< 0.2 ppm 1σ at 1 Hz) for UAV applications are emerging currently such as the ABB light micro-portable greenhouse gas analyzer (pMGGA) (Shah et al., 2020). However, the weight (about 3 kg) is much larger and the price is more expensive compared to the NDIR sensors mentioned in the above literature.

These works have faced the difficulty of miniaturizing high-precision, fast-response CO_2 sensors. Few studies among them could reach a CO_2 measurement accuracy below 2 ppm with light payload (2 kg) on board UAVs. It is also challenging to have stable and high-frequency measurements against rapid changes in pressure and temperature, which is also the main reason for UAV CO_2 measurements not being widely applied. Therefore, this study aims to develop a cost-effective, compact, lightweight CO_2 measurement system with high frequency and accuracy that can be widely used in different UAV applications. Targeted applications include emission estimates from point sources, stack emission factor measurements, and mapping CO_2 distribution in mixed natural–urban environments.

Towards this goal, a portable CO_2 sensor system has been developed based on a low-cost commercial NDIR CO_2 sensor (Senseair AB, Sweden). Prior to integration, the accuracy and linearity of the instrument were ensured with a series of laboratory tests. The performance of the system was validated during laboratory (chamber) and ambient conditions. For the latter, the system was installed on board a manned aircraft and unmanned aerial vehicle platforms. As a proof of concept, intensive flights of the developed UAV CO_2 sensor system were presented in the urban area (Nicosia, Cyprus). It is shown that our system is easy to reproduce, enabling a wide range of field applications, such as urban and pointsource emissions monitoring. Moreover, the system developed in this study has the potential to accommodate other sensors to make stack emission ratio measurements.

2 Methodology

2.1 CO₂ sensor

The sensor used in this study is a non-dispersive nearinfrared (NDIR) sensor from Senseair AB based on their High-Performance Platform (HPP) 3.2 version for gas detection below parts per million. These sensors measure the molar fraction of CO_2 in the optical cell based on infrared (IR) light absorption, based on the Beer–Lambert law (Barritault et al., 2013). The multi-pass cell of the sensor provides eight round trips of the beam with a total path length of 1.28 m. Temperature-controlled molded optics in the sensors are used to keep the temperature of the sensor cell constant to prevent condensation on the mirrors (Hummelgård et al., 2015). This study involved two CO_2 sensor units using this technology (named SaA and SaB hereafter). More information on the sensor can be found in Arzoumanian et al. (2019).

2.2 Laboratory tests

The schematic diagram of the measurement setup used for laboratory testing is shown in Fig. 1. In this setup, the sampled air first passes through a 15 cm cartridge filled with magnesium perchlorate ($Mg(ClO_4)_2$), which is sufficient to dry air at room temperature (24 °C) and a flow rate of 500 mL min^{-1} to a water mole fraction of 20 ppm for 2 h, and then through a 0.5 µm membrane filter to remove particles. A diaphragm micro-pump (GardnerDenverThomas, USA, Model 1410VD/1.5/E/BLDC/12V) drives the air through the gas line towards SaA and SaB. Temperature and relative humidity are continuously monitored via a SHT75 sensor placed between the micro-pump and the two sensors. Finally, a Raspberry Pi3 acquires the data from all the sensors. The integrated system is powered by a 12 V DC supply, isolated from the UAV power system. Parallel to the two sensors, a Picarro model G2401 instrument (Picarro, USA) based on cavity ring-down spectroscopy (CRDS) (Crosson, 2008) served as a reference instrument in this setup (see Fig. 1).

Figure 2 presents the data quality control procedure flowchart. SaA and SaB were first tested in the metrology labo-

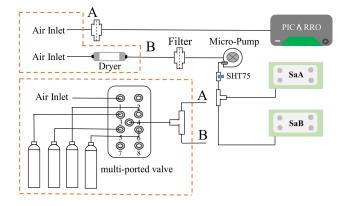


Figure 1. The schematic of the developed system for lab tests and field deployment (A and B represent air flows to G2401 and CO_2 sensors, respectively).

ratory of the Integrated Carbon Observation System (ICOS) Atmosphere thematic center (ICOS ATC). Then the system was integrated into a manned aircraft and UAVs to be validated and evaluated under ambient conditions. Table 1 is a summary of all the laboratory and field tests performed for the system, and all results are presented in Sect. 3. In the laboratory, four calibration sequences were performed to determine the calibration function that linked the measured values to the assigned values (Yver Kwok et al., 2015). Four high-pressure calibration standard gas cylinders with known amounts of CO₂, ranging from 380.096 to 459.773 ppm, were used. The standard gases were calibrated using the international primary standard for GHG, maintained in NOAA CMDL, Boulder, Colorado, USA (https://gml.noaa.gov/ccl/, last access: 9 February 2022). To ensure stabilization after adequate flushing of each sensor's cell with CO₂, each standard gas ran for 30 min continuously and only the last 10 min of data was used. Then the calibration function using a linear fit was calculated for the sensors and the Picarro instrument. The cylinder with 459.773 ppm CO₂ was considered to resemble ambient atmospheric conditions. During the Allan deviation test (Hummelgård et al., 2015), the CO₂ sensors continuously measured a cylinder filled with dry air for 24 h.

Temperature (*T*) and pressure (*P*) sensitivity tests were performed in a closed automated climate chamber at the Observatoire de Versailles Saint-Quentin-en-Yvelines (OVSQ) Guyancourt, France, using the Plateforme d'Integration et de Tests (PIT). The temperature (from -60 to $100 \,^{\circ}$ C) and pressure (from 10 to $1000 \,^{h}$ Pa) ranges inside the chamber can be controlled and supervised by the Spirale 2 software (https:// www.ovsq.uvsq.fr/essais-thermiques, last access: 9 February 2022). We implemented repeated sequences of variable temperature and pressure following Arzoumanian et al. (2019). These tests allow determining the linear response of SaA and SaB sensors against temperature and pressure (as shown in Sect. 3.2). 4434

2.3 Aircraft test

After a series of laboratory tests, the sensors were moved to a manned aircraft together with a reference instrument Picarro G2401-m to test the performance of SaA and SaB under real atmospheric conditions.

SaA, SaB, and the reference Picarro instrument G2401-m were flown on board a manned aircraft on 8 April 2019 in the vicinity of Orléans forest (150 km south of Paris), France. All instruments were calibrated using standard cylinders from ICOS ATC before and after the flight (Hazan et al., 2016). The setup used and the aircraft are shown in Fig. S1 in the Supplement. These flights aimed to confirm the accuracy of SaA and SaB in real flight conditions.

2.4 Unmanned aerial vehicle (UAV) system integration

Then, for further validation the system was miniaturized and integrated into a small-size unmanned aerial system (UAS), developed at the Unmanned Systems Research Laboratory (USRL) of the Cyprus Institute (CyI) (https://usrl.cyi.ac.cy/, last access: 28 March 2022). The components of the integrated system are shown in Fig. 3a. The CO₂ sensor setup weighs 1058 g with dimensions of $15 \text{ cm} \times 9.5 \text{ cm} \times 11 \text{ cm}$, including the battery. A 15 cm customized cartridge was used here to reduce volume and weight. The impact of water vapor dilution on dry CO₂ mole fraction is within 40 ppb by using the dryer. It does not depend on external systems, allowing for its integration into various small UAVs. The system was successfully integrated into the USRL small-sized quad-rotor UAS (Fig. 3b), optimally developed in terms of minimum size and maximum performance, to accomplish the desired CO₂ unmanned measurements. Multi-rotors allow vertical take-off and landing (VTOL) in urban and remote regions (Kezoudi et al., 2021). The UAS has up to 30 min flight endurance for atmospheric measurements with the selected sensor. In order to improve accuracy and response time for in-flight temperature measurements (critical for CO₂ correction), a Rotronic HC2-ROPCB sensor (Rotronic, Switzerland) replaced the SHT75 sensor. To validate the system on site, calibration sequences were performed before and after the flights in the laboratory. In addition, a target gas cylinder was performed for 20 min between each flight to determine and correct the instrument's drift over time.

3 Results

3.1 Sensor calibration

The response curves obtained from the CO_2 calibration are shown in Fig. S2. The stability of successive CO_2 calibrations is shown in Fig. 4, which presents the difference between CO_2 mole fraction measured by sensors and CO_2 mole fraction assigned to each calibration cylinder. The biases of SaA and SaB against the four calibration standards are nega-

Code name	Name	Purpose	Parameters	Range of T (°C) and P (hPa)	Duration (h) Sensors testee	Sensors tested
CA	Calibration	Test calibration frequency and stability	CO ₂	N/A		SaA, SaB
AD	Allan deviation	se, stability,	CO_2	N/A	24	SaA, SaB
		and detection limit				
PT1	Temperature and pressure	Correlation between	T, P	0–45 and 600–1000	72	SaA, SaB
	sensitivity tests	CO_2 and P/T				
PT2	Temperature and pressure	Correlation between	T, P	0–45 and 600–1000	72	SaB
	sensitivity tests	CO_2 and P/T				
SF1	Simulation flight	Estimate P/T impact	T, P	15-25 and 800-1000	4	SaB
SF2	Simulation flight	Estimate P/T impact	T, P	15-35 and 600-1000	10	SaB
Aircraft test	Manned aircraft test	Estimate precision on board	T, P, CO_2	Real flight conditions	2.5	SaA, SaB
UAV test	UAV test	Test and evaluate the system on UAVs	T, P, CO_2	Real flight conditions	~ 0.3	SaB

Table 1. A summary of the laboratory tests and field deployment

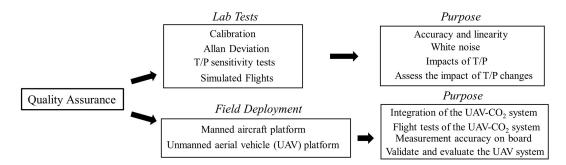


Figure 2. The flow chart of data quality control procedures.

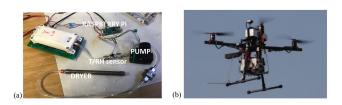


Figure 3. Components of the portable CO_2 sensor system setup (a) and the selected UAV (b).

tive and positive, respectively, during the calibration (Fig. 4). Additionally, the biases increased by 0.2 ppm on average between calibration sequences (2 h of each sequence). This drift against the sensors' running time is further investigated and validated in the field deployment (Sect. 3.4). The result of the Allan deviation (AD) test is shown in Fig. S3. The plot shows the stability as a function of integration time (Hummelgård et al., 2015). The unfiltered data were used from the HPP data set. The precision improved by increasing the integration time. However, the sensors were intended for mobile platforms, their performance at 1 Hz was chosen as the most significant. The precision is, respectively, ± 0.36 ppm (1 σ) and ± 0.85 ppm (1 σ) for SaA and SaB at 1 Hz (S3), which shows the precision of the sensors in the laboratory is below 1 ppm (1 σ) at 1 Hz.

3.2 Temperature and pressure dependence

3.2.1 Temperature sensitivity test

During temperature sensitivity tests, the chamber pressure was kept constant at 950 hPa, while the temperature was gradually changed, as seen in Fig. S4. The temperature ranged between 0 and 45 °C, following 9 °C increment steps, lasting for 20 min. The sensors' cell temperature exhibited an unstable behavior for chamber temperatures below 25 °C, while it was stable, at approximately 57 °C, for chamber temperatures above 25 °C. However, SaA and SaB behaved oppositely when their cells' temperature changed. Therefore, two scenarios were considered for both sensors.

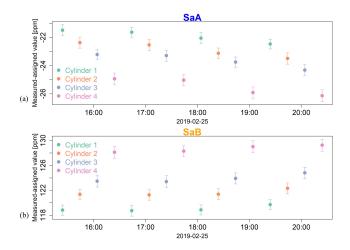


Figure 4. Stability of successive CO_2 calibrations for SaA (**a**) and SaB (**b**); the error bars represent the standard deviation of 2 s averages.

The first scenario is when the analyzer's cell temperature is stable while the ambient air temperature changes (above 25 °C). The trend coefficients of CO₂ mole fraction over ambient temperatures were -0.564 and -0.527 for SaA and SaB, respectively (shown in Fig. 5a and c). The second scenario is when both the analyzer's cell and ambient temperatures change simultaneously. In this case, the impact of ambient air temperature changes obtained from the first scenario has been corrected prior to considering the cell temperature changes. The trend coefficients of CO₂ mole fraction over cell temperatures were -0.979 and 0.378 for SaA and SaB, respectively (shown in Fig. 5b and d). Consequently, SaA performed better when applying the temperature sensitivity test (high R^2 , lower standard error).

3.2.2 Pressure sensitivity test

During the pressure tests, the chamber temperature was maintained at $25 \,^{\circ}$ C, and pressure ranged from 600 hPa corresponding to 3 km above sea level (a.s.l.) to 1000 hPa in 100 hPa steps, repeated twice. SaA and SaB performed significantly differently in this test, with the SaB sensor show-

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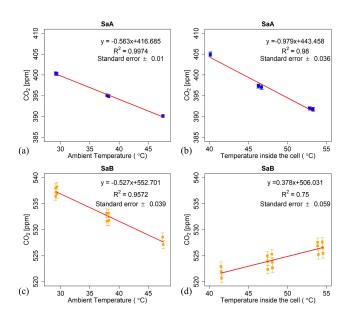


Figure 5. Temperature sensitivity tests in the environment chamber: (a) and (c) represent the first scenario; (b) and (d) represent the second scenario.

ing increased sensitivity to pressure changes (Fig. 6). Generally, the sensors have an internal pressure correction from the manufacturer, and it is apparently not implemented in SaB. However, SaB performed better in the pressure sensitivity test, with tighter linearity (higher R^2) when both tests were accounted for.

From the sensitivity tests presented above, we derived the following equations for both sensors:

SaA:
$$C_{\rm cor} = C_{\rm obs} + 0.564 \times (T_{\rm a} - T_{\rm a0}) + 0.979$$

 $\times (T_{\rm c} - T_{\rm c0}) - 0.013 \times (P - P_0),$ (1)

SaB:
$$C_{\rm cor} = C_{\rm obs} + 0.527 \times (T_{\rm a} - T_{\rm a0}) - 0.378$$

 $\times (T_{\rm c} - T_{\rm c0}) - 0.607 \times (P - P_0),$ (2)

where C_{cor} is the mole fraction after correction for P/T changes. C_{obs} is the observed mole fraction. T_c represents the analyzer's measurement cell temperature, and T_{c0} is the original cell temperature at the start of the measurements. T_a represents the ambient temperature, and T_{a0} is the ambient temperature at the start of the measurement. P represents the ambient pressure, and P_0 is the ambient pressure at the start of the measurements. The equations are also applied for calibrations.

Replications of temperature and pressure sensitivity tests for SaB at a later stage showed high consistency with the initial results presented above. Both sensors have shown different responses in the tests. Therefore, it is essential to perform both temperature and pressure sensitivity tests for individual sensors to obtain their individual correction equations against

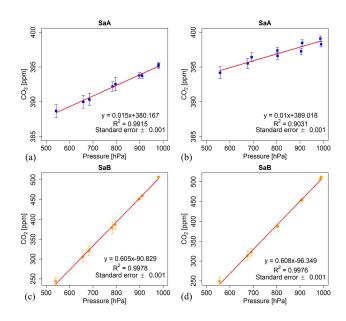


Figure 6. Pressure sensitivity tests in the environment chamber: (a) and (b) represent SaA results of the repeated pressure tests; (c) and (d) represent SaB results of the repeated pressure tests.

temperature and pressure changes. Here, we highly recommend characterizing every individual sensor at least once before any use. We also recommend repeating (e.g., annually) these tests regularly as sensor performances tend to change over time.

3.3 Manned aircraft test results

SaA and SaB measured consistently with the Picarro G2401-m for atmospheric pressure above 800 hPa (equal to 1.5 km a.s.l.) (see Fig. 7a). Their precision was ± 1.4 ppm (1 σ) and ± 1.7 ppm (1 σ) at 1 Hz and 0.78 ppm (1 σ) and ± 1.1 ppm (1 σ) with minute-averaged data, respectively (Fig. 7b), larger than the precisions calculated during the laboratory tests. This degradation was expected due to less optimal measurement conditions. Therefore, the test on the piloted aircraft shows the sensors' precision on board under real flight conditions is within 2 ppm (1 σ) at 1 Hz and improves to about 1 ppm (1 σ) with minute-averaged data.

3.4 Unmanned aerial vehicle (UAV) tests and validation

SaB was chosen for field deployments due to technical issues with SaA. SaB was integrated into a quad-rotor to evaluate and validate the performance of the sensor on board a UAV platform during flights. The flight path was over the Athalassa National Forest Park (35.1294° N, 33.3916° E) in Nicosia, Cyprus (Fig. 8). Four flights were performed on 10 June 2021 from 15:00 to 18:00 LT. The procedure was the following: calibration response curves were obtained before and after the flights. A target gas cylinder was measured for 20 min between each flight to characterize the instrument

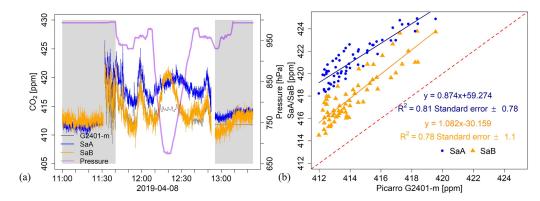


Figure 7. Manned aircraft results: (a) is the time series, and the gray shaded parts present measurements on the ground and measurements of the gas cylinder; (b) is the correlation between CO_2 sensors and G2401-m.



Figure 8. The map presents the locations of the Picarro G2401 at CyI, the UAV flight path, Athalassa National Forest Park, and the residential area in Nicosia (© Google Earth 2022).

drift. The sensitivity correction Eq. (2) was then applied to the raw data. It was noted that the measured target gas mole fraction drifted linearly throughout the day (Fig. S5a). To account for that, a time-dependent correction, based on running time, was calculated and applied for calibration sequences (Fig. S5a). Practically, this correction was applied to obtain flight-specific calibration response curves according to the sensor running time and confirmed by the target linear drift (Fig. S5).

Reference CO_2 measurements were additionally conducted with another Picarro G2401 on the roof of the Novel Technologies Building (NTL) at the Cyprus Institute (CyI) (Fig. 8), at 174 m a.s.l., 1.82 km northwest upwind from the UAV launching location (187 m a.s.l.). Therefore, the flight path was downwind from the Picarro G2401. The residual values of CO_2 between the Picarro and UAV CO_2 systems varied from 0.2 to 2.1 ppm (median = 1.1 ppm) during the experiment.

4 Case study for CO₂ measurements in an urban environment (Nicosia)

The field campaign to test operation in real conditions of our UAV CO₂ system was performed on 14 May 2021 from early morning 06:00 LT to late afternoon 17:30 LT. It took place above the Athalassa National Forest Park located southeast of CyI in Nicosia, where 16 flights were performed. Each flight lasted approximately 15 min with most of the flight performed at a constant altitude of 50 and 100 m above ground level (a.g.l.) alternatively. The altitudes were determined following security rules. Firstly, the UAV had to maintain a safe distance above the treeline of the forest park. Therefore, the lowest safe altitude to fly the drone was 50 m a.g.l. Secondly, the ceiling of the UAV CO₂ flights was set to 100 m a.g.l., following the European regulations (2019/947 and 2019/945; EASA, 2022) for UAV operations in sparsely populated areas (open category A2), with flights permitted up to 120 m a.g.l. The two selected altitudes were used alternatively in order to obtain representative measurements for either horizontal "mapping" or vertical gradients. The vertical gradients were completed at lower altitudes by rooftop measurements in a nearby building. CO₂ mole fractions, as well as meteorological conditions, were measured during the flights on the roof of NTL at CyI. CO₂ measurements were done using a Picarro G2401 (174 m a.s.l., 16 m a.g.l., 35.141° N, 33.381° E); wind speed and wind direction were measured using a sonic anemometer Clima Sensor US model 4.920x.x0.00x with a resolution of a wind speed of 0.1 m s^{-1} and a wind direction of 1°.

Each pair of 50 and 100 m altitude flights lasted approximately 1 h (including flight time and the time needed to change the dryer and battery on the ground). The 15 cm cartridge filled with magnesium perchlorate (Mg(ClO₄)₂) was changed every two flights. The first six flights (three pairs) were performed continuously from 06:00 to 09:00 LT, as well as the last six flights from 15:00 to 17:30 LT. In between, four flights (two pairs) took place between 10:00 and 11:00 LT and between 13:00 and 14:00 LT.

According to the meteorological station data, the wind direction in the morning (before 08:00 LT) was from the northwest, with an average wind speed of 1.2 m s^{-1} . Then the wind direction shifted to northeast and southeast during the day before 13:00 LT, with an average wind speed of 0.9 m s^{-1} . Afterwards, the wind shifted back to northwest but with stronger wind speeds (average of 5.3 m s^{-1}).

Figure 9a displays the measured CO_2 (ppm) time series from all UAV flights and the Picarro. The CO_2 mole fraction measured during the flights in the early morning and evening, when northwesterly winds occurred, was consistent with that measured by the G2401. A CO_2 enhancement linked to the morning traffic peak (from 07:00 to 08:00 LT) was detected at all altitudes. Interestingly, the two measurements eventually differed at 10:00 LT, creating a vertical gradient: the CO_2 mole fraction measured on board the UAV remained constant, whereas a decrease of about 5 ppm was measured by the G2401 on the ground.

During the day, with the surface wind direction shifting starting 08:00 LT from northwest to northeast and then southeast, the Picarro G2401 progressively sampled air from the Athalassa National Forest Park. The park, with a total area of 8.4 km², is an oasis of greenery with many trees, shrubs, and grasses located on the southeastern edge of Nicosia. Considering that the inlet of the G2401 is at the same altitude above sea level as the UAV launching location, the lower observed CO₂ mole fraction by the G2401 can most likely be attributed to the Athalassa National Forest Park acting as a surface sink taking up CO_2 . The reduction of traffic after the peak hour can also play a role in the first part of the day, when the air was blowing from the north. At 50 or 100 m height, the constancy of CO₂ mole fractions during the day may suggest a different origin for the air sampled depending on the wind direction at these altitudes (wind was not measured on board the UAV). Potential origins may include "regional" air moving above the surface layer or a plume of emissions from the city lofted at a few tens of meters with a stratified air mass above the park.

During the afternoon, the progressive convergence of surface and UAV observations, with a decrease in UAV CO_2 values, suggests either a diffusion of the surface signals at altitude or an enhanced atmospheric mixing. This explanation could be supported using an anemometer integrated on board the UAV to provide additional wind data at various heights. UAV-integrated wind measurements would have to be considered for future applications.

A CO₂ mapping during the peak traffic hour is shown in Fig. 9c combined with the flight path at 100 m (the red dot represents the launching site). Figure 9b shows the corresponding CO₂ time series combined with wind direction (arrow head) and wind speed (arrow length) information. The high mole fraction (20 ppm above background levels) probably originated from local traffic emissions from the main road

southwest of the Athalassa National Forest Park (Fig. 8). This finding highlights the capability of the developed UAV CO_2 sensor system to detect fast mole fraction changes and the potential to provide useful insights into CO_2 emissions close to the ground in urban areas.

From the vertical profiles (Fig. 10), the difference between the 06:00 and 07:00 LT profiles highlights the peak traffic hour. Additionally, we observed an increasing difference (about 3 ppm) between ground level and 50 m a.g.l., followed by a difference (about 0.5 ppm) between 50 m and 100 m a.g.l. from 08:00 to 13:00 LT when the air mass came from the Athalassa National Forest Park with an average wind speed of $0.9 \,\mathrm{m \, s^{-1}}$. This suggests that the CO₂ mole fraction measured by the G2401 and UAV CO2 system represents local CO₂ characteristics and that the Athalassa National Forest Park acted as a CO2 sink. Later on, between 15:00 and 17:00 LT when the average wind speed increased (5.3 ms^{-1}) , the CO₂ mole fraction at 50 and 100 m a.g.l. converged towards surface values. This suggests that the observed wind speed enhancement enabled a better mixing of surface signals at altitude. However, the transport of wellmixed regional background air masses at the measurement area could also be an alternative explanation (background CO₂ mole fraction is 418.9 ppm). Although we demonstrated the usefulness of UAV measurements to capture horizontal and vertical CO₂ gradients in the planetary boundary layer in an urban or peri-urban environment, a definitive explanation of this particular observation would be beyond the scope of this paper.

5 Conclusions

Following the integration of an NDIR CO₂ sensor, we developed and validated an autonomous system that can be regarded as a portable package (1058 g), suitable for CO₂ measurements on board small UAVs (or other platforms) with good field performance after applying calibration and data corrections (± 1 ppm accuracy for 1 min averages). Prior to deployment, and in order to acquire high-quality observations, the sensor followed a series of quality control procedures. The laboratory tests indicated that the precision was within $\pm 1 \text{ ppm} (1\sigma)$ at 1 Hz. Two CO₂ sensors (SaA and SaB) were tested. It is essential to conduct calibrations before any measurements as shown in this study. NDIR CO2 sensors should not be regarded as plug and play without conducting calibrations and bias correction prior to any measurement campaigns as measurement data would suffer from large, unknown biases without that important step. In general, we advocate that low- and mid-cost sensor units should systematically be characterized for their dependence on pressure and temperature and their factory correction and calibration verified. Strategies for field deployment should also take into account the significant drift that can be observed at the hourly scale. Using a single target gas between flights is

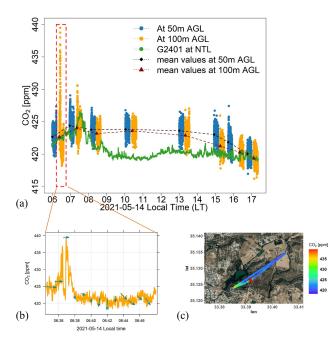


Figure 9. (a) Time series of CO_2 mole fraction measured by the UAV CO_2 sensor (at 50 m in blue and 100 m a.g.l. in orange) and by the Picarro G2401 at CyI (in green). The black diamonds represent the averaged CO_2 mole fraction measured by SaB during the flights at 50 m, and the dark red triangles represent the averaged CO_2 mole fraction measured by SaB during the flights at 100 m. (b) The corresponding CO_2 time series combined with wind direction (arrow head) and wind speed (arrow length) information obtained from the nearby meteorological station, which is a zoom of the second flight marked in the red dashed box in (a). Panel (c) presents the CO_2 mapping (the red triangle represents the launching location) during the rush hour (map data: © Google, Maxar Technologies).

sufficient to cope with this drift. Alternative strategies to correct the drift without using gas cylinders on the field remain to be explored, such as comparison against a high-precision instrument at regular intervals during the deployment. Each sensor's performance is impacted by changes in pressure and temperature; therefore, it is necessary to perform pressure and temperature sensitivity tests before any field applications.

Further validation on board a manned aircraft resulted in an estimated precision of $\pm 2 \text{ ppm}(1\sigma)$ at 1 Hz and $\pm 1 \text{ ppm}(1\sigma)$ at a 1 min time resolution. During the integration of our system on board a small quadcopter, the calibration strategy was extended to account for running-time-dependent instrumental drifts. Due to its simplicity, the developed system can be replicated easily for wider applications since it has compact, cost-effective, and lightweight advantages. It is anticipated that the integrated portable package can be used in the investigation of emission ratios and fluxes, especially when combined with other sensors on board the UAV platform.

As a proof of concept, the developed system was deployed in a UAV-based flight campaign, where several horizontal

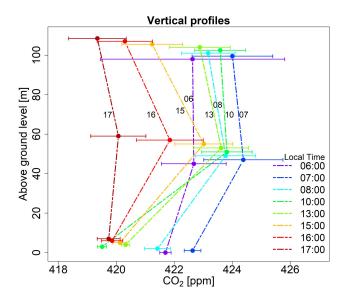


Figure 10. Vertical profiles from the eight pairs of flights. The ground level values are from the Picarro G2401 at CyI. CO_2 at 50 and 100 m a.g.l. are from the UAV CO_2 sensor horizontal flights; the error bars represent the standard deviation of the duration of each flight.

flights were performed near the ground and up to 100 m in height. The mole fraction of CO₂ up to 440 ppm (20 ppm above the background levels) was detected during the morning traffic rush hour, attributed to emission from a major road located on the southwest of the Athalassa National Forest Park. The CO₂ mole fraction measured by the UAV system was consistent with that measured by the Picarro G2401 at CyI when the flight path was downwind of CyI. The system also revealed its ability to capture the temporal variability of the vertical CO₂ gradient between the surface and the lower atmosphere. The observed CO2 profiles depict the contribution of traffic emission in the morning from 06:00 to 08:00 LT and also a probable sink due to the Athalassa National Forest Park during the course of the day from 08:00 to 13:00 LT. Furthermore, the measurement system captured the mole fraction drop from 15:00 to 17:00 LT observed at different height levels due to the intensification in the wind speed leading to more horizontal and vertical mixing. In conclusion, the designed system demonstrated its capability to measure fast mole fraction changes and spatial gradients and to provide accurate plume dispersion maps. It proved to be a good complementary measurement tool to the in situ observations performed at the surface.

Data availability. The data presented in this study are based on many different experiments, given the fact that our experiments and field deployments were aimed at characterizing the two sensors used here. The data are not made publicly available in a repository but can be requested from the corresponding author.

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