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2-D tomography of volcanic CO$_2$ from scanning hard target differential absorption LIDAR: The case of Solfatara, Campi Flegrei (Italy)

Manuel Queisser$^1$, Domenico Granieri$^2$, Mike Burton$^1$

$^1$School of Earth, Atmospheric and Environmental Sciences, University of Manchester, Oxford Road, Manchester M13 9PL, UK
$^2$Istituto Nazionale di Geofisica e Vulcanologia (INGV), Sezione di Pisa, 50126 Pisa, Italy

Correspondence to: Manuel Queisser (manuel.queisser@manchester.ac.uk), Tel.: +44(0)161 2750778, Fax.: +44(0)161 306 9361

Abstract. Solfatara is part of the active volcanic zone of Campi Flegrei (Italy), a densely populated urban area where ground uplift and increasing ground temperature are observed, connected with rising rates of CO$_2$ emission. A major pathway of CO$_2$ release at Campi Flegrei is diffuse soil degassing, and therefore quantifying diffuse CO$_2$ emission rates is of vital interest. Conventional in-situ probing of soil gas emissions with accumulation chambers is accurate over a small footprint but requires significant time and effort to cover large areas. An alternative approach is differential absorption LIDAR, which allows for a fast and spatially integrated measurement. Here, a portable hard-target differential absorption LIDAR has been used to acquire horizontal 1-D profiles of CO$_2$ concentration at the Solfatara crater. To capture the non-isotropic nature of the diffuse degassing activity, a 2-D tomographic map of the CO$_2$ distribution has been inverted from the 1-D profiles. The acquisition was performed from a single half space only, which increases the non-linearity of the inverse problem. Nonetheless, the result is in agreement with independent measurements and furthermore confirms an area of anomalous CO$_2$ degassing along the eastern edge as well as the center of the Solfatara crater. The method has important implications for measurements of degassing features that can only be accessed from limited angles, such as airborne sensing of volcanic plumes. CO$_2$ fluxes retrieved from the 2-D map are comparable, but modestly higher than emission rates from previous studies, perhaps reflecting a more integrated measurement.

1 Introduction

Subaerial volcanoes emit a variety of gaseous species, dominated by water vapor and CO$_2$, and aerosols. Originating from exsolution processes that may take place deep in the crust due to the low solubility of CO$_2$ in magmas, volcanic CO$_2$ is a powerful tracer for magmatic recharge and ascent processes (Burton et al., 2013; Frezzotti et al., 2014; Chiodini et al., 2015; La Spina et al., 2015). Measuring volcanic CO$_2$ emission rates is therefore also a feasible pathway towards improved forecasting of volcanic activity, such as seismicity or eruptions (Petrazzuoli et al., 1999; Carapezza et al. 2004; Aiuppa et al.,...
2011). Unfortunately, magmatic CO$_2$ is not only released actively via vents such as the volcano mouth, but also diffusively via soil or flank degassing (Baubron et al., 1991; Hards, 2005; Chiodini et al., 2007). In addition, in most cases the volcanic CO$_2$ signal is modest compared with ambient concentrations (Burton et al., 2013) and quickly diluted into the atmosphere. A common approach to determine the magmatic CO$_2$ flux is based on a gridded sampling of the CO$_2$ distribution in the volcanic plume itself (Gerlach et al., 1997; Lewicki et al., 2005; Diaz et al., 2010; Lee et al., 2016) from which 2-D CO$_2$ concentration maps are retrieved by secondary data processing, such as statistical methods (Lewicki et al., 2005; McGee et al., 2008) and dispersion modeling (Aiuppa et al., 2013; Granieri et al., 2014). Integrating the CO$_2$ concentrations over the cross sectional plume area and multiplying the result with the transport speed perpendicular to the cross section yields CO$_2$ fluxes. The in situ method has two drawbacks. Firstly, it may be dangerous to perform in situ measurements from within the volcanic plume (e.g. due to toxic gases or low visibility near the crater mouth). Secondly, in situ methods allow for a very accurate estimation of CO$_2$ concentration, but only in the vicinity of the measurement point, potentially missing significant contributions from in between the measurement points.

Remote sensing techniques (see Platt et al., 2015 for overview of state-of-the-art), notably active remote sensing platforms, including differential absorption LIDAR (DIAL) and spectrometers (Menzies and Chahine, 1974; Weibring et al., 1998; Koch et al., 2004; Kameyama et al., 2009) acquire columns of range resolved (Sakaizawa et al., 2009; Aiuppa et al., 2015) or column averaged (Amediek et al., 2008; Kameyama et al., 2009) CO$_2$ concentrations. They provide a powerful tool to overcome the aforementioned drawbacks of in situ measurement techniques by offering a faster, safer and comprehensive acquisition (spatial coverage yields inclusive CO$_2$ concentration profiles). Moreover, there is no need for receivers or retroreflectors at the opposite end of the measurement column, which increases not only flexibility and timeliness of the acquisition, but is crucial for some measurements, including airborne or spaceborne acquisitions.

Active remote sensing platforms based on hard target DIAL (topographic target DIAL) can use continuous wave lasers. This allows for high signal return and compact, rugged and portable instruments, which is desirable for platform independent measurement of atmospheric CO$_2$, be it ground based or air-borne (Sakaizawa et al., 2013; Queißer et al., 2015a). Yet, the drawback compared with “traditional”, pulsed DIAL is that no range resolved CO$_2$ concentrations are measured, but column densities (in m$^2$) or, as in this work, path length concentration products (called “path amount” hereafter, in ppm.m). By scanning across the emission feature one obtains 1-D profiles of path amounts. Using these profiles to determine CO$_2$ fluxes is straightforward only for gas plumes for which a homogeneous cross section can be assumed (Galle et al., 2010). However, particularly diffuse degassing activities are often not associated with homogeneous, but an unknown CO$_2$ distribution within the scanned plume cross section. Therefore, the assumption of homogenous CO$_2$ distribution may lead to under or overestimated CO$_2$ fluxes when probed from different directions, since path amounts are measured, which represent path averaged CO$_2$ concentrations. It would be very desirable, and this was the main motivation of this work, to have a 2-D map that at least contains the geometry of the anomalous CO$_2$ release, let aside precise CO$_2$ mixing ratios. This would allow to geometrically correct the fluxes derived from CO$_2$ path amounts delivered by hard target DIAL systems. Provided the 2-D map contains correct CO$_2$ mixing ratios, the CO$_2$ flux can be conveniently obtained by simple integration over the 2-D map.
Note that tomographic reconstructions of volcanic gas plumes have already been performed, however, for SO$_2$ and using passive remote sensing techniques (Kazahaya et al., 2008; Wright et al., 2008; Johansson et al., 2009).

The study was focusing on a zone of diffuse degassing of magmatic CO$_2$ within the Solfatara crater (Italy) reported previously (e.g. Bagnato et al., 2014). Solfatara is a fumarolic field and part of the active volcanic area of Campi Flegrei (CF, Fig. 1). CF is a nested caldera, resulting from two large collapses, the last one ~15 ka ago (Scarpa et al., 1993). CF is in direct vicinity to the metropolis of Naples and thus a direct threat to millions of residents. Thanks to its accessibility and strong CO$_2$ degassing Solfatara provides almost a model like volcano, a natural laboratory, to test new sensing approaches. On the other hand, it is part of one of the most dangerous volcanic zones in the world, showing ground uplift coupled with seismic activity with magma degassing likely having a significant role in triggering unrest (Chiodini et al., 2010). Solfatara therefore merits particular monitoring efforts and any new results on observables, may they stem from well-tried or new methods, are of direct importance to understand the fate of this active volcanic system.

2 Methods

2.1 Measuring 1-D profiles of CO$_2$ path amounts

The CO$_2$DIAL (Fig. 2) is an active remote sensing platform based on the differential absorption LIDAR principle (Koch et al., 2004; Amediek et al., 2008). It is a further development of the portable instrument described in Queißer et al. (2015a, 2015b). By taking the ratio of the optical powers associated with the received signals for the wavelengths coinciding with an absorption line of CO$_2$ and the wavelength at the line edge, $\lambda_{ON}$ and $\lambda_{OFF}$, respectively, one arrives at

$$\int_0^R dr \Delta \sigma(r) N_{CO_2}(r) = -\ln \left( \frac{P(\lambda_{ON})P(\lambda_{OFF})_{ref}}{P(\lambda_{OFF})P(\lambda_{ON})_{ref}} \right),$$

$$\equiv \Delta \tau$$

where $N_{CO_2}$ is the CO$_2$ number density, $R$ is the range, i.e. the distance between the instrument and the hard target, $\Delta \sigma$ is the difference between the molecular absorption cross sections of CO$_2$ associated with $\lambda_{ON}$ and $\lambda_{OFF}$, $P(\lambda)$ is the received (“science”) and $P(\lambda)_{ref}$ the transmitted optical power (“reference”). The latter is measured as a reference to normalize fluctuations of the transmitted power. The normalized optical power in Eq. (1) is referred to as grand ratio (GR),

$$GR = \frac{P(\lambda_{ON})P(\lambda_{OFF})_{ref}}{P(\lambda_{OFF})P(\lambda_{ON})_{ref}}$$

$\Delta \tau$ is the differential optical depth. The two distributed feedback (DFB) fiber seed lasers emit at $\lambda_{ON}$=1572.992 nm and $\lambda_{OFF}$=1573.173 nm (Rothman et al., 2013). To be able to easily reject background noise (such as solar background) lock-in detection is used. Consequently, both seed laser beams (for $\lambda_{ON}$ and $\lambda_{OFF}$) are amplitude modulated using two LiNbO$_3$ electro-optical modulators (EOM) at slightly different sine tones near 5 kHz and simultaneously amplified by an Erbium doped fiber amplifier (EDFA) before being transmitted. The transmitted optical power can be adjusted between ~80 mW and a maximum of 1.5 W. A glass wedge scatters a fraction of the transmitted light into an integrating sphere where the reference detector is mounted.
The transmitted light is diffusively backscattered by a hard target, which can be any surface located up to ~2000 m away from the instrument, and is received by a 200 mm diameter Schmidt-Cassegrain Telescope with a focal length of 1950 mm. Typically the received optical power is a couple of nW at a bandwidth integrated noise of ~1 pW (root mean squared noise equivalent power). The analog to digital converter (ADC) operates at 250 kSamples s⁻¹ and has a resolution of 16-bit. The integration time per scan angle was set to 4000 EOM modulation periods, which corresponds to data chunks of length 784 ms (integration time) for both science and reference channel. Each of these four chunks of data is demodulated using a digital lock-in routine following Dobler et al. (2013). After the lock-in operation one arrives at four DC signals, associated with the optical powers \( P(\lambda_{ON}) \), \( P(\lambda_{OFF}) \), \( P(\lambda_{ON})_{ref} \) and \( P(\lambda_{OFF})_{ref} \). \( \Delta \tau \) is calculated using the right hand side of Eq. (1), after taking the mean of each of the four signals. To account for the instrumental offset of \( \Delta \tau \), prior to scanning the volcanic plume, values of \( \Delta \tau \) were acquired for different \( R \) in the ambient atmosphere. The points were used to fit a calibration curve. The ordinate at \( R = 0 \) gave the instrumental offset. The calibration curve was also used to convert the measured in-plume \( \Delta \tau \) to CO₂ path amounts \( X_{CO_2}^{col} \) (in ppm.m). Column averaged CO₂ mixing ratios \( X_{CO_2,av} \) (in ppm) were obtained by dividing path amounts by \( R \). The range was measured by an onboard range finder (DLEM, Jenoptik, Germany), based on a 1550 nm LIDAR with pulse energy of 500 \( \mu \)J and accuracy <1 m. By pivoting the receiver/transmitter unit using a step motor values for \( X_{CO_2}^{col} \) (or \( X_{CO_2,av} \)) per heading were attained, and hence 1-D profiles.

The precision of the column averaged CO₂ mixing ratio was evaluated as

\[
\left( \frac{\Delta X_{CO_2,av}}{X_{CO_2,av}} \right)^2 = \text{SNR}^{-2} + \left( \frac{\sigma_R}{\langle R \rangle} \right)^2 + \delta_{\text{Speckle}}^2,
\]

with the signal-to-noise-ratio (SNR)

\[
\text{SNR} = \left[ \frac{\sigma_{GR}}{\langle GR \rangle \ln(\langle GR \rangle)} \right]^{-1},
\]

where \( \langle GR \rangle \) and \( \sigma_{GR} \) are the mean and standard deviation of the grand ratio, respectively. They were estimated from time series acquired at fixed angles in between the scans at CF. The SNR accounts for all noise sources occurring during acquisition, including instrumental noise, non-stationary baseline drift, solar background noise, atmospheric noise (mostly air turbulence) and perturbation by aerosol scattering (e.g. condensed water vapor). The second term depicts the relative range uncertainty (standard deviation of ranges \( \sigma_R \) over mean of ranges \( \langle R \rangle \)) which is typically ~1 m. The relative uncertainty due to hard target speckle was estimated as (MacKerrow et al., 1997)

\[
\delta_{\text{Speckle}} = \frac{1.22 \lambda_{OFF} R}{D \xi},
\]

where \( D \) is the spot diameter on the hard target (in m) and \( \xi \) is the dimension of the telescope field of view (in m) on the hard target.
2.2 Reconstructing a 2-D CO$_2$ concentration map

Ranges and their respective heading angles (i.e. range vectors, referred to as rays in the following) from the scans were converted to absolute Cartesian coordinates ($x$, $y$). The goal is to obtain CO$_2$ mixing ratios ($X_{CO_2}$, in ppm) at a given point ($x$, $y$). Due to the finite spatial resolution of every measurement system this will always be an average mixing ratio within a confined space, in this case a 2-D grid cell. The region of interest (area bounding the scans) was divided into grid cells with length $\Delta x$ (in $x$ direction) and $\Delta y$ (in $y$ direction). $X_{CO_2}$ were inferred from the measured $X_{CO_2}^{col}$ using an inverse technique following Pedone et al. (2014). Thereby one uses the fact that the CO$_2$ path amount is associated with the product of a range segment and a uniform CO$_2$ mixing ratio $X_{CO_2}$ along that range segment. For a given ray and for $n$ grid cells traversed by the ray this can be written as

$$\sum_{i=1}^{n} r_i X_{CO_2,i} = X_{CO_2}^{col},$$

where $r_i$ depicts the length of the ray segment in grid cell $i$ ($\sum_{i=1}^{n} r_i = R$). $X_{CO_2,i}$ is the (unknown) CO$_2$ mixing ratio within grid cell $i$ (in ppm). Including all rays, one arrives at a system of linear equations, which can be written as

$$Lc = a,$$

where $L$ is a $m \times n$ matrix, called geometry matrix, containing all $m$ rays for all $n$ grid cells, $c$ is a $n \times 1$ matrix containing the uniform $X_{CO_2}$ per grid cell and is the desired quantity to be inverted. $a$ is a $m \times 1$ matrix containing the measured (observed) $X_{CO_2}^{col}$ for each ray. For simplicity, $n_x = n_y$, where $n_x, n_y$ are the number of grid cells in $x$- and $y$-direction, respectively. Thus, $n = n_x^2$.

To invert Eq. (7) for $c$ a least square solver, the MATLAB LSQR routine, was used. The algorithm iteratively seeks values for $c$, which minimize the misfit $\|a - Lc\|$. Therefore, $c$ represents a model with a maximized likelihood of explaining the observed data $a$. By reshaping $c$ into the measurement 2-D grid a 2-D map was obtained.

2.3 CO$_2$ flux retrieval

From the inverted 2-D map of $X_{CO_2}$ the CO$_2$ flux was computed as

$$\phi_{CO_2} = 10^{-6} \ u N_{air} \ \frac{M_{CO_2}}{N_A} \ \iint_{\text{plume}} dx dy X_{CO_2,pl}(x,y)$$

where $X_{CO_2,pl}$ are the inverted, background corrected CO$_2$ mixing ratios computed as

$$X_{CO_2,pl}(x,y) = X_{CO_2}(x,y) - X_{CO_2,bg},$$

where $X_{CO_2,bg} = 380$ ppm is the background CO$_2$ mixing ratio at Solfatara measured in situ. $u$ is the magnitude of the component of the plume transport speed perpendicular to the scanned cross section (in m s$^{-1}$), $N_{air}$ is the number density of air (in m$^{-3}$), computed using meteorological data (pressure, temperature, humidity) acquired by a portable meteorological station close to the instrument. $M_{CO_2}$ is the molar mass of CO$_2$ (in kg mol$^{-1}$) and $N_A$ is Avogadro’s constant (in mol$^{-1}$).
The plume transport speed was evaluated from digital video footage acquired during the measurement, employing a video analysis program (Tracker from Open Source Physics). Condensed water vapor aerosol emitted by various vents in the region of interest was assumed to propagate with the same velocity as the volcanic CO$_2$. At a given video frame a pixel was fixed and the calibrated propagated distance (in pixels) was measured as the video proceeded. Since the frame rate of the video was known (30 frames per second), the speed by which the tracked point and hence a parcel of gas was transported could be estimated.

The relative error of the CO$_2$ flux was estimated as

$$\left( \frac{\Delta \Phi_{CO_2}}{\Phi_{CO_2}} \right)^2 = \left( \frac{\Delta u}{u} \right)^2 + \left( \frac{\int_{\text{plume}} dxdy \Delta X_{\text{CO}_2,p}(x,y)}{\int_{\text{plume}} dxdy X_{\text{CO}_2,p}(x,y)} \right)^2,$$

(10)

where $\Delta X_{\text{CO}_2,p}$ is the absolute error of the CO$_2$ mixing ratio at a given point within the integrated area and $\Delta u$ is the absolute uncertainty of the plume speed.

3 Results

The experiment took place on 4 March 2016 inside the crater of Solfatara (Fig. 1) and was focusing on the diffuse CO$_2$ release alongside the Solfatara crater edge, located south of the main vents Bocca Nuova (BN) and Bocca Grande (BG), although they were included in the scans. Elevated CO$_2$ mixing ratios, up to 1500 ppm at places, could be affirmed by means of in situ measurements using a LICOR CO$_2$ analyzer with 4% accuracy. The LICOR analyzer was measuring at the same height as the propagation height of the laser beam (ca. 2 m above ground). Due to logistical constraints the in situ measurements could only be measured the day before the experiment. Five scans were performed between 9:35 and 11:57 LT (duration 142 min) from five different locations with a total of $m = 627$ beam paths (rays), which are shown along with the respective five instrument locations in Fig 1c. It is assumed that during the complete acquisition the CO$_2$ distribution did not change (“frozen plume”).

For each scan and for each heading differential optical depths $\Delta \tau$ have been retrieved and converted into $X_{\text{CO}_2,col}$ (and $X_{\text{CO}_2,av}$), as detailed in the method section. The resulting 1-D concentration profiles are shown in Fig. 3. Numerous wiggles indicate vigorous degassing activity, suggesting diffuse degassing or CO$_2$ advected by local wind eddies. In addition, there are symmetric features, such as around $26^\circ$ in Fig. 3a, which appeared in scans carried out prior to the experiment and the day before, thus suggesting vented degassing activity. The angular scanning velocity was 2.1 mrad s$^{-1}$, associated with an angular resolution of 1.65 mrad, which corresponds to a lateral resolution of around 24 cm between points in Fig. 3.

To invert for $X_{\text{CO}_2}$, ranges and headings were converted to Cartesian coordinates. The coordinate system was chosen such that the instrument positions of all five scans were located on the y-axis (Fig. 1c). It proved to be useful to plot the measured data, i.e. $X_{\text{CO}_2,av}$ against their associated coordinates. The result (Fig. 4) is a semi-quantitative map indicating where high CO$_2$ concentrations are likely to be expected. This image therefore provides valuable a-priori information for the inversion.
The LSQR algorithm was tested using a synthetic realistic scenario. Synthetic data $X_{\text{CO}_2}^{\text{col}}$ were generated from a true model comprising of known $X_{\text{CO}_2}$ at each grid point using the real geometry matrix $L$, which contained the actual instrument positions and measured ranges. $X_{\text{CO}_2}$ of the true model were starting at 380 ppm at grid 1 and increasing by 60 ppm per increase in grid number (Fig. 5a). By running the inversion with varying number of grid cells the viable number of grid cells was found to be between $n = 4$ up to 36 without considerable loss of capability to recover the true $X_{\text{CO}_2}$ (Fig. 5b).

For $n > 36$ the inverted $X_{\text{CO}_2}$ oscillated, that is, they were over and under shooting the true $X_{\text{CO}_2}$. For the real data, however, already for $n > 16$ the inversion yielded unreasonable high $X_{\text{CO}_2}$, indicating an oscillation. The inverse problem is over determined since $m > n$, i.e., the number of beam paths traversing most of the grid cells is much higher than any practical number of grid cells usable for the inversion. Increasing the number of grid cells would reduce the number of rays traversing a given grid cell, but the problem would become highly non-linear. Generally, a viable strategy to tackle non-linearity in situations like that is a gradual introduction of non-linearity, such as by splitting up the inversion into sub-steps, using a starting model close enough to the true solution at each step (Queißer et al., 2012). With each increase in sub-step, the starting model contains more small-scale information. This approach was tested in the real data inversion. Starting with $n = 4$ grid cells, the inversion result was interpolated, smoothed and used as the starting model for the inversion with $(n_x + 1)^2$ grid cells. At $n = 25$ the location of the peak $X_{\text{CO}_2}$ were in strong disagreement with the LICOR data, indicating that the inversion was trapped in a local minimum. A similar outcome was obtained by reducing the number of rays used for the inversion (using every 2nd up to every 10th ray).

That left $n = 16$ the maximum feasible number of grid cells for a robust inversion. The resulting grid length was $\Delta x = 38$ m and $\Delta y = 33$ m. As for the synthetic tests, a constant $X_{\text{CO}_2}$, the mean of the raw data (Fig. 4), was used as a starting model. The inversion result is shown in Fig. 6a. To increase spatial resolution the inverted model was interpolated onto a grid with grid spacing $\Delta x/8$ and $\Delta y/8$ using ordinary Kriging interpolation (Oliver, 1996). The result is shown in Fig. 6b. Overlaying the 2-D map of CO$_2$ mixing ratios with the map of Solfatara reveals a zone of increased anomalous CO$_2$ degassing activity along the southeastern edge of Solfatara, which is in reasonable agreement with in situ data from the LICOR CO$_2$ analyzer (Fig. 6c).

The resulting 2-D map of CO$_2$ mixing ratios was used to compute the CO$_2$ flux. Since zones with poor ray coverage were prone to inversion artifacts (see Fig. 4c) zones without ray coverage were excluded from integration. The plume transport speed was estimated to be $1.1 \pm 0.2$ m s$^{-1}$. The plume speed uncertainty was retrieved from the standard deviation of various plume speeds retrieved from different tracks carried out across the plume. To estimate the flux uncertainty (Eq. 10), a constant $\Delta X_{\text{CO}_2,pt} = \max(\Delta X_{\text{CO}_2,av})$ was considered (maximum error of all five scans). Using Eq. (8) the resulting CO$_2$ flux was computed as $12.8 \pm 3.3$ kg s$^{-1}$ (± 1 SD) or $1106 \pm 288$ tons day$^{-1}$.
4 Discussion

The retrieved 2-D map (Fig. 6c) indicates an elongated zone of intense anomalous degassing along the eastern edge of the Solfatara crater. Encouragingly, this is a persistent feature in different inversions performed with different number of grid cells and beam paths (and thus degree of non-linearity) and underpins that it is real. Previous measurements sampling the Solfatara area with accumulation chambers yielded an increased anomalous CO$_2$ degassing activity in the corresponding area too (Granieri et al., 2010; Tassi et al., 2013; Bagnato et al., 2014). The retrieved elongated zone of anomalous CO$_2$ degassing likely encompasses at least two major vents (Fig. 6c). The locations of the peaks in CO$_2$ mixing ratio in Fig. 6c fairly agree with the 1-D input data. For instance, the peak near the center of the crater corresponds to the peak near 26° in the first scan in Fig. 3a. The second scan (Fig. 3b) indicates a rather abrupt decrease in $X_{CO_2,av}$ at 28°, in line with the edge of the zone of elevated CO$_2$ concentrations at the crater center (Fig. 6c). This central degassing feature is coherent with results of recent campaigns (Granieri et al., 2010; Tassi et al., 2013; Bagnato et al., 2014). The symmetric increase in $X_{CO_2,av}$ near 9° in Fig. 3d corresponds to the position of the local peak in $X_{CO_2}$ between in situ points 7 and 8 in Fig. 6c. Provided sufficient ray coverage and angle diversity, which is the case for the zones away from the edges of the 2-D map, disagreement between the peaks in the 1-D data (Fig. 3) and those in the 2-D map (Fig. 6) are likely due to physical fluctuations in CO$_2$ concentration. The plume was assumed to be “frozen”, but the measurement duration of 142 min was certainly larger than the time scale of alterations in the dispersion pattern of the plume. During acquisition one could visually identify at least 5 small vents emitting water vapor and therefore most likely also CO$_2$. Though not recovered due to the limited spatial resolution of the inversion this advocates that there are in fact separate vents south of the main vents, near the edge of the Solfatara crater.

Retrieved $X_{CO_2}$ peak near 1300 ppm (2 m above ground), in line with the in situ LICOR data, although not spatially matching them in places. Again, this can be explained by the fact that the in situ values were acquired the day before so that local wind and thus dispersion patterns were different. Nevertheless, both the LICOR in situ data and the inversion result indicate high $X_{CO_2}$ near the main vents and along the crater edge. Near the main vents highest CO$_2$ mixing ratios in the 2-D map are located ca. 20 m west of BN. In fact, the whole zone of high $X_{CO_2}$ is shifted 20 m northwest from where one would expect it. Since the predominant wind direction at the time of acquisition was around 300°, to first order one would expect the CO$_2$ to disperse rather towards southeast, along the crater edge. The main vent area was at the edge of the scanned area. Note that the relative inversion residual $\|a - Lc\|/\|a\|$ was 0.18, which means on average 18% of $X_{CO_2,av}$ are unexplained by the model in Fig. 6a. This mismatch is therefore likely due to poor ray coverage and angle diversity for the zone containing the main vents, since the acquisition focused on the zone south of the main vents. Possibly, but less likely, CO$_2$ was advected slightly towards west due to dispersive mechanism related to local wind eddies decoupled from the main wind direction. These dispersive mechanisms take place in any case and make a distinction between CO$_2$ from the main vents and the surrounding diffuse degassing challenging. For that reason, in future acquisitions at that site the region of interest shall be scanned from instrument positions aligned along a half circle around the zone rather than using a “flat” scan geometry as chosen here.
For a comparison, CO$_2$ fluxes were computed directly from the 1-D profiles, that is, similar to Eq. (8) but using path amounts, ignoring any heterogeneity in the CO$_2$ distribution. The average flux of all five scans (1055 ± 389 tons/day) is in good agreement with the result obtained from the 2-D map (1106 ± 288 tons/day). Note that disagreement with the flux result from the 2-D map may partly be due to the frozen plume assumption, since this assumption is better fulfilled for the acquisition of a single 1-D profile, which takes much less time. Future scans shall thus be acquired with higher scan velocity or from further away.

Yet, both the CO$_2$ flux from the 2-D map and from the 1-D profiles are higher than fluxes previously estimated. To our knowledge, all former studies except one (Pedone et al., 2014) inferred $X_{CO_2}$ and hence CO$_2$ fluxes from a grid of point measurements, which may have missed degassing activity in between the measurement points and so tended to yield lower flux values. Spatially comprehensive sounding by Pedone et al. (2014) resulted in a CO$_2$ flux of only ~300 tons/day in early 2013, however, it focused on the area around the Solfatara main vents, that is, 8000 m$^2$. In this study the area considered for flux computation was over 21000 m$^2$. The average degassing rate at Solfatara has been increasing by ~9% each year over the past 10 years or so (Chiodini et al., 2010; d’Auria, 2015). Extrapolating the 300 tons/day would yield a flux of 390 tons/day in early 2016. Integrating CO$_2$ mixing ratios of the area around the main vents only (bounded to the south by in situ point 6, Fig. 6c) yields a flux of 399 ± 104 tons/day of CO$_2$, in excellent agreement with the extrapolated flux. However, as mentioned before, CO$_2$ from the main vents mixes with surrounding volcanic CO$_2$ and furthermore the scans focused on the area south of the main vents (poor ray coverage at BN and BG). So this value should be interpreted with care. It deems to be reasonable to exclude the zone of high anomalous degassing in the north of the 2-D map, which leads to a flux of 675 ± 175 tons/day, representing any degassing activity (vented and diffuse) within the investigated area, excluding the main vent area (BN and BG). This magnitude equals roughly 45% of the total CO$_2$ flux of the DDS (diffuse degassing structure) reported by Granieri et al. (2003), 13 years prior to this study.

All five scans were performed one-sided, i.e. from a single half space, as often the case in geophysical tomography problems (e.g. Hobro et al., 2003). This is not ideal for any inversion technique as it makes the inverse problem highly non-linear with a non-unique solution, meaning that many models may explain the observed data equally well. However, for Solfatara there is an abundance of hard data available, which extremely facilitated the rejection of unlikely models. This case therefore enabled to demonstrate that one may obtain useful tomographic results from one-sided scanning of a degassing feature. The inverted model is missing small-scale features, since to linearize the inversion the grid spacing had to be rather coarse. Yet, given the fair agreement with the hard data, the inverted 2-D model (Fig. 6c) is quantitatively sound and outlines the geometry of the diffuse degassing probed at Solfatara. Future measurements of this type at Solfatara are envisaged, including a more systematic study, using a wider variety of viewing angles, which will allow a more quantitative picture as to which extent this method is useful for one-sided tomography of highly non-isotropic volcanic CO$_2$ plumes. In particular, we expect an enlarged angle diversity to increase the maximum number of grids usable for stable inversion, boosting 2-D resolution. The outcome indicates this method to be particularly useful for future measurement campaigns using hard target DIAL to scan volcanic plumes from an aircraft or similar acquisition geometries sensing other types of gas emission.
5 Conclusions

As magmatic CO$_2$ degassing rates are tracers for the dynamics and chemistry of the magma plumbing system beneath Campi Flegrei and at volcanic areas in general, a comprehensive quantification of magmatic CO$_2$ degassing strength is of interest for volcanology and of vital importance for civil protection.

Scanning hard target DIAL measurements have been performed at Solfatara crater (Campi Flegrei, Italy), which allowed an inclusive measurement of CO$_2$ amounts in the form of 1-D profiles of CO$_2$ path amounts. From the 1-D profiles a 2-D map of CO$_2$ mixing ratios has been reconstructed outlining the main CO$_2$ distribution. Such a map is useful to geometrically correct the CO$_2$ flux obtained from 1-D concentration profiles for heterogeneous CO$_2$ distribution. Since it was in line with in situ hard data, the 2-D map was directly used to retrieve the CO$_2$ flux, which is compatible with previous results.

The 1-D profiles have been acquired from a single half space, which indicates this tomography method to be beneficial for scanning strongly non-isotropic CO$_2$ distributions, such as diffuse emissions, that can be viewed from limited angles only. To fully assess the potential of this method, future acquisitions should involve different scanning geometries, potentially allowing for an enhanced resolution of the 2-D map and thus more accurate gas flux estimation.

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References


Figures

Figure 1: Geography and measurement geometry. (a) Location of the Solfatara crater as part of the volcanic area of Campi Flegrei, near Naples (Italy). (b) Nadir photo of Solfatara crater. The rectangle contains the region of interest. (c) Zoom of area outlined by the rectangle depicting the five instrument positions P1 to P5 with the following UTM-coordinates: P1: (427476, 4519921), P2: (427485, 4519935), P3: (427495, 4519949), P4: (427507, 4519967), P5: (427520, 4519986). Also shown are the respective range vectors (rays) for all five scans and the numbered locations of the LICOR measurements. (d) Photo taken during the scan at P5 looking towards east. The largest clouds of condensed water aerosol appeared near the main vents (Bocca Nuova, BN and Bocca Grande, BG) on the left. The CO₂DIAL, visible in the lower right corner, comprised of the tripod carrying the telescope (with transmitter unit) and the main unit (red box).
Figure 2: Scheme of the CO$_2$ DIAL as used for this experiment. EOM: Electro-optical modulator, DLEM: range finder module, EDFA: Erbium doped fiber amplifier. ADC: analog-to-digital converter, DAC: digital-to-analog converter. The CO$_2$ cell is used to calibrate the seed laser wavelengths. To minimize hard target and turbulence related speckle noise the collimator used had a relatively high divergence of 1.7 mrad while the telescope field of view was 1.5 mrad. For mechanical reasons the optical band pass filter was mounted before the collimating lens. The change in transmission spectrum can be neglected.
Figure 3: 1-D profiles of $X_{\text{CO}_2, \text{av}}$, the total (not background corrected) CO$_2$ mixing ratios, derived by dividing the path amounts $X_{\text{CO}_2}^{\text{col}}$ (ppm.m) per angle by the associated range. Each value therefore represents a column-averaged concentration. Each point corresponds to 784 ms integration time. For each profile and heading ranges are indicated by the red dashed line. (a) Profile acquired between 9:35:36 and 9:41:54. (b) Profile acquired between 10:04:08 and 10:10:54. (c) Profile acquired between 10:31:24 and 10:37:28. (d) Profile acquired between 11:01:46 and 11:07:46. (e) Profile acquired between 11:50:39 and 11:57:15. The grey envelope depicts precision (1 SD, Eq. 3).
Figure 4: Contour plot of $X_{\text{CO}_2,\text{av}}$ ($X_{\text{CO}_2}^{\text{col}}$ divided by the range) for all 627 beam paths. Also shown are the instrument positions (squares on y-axis) starting with P1 at $y=20$ m. The data has been regridded on a regular grid of $90 \times 90$ points using natural interpolation. One would expect high anomalous CO$_2$ mixing ratios near the main vents (BN, BG near $x=120$ m, $y=140$ m) and the southern part of the area. Low anomalous CO$_2$ mixing ratios are to be expected in the northwestern part. Note that due to the abundance of data some data points were masking each other. They were thus averaged, leading to a maximum mixing ratio lower than actually observed (e.g. in Fig. 3b).
Figure 5: Synthetic inversion result with $n = 16$ grid cells. (a) True model used to generate synthetic column averaged $X_{CO_2}^{col}$. Each grid cell is identified by a grid number. The dotted line outlines the ray coverage. The instrument positions are indicated. (b) Inverted model. (c) True and inverted $X_{CO_2}$ versus grid cell. The inverted $X_{CO_2}$ for grid 13 is off since the ray coverage associated with that area was poor.
Figure 6: Retrieved 2-D model of $X_{CO_2}$. (a) Inverted model of $X_{CO_2}$. (b) Inverted $X_{CO_2}$ in after ordinary Kriging interpolation. The ray coverage is depicted by the dotted line. (c) $X_{CO_2}$ superposed onto nadir photo of Solfatara for those grid cells covered by the rays. Also shown are the $X_{CO_2}$ from in situ measurements (measurement points 3 to 10) using the LICOR CO$_2$ analyzer. Note that the in situ values had been acquired a day before the scans and thus serve as an approximate reference only.