Towards standardized processing of eddy covariance flux measurements of carbonyl sulfide

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Abstract. Carbonyl sulfide (COS) flux measurements with the eddy covariance (EC) technique are becoming popular for estimating gross primary productivity. To compare COS flux measurements across sites, we need standardized protocols for data processing. In this study, we analyze how various data processing steps affect the calculated COS flux, how they differ from carbon dioxide (CO_2) flux processing steps and we provide a method for gap-filling COS fluxes. Different methods for determining the time lag between COS mixing ratio and the vertical wind velocity (w) resulted in a maximum of 15.9 % difference in the median COS flux over the whole measurement period. Due to limited COS measurement precision, small COS fluxes (below approximately 3 pmol m^{-2} s^{-1}) could not be detected when the time lag was determined from maximizing the covariance between COS and w. The difference between two high-frequency spectral corrections was 2.7 % in COS flux calculations, whereas omitting the high-frequency spectral correction resulted in a 14.2 % lower median flux and different detrending methods caused a spread of 6.2 %. Relative total uncertainty was more than five times higher for low COS fluxes (lower than ± 3 pmol m^{-2}s^{-1}) than for low CO_2 fluxes (lower than ± 1.5 μmol m^{-2}s^{-1}), indicating a low signal-to-noise ratio of COS fluxes. Due to similarities in ecosystem COS and CO_2 exchange we recommend applying storage change flux correction and friction velocity filtering as usual in EC flux processing, but due to the low signal-to-noise ratio of COS fluxes, we recommend using CO_2 data for time lag and high-frequency corrections of COS fluxes, due to higher signal-to-noise ratio of CO_2 measurements.

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1 Introduction

Carbonyl sulfide (COS) is the most abundant sulfur compound in the atmosphere, with tropospheric mixing ratios around 500 ppt (Montzka et al., 2007). During the last decade, studies on COS have grown in number, mainly motivated by the use of

COS exchange as a tracer for photosynthetic carbon uptake (also known as gross primary productivity, GPP) (Sandoval-Soto et al., 2005; Blonquist et al., 2011; Asaf et al., 2013; Whelan et al., 2018). COS shares the same diffusional pathway in leaves as carbon dioxide (CO_2), but in contrast to CO_2, COS is destroyed completely by hydrolysis and is not emitted. This one-way flux makes it a promising proxy for GPP (Asaf et al., 2013; Yang et al., 2018; Kooijmans et al., 2019).

Eddy covariance (EC) measurements are the backbone of gas flux measurements at the ecosystem scale (Aubinet et al., 2012). Protocols for instrument setup, monitoring and data processing have been recently harmonized for CO_2 (Rebmann et al., 2018; Sabbatini et al., 2018) as well as for methane (CH_4) and nitrous oxide (N_2O) (Nemitz et al., 2018) within the Integrated Carbon Observation System (ICOS) flux stations (Franz et al., 2018). EC data processing chain includes despiking and filtering raw data, rotation of the coordinate system to align with the prevailing streamlines, determining the time lag between the sonic anemometer and the gas analyzer signals, trend removal to separate the turbulent fluctuations from the mean trend, calculating covariances and correcting for flux losses at low and high frequencies. After processing, fluxes are quality filtered and flagged according to atmospheric turbulence characteristics and stationarity.

Studies on ecosystem COS flux measurements with the EC technique are still limited (Asaf et al., 2013; Billesbach et al., 2014; Maseyk et al., 2014; Commane et al., 2015; Gerdel et al., 2017; Wehr et al., 2017; Yang et al., 2018; Kooijmans et al., 2019; Spielmann et al., 2019) and there is no standardized flux processing protocol for COS EC fluxes. Table 1 summarizes the processing steps used in earlier studies. Most studies do not report all necessary steps, and in particular, often ignore the storage change correction. COS EC flux measurements and data processing have similarities with other trace gases (e.g., CH_4 and N_2O) that often have low signal-to-noise ratios, especially regarding time lag determination and frequency response corrections. Time lag determination is essential for aligning wind and gas concentration measurements to minimize biases in flux estimates. Frequency response corrections, on the other hand, are needed for correcting flux underestimation due to signal losses both at high and low frequencies (Aubinet et al., 2012). Unlike for CH_4 or N_2O, there are no sudden bursts or sinks expected for COS, and in that sense some processing steps for COS are more like those for CO_2 (e.g., despiking, storage change correction and friction velocity (u_*) filtering). Gerdel et al. (2017) describes the issues of different detrending methods, high-frequency spectral correction, time lag determination and u_* filtering. However, there has not been any study comparing different methods for time lag determination or high-frequency spectral correction in terms of their effects on COS fluxes. This weakens our ability to assess uncertainties in COS flux measurements.

In this study, we compare different methods for detrending, time lag determination and high frequency spectral correction. In addition, we compare two methods for storage change flux calculation, discuss the nighttime low turbulence problem in the context of COS EC measurements, introduce a method for gap-filling COS fluxes for the first time and discuss the most important sources of random and systematic errors. Through the evaluation of these processing steps, we aim to settle on a set of recommended protocols for COS flux calculation.

Table 1. Processing steps used in previous COS eddy covariance studies. Detrending methods include linear detrending (LD), block averaging (BA), and recursive filtering (RF). Spectral corrections include an experimental method and a theoretical method by Moore (1986). Processing steps not specified in the original articles are marked with a hyphen. The last row summarizes the recommended processing options in this study.

Reference	Sampling frequency	Detrending	Time lag	Spectral corrections	Storage change correction	u_* filtering
Asaf et al. (2013)	1 Hz	LD	$\max \overline{|w'\chi'_{COS}|}$	Exp. method	-	-
Billesbach et al. (2014)	10 Hz	-	$\max \overline{|w'\chi'_{COS}|}$	Moore (1986)	-	$u_* > 0.15$ ms^{-1}
Maseyk et al. (2014)	10 Hz	-	$\max \overline{|w'\chi'_{COS}|}$	Moore (1986)	-	$u_* > 0.15$ ms^{-1}
Commane et al. (2015)	4 Hz	BA	$\max \overline{|w'\chi'_{H2O}|}$	-	Neglected	$u_* > 0.17$ ms^{-1}
Gerdel et al. (2017)	10 Hz	BA/ LD/ RF	$\max \overline{|w'\chi'_{COS}|}$	Exp. method	EC meas.	$u_* > 0.12$ ms^{-1}
Kooijmans et al. (2017)	10 Hz	LD	$\max \overline{|w'\chi'_{CO2}|}$	Exp. method	Profile meas.	$u_* > 0.3$ ms^{-1}
Wehr et al. (2017)	4 Hz	BA	-	Exp. method, CO_2 spectrum	Profile meas.	$u_* > 0.17$ ms^{-1}
Yang et al. (2018)	1 Hz	-	-	-	Neglected	-
Kooijmans et al. (2019)	10 Hz	LD	$\max \overline{|w'\chi'_{CO2}|}$	Exp. method	Profile meas.	$u_* > 0.3$ ms^{-1}
Spielmann et al. (2019)	5 or 10 Hz	LD	$\max \overline{|w'\chi'_{COS}|}$	Exp. method	-	-
Recommendation by this study	10 Hz	Site specific	$\max \overline{|w'\chi'_{CO2}|}$	Exp. method, CO_2 spectrum	Profile meas.	Site specific threshold

2 Materials and Methods

In this study we used COS and CO_2 EC flux datasets collected at the Hyytiälä ICOS station in Finland from 26 June to 2 November 2015. The site has a long history of flux and concentration observations (Hari and Kulmala, 2005) and a COS analyzer was introduced to the site in March 2013. In this section, we describe methods used in the reference and alternative data processing schemes.

2.1 Site description

Measurements were made in a boreal Scots pine (*Pinus sylvestris*) stand at the Station for Measuring Forest Ecosystem-Atmosphere Relations (SMEAR II) in Hyytiälä, Finland (61°51′ N, 24°17′ E, 181 m above sea level). The Scots pine stand was established in 1962 and reaches at least 200 m to all directions and about 1 km to the north (Hari and Kulmala, 2005). The site is characterized by modest height variation and an oblong lake is situated about 700 m to the southwest of the forest station (Rannik, 1998; Vesala et al., 2005). Canopy height was 17 m and the all-sided leaf area index (LAI) was approximately 8 m^{2}m^{-2} in 2015. EC measurements were done at 23 m height. Sunrise time varied from 2:37 am in June to 7:55 am in November, while sunset was at 10:14 pm in the beginning and 4:17 pm in the end of the measurement period. All results are presented in Finnish winter time (UTC +2) and nighttime is defined as periods when photosynthetic photon flux density PPFD < 3 μmol m^{-2}s^{-1}.

2.2 EC measurement setup

The EC setup consisted of an ultrasonic anemometer (Solent Research HS1199, Gill Instruments Ltd., England, UK) for measuring wind speed in three dimensions and sonic temperature, an Aerodyne quantum cascade laser spectrometer (QCLS) (Aerodyne Research Inc., Billerica, MA, USA) for measuring COS, CO_2 and water vapor (H_2O) mole fractions and an LI-6262 infrared gas analyzer (Licor, Lincoln, NE, USA) for measuring H_2O and CO_2 mole fractions. All measurements were recorded at 10 Hz frequency and were made with a flow rate of approximately 10 liters per minute (LPM) for the QCLS and 14 LPM for LI-6262, respectively. The PTFE sampling tubes were 32 m and 12 m long for QCLS and LI-6262, respectively, and both had an inner diameter of 4 mm. Two PTFE filters were used upstream of the QCLS inlet to prevent any contaminants from entering the analyzer sample cell: one coarse filter (0.45 μm, Whatman), followed by a finer filter (0.2 μm, Pall corporation), at approximately 50 cm distance to the analyzer inlet. The Aerodyne QCLS used an electronic pressure control system to control the pressure fluctuations in the sampling cell. The QCLS was run at 20 Torr sampling cell pressure. An Edwards XDS35i scroll pump (Edwards, England, UK) was used to pump air through the sampling cell, while LI-6262 had flow control by a LI-670 flow control unit.

Background measurements of high-purity nitrogen (N_2) were done every 30 min for 26 s to remove background spectral structures in the QCLS (Kooijmans et al., 2016). In addition, a calibration cylinder was sampled each night at 00:00:45 for 15 s. The calibration cylinder consisted of COS at 429.6 $\pm$ 5.6 ppt, CO_2 at 408.37 $\pm$ 0.05 ppm and CO at 144.6 $\pm$ 0.2 ppb. The cylinder was calibrated against the NOAA-2004 COS scale, WMO-X2007 CO_2 scale and WMO-X2004 CO scale using cylinders that were calibrated at the Center for Isotope Research of the University of Groningen in the Netherlands (Kooijmans et al., 2016). The standard deviation calculated from the cylinder measurements was 19 ppt for COS mixing ratios and 1.3 ppm for CO_2 at 10 Hz measurement frequency.

It has previously been shown that water vapor in the sample air can affect the measurements of COS through spectral interference of the COS and H_2O absorption lines (Kooijmans et al., 2016). This spectral interference was corrected for by fitting the COS spectral line separately from the H_2O spectral line.

The computer embedded in the Aerodyne QCLS and the computer that controlled the sonic anemometer and logged the LI-6262 data were synchronized once a day with a separate server computer. Analogue data signals from LI-6262 were acquired by the Gill anemometer sensor input unit, which digitised the analogue data and appended it to the digital output data string. Digital Aerodyne data were collected to the same computer with a serial connection and recorded in separate files with home-made software (COSlog).

2.3 Profile measurements

Atmospheric concentration profiles were measured with another Aerodyne QCLS at a sampling frequency of 1 Hz. Air was sampled at 5 heights: 125 m, 23 m, 14 m, 4 m and 0.5 m. A multi-position Valco valve (VICI, Valco Instruments Co. Inc.) was used to switch between the different profile heights and calibration cylinders. Each measurement height was sampled for 3 min each hour. One calibration cylinder was measured twice for 3 min each hour to correct for instrument drift, and two

other calibration cylinders were measured once for 3 min each hour to assess the long-term stability of the measurements. A background spectrum was measured once every six hours using high-purity nitrogen (N 7.0) (for more details, see Kooijmans et al. (2016)). The overall uncertainty of this analyzer was determined to be 7.5 ppt for COS and 0.23 ppm for CO_2 at 1 Hz frequency (Kooijmans et al., 2016). The measurements are described in more detail in Kooijmans et al. (2017).

2.4 Eddy covariance fluxes

In this section, we describe the processing steps of EC flux calculation from raw data handling to final flux gap-filling and uncertainties. Fig. 1 provides a graphical outline of all processing steps. The different processing options presented here are applied and discussed in Sect. 3. In the next section, the different processing schemes are compared to a "reference scheme", which consists of linear detrending, planar fit coordinate rotation, using CO_2 time lag for COS and experimental spectral correction. A subset of the data – nighttime fluxes processed with the reference scheme – was published in Kooijmans et al. (2017).

2.4.1 Pre-processing

For flux calculation, the sonic anemometer and gas analyzer signals need to be synchronized. This is particularly relevant for fully digital systems where digital data streams are gathered from different instruments that can be completely asynchronous to each other (Fratini et al., 2018). The following procedure was used to combine two data files of 30 min length (of which one includes sonic anemometer and LI-6262 data and the other includes Aerodyne QCLS data): 1) the cross-covariance of the two CO_2 signals (QCLS and LI-6262) was calculated 2) the QCLS data were shifted so that the cross-covariance of the CO_2 signals was maximized. Note that this will result in having the same time lag for QCL and LI-6262. The time shift between the two computers was a maximum of 10 seconds, with most varying between 0 s to 2 s during one day. It is also possible to shift the time series by maximizing the covariance of CO_2 and w, which will then already account for the time lag (Fig. S1) or combine files according to their time stamps and allow a longer window in which the time lag is searched. However, in this case it is important that the time lag (and computer time shift) is determined from CO_2 measurements only, as using COS data might result in several covariance peaks in longer time frames due to low signal-to-noise ratios and small fluxes.

Raw data were then despiked so that the difference between subsequent data points was maximum 5 ppm for CO_2, 1 mmol mol^{-1} for H_2O, 200 ppt for COS and 5 ms^{-1} for w. After despiking, the missing values were gap-filled by linear interpolation.

We used the planar fit method to rotate the coordinate frame so that the turbulent flux divergence is as close as possible to the total flux divergence. In this method, $\bar{w}$ is assumed to be zero only on longer time scales (weeks or even longer). A mean streamline plane is fitted to a long set of wind measurements. Then the z-axis is fixed as perpendicular to the plane and $\bar{v}$ wind component to be zero (Wilczak et al., 2001). In addition, we used the 2D coordinate rotation for coordinate rotation in two processing schemes to determine the flux uncertainty that is related to flux data processing (Sect. 2.4.6). First, the average u-component was forced to be along the prevailing wind direction. The second rotation was performed to force the mean vertical wind speed ($\bar{w}$) to be zero (Kaimal and Finnigan, 1994). In this way, the x-axis is parallel and z-axis perpendicular to the mean flow. While 2D coordinate rotation is the most commonly used rotation method, the planar fit method brings benefits
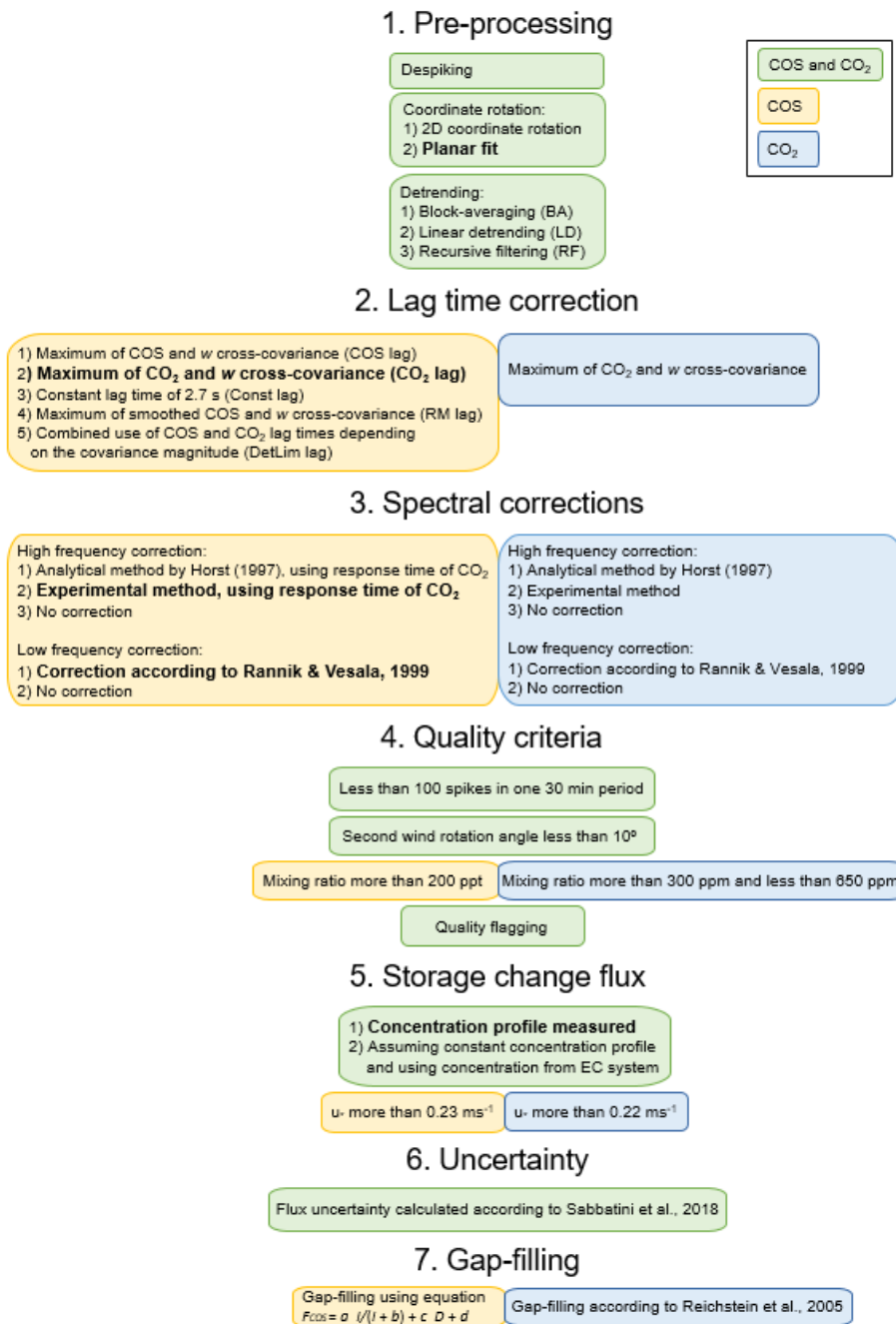
Figure 1. Different EC processing steps summarized. Yellow boxes refer to steps only used for COS data processing, blue boxes to steps used only to CO_2 data and green boxes to steps that are relevant for both gases. Recommended options are written in bold. Options that are used in the reference processing scheme for COS in this study are: planar fit coordinate rotation, linear detrending, CO_2 time lag, experimental high-frequency correction, low-frequency correction according to Rannik and Vesala (1999) and storage change flux from measured concentration profile. The abbreviations most commonly used throughout the text are written in parenthesis.

especially in complex terrain (Lee and Finnigan, 2004) and is nowadays recommended as the preferred coordinate rotation method (Sabbatini et al., 2018).

To separate the mixing ratio time series into mean and fluctuating parts, we tested three different detrending options: 1) 30 min block averaging (BA), 2) linear detrending (LD) and 3) recursive filtering (RF) with a time constant of 30 s. BA is the most commonly used method for averaging the data with the benefit of dampening the turbulent signal the least. On the other hand, BA may lead to an overestimation of the fluctuating part (and thus overestimating the flux), for example due to instrumental drift or large scale changes in atmospheric conditions, that are not related to turbulent transfer (Aubinet et al., 2012). The LD method fits a linear regression to the averaging period and thus gets rid of instrumental drift and to some extent weather changes, but may lead to underestimation of the flux if the linear trend was related to actual turbulent fluctuations in the atmosphere. The third method, RF, uses a time window (here 30 s) for a moving average over the whole averaging period. RF brings the biggest correction and thus lowest flux estimate compared to other methods, but effectively removes biased low-frequency contributions to the flux. An Allan variance was determined for a time period when the instrument was sampling from a gas cylinder (Werle, 2010). The time constant of 30 s for RF was determined from the Allan plot (Fig. S4), as the system starts to drift in non-linear fashion at 30 s, following the approach suggested by Mammarella et al. (2010). The effect of different detrending methods is shown and discussed in Sect. 3.

2.4.2 Time lag determination

The time lag between w and COS signals was determined using the following five methods:

1) From the maximum difference of the cross-covariance of the COS mixing ratio and w ($\overline{w'\chi'_{COS}}$) to a line between covariance values at the lag window limits (referred hereafter as COS_{lag}). This applies also to other covariance methods explained below, and prevents the time lag to be exactly at the lag window limits. Lag window limits (form 1.5 s to 3.8 s) were determined based on the nominal time lag of 2.6 s calculated from the flow rate and tube dimensions. More flexibility was given to the upper end of the lag window as time lags have been found to be longer than the nominal time lag (Massman, 2000; Gerdel et al., 2017).

2) From the maximum difference of the cross-covariance of the CO_2 mixing ratio and w ($\overline{w'\chi'_{CO_2}}$) to a line between covariance values at the lag window limits within the lag window 1.5–3.8 s (referred hereafter as $CO2_{lag}$).

3) Using a constant time lag of 2.6 s, which was the nominal time lag and the most common lag for CO_2 with our setup (referred hereafter as $Const_{lag}$).

4) From the maximum difference of the smoothed $\overline{w'\chi'_{COS}}$ to a line between covariance values at the lag window limits. The cross-covariance was smoothed with a 1.1 s running mean (referred hereafter as RM_{lag}). The averaging window was chosen so that it provided a more distinguishable covariance maximum while still preventing a shift in the timing of the maximum.

5) A combination of COS_{lag} and $CO2_{lag}$. First, the random flux error due to instrument noise was calculated according to Mauder et al. (2013):

$$RE = \sqrt{\frac{(\sigma_c^{noise})^2 \sigma_w^2}{N}} \tag{1}$$

where instrumental noise σ_c^{noise} was estimated from the method proposed by Lenschow et al. (2000), σ_w is the standard
deviation of the vertical wind speed and N the number of data points in the averaging period. The random error was then
compared to the raw maximum covariance. If the maximum covariance was higher than three times the random flux error,
then the COS_{lag} method was used for time lag determination. If the covariance was dominated by noise (the covariance being
smaller than three times the random error) or COS_{lag} was at the lag window limit, then the $CO2_{lag}$ lag method was selected, as
proposed in (Nemitz et al., 2018) (referred hereafter as the $DetLim_{lag}$).

2.4.3 Frequency response correction

Some of the turbulence signal is lost both at high and low frequencies due to losses in sampling lines, inadequate frequency
response of the instrument and inadequate averaging times, among other reasons (Aubinet et al., 2012). In this section, we
describe both high and low frequency loss corrections in detail. We tested two high-frequency correction methods, described
below, simultaneously correcting for low frequency losses. One run was performed without both low-frequency and high-
frequency response corrections.

High-frequency correction

Especially in closed-path systems the high frequency turbulent fluctuations of the target gas dampen at high frequencies due
to long sampling lines. Other reasons for high-frequency losses include sensor separation and inadequate frequency response
of the sensor. In turn, high-frequency losses cause the normalized co-spectrum of the gas with w to be lower than expected at
high frequencies, resulting in lower flux. The flux attenuation factor (FA) for a scalar s is defined as

$$FA = \frac{\overline{w's'}_{meas}}{\overline{w's'}} = \frac{\int_0^\infty T_{ws}(f)Co_{ws}(f)df}{\int_0^\infty Co_{ws}(f)df} \tag{2}$$

where $\overline{w's'_{meas}}$ and $\overline{w's'}$ are the measured and unattenuated covariances, respectively, $T_{ws}(f)$ is the net transfer function,
specific to the EC system and scalar s, and $Co_{ws}(f)$ is the cospectral density of the scalar flux $\overline{w's'}$. For solving FA, a
cospectral model and transfer function are needed. In this study, we tested the effect of high-frequency spectral correction by
applying either an analytical correction for high-frequency losses (Horst, 1997) or an experimental correction (Aubinet et al.,
2000). The analytical correction was based on scalar co-spectra defined in Horst (1997) and the experimental approach was
based on the assumption that temperature co-spectrum is measured without significant error and the normalized scalar co-
spectra were compared to the normalized temperature co-spectrum (Aubinet et al., 2000; Wohlfahrt et al., 2005; Mammarella
et al., 2009).

In the analytical approach by Horst (1997) the integral in Eq. 2 is solved analytically by assuming a model cospectrum of
the form $\frac{fCo_{ws}(f)}{w's'} = \frac{2}{\pi}\frac{f/f_m}{1+(f/f_m)^2}$ and a transfer function $T_{ws}(f) = \frac{1}{1+(2\pi f\tau_s)^2}$. The flux attenuation then results to

$$FA_H = [1 + (2\pi\tau_s f_m)^\alpha]^{-1} \tag{3}$$

where $\alpha = 7/8$ for neutral and unstable stratification and $\alpha = 1$ for stable stratification in the surface layer, τ_s is the overall EC system time constant and f_m is the frequency of the logarithmic co-spectrum peak estimated from $f_m = \dfrac{n_m \bar{u}}{z_m - d}$, where n_m is the normalized frequency of the co-spectral peak, $\bar{u}$ the mean wind speed, z_m the measurement height and d the displacement height. The normalized frequency of the co-spectral peak n_m is dependent on atmospheric stability $\zeta = \dfrac{z_m - d}{L}$ (Horst, 1997):

$$n_m = 0.085, \text{ for } \zeta \leq 0 \qquad (4)$$

$$n_m = 2.0 - 1.915/(1 + 0.5\zeta), \text{ for } \zeta > 0 \qquad (5)$$

where L is the Obukhov length (Fig. S8). The model cospectrum for the analytical high-frequency spectral correction was adapted from Kaimal et al. (1972) and given as

$$\frac{fCo_{w\theta}(n)}{\overline{w'\theta'}} = \begin{cases} \frac{1.05n/n_m}{(1+1.33n/n_m)^{7/4}} & \text{for } \zeta \leq 0, \ n < 1 \\ \frac{0.387n/n_m}{(1+0.38n/n_m)^{7/3}} & \text{for } \zeta \leq 0 \ n \geq 1 \\ \frac{0.637n/n_m}{(1+0.91n/n_m)^{2.1}} & \text{for } \zeta > 0 \end{cases} \qquad (6)$$

In the experimental approach, we solved the Eq. 2 numerically and used the fitting of the measured temperature cospectra to define a site-specific scalar cospectral model (De Ligne et al., 2010) as

$$\frac{fCo_{w\theta}(n)}{\overline{w'\theta'}} = \begin{cases} \frac{10.36n/n_m}{(1+4.82n/n_m)^{3.05}} & \text{for } \zeta \leq 0, \ n < 1 \\ \frac{1.85n/n_m}{(1+3.80n/n_m)^{7/3}} & \text{for } \zeta \leq 0, \ n \geq 1 \\ \frac{0.094n/n_m}{(1+9.67n/n_m)^{1.74}} & \text{for } \zeta > 0 \end{cases} \qquad (7)$$

where the stability-dependency of the cospectral peak frequency n_m (Fig. S8) followed the equation

$$n_m = 0.0956, \text{ for } \zeta \leq 0 \qquad (8)$$

$$n_m = 0.0956(1 + 2.4163\zeta^{0.7033}), \text{ for } \zeta > 0 \qquad (9)$$

In both approaches (analytical and experimental), the time constant τ_s was empirically estimated by fitting the transfer function $T_{ws}(f)$ to the normalized ratio of cospectral densities

$$T_{ws} = \frac{N_\theta Co_{ws}(f)}{N_s Co_{w\theta}(f)} \qquad (10)$$

where N_θ and N_s are normalization factors and Co_{ws} and $Co_{w\theta}$ the scalar and temperature cospectra, respectively. The estimated time constant was 0.68 s for the Aerodyne QCLS and 0.35 s for LI-6262.

Low-frequency correction

Detrending the turbulent time series, especially with LD or RF methods, may also remove part of the real low frequency variations in the data (Lenschow et al., 1994; Kristensen, 1998), which should be corrected for in order to avoid flux underestimation. Low frequency correction in this study for different detrending methods was done according to Rannik and Vesala (1999).

2.4.4 Flux quality criteria

The calculated fluxes were accepted when following criteria were met: the second wind rotation angle (θ) was below 10 °, the number of spikes in one half hour was less than 100, the COS mixing ratio was higher than 200 ppt, the CO_2 mixing ratio ranges between 300 ppm and 650 ppm and the H_2O mixing ratio was higher than 1 ppb.

We used a similar flagging system for micrometeorological quality criteria as Mauder and Foken (2006) for COS and CO_2: flag 0 was given if flux stationarity was less than 0.3 (meaning that covariances calculated over 5 min intervals deviate less than 30 % from the 30 min covariance), kurtosis was between 1 and 8 and skewness was within a range from -2 to 2. Flag 1 was given if flux stationarity was from 0.3 to 1 and kurtosis and skewness were within the ranges given earlier. Flag 2 was given if these criteria were not met.

In addition to these filtering and flagging criteria, we added friction velocity (u_*) filtering to screen out data collected under low turbulence conditions. A decrease in measured EC flux is usually observed under low turbulence conditions, although it is assumed that gas exchange should not decrease due to low turbulence. While this assumption holds for CO_2, it may not be justified for COS (Kooijmans et al., 2017), as will be further discussed in Sect. 3.5. The appropriate u_* threshold was derived from a 99 % threshold criterion (Papale et al., 2006; Reichstein et al., 2005). The lowest acceptable u_* value was determined from both COS and CO_2 nighttime fluxes.

2.4.5 Storage change flux calculation

Storage change fluxes were calculated from gas mixing ratio profile measurements and by assuming a constant profile throughout the canopy using EC system mixing ratio measurements. Storage change fluxes from mixing ratio profile measurements were calculated using the formula

$$F_{stor} = \frac{p}{RT_a} \int_0^{z_m} \frac{\partial \chi_c(z,t)}{\partial t} dz,$$ (11)

where p is the atmospheric pressure, T_a air temperature, R the universal gas constant and $\chi_c(z)$ the gas mixing ratio at each measurement height. The integral was determined from hourly measured χ_c profile at 0.5 m, 4 m, 14 m, and 23 m by integrating an exponential fit through the data (Kooijmans et al., 2017). When the profile measurement was not available, storage was calculated from COS (or CO_2) mixing ratio measured by the EC setup.

Another storage change flux calculation was done assuming a constant profile from the EC measurement height (23 m) to the ground level. A running average over a 5 hour window was applied to the COS mixing ratio data to reduce the random noise of measurements.

The storage change fluxes were used to correct the EC fluxes for storage change of COS and CO_2 below the flux measurement height.

2.4.6 Flux uncertainty

The flux uncertainty was calculated according to ICOS recommendations presented by Sabbatini et al. (2018). First, flux random error was estimated as the variance of covariance, according to Finkelstein and Sims (2001):

$$\epsilon_{rand} = \frac{1}{N} \left(\sum_{j=-m}^{m} \hat{\gamma}_{w,w}(j)\hat{\gamma}_{c,c}(j) + \sum_{j=-m}^{m} \hat{\gamma}_{w,c}(j)\hat{\gamma}_{c,w}(j) \right) \tag{12}$$

where N is the number of datapoints (18 000 for 30 min of EC measurements at 10 Hz), m number of samples sufficiently large to capture the integral timescale (18 000 was used in this study), $\hat{\gamma}_{w,w}$ is the variance and $\hat{\gamma}_{w,c}$ the covariance of the measured variables w and c (in this case, the vertical wind velocity and gas mixing ratio).

As the chosen processing schemes affect the resulting flux, the uncertainty related to the used processing options have to be accounted for. This uncertainty was estimated as

$$\epsilon_{proc} = \frac{max(F_{c,j}) - min(F_{c,j})}{\sqrt{12}} \tag{13}$$

where $F_{c,j}$ is the flux calculated according to j=1,...,N different processing schemes and N is the number of possible processing scheme combinations that are equally reliable but cause variability in fluxes. For simplicity, the processing steps we considered here are detrending, coordinate rotation and high-frequency spectral correction, leading to $N = 8$ processing schemes: BA with 2D coordinate rotation and experimental spectral correction, BA with 2D rotation and Horst (1997) spectral correction, BA with planar fitting and experimental correction, BA with planar fitting and Horst (1997) correction, LD with 2D rotation and experimental correction, LD with 2D rotation and Horst (1997) correction, LD with planar fitting and experimental correction and LD with planar fitting and Horst (1997) correction. As all the different processing schemes will lead to slightly different random errors as well, the flux random error was estimated to be the mean of Eq. (12) for different processing schemes:

$$\overline{\epsilon_{rand}} = \frac{\sqrt{\sum_{j=1}^{N} \epsilon_{rand,j}^2}}{N} \tag{14}$$

The combined flux uncertainty is then the summation of ϵ_{rand} and ϵ_{proc} in quadrature:

$$\epsilon_{comb} = \sqrt{\overline{\epsilon_{rand}}^2 + \epsilon_{proc}^2} \tag{15}$$

To finally get the total uncertainty as the 95th percentile confidence interval, the total uncertainty becomes

$$\epsilon_{total} = 1.96\epsilon_{comb} \tag{16}$$

It should be noted, though, that this uncertainty estimate holds for single 30 min fluxes only. When working with fluxes averaged over time, the total uncertainty cannot be directly propagated to the long-term averages because the two uncertainty sources behave differently. The random uncertainty is expected to decrease with increasing number of observations while

processing related uncertainty would not be much affected by time averaging. The random uncertainty of a flux averaged over multiple observations is obtained as

$$\langle \overline{\epsilon_{rand}} \rangle = \sqrt{\frac{\Sigma_{i=1}^{N}(\overline{\epsilon_{rand,i}})^2}{N^2}} \tag{17}$$

where N is the number of observations and $\overline{\epsilon_{rand,i}}$ the random uncertainty of each observation (Rannik et al., 2016). In this study, we calculated the time averaged processing uncertainty from Eq. 13 with time averaged fluxes from the four different processing schemes. The total uncertainty of time averaged flux was then calculated similarly from Eq. 15 and 16 with time averaged random and processing uncertainties.

2.4.7 Flux gap-filling

Missing CO_2 fluxes were gap-filled according to Reichstein et al. (2005) while missing COS fluxes were replaced by simple model estimates or by hourly mean fluxes if model estimates were not available, in a way comparable to gap-filling of CO_2 fluxes (Reichstein et al., 2005).

The COS gap-filling function was parameterised in a moving time window of 15 days to capture the seasonality of the fluxes. To calculate gap-filled fluxes, the parameters were interpolated daily. Gaps where any driving variable of the regression model was missing were filled with the mean hourly flux during the 15-day period.

We tested different combinations of linear or saturating (rectangular hyperbola) functions of the COS flux on PPFD and linear functions of the COS flux against vapor pressure deficit (VPD) or relative humidity (RH). The saturating light response function with the mean nighttime flux as a fixed offset term explained the short-term variability of COS flux relatively well but the residuals as a function of temperature, RH and VPD were clearly systematic. Therefore, for the final gap-filling, we used a combination of saturating function on PPFD and linear function on VPD that showed good agreement with the measured fluxes while having a relatively small number of parameters:

$$F_{COS} = a * I/(I+b) + c * D + d \tag{18}$$

where I is PPFD (μmol m^{-2}s^{-1}), D is VPD (kPa) and a (pmol m^{-2}s^{-1}), b (μmol m^{-2}s^{-1}), c (pmol m^{-2} s^{-1}kPa^{-1}), and d (pmol m^{-2}s^{-1}) are fitting parameters. Parameter d was set as the median nighttime COS flux over the 15-day window and other parameters estimated accordingly.

 ## 3   Results and Discussion

3.1   Detrending methods

In order to check the contribution of different detrending methods on the resulting flux, we made flux calculations with different methods: block averaging (BA), linear detrending (LD) and recursive filtering (RF) using the same time lag (CO_2 time lag) for all runs (Fig. S5).

305   The largest median COS flux (most negative) was obtained from RF (-12.0 pmol m^{-2}s^{-1}) while the smallest median (least negative) flux resulted from BA (-10.7 pmol m^{-2}s^{-1}) and LD (-11.3 pmol m^{-2}s^{-1}) differed by 5.3 % from BA (Table 2). The range of the 30 min COS flux was the largest (from -183.2 to 82.56 pmol m^{-2}s^{-1}) when using the BA detrending method, consistent with a similar comparison in Gerdel et al. (2017). In comparsion, it was from -107.3 to 73.1 pmol m^{-2}s^{-1} for LD and from -164.9 to 36.8 pmol m^{-2}s^{-1} for RF. While it was surprising that RF resulted in a more negative median flux than 310   BA, it is likely explained by the large variation in BA with compensating high negative and positive flux values, as the positive flux values with BA and LD are higher than with RF. In addition, it has to be kept in mind that the fluxes were corrected for low frequncy losses, which in part also accounts for the effects of detrending.

For CO_2, the largest median flux resulted from BA (-0.62 μmol m^{-2}s^{-1}). The smallest median flux resulted from LD (-0.56 μmol m^{-2}s^{-1}) and RF (-0.59 μmol m^{-2}s^{-1}). The difference between median CO_2 flux with BA and LD was 10.7 %.

315   The most commonly recommended averaging methods are BA (Sabbatini et al., 2018; Nemitz et al., 2018; Moncrieff et al., 2004) and LD (Rannik and Vesala, 1999) because they have less impact on spectra (Rannik and Vesala, 1999) and require less low-frequency corrections. These are also the most used detrending methods in previous COS EC flux studies (Table 1). RF may underestimate the flux (Aubinet et al., 2012) but as spectroscopic analyzers are prone to e.g. fringe effects under field conditions, the use of RF might still be justified (Mammarella et al., 2010). Regular checks of raw data provides information 320   on instrumental drift and helps to determine the optimal detrending method. It is also recommended to check the contribution of each detrending method on the final flux to better understand what is the low frequency contribution in each measurement site and setup.

3.2   Effect of time lag correction

325   Different time lag methods resulted in slightly different time lags and COS fluxes. The most common time lags were 2.6 s from the COS$_{lag}$, CO2$_{lag}$ and DetLim$_{lag}$ methods, and 1.5 s from RM$_{lag}$, which was the lag window limit (Fig. 2). Time lags determined from the COS$_{lag}$ and RM$_{lag}$ methods were distributed evenly throughout the lag window, whereas the lag from the CO2$_{lag}$ and DetLim$_{lag}$ methods were normally distributed with most lags detected at the window center.

We tested determining time lags also from the most commonly used method of maximizing the absolute covariance. If 330   the time lag was determined from the absolute covariance maximum instead of the maximum difference to a line between covariance values at the lag window limits, the COS$_{lag}$ and RM$_{lag}$ had most values at the lag window limits (Fig. S2). This resulted in a "mirroring effect" (Langford et al., 2015), i.e., fluxes close to zero were not detected as often as with other

Table 2. Median COS and CO_2 fluxes (in pmol m^{-2}s^{-1} and μmol m^{-2}s^{-1}, respectively) during 26 June to 2 November 2015 and their difference to the reference fluxes when using different processing options. Median reference fluxes are -11.3 pmol m^{-2}s^{-1} and -0.56 μmol m^{-2}s^{-1} for COS and CO_2, respectively, and median daytime (nighttime) fluxes -16.8 (-4.1) pmol m^{-2}s^{-1} and -4.58 (1.23) μmol m^{-2}s^{-1} when using linear detrending, CO_2 time lag and experimental high frequency spectral correction.

	Detrending		Time lag				Spectral corrections		Coordinate rotation
	BA	RF	COS_{lag}	$Const_{lag}$	RM_{lag}	$DetLim_{lag}$	Horst (1997)	None	2D
Median F_{COS}	-10.7	-12.0	-11.5	-11.1	-11.1	-13.1	-11.0	-9.7	-11.7
Difference to reference	5.3 %	6.2 %	1.8 %	1.8 %	1.8 %	15.9 %	2.7 %	14.2 %	3.5 %
Daytime median F_{COS}	-16.0	-17.6	-17.4	-16.6	-16.6	-19.2	-16.9	-15.0	-17.8
Nighttime median F_{COS}	-4.2	-4.1	-4.4	-4.1	-4.3	-4.9	-4.0	-3.4	-4.7
Median F_{CO2}	-0.62	-0.59	NA	NA	NA	NA	-0.54	-0.48	-0.55
Difference to reference	10.7 %	5.4 %	NA	NA	NA	NA	3.6 %	14.3 %	1.8 %
Daytime median F_{CO2}	-4.49	-5.00	NA	NA	NA	NA	-4.77	-4.18	-4.88
Nighttime median F_{CO2}	1.17	1.26	NA	NA	NA	NA	1.18	1.02	1.33

methods, since the covariance is always maximized and the derived flux switches between positive and negative values of similar magnitude (Fig. S3). This issue should be taken into account in COS EC flux processing, as absolute COS covariance maximum is by far the most commonly used method in determining the time lag in COS studies (Table 1). To make the different methods more comparable, the time lags were, in the end, determined from the maximum difference to a line between covariance values at the lag window limits in this study. In this way, time lags were not determined at the window borders and most of the methods had the final flux PDF peak approximately at the same flux values and had otherwise small differences in the distributions (Fig. 3). The only method that was clearly different from the others was $DetLim_{lag}$, that produced higher fluxes than other methods.

A constant time lag has been found to bias the flux calculation as the time lag likely can vary over time due to, for example, fluctuations in pumping speed (Langford et al., 2015; Taipale et al., 2010; Massman, 2000). However, as the $CO2_{lag}$ was often the same as the chosen constant lag 2.6 s, we did not observe major differences between these two methods. A reduced bias in the flux calculation with smoothed cross-covariance was introduced by Taipale et al. (2010), who recommended using this method for any EC system with a low signal-to-noise ratio. However, we do not recommend this as a first choice since the time lags do not have a clear distribution and if maximum covariance method is used, we find a mirroring effect with the RM_{lag} in the final flux distributions.

By using the $DetLim_{lag}$ method, the COS time lag was estimated for 54 % of cases from COS_{lag}, while the $CO2_{lag}$ was used as proxy for the COS time lag in about 46 % of cases. Fig. 3 shows that the raw covariance of COS only exceeds the noise level at higher COS flux values and thus the COS_{lag} is chosen by this method only at higher fluxes, as expected. At lower flux rates, and especially close to zero, the COS fluxes are not high enough to surpass the noise level and thus the $CO2_{lag}$ is chosen.
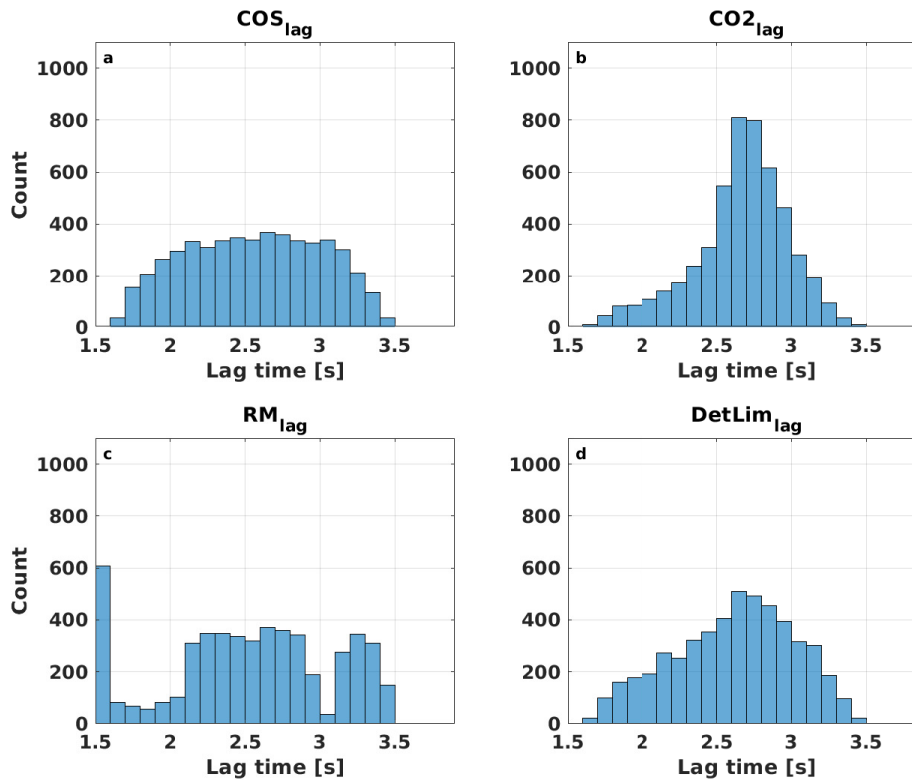
Figure 2. Distribution of time lags derived from different methods: COS_{lag} (a), $CO2_{lag}$ (b), COS time lag from a running mean cross-covariance (RM_{lag}, c) and combination of COS_{lag} and $CO2_{lag}$ ($DetLim_{lag}$, d).

The median COS fluxes were highest when the time lag was determined from the $DetLim_{lag}$ (-13.1 pmol m^{-2}s^{-1}) and COS_{lag} (-11.5 pmol m^{-2}s^{-1}) methods, respectively (Table 2). Using the COS_{lag} results in both higher positive and negative fluxes and might thus have some compensating effect on the median fluxes. $Const_{lag}$ and RM_{lag} produced the smallest median
355 uptake (-11.1 pmol m^{-2}s^{-1}) while the $CO2_{lag}$ had a median flux of -11.3 pmol m^{-2}s^{-1}. The difference between the reference flux and the flux from $DetLim_{lag}$ method was clearly higher (15.9 %) than with other methods (1.8 %) and had largest variation (from -113.8 pmol m^{-2}s^{-1} to 81.6 pmol m^{-2}s^{-1}) and standard deviation (16.1 pmol m^{-2}s^{-1}). This difference might become important in the annual scale, and as the most commonly used covariance maximization method does not produce a clear time lag distribution for $DetLim_{lag}$ or COS_{lag}, we recommend using the $CO2_{lag}$ for COS fluxes, as in most ecosystems the CO_2
360 cross-covariance with w is more clear than the cross-covariance of COS and w signals.
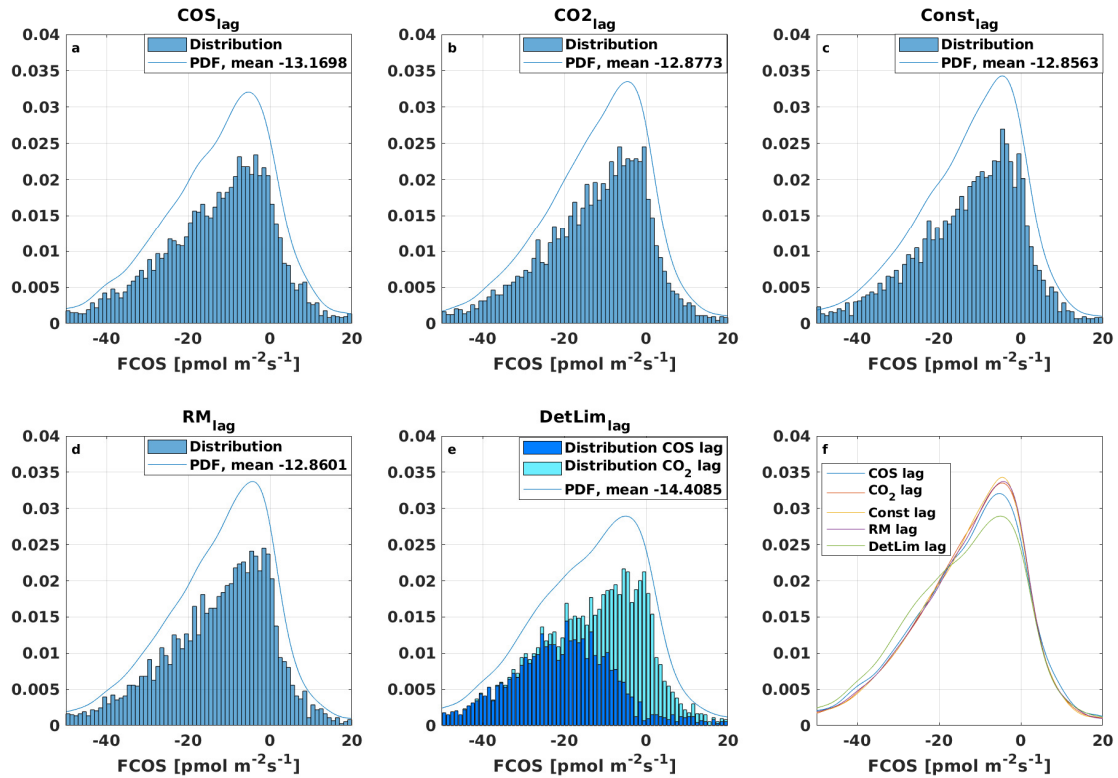
Figure 3. Normalized COS flux distributions using different time lag methods: COS$_{lag}$ (a), CO2$_{lag}$ (b), constant time lag of 2.6 s (Const$_{lag}$, c), time lag from a running mean COS cross-covariance (RM$_{lag}$, d) and combination of COS and CO$_2$ time lags (DetLim$_{lag}$, e), and a summary of all probability distribution functions (PDFs) (f).

3.3 High-frequency response correction

The mean COS cospectrum was close to the normal mean CO$_2$ cospectrum (compare Fig. 4a and 4c). The power spectrum of COS was dominated by noise as can be seen from the increasing power spectrum with increasing frequency for normalized frequencies greater than 0.2, which is similar to what was measured for COS by Gerdel et al. (2017) and for N$_2$O by Eugster et al. (2007). The fact that COS measurements are dominated by noise at high frequencies means that those measurements are limited by precision and that they likely do not capture the true variability in COS turbulence signals. This is less of a problem for CO$_2$, where white noise only starts to dominate at higher frequencies (normalized frequency higher than 3). Cospectral attenuation was found for both COS and CO$_2$ at high frequency.

High frequency losses due to e.g. attenuation in sampling tubes and limited sensor response times are expected to decrease fluxes if not corrected for (Aubinet et al., 2012). The median COS flux, when using the CO$_2$ time lag and keeping low-

frequency correction and quality filtering the same, was the least negative without any high-frequency correction (-9.7 pmol $m^{-2}s^{-1}$), most negative with the experimental correction (-11.3 pmol $m^{-2}s^{-1}$) and in between with the analytical correction (-11.0 pmol $m^{-2}s^{-1}$) (Fig. S6). Correcting for the high frequency attenuation thus made a maximum of 14.2 % difference in the median COS flux. In addition, daytime median flux magnitudes increased from -15.0 to -16.9 pmol $m^{-2}s^{-1}$ with the analytical correction, and to -16.8 pmol $m^{-2}s^{-1}$ with the experimental high frequency correction. However, the relative difference was larger during nighttime when flux magnitudes increased from -3.4 to -4.0 and -4.1 pmol $m^{-2}s^{-1}$ with analytical and experimental methods, respectively.

Similar results were found for the CO_2 flux but the differences were smaller: without any high frequency correction the median flux was the least negative at -0.48 μmol $m^{-2}s^{-1}$, most negative with experimental correction (-0.56 μmol $m^{-2}s^{-1}$) and in between with the analytical correction (-0.54 μmol $m^{-2}s^{-1}$), thus making a maximum of 14.3 % difference in the median CO_2 flux, similar to COS. Very similar results were found for CH_4 and CO_2 fluxes in Mammarella et al. (2016), where the high frequency correction made the largest difference in the final flux processing for closed-path analysers. Similarly to COS, also CO_2 flux magnitudes were increased more during nighttime due to spectral correction than during daytime. Daytime median flux magnitude increased from -4.18 to -4.77 and -4.58 μmol $m^{-2}s^{-1}$ and nighttime fluxes increased from 1.02 to 1.18 and 1.23 μmol $m^{-2}s^{-1}$ when using analytical and experimental high frequency spectral corrections, respectively. Flux attenuation was dependent on stability and wind speed for both correction methods, as also found in Mammarella et al. (2009) (Fig. S7).

The site specific model captures the cospectrum better than the model cospectrum by Horst (1997), as shown in Fig. 4. For high-frequency spectral corrections, it is recommended to use the site specific cospectral model, as has been done in most previous COS studies (Table 1).

3.4 Storage change fluxes

In the following, storage change fluxes based on profile measurements are listed as default, with fluxes based on the constant profile assumption listed in brackets.

The COS storage change flux was negative from 15:00 in the afternoon until 06:00 in the morning with a minimum of -1.0 pmol $m^{-2}s^{-1}$ (-0.6 pmol $m^{-2}s^{-1}$) reached at 20:00 in the evening. A negative storage change flux of COS indicates that there is a COS sink in the ecosystem when the boundary layer and effective mixing layer is shallow. Neglecting this effect would lead to overestimated uptake at the ecosystem level later when the air at the EC sampling height is better mixed. The COS storage change flux was positive from around 6:00 in the morning until 15:00 in the afternoon and peaked at 9:00 with a magnitude of 1.9 pmol $m^{-2}s^{-1}$ (0.8 pmol $m^{-2}s^{-1}$). The storage change flux made the highest relative contribution to the sum of measured EC and storage change fluxes at midnight with 18 % (13 %) (Fig. 5 c). The difference between the two methods was minimum 13 % at 11:00 and maximum 56 % at 09:00. The two methods made a maximum of 7 % difference on the resulting cumulative ecosystem flux, as already reported in Kooijmans et al. (2017).

The CO_2 storage change flux was positive from 15:00 in the afternoon until around 4:00 in the morning with a maximum value of 0.62 μmol $m^{-2}s^{-1}$ (0.38 μmol $m^{-2}s^{-1}$) reached at 21:00 in the evening. A positive storage change flux indicates that the ecosystem contains a source of carbon when the boundary layer is less turbulent and accumulates the respired CO_2
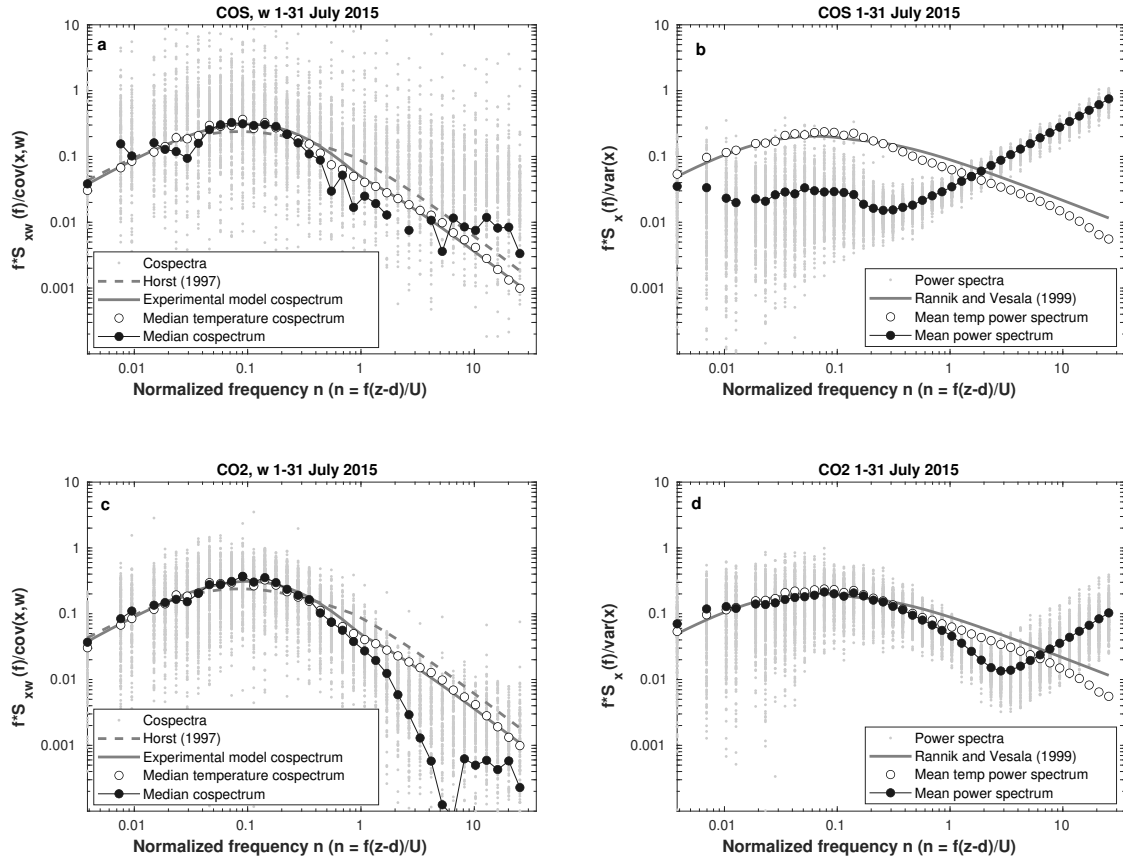
Figure 4. Cospectrum and power spectrum for COS (a and b, respectively) and CO_2 (c and d) in July 2015. All data were filtered by the stability condition $-2 < \zeta < -0.0625$ and COS data were only accepted when the covariance was higher than three times the random error due to instrument noise (Eq. 1). The cospectrum models by experimental method and Horst (1997), that were used in the high-frequency spectral correction, are shown in grey continuous and dashed lines, respectively.

within the canopy. As turbulence would increase later in the morning, the accumulated CO_2 would result in an additional flux that could mask the gas exchange processes occurring at that time step. The CO_2 storage change flux minimum was reached with both methods at 08:00 with magnitude -1.01 μmol m^{-2}s^{-1} (-0.52 μmol m^{-2}s^{-1}) when the boundary layer has already started expanding, and leaves are assimilating CO_2. The maximum contribution of the storage change flux was as high as 89 % (36 %) compared to the EC flux at 18:00, when the CO_2 exchange is turning to respiration and storage change flux increases its relative importance (Fig. 5 d). The difference between the two storage change flux methods for CO_2 was maximum 53 % at 21:00 and minimum 13 % at midnight. The maximum difference of 5 % was found in the cumulative ecosystem CO_2 flux when the different methods were used.

In conclusion, the storage change fluxes are not relevant for budget calculations – as expected – and have not been widely applied in previous COS studies (Table 1), even though storage change flux measurements are mandatory in places where the EC system is placed at a height of 4 m or above according to the ICOS protocol for EC flux measurements (Montagnani et al., 2018). In addition, storage change fluxes are important in the diurnal scale to account for the delayed capture of fluxes by the EC system under low turbulence conditions.

3.5 u_* filtering

Calm and low turbulence conditions are especially common during nights with stable atmospheric stratification. In this case, storage change and advective fluxes have an important role and the measured EC flux of a gas does not only reflect the atmosphere-biosphere exchange, typically underestimating the exchange. This often leads to a systematic bias in the annual flux budgets (Moncrieff et al., 1996; Aubinet et al., 2000; Aubinet, 2008). Even after studies of horizontal and vertical advection, the u_* filtering still keeps its place as the most efficient and reliable tool to filter out data that is not representative of the surface-atmosphere exchange under low turbulence (Aubinet et al., 2010).

For COS, the nighttime filtering is a more complex issue than it is for CO_2. In contrast to CO_2, COS is taken up by the ecosystem during nighttime (Kooijmans et al., 2017, 2019), depending on stomatal conductance and the concentration gradient between the leaf and the atmosphere. When the atmospheric COS mixing ratio decreases under low turbulence conditions (due to nighttime COS uptake in the ecosystem), the concentration gradient between the leaf and the atmosphere goes down, such that a decrease in COS uptake can be expected (Kooijmans et al., 2017). Thus, the assumption that fluxes do not go down under low turbulence conditions, as is the case for respiration of CO_2, does not necessarily apply to COS uptake. Gap-filling the u_* filtered COS fluxes may therefore create a bias due to false assumptions, if the gap-filling is only based on data from periods with high turbulence. However, as we did not see u_* dependency disappearing even with a concentration gradient-normalized flux, the u_* filtering and proceeding gap-filling were applied here as usual (Papale et al., 2006) to overcome the EC measurement limitations under low turbulence conditions.

We determined u_* limits of 0.23 ms^{-1} for COS and 0.22 ms^{-1} for CO_2 (Fig. 6). Filtering according to these u_* values would remove 12 % and 11 % of data, respectively. If the storage change flux was excluded when determining the u_* threshold, the limits were 0.39 ms^{-1} and 0.24 ms^{-1} from CO_2 and COS fluxes, respectively. The increase in the u_* threshold with CO_2 is because the fractional storage change flux is larger for CO_2 than for COS (Fig. 5, panels c and f). On the other hand, the u_*
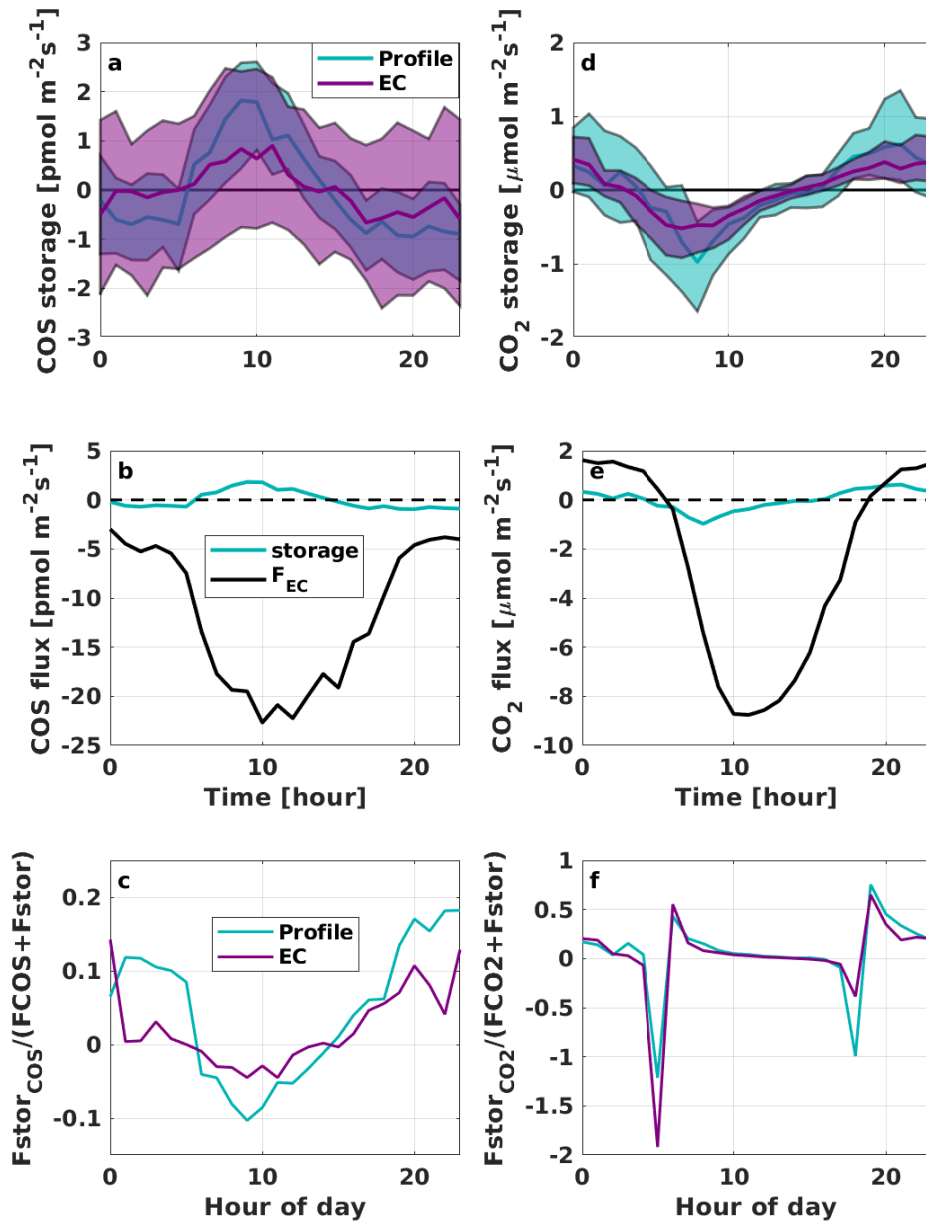
Figure 5. Diurnal variation of the storage change flux, determined from COS (a) and CO_2 (d) profile measurements (blue) and by assuming a constant profile up to 23 m height (purple), and diurnal variation of the EC flux (black) and storage change flux with the profile method (blue) (b and e for COS and CO_2 fluxes, respectively) during the measurement period 26 June to 2 November 2015. Contribution of storage change flux to the total ecosystem EC flux with the profile measurements and assuming a constant profile for COS (c) and CO_2 (f).

limit for COS stayed similar to the previous one. With these u_* thresholds the filtering would exclude 30 % and 13 % of the data for COS and CO_2 respectively.

If fluxes are not corrected for storage before deriving the u_* threshold, there is a risk of flux overestimation due to double accounting. The flux data filtered for low turbulence would be gap-filled, thereby accounting for storage by the canopy, but then accounted for again when the storage is released and measured by the EC system during the flushing hours in the morning (Papale et al., 2006). Thus, it is necessary to make the storage change flux correction before deriving u_* thresholds and applying the filtering.

3.6 Gap-filling

Three combinations of environmental variables (PAR, PAR and RH, PAR and VPD) were tested using the gap-filling function Eq. 18. These environmental parameters were chosen because COS exchange has been found to depend on stomatal conductance, which in turn depends especially on radiation and humidity (Kooijmans et al., 2019). Development of the gap-filling parameters a, b, c and d over the measurement period is presented in the Supplementary material Fig. S10. While saturating function of PAR only captured the diurnal variation already relatively well, adding a linear dependency on VPD or RH made the diurnal pattern even closer to the measured one, although some deviation is still observed especially in early morning (Fig. 7). Therefore, the combination of saturating light response curve and linear VPD dependency was chosen. Furthermore, we chose a linear VPD dependency instead of a linear RH dependency due to smaller residuals in the former (Fig. S9). The mean residual of the chosen model was -0.54 pmol m^{-2}s^{-1} and root-mean-square error (RMSE) 18.7 pmol m^{-2}s^{-1}, while saturating PAR function with linear RH dependency had a mean residual of -0.84 pmol m^{-2}s^{-1} and RMSE 19.3 pmol m^{-2}s^{-1}, and saturating PAR function had a residual of 0.97 pmol m^{-2}s^{-1} and RMSE 22.8 pmol m^{-2}s^{-1}.

For COS fluxes, 44 % of daytime flux measurements were discarded due to the above-mentioned quality criteria (Sect. 2.4.4) and low-turbulence filtering. As expected, more data (66 %) were discarded during nighttime. Altogether 52 % of all COS flux data were discarded and gap-filled with the gap-filling function presented in Eq. (18). The average of the corrected and gap-filled COS fluxes during the whole measurement period was -12.3 pmol m^{-2}s^{-1}, while without gap-filling the mean flux was 14 % more negative, -14.3 pmol m^{-2}s^{-1}. This indicates that most gap-filling was done for the nighttime data, when COS fluxes are less negative than during daytime.

For CO_2, 41 % daytime CO_2 fluxes were discarded, while 67 % of fluxes were discarded during nighttime. Altogether comprising 53 % of all CO_2 flux data. CO_2 fluxes were gap-filled according to standard gap-filling procedures presented in (Reichstein et al., 2005). The average NEE after all corrections and gap-filling was -2.14 μmol m^{-2}s^{-1}, while without gap-filling the mean flux was 39 % more negative, -3.53 μmol m^{-2}s^{-1}. Similar to COS, CO_2 fluxes are also mostly gap-filled during nighttime. As nighttime CO_2 fluxes are positive, gap-filled fluxes include more positive values than non-gap-filled fluxes, thus making the mean flux less negative.
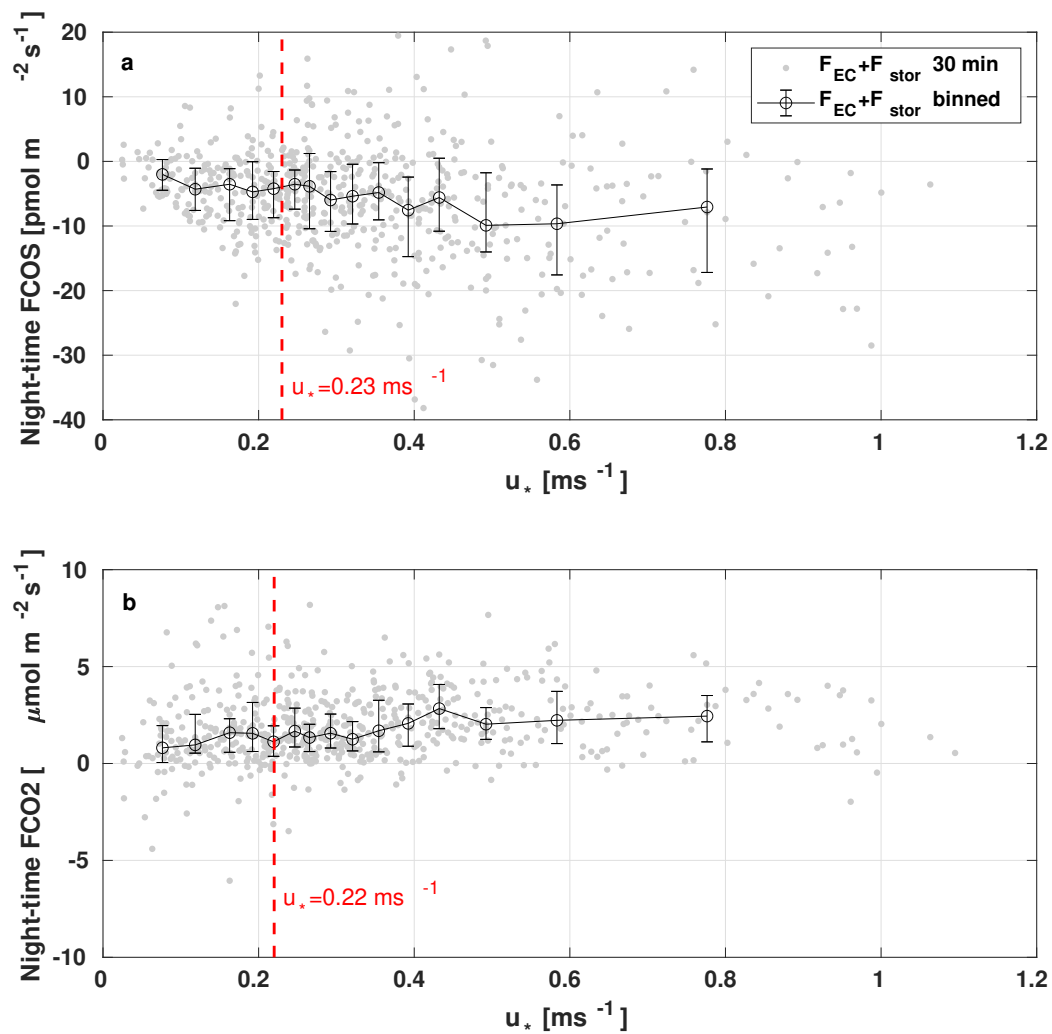
Figure 6. Nighttime median ecosystem fluxes (black) of COS (a) and CO_2 (b) binned into 15 equal-sized friction velocity bins. Error bars indicate ranges between 25th and 75th percentiles. Friction velocity thresholds are shown with red dashed lines and single data points of 30-min fluxes with light gray dots.
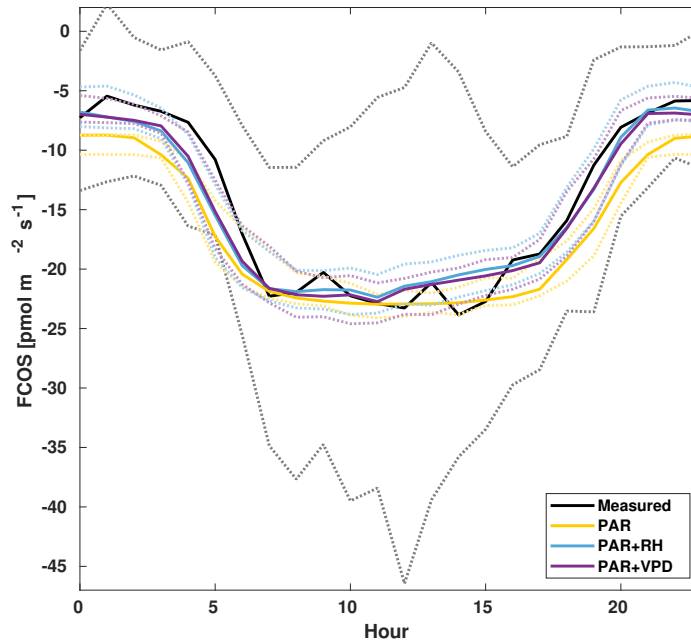
Figure 7. Diurnal variation of the measured COS flux (black) and the flux from different gap-filling methods: gap-filling with only saturating PAR function (yellow), saturating PAR and linear dependency on RH (blue) and saturating PAR and linear dependency on VPD (purple). Diurnal variation is calculated from 1 July to 31 August 2015 for periods when measured COS flux existed. Dashed lines represent the 25th and 75th percentiles.

Although the COS community has not been interested in the cumulative COS fluxes or yearly COS budget so far, it is important to fill short-term gaps in COS flux data to properly capture e.g. the diurnal variation. The gap-filling method presented here is one option to be tested also at other measurement sites.

3.7 Errors and uncertainties

475 The uncertainty due to chosen processing scheme was determined from a combination of eight different processing schemes, as described in Sec. 2.4.6, that were equally reliable but caused most variation in the final COS flux (Table 2). This processing uncertainty contributed 36 % to the total uncertainty of the 30 min COS flux, while the rest was composed of the random flux uncertainty (Fig. 8). For the CO_2 flux uncertainty, the processing was more important than for COS (48 %), but the random uncertainty was still dominating the combined flux uncertainty. The random error of the CO_2 flux was found to be lower than 480 in Rannik et al. (2016) for the same site, probably related to differences in the gas analyzers and overall setup. The mean noise estimated from Lenschow et al. (2000) was 0.06 μmol m^{-2}s^{-1} for our QCLS CO_2 fluxes while in Rannik et al. (2016) it

was approximately 0.08 μmol m^{-2}s^{-1} for LI-6262 CO_2 fluxes at the same site. Gerdel et al. (2017) found the total random uncertainty of COS fluxes to be mostly around 3–8 pmol m^{-2}s^{-1}, comparable to our results.

The relative flux uncertainty for COS was very high at low flux (-3 pmol m^{-2}s^{-1} < F_{COS} < 3 pmol m^{-2}s^{-1}) values (8 times the actual flux value) but leveled off to 45 % at fluxes higher (meaning more negative fluxes) than -27 pmol m^{-2}s^{-1} (Fig. 8c). The total uncertainty of the CO_2 flux was also high at low fluxes (-1.5 μmol m^{-2}s^{-1} < F_{CO2} < 1 μmol m^{-2}s^{-1}, uncertainty reaching 130 % of the flux at 0.17 μmol m^{-2}s^{-1}) and decreased to 15 % at fluxes more negative than -11 μmol m^{-2}s^{-1} (Fig. 8d). Higher relative uncertainty at low flux levels is probably due to the detection limit of the measurement system.

The median relative random uncertainty of COS flux decreased from 0.35 for single 30 min flux to 0.013 for monthly averaged flux (Fig. 8e). The processing uncertainty had less prominent decrease from 0.15 for 30 min fluxes to 0.05 for monthly fluxes. The strongest decrease in processing uncertainty was when moving from single 30 min flux values to daily average fluxes, after which the averaging period did not affect the processing uncertainty. This is probably due to the large scatter between the two detrending methods, that levels off when averaging over several flux values (Fig. S11 a). Relative random uncertainty of CO_2 flux decreased from 0.11 for 30 min fluxes to 0.021 for monthly fluxes. The processing uncertainty, however, did not change significantly between the different averaging periods, as would be expected.

4 Conclusions

In this study, we examined the effects of different processing steps on COS EC fluxes and compared them to CO_2 flux processing. COS fluxes were calculated with five time lag determination methods, three detrending methods, two high-frequency spectral correction methods and with no spectral corrections. We calculated the storage change fluxes of COS and CO_2 from two different concentration profiles and investigated the diurnal variation in the storage change fluxes. We applied u_* filtering and introduced a gap-filling method for COS fluxes. We also quantified the uncertainties of COS and CO_2 fluxes.

The largest differences in the final fluxes came from time lag determination and detrending. Different time lag methods made a difference of maximum 15.9 % in the median COS flux. Different detrending methods, on the other hand, made a maximum of 6.2 % difference in the median COS flux while it was more important for CO_2 (10.7 % difference between linear detrending and block averaging). Omitting high-frequency spectral correction resulted in a 14.2 % lower median flux, while different methods used in high-frequency spectral correction resulted in only 2.7 % difference in the final median fluxes. We suggest to use CO_2 time lag for COS flux calculation so that potential biases due to low signal-to-noise ratio of COS mixing ratio measurements can be eliminated. CO_2 mixing ratio is measured simultaneously with COS mixing ratio with the Aerodyne QCLS and in most cases has a higher signal-to-noise ratio and more clear cross-covariance with w than COS. Experimental high frequency correction is recommended for accurately correcting for site specific spectral losses. We recommend comparing the effect of different detrending methods on the final flux for each site separately, to determine the site and instrument specific trends in the raw data.

Flux uncertainties of COS and CO_2 followed a similar trend of higher relative uncertainty at low flux values and random flux uncertainty dominating over uncertainty related to processing in the total flux uncertainty. The relative uncertainty was more
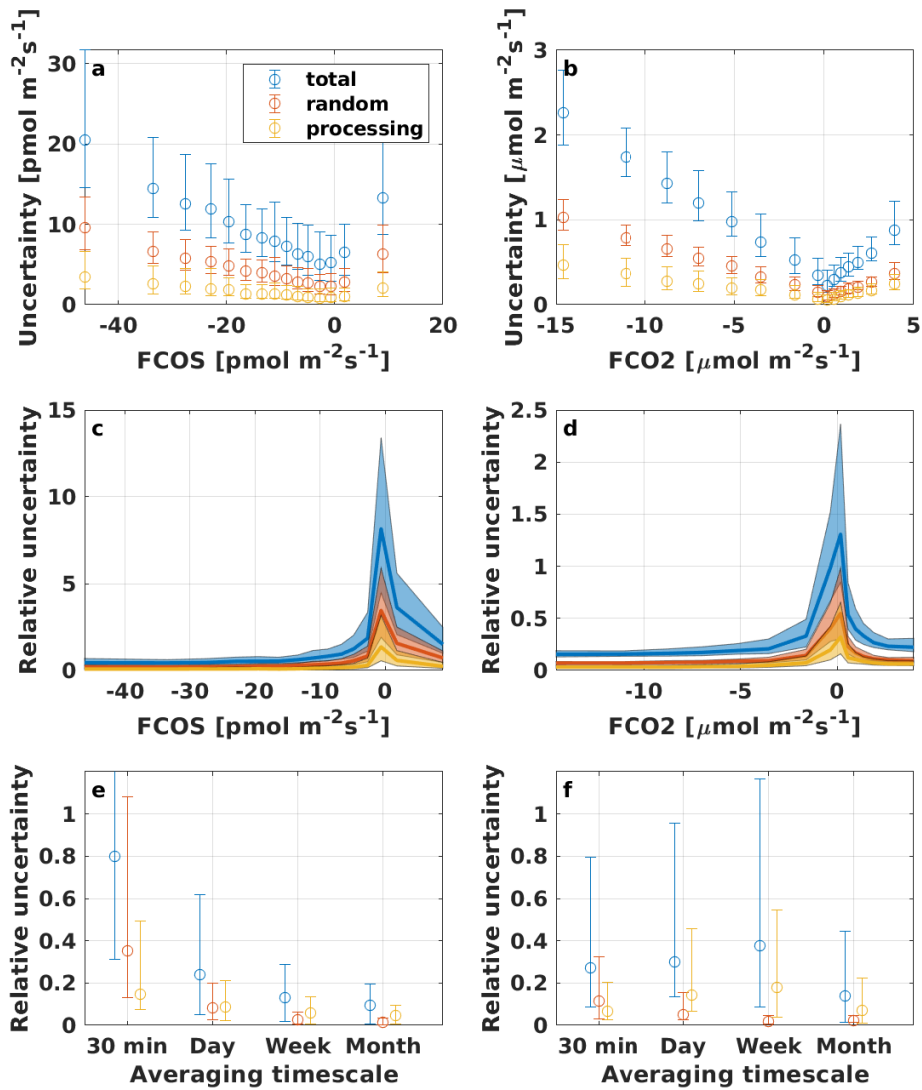
Figure 8. Uncertainty of COS and CO$_2$ fluxes, binned to 15 equal-sized bins that represent median values (a,b). Error bars show the 25th and 75th percentiles. Total uncertainty is represented as the 95 % confidence interval (1.96 ϵ_{comb}) Panels c and d represent the relative uncertainty, i.e. the uncertainty divided by the flux, for COS and CO$_2$, respectively. Panels e and f represent the relative uncertainties of time averaged COS and CO$_2$ fluxes, respectively.

than 5 times higher for COS than for CO_2 at low flux values (absolute COS flux less than 3 pmol m^{-2} s^{-1}), while at higher fluxes they were more similar. When averaging fluxes over time, the relative random uncertainty decreased with increasing averaging period both for COS and CO_2. Relative processing uncertainty decreased from single 30 min COS fluxes to daily averages but remained at the same level for longer averaging periods.

We emphasize the importance of time lag method selection for small fluxes, whose uncertainty may exceed the flux itself, to avoid systematic biases. COS EC flux processing follows similar steps as other fluxes with low signal-to-noise ratio, such as CH_4 and N_2O, but as there are no sudden bursts of COS expected and its diurnal behaviour is close to CO_2, some processing steps are more similar to CO_2 flux processing. In particular, time lag determination and high-frequency spectral corrections should follow the protocol of low signal-to-noise ratio fluxes (Nemitz et al., 2018), while QA/QC, despiking, u_* filtering and storage change correction should follow the protocol produced for CO_2 flux measurements (Sabbatini et al., 2018). Our recommendation for time lag determination (CO_2 cross-covariance) differs from the most commonly used method so far (COS cross-covariance), while experimental high frequency spectral correction has been widely applied already before (Table 1). Many earlier studies have neglected the storage change flux, but we emphasize its importance in the diurnal variation of COS exchange. In addition, we encourage implementing gap-filling to future COS flux calculations for eliminating short-term gaps in data.

Data availability. The flux data used in this study is available in reference Kohonen (2020). Environmental data are available in the AVAA – Open research data publishing platform (https://avaa.tdata.fi/). The metadata of the environmental observations are available via the ETSIN service. Raw data and flux data processed with different schemes other than the standard scheme are available upon request from the first author.

Author contributions. KMK and IM designed the study and KMK processed and analyzed the data. PK developed the gap-filling function for COS and participated in data processing. IM, LK, US and HC participated in field measurements. KMK and IM wrote the manuscript with contributions from all co-authors.

Competing interests. The authors declare that they have no conflict of interest.

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