

	ESA Climate Change Initiative “Plus” (CCI+)	
	End-to-End ECV Uncertainty Budget (E3UB) – TROPOMI/WFMD	
	for the Essential Climate Variable (ECV) Greenhouse Gases (GHG)	Version 2 - Final 19 Dec 2019

ESA Climate Change Initiative “Plus” (CCI+)

End-to-End ECV Uncertainty Budget (E3UB)

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TROPOMI WFM-DOAS XCH₄

for the Essential Climate Variable (ECV)

Greenhouse Gases (GHG)

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Change log:

Version Nr.	Date	Status	Reason for change
Version 1	23 Oct 2019	Final	New document
Version 2	19 Dec 2019	Final	Minor changes in the text according to ESA's review

End-to-End ECV Uncertainty Budget (E3UB)

TROPOMI WFM-DOAS (TROPOMI/WFMD) XCH_4

Prepared by:

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Valid for:

TROPOMI WFM-DOAS

Product

Methane column-averaged dry air mole fraction (XCH_4)

Version

v1.2

Contents

1	Introduction	3
1.1	Purpose of document	3
1.2	Intended Audience	3
1.3	Error term definition	3
2	Error sources	4
2.1	Systematic	4
2.2	Random	4
3	Methodology	4
4	Error results	6
4.1	Error analysis based on synthetic data	6
4.1.1	Systematic error	6
4.1.2	Random error	7
4.2	Error analysis based on real data	9
4.2.1	Systematic error	9
4.2.2	Random error	11
4.2.3	Correlations	11
4.2.4	Reported uncertainty	11
5	Summary	12
	References	14

1 Introduction

1.1 Purpose of document

This E3UB provides an overview of random and systematic errors affecting the WFMD retrievals submitted for the ESA GHG-CCI+ Climate Research Data Package version 5 (CRDP#5). Application of confidence limits to the retrieval is required to translate remotely sensed data presented here into modelled estimations with a known degree of confidence, allowing detection of climate change impacts additional to the natural variability of greenhouse gases. In particular the GHG-CCI User Requirements have placed strict measurement error requirements on the participating GHG retrievals (Chevallier et al., 2014).

1.2 Intended Audience

This document is intended for users in the modelling community applying the WFMD products for inversions, as well as remote sensing experts interested in atmospheric soundings of XCH₄. In both cases the work presented here will give the user a more thorough understanding of error implicit in this GHG-CCI product.

1.3 Error term definition

Error terms used in this report are defined to maintain consistency with other CCI user group error terms recommended at the 2014 CCI co-location meeting. Following the descriptions of Chevallier et al. (2014) and references therein:

Error	Difference between measured values and reality.
Systematic error	Component of measurement error that in replicate measurements remains constant or varies in a predictable manner.
Bias	Estimate of a systematic measurement error.
Precision	Reproducibility or repeatability of a measurement. Precision is a measure of the random error and can be improved by suitable averaging.
Stability	Systematic error over time, with random errors largely removed by averaging of observations.
Sensitivity	Change of measurement due to instrumental and algorithmic response to physical or simulated input parameters.

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2 Error sources

The majority of error is added to measurements from sources grouped into two themes - scattering of radiation into and out of the sensed light path by poorly quantified aerosol loading and cloud parameters in combination with surface reflectivity and viewing geometry; and instrumental or forward model uncertainties (e.g., calibration, spectroscopy). The aforementioned errors can be further grouped into systematic and random error components.

2.1 Systematic

Systematic retrieval errors include algorithmic effects such as inaccuracy in the solar and radiative transfer models, which will not change with the duration of the satellite’s sensing. The same applies to restrictions in instrument calibration accuracy. Viewing geometry also affects retrievals in a regular fashion by modifying the light path of sensed radiation as a function of the position of the instrument and the sun. Interplay between increased path lengths and random error components such as aerosol optical depth add complications to the issue of measurement geometry.

2.2 Random

Random errors are introduced to observations at the sensing stage of a measurement by detector noise. In addition to instrument noise, atmospheric parameters are able to have effects on sounding measurements by scattering light in and out of the sensed column. Errors due to unknown aerosol parameters are particularly pronounced where the scattering and absorption effects of suspended particulate matter are poorly modelled. Scattering due to clouds which are not screened from observation record present similar problems. In addition to atmospheric parameters, specific instrument features, such as potential sensitivity nonuniformities of the detector array, also contribute to the pseudo-noise component.

3 Methodology

The Weighting Function Modified Differential Optical Absorption Spectroscopy (WFM-DOAS) algorithm (Buchwitz et al., 2005a,b; Schneising et al., 2008, 2009, 2011, 2012; Heymann et al., 2012a,b; Schneising et al., 2013, 2014a,b, 2019) is a least-squares method based on scaling (or shifting) pre-selected atmospheric vertical profiles. The column-averaged dry air mole fractions of methane (denoted XCH₄) are obtained from the vertical column amounts of CH₄ by normalising

Station	Latitude [°]	Longitude [°]	Altitude [km]	Reference
Eureka	80.05	−86.42	0.61	Strong et al. (2019)
Ny-Ålesund	78.92	11.92	0.02	Notholt et al. (2017)
Sodankylä	67.37	26.63	0.19	Kivi et al. (2014)
East Trout Lake	54.35	−104.99	0.50	Wunch et al. (2018)
Białystok	53.23	23.03	0.19	Deutscher et al. (2015)
Karlsruhe	49.10	8.44	0.11	Hase et al. (2015)
Orléans	47.97	2.11	0.13	Warneke et al. (2014)
Garmisch	47.48	11.06	0.75	Sussmann and Rettinger (2018)
Park Falls	45.94	−90.27	0.44	Wennberg et al. (2017)
Lamont	36.60	−97.49	0.32	Wennberg et al. (2016b)
Tsukuba	36.05	140.12	0.03	Morino et al. (2018a)
Edwards	34.96	−117.88	0.70	Iraci et al. (2016)
JPL	34.20	−118.18	0.39	Wennberg et al. (2016a)
Caltech	34.14	−118.13	0.24	Wennberg et al. (2015)
Saga	33.24	130.29	0.01	Shiomi et al. (2014)
Burgos	18.53	120.65	0.04	Morino et al. (2018b)
Ascension Island	−7.92	−14.33	0.03	Feist et al. (2014)
Darwin	−12.46	130.93	0.04	Griffith et al. (2014a)
Réunion	−20.90	55.49	0.09	De Mazière et al. (2017)
Wollongong	−34.41	150.88	0.03	Griffith et al. (2014b)
Lauder	−45.04	169.68	0.37	Sherlock et al. (2014)

Table 3.1: TCCON sites used in the validation ordered according to latitude from north to south.

with the air column, which is obtained from the European Centre for Medium-Range Weather Forecasts (ECMWF) analysis. The corresponding vertical column amounts of CH₄ are retrieved from the measured sun-normalised radiance using spectral fitting windows in the SWIR spectral region (2311-2315.5 nm and 2320-2338 nm). A post-processing machine learning-based quality filter is applied for removal of low quality retrievals (Schneising et al., 2019). An additional (post-processing) shallow learning calibration procedure is applied to minimise residual systematic retrieval biases (Schneising et al., 2019).

The error analysis is based on synthetic data and validation of the results based on real TROPOMI data with independent reference data. The validation data set is the GGG2014 collection of the Total Carbon Column Observing Network (TCCON) (Wunch et al., 2011a) (available from <https://tccondata.org/>). From the validation with TCCON data at the 21 TCCON sites listed in Table 3.1, realistic error estimates of the satellite data are provided.

The error analysis presented in the following is largely adopted from Schneising et al. (2019). To compare the satellite data with TCCON quantitatively, it has to be taken into account that the sensitivities of the instruments differ from each other and that individual apriori profiles are used to determine the best estimate of the true atmospheric state, respectively. The first step is to correct for the apriori contribution to the smoothing equation by adjusting the measurements for a common apriori profile (Rodgers, 2000; Schneising et al., 2012; Dils et al., 2014). Here we use the TCCON prior as the common apriori profile for all measurements:

$$\hat{c}_{\text{adj}} = \hat{c} + \frac{1}{m_0} \sum_l m_l (1 - A_l) (x_{a,T}^l - x_a^l) \quad (1)$$

In this equation, $\hat{c}$ represents the originally retrieved TROPOMI column-averaged dry air mole fraction, l is the index of the vertical layer, A_l the corresponding column averaging kernel of the TROPOMI algorithm, x_a and $x_{a,T}$ the TROPOMI and TCCON apriori dry air mole fraction profiles. m_l is the mass of dry air determined from the dry air pressure difference between the upper and lower boundary of layer l via $\frac{\Delta p_l}{g_l}$ with (latitude-dependent) gravitational acceleration g_l and $m_0 = \sum_l m_l$ is the total mass of dry air. To minimise the smoothing error introduced by the averaging kernels we do not compare $\hat{c}_{\text{adj}}$ directly with the retrieved TCCON mole fractions $\hat{c}_T$ but rather with the adjusted expression (Rodgers and Connor, 2003; Wunch et al., 2011b)

$$\hat{c}_{T,\text{adj}} = c_{a,T} + \left(\frac{\hat{c}_T}{c_{a,T}} - 1 \right) \frac{1}{m_0} \sum_l m_l A_l x_{a,T}^l \quad (2)$$

Thereby, $c_{a,T}$ represents the TCCON apriori column-averaged dry air mole fraction associated with the apriori profile $x_{a,T}$.

4 Error results

4.1 Error analysis based on synthetic data

4.1.1 Systematic error

Several error sources were analysed using simulated measurements (Schneising et al., 2019). That means that for different scenarios defined by specific atmospheric conditions, radiances and irradiances are calculated with the radiative transfer model, which are subsequently used as measurement input in the retrieval. The errors are then defined as the deviation of the retrieved from the true

quantities. The corresponding results for several scenarios are summarised in Table 4.1. All scenarios already include interpolation between different wavelength grids (for measured and reference spectra) unless otherwise stated.

The analysis includes *Basic* scenarios testing if perturbations of the state vector elements can be retrieved, quantifying look-up table interpolation errors, and analysing errors caused by off-nadir conditions. In order to examine the sensitivity to vertical profile variations, the scenario class of *Profiles* includes several realistic model atmospheres based on measurements and theoretical predictions (Anderson et al., 1986), with all methane profiles scaled to have surface values of 1850 ppb in each case to better represent current atmospheric conditions. The respective atmospheres differ from the U.S. Standard Atmosphere with respect to temperature, pressure, water vapour, carbon monoxide, and methane profiles. These scenarios are more difficult to deal with than the basic ones, because the perturbations are not consistent with the scaling assumption, i.e., they include proper variations of the profile shape.

Also examined is the sensitivity to the *Spectral albedo* of typical natural surface types. The analysed *Aerosol* scenarios are largely described in Schneising et al. (2008, 2009) with aerosol type definitions in the different atmospheric layers based on Optical Properties of Aerosols and Clouds (OPAC) (Hess et al., 1998). The retrieval errors due to undetected *Subvisual clouds* are also investigated for different ice and water clouds. Larger systematic errors occur in the case of thick clouds because clouds are not explicitly considered in the forward model of the retrieval algorithm to retain the high processing speed. However, these cases are typically filtered out reliably by the implemented quality filter.

Thus, the scenarios of Table 4.1 give an impression of the magnitude of errors one can theoretically expect after quality filtering: Typical systematic retrieval errors are below 1% for CH₄ even for challenging scenarios. Corresponding errors for the simultaneously retrieved CO are below 2%.

4.1.2 Random error

The retrieval noise is determined via error propagation from the measurement noise. To assess the theoretical precision performance, we assume a simple shot noise limited noise model, which is defined in the following way: The reference signal-to-noise ratio is $SN_{ref} = 100$ in the continuum (radiance $L_{ref} = 4.3 \cdot 10^{11}$ phot/s/cm²/nm/sr) for a dark scene (albedo = 0.05) with low sun (solar zenith angle of 70°) and is scaled according to

$$SN(L) = SN_{ref} \sqrt{\frac{L}{L_{ref}}} \quad (3)$$

scenario		CH ₄ error [%]	CO error [%]
Basic	dry run (no λ interpol.)	0.00	0.00
	dry run	0.00	−0.03
	dry run $\oplus$	−0.08	−0.15
	dry run $\angle$	−0.09	−0.20
	$T + 30$ K	0.25	−0.24
	$T - 30$ K	0.06	−0.42
	$p + 5\%$	−0.01	−0.06
	$p - 5\%$	−0.04	−0.10
	albedo 0.2	−0.01	−0.04
Profiles	midlatitude summer	0.12	0.35
	midlatitude winter	−0.13	0.68
	subarctic summer	0.09	0.60
	subarctic winter	0.63	−0.59
	tropical	0.15	−0.94
Spectral albedo	sand	−0.03	−0.04
	soil	0.01	−0.03
	rangeland	0.02	−0.11
	deciduous	−0.07	0.01
	conifers	0.01	−0.19
	snow	−0.25	−0.30
	ocean	0.00	−0.07
Aerosols	no aerosol	0.01	0.10
	urban	0.11	0.04
	desert (sand albedo)	0.41	0.40
	arctic (snow albedo)	−0.19	−0.41
	extreme in boundary layer	0.34	−0.43
	extreme in boundary layer $\oplus$	0.24	−0.51
	extreme in boundary layer $\angle$	−0.28	−1.38
Subvisual clouds	cirrus	−0.29	−0.87
	cirrus $\oplus$	−0.41	−0.99
	cirrus $\angle$	−0.86	−1.84
	cirrus (fractal 50)	−0.33	−0.94
	cirrus (hexagonal 50×100)	0.11	−0.20
	cirrus ($\tau=0.05$)	−0.56	−1.53
	cumulus	−0.31	−0.86
	cumulus ($R=6\ \mu\text{m}$)	−0.21	−0.74
	cumulus ($R=14\ \mu\text{m}$)	−0.32	−0.87
	cumulus ($\tau=0.05$)	−0.55	−1.44

Table 4.1: Error analysis for different scenarios. Standard settings are direct nadir, sea level, solar zenith angle 50° , albedo 0.1, and U.S. Standard atmosphere. Scenarios with $\oplus$ include scaling of the CH₄ and CO profiles by 10%, for scenarios with $\angle$ the sensor zenith angle is set to 30° (relative azimuth 60°). Standard cirrus are located between 11 and 12 km (cloud optical thickness $\tau=0.03$) consisting of fractal ice crystals with an edge length of $100\ \mu\text{m}$. Standard cumulus are located between 3 and 4 km ($\tau=0.03$) consisting of water droplets with an effective radius R of $10\ \mu\text{m}$.

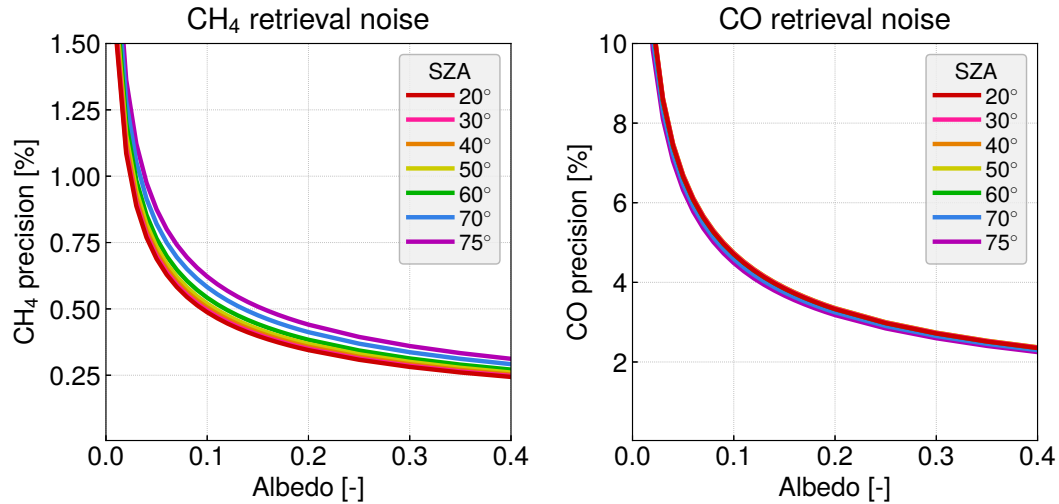


Figure 4.1: TROPOMI/WFM-DOAS CH₄ and CO relative retrieval noise for U.S. Standard atmosphere conditions.

for other radiances. The resulting absolute precision is widely independent of the current concentrations. For U.S. Standard atmosphere values, the corresponding relative retrieval noise for different albedos and solar zenith angles is shown in Figure 4.1. It is below 1% for solar zenith angles smaller than 75° and albedos larger than 0.03 in the case of CH₄. For the interested reader, the results for the simultaneously retrieved CO are also shown. As the CO absorption is considerably weaker than the CH₄ absorption, the CO retrieval exhibits larger relative noise, which is below 8% for albedos larger than 0.03.

4.2 Error analysis based on real data

4.2.1 Systematic error

The systematic error is quantified by validation with the 2014 release of ground-based Fourier Transform Spectroscopy (FTS) measurements of the Total Carbon Column Observing Network (TCCON) (Wunch et al., 2011a). To ensure comparability, all TCCON sites use similar instrumentation (Bruker IFS 125HR) and a common retrieval algorithm. The TCCON data are tied to the WMO trace gas scale using airborne in situ measurements applying individual scaling factors for each species. The estimated station-to-station accuracy (1σ) is about 3.5 ppb for XCH₄ (Wunch et al., 2010).

The validation results are summarised in Figure 4.2 including the mean bias μ and the scatter σ relative to TCCON for each site. The spatial systematic error is

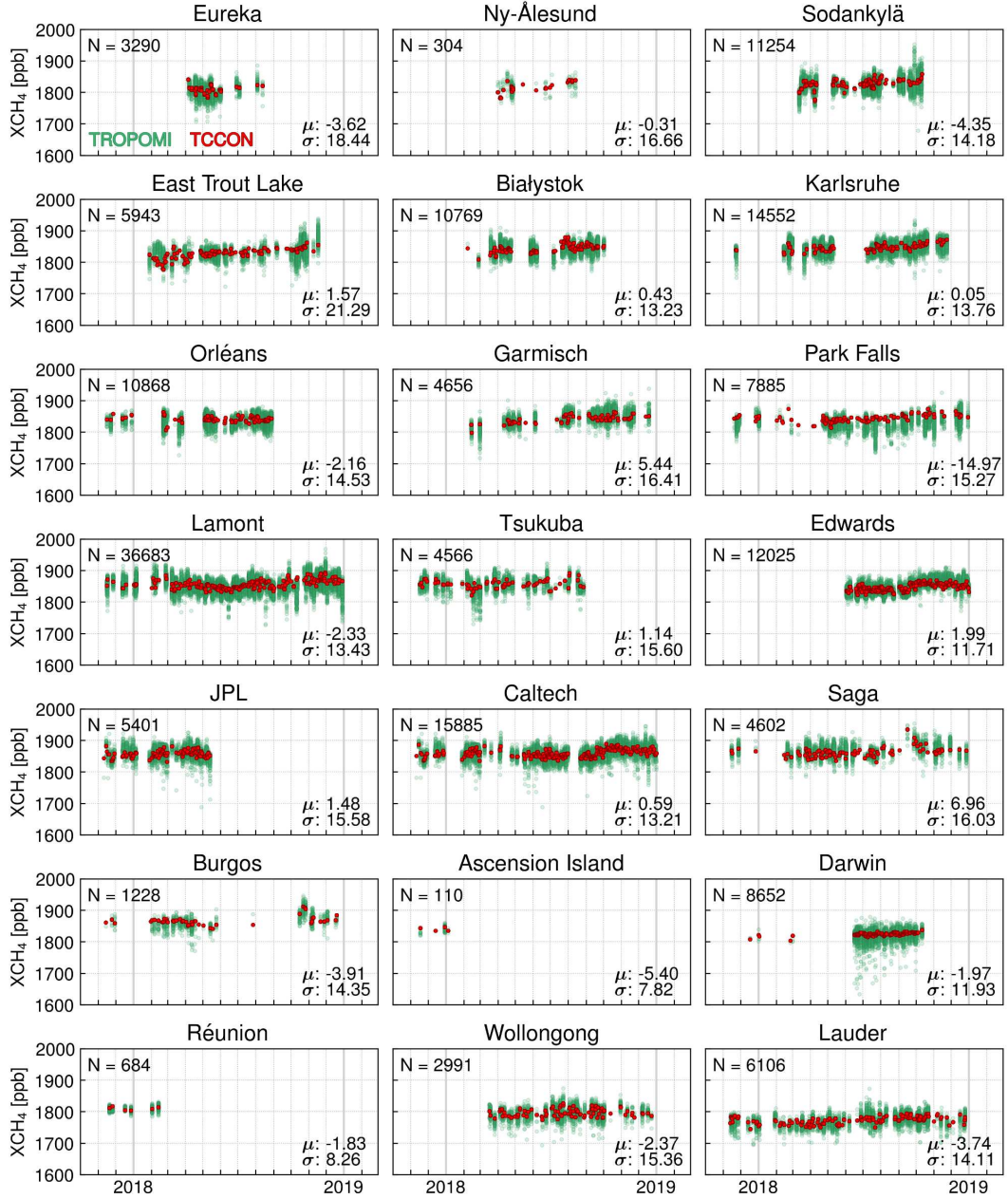


Figure 4.2: Comparison of the TROPOMI/WFMD v1.2 XCH₄ time series (green) with ground based measurements from the TCCON (red). For each site, N is the number of collocations, μ corresponds to the mean bias and σ to the scatter of the satellite data relative to TCCON in ppb.

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then defined as the the standard deviation of the local offsets μ relative to TCCON at the individual sites and amounts to 4.31 ppb. The seasonal systematic error is defined as the standard deviation of the four overall seasonal offsets (using all sites combined after subtraction of the respective local offsets) relative to TCCON and amounts to 0.57 ppb. The spatio-temporal systematic error (defined as the the root-sum-square of the spatial and seasonal systematic errors) amounts to 4.35 ppb, which is on the order of the estimated (station-to-station) accuracy of the TCCON of about 3.5 ppb. The global offset relative to the validation data aggregates to -1.30 ppb.

Long-term drift stability, and year-to-year stability are not determined at this juncture because the temporal coverage of the analysed time series is too short for a sound and stable estimation of these figures of merit.

4.2.2 Random error

The random error is estimated from the global scatter of the differences to TCCON after subtraction of the respective regional biases and amounts to 14.04 ppb. The corresponding scatter per site σ is shown in Figure 4.2.

4.2.3 Correlations

To assess the sensitivity of the retrieval results to several parameters, it is analysed to what extent the difference to TCCON is correlated with these parameters. The square of the correlation coefficient r^2 is a measure of how much of the difference to TCCON can potentially be explained by sensitivity to the respective parameter. As can be seen in Table 4.2, there are no indications for significant biases caused by the analysed parameters.

4.2.4 Reported uncertainty

The reported uncertainty of TROPOMI/WFMD v1.2 XCH₄ is estimated during the inversion procedure via error propagation from the uncorrelated spectral measurement errors σ_i given in the TROPOMI Level 1 files. The (unknown) pseudo-noise component determined by specific atmospheric parameters or instrumental features (e.g., potential striping in flight direction due to different offsets and gains of pixels of the detector array used for the staring push-broom concept of TROPOMI) is not considered and thus the reported uncertainty is typically underestimating the actual uncertainty.

To estimate the extent of this discrepancy, the reported uncertainty of XCH₄ is compared to the measured scatter relative to the TCCON for different sites

Parameter	r^2 [%]
Albedo (2313 nm)	0.8081
Cloud parameter r_{cld}	0.6144
Solar zenith angle	0.1248
Sensor zenith angle	0.0298
Altitude	0.0080
H ₂ O column	0.0001

Table 4.2: Sensitivity analysis of TROPOMI/WFMD v1.2 XCH₄ to several parameters by analysing the correlation of the difference to TCCON with these parameters.

and seasons in Figure 4.3. A robust regression (Huber and Ronchetti, 2009) provides the following statistical error parameterisation to obtain a more realistic uncertainty estimate $\hat{\sigma}$ from the reported uncertainty σ in the product files (neglecting the random and systematic errors of the TCCON measurements):

$$\hat{\sigma} = \sigma + 9 \text{ ppb}$$

5 Summary

An analysis based on simulated measurements suggests that typical systematic retrieval errors after quality filtering are below 1%. The validation with TCCON provides realistic error estimates. The corresponding error characteristics are summarised in Table 5.1. A correlation analysis confirms that there are no indications for significant biases caused by the analysed parameters.

Based on the comparison to TCCON it is concluded that the reported uncertainties are too optimistic because they only comprise the propagated measurement errors given in the Level 1 files. More realistic uncertainty estimates can be obtained by applying the error parameterisation determined from the regression shown in Figure 4.3 (adding a constant value of 9 ppb to the reported uncertainties). The improved error parameterisation will be explicitly implemented in the next version of the TROPOMI/WFMD XCH₄ product. In the meantime, this can easily be done by the user himself, if the uncertainties are of concern for the intended application.

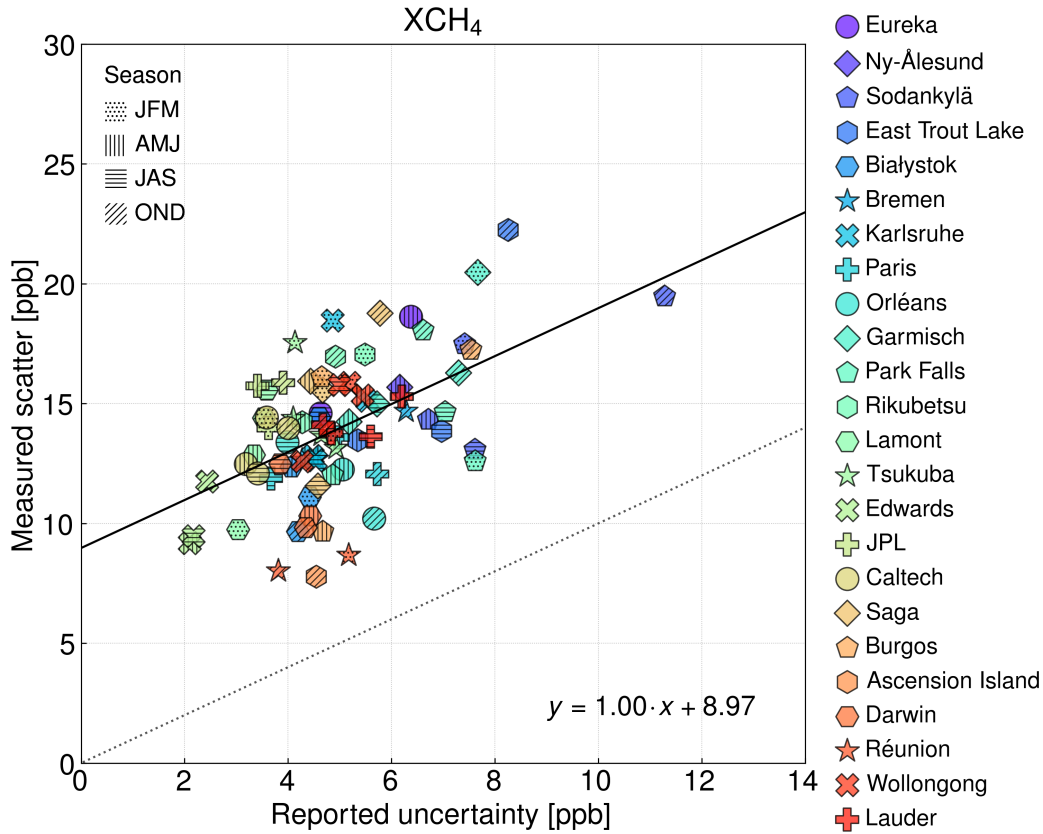


Figure 4.3: Comparison of the reported uncertainty of TROPOMI/WFMD v1.2 XCH₄ with the measured scatter relative to the TCCON for different sites and seasons. An improved error parameterisation is determined by a robust regression.

Sensor	Algorithm	Random error [ppb]	Systematic error (spatio-temporal) [ppb]	Stability [ppb·yr ⁻¹]
				Long-term Drift: n/a
TROPOMI	WFMD v1.2	14.04	4.35	Year-to-year: n/a

Table 5.1: TROPOMI/WFMD v1.2 XCH₄ error characteristics.

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