Harmonisation and trends of 20-years tropical tropospheric ozone data

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Abstract. Using a convective clouds differential (CCD) method, developed in house and applied to retrievals of total ozone and cloud data from three European satellite instruments (viz. GOME/ERS-2 (1995–2003), SCIAMACHY/Envisat (2002–2012), and GOME-2/MetOp-A (2007–2015)) monthly mean tropical tropospheric columns of ozone (TTCO) have been retrieved, which are in good agreement with ozonesondes (biases less than 6 DU). As small differences in TTCO between the individual instruments were evident, it was necessary to develop a scheme to harmonise the three datasets into one consistent time-series starting from 1996 until 2015. Correction offsets (bias) between the instruments using SCIAMACHY as intermediate reference have been calculated and six different harmonisation/merging scenarios have been evaluated. Depending on the merging approach, the magnitude, pattern, and uncertainty of the trends strongly vary. The harmonisation/merging represents an additional source of uncertainty in the trends (2 DU/decade on average, exceeding in most of the cases the uncertainty from the regression). For studying further details on tropospheric ozone trends on various spatial scales in the tropics we stick with one preferred merged dataset that shows best agreement with ozonesondes. In this merged dataset no correction was applied for GOME, and mean biases with respect to SCIAMACHY in the overlapping period (2007–2012) were calculated and applied for GOME-2 in each grid-box ($2.5^\circ \times 5^\circ$). In contrast with other studies we found that the tropospheric trend averaged over the tropics ($15^\circ S-15^\circ N$) is not statistically significant. The mean tropospheric ozone trend equals $-0.2 \pm 0.6$ DU decade\textsuperscript{-1} ($2\sigma$). Regionally, tropospheric ozone has a statistically significant increase of $\sim 3$ DU decade\textsuperscript{-1} over southern Africa ($\sim 1.5$ \% year\textsuperscript{-1}), the southern tropical Atlantic ($\sim 1.5$ \% year\textsuperscript{-1}), southeastern tropical Pacific Ocean ($\sim 1$ \% year\textsuperscript{-1}), and central Oceania ($\sim 2$ \% year\textsuperscript{-1}) and by $\sim 2$ DU decade\textsuperscript{-1} over central Africa ($2$–$2.5$ \% year\textsuperscript{-1}) and south India ($\sim 1.5$ \% year\textsuperscript{-1}). On the other hand, tropospheric O$_3$ decreases by $\sim 3$ DU decade\textsuperscript{-1} over the Caribbean sea and parts of the North Pacific Ocean ($\sim 2$ \% year\textsuperscript{-1}), and by less than 2 DU decade\textsuperscript{-1} over some regions of the southern Pacific and Indian Ocean ($\sim 0.5$ – $1$ \% year\textsuperscript{-1}).
1 Introduction

As is well known since the industrial revolution the earth’s population and its standard of living have grown dramatically. At the same time the urban population has grown. Since 2011, more than 50% of the world’s population live in urban areas and the population has now passed 7.5 Billion. In the past two decades, the population has grown by more than 2 Billion. An increasing population and standard of living inevitably leads to increased energy consumption, which is used in industry, transportation, and food production. These human activities release a large number of atmospheric pollutants which can be harmful to public health and/or vegetation and modify the terrestrial climate (Crutzen, 2002). Climate change may also impact air pollution events (WMO/IGAC, 2012). Tropospheric ozone (O₃) is regarded as one of the most important surface pollutants. This is because it oxidizes the biological tissues causing respiratory problems or even death (WHO, 2006), acts as a greenhouse gas (IPCC, 2007), and controls the oxidizing capacity of the troposphere (Jacob, 2000). O₃ in the troposphere is expected to increase by 60 to 80% by 2050 in Southeast Asia, India and Central America under the A2 IPCC (2013) scenario. However, the effects of climate change, especially the increased tropospheric temperatures and water vapor, may offset this increase by 10% to 17% (Stevenson et al., 2000; Grewe et al., 2001; Hauglustaine et al., 2005; IPCC, 2013).

Ozone is not directly emitted in the troposphere but it is a byproduct of the oxidation of volatile organic compounds (VOCs) in the presence of nitrogen oxides (NOₓ) and sunlight (Crutzen, 1970; Chameides and Walker, 1973). Young et al. (2013) estimated that 4877 ± 1706 (2σ) Tg of O₃ are chemically produced every year. Additionally, 477 ± 392 Tg · yr⁻¹ are transported from the stratosphere to the troposphere via the stratosphere to troposphere exchange (STE) (Holton and Lelieveld, 1996; Young et al., 2013). Tropospheric ozone loss is controlled by deposition to the Earth’s surface and chemical destruction, mainly by photolysis to atomic oxygen (O(¹D)), followed by the reaction of O(¹D) with water (H₂O) to produce two hydroxyl radicals (2OH) (Levy, 1972). The net chemical production (production minus loss) is estimated at 618 ± 550 Tg · yr⁻¹ (2σ) (Young et al., 2013; IPCC, 2013). The mean tropospheric ozone burden is 337±46 Tg (2σ) today, which is about 30% more than in 1850 (Young et al., 2013).

The sources of ozone precursors (VOCs and NOₓ) can be both of anthropogenic and natural origin. Various efforts towards reducing NOₓ and VOC emissions have been taken in developed countries, particularly in Europe and North America, leading to negative surface ozone trends on a local scale (Derwent et al., 2003; Cooper et al., 2014; Parrish et al., 2014). Nevertheless, tropospheric ozone pollution is a matter of global concern, because ozone and its precursors are transported from polluted areas to clean regions over continental distances and into the free troposphere through atmospheric dynamics, increasing the tropospheric ozone abundances over remote areas. For example, air masses originated from eastern China have increased ozone abundance over Japan and North America’s West Coast, despite the US legislation of reducing NOₓ emissions (Parrish et al., 2009; Cooper et al., 2010; Oltmans et al., 2013; Verstraeten et al., 2016). Additionally, the high tropospheric ozone amounts noticed over the south Atlantic ocean, the so-called "tropical Atlantic paradox", arise from ozone precursor emissions by biomass burning taking place in south America and Africa (Thompson et al., 2000; Diab et al., 2003).

The long-term evolution of tropospheric ozone is complex and depends upon the evolution of precursor emissions and climate change. As the predicted increase of trace gases emissions for the next years is mainly located over low latitudes (Grenfell
et al., 2003), long term observations of tropospheric ozone in the tropics should receive particular attention. Various studies have been performed in urban and rural sites using in situ data in order to estimate tropical tropospheric ozone trends. Lelieveld et al. (2004) noticed an increase in surface ozone in the order of 0.4 ppbv year\(^{-1}\) over the northeastern tropical Atlantic, 0.4 ppbv year\(^{-1}\) over the southeastern tropical Atlantic, and a smaller trend of 0.1 ppbv decade\(^{-1}\) over the southwestern tropical Atlantic Ocean, based on ship-borne measurements (1977–2002). Oltmans et al. (2013) observed an increase of 3.8 % decade\(^{-1}\) (0.16 ppbv year\(^{-1}\)) in surface ozone in Mauna Loa, Hawaii (19.5°N) in the North Pacific since 1974 and a smaller insignificant trend in the order of 0.7 % decade\(^{-1}\) (0.01 ppbv year\(^{-1}\)) in American Samoa (14.5°S) after 1976. Additionally, Cooper et al. (2014) report a significant increase of 0.19 ppbv year\(^{-1}\) in the subtropical site of Cape Point in South Africa from 1983 to 2011. Thompson et al. (2014) using ozonesonde data from the SHADOZ stations in Irene (25.9°S, 28.2°W) and Réunion (21.1°S, 55.5°W) noticed statistically significant trends in the middle and upper troposphere of ∼ 25 % decade\(^{-1}\) (1 ppbv year\(^{-1}\)) and ∼35–45 % decade\(^{-1}\) (2 ppbv year\(^{-1}\)) respectively during winter (June-August). Smaller positive trends appear, close to the tropopause in summer.

Satellite remote sensing is required to perform trend analysis up to global scale. One key challenge to retrieve tropospheric ozone column amounts from the measurements of satellite remote sensing instrumentation is the accurate subtraction of stratospheric ozone from the total column ozone. This requires accurate knowledge of the pressure/altitude level at which the tropopause is located. However in the tropics, where the the tropopause is not strongly modulated by frontal systems, the retrieval uncertainties due to the day to day variability of the tropopause can be reduced using monthly averages (Jensen, 2012). Most of the methods of estimating tropospheric ozone columns from space in the tropics derive from the residual approach (TOR) of Fishman and Larsen (1987) and Fishman et al. (1990). Later, more methods were developed such as the scan angle method from Kim (et al.), a modified residual method from Thompson and Hudson (1999), the convective clouds differential (CCD) from Ziemke et al. (1998), the cloud slicing (CS) technique from Ziemke et al. (2005), a modified trajectory enhanced tropospheric ozone residual method (TTOR) from Schoeberl et al. (2007) and Doughty et al. (2011), and the limb nadir matching (LNM) from Ebojie et al. (2014).

These methods have provided valuable datasets with which tropospheric ozone trends have been derived in the tropics. For example, Ziemke et al. (2005) using the CCD method on Total Ozone Mapping Spectrometer (TOMS) version 8 data from 1979 to 2003, found a statistically significant positive linear trend in the mid-latitudes but not in the tropics. Beig and Singh (2007) using the same data found an increasing trend of 7–9 % decade\(^{-1}\) over some parts of south Asia, 4–6 % decade\(^{-1}\) over the Bay of Bengal and 2–3 % decade\(^{-1}\) over the central Atlantic ocean and central Africa up to 2005. Kulkarni et al. (2010) using Tropospheric Ozone Residual (TOR) data from TOMS, SAGE and SBUV instruments, calculated statistically significant trends over three Indian mega-cities during 1979–2005. They showed that ozone increased by 3.4 % decade\(^{-1}\) in Delhi during the monsoon period, while it increased by 3.4–4.7 % decade\(^{-1}\) in Hyderabad and 5–7.8 % decade\(^{-1}\) in Bangalore during pre-monsoon and post-monsoon, respectively. One objectives of the SCIAMACHY proposal in 1988 Burrows et al. (1995) and references therein) was the retrieval of tropospheric ozone by making limb and nadir observation in the back scattered and reflected solar radiation. Ebojie et al. (2016) using the full record of SCIAMACHY limb-nadir matching data (2002-2011) retrieved regional and global tropospheric ozone trends. An insignificant positive trend on the order of 0.5 DU decade\(^{-1}\) was
noticed for the northern tropics (0-20°N) and in the order of 0.3 DU decade$^{-1}$ in the southern tropics (0–20°S). Regionally, they reported statistically significant trends of -1.6 % year$^{-1}$ over Northern South America (0–10°S, 75-45°W), of 1.6 % year$^{-1}$ in Southern Africa (5-15°S, 25-35°E), of 1.9 % year$^{-1}$ over Southeast Asia (15-35°N, 80-115°E), and a trend of 1.2 % year$^{-1}$ over Northern Australia (20-10°S, 100-130°E). Most recently, Heue et al. (2016) published a study about tropical tropospheric ozone trends using the CCD method on a harmonised dataset consisting of data retrieved from GOME, SCIAMACHY, GOME-2 and OMI satellite instruments from July 1995–December 2015 which are based upon different total ozone and cloud retrievals as well as merging approaches. The mean tropical tropospheric ozone trend that they found is 0.7 DU decade$^{-1}$ and regionally the trend reaches 1.8 DU decade$^{-1}$ near the African Atlantic coast, and -0.8 DU decade$^{-1}$ over the western Pacific. Seasonally, they found that the trend over the South African coast maximises in summer, whereas the negative trend over the southwest Pacific ocean maximises during autumn. As discussed earlier, the trend results from the various studies vary significantly, and in some cases they do not agree with each other, even though the same dataset was used.

Using a convective clouds differential (CCD) method, developed at the Institute of Environmental Physics (IUP)/University of Bremen and applied to retrievals of total ozone and cloud data from GOME/ERS-2 (1995–2003), SCIAMACHY/Envisat (2002–2012), and GOME-2/MetOp-A (2007–2015), new datasets of monthly mean tropical tropospheric columns of ozone (TTCO) have been created (Leventidou et al., 2016). The main differences between our CCD algorithm and the one developed by Heue et al. (2016) mainly originate from the different corrections that we have applied in the above cloud column calculation of GOME and GOME-2 data and handling of the outlier data (Leventidou et al., 2016). The main goal of this study is to derive long-term trends from our merged CCD tropical tropospheric ozone datasets. In a first step the three satellite data are merged into a consistent long-term dataset. Six possible approaches for merging the data are considered and evaluated by comparisons to SHADOZ ozonesondes and by trend evaluations (Section 2). The comparisons to ozonesonde, among other criteria, are used to identify the preferred merging scenario. The trend evaluation of the six merging scenarios will allow us to roughly estimate the contribution of the merging approach to trend uncertainties. In Section 3 the multiple linear regression model is briefly described. Detailed trend results for the tropics 15S -15N as well as for selected regions are presented in Section 4 for the preferred merged dataset. This paper ends with a summary and discussion (Section 5).

2 Harmonisation/merging of the TTCO datasets

2.1 Tropical tropospheric O$_3$ data

Monthly mean TTCO data have been retrieved as reported by Leventidou et al. (2016) using the Convective Clouds Differential (CCD) method on GOME (Burrows et al., 1999), SCIAMACHY (Burrows et al., 1995; Bovensmann et al., 1999), and GOME-2 (Callies et al., 2000) total ozone and cloud data from 1996 to 2015. These instruments have different properties such as spatial resolution, cloud algorithms, overpass time, etc. The individual TTCO datasets have been created taking into account these specific characteristics. The individual TTCO datasets have been separately validated with integrated (up to 200 hPa) tropospheric ozone columns by ozonesondes from the SHADOZ network (Thompson et al., 2003) (see: Leventidou et al. (2016)). The biases between them have been found to be within the uncertainties of the mean biases of 6 DU (1 sigma). Large
source of uncertainties in these comparisons are the low sampling of the sondes (typically less than five launches in a month) and the fact that CCD ozone is only derived as monthly means covering rather large areas (grid box). The uncertainty of the tropospheric ozone column retrieval with the CCD method is in the order of 3 DU (∼10%). For most of the stations, the bias with the ozonesondes is within the retrieval uncertainty, with the exception of GOME-2 TTCO which is in the order of 5 DU.

Finally, the CCD TTCO from SCIAMACHY data have been compared with the Limb-Nadir-Matching (LNM) tropospheric O₃ columns up to 200 hPa altitude from the same satellite instrument, showing that the bias and the RMS values are within the ones calculated for the comparison with ozonesondes.

2.2 Correction offsets between GOME and GOME-2 with respect to SCIAMACHY TTCO

For trend calculations the existence of a constant bias (in clouds and ozone) between the instruments, caused by the spatial and temporal differences of the individual instruments, can be removed by using a suitable merging approach as will be shown here. Correction offsets have been calculated in order to create one consistent tropical tropospheric columns dataset from the CCD method for the whole timespan of the operation of the European satellites (1996–2015). SCIAMACHY TTCO were used as the reference, because SCIAMACHY is the only instrument that overlaps (2002-2012) both with GOME and GOME-2 and has the smallest bias with respect to the ozonesondes (< 2 DU). The average difference (bias) for each grid-box during the common years of the instruments operation (2002 for SCIAMACHY–GOME and 2007-2012 for SCIAMACHY–GOME-2) was computed and applied (added) to GOME and GOME-2 TTCO data. The mean biases, shown in Fig. 1, range between -6 and 6 DU for GOME, with positive differences (3–6 DU) located mainly over land. There are also two regions with positive biases appearing north of 7.5°N until 20°N, and between -5 and -7.5°S. For GOME-2, the bias ranges between -8 and 0 DU, with the biases being smaller over land, especially over south America and north/central Africa. Possible reasons for the biases are the different cloud algorithms used for each instrument (SACURA for SCIAMACHY and FRESCO for GOME and GOME-2) and the small biases noticed in the total ozone columns (e.g. ∼ -2.5 DU between SCIAMACHY and GOME-2). Differences in spatial resolution and overpass time of the instruments have also minor contributions in the biases.

The latitudinal dependence of the mean bias is shown at the bottom of Fig. 1. The average differences between GOME and GOME-2 with SCIAMACHY are generally negative (less than 5 DU) in all latitude bands with the exception of the northern tropical latitudes, where GOME mean biases are positive (0–2 DU). GOME mean biases have stronger latitudinal variability than those of GOME-2. This behavior may be explained by the short time of common operation (Jan. 2002–Jun. 2003) between GOME and SCIAMACHY instruments. The 1σ standard deviation (uncertainty bars) of the mean bias per latitude band is comparable to the magnitude of the biases, ranging from less than 5 DU close to the equator to 7 DU for latitude bands close to the tropical borders. For the case of GOME, the mean correction offset is -1.2 DU, whereas for GOME-2, it is -5.7 DU.

The mean offset of GOME-2 is almost twice the CCD retrieval uncertainty (∼3 DU). For this reason and because of the large biases with the ozonesonde data, it seems reasonable to apply a correction for the GOME-2 TTCO dataset.

The drift on the average differences (bias), β, has been estimated using a simple linear regression model such as: \( Y = \alpha + \beta \cdot X_t \), where \( Y \) is the time-series of the biases, \( X_t \) is the time variable in months, and \( \alpha \) is the offset. The drift between SCIAMACHY and GOME-2 is shown in Fig. 2. There are not enough overlapping years to calculate a trend in the GOME-
Figure 1. Correction offsets using SCIAMACHY TTCO as reference. (up) Correction offset for GOME: average difference of GOME from SCIAMACHY TTCO for the years 2002-2003. (down) Correction offset for GOME-2: average difference of GOME-2 from SCIAMACHY TTCO (in DU) for the years 2007-2012. On the right are shown the biases per latitude band. The error bars denote the 1σ standard deviations of the latitudinally averaged biases.
Figure 2. Top: drift in the correction offset for GOME-2. Black "x" denotes statistically significant trend. Bottom: average difference and drift in the correction offset for GOME-2 between 2007 and 2012.

SCIAMACHY difference time-series. The drift is generally less than $\sim 0.4$ DU per year and is statistically not significant ($\beta/\sigma_\beta < 2$ (Weatherhead et al., 1998; Wilks, 2011)) for nearly all grid boxes, with the exception of the 17.5–20 °N latitude band, where it is statistically significant and exceeds 1 DU year$^{-1}$. During local winter months at the tropical borders, there are often missing TTCO data owing to the movement of the ITCZ and the inability to retrieve a reliable stratospheric $O_3$ column. For this reason, calculated drifts for these latitudes are not reliable, in spite of the fact that they might appear to be statistically significant. Consequently, the trend of the correction offsets is considered to be negligible.

2.3 Six Harmonisation scenarios

The creation of a consistent tropical tropospheric ozone column dataset from multiple satellite instruments demands a careful selection of the optimal harmonisation approach, since it introduces additional uncertainty in the merged dataset. Six harmoni-
isation scenarios have been tested. They all use the SCIAMACHY TTCO dataset as a reference, which is in the middle of the time period, as follows:

- **Scenario 1**: No correction applied to GOME data (which maybe justified by the very short overlap period), while GOME-2 is corrected using for each grid-box the mean bias with respect SCIAMACHY for the common years of operation (2007–2012 for GOME-2).

- **Scenario 2**: No correction applied to GOME data and the average bias (-5.7 DU) with respect to SCIAMACHY is added to all GOME-2 TTCO data.

- **Scenario 3**: GOME and GOME-2 have been corrected using for each grid-box the mean bias with respect to SCIAMACHY for the common years of operation.

- **Scenario 4**: The average bias with respect to SCIAMACHY (-1.2 DU) is added to all GOME TTCO data, whereas GOME-2 TTCO has been corrected using for each grid-box the mean bias with respect to SCIAMACHY for the common years of operation (2002 for GOME and 2007-2012 for GOME-2).

- **Scenario 5**: The average bias with respect to SCIAMACHY (-1.2 DU) for GOME and for GOME-2 (-5.7 DU) is added to all GOME and GOME-2 TTCO data respectively.

- **Scenario 6**: No correction applied to GOME, whereas for GOME-2 both the bias and the drift is included in the correction of GOME-2 TTCO in each grid-box.

After the correction terms for all scenarios have been applied to the original data, the "corrected" GOME (1996-2002) and GOME-2 (2007-2015) TTCO were averaged with the ones from SCIAMACHY (2003-2012) for the overlapping months (Jan. 2002–Jun. 2003 and Jan. 2007–Dec. 2012, respectively).

In order to decide which is the most suitable harmonisation scenario, the various merged datasets were compared with integrated ozone columns up to 200 hPa altitude from nine ozonesonde stations: (a) Ascension (8°S, 14.4°W), b) Paramaribo (5.8°N, 55.2°W), c) Java (7.6°S, 111°E), d) Natal (5.4°S, 35.4°W), e) Samoa (14.4°S, 170.6°W), f) Nairobi (1.4°S, 36.8°E), and g) Kuala Lumpur (2.7°S, 101.7°E). Fiji (18.1S, 178.4E)) station is not included in the comparison because it is highly influenced by air coming in from mid-latitudes and the upper troposphere (Thompson et al., 2017). Hilo (19.4N, 155.4W) is strongly affected by volcanic outgassing which interferes with the ozonesondes’ electrochemical concentration cells (ECC), resulting in negligible ozone concentrations being measured in the boundary layer (Morris et al., 2010). Therefore, this station is also not included. As seen in Table 1, the mean bias between the six harmonised TTCO datasets and the ozonesondes range between -1.1 and 0.9 DU which is well within the retrieval uncertainty showing that for most scenarios the spatio-temporal offsets with respect to ozonesondes are minimised. However, the biases of each scenario with ozonesondes are very close to each other for every station. The same occurs for the correlation between the harmonised TTCO datasets and the ozonesondes (not shown here). Although the comparison between the TTCO from the individual harmonised scenarios and the ozonesonde data does not favor clearly any harmonisation scenario, the scenarios that can be confidently rejected are scenarios 3, 4 and 5.
Table 1. Mean differences (in DU) between merged TTCO data, retrieved with the CCD method using six possible harmonisation scenarios, with integrated ozone columns up to 200 hPa from nine SHADOZ stations. The stations marked with asterisk present data from the newest reprocessed (V05.1_R) version (Thompson et al., 2007; Witte, 2017). The regions where the merged scenarios have the smallest biases with the ozonesondes are marked with bold. Scenario 1 has the smallest mean bias for all the stations.

<table>
<thead>
<tr>
<th>Site</th>
<th>scenario 1</th>
<th>scenario 2</th>
<th>scenario 3</th>
<th>scenario 4</th>
<th>scenario 5</th>
<th>scenario 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Am. Samoa (14.4S,170.6W)*</td>
<td>-0.89</td>
<td>-0.92</td>
<td>-1.99</td>
<td>-0.61</td>
<td>-0.93</td>
<td>4.59</td>
</tr>
<tr>
<td>Ascension (8S,14.4W)</td>
<td>0.03</td>
<td>-0.14</td>
<td>-0.77</td>
<td>-0.42</td>
<td>-0.60</td>
<td>0.03</td>
</tr>
<tr>
<td>Java (7.6S,111E)</td>
<td>-0.11</td>
<td>-0.12</td>
<td>-1.12</td>
<td>-0.54</td>
<td>-0.55</td>
<td>-0.11</td>
</tr>
<tr>
<td>Kuala Lumpur (2.7N,101.7E)</td>
<td>-1.81</td>
<td>-2.12</td>
<td>-2.12</td>
<td>-2.14</td>
<td>-2.48</td>
<td>-1.78</td>
</tr>
<tr>
<td>Nairobi (1.3S,36.8E)</td>
<td>1.81</td>
<td>1.10</td>
<td>1.80</td>
<td>1.48</td>
<td>0.74</td>
<td>1.84</td>
</tr>
<tr>
<td>Natal (5.4S,35.4W)</td>
<td>0.56</td>
<td>0.63</td>
<td>-0.21</td>
<td>0.22</td>
<td>0.28</td>
<td>0.57</td>
</tr>
<tr>
<td>Paramaribo (5.8N,55.2W)*</td>
<td>-2.98</td>
<td>-2.95</td>
<td>-3.02</td>
<td>-4.11</td>
<td>-4.34</td>
<td>-0.11</td>
</tr>
<tr>
<td>Mean bias for all stations</td>
<td>-0.48</td>
<td>-0.64</td>
<td>-1.06</td>
<td>0.87</td>
<td>-1.13</td>
<td>0.72</td>
</tr>
</tbody>
</table>

where GOME data are corrected with respect to SCIAMACHY since the overlap period between GOME and SCIAMACHY is very short (10 months, 8/2002-6/2003). Scenario 6 can also be rejected due to the fact that the drift in GOME-2 correction offset at 81% of the grid-boxes is statistically non significant. Lack of significant drifts in the comparison between GOME-2 and SCIAMACHY over the overlapping period shows that the data records are quite stable. Scenarios 1 and 2 have the smallest bias with respect ozonesondes, however both show some differences in regional trends with scenario 1 having larger regions with statistically significant trends, while scenario 2 shows mostly zero trends across the tropics within their uncertainties (see: Supplement). The main difference between scenario 2 and one is that the former uses a single (global) bias correction for GOME-2, while scenario 1 biases are corrected individually for each grid-box, which we believe makes physically more sense given the large overlap period. For these reasons, scenario1 (no drift corrections and bias correction for GOME-2) which also has the smallest mean bias with the ozonesondes (-0.4 DU) has been selected to be the preferred harmonisation scenario for merging the TTCO datasets. Before we discuss in details tropical tropospheric trends using the preferred scenario (Section 5), we try to estimate the potential contribution of the merging approaches to trend uncertainties in tropical tropospheric ozone.

2.4 Sensitivity of the trend to the merging approach

The statistical trend uncertainty derived from a single dataset usually does not account for uncertainties due to the merging approach applied. Here we will provide a rough estimate on how large the trend uncertainties may be. We applied the multi-variate linear regression model (see Section 3, Eq. 1 for details on the regression) to derive trends from all six merged datasets. The tropospheric $O_3$ trends from all scenarios range between $\sim-4$ and 4 DU decade$^{-1}$, with mean values between 0 and 0.8 DU decade$^{-1}$, without any of them being statistically significant for the global tropics (see Fig. S1 in the supplement). The maximum trend difference among all six harmonisation scenarios is on average 2 DU decade$^{-1}$ exceeding the $2\sigma_\beta$ uncertainty.
of the trends which is \( \sim 1.2 \) DU decade\(^{-1}\) (see Fig. S2 in the supplement). These differences in the trends among the differently harmonised datasets reveal the additional uncertainty which results from the harmonisation procedure of multiple TTCO datasets.

3 The multi-linear regression trend model

Changes in ozone precursor emissions due to urbanization and land use, along with changes in the atmospheric dynamics which impact tropical upwelling or the horizontal ozone transport, may cause long-term changes in the tropospheric ozone burden. This in turn impacts the photochemical ozone production and loss in the troposphere (Ziemke and Chandra, 2003; Solomon et al., 2007; Chandra et al., 2009; Voulgarakis et al., 2010; WMO, 2011; Neu et al., 2014; Monks et al., 2015). Some of these factors can be represented by periodic seasonal proxies, such as the El Niño Southern Oscillation (ENSO), the quasi-biennial oscillation (QBO), and the solar cycle (SC). These indexes are embodied in the trend model described here.

The time series of the monthly mean tropical tropospheric ozone columns \( Y_t \) at a specific latitude and longitude \((i,j)\) (running every 2.5\(^\circ\) and 5\(^\circ\), respectively) is described by the following trend model:

\[
Y_t(i,j) = \alpha(i,j) + \beta(i,j) \cdot X_t + S_t(i,j) + R_t(i,j) + N_t(i,j),
\]

(1)

where \( \alpha \) is the offset for the first month \( t=1 \), \( \beta \) the linear trend in DU month\(^{-1}\), \( X \) the time variable (months running from zero to 239) covering the years 1996–2015, \( S_t \) is the seasonal variation, \( R_t \) are the terms with the various proxies (ENSO, QBO, solar cycle) and \( N_t \) is the noise of the time series, representing the unexplained portion of the variability in the fit. The seasonal cycle is modeled by a Fourier series (see Eq. 2), with \( \gamma_{11}, \gamma_{21}, \gamma_{12}, \gamma_{22}, \gamma_{13}, \gamma_{23} \) being the regression coefficients for 12-, 6- and 4-month periodicities, with sine and cosine terms for each periodicity, respectively:

\[
S_t(i,j) = \sum_{n=1}^{3} (\gamma_{1n} \cdot \sin(\frac{2 \cdot \pi \cdot n \cdot t}{12}) + \gamma_{2n} \cdot \cos(\frac{2 \cdot \pi \cdot n \cdot t}{12})).
\]

(2)

\( R_t \), represents the time dependent regression coefficients for the ENSO, QBO, and solar cycle proxies which can be expressed as:

\[
R_t = \delta \cdot ENSO_t + \varepsilon \cdot QBO_{30t} + \zeta \cdot QBO_{50t} + \eta \cdot SC_t.
\]

(3)

Because the tropospheric ozone lifetime approaches a month, the pattern of tropospheric ozone for a month has the tendency to persist into the next month. Even after removing the seasonal and other effects in the time series shown in Eq. 1, there is still a month-to-month correlation (\( \phi \)) in residuals. This phenomena is called persistence (Wilks, 2011) and is quantified by the degree of autocorrelation of a parameter, shifted by p time steps (lag p). Therefore, the first order autocorrelation of the noise (AR[1]) is included in the model, as explained by Weatherhead et al. (1998).
4 Tropical tropospheric ozone trends

For the rest of the discussion about tropical tropospheric ozone, the trend refers to the preferred harmonisation scenario (scenario 1).

4.1 Tropical distribution of tropospheric O3 trends and mean tropical trend

Figure 3 summarises the tropical tropospheric ozone trends calculated in a $2.5^\circ \times 5^\circ$ grid as derived from the preferred merged CCD TTCO dataset using the multivariate regression model (Eq. 1) between 1996 and 2015. As shown in 3a, the trend varies between -3.2 and 3.7 DU decade$^{-1}$, and the average trend for the period 1996–2015 is statistically not-significant and equal to $-0.1 \pm 1.2$ DU decade$^{-1}$ ($2\sigma$). Fig. 3b shows the $2\sigma$ of the trend, which is in the order of $\sim0–4$ DU decade$^{-1}$ (mean: 1.2 DU decade$^{-1}$), with higher values at the tropical borders and values close to zero along the equator. Fig. 3c shows the correlation between the model and the time-series. The correlation coefficient reaches 1 over the north and central-east Pacific and the southern Atlantic Ocean. The regions of smaller correlations are mostly over the west Pacific, the Caribbean sea, the south-east Asia, and over the central African continent. The main reason for the low correlation is the weak seasonal cycle observed in these regions. Fig. 3d shows the RMS between the time-series and the model fit. The RMS is less than 3 DU close to equator and reaches 7 DU at the tropical borders. Fig. 3e presents only those grid boxes where the trend is statistically significant and exceeds the maximum difference of the trends calculated from all six scenarios. This additional criterion (to exceed the differences between harmonisation scenarios) allows us to identify grid boxes that have significant trends with higher confidence. Using this stricter criterion, tropospheric ozone trends are positive over some parts of central Africa ($\sim2$ DU decade$^{-1}$), southern Africa and Atlantic Ocean ($\sim2–3$ DU decade$^{-1}$), India ($\sim2$ DU decade$^{-1}$) and Oceania ($\sim3–4$ DU decade$^{-1}$) but are negative over the Caribbean sea and parts of North Pacific Ocean ($\sim2–3$ DU decade$^{-1}$), as well as over some regions of the southern Pacific Ocean ($\sim-2$ DU decade$^{-1}$) seem to be relevant, however, for all other grid boxes trends are highly uncertain and mainly dependent on the choice of the harmonisation scenario. The negative trends appearing in a region at the northern latitudes (Caribbean sea and northern Pacific) may be an artifact of the data-set (low sampling of data, 54 out of 240 months of data). Finally, Fig. 3f shows the tropical tropospheric ozone trends in per cent per year ($\%$ year$^{-1}$) that are statistically significant. Here the maximum increase is observed over central Africa, $\sim3\%$ year$^{-1}$, over southern Africa, south tropical Atlantic and Oceania $\sim1.5\%$ year$^{-1}$, and finally over India and south-east Asia $\sim1\%$ year$^{-1}$. The maximum tropospheric ozone decreasing trend is observed over the Caribbean sea and the north-east tropical Pacific, about $\sim2\%$ year$^{-1}$, followed by the central-south Pacific and Indian Ocean, $\sim1\%$ year$^{-1}$.

The southern and northern boundary of the tropics ($15^\circ$–$20S$ and $15^\circ$–$20N$) is strongly influenced by stratospheric intrusions via tropopause foldings and air masses being transported from the mid-latitudes and the upper troposphere (Pickering et al., 2001; Thompson et al., 2017). Therefore, in order to estimate a more reliable mean trend for the tropics the multivariate regression model (1) has been applied to the mean tropical time-series between $15^\circ$–$15^\circ$. The fit results are shown in Fig. 4. The mean ("global") tropical trend equals $-0.2 \pm 0.6$ DU decade$^{-1}$ (2 sigma). This means that there is no significant trend for tropospheric ozone in the tropics. The mean tropospheric ozone trend is in agreement with Ziemke et al. (2005) (using
Figure 3. (a) Tropical tropospheric ozone trends using a linear multivariate first order auto-regression model for the selected harmonised scenario 1 in DU decade$^{-1}$. Grid-boxes marked with "x" are statistically non-significant at the 95% confidence level (b>2$\sigma$). b) 2$\sigma$ standard deviation of the trend. c) The correlation coefficient, R, between the multi-linear trend model fit and the original time-series. d) The RMS error between the trend model and the time-series. e) The statistically significant trend that exceeds the maximum absolute difference of the trends calculated for all six scenarios. f) The significant tropical tropospheric ozone trend in % year$^{-1}$. 
solar backscatter ultraviolet (SBUV) and Total Ozone Mapping Spectrometer (TOMS) version 8 data from 1979 to 2003) and Ebojie et al. (2016) (using SCIAMACHY limb-nadir-matching (LNM) observations during the period 2003–2011) who also indicated insignificant and near zero global trends in the tropics, although their analysis was based on different datasets and covered shorter time periods. Nevertheless, Heue et al. (2016), using a similar CCD method on the same period and satellite instruments, reported a significant average increase of $0.7 \pm 0.1$ DU decade$^{-1}$.

\[\text{mean trend from 15°S to 15°N}\]

![Graph showing mean trend from 15°S to 15°N for the period 1996 to 2015.](image)

**Figure 4.** Mean tropical tropospheric ozone trends between 15°S – 15°N for the period 1996 to 2015. Top: The multivariate linear trend (black), the fit (red) and the residual (orange) are over-plotted. The mean tropical tropospheric ozone trend is equal to -0.18 and the 2σ uncertainty of the trend is ±0.62 DU decade$^{-1}$. The next panels show the harmonic functions (green), ENSO (light blue), QBO (red), solar (orange). Overlaid in black for all proxies are the time series with all fit terms removed except the particular fit parameter.

The tropical mean tropospheric ozone time-series (black stars) shows a seasonal cycle with higher values in Jul–Oct. The time-series are well followed by the regressed tropospheric ozone (red line) and the residual (orange line in upper panel) is less than 5 DU. The seasonal cycle contributes the most to the TTCO variability in the tropics by about ±2 DU. Tropical tropospheric ozone is reduced by 4 DU during El Niño years (1997-98, 2006-07, 2009-10, 2015) and slightly increases by 1 – 2 DU during strong La Niña years (1999-00, 2007-08, 2010-11). QBO and the solar cycle, do not contribute to the inter-annual
mean tropical tropospheric ozone variability. Overlaid in black for all proxies are the time series with all fit terms removed except the particular fit parameter. This allows us to relate the magnitude of changes due to a selected process to the observed residuals (or unexplained variations).

4.2 Regionally averaged tropical tropospheric ozone trends

We also studied regional trends focusing on the regions where the trends are statistically significant. The TTCO have been regionally averaged for eight regions and the regression analysis applied to them. The regions are: A: Caribbean Sea (15° – 17.5°, -85° – -45°), B: India(10° – 20°, 70° – 85°), C: north-south America (0° – 10°, -75° – -60°), D: North Africa (5° – 15°, -17.5° – 50°), E: east Pacific Ocean (0° – 7.5°, -180° – -110°), F: Indian Ocean (0° – 7.5°, 50° – 100°), G: west Pacific Ocean (0° – 7.5°, 160° – 180°), and H: southern Africa (-20° – -12.5°, 10° – 50°).

Table 2. Regional tropospheric ozone trends in 8 tropical regions. Bold are the regions where the trend is greater than three times the standard deviation of the trend (3σ).

<table>
<thead>
<tr>
<th>Area</th>
<th>Tropospheric O3 trend ±2σ in DU decade⁻¹</th>
</tr>
</thead>
<tbody>
<tr>
<td>A) Caribbean sea</td>
<td>-1.59 ± 1.30</td>
</tr>
<tr>
<td>B) India</td>
<td>1.10 ± 0.86</td>
</tr>
<tr>
<td>C) North South America</td>
<td>0.99 ± 0.94</td>
</tr>
<tr>
<td>D) North Africa</td>
<td>1.54 ± 1.09</td>
</tr>
<tr>
<td>E) East Pacific Ocean</td>
<td>-1.21 ± 0.65</td>
</tr>
<tr>
<td>F) Indian Ocean</td>
<td>-1.61 ± 0.83</td>
</tr>
<tr>
<td>G) West Pacific Ocean</td>
<td>-1.87 ± 0.72</td>
</tr>
<tr>
<td>H) South Africa</td>
<td>1.44 ± 1.28</td>
</tr>
</tbody>
</table>

As shown in Figure 5 and Table 2, regions B, C, D and H show significant increase in the order of 1–1.5 DU decade⁻¹ and regions A, E, F, and G a significant ozone decrease in the order of 1.2 – 1.9 DU decade⁻¹. The observed significant positive changes in tropospheric O₃ over north Africa and parts of the Arabian sea (D), south Africa and the southern African outflow (H), parts of India (B), and north south America (C) agree well with results of Lelieveld et al. (2004), Beig and Singh (2007), Kulkarni et al. (2010), Ebojie et al. (2016) and Heue et al. (2016) who also observed an increasing ozone trend over these regions. Although, Ebojie et al. (2016) observe a decreasing trend of -0.5 DU decade⁻¹ in tropospheric ozone over north-east Africa (D).

The negative changes in TTCO over the Caribbean sea (A) are in agreement with the results of Ebojie et al. (2016). However, the observed trends over the northern and southern tropical latitudes (18° –20° in SH and NH) should be generally interpreted
Figure 5. Tropical tropospheric ozone trend in A) central America, B) India, C) east Pacific Ocean, D) South America, E) central Atlantic Ocean, F) Indian Ocean, G) south Atlantic Ocean, and H) southern Africa.
with caution because they are influenced by low sampling of data due to the movement of the ITCZ, which reduces the cloudy data during local winters and makes the above cloud ozone column (ACCO) retrieval difficult, violating in some cases the invariance of the ACCO per latitude band. The decreasing tropospheric ozone trend over the western Pacific (G) and Indian (F) Oceans agrees well with Heue et al. (2016). On the other hand, the decreasing trend over the eastern Pacific Ocean (E) is in disagreement with Heue et al. (2016) who reported a significant increase in the order of 0.5 – 1 DU decade\(^{-1}\).

4.3 **Seasonal tropospheric O\(_3\) trends**

![Figure 6](image.png)

*Figure 6.* Tropical tropospheric ozone trends for winter (DJF), spring (MAM), summer (JJA), and autumn (SON) for the years 1996 to 2015. Black "x" denotes statistically significant trend.

Seasonal tropospheric O\(_3\) trends can be useful for understanding the connection between the factors (e.g. meteorology or emissions) that contribute to tropospheric ozone changes and its distribution. For this reason, the multi-linear regression model has been applied to monthly time-series containing only the following months: Dec.–Feb., Mar.–May, Jun.–Aug., and Sep.–Nov respectively. For these time series no seasonal terms are used in the regression. In Fig. 6, the maximum decreasing trends appear during December to February over the northern tropical Atlantic and Pacific Oceans (~4 DU decade\(^{-1}\)). These air masses are more affected by changes occurring in the mid-latitudes due to the southward movement of the ITCZ in these months and the strong westerly air flow over the tropical borders (Oltmans et al., 2004). Therefore, it is possible that changes in ozone precursors, such as NO\(_2\) over North America and Europe may have affected the O\(_3\) trends over these tropical latitudes (Logan et
This decrease might also be associated with the limited number of TTCO measurements on the northern tropical borders, thus it demands a more careful investigation. The trends are mostly insignificant between March and May, with the exception of Africa where ozone is increasing by $\sim 1$ DU decade$^{-1}$ and some parts over South America where ozone is decreasing by less than 1 DU decade$^{-1}$. During June to August, ozone shows a small statistically significant decrease over the Pacific and Indian Oceans (1-2 DU decade$^{-1}$). Possible reasons for tropospheric ozone decrease over the oceans may be related to changes in sea surface temperatures (SSTs), which are closely tied to the tropospheric humidity (Trenberth, 2011; IPCC, 2007). As discussed earlier, the production of HOx (OH and OH$_2$) from water vapor in the troposphere accounts for one of the most important sinks of tropospheric ozone (Jacob, 2000). An increase in vertical convective patterns over the tropical oceans may result in lower ozone mixing ratios in the upper troposphere where the WFDOAS retrieval is more sensitive (Morris et al., 2010; Wai et al., 2014; Fontaine et al., 2011; Ziemke et al., 2008; Coldewey-Egbers et al., 2005). Several studies have shown that the total column of water vapour (TCWV) has increased over the tropics. Mieruch et al. (2014) and Trenberth et al. (2005) found that the TCWV has increased by $\sim 1-2$% decade$^{-1}$ over the oceans. Chen and Liu (2016) found that also the precipitable water vapor (PWV) increased by 1-2% in the tropics between 1992–2014. The precipitation increase is about 4% over the ocean, while a decrease of 2% is found over land in the latitude range 25$^\circ$S to 25$^\circ$N, between 1979 and 2001 (Adler et al., 2003). The significant positive trend of ozone at the southern tropical Atlantic, southern Africa, South America, and Oceania maximise during September to November ($\sim 4$ DU decade$^{-1}$). According to MODIS/TERRA Fire Radiative Power (mW/m$^2$) data (https://disc.gsfc.nasa.gov/neespi/data-holdings/mod14cm1.shtml) boreal autumn is the season with the most intense fires over southern Africa and South America. The burned area in southern tropical Africa increased by 1.8%/yr during the period 2000 to 2011 Giglio et al. (2013). Ziemke et al. (2009b) and Wai et al. (2014) estimated that biomass burning can contribute to an increase in tropospheric ozone column by $\sim 20\%$. Hence, it is very likely that biomass burning could be the origin of the observed ozone increase.

5 Summary and discussion

The new harmonised dataset of tropical tropospheric ozone columns for the last 20 years between 1996 and 2015, makes it possible to calculate and study long-term tropospheric O$_3$ variability and trends. Correction offsets have been calculated for GOME and GOME-2 TTCO using SCIAMACHY as reference (in the middle of the time-series) in order to reduce the instrumental effects in the long-term time series. Nevertheless, the short overlap period between GOME and SCIAMACHY limits the harmonisation of the GOME dataset. The correction offsets for GOME presented artificial features which are also visible afterwards in the trend (see Fig. S1). In order to identify the best way to merge the CCD data and also to investigate how the harmonisation approach may affect the observed trends, six different harmonisation scenarios have been evaluated by comparing with ozonesondes. The merging scenario, using no correction for GOME (short overlap) and the mean bias correction of GOME-2 with respect to SCIAMACHY in each grid box was found to show slightly smaller differences to ozonesondes and therefore, was considered to be the preferred scenario. From the trend analysis of all merged datasets a rough
estimate on the variability of trends due to merging approaches was provided (∼ ± 2 DU decade\(^{-1}\)). After the harmonisation, the data obtained from the different instruments agree better with each other and with ozonesondes.

Harmonisation and merging of multi-instrument datasets is one of the largest sources of uncertainty. Most of the trend studies that use multiple satellite data (e.g. Xu et al. (2011), Loyola et al. (2009), Heue et al. (2016), and TOAR) do not account for uncertainties related to the merging approach. Therefore, in order to quantify the uncertainty due to harmonisation, multi-linear tropospheric ozone trends using all six harmonised datasets have been derived and the maximum deviation between them has been calculated. The trends range between about -4 and 4 DU decade\(^{-1}\) and the average difference between the trends from the six scenarios has been found to be ∼2 DU decade\(^{-1}\), exceeding locally the 2\(\sigma\) of the individual trends (0 to 4 DU decade\(^{-1}\)). We conclude that the overall uncertainties in the trends are larger than the statistical ones reported.

Despite the fact that the trend results using the preferred merged dataset are small (< ±4 DU decade\(^{-1}\) or 3 % year\(^{-1}\)) and mostly uncertain (66 % are statistically insignificant), there are regions such as over southern Africa, the southern tropical Atlantic, south-east tropical Pacific Ocean, and central Oceania where tropospheric O\(_3\) increased significantly by ∼3 DU decade\(^{-1}\). In central Africa and southern India, tropospheric ozone increased by ∼2 DU decade\(^{-1}\). Regional positive tropospheric ozone trends of similar magnitude were also observed in other studies (e.g. Lelieveld et al., 2004; Beig and Singh, 2007; Kulkarni et al., 2010; Cooper et al., 2014; Ebojie et al., 2016; Heue et al., 2016). On the other hand, tropospheric O\(_3\) decreases by ∼ 3 DU decade\(^{-1}\) over the Caribbean sea and parts of North Pacific Ocean, and by less than 2 DU decade\(^{-1}\) over some regions of the southern Pacific Ocean. The most important limitation in interpreting the observed trends over the northern and southern tropical latitudes (18°–20° in SH and NH) is the low data sampling at these latitudes. Due to the ITCZ movement, cloudy data during local winters are reduced, making the above cloud ozone column (ACCO) retrieval difficult or violating the invariance of the ACCO per latitude band. Therefore, even though they might appear to be statistically significant, they should be referred to with caution.

The mean tropospheric ozone trend has been estimated between 15S and 15N during the period 1996–2015. This restriction has been applied in order to avoid the influence of sub-tropical air masses on tropospheric ozone abundances at the tropical borders (Thompson et al., 2017). The mean trend for global tropics (15°S – 15°N) is found to be almost equal to zero (-0.2±0.6 DU decade\(^{-1}\)) and statistically non significant. This is in agreement with studies of Ziemke et al. (2005) (nearly zero trend) and Ebojie et al. (2016) (0.3±0.4 % year\(^{-1}\) for the southern tropics and 0.1±0.5 % year\(^{-1}\) for the northern tropics) who also found no trend or insignificant trends. This is in contrast with the results of Heue et al. (2016) who found a mean increase of 0.7±0.1 DU decade\(^{-1}\) for the entire tropics.

Comparison of several independent studies conducted on tropospheric ozone trends shows that the trends vary in sign and magnitude for the past few decades in the tropics (Cooper et al., 2014; Ziemke et al., 2005; Monks et al., 2015; Oltmans et al., 2013; Lelieveld et al., 2004; Lin et al., 2014; Beig and Singh, 2007; Kulkarni et al., 2010; Thompson et al., 2014; Heue et al., 2016; Ebojie et al., 2016). This is a significant issue for the scientific community, especially climate modelers who try to use recent past data to evaluate the performance of climate and global atmospheric chemistry models for future prediction (Zhang et al., 2016; Young et al., 2018). At the moment, there is a new activity of the International Global Atmospheric Chemistry Project (IGAC), named Tropospheric Ozone Assessment Report (TOAR), which aims to assess our knowledge of
the tropospheric ozone distribution, pattern and trends, using the available surface ozone data, ozone sonde, aircraft and satellite observations (currently under review in Elementa: https://collections.elementascience.org/toar/).

The accurate interpretation of the trend results is challenging and requires the parallel investigation of changes in numerous factors that impact on ozone production, loss, and transport in the troposphere, including various feedbacks (e.g. Cooper et al., 2014, and references therein) Finally, the attribution of observed TTCO trends in specific regions to the various processes is not possible without the additional use of chemistry-transport models that can potentially disentangle the different contributions to tropospheric ozone variability (dynamics and chemistry)(Grewe et al., 2012; Coates et al., 2015).

The launch of Sentinel 5 precursor(S5p) satellite in 2017 and the planned launches of three consecutive Sentinel 5 instruments until 2030 will extend the TTCO record which will likely result in more reliable trends. The grid box size used in this study was relatively coarse (2.5° × 5° degrees), due to the instruments spatial resolution (GOME pixel ∼320 km), and in order to remove the residual noise. The high spatial resolution (7×7 km) of the TROPOMI instrument aboard S5p will improve the trend estimates of tropospheric ozone in particular over mega-cities.

6 Data availability

Data used in this publication can be accessed via the IUP website: http://www.iup.uni-bremen.de/UVSAT/datasets or by contacting the corresponding author.

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