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A multi-model intercomparison of halogenated very short-lived substances (TransCom-VSLS): linking oceanic emissions and tropospheric transport for a reconciled estimate of the stratospheric source gas injection of bromine


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Abstract. The first concerted multi-model intercomparison of halogenated very short-lived substances (VSLS) has been performed, within the framework of the ongoing Atmospheric Tracer Transport Model Intercomparison Project (TransCom). Eleven global models or model variants participated (nine chemical transport models and two chemistry-climate models) by simulating the major natural bromine VSLS, bromoform (CHBr$_3$) and dibromomethane (CH$_2$Br$_2$), over a 20-year period (1993–2012). Except for three model simulations, all others were driven offline by (or nudged to) reanalysed meteorology. The overarching goal of TransCom-VSLS was to provide a reconciled model estimate of the stratospheric source gas injection (SGI) of bromine from these gases, to constrain the current measurement-derived range, and to investigate inter-model differences due to emissions and transport processes. Models ran with standardised idealised chemistry, to isolate differences due to transport, and we investigated the sensitivity of results to a range of VSLS emission inventories. Models were tested in their ability to reproduce the observed seasonal and spatial distribution of VSLS at the surface, using measurements from NOAA’s long-term global monitoring network, and in the tropical troposphere, using recent aircraft measurements – including high-altitude observations from the NASA Global Hawk platform.

The models generally capture the observed seasonal cycle of surface CHBr$_3$ and CH$_2$Br$_2$ well, with a strong model–measurement correlation ($r \geq 0.7$) at most sites. In a given model, the absolute model–measurement agreement at the surface is highly sensitive to the choice of emissions. Large inter-model differences are apparent when using the same emission inventory, highlighting the challenges faced in evaluating such inventories at the global scale. Across the ensemble, most consistency is found within the tropics where most of the models (8 out of 11) achieve best agreement to surface CHBr$_3$ observations using the lowest of the three CHBr$_3$ emission inventories tested (similarly, 8 out of 11 models for CH$_2$Br$_2$). In general, the models reproduce observations of CHBr$_3$ and CH$_2$Br$_2$ obtained in the tropical tropopause layer (TTL) at various locations throughout the Pacific well. Zonal variability in VSLS loading in the TTL is generally consistent among models, with CHBr$_3$ (and to a lesser extent CH$_2$Br$_2$) most elevated over the tropical western Pacific during boreal winter. The models also indicate the Asian monsoon during boreal summer to be an important pathway for VSLS reaching the stratosphere, though the strength of this signal varies considerably among models.

We derive an ensemble climatological mean estimate of the stratospheric bromine SGI from CHBr$_3$ and CH$_2$Br$_2$ of 2.0 (1.2–2.5) ppt, $\sim 57\%$ larger than the best estimate from the most recent World Meteorological Organization (WMO) Ozone Assessment Report. We find no evidence for a long-term, transport-driven trend in the stratospheric SGI of bromine over the simulation period. The transport-driven interannual variability in the annual mean bromine SGI is of the order of $\pm 5\%$, with SGI exhibiting a strong positive correlation with the El Niño–Southern Oscillation (ENSO) in the eastern Pacific. Overall, our results do not show systematic differences between models specific to the choice of reanalysis meteorology, rather clear differences are seen related to differences in the implementation of transport processes in the models.

1 Introduction

Halogenated very short-lived substances (VSLS) are gases with atmospheric lifetimes shorter than, or comparable to, tropospheric transport timescales (\(\sim 6\) months or less at the surface). Naturally emitted VSLS, such as bromoform (CHBr$_3$), have marine sources and are produced by phytoplankton (e.g. Quack and Wallace, 2003) and various species of seaweed (e.g. Carpenter and Liss, 2000) – a number of which are farmed for commercial application (Leedham et al., 2013). Once in the atmosphere, VSLS (and their degradation products) may ascend to the lower stratosphere (LS), where they contribute to the inorganic bromine (Br$_x$) budget (e.g. Pfeilsticker et al., 2000; Sturges et al., 2000) and thereby enhance halogen-driven ozone (O$_3$) loss (Salawitch et al., 2005; Feng et al., 2007; Sinnhuber et al., 2009; Sinnhuber and Meul, 2015). On a per molecule basis, O$_3$ perturbations near the tropopause exert the largest radiative effect (e.g. Lacis et al., 1990; Forster and Shine, 1997; Riese et al., 2012), and recent work has highlighted the climate relevance of VSLS-driven O$_3$ loss in this region (Hossaini et al., 2015a).

Quantifying the contribution of VSLS to stratospheric Br$_x$ (Br$_x^{\text{VSLS}}$) has been a major objective of numerous recent observational studies (e.g. Dorf et al., 2008; Laube et al., 2008; Brinckmann et al., 2012; Sala et al., 2014; Wischer et al., 2014) and modelling efforts (e.g. Warwick et al., 2006; Hossaini et al., 2010; Liang et al., 2010; Aschmann et al., 2011; Tegtmeier et al., 2012; Hossaini et al., 2012b, 2013; Aschmann and Sinnhuber, 2013; Fernandez et al., 2014). However, despite a wealth of research, Br$_x^{\text{VSLS}}$ remains poorly constrained, with a current best-estimate range of 2–8 ppt reported in the most recent World Meteorological Organization (WMO) Ozone Assessment Report (Carpenter and Reimann, 2014). Between 15 and 76\% of this supply comes from the stratospheric source gas injection (SGI) of VSLS, i.e. the transport of a source gas (e.g. CHBr$_3$) across the tropopause, followed by its breakdown and in situ release of Br$_x^{\text{VSLS}}$ in the LS. The remainder comes from the troposphere-to-stratosphere transport of both organic and inorganic product gases, formed following the breakdown of VSLS below the tropopause; termed product gas injection (PGI).

Owing to their short tropospheric lifetimes, combined with significant spatial and temporal inhomogeneity in their emissions (e.g. Carpenter et al., 2005; Archer et al., 2007; Or-
likowska and Schulz-Bull, 2009; Ziska et al., 2013; Stemmler et al., 2015), the atmospheric abundance of VSLS can exhibit sharp tropospheric gradients. The stratospheric SGI of VSLS is expected to be most efficient in regions where strong uplift, such as convectively active regions, coincides with regions of elevated surface mixing ratios (e.g. Tegtmeier et al., 2012, 2013; Liang et al., 2014), driven by strong localised emissions or hotspots. Both the magnitude and distribution of emissions, with respect to transport processes, could be, therefore, an important determining factor for SGI. However, current global-scale emission inventories of CHBr_3 and CH_3Br_2 are poorly constrained, owing to a paucity of observations used to derive their surface fluxes (Ashfold et al., 2014), contributing significant uncertainty to model estimates of Br_y^{VSLS} (Hossaini et al., 2013). Given the uncertainties outlined above, it is unclear how well preferential transport pathways of VSLS to the LS are represented in global-scale models.

Strong convective source regions, such as the tropical western Pacific during boreal winter, are likely important for the troposphere-to-stratosphere transport of VSLS (e.g. Levine et al., 2007; Aschmann et al., 2009; Pisco et al., 2010; Hossaini et al., 2012b; Liang et al., 2014). The Asian monsoon also represents an effective pathway for boundary layer air to be rapidly transported to the LS (e.g. Randel et al., 2010; Vogel et al., 2014; Orbe et al., 2015; Tissier and Légras, 2016), though its importance for the troposphere-to-stratosphere transport of VSLS is largely unknown, owing to a lack of observations in the region. While global models simulate broadly similar features in the spatial distribution of convection, large inter-model differences in the number of tracers transported to the tropopause have been reported by Hoyle et al. (2011), who performed a model intercomparison of idealised (“VSLS-like”) tracers with a uniform surface distribution. In order for a robust estimate of the stratospheric SGI of bromine to be obtained, it is necessary to consider spatial variations in VSLS emissions, and how such variations couple with transport processes. However, a concerted model evaluation of this type has yet to be performed.

Over a series of two papers, we present results from the first VSLS multi-model intercomparison project (Atmospheric Tracer Transport Model Intercomparison Project; TransCom- VSLS). The TransCom initiative was set up in the 1990s to examine the performance of chemical transport models. Previous TransCom studies have examined non-reactive tropospheric species, such as sulfur hexafluoride (SF_6) (Denning et al., 1999) and carbon dioxide (CO_2) (Law et al., 1996, 2008). Most recently, TransCom projects have examined the influence of emissions, transport and chemical loss on atmospheric CH_4 (Patra et al., 2011) and N_2O (Thompson et al., 2014). The overarching goal of TransCom-VSLS was to constrain estimates of Br_y^{VSLS}, towards closure of the stratospheric bromine budget, by (i) providing a reconciled climatological model estimate of bromine SGI, to reduce uncertainty on the measurement-derived range (0.7–3.4 ppt Br) – currently uncertain by a factor of ~5 (Carpenter and Reimann, 2014) – and (ii) quantify the influence of emissions and transport processes on inter-model differences in SGI. In this regard, we define transport differences between models as the effects of boundary layer mixing, convection and advection, and the implementation of these processes. The project was not designed to separate the contributions of each transport component in the large model ensemble clearly, but this can be inferred as the boundary layer mixing affects tracer concentrations mainly near the surface, convection controls tracer transport to the upper troposphere and advection mainly distributes tracers horizontally (e.g. Patra et al., 2009). Specific objectives were to (a) evaluate models against measurements from the surface to the tropical tropopause layer (TTL) and (b) examine zonal and seasonal variations in VSLS loading in the TTL. We also show interannual variability in the stratospheric loading of VSLS (limited to transport) and briefly discuss possible trends related to the El Niño–Southern Oscillation (ENSO). Section 2 gives a description of the experimental design and an overview of participating models. Model–measurement comparisons are given in Sects. 3.1 to 3.3. Section 3.4 examines zonal/seasonal variations in the troposphere–stratosphere transport of VSLS and Sect. 3.5 provides our reconciled estimate of bromine SGI and discusses interannual variability.

2 Methods, models and observations

Eleven models, or their variants, took part in TransCom-VSLS. Each model simulated the major bromine VSLS, bromomethane (CHBr_3) and dibromomethane (CH_3Br_2), which together account for 77–86% of the total bromine SGI from VSLS reaching the stratosphere (Carpenter and Reimann, 2014). Participating models also simulated the major iodine VSLS, methyl iodide (CH_3I), though results from the iodine simulations will feature in a forthcoming, stand-alone paper (Hossaini et al., 2016). Each model ran with multiple CHBr_3 and CH_3Br_2 emission inventories (see Sect. 2.1) in order to (i) investigate the performance of each inventory, in a given model, against observations and (ii) identify potential inter-model differences whilst using the same inventory. Analogous to previous TransCom experiments (e.g. Patra et al., 2011), a standardised treatment of tropospheric chemistry was employed, through the use of prescribed oxidants and photolysis rates (see Sect. 2.2). This approach (i) ensured a consistent chemical sink of VSLS among models, minimising the influence of inter-model differences in tropospheric chemistry on the results, and thereby (ii) isolated differences due to transport processes. Long-term simulations, over a 20-year period (1993–2012), were performed by each model in order to examine trends and transport-driven interannual variability in the stratospheric SGI of CHBr_3 and CH_3Br_2. Global monthly mean model output over the full simula-
tion period, along with output at a higher temporal resolution (typically hourly) over measurement campaign periods, was requested from each group. A brief description of the models is given in Sect. 2.3 and a description of the observational data used in this work is given in Sect. 2.4. Figure 1 summarises the approach of TransCom-VSLS and its broad objectives.

2.1 Tracers and oceanic emission fluxes

Owing to significant differences in the magnitude and spatial distribution of VSLS emission fluxes, among previously published inventories (Hossaini et al., 2013), all models ran with multiple CHBr$_3$ and CH$_2$Br$_2$ tracers. Each of these tracers used a different set of prescribed surface emissions. Tracers named “CHBr$_3$-L”, “CHBr$_3$-O” and “CHBr$_3$-Z” used the inventories of Liang et al. (2010), Ordóñez et al. (2012) and Ziska et al. (2013), respectively. These three studies also reported emission fluxes for CH$_2$Br$_2$, and thus the same (L/O/Z) notation applies to the model CH$_2$Br$_2$ tracers, as summarised in Table 1. As these inventories were recently described and compared by Hossaini et al. (2013), only a brief description of each is given below. Surface CHBr$_3$/CH$_2$Br$_2$ emission maps for each inventory are given in the Supplement (Figs. S1 and S2).

The Liang et al. (2010) inventory is a top-down estimate of VSLS emissions based on aircraft observations, mostly concentrated around the Pacific and North America between 1996 and 2008. Measurements of CHBr$_3$ and CH$_2$Br$_2$ from the following National Aeronautics and Space Administration (NASA) aircraft campaigns were used to derive the ocean fluxes: PEM-Tropics, TRACE-P, INTEX, TC4, ARCTAS, STRAT, Pre-AVE and AVE. This inventory is aseasonal and assumes the same spatial distribution of emissions for CHBr$_3$ and CH$_2$Br$_2$. The Ordóñez et al. (2012) inventory is also a top-down estimate, based on the same set of aircraft measurements with the addition of the NASA POLARIS and SOLVE campaigns. This inventory weights tropical (±20° latitude) CHBr$_3$ and CH$_2$Br$_2$ emissions according to a monthly varying satellite climatology of chlorophyll a (chl a), a proxy for oceanic bio-productivity, providing some seasonality to the emission fluxes. The Ziska et al. (2013) inventory is a bottom-up estimate of VSLS emissions, based on a compilation of seawater and ambient air measurements of CHBr$_3$ and CH$_2$Br$_2$. Climatological, aseasonal emission maps of these VSLS were calculated using the derived sea-air concentration gradients and a commonly used sea-to-air flux parameterisation, considering wind speed, sea surface temperature and salinity (Nightingale et al., 2000).

2.2 Tropospheric chemistry

Participating models considered chemical loss of CHBr$_3$ and CH$_2$Br$_2$ through oxidation by the hydroxyl radical (OH) and by photolysis. These loss processes are comparable for CHBr$_3$, with photolysis contributing ~60% of the CHBr$_3$ chemical sink at the surface (Hossaini et al., 2010). For CH$_2$Br$_2$, photolysis is a minor tropospheric sink, with its loss dominated by OH-initiated oxidation. The overall local lifetimes of CHBr$_3$ and CH$_2$Br$_2$ in the tropical marine boundary layer have recently been evaluated to be 15 (13–17) and 94 (84–114) days, respectively (Carpenter and Reimann, 2014). These values are calculated based on [OH]=1×10$^6$ molecules cm$^{-3}$, $T=275$ K and with a global annual mean photolysis rate. For completeness, models also considered loss of CHBr$_3$ and CH$_2$Br$_2$ by reaction
were the same as those used in the previous TransCom framework, these fields have been extensively evaluated and shown to give a realistic simulation of tropospheric burden and lifetime of methane and also methyl chloroform. Models also ran with the same monthly mean oxidant fields. For OH, O(1D) and Cl, these fields were the same as those used in the previous TransCom-CH4 model intercomparison (Patra et al., 2011). Within the TransCom framework, these fields have been extensively used and evaluated and shown to give a realistic simulation of the tropospheric burden and lifetime of methane and also methyl chloroform. Models also ran with the same monthly mean CHBr3 and CH2Br2 photolysis rates, calculated offline from the TOMCAT chemical transport model (Chipperfield, 2006). TOMCAT has been used extensively to study the tropospheric chemistry of VSLS (e.g. Hossaini et al., 2010, 2012b, 2015b), and photolysis rates from the model were used to evaluate the lifetime of VSLS for the recent WMO Ozone Assessment Report (Carpenter and Reimann, 2014).

2.3 Participating models and output

Eight global models (ACTM, B3DCTM, EMAC, MOZART, NIES-TM, STAG, TOMCAT and UKCA) and three of their variants (see Table 2) participated in TransCom-VSLS. All the models are offline chemical transport models (CTMs), forced with analysed meteorology (e.g. winds and temperature fields), with the exception of EMAC and UKCA, which are free-running chemistry–climate models (CCMs), calculating winds and temperature online. The horizontal resolution of models ranged from \( \sim 1^\circ \times 1^\circ \) (longitude \times latitude) to \( 3.75^\circ \times 2.5^\circ \). In the vertical, the number of levels varied from 32 to 85, with various coordinate systems. A summary of the models and their salient features is given in Table 2. Note, these features do not necessarily link to model performance as evaluated in this work. Note also, approximately half of the models used ECMWF ERA-Interim meteorological data. In terms of mean upwelling in the tropics, where stratospheric bromine SGI of VSLS takes place, there is generally good agreement between the most recent major reanalysis products from ECMWF, JMA and NCEP (e.g. Harada et al., 2016). Therefore, we do not expect a particular bias in our results from the use of ERA-Interim.

Three groups, the Karlsruhe Institute of Technology (KIT), the University of Leeds (UoL) and the University of Cambridge (UoC), submitted output from an additional set of simulations using variants of their models. KIT ran the EMAC model twice, as a free-running model (here termed “EMAC_F”) and also in nudged mode (EMAC_N). The UoL performed two TOMCAT simulations, the first of which used the model’s standard convection parameterisation, based on the mass flux scheme of Tiedtke (1989). The second TOMCAT simulation (“TOMCAT_conv”) used archived convective mass fluxes, taken from the ECMWF ERA-Interim reanalysis. A description and evaluation of these TOMCAT variants is given in Feng et al. (2011). In order to investigate the influence of resolution, the UoC ran two UKCA model simulations with different horizontal/vertical resolutions. The horizontal resolution in the “UKCA_high” simulation was a factor of 4 (2 in two dimensions) greater than that of the standard UKCA run (Table 2).

All participating models simulated the six CHBr3 and CH2Br2 tracers (see Sect. 2.1) over a 20-year period, 1 January 1993 to 31 December 2012. This period was chosen as it (i) encompasses a range of field campaigns during which VSLS measurements were taken and (ii) allows the strong El Niño event of 1997/1998 to be investigated in the analysis of SGI trends. The monthly mean volume mixing ratio (vmr) of each tracer was archived by each model on the same 17 pressure levels, extending from the surface to 10hPa over the full simulation period. The models were also sampled hourly at 15 surface sites over the full simulation period and during periods of recent ship/aircraft measurement campaigns, de-

### Table 1. Summary of the VSLS tracers simulated by the models, the global total emission flux (Gg VSLS yr\(^{-1}\)) and the rate constant for their reaction with OH (Sander et al., 2011). See text for details of emission inventories.

<table>
<thead>
<tr>
<th>Tracer no.</th>
<th>Species</th>
<th>Tracer name</th>
<th>Global flux (Gg yr(^{-1}))</th>
<th>Reference</th>
<th>Rate constant (VSLS + OH reaction) ( k(T) ) (cm(^3) molec(^{-1}) s(^{-1}))</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Bromoform</td>
<td>CHBr3(_L)</td>
<td>450</td>
<td>Liang et al. (2010)</td>
<td>( 1.35 \times 10^{-12} \exp(-600/T) )</td>
</tr>
<tr>
<td>2</td>
<td></td>
<td>CHBr3(_O)</td>
<td>530</td>
<td>Ordóñez et al. (2012)</td>
<td>( 2.00 \times 10^{-12} \exp(-840/T) )</td>
</tr>
<tr>
<td>3</td>
<td></td>
<td>CHBr3(_Z)</td>
<td>216</td>
<td>Ziska et al. (2013)</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>Dibromomethane</td>
<td>CH2Br2(_L)</td>
<td>62</td>
<td>Liang et al. (2010)</td>
<td></td>
</tr>
<tr>
<td>5</td>
<td></td>
<td>CH2Br2(_O)</td>
<td>67</td>
<td>Ordóñez et al. (2012)</td>
<td></td>
</tr>
<tr>
<td>6</td>
<td></td>
<td>CH2Br2(_Z)</td>
<td>87</td>
<td>Ziska et al. (2013)</td>
<td></td>
</tr>
</tbody>
</table>
2.4 Observational data and processing

2.4.1 Surface

Model output was compared to and evaluated against a range of observational data. At the surface, VSLS measurements obtained from whole air samples, collected approximately weekly using gas chromatography/mass spectrometry within the Central Observatory at Boulder, were calculated from each model's hourly output sampled at each location. Similar climatological fields of CHBr$_3$ and CH$_2$Br$_2$ were calculated over the period January 1998 to December 2012. Simulated tracer mixing ratios were also compared to and evaluated against observations from the VSLS network at 15 sites were considered (Table 3). All sites except one (the northern coast of Borneo) were sampled within the South China Sea and along the northern coast of Borneo. The site "TAW", Table 3, were also considered. A description of the climatological fields of CHBr$_3$ and CH$_2$Br$_2$ is given in Robinson et al. (2004). A description of the modelled boundary layer mixing is given in Sect. 2.4 below.

Table 2. Overview of TransCom-VSLS model variants.

<table>
<thead>
<tr>
<th>No.</th>
<th>Model$^a$</th>
<th>Institution$^b$</th>
<th>Resolution</th>
<th>Meteorology$^c$</th>
<th>Boundary layer mixing</th>
<th>Convection</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>B3DCTM</td>
<td>UoB</td>
<td>3.75° × 2.5°</td>
<td>ECMWF ERA-Interim</td>
<td>Simple</td>
<td>ERA-Interim, archived$^b$</td>
<td>Aschmann et al. (2014)</td>
</tr>
<tr>
<td>3</td>
<td>EMAC$^c$</td>
<td>KIT</td>
<td>2.8° × 2.8°</td>
<td>ECMWF ERA-Interim</td>
<td>Jöckel et al. (2006)</td>
<td>Jöckel et al. (2000, 2006)</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>EMAC (nudging)</td>
<td>KIT</td>
<td>2.8° × 2.8°</td>
<td>Nudged to ERA-Interim</td>
<td>Jöckel et al. (2006)</td>
<td>Jöckel et al. (2006, 2010)</td>
<td></td>
</tr>
<tr>
<td>5</td>
<td>MOZART</td>
<td>EMU</td>
<td>2.5° × 1.9°</td>
<td>MERRA</td>
<td>Holtslag and Boville (1993)</td>
<td>1</td>
<td>Emmons et al. (2000)</td>
</tr>
<tr>
<td>6</td>
<td>NIES-TM</td>
<td>NIES</td>
<td>2.5° × 2.5°</td>
<td>JCDAS (JRA-25)</td>
<td>Belikov et al. (2013)</td>
<td>Belikov et al. (2011, 2013)</td>
<td></td>
</tr>
<tr>
<td>8</td>
<td>TOMCAT</td>
<td>UoL</td>
<td>2.8° × 2.8°</td>
<td>ECMWF ERA-Interim</td>
<td>Holtslag and Boville (1993)</td>
<td>1</td>
<td>Chipperfield (2009)</td>
</tr>
<tr>
<td>9</td>
<td>TOMCAT (conv)</td>
<td>UoL</td>
<td>2.8° × 2.8°</td>
<td>ECMWF ERA-Interim</td>
<td>Holtslag and Boville (1993)</td>
<td>1</td>
<td>Chipperfield (2009)</td>
</tr>
<tr>
<td>10</td>
<td>UKCA (low)</td>
<td>UoC/NCAS</td>
<td>3.75° × 2.5°</td>
<td>ECMWF ERA-Interim</td>
<td>EMAA</td>
<td>Gregory and Rowntree (1990)</td>
<td>Morgenstern et al. (2009)</td>
</tr>
</tbody>
</table>

$^a$All models are offline CTMs except bold entries which are CCMs. Model variants are shown in italics. CCMs ran using prescribed sea surface temperatures from observations. $^b$ JAMSTEC: Japan Agency for Marine-Earth Science and Technology, Japan; UoB: University of Bremen, Germany; KIT: Karlsruhe Institute of Technology, Germany; EMU: Emory University, USA; NIES: National Institute for Environmental Studies, Japan; AIST: National Institute of Advanced Industrial Science and Technology, Japan; UoC: University of Leeds, UK; UoK: University of Keele, UK; NCAS: National Centre for Atmospheric Science, UK. $^c$Longitude × latitude. $^d$ climate-following sigma levels (pressure divided by surface pressure); $^e$ hybrid sigma-pressure; $^f$ hybrid sigma-potential temperature; $^g$ hybrid sigma-height. $^c$ MERRA: Modern-era Retrospective Analysis for Research and Applications. JCDAS: Japan Meteorological Agency Climate Data Assimilation System. ERA-25: Japanese 25-year ReAnalyses. ECMWF: European Centre for Medium-range Weather Forecasts. $^e$ ECHAM/MESy Atmospheric Chemistry (EMAC) model (Roekner et al., 2006). ECHAM5 version 5.3.02. MESy version 2.42. $^f$ Simple averaging of tracer mixing ratio below ERA-Interim boundary layer height. $^g$ Read-in convective mass fluxes from ECMWF ERA-Interim. See Aschmann et al. (2011) for B3DCTM implementation and Feng et al. (2011) for TOMCAT implementation. With modifications from Nodden (1994). $^h$ Shallow $n$-level convection (Bach, 1994); deep convection (Zhang and McFarlane, 1995).


2.4.2 Aircraft

Observations of CHBr$_3$ and CH$_2$Br$_2$ from a range of aircraft campaigns were also used (Fig. 2). As (i) the troposphere-to-stratosphere transport of air (and VSLS) primarily occurs in the tropics, and (ii) because VSLS emitted in the extratropics have a negligible impact on stratospheric ozone (Tegtmeier et al., 2015), TransCom-VSLS focused on aircraft measurements obtained in the latitude range 30° N to 30° S. Hourly model output was interpolated to the relevant aircraft sampling location, allowing for point-by-point model–measurement comparisons. A brief description of the aircraft campaigns follows.

The HIAPER Pole-to-Pole Observations (HIPO) project (http://www.eol.ucar.edu/projects/hippo) comprised a series of aircraft campaigns between 2009 and 2011 (Wofsy et al., 2011), supported by the National Science Foundation (NSF). Five campaigns were conducted: HIPPO-1 (January 2009), HIPPO-2 (November 2009), HIPPO-3 (March/April 2010), HIPPO-4 (June 2011) and HIPPO-5 (August/September 2011). Sampling spanned a range of latitudes, from near the North Pole to coastal Antarctica, on board the NSF Gulfstream V aircraft, and from the surface to ~14 km over the Pacific Basin. Whole air samples, collected in stainless steel and glass flasks, were analysed by two different laboratories using GC/MS (NOAA/ESRL and the University of Miami). HIPO results from both laboratories are provided on a scale consistent with NOAA/ESRL.

The SHIVA aircraft campaign, based in Miri (Malaysian Borneo), was conducted during November–December 2011. Measurements of CHBr$_3$ and CH$_2$Br$_2$ were obtained during 14 flights of the DLR Falcon aircraft, with sampling over much of the northern coast of Borneo, within the South China and Sulu seas, up to an altitude of ~12 km (Sala et al., 2014; Fuhlbrügge et al., 2016). VSLS measurements were obtained by two groups, the University of Frankfurt (UoF) and the University of East Anglia (UEA). UoF measurements were made using an in situ GC/MS system (Sala et al., 2014), while UEA analysed collected whole air samples, using GC/MS.

CAST (Coordinated Airborne Studies in the Tropics) is an ongoing research project funded by the UK Natural Environment Research Council (NERC) and is a collaborative initiative with the NASA ATTREX programme (see below). The CAST aircraft campaign, based in Guam, was conducted in January–February 2014 with VSLS measurements made by the University of York on board the FAAM (Facility for Airborne Atmospheric Measurements) BAe-146 aircraft, up to an altitude of ~8 km. These observations were made by GC/MS collected from whole air samples as described in Andrews et al. (2016).

Observations of CHBr$_3$ and CH$_2$Br$_2$ within the TTL and lower stratosphere (up to ~20 km) were obtained during the NASA (i) Pre-Aura Validation Experiment (Pre-AVE), (ii) Costa Rica Aura Validation Experiment (CR-AVE) and (iii) Airborne Tropical TRopopause EXperiment (ATTREX) missions. The Pre-AVE mission was conducted in 2004 (January–February), with measurements obtained over the equatorial eastern Pacific during eight flights of the high-altitude WB-57 aircraft. The CR-AVE mission took place in 2006 (January–February) and sampled a similar region around Costa Rica (Fig. 2), also with the WB-57 aircraft (15 flights). The ATTREX mission consists of an ongoing series of aircraft campaigns using the unmanned Global Hawk aircraft. Here, CHBr$_3$ and CH$_2$Br$_2$ measurements from 10 flights of the Global Hawk, over two ATTREX campaigns, were used. The first campaign (February–March 2013) sampled large stretches of the north-east and central Pacific Ocean, while the second campaign (January–March 2014) sampled predominantly the western Pacific, around Guam. During Pre-AVE, CR-AVE and ATTREX, VSLS measurements were obtained by the University of Miami following GC/MS analysis of collected whole air samples.

3 Results and discussion

3.1 Model–observation comparisons: surface

In this section, we evaluate the models in terms of (i) their ability to capture the observed seasonal cycle of CHBr$_3$ and CH$_2$Br$_2$ at the surface and (ii) the absolute agreement to the observations. We focus on investigating the relative performance of each of the tested emission inventories, within a given model, and the performance of the inventories across the ensemble.

3.1.1 Seasonality

We first consider the seasonal cycle of CHBr$_3$ and CH$_2$Br$_2$ at the locations given in Table 3. Figure 3 compares observed and simulated (CHBr$_3$ L tracer) monthly mean anomalies, calculated by subtracting the climatological monthly mean CHBr$_3$ surface mole fraction from the climatological annual...
Figure 3. Comparison of the observed and simulated seasonal cycle of surface CHBr$_3$ at ground-based measurement sites (see Table 3). The seasonal cycle is shown here as climatological (1998–2011) monthly mean anomalies, calculated by subtracting the climatological monthly mean CHBr$_3$ mole fraction (ppt) from the climatological annual mean, in both the observed (black points) and model (coloured lines; see legend) data sets. The location of the surface sites is summarised in Table 3. Model output based on CHBr$_3$ L tracer (i.e. using aseasonal emissions inventory of Liang et al., 2010). Horizontal bars denote ±1σ.

Table 3. Summary and location of ground-based surface VSLS measurements used in TransCom-VSLS, arranged from north to south. All sites are part of the NOAA/ESRL global monitoring network, with the exception of TAW, at which measurements were obtained by the University of Cambridge (see main text). * Stations SUM, MLO and SPO are elevated at ~3210, 3397 and 2810 m, respectively.

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<th>Longitude</th>
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Table 3, Summary and location of ground-based surface VSLS measurements used in TransCom-VSLS, arranged from north to south. All sites are part of the NOAA/ESRL global monitoring network, with the exception of TAW, at which measurements were obtained by the University of Cambridge (see main text). * Stations SUM, MLO and SPO are elevated at ~3210, 3397 and 2810 m, respectively.

To mean (to focus on the seasonal variability). Based on photochemistry alone, in the Northern Hemisphere (NH), one would expect a CHBr$_3$ winter (December–February) maximum owing to a reduced chemical sink (e.g. slower photolysis rates and lower [OH]) and thereby a relatively longer CHBr$_3$ lifetime. This seasonality, apparent at most NH sites shown in Fig. 3, is particularly pronounced at high latitudes (> 60° N, e.g. ALT, BRW and SUM), where the amplitude of the observed seasonal cycle is greatest. A number of features are apparent from these comparisons. First, in general, most models reproduce the observed phase of the CHBr$_3$ seasonal cycle well, even with emissions that do not vary seasonally, suggesting that seasonal variations in the CHBr$_3$ chemical sink are generally well represented. For example, model–measurement correlation coefficients ($r$), summarised in Table 4, are > 0.7 for at least 80% of the models at 7 of 11 NH sites. Second, at some sites, notably MHD, THD, CGO and PSA, the observed seasonal cycle of CHBr$_3$ is not captured well by virtually all of the models (see discussion below). Third, at most sites the amplitude of the seasonal cycle is generally consistent across the models (within a few percent, excluding clear outliers). The cause of outliers at a given site is likely in part related to the model sampling error, including distance of a model grid from the measurement...
Table 4. Correlation coefficient (r) between the observed and simulated climatological monthly mean surface CHBr3 volume mixing ratio (at ground-based monitoring sites, Table 3). Model output based on CHBr3_L tracer (i.e. using aseasonal emissions inventory of Liang et al., 2010). Stations in bold denote where virtually all models fail to reproduce phase of the observed CHBr3 seasonal cycle.

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<th>EMAC_N</th>
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<th>NIES</th>
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site and resolution (as was shown for CO2 in Patra et al., 2008). These instances are rare for VSLS but can be seen in B3DCTM’s output in Fig. 3 for CHBr3 at SMO. B3DCTM ran at a relatively coarse horizontal resolution (3.75°) and with fewer vertical layers (40) compared to most other models. Note, it also has the simplest implementation of boundary layer mixing (Table 2). The above behaviour is also seen at SMO but to a lesser extent for CH2Br2, for which the seasonal cycle is smaller (see below). The STAG model also produces distinctly different features in the seasonal cycle of both species at some sites (prominently at CGO, SMO and HFM). We attribute these deviations to STAG’s parameterisation of boundary layer mixing, noting that differences for CHBr3 are greater at KUM than at MLO – two sites in very close proximity but with the latter elevated at ~ 3000 m above sea level (i.e. above the boundary layer). With respect to the observations, the amplitude of the seasonal cycle is either under- (e.g. BRW) or overestimated (e.g. KUM) at some locations, by all of the models. This possibly reflects a more systematic bias in the prescribed CHBr3 loss rate and/or relates to emissions, though this effect is generally small and localised.

A similar analysis has been performed to examine the seasonal cycle of surface CH2Br2. Observed and simulated monthly mean anomalies, calculated in the same fashion as those for CHBr3 above, are shown in Fig. 4 and correlation coefficients are given in Table 5. The dominant chemical sink of CH2Br2 is through OH-initiated oxidation, and thus its seasonal cycle at most stations reflects seasonal variation in [OH] and temperature. At most sites, this gives rise to a minimum in the surface mole fraction of CH2Br2 during summer months, owing to greater [OH] and temperature, and thereby a faster chemical sink. Relative to CHBr3, CH2Br2 is considerably longer lived (and thus well mixed) near the surface, meaning the amplitude of the seasonal cycle is far smaller. At most sites, most models capture the observed phase and amplitude of the CH2Br2 seasonal cycle well, though as was the case for CHBr3, agreement in the Southern Hemisphere (SH, e.g. SMO, CGO, PSA) seems poorest. For example, at SMO and CGO only 40% of the models are positively correlated to the simulations with r > 0.5 (Table 5). The NIES-TM model does not show major differences from other models for CHBr3, but outliers for CH2Br2 at SH sites (SMO to SPO) are apparent. We were unable to assign any specific reason for the inter-species differences seen for this model.

At two sites (MHD and THD) almost none of the models reproduce the observed CHBr3 seasonal cycle, exhibiting an anti-correlation with the observed cycle (see bold entries in Table 4). Here, the simulated cycle follows that expected from seasonality in the chemical sink. At MHD, seasonality in the local emission flux is suggested to be the dominant factor controlling the seasonal cycle of surface CHBr3 (Carpenter et al., 2005). This leads to the observed summer maximum (as shown in Fig. 3) and is not represented in the models’ CHBr3_L tracer which, at the surface, is driven by the aseasonal emission inventory of Liang et al. (2010). A similar summer maximum seasonal cycle is observed for CH2Br2, also not captured by the models’ CH2Br2 L tracer. To investigate the sensitivity of the model–measurement correlation to the prescribed surface fluxes, multi-model mean (MMM) surface CHBr3 and CH2Br2 fields were calculated for each tracer (i.e. for each emission inventory considered) and each site. Figure 5 shows calculated MMM r values at each site for CHBr3 and CH2Br2. For CHBr3, r generally has a low sensitivity to the choice of emission fluxes at most sites (e.g. ALT, SUM, BRW, LEF, NWR, KUM,
MLO, SPO), though notably at MHD, use of the Ziska et al. (2013) inventory (which is aseasonal) reverses the sign of $r$ to give a strong positive correlation (MMM $r > 0.70$) against the observations. Individual model $r$ values for MHD are given in Table S1 of the Supplement. With the exception of TOMCAT, TOMCAT_CONV and UKCA_HI, the remaining seven models each reproduce the MHD CHBr$_3$ seasonality well (with $r > 0.65$). That good agreement obtained with the Ziska aseasonal inventory, compared to the other aseasonal inventories considered, highlights the importance of the CHBr$_3$ emission distribution, with respect to transport processes, serving this location. We suggest that the summertime transport of air that has experienced relatively large CHBr$_3$ emissions north/north-west of MHD is the cause of the apparent seasonal cycle seen in most models using the Ziska inventory (example animations of the seasonal evolution of surface CHBr$_3$ are given in the Supplementary Information to visualise this). Note also, the far better abso-

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**Figure 4.** As Fig. 3 but for CH$_2$Br$_2$.

**Table 5.** As Table 4 but for CH$_2$Br$_2$.

<table>
<thead>
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<th>Site</th>
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<td>0.20</td>
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</tr>
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<td>0.59</td>
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<tr>
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<td>0.80</td>
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<td>0.97</td>
<td>0.90</td>
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<td>0.88</td>
</tr>
</tbody>
</table>
observed background mixing ratios of CHBr$_3$ from models using this inventory (Supplement Fig. S3). At other sites, such as TAW, no clear seasonality is apparent in the observed background mixing ratios of CHBr$_3$ and CH$_2$Br$_2$ (Robinson et al., 2014). Here, the models exhibit little or no significant correlation to measured values and are unlikely to capture small-scale features in the emission distribution (e.g. the contribution from local aquaculture) that conceivably contribute to observed levels of CHBr$_3$ and CH$_2$Br$_2$ in this region (Robinson et al., 2014).

### 3.1.2 Absolute agreement

To compare the absolute agreement between a model (M) and an observation (O) value, for each monthly mean surface model–measurement comparison, the mean absolute percentage error (MAPE, Eq. 1) was calculated for each model tracer. Figure 6 shows the CHBr$_3$ and CH$_2$Br$_2$ tracer that provides the lowest MAPE (i.e. best agreement) for each model (indicated by the fill colour of cells). The numbers within the cells give the MAPE value itself, and therefore correspond to the “best agreement” that can be obtained from the various tracers with the emission inventories that were tested.

\[
\text{MAPE} = \frac{100}{n} \sum_{r=1}^{n} \left| \frac{M_r - O_r}{O_r} \right|
\]

Figure 5. Correlation coefficient (r) between observed and multi-model mean (a) CHBr$_3$ and (b) CH$_2$Br$_2$, at ground-based monitoring sites. The correlation here represents the mean annual seasonal variation. At each site, 3x3 values are given, reflecting the three different model CHBr$_3$ tracers; green squares denote the CHBr$_3$ L-tracer (top-down derived, Liang et al., 2010, emissions), blue diamonds denote the CHBr$_3$ O-tracer (top-down Ordóñez et al., 2012, emissions) and red circles denote the CHBr$_3$ Z-tracer (bottom-up Ziska et al., 2013, emissions).

For both CHBr$_3$ and CH$_2$Br$_2$, within any given model, no single emission inventory is able to provide the best agreement at all surface locations (i.e. from the columns in Fig. 6). This was previously noted by Hossaini et al. (2013) using the TOMCAT model, and to some degree likely reflects the geographical coverage of the observations used to create the emission inventories. Hossaini et al. (2013) also noted significant differences between simulated and observed CHBr$_3$ and CH$_2$Br$_2$, using the same inventory; i.e. at a given location, low CHBr$_3$ MAPE (good agreement) does not necessarily accompany a corresponding low CH$_2$Br$_2$ MAPE using the same inventory.

A key finding of this study is that significant inter-model differences are also apparent (i.e. see rows in Fig. 6 grid). For example, for CHBr$_3$, no single inventory performs best across the full range of models at any given surface site. TOMCAT and B3DCTM – both of which are driven by ERA-Interim – agree on the best CHBr$_3$ inventory (lowest MAPE) at approximately half of the 17 sites considered. This analysis implies that, on a global scale, the “performance” of emission inventories is somewhat model-specific and highlights the challenges of evaluating such inventories. Previous conclusions as to the best performing VSLS inventories, based on single model simulations (Hossaini et al., 2013), must therefore be treated with caution. When one considers that previous modelling studies (Warwick et al., 2006; Liang et al., 2010; Ordóñez et al., 2012), each having derived different VSLS emissions based on aircraft observations, and having different tropospheric chemistry, report generally good agreement between their respective model and observations, our findings are perhaps not unexpected. However, we also note that few VSLS modelling studies have used long-term surface observations to evaluate their models, as performed here. This suggests that any attempts to reconcile estimates of global VSLS emissions, obtained from different modelling studies, need to consider the influence of inter-model differences.

As the chemical sink of VSLS was consistent across all models, the inter-model differences discussed above are attributed primarily to differences in the treatment and implementation of transport processes. This includes convection and boundary layer mixing, both of which can significantly influence the near-surface abundance of VSLS in the real atmosphere (Fuhlbrügge et al., 2013, 2016) and model (Zhang et al., 2008; Feng et al., 2011; Hoyle et al., 2011) atmospheres, and are parameterised in different ways (Table 2). On this basis, it is not surprising that different CTM set-ups lead to differences in the surface distribution of VSLS, nor that differences are apparent between CTMs that use the same meteorological input fields. Indeed, such effects have also been observed in previous model intercomparisons (Hoyle et al., 2011). Large-scale vertical advection, the native grid of a model and its horizontal/vertical resolution may also be contributing factors, though quantifying their relative influence was beyond the scope of TransCom-VSLS. At some
CHBr

The MAPE value (%), for each model compared to the observations. CHBr for that model, i.e. the lowest mean absolute percentage error (MAPE, see main text for details), and the numbers within the cells give see Table 3, and TransBrom/SHIVA ship cruises). The fill colour of each cell (see legend) indicates the tracer giving the best agreement otherwise identical, i.e. TOMCAT and TOMCAT_CONV entries of Fig. 6.

Despite the inter-model differences in the performance of emission inventories, some generally consistent features are found across the ensemble. First, for CHBr3 the tropical MAPE (see Fig. 7), based on the model–measurement comparisons in the latitude range ±20°, is lowest when using the emission inventory of Ziska et al. (2013), for most (8 out of 11, ~70%) of the models. This is significant as troposphere-to-stratosphere transport primarily occurs in the tropics and the Ziska et al. (2013) inventory has the lowest CHBr3 emission flux in this region (and globally, Table 1). Second, for CH2Br2, the tropical MAPE is lowest for most (also ~70%) of the models when using the Liang et al. (2010) inventory, which also has the lowest global flux of the three inventories tested. For a number of models, a similar agreement is also obtained with the Ordóñez et al. (2012) inventory, as the two are broadly similar in magnitude/distribution (Hossaini et al., 2013). For CH2Br2, the Ziska et al. (2013) inventory performs poorest across the ensemble (models generally overestimate CH2Br2 with this inventory). Overall, the tropical MAPE for a given model is more sensitive to the choice of emission inventory for CHBr3 than CH2Br2 (Fig. 7). Based on each model’s preferred inventory (i.e. from Fig. 7), the tropical MAPE is generally ~40% for CHBr3 and <20% for CH2Br2 (in most models). One model (STAG) exhibited a MAPE of >50% for both species, regardless of the choice of emission inventory, and was therefore omitted from the subsequent model–measurement comparisons to aircraft data and also from the multi-model mean SGI estimate derived in Sect. 3.5.

For the five models that submitted hourly output over the period of the SHIVA (2011) and TransBrom (2009) ship cruises, Figs. 8 and 9 compare the multi-model mean (MMM) CHBr3 and CH2Br2 mixing ratio (and the model spread) to the observed values. Note, the MMM was calcu-
We now evaluate modelled profiles of CHBr$_3$. Model–observation comparisons: free troposphere The boundary layer; see Sect. 3.2 below. Agreement between models and aircraft observations above distribution at most locations, and thereby provides good emission inventories. Generally, this leads to a realistic surface emissions inventory), as was diagnosed in the previous discussion (i.e. from Fig. 7, see also Sect. 3.1.2). This approach ensures that an estimate of stratospheric bromine SGI, from a given model, is based on a simulation in which the optimal CHBr$_3$/CH$_2$Br$_2$ model–measurement agreement at the surface was achieved. The objective of the comparisons below is to show that the models produce a realistic simulation of CHBr$_3$ and CH$_2$Br$_2$ in the tropical free troposphere and to test model transport of CHBr$_3$ and CH$_2$Br$_2$ from the surface to high altitudes, against that from atmospheric measurements. Intricacies of individual model–measurement comparison are not discussed. Rather, Fig. 10 compares MMM profiles (and the model spread) of CHBr$_3$ and CH$_2$Br$_2$ mixing ratio to observed campaign means within the tropics ($\pm 20^\circ$ latitude). Generally model–measurement agreement, diagnosed by both the campaign-averaged MAPE and the correlation coefficient ($r$) is excellent during most campaigns. For all of the seven campaigns considered, the modelled MAPE for CHBr$_3$ is $\leq 35\%$ ($\leq 20\%$ for CH$_2$Br$_2$). The models also capture much of the observed variability throughout the observed profiles, including, for example, the signature “c-shape” of convection in the measured CHBr$_3$ profile from SHIVA and HIPPO-I (panel a, second and third rows of Fig. 10). Correlation coefficients between modelled and observed CHBr$_3$ are $\geq 0.8$ for five of the seven campaigns and for CH$_2$Br$_2$ are generally $> 0.5$.

It is unclear why model–measurement agreement (particularly the CHBr$_3$ MAPE) is poorest for the HIPPO-4 and HIPPO-5 campaigns. However, we note that at most levels MMM CHBr$_3$ and CH$_2$Br$_2$ falls within ±1 standard deviation ($\sigma$) of the observed mean. Note, an underestimate of surface CHBr$_3$ does not generally translate to a consistent underestimate of measured CHBr$_3$ at higher altitude. Critically, for the most part, the models are able to reproduce observed values of both gases well at $\sim 12–14$ km, within the lower TTL. Recall that the TTL is defined as the layer between the level of main convective outflow ($\sim 200$ hPa, $\sim 12$ km) and
Overall, given the large spatial/temporal variability in observed VSLS mixing ratios, in part due to the influence of transport processes, global-scale models driven by aseasonal emissions and using parameterised sub-grid-scale transport schemes face challenges in reproducing VSLS observations in the tropical atmosphere; yet despite this, we find that the TransCom-VSLS models generally provide a very good simulation of the tropospheric abundance of CHBr$_3$ and CH$_2$Br$_2$, particularly in the important tropical western Pacific region (e.g. SHIVA comparisons).

### 3.3 Model–observation comparisons: TTL and lower stratosphere

Figure 11 compares model profiles of CHBr$_3$ and CH$_2$Br$_2$ with high-altitude measurements obtained in the TTL, extending into the tropical lower stratosphere. Across the ensemble, model–measurement agreement is varied but generally the models capture observed CHBr$_3$ from the Pre-AVE and CR-AVE campaigns, in the eastern Pacific, well. It should be noted that the number of observations varies significantly between these two campaigns; CR-AVE had almost twice the number of flights as Pre-AVE and this is reflected in the larger variability in the observed profile, particularly in the lower TTL. For both campaigns, the models capture the observed gradients in CHBr$_3$ and variability throughout the profiles; model–measurement correlation coefficients ($r$) for all of the models are $> 0.93$ and $> 0.88$ for Pre-AVE and CR-AVE, respectively. In terms of absolute agreement, 100% of the models fall within ±1σ of the observed CHBr$_3$ mean at the tropopause during Pre-AVE (and ±2σ for CR-AVE). For both campaigns, virtually all models are within the measured (min–max) range (not shown) around the tropopause.

During both ATTREX campaigns, larger CHBr$_3$ mixing ratios were observed in the TTL (panels c and d of Fig. 11). This reflects the location of the ATTREX campaigns compared to Pre-AVE and CR-AVE; over the tropical western Pacific, the level of main convective outflow extends deeper into the TTL compared to the eastern Pacific (Gettelman and Forster, 2002), allowing a larger portion of the surface CHBr$_3$ mixing ratio to detrain at higher altitudes. Overall, model–measurement agreement of CHBr$_3$ in the TTL is poorer during the ATTREX campaigns, with most models exhibiting a low bias between 14 and 16 km altitude. MOZART and UKCA simulations (which prefer the Liang CHBr$_3$ inventory) exhibit larger mixing ratios in the TTL, though are generally consistent with other models around the tropopause. Most ($\geq 70\%$) of the models reproduce CHBr$_3$ at the tropopause to within ±1σ of the observed mean and all the models are within the measured range (not shown) during both ATTREX campaigns. Model–measurement CHBr$_3$ correlation is $> 0.8$ for each ATTREX campaign, showing that again much of the observed variability throughout the CHBr$_3$ profiles is captured. The same is true for CH$_2$Br$_2$, with $r > 0.84$ for all but one of the models during Pre-AVE.
and $r > 0.88$ for all of the models in each of the other campaigns.

Overall, mean CHBr$_3$ and CH$_2$Br$_2$ mixing ratios around the tropopause, observed during the 2013/2014 ATTREX missions, are larger than the mean mixing ratios (from previous aircraft campaigns) reported in the latest WMO Ozone Assessment Report (Tables 1–7 of Carpenter and Reimann, 2014). As noted, this likely reflects the location at which the measurements were made; ATTREX 2013/2014 sampled in the tropical West and central Pacific, whereas the WMO estimate is based on a compilation of measurements with a paucity in that region. From Fig. 11, observed CHBr$_3$ and CH$_2$Br$_2$ at the tropopause were (on average) $\sim 0.35$ ppt and $\sim 0.8$ ppt, respectively, during ATTREX 2013/2014, compared to the 0.08 (0.00–0.31) ppt CHBr$_3$ and 0.52 (0.3–0.86) ppt CH$_2$Br$_2$ ranges reported by Carpenter and Reimann (2014).

3.4 Seasonal and zonal variations in the troposphere-to-stratosphere transport of VSLS

In this section we examine seasonal and zonal variability in the loading of CHBr$_3$ and CH$_2$Br$_2$ in the TTL and lower stratosphere, indicative of transport processes. In the tropics, a number of previous studies have shown a marked seasonality in convective outflow around the tropopause, owing to seasonal variations in convective cloud top heights (e.g. Folkins et al., 2006; Hosking et al., 2010; Bergman et al., 2012). Such variations influence the near-tropopause abundance of brominated VSLS (Hoyle et al., 2011; Liang et al., 2014) and other tracers, such as CO (Folkins et al., 2006).

Figures 12 and 13 show the simulated seasonal cycle of CHBr$_3$ and CH$_2$Br$_2$, respectively, at the base of the TTL and the cold point tropopause (CPT). CHBr$_3$ exhibits a pronounced seasonal cycle at the CPT, with virtually all models showing the same phase; with respect to the annual mean and integrated over the tropics, CHBr$_3$ is most elevated during boreal winter (DJF). The amplitude of the cycle varies considerably between models, with departures from the annual mean ranging from around ±10 to ±40%, in a given month (panel b of Fig. 12). Owing to its relatively long tropospheric lifetime, particularly in the TTL (> 1 year) (Hossaini et al., 2010), CH$_2$Br$_2$ exhibits a weak seasonal cycle at the CPT as it is less influenced by seasonal variations in transport.

Panels c and d of Figs. 12 and 13, also show the modelled absolute mixing ratios of CHBr$_3$ and CH$_2$Br$_2$ at the TTL base and CPT. Annually averaged, for CHBr$_3$, the model spread results in a factor of 3 difference in simulated CHBr$_3$ at both levels (similarly, for CH$_2$Br$_2$ a factor of 1.5). The modelled mixing ratios fall within the measurement-derived range reported by Carpenter and Reimann (2014). The MMM CHBr$_3$ mixing ratio at the TTL base is 0.51 ppt, within the 0.2–1.1 ppt measurement-derived range. At the CPT, the MMM CHBr$_3$ mixing ratio is 0.20 ppt, also within the measured range of 0.0–0.31 ppt. On average, the models suggest a $\sim 60\%$ gradient in CHBr$_3$ between the TTL base and tropopause. Similarly, the annual MMM CH$_2$Br$_2$ mixing ratio is 0.82 ppt at the TTL base, within the measured range of 0.6–1.2 ppt, and at the CPT is 0.73 ppt, within the measured range of 0.3–0.86 ppt. On average, the models show a CH$_2$Br$_2$ gradient of 10% between the two levels. These model absolute values are annual means over the whole
Figure 12. Simulated monthly mean anomalies of CHBr$_3$ volume mixing ratio (vmr), expressed as a percentage with respect to the annual mean, for (a) 200 hPa, the approximate base of the tropical tropopause layer (TTL) and (b) 100 hPa, the cold point tropopause (CPT). Panels (c, d) show the CHBr$_3$ vmr (ppt) at these levels. All panels show tropical (±20° latitude) averages over the full simulation period (1993–2012). See Fig. 3 for legend. The thick black line denotes the multi-model mean.

Figure 13. As Fig. 12 but for CH$_2$Br$_2$.

tropical domain. However, zonal variability in VSLS loading within the TTL is expected to be large (e.g. Aschmann et al., 2009; Liang et al., 2014), owing to inhomogeneity in the spatial distribution of convection and oceanic emissions. The Indian Ocean, the Maritime Continent (incorporating Malaysia, Indonesia, and the surrounding islands and ocean), central America, and central Africa are all convectively active regions, shown to experience particularly deep convective events, with the potential, therefore, to rapidly loft VSLS from the surface into the TTL (e.g. Gettelman et al., 2002, 2009; Hosking et al., 2010). As previously noted, the absolute values can vary, though generally the TransCom-VSLS models agree on the locations with the highest VSLS mixing ratios, as seen from the zonal CHBr$_3$ anomalies at the CPT shown in Fig. 14. These regions are consistent with the convective source regions discussed above. The largest CHBr$_3$ mixing ratios at the CPT are predicted over the tropical western Pacific (20° S–20° N, 100–180° E), particularly during DJF. Integrated over the tropical domain, this signal exerts the largest influence on the CHBr$_3$ seasonal cycle at the CPT. This result is consistent with the model intercomparison of Hoyle et al. (2011), who examined the seasonal cycle of idealised VSLS-like tracers around the tropopause, and reported a similar seasonality.

While meridionally, the width of elevated CHBr$_3$ mixing ratios during DJF is similar across the models, differences
during boreal summer (JJA) are apparent, particularly in the vicinity of the Asian monsoon (5–35°N, 60–120°E). Note, the CHBr$_3$ anomalies shown in Fig. 14 correspond to departures from the mean calculated in the latitude range of ±30°, and therefore encompass most of the monsoon region. A number of studies have highlighted (i) the role of the monsoon in transporting pollution from east Asia into the stratosphere (e.g. Randel et al., 2010) and (ii) its potential role in the troposphere-to-stratosphere transport of aerosol precursors, such as volcanic SO$_2$ (e.g. Bourassa et al., 2012; Fromm et al., 2014). For VSLS, and other short-lived tracers, the monsoon may also represent a significant pathway for transport to the stratosphere (e.g. Vogel et al., 2014; Orbe et al., 2015; Tissier and Legras, 2016). Here, a number of models show elevated CHBr$_3$ in the lower stratosphere over the monsoon region, though the importance of the monsoon with respect to the tropics as a whole varies substantially between the models. For example, from Fig. 14, models such as ACTM and UKCA show far greater enhancement in CHBr$_3$ associated with the monsoon during JJA, compared to others (e.g. MOZART, TOMCAT). A comparison of CHBr$_3$ anomalies at 100hPa but confined to the monsoon region, as shown in Fig. 15, reveals a monsoon signal in most of the models, but as noted above, the strength of this signal varies considerably. The STAG model, which does not include a treatment of deep convection and has been shown to have weak ventilation through the boundary layer (Law et al., 2008), exhibits virtually no CHBr$_3$ enhancement over the monsoon region.

The high-altitude model–model differences in CHBr$_3$, highlighted in Figs. 14 and 15, are attributed predominately to differences in the treatment of convection. Previous studies have shown that (i) convective updraft mass fluxes, including the vertical extent of deep convection (relevant for bromine SGI from VSLS), vary significantly depending on the implementation of convection in a given model (e.g. Feng et al., 2011) and (ii) that significantly different short-lived tracer distributions are predicted from different models using different convective parameterisations (e.g. Hoyle et al., 2011). Such parameterisations are often complex, relying on assumptions regarding detrainment levels, they trigger thresholds for shallow, mid-level and/or deep convection and they vary in their approach to computing updraft (and downdraft) mass fluxes. Furthermore, the vertical transport of model tracers is also sensitive to interactions of the convective parameterisation with the boundary layer mixing scheme (also parameterised) (Rybka and Tost, 2014). On the above basis and considering that the TransCom-VSLS models implement these processes in different ways (Table 2), it was not possible to disentangle transport effects within the scope of this project. However, no systematic similarities/differences between models according to input meteorology were apparent. Examining the difference between UKCA_HI and UKCA_LO reveals that horizontal resolution is a significant factor. The UKCA_HI simulation shows a greater role of the monsoon region, likely due to differences in the distribution of surface emissions (e.g. along longer coastlines in the higher resolution model) with respect to the occurrence of convection, as shown by Russo et al. (2015). Overall, aircraft VSLS observations within this poorly sampled region are required in order to elucidate further the role of the monsoon in the troposphere-to-stratosphere transport of brominated VSLS.

3.5 Stratospheric source gas injection of bromine and trends

In this section we quantify the climatological SGI of bromine from CHBr$_3$ and CH$_2$Br$_2$ to the tropical LS and examine interannual variability. The current measurement-derived range of bromine SGI ([3 × CHBr$_3$] + [2 × CH$_2$Br$_2$] at the tropical tropopause) from these two VSLS is 1.28 (0.6–2.65) ppt Br, i.e. uncertain by a factor of ~4.5 (Carpenter and Reimann, 2014). This uncertainty dominates the overall uncertainty on the total stratospheric bromine SGI range (0.7–3.4 ppt Br), which includes relatively minor contributions from other VSLS (e.g. CHBr$_2$Cl, CH$_2$BrCl and CHBrCl$_2$). Given that SGI may account for up to 76% of stratospheric Br$_y^{VSL}$ (Carpenter and Reimann, 2014) (note, Br$_y^{VSL}$ also includes the contribution of product gas injection), constraining the contribution from CHBr$_3$ and CH$_2$Br$_2$ is, therefore, desirable.

The TransCom-VSLS climatological MMM estimate of Br SGI from CHBr$_3$ and CH$_2$Br$_2$ is 2.0 (1.2–2.5) ppt Br, with the reported uncertainty from the model spread. CH$_2$Br$_2$ accounts for ~72% of this total, in good agreement with the ~80% reported by Carpenter and Reimann (2014). The model spread encompasses the best estimate reported by Carpenter and Reimann (2014), though our best estimate is 0.72 ppt (57%) larger. The spread in the TransCom-VSLS models is also 37% lower than the Carpenter and Reimann (2014) range, suggesting that their measurement-derived range in bromine SGI from CHBr$_3$ and CH$_2$Br$_2$ is possi-
bly too conservative, particularly at the lower limit (Fig. 16), and from a climatological perspective. We note that (i) the TransCom-VSLS estimate is based on models, shown here, to simulate the surface to tropopause abundance of CHBr₃ and CH₂Br₂ well and (ii) represents a climatological estimate over the simulation period, 1995–2012. The measurement-derived best estimate and range (i.e. that from Carpenter and Reimann, 2014) does not include the high-altitude observations over the tropical western Pacific obtained during the most recent NASA ATTREX missions. As noted in Sect. 3.3, mean CHBr₃ and CH₂Br₂ measured around the tropopause during ATTREX (2013/2014 missions), are at the upper end of the compilation of observed values given in the recent WMO Ozone Assessment Report (Tables 1–7 of Carpenter and Reimann, 2014). Inclusion of these data would bring the WMO SGI estimate from CHBr₃ and CH₂Br₂ closer to the TransCom-VSLS estimate reported here. For context, the TransCom-VSLS MMM estimate of Br SGI from CHBr₃ and CH₂Br₂ (2.0 ppt Br) represents 10 % of total stratospheric Br, (i.e. considering long-lived sources gases also) – estimated at ~20 ppt in 2011 (Carpenter and Reimann, 2014).

The TransCom-VSLS MMM SGI range discussed above is from CHBr₃ and CH₂Br₂ only. Minor VSLS, including CHBr₂Cl, CH₃BrCl, CHBrCl₂, C₂H₃Br, C₃H₅Br and C₃H₇Br, are estimated to contribute a further 0.08 to 0.71 ppt Br through SGI (Carpenter and Reimann, 2014). If we add this contribution on to our MMM estimate of bromine SGI from CHBr₃ and CH₂Br₂, a reasonable estimate of 1.28 to 3.21 ppt Br is derived from our results for the total SGI range. This range is 28 % smaller than the equivalent estimate of total SGI reported by Carpenter and Reimann (2014), because of the constraint on the contribution from CHBr₃ and CH₂Br₂, as discussed above.

Our uncertainty estimate on simulated bromine SGI (from the model spread) reflects inter-model variability, primarily due to differences in transport, but does not account for uncertainty on the chemical factors influencing the loss rate and lifetime of VSLS (e.g. tropospheric [OH]) – as all of the models used the same prescribed oxidants. However, Aschmann and Sinnhuber (2013) found that the stratospheric SGI of Br exhibited a low sensitivity to large perturbations to the chemical loss rate of CHBr₃ and CH₂Br₂; a ±50 % perturbation to the loss rate changed bromine SGI by 2 % at most in their model sensitivity experiments. Furthermore, our SGI range is compatible with recent model SGI estimates that used different [OH] fields; for example, Fernandez et al. (2014) simulated a stratospheric SGI of 1.7 ppt Br from CHBr₃ and CH₂Br₂.

We found no clear long-term transport-driven trend in the stratospheric SGI of bromine. Clearly, this result is limited to the study period examined and does not preclude potential future changes due to climate change, as suggested by some studies (e.g. Hossaini et al., 2012b). In terms of interannual variability, the simulated annual mean bromine SGI varied by ±5 % around the climatological mean (panel b of Fig. 16) over the simulation period (small in the context of total stratospheric Br, see above). Naturally, this encompasses interannual variability of both CHBr₃ and CH₂Br₂ reaching the tropical LS, the latter of which is far smaller, and given that CH₂Br₂ is the larger contributor to SGI, dampens the overall interannual variability. Note, interannual changes in emissions, [OH] or photolysis rates were not quantified here (only transport). On a monthly basis, the amount of CHBr₃ reaching the tropical LS can clearly exhibit larger variability. CHBr₃ anomalies (calculated as monthly departures from the climatological monthly mean mixing ratio) at the tropical tropopause are shown in Fig. 17. The Multivariate ENSO Index (MEI) – a time series which characterises ENSO intensity based on a range of meteorological and oceanographic components (Wolter and Timlin, 1998) – is also shown in Fig. 17. See also http://www.esrl.noaa.gov/psd/enso/mei/. The transport of CHBr₃ (and CH₂Br₂, not shown) to the tropical LS is strongly correlated (r values ranging from 0.6 to 0.75 across the ensemble) to ENSO activity over the eastern Pacific (owing to the influence of sea surface temperature on convection). For example, a clear signal of the very strong El Niño event of 1997/1998 is apparent in the models (i.e. with enhanced CHBr₃ at the tropopause) supporting the notion that bromine SGI is sensitive to such climate modes, in this region (Aschmann et al., 2011). However, when aver-
aged over the tropics, no strong correlation between VSLS loading in the LS and the MEI (or just sea surface temperature) was found across the ensemble. We suggest that zonal variations in sea surface temperature anomalies (and convective activity) associated with ENSO, with warming in some regions and cooling in others, has a cancelling effect on the tropical mean bromine SGI. Indeed, previous model studies have shown a marked zonal structure in CHBr$_2$ loading in the LS in strong ENSO years, with relative increases and decreases with respect to climatological averages depending on region (Aschmann et al., 2011). Further investigation, beyond the scope of this work, is needed to determine the sensitivity of total stratospheric Br$^\text{YSL}$ (i.e. including the contribution from product gas injection), to this and other modes of climate variability.

4 Summary and conclusions

Understanding the chemical and dynamical processes which influence the atmospheric loading of VSLS in the present, and how these processes may change in the future, is important to understand the role of VSLS in a number of issues. In the context of the stratosphere, it is important to (i) determine the relevance of VSLS for assessments of O$_3$ layer recovery timescales (Yang et al., 2014), (ii) assess the full impact of proposed stratospheric geoengineering strategies (Tilmes et al., 2012) and (iii) accurately quantify the ozone-driven radiative forcing of climate (Hossaini et al., 2015a). Here we performed the first concerted multi-model intercomparison of halogenated VSLS. The overarching objective of TransCom-VSLS was to provide a reconciled model estimate of the SGI of bromine from CHBr$_3$ and CH$_2$Br$_2$ to the lower stratosphere and to investigate inter-model differences due to emissions and transport processes. Participating models performed simulations over a 20-year period, using a standardised chemistry set-up (prescribed oxidants/photolysis rates) to isolate, predominantly, transport-driven variability between models. We examined the sensitivity of results to the choice of CHBr$_3$/CH$_2$Br$_2$ emission inventory within individual models, and also quantified the performance of emission inventories across the ensemble. The main findings of TransCom-VSLS are summarised below.

- The TransCom-VSLS models reproduce the observed surface abundance, distribution and seasonal cycle of CHBr$_3$ and CH$_2$Br$_2$, at most locations where long-term measurements are available, reasonably well. At most sites, (i) the simulated seasonal cycle of these VSLS is not particularly sensitive to the choice of emission inventory, and (ii) the observed cycle is reproduced well simply from seasonality in the chemical loss (a notable exception is at Mace Head, Ireland). Within a given model, absolute model–measurement agreement at the surface is highly dependent on the choice of VSLS emission inventory, particularly for CHBr$_3$ for which the global emission distribution and magnitude is somewhat poorly constrained. We find that at a number of locations, no consensus among models as to which emission inventory performs best can be reached. This is due to differences in the representation/implementation of transport processes between models which can significantly influence the boundary layer abundance of short-lived tracers. This effect was observed between CTM variants which, other than tropospheric transport schemes, are identical. A major implication of this finding is that care must be taken when assessing the performance of emission inventories in order to constrain global VSLS emissions, based on single model studies alone. However, we also find that within the tropics – where the troposphere-to-stratosphere transport of VSLS takes place – most models (~70%) achieve best agreement with measured surface CHBr$_3$ when using a bottom-up derived inventory, with the lowest CHBr$_3$ emission flux (Ziska et al., 2013). Similarly for CH$_2$Br$_2$, most (also ~70%) of the models achieve optimal agreement using the CH$_2$Br$_2$ inventory with the lowest tropical emissions (Liang et al., 2010), though agreement is generally less sensitive to the choice of emission inventory (compared to CHBr$_3$). Recent studies have questioned the effectiveness of using aircraft observations and global-scale models (i.e. the top-down approach) in order to constrain regional VSLS emissions (Russo et al., 2015). For this reason and given growing interest as to possible climate-driven changes in VSLS emissions (e.g. Hughes et al., 2012), online calculations (e.g. Lennartz et al., 2015) which (i) consider interactions between the ocean/atmosphere state (based on ob-
served seawater concentrations) and (ii) produce seasonally resolved sea-to-air fluxes, may prove a more insightful approach, over the use of prescribed emission climatologies, in future modelling work.

- The TransCom-VSLS models generally agree on the locations where CHBr$_3$ and CH$_2$Br$_2$ are most elevated around the tropopause. These locations are consistent with known convectively active regions and include the Indian Ocean, the Maritime Continent and wider tropical western Pacific and the tropical eastern Pacific, in agreement with a number of previous VSLS-focused modelling studies (e.g. Aschmann et al., 2009; Pioso et al., 2010; Hossaini et al., 2012b; Liang et al., 2014). Owing to significant inter-model differences in transport processes, both the absolute tracer amount transported to the stratosphere and the amplitude of the seasonal cycle varies among models. However, of the above regions, the tropical western Pacific is the most important in all of the models (regardless of the emission inventory), due to rapid vertical ascent of VSLS simulated during boreal winter. In the free troposphere, the models reproduce observed CHBr$_3$ and CH$_2$Br$_2$ from the recent SHIVA and CAST campaigns in this region to within $\pm$16 and $\pm$32 ppt Br, respectively. However, at higher altitudes in the TTL the models generally (i) underestimated CHBr$_3$ between 14 and 16 km observed during the 2014 NASA ATTREX mission in this region but (ii) fell within $\pm$1$\sigma$ of the observed mean around the tropical tropopause ($\sim$17 km). Generally good agreement with high-altitude aircraft measurements of VSLS around the tropopause in the eastern Pacific was also obtained. During boreal summer, most models show elevated CHBr$_3$ around the tropopause above the Asian monsoon region. However, the strength of this signal varies considerably among the models, with a spread that encompasses virtually no CHBr$_3$ enhancement over the monsoon region to strong (85%) CHBr$_3$ enhancements at the tropopause, with respect to the zonal average. Measurements of VSLS in the poorly sampled monsoon region from the upcoming StratoClim campaign (http://www.stratoclim.org/) will prove useful in determining the importance of this region for the troposphere-to-stratosphere transport of VSLS.

- Climatologically, we estimate that CHBr$_3$ and CH$_2$Br$_2$ contribute 2.0 (1.2–2.5) ppt Br to the lower stratosphere through SGI, with the reported uncertainty due to the model spread. The TransCom-VSLS best estimate of 2.0 ppt Br is (i) $\sim$57% larger than the measurement-derived best estimate of 1.28 ppt Br reported by Carpenter and Reimann (2014), and (ii) the TransCom-VSLS range (1.2–2.5 ppt Br) is $\sim$37% smaller than the 0.6–2.65 ppt Br range reported by Carpenter and Reimann (2014). From this we suggest that, climatologically, the Carpenter and Reimann (2014) measurement-derived SGI range, based on a limited number of aircraft observations (with a particular paucity in the tropical western Pacific), is potentially too conservative at the lower limit, although we acknowledge that our uncertainty estimate (the model spread) does not account for a number of intrinsic uncertainties within global models, for example, tropospheric [OH] (as the models used the same set of prescribed oxidants). No significant transport-driven trend in stratospheric bromine SGI was found over the simulation period, though interannual variability was of the order of $\pm$5%. Loading of both CHBr$_3$ and CH$_2$Br$_2$ around the tropopause over the eastern Pacific is strongly coupled to ENSO activity, but no strong correlation to ENSO or sea surface temperature was found when averaged across the wider tropical domain.

Overall, results from the TransCom-VSLS model intercomparison support the large body of evidence that natural VSLS contribute significantly to stratospheric bromine. Given suggestions that emissions of VSLS from the growing aquaculture sector will likely increase in the future (WMO, 2014; Phang et al., 2015) and that climate-driven changes to ocean emissions (Tegtmeier et al., 2015), tropospheric transport and/or oxidising capacity (Dessens et al., 2009; Hossaini et al., 2012a) could lead to an increase in the stratospheric loading of VSLS, it is paramount to constrain the present-day Br$_{VSLS}$ contribution to allow any possible future trends to be determined. In addition to SGI, this will require constraint on the stratospheric product gas injection of bromine which conceptually presents a number of challenges for global models given its inherent complexity.

5 Data availability

The observational data used in this paper are available at http://www.esrl.noaa.gov/gmd/dv/ftpdata.html (NOAA surface data; NOAA, 2016), http://dx.doi.org/10.3334/CDIAC/hippo_012 (HIPO aircraft data; Wofsy et al., 2016), and https://espoarchive.nasa.gov/archive/browse (NASA aircraft data; NASA, 2016). Model data are available on request: please contact Ryan Hossaini (r.hossaini@lancaster.ac.uk).

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