Changes Overall:

We used latexdiff to create the changes file, hope this worked out. Many figures that required changes are updated and all text changes should be reflected in the latexdiff PDF.

We attached the responses + the changed document at the end here.

Thanks,
Christian Frankenberg
We thank Anonymous Referee #1 for a positive and thorough review. In the following, we will respond to the Reviewers comments step by step.

Minor issues: Page 10, line 24: I think “an excellent agreement” is a bit too strong taking into account that there are differences up to 4 ppm (same page, line 20). I recommend to change this to “good agreement” or so.

→ Changed to “good”. We also removed the qualitative comparison statement between MACC and CT2013B as it wasn’t really too substantiated.

Similar on page 11, line 21, with the statement “compare extremely well” already in the first sentence before any comparison results are shown and discussed. I also do not think that agreement within 1 ppm and outliers up to 3 ppm is best characterized by “extremely well” (NOTE: in the text on page 11 the unit ppb is given two times (lines 24 and 25) but I guess this should be ppm!).

→ Removed “extremely” and changed ppb to ppm (old methane habit).

Page 10, line 25: “In some cases, MACC seems to compare somewhat better, . . .”. A MACC colleague is co-author but no CarbonTracker colleague. I wonder if NOAA would agree with this statement. I also wonder if NOAA needs to be acknowledged for their data.

→ We removed this statement altogether and added an acknowledgement to that effect, esp. as Andy Jacobson was involved in our discussions but not listed as co-author.

Page 11, line 6: “SCIAMACHY data over the oceans is not yet matured as is has no dedicated Glint mode.” Sounds a bit strange (even if “is” typo corrected). I recommend to replace this with “SCIAMACHY data have not been used as it has no dedicated glint mode and the SCIAMACHY products (e.g., Reuter et al., 2011) are limited to retrievals over land”. Reuter et al., 2011: "Retrieval of atmospheric CO2 with enhanced accuracy and precision from SCIAMACHY: Validation with FTS measurements and comparison with model results", J. Geophys. Res.

→ done

Page 16, line 8: “. . . indicates that GOSAT compares slightly better overall.” Compared to what?

→ removed that sentence and added “comparable to those with models” to the previous sentence.

Page 21, Tab. 1: Why is the GOSAT sigma only 0.45 ppm (as far as I know the GOSAT XCO2 single measurement precision is about 2 ppm; or have data been averaged?)? Please check and add additional explanation if necessary.
Yes, multiple GOSAT soundings are used per HIPPO profile and averaged (as stated before, “For the GOSAT comparison, we require more than 5 co-located GOSAT measurement per HIPPO profile.”). We changed that sentence to “For the GOSAT comparison, we require at least 5 co-located GOSAT measurement per HIPPO profile, all of which are subsequently averaged before comparison against HIPPO”. It was also stated before that “For each match, the standard error in the GOSAT XCO2 average is computed using the standard deviation of all corresponding GOSAT colocations divided by the square root of the number of colocations.”

Fig. 3: Bottom, middle: Profiles only partially visible as overplotted by legend. Please improve.
  ➔ done

Fig. 5, left: Possibly data points only partially visible as overplotted by legend. Please improve.
  ➔ done

Fig. 6, left: Data points only partially visible as overplotted by legend.
  ➔ done

Please improve. Fig. 7, right: Symbols for models very difficult to see in printout.
  ➔ We would ask the editorial office to check into that issue.

No reference to Figs. 5 and 6 in text (should be somewhere in Sect. 4).
  ➔ Added “In terms of XCO$_2$, both atmospheric models used here compare well against HIPPO, as can be seen in Figures 5 and 6. (at beginning of Sec. 4). Thanks for noticing this!

No reference to Fig. 11 in text (should be somewhere in Sect. 5.2).
  ➔ Added, thanks

Typos: Page 10, line 6: “are usually 162253”?
  ➔ Typical LaTeX typing error (accidentally copying something without noticing it), apologies.

Caption Tab. 1: “of different compared to” -> “of the difference compared to”
  ➔ Done, thanks
We thank Anonymous Referee #2 for a positive and thoughtful review. In the following, we will respond to the Reviewers comments step by step.

Minor revisions: p5, l8-10: This enables ... denoted XCO2 I think this statement would be more clear to the reader when a line is added to indicate that an extension is needed above 14 km. Then you can indeed state that this extension is of limited consequence since most of the variability in XCO2 stems from the troposphere which is covered by the HIPPO profiles.

⇒ Good point. We changed to: “This enables a comparison of individual sub-columns of air but also of column-averaged mixing ratios of CO$_2$, denoted XCO$_2$, if the profile can be reliably extended above 14 km. As the troposphere dominates the variability in XCO$_2$, errors induced by extending profiles are expected to be small.”

p7, l11-12: As most ... analysis. Add a line why SCIAMACHY does not provide data over oceans

⇒ Added “…because it lacks a dedicated Glint measurement mode” and explained it better later as well, as requested by Rev. #1.

p7, l16: short-wave → short-wave infrared

⇒ done

p8, l11: How can averaging lead to the reduction of systematic errors?

⇒ Removed systematic here.

p9, l9-11: Validation ... (Olsen and Licata, 2014). If Olsen and Licata already have compared IR/MW L2 and IR-Only L2 against HIPPO, then I would expect a sentence explaining how the current study differs and/or extends wrt. the cited paper.

⇒ We rephrased and extended that sentence to reflect the main differences (using models to fill up the profile).: Olsen and Licata (2014) compare the IR/MW based and IR-Only based CO2 retrievals over the globe for 2010-2011 and for collocations with the deep-dip HIPPO-2, HIPPO-3, HIPPO-4 and HIPPO-5 profiles. Their global analysis reveals that the zonal monthly average difference rarely exceeds 0.5 ppm save at the high northern latitudes in January and October where fluctuations resulting from small number statistics dominate. Their analysis against HIPPO employs only the deep-dip measured profiles, i.e. those in which the aircraft reached the 190 hPa pressure level, to ensure good in situ measurement coverage of the AIRS sensitivity profile and to minimize the error introduced by their simple approximation of extending the aircraft profile into the stratosphere by replicating the highest altitude measurement. During the HIPPO-2 and HIPPO-3 campaigns, the AMSU channel 5 noise figure was acceptable, whereas during the HIPPO-4 and HIPPO-5 campaigns it progressively degraded at a rapid rate. For all campaigns, the two sets of collocations, averaging AIRS retrievals within ±24 hours and 500 km of the aircraft profile, exhibit the same bias and RMS to within 1 ppm for |lat| ≤ 60°. The current study extends the in situ measurements to higher altitude by the means of CarbonTracker and MACC model output, thereby allowing use of all HIPPO profiles rather than only the deep-dip profiles. Our results are statistically consistent with the
latitude-dependent biases reported by Olsen and Licata (2014) and give a more detailed view of the scatter as a function of latitude.

p9, l14-18: For the differences ... should dominate If you first extend the HIPPO profile with model data, then integrate, and finally subtract the integrated model data, does the part above flight altitude not exactly cancel? HIPPO (< 14 km) + model (> 14 km) - model (0-TOA) = HIPPO (< 14 km) + model (> 14 km) - model (< 14 km) - model (> 14 km) = C2 HIPPO (< 14 km) - model (< 14 km). So, I do not see how the extension can contribute to the difference between HIPPO and model.

⇒ About 20% of the total column is located above 14km and not all HIPPO profiles extended that far. If we use part of the model, these values indeed cancel and yield exactly 0 difference, making the agreement somewhat better. With the 80/20 weighting, it is similar to saying that delta-XCO2 is 0.8*(Model-HIPPO)+0.2*(model-model=0), thus potentially always dampening the differences. Or, if the profile extended only to 10km, dampening it even further.

p10, l14-23: Figure 4 ... potential biases. HIPPO 3 is nicely explained in this paragraph, but HIPPO 5 is depicted in the Figure but not mentioned. Any comment that the authors can make on the MACC and CT differences/similarities?

⇒ Added “In HIPPO 5, at the end of the growing season, the situation is reversed as the profile slopes change sign after the large CO$_2$ uptake during summer.” And “For HIPPO 5, the deviations for CT2013B are somewhat smaller but it can be seen that most models suffer from these potential biases if large vertical gradients exist.”

p11, l2-10: Here, we look ... in the future. This alinea is mostly about measurements and campaigns that are not treated in the paper. I understand why the authors like to mention this, but maybe the conclusion, which includes a future outlook mentioning OCO-2, is the better spot for this.

⇒ This is indeed better, we moved this to the Conclusions.

p11, l11-19: For the comparison ... were the truth). This my strongest comment on the paper: Since the requirements on XCO2 are so stringent, it matters for the comparisons in this paper how exactly 1) the HIPPO profiles are extended, 2) the averaging kernel is applied, and 3) the null-space is attributed. I would recommend to incorporate a small section/paragraph explaining the mathematical details. Questions that come to mind: Is the model information just attached to the HIPPO pro- file? If a jump would appear in such a profile, how is that treated? Is the smoothed (extended) HIPPO profile compared to the GOSAT profile without null-space contribution, or is there also a null-space contribution to the smoothed HIPPO profile? If the latter, which reference is used? The same as in the GOSAT retrievals, or the model?
This is a good point even though we prefer to keep this short in the paper. Re 1). The HIPPO profiles are extended with the model data before applying the averaging kernel correction. 2). The AK corrected HIPPO values are computed as xa+A(xt-xa) with the a priori profile xa and the “true” profile xt (HIPPO + model). For GOSAT, the column averaging kernel was used, for TES and AIRS the averaging kernel for the respective retrieval layer.

We have not tested the impact of a jump in a profile; in the manuscript, a simple profile extension was performed without testing smoothness. In most cases, the impact should be relatively small. The null space contribution in GOSAT comparisons should be small as the column averaging kernels are relatively large throughout the entire column. In general, HIPPO data has always been filled in with model data, not satellite priors. We added

For GOSAT: “For the HIPPO comparison against GOSAT data, we take the instrument sensitivity into account by applying the averaging kernel to the difference of the true profile (using the model-extended HIPPO dataset as truth) and the respective a priori profile. We perform this correction using both model extensions independently and then use the average of the two.”

For TES: “For the comparison with TES, we use the 510\,hPa retrieval layer and apply averaging kernel corrections using model-extended HIPPO data as \textit{truth}, using both models independently and averaging results after averaging kernel correction.”

For AIRS: “For the comparison with AIRS (Fig\textsubscript{`\ref{fig:HIPPO_AIRS}}), the sensitivity maximum varies around 300\,hP and we apply the averaging kernels similarly to TES.”

We hope this will clarify the issue.

p11, l22-24: Even after ... for MACC. Please refer to Figs 5 and 6

--- done

p11, l22: Even after normalization It is clear how the HIPPO data is corrected, but how is the other data corrected? With the HIPPO value, or with the average value of the particular model?

--- With the HIPPO value. We added a sentence “For each campaign, we also normalize all data with the respective campaign average of the HIPPO dataset.”

p13, l23: lower left quadrant Maybe the authors would like to note that these points are also outliers in the CT comparison. Not as strong as in the case of MACC, but still in C3 the same quadrant, which may be an indication that the transport errors in both models are roughly equal and/or the GFED data is somewhat off.

--- We mentioned that “both models” show that feature.

p24, Fig 3: There are some strong excursions in the HIPPO profiles close to the surface; any explanation for these?

--- These might be caused by dips close to the surface with HIPPO, potentially coming from the land data. It should not really affect XCO2 a lot as it only affects a small subcolumn.
HIPPO-1, 3, and 4 (and possibly 5), the differences between HIPPO and MACC resp. CT differs significantly for > 70N. Any explanation for this behaviour?

→ We agree, there seem to be substantial differences but we don’t have any explanation yet for this and would not like to speculate too much.

Please, reposition the legend box; CT-HIPPO 5 is barely visible.

→ done

p26-p28, Fig 5-7: Mention the shift for both axes

→ we now state “Scatterplot of normalized (with campaign average) CO2…”

p31,p32, Fig 10,11: Mention the shift for both axes

→ see above

p4, l5: Greenhouse Gas Observing → Greenhouse Gases Observing

→ done

p4, l5: haven → have

→ fixed, thanks.

p4, l11: sensing measurement → sensing measurements

→ done

p5, l5-8: This sentence does not have a verb. Suggestion: The HIAPER Polo-to-Pole Observations (HIPPO) project consists of a sequence of ...

→ replaced “sampling” with “sampled”

p6, l23: LSCE. To be on the safe side I would explicitly write out this acronym

→ Done

p9, l21-22: consistent between model, → consistent between the two models,

→ done

p10, l6: usually 162 253 → usually

→ done

There are several places where ppb is used in stead of ppm: p11, l24 p11, l25 p21, Table 1 (2 instances)

→ done

p16, l11: that → than

→ done, thanks
Using airborne HIAPER Pole-to-Pole Observations (HIPPO) to evaluate model and remote sensing estimates of atmospheric carbon dioxide

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Abstract. In recent years, space-borne observations of atmospheric carbon-dioxide (CO2) have become increasingly used in global carbon-cycle studies. In order to obtain added value from space-borne measurements, they have to suffice stringent accuracy and precision requirements, with the latter being less crucial as it can be reduced by just enhanced sample size. Validation of CO2 column averaged dry air mole fractions (XCO2) heavily relies on measurements of the Total Carbon Column Observing Network TCCON. Owing to the sparseness of the network and the requirements imposed on space-based measurements, independent additional validation is highly valuable. Here, we use observations from the HIAPER Pole-to-Pole Observations (HIPPO) flights from 01/2009 through 09/2011 to validate CO2 measurements from satellites (GOSAT, TES, AIRS) and atmospheric inversion models (CarbonTracker CT2013B, MACC v13r1). We find that the atmospheric models capture the XCO2 variability observed in HIPPO flights very well, with correlation coefficients (r2) of 0.93 and 0.95 for CT2013B and MACC, respectively. Some larger discrepancies can be observed in profile comparisons at higher latitudes, esp. at 300 hPa during the peaks of either carbon uptake or release. These deviations can be up to 4 ppm and hint at misrepresentation of vertical transport.

Comparisons with the GOSAT satellite are of comparable quality, with an r2 of 0.85, a mean bias μ of -0.06 ppm and a standard deviation σ of 0.45 ppm. TES exhibits an r2 of 0.75, μ of 0.34 ppm and σ of 1.13 ppm. For AIRS, we find an r2 of 0.37, μ of 1.11 ppm and σ of 1.46 ppm, with latitudedependent biases. For these comparisons at least 6,20 and 50 atmospheric soundings have been averaged for GOSAT, TES and AIRS, respectively. Overall, we find that GOSAT soundings over the remote pacific ocean mostly meet the stringent accuracy requirements of about 0.5 ppm for space-based CO2 observations.

1 Introduction

Space-borne measurements of atmospheric carbon dioxide can provide unique constraints on carbon exchanges between land, ocean, and atmosphere on a global scale. Results from the Scanning Imaging Absorption Spectrometer for Atmospheric CHartography SCIAMACHY (e.g. Schneising et al., 2014) and the Greenhouse Gas Observing Satellite GOSAT (Lindqvist et al., 2015) have shown to reproduce the seasonal cycle as well as the secular trend of total column CO2 abundances reasonably well (Kulawik et al., 2015). However, accuracy requirements are very stringent (Miller et al., 2007), warranting large scale biases of less than 0.5–1 ppm, being less than 0.3% of the global background concentration. This is one of the most challenging remote sensing measurements from space as we not only want to reproduce known average seasonal cycles.
and trends but also small inter-annual deviations, resolved to subcontinental scales. There have been successes in doing so (e.g. Basu et al. (2014); Guerlet et al. (2013)) but controversies regarding overall retrieval accuracy on the global scale still remain (Chevallier, 2015) and can neither be fully refuted nor confirmed with validations against the Total Column Carbon Observing Network (TCCON) (e.g. Kulawik et al., 2015). In addition, total uncertainties might be a mix of measurement and modeling biases (Houweling et al., 2015), for which uncertainties in vertical transport can play a crucial role (Stephens et al., 2007; Deng et al., 2015).

In this manuscript, we use the term accuracy to refer to systematic errors that remain after infinite averaging and can vary in space and time. Globally constant systematic errors are easy to correct for but those with spatio-temporal dependencies can have a potentially large impact on flux inversions.

Given the importance of the underlying scientific questions regarding the global carbon cycle and the challenging aspect of both the remote sensing aspect as well as the atmospheric inversion, every additional independent validation beyond ground-based data can be crucial. Here, we use measurements from the HIAPER Pole-to-Pole Observations (HIPO) programme (Wofsy, 2011) to evaluate both atmospheric models as well as remotely sensed estimates of atmospheric CO₂.

2 Data description

2.1 HIPPO

The HIAPER Pole-to-Pole Observations (HIPPO) project, a sequence of five global aircraft measurement programs, sampled the atmosphere from (almost) the North Pole to the coastal waters of Antarctica, from the surface to 14 km, spanning the seasons (Wofsy, 2011). This enables a comparison of both individual sub-columns of air but also integrating the atmosphere across the troposphere, which dominates variability in the column-averaged mixing ratios of CO₂, denoted XCO₂, if the profile can be reliably extended above 14 km. As the troposphere dominates the variability in XCO₂, errors induced by extending profiles are supposed to be small. The campaigns covered different years as well as different seasons, namely: HIPPO 1: 8 January-30 January 2009, HIPPO 2: 31 October-22 November 2009, HIPPO 3: 24 March-16 April 2010, HIPPO 4: 14 June-11 July 2011, HIPPO 5: 9 August-9 September 2011.

Figure 1 shows an overview of the locations of the HIPPO profiles taken during different campaigns. As the 5 campaigns covered the years 2009 through 2011, we normalized the latitudinal cross section plot by subtracting the average XCO₂ around 50 degrees south. In the southern hemisphere, the shape of the latitudinal gradients only changes marginally between seasons while the amplitude at the higher latitudes in the north spans about 10 ppm, with the strongest drawdown during Aug/Sep for HIPPO 5 and the highest concentrations during HIPPO 3 in Mar/Apr. The dataset thus covers a wide range of atmospheric CO₂ profiles especially in the northern hemisphere where the strong biogenic cycle causes strong seasonality in CO₂ fluxes.

2.2 Atmospheric models

For the comparison of HIPPO against model data as well as for a more robust comparison of HIPPO against total column satellite CO₂ observations, we use two independent atmospheric models that both provide 4D CO₂ fields (space and time) that are consistent with in-situ measurements of atmospheric CO₂. The main differences between those are the use of a different inversion scheme as well as underlying transport model. In addition, both models were used to extend individual HIPPO profiles from the highest flight altitude to the top of atmosphere when comparing to total column estimates from the satellite.

2.2.1 CarbonTracker CT2013B

CarbonTracker (Peters et al. (2007) with updates documented at http://carbontracker.noaa.gov) is a CO₂ modeling system developed by the NOAA Earth System Research Laboratory. CarbonTracker (CT) estimates surface emissions of carbon dioxide by assimilating in situ data from NOAA observational programs, monitoring stations operated by Environment Canada, and numerous other international partners using an ensemble Kalman filter optimization scheme built around the TM5 atmospheric transport model (Krol et al. (2005); http://www.phys.uu.nl/~tm5/). Here we use the latest release of CarbonTracker, CT2013B, which provides CO₂ mole fraction fields globally from 2000-2012. In this study, we interpolate modeled CO₂ mole fractions to the times and locations of individual HIPPO observations.

2.2.2 MACC v13r1

Monitoring Atmospheric Composition and Climate (MACC, http://www.copernicus-atmosphere.eu/) is the European Union-funded project responsible for the development of the pre-operational Copernicus atmosphere monitoring service. Its CO₂ atmospheric inversion product relies on a variational Bayesian formulation, developed by LSCE (Le Laboratoire des Sciences du Climat et de l’Environnement), that estimates 8-day grid-point daytime/nighttime CO₂ fluxes and the grid point total columns of CO₂ at the initial time step of the inversion window. It uses the global tracer transport model LMDZ (Hourdin et al., 2006), driven by the wind analyses from the ECMWF. Version 13r1 of the product covers the period from 1979 to 2013, at horizontal resolution 3.75° × 1.9° (longitude–latitude). It assimilated the dry air mole fraction measurements from 131 CO₂ stations over the
globe in a unique 35-year assimilation window (see the list of sites in Tables S1 and S2 of Chevallier 2015). For this study, the model simulation has been interpolated to the time and location of the individual observations using the subgrid parametrization of the LMDZ advection scheme in the 3 dimensions of space (Hourdin and Armengaud, 1999). For the sake of brevity, we refer to MACC version 13r1 simply as MACC.

2.3 Satellite data

We use remotely sensed CO_2 observations from three different instruments, namely GOSAT, the Thermal Emission Sounder TES and the Atmospheric Infrared Sounder AIRS.

As most HIPPO profiles took place over the oceans, SCIAMACHY was not included in the analysis because it lacks a dedicated Glint measurement mode. While GOSAT CO_2 is representative of the column averaged dry mole fraction (XCO_2), both TES and AIRS are most sensitive to the atmosphere around 500 and 300 hPa, respectively.

2.3.1 GOSAT (ACOS B3.5)

GOSAT takes measurements of reflected sunlight in three short-wave infrared bands with circular footprints (diameter of 10.5 km) at nadir (Hamazaki et al., 2005; Kuze et al., 2009). Science data is starting in July 2009. In this work, we use column averaged dry air mole fraction (XCO_2) retrievals produced by NASA’s Atmospheric CO_2 Observations from Space (ACOS) project, version 3.5 (see O’Dell et al. (2012) for retrieval details), which is very similar to the B3.4 version described in https://co2.jpl.nasa.gov/static/docs/v3.4_DataUsersGuide-RevB_131028.pdf. The data and bias correction as used here is identical to the dataset investigated in Kulawik et al. (2015).

2.3.2 TES

TES is on the Earth Observing System Aura (EOS-Aura) satellite and makes high spectral resolution nadir measurements in the thermal infrared (660 cm^{-1} – 2260 cm^{-1}, with unapodized resolution of 0.06 cm^{-1}, apodized resolution of 0.1 cm^{-1}). TES was launched in July 2004 in a sun-synchronous orbit at an altitude of 705 km with an equatorial crossing time of 13:38 (local mean solar time) and with a repeat cycle of 16 days. From September, 2004 through June, 2011, TES collected “global survey” observations, averaging ≈500 good quality CO_2 day/night and land/ocean observations with cloud optical depth less than 0.5 between...
40S and 45N. The peak sensitivity of CO$_2$ is about 500 hPa, with full-width half-maximum sensitivity between 200 and 800 hPa. TES CO$_2$ requires averaging to reduce random and systematic errors, which can approach $\approx$6 ppm for a single observation to $\approx$1.3 ppm for monthly regional scales. For more details on TES CO$_2$, see Kulawik et al. (2013).

### 2.3.3 AIRS (v5)

The AIRS Version 5 (V5) tropospheric CO$_2$ product is a retrieval of the weighted partial-column dry volume mixing ratio characterizing the mid- to upper-tropospheric CO$_2$ concentration. The product is derived by the technique of Vanishing Partial Derivatives (VPD) described in Chahine et al. (2005) and is reported at a nominal nadir resolution of 90 km x 90 km over the globe over the latitude range 60S to 90N and time span September 2002 to present.

The VPD method assumes a CO$_2$ profile that is a linearly time-dependent global average constant volume mixing ratio throughout the atmosphere. Using that prior profile, the VPD derives CO$_2$ by shifting the CO$_2$, T, q and O$_3$ profiles and minimizing the residuals between the cloud-cleared radiances and those resulting from the forward calculation for channel subsets selected to avoid contamination by surface emission (except in regions of high topography). Further, it localizes the maximum sensitivity to variations of CO$_2$ concentration to the pressure regime spanning 300 hPa to 700 hPa.

In normal practice, the AIRS Level 2 products ingested by the CO$_2$ post-processing retrieval stage are retrieved using the combination of the infrared instrument and a companion Advanced Microwave Sounding Unit (AMSU). The 5-7 year expected lifetime of AMSU based on NOAA experience is much shorter than that of the AIRS instrument, so an alternate Level 2 retrieval using only the infrared radiances (AIRS_Only) was developed. The VPD retrieval normally ingests the combined IR/MW retrieval system products. Beginning in January 2011 the degradation of AMSU channel 5 noise figure significantly reduced the IR/MW L2 product yield so that the ingest was shifted to the IR-Only L2 product. Validation against HIPPO measurements of the CO$_2$ retrievals resulting from ingesting.

Olsen and Licata (2014) compare the IR/MW L2-based and IR-Only L2 products indicates that the products are equivalent (Olsen and Licata, 2014) based CO$_2$ retrievals.
over the globe for 2010-2011 and for collocations with the deep-dip HIPPO-2, HIPPO-3, HIPPO-4 and HIPPO-5 profiles. Their global analysis reveals that the zonal monthly average difference rarely exceeds 0.5 ppm save at the high northern latitudes in January and October where fluctuations resulting from small number statistics dominate. Their analysis against HIPPO employs only the deep-dip measured profiles, i.e. those in which the aircraft reached the 190 hPa pressure level, to ensure good in situ measurement coverage of the AIRS sensitivity profile and to minimize the error introduced by their simple approximation of extending the aircraft profile into the stratosphere by replicating the highest altitude measurement. During HIPPO-2 and HIPPO-3 campaigns AMSU channel 5 noise figure was acceptable, whereas during HIPPO-4 and HIPPO-5 campaigns it had progressively degraded at a rapid rate. For all campaigns, the two sets of collocations, averaging AIRS retrievals within ±24 hours and 500 km of the aircraft profile, exhibit the same bias and RMS to within 1 ppm for |lat| ≤ 60°. The current study extends the in situ measurements to higher altitude by the means of CarbonTracker and MACC model output thereby allowing use of all HIPPO profiles rather than only the deep-dip profiles. Our results are statistically consistent with the latitude dependent biases reported by Olsen and Licata (2014) and give a more detailed view of the scatter as a function of latitude.

3 HIPPO – Model inter-comparisons

Figure 2 shows an overview of model-HIPPO differences at 3 pressure levels as well as XCO₂, the total column average. For the differences in XCO₂, the respective model has been used to extend the HIPPO profiles from its highest altitude to the top of the atmosphere, hence part of the smaller differences observed in XCO₂ comparisons can stem from the fact that the model contributes slightly to the HIPPO based XCO₂ as well, though the tropospheric variability should dominate. As can be seen in the left panels, not all HIPPO profiles extend up to 300hPa.

Unsurprisingly, model-data mismatches at individual levels are substantially higher than in the total column, about a factor 2. Many differences are not consistent between model the two models, for example during HIPPO 4N, extending from West Papua northwards. In MACC, there is first a substantial underestimation throughout the profile and then an overestimation further north. In CT2013B, no obvious discrepancies can be observed. In other areas, such as the same HIPPO 4N path south of Alaska, MACC appears rather consistent but CT2013B is much higher at 800hPa but much lower at 500hPa, with a slight underestimate in the total column.

Figure 3 provides an in-depth review of HIPPO – model comparisons for profiles averaged by latitudinal bands and campaign. In most cases, profiles agree to within 1 ppm with a few notable exceptions, mostly at higher latitudes during the drawdown or respiration maximum in HIPPO 5 and 3, respectively. These are typically associated with steep vertical gradients around 300hPa, both in HIPPO 5 and 3, albeit with different signs. In most other cases, the differences even in the profiles are usually below 1 ppm, underlining the stringent accuracy requirements for space based CO₂ measurements, as atmospheric models optimized with respect to the ground-based network already model oceanic background concentrations fairly well. However, the caveat is that also these ground-based stations are located in remote regions, ideally not affected by local sources. On smaller spatial scales near sources, space-based measurements can provide valuable information even in the presence of potential large-scale biases.

Figure 4 shows an in-depth comparison of the largest model-HIPPO discrepancies, namely the high latitude profiles during HIPPO 3 and 5. As one can see on the left panels, the seasonal cycles in the mid-troposphere and at 200hPa can be opposite, with large CO₂ values in the upper atmosphere during the largest CO₂ drawdown and vice versa during the peak of respiration. Model-HIPPO mismatches are most obvious and similar between models in HIPPO 3 (Mar/Apr 2010), with differences reaching up to 4 ppm at 300 hPa. This is consistent with a comparison against the GEOS-Chem model by Deng et al. (2015), who studied the impact of discrepancies in stratosphere–troposphere exchange on inferred sources and sinks of CO₂. It is in HIPPO 5, at the end of the growing season, the situation is reversed as the profile slopes change sign after the large CO₂ uptake during summer. For HIPPO 5, the deviations for CT2013B are somewhat smaller but it can be seen that most models suffer from these potential biases.

if large vertical gradients exist. Overall, both CT2013B as well as MACC show an excellent a good agreement with HIPPO over the oceans. In some cases, MACC seems to compare somewhat better, which might be related to the longer inversion window of MACC, which can have an impact over remote areas such as the Pacific Ocean. However, this statement cannot be generalized as it may be specific to remote areas with low measurement density and be very different elsewhere.

4 Comparisons of column-averaged mixing ratios

Here, we look at column-averaged dry air mole fractions XCO₂, derived using absorption spectroscopy of reflected sun-light recorded by near-infrared spectrometers such as SCIAMACHY, GOSAT or OCO-2. In this paper, we only used GOSAT data as it is the only instrument having sampled in Glint mode during the HIPPO investigation. SCIAMACHY data over the oceans is not yet matured as is have not been used as it has no dedicated Glint mode glint mode and the SCIAMACHY products (e.g., Reuter et al. (2011))
Figure 3. Summary of averaged CO$_2$ HIPPO profiles in ppm (left column) and model-HIPPO differences (middle and right column), separated by latitudinal bands (color-coded) and HIPPO campaign (separate rows).
Averaging kernel to the take campaign period and not by the secular trend. For this, we use the respective atmospheric model to compare against each other because, as evident in Fig. 2, there are regions where either one or the other model is in better agreement with HIPPO. In conclusion, one can state that most regions where either one or the other model is in better agreement, the correlation coefficients and slopes are $r^2=0.93$ (slope=1.00) for CT2013B and $r^2=0.95$ (slope=1.00) for MACC. South of 20N, almost all data-points lie within ±1 ppm with some outliers of up to 3 ppm at higher latitudes, mostly over the continents (see Fig. 2). These numbers should not be used to compare the models against each other because, as evident in Fig. 2, there are regions where either one or the other model is in better agreement with HIPPO.

4.1 Atmospheric Models

In terms of XCO₂, both atmospheric models used here compare extremely well against HIPPO, as can be seen in Figures 5 and 6. Even after normalization with the campaign average, the correlation coefficients and slopes are $r^2=0.93$ (slope=1.00) for CT2013B and $r^2=0.95$ (slope=1.00) for MACC. South of 20N, almost all data-points lie within ±1 ppm with some outliers of up to 3 ppm at higher latitudes, mostly over the continents (see Fig. 2). These numbers should not be used to compare the models against each other because, as evident in Fig. 2, there are regions where either one or the other model is in better agreement with HIPPO. In conclusion, one can state that most model mismatches are below 1 ppm in remote areas such as the oceans and can reach 2-3 ppm over the continents with potentially higher values in under-sampled areas with high CO₂ uptake such as the US corn belt. In addition, it should be mentioned that both models ingest a multitude of CO₂ measurements at US ground-based stations and areas further away might be less well modeled. However, the excellent

Figure 4. Averaged HIPPO and matched model profiles for latitudes >70N during HIPPO 3 and 5, respectively. The left panels show model and HIPPO profiles and the right panels show model-HIPPO average differences as well as their range in the thinner and somewhat transparent colors.

are limited to retrievals over land. OCO-2 could largely improve on GOSAT’s data density over the oceans but didn’t overlap with the HIPPO measurement campaign period. The new Atmospheric Tomography Mission (ATom), selected as one of NASA’s Earth Venture airborne missions, will potentially allow for similar comparisons to OCO-2 in the future.

For the comparison of column-averaged mixing ratios, we need to extend the HIPPO profiles to the top-of-atmosphere. For this, we use the respective atmospheric model to compare with. In addition, we computed the average HIPPO XCO₂ for each campaign using all the data and subsequently removed it from individual measurements, both from the HIPPO, model, and satellite data. This ensures that observed correlations are driven pre-dominantly by spatial gradients within a campaign period and not by the secular trend. For satellite the HIPPO comparison against GOSAT data, we include take the instrument sensitivity into account by applying the averaging kernel to the measured profile (in other words, this conversion computes what the respective instruments should measure if HIPPO were the truth). Difference of the true profile (using the model-extended HIPPO dataset as truth) and the respective a priori profile. We perform this correction using both model extensions independently and then use the average of the two.
ings. For each match, the standard error in the GOSAT XCO₂ 
0.5 ppm, thereby bounding the error introduced by the spa-
tat the HIPPO and the actual GOSAT location is less than 
Wunch et al. (2011); Keppel-Aleks et al. (2011, 2012). In ad-
For coincidence criteria, we follow exactly Kulawik et al. 
a matching GOSAT measurement with each HIPPO profile. 
The comparison of GOSAT satellite data against HIPPO is 
4.2 GOSAT 
agreement provides a benchmark against which satellite re-
trees have to be measured. 

4.2 GOSAT 
The comparison of GOSAT satellite data against HIPPO is 
390 somewhat more complicated because there is not necessarily 
a matching GOSAT measurement with each HIPPO profile. 

For coincidence criteria, we follow exactly Kulawik et al. 
395 (2015), based on the dynamic co-location criteria detailed in 
Wunch et al. (2011); Keppel-Aleks et al. (2011, 2012). In ad-
390 dition, we require that the difference of CT2013B sampled 
at the HIPPO and the actual GOSAT location is less than 
0.5 ppm, thereby bounding the error introduced by the spa-
tal mismatch between HIPPO and respective GOSAT sound-
395 ings. For each match, the standard error in the GOSAT XCO₂ 

average is computed using the standard deviation of all cor-
395 responding GOSAT colocations divided by the square root of 
the number of colocations. 

For the GOSAT comparison, we require more than at least 
5 co-located GOSAT measurement per HIPPO profile, all of 
which are subsequently averaged before comparison against 
HIPPO. HIPPO XCO₂ is computed as the average of MACC 
and CT2013B extended HIPPO profiles with the difference 
between the two used as uncertainty range for HIPPO. 

In Figure 7, the scatterplot of HIPPO vs. GOSAT is de-
picted. It is obvious that the data density is far lower than 
for the models because a) HIPPO 1 is not overlapping in 
time and b) only a subset of HIPPO profiles is matched with 
enough co-located GOSAT soundings. This gives rise to a 
reduced dynamic range in XCO₂, from about -1.5 to 3 ppm 
difference to the campaign average. However, both slope and 
r² are also in excellent agreement with HIPPO and only very 
few points are exceeding 1 ppm difference. Those that are 
< -1 ppm are also associated with larger uncertainties in-
duced by model extrapolation, as seen in the larger error-bars 
for HIPPO in the left panel (esp. for HIPPO 2S). The right panel 
shows the discrepancies for the models as well, just for 
the subset that could be compared against GOSAT and using 
the model sampled at the GOSAT locations. 

One can see that it is hard to make a clear statement on 
whether GOSAT or the models compare better with HIPPO. 
Figure 8 shows this comparison in more detail, plotting 
model-HIPPO differences on the x-axis and GOSAT-model 
differences on the y-axis. As before, the error-bar for GOSAT 
is derived as the standard error in the mean and the model 
error-bar by using the variability of HIPPO XCO₂ using the 2 
different models to extrapolate to the top-of-atmosphere (and 
the average of the 2 is defined as HIPPO XCO₂. The center
box spans the range from -0.5–0.5 ppm, a strict requirement for systematic biases (GHG-CCI, 2014). The green and red shaded areas indicated regions where either the GOSAT data meets the 0.5 ppm requirement but the models not (green) or vice versa (red). Given the small amount of samples, it is premature to draw strong conclusions but it appears that somewhat more points lie in the green area. It also has to be pointed out that pure measurement unsystematic noise also contributes to the scatter in GOSAT.

Figure 8. Left: Scatterplot of $\Delta$ XCO$_2$ (CT-HIPPO) against $\Delta$ XCO$_2$ (GOSAT-HIPPO), using just the GOSAT subsets. Right: Same as left but using MACC instead of CT2013B. The inner box represent the area where both model and GOSAT are within 0.5 ppm compared to HIPPO, which corresponds to the very stringent accuracy requirement. The green and red shaded areas correspond to regions where the satellite deviates less than the models and is within 0.5 ppm (green) as well as where the models deviate less than GOSAT (red). The white cells on the outer edges indicate areas where both deviate more than 0.5 ppm overall.

For MACC, there is even a noticeable correlation between MACC-HIPPO and GOSAT-HIPPO with an $r^2$ of 0.26. This can hint at either small-scale features caught by HIPPO and missed by both GOSAT and models or small systematic variability between the exact HIPPO and MACC co-location. Most of the samples causing the high $r^2$ are located in the lower left quadrant, underestimated by GOSAT and both models and apparently all within HIPPO 2S, located between 40S and 20S.

Figure 9 depicts the HIPPO 2S campaign in more detail, showing the exact flight patterns and the differences with respect to MACC (MACC-HIPPO) at each measurement point (upper panel). For the sake of simplicity, we only show MACC here. The measured CO concentrations are shown in the lower panel. There is enhanced Carbon Monoxide (CO) at higher altitudes, indicating long-range transport of biomass burning at the time of overflight, which can explain the apparent model-HIPPO mismatch. The features span several degrees of latitude, excluding coarse model resolution as a reason for missing the plume. Thus, we hypothesize that the mismatch is caused by either underestimated CO emissions from the GFED (Randerson et al., 2013) emission database (which is used by both models) or transport errors in the models. For GOSAT, the mismatch is most likely caused by too lenient coincidence criteria, missing most of the biomass burning plume.

Overall, it can be concluded that GOSAT measurements can provide valuable and accurate information on the global CO$_2$ distribution and meets the 0.5 ppm bias criterion in most cases over the ocean. However, small sampling sizes precludes an in-depth analysis of potential large-scale biases in the datasets. In the future, OCO-2 with its much higher sampling density will help to disentangle measurement and modeling bias and guide inversion studies.

5 Comparisons of mid to upper tropospheric CO$_2$

5.1 TES (~510 hPa)

For the comparison with TES, we use the 510 hPa retrieval layer and apply averaging kernels accordingly. Kernel corrections using model-extended HIPPO data as truth, using both models independently and averaging results after averaging kernel correction. Coincidence criteria are identical to the GOSAT analysis but we require at least 20 valid TES soundings per HIPPO profile to reduce measurement noise. Similar to before, the TES error-bars are empirically
derived using the standard deviation of the co-located soundings itself.

Figure 10 shows the comparison of TES against HIPPO in the same way as done for GOSAT. The correlation ($r^2$) is somewhat lower than for GOSAT but still very significant. Some differences exceed 2 ppm, albeit with a relatively high standard error, i.e., barely significant at the 2-$\sigma$ level (see right panel, error-bars indicate 1-$\sigma$).

![Figure 11](image_url)

**Figure 11.** Left: Scatterplot of normalized (with campaign average) CO$_2$ from individual HIPPO profiles (x-axis) against corresponding TES data. Right: Difference plot of CO$_2$ against latitude. Campaigns as well as North and Southbound tracks are color-coded, model-HIPPO differences are plotted as well. Please refer to Fig. 7 for a detailed legend.

6 Conclusions

In this study, we compared atmospheric models as well as satellite data of CO$_2$ against HIPPO profiles. Table 1 provides a high level overview of the derived statistics. Both atmospheric models compare very similarly, both showing a very high correlation with respect to HIPPO, even with subtracting the campaign average XCO$_2$, as is done throughout all comparisons. Largest discrepancies are found near 300 hPa at higher latitudes during peak wintertime CO$_2$ accumulation as well as the summer uptake period. These may be related to steep vertical gradients poorly resolved by the models. In addition, a biomass burning event in the southern hemisphere seems to have been underestimated by the models, causing discrepancies of around 1 ppm.

For GOSAT comparisons, results are comparable to those with models but the sample size is much smaller. A comparison of GOSAT and model mismatches with respect to HIPPO indicates that GOSAT compares slightly better overall. In the future, OCO-2 with its much higher sampling density and expanded latitudinal coverage over the oceans could largely improve on GOSAT’s data density over the oceans but didn’t overlap with the HIPPO measurement campaign period. The new Atmospheric
C. Frankenberg: HIPPO model-satellite comparison

Table 1. Summary of all HIPPO comparisons. \( \#_{\text{profiles}} \) shows how many HIPPO profiles were used for the comparison. Correlation coefficients, fitted slope, mean difference \( \mu \) and standard deviation \( \sigma \) of differences in the difference compared to HIPPO of all comparisons are computed using measurements normalized by the respective campaign average. For comparison, \( \sigma \) of model-HIPPO for the satellite colocations and respective sensitivity are provided as well.

|       | \( \#_{\text{profiles}} \) | \( r^2 \) | slope \( \mu \) (ppb ppm) | \( \sigma \) (ppb ppm) | \( \sigma_{CT} \) | \( \sigma_{MACC} \) |
|-------|-----------------|--------|------------------|---------------------|-------------|----------------|
| GOSAT | 94              | 0.85   | 0.99             | -0.06               | 0.45        | 0.42 0.36     |
| TES   | 135             | 0.75   | 1.45             | 0.34                | 1.13        | 0.36 0.3      |
| AIRS  | 200             | 0.37   | 0.66             | 1.11                | 1.46        | 0.63 0.47     |
| CT2013B | 676        | 0.93   | 0.95             | 0.10                | 0.51        | N/A  N/A      |
| MACC  | 674             | 0.95   | 1.00             | 0.06                | 0.43        | N/A  N/A      |

Tomography Mission (ATom), selected as one of NASA’s Earth Venture airborne missions, will potentially allow for similar comparisons to OCO-2 in the future and should provide enough data to draw more robust conclusions that using GOSAT, for which the data density is fairly low, than using TES.

In general, GOSAT compares very well to HIPPO, followed by TES and AIRS. For TES, most deviations can be explained by pure measurement noise but AIRS appears to exhibit some latitudinal biases that need to be accounted for if used for source-inversion studies. On the other hand, systematic model transport errors that can affect source inversions (Deng et al., 2015) were confirmed here for both atmospheric models used. Despite initial scepticism towards using remotely sensed CO\(_2\) data for global carbon cycle inversion, we are now reaching a state where potential systematic errors in both remote sensing as well as atmospheric modeling can play an equally crucial part. Innovative methods to characterize and ideally minimize both of these error sources will be needed in the future. One option is to apply flux inversion schemes that co-retrieve systematic biases alongside fluxes, such as in Bergamaschi et al. (2007), using prior knowledge on potential physical insight into systematic biases, such as aerosol interference, land/ocean biases or air mass factors.

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