

Reply to comments, Referee #1

We would like to thank Referee #1 for his/her time, constructive and helpful comments and suggestions. Note, throughout the document R1 stands for reviewer 1, and figures named R1.x point to the x'th figure in "Reply to comments, Referee #1".

General comments:

1. Since the mean number of degrees of freedom of the NH₃ retrieval for both instruments seems to be near 1, comparison of the total NH₃-column amounts is a central part of the paper. However, some major retrieval characteristics should additionally be provided. Especially a kind of total column operator, like the one shown by Rodgers and Conner, 2003, Fig. 11. E.g. Fig 5 of the actual draft could be modified such that the AK for absolute concentrations or partial column amounts is provided.

We agree it would be a good idea to show total column averaging kernels, but with this version of the CrIS retrieval we did not save the required temperature and water vapour input profiles in the output retrieval files to compute accurate total column averaging kernels. In a future version of the retrieval output files we will compute and provide the total column averaging kernels to go along with the total column values. In the case of the FTIR retrievals, the partial column averaging kernel is provided in the datasets. Figure R1.1 shows the partial column averaging kernel for the example as shown in Figure 5.

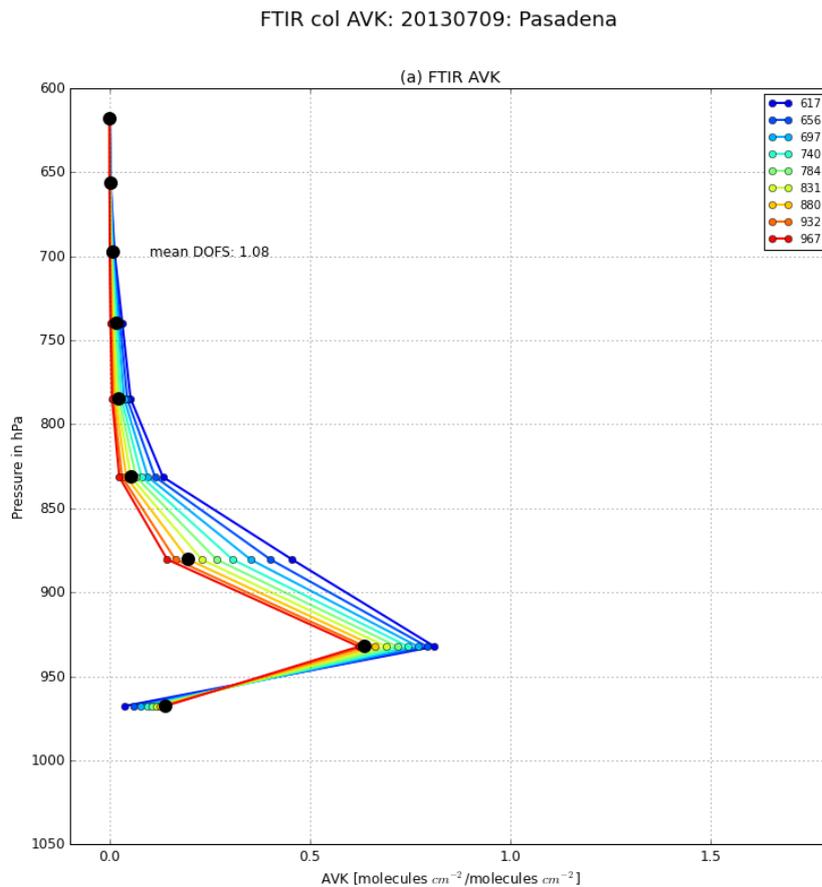


Fig R1.1. FTIR averaging kernel in [molecules cm⁻² / molecules cm⁻²]. The black dots show the matrices diagonal values.

2. Further, in Fig. 5 I wonder why the FTIR AK does not peak at the ground level: is there some problem with half-levels there? This should be explained in the paper.

The FTIR averaging kernel does not seem to always have complete sensitivity for the ammonia near the surface and varies from observation to observation. As mentioned in the text it usually peaks in between the surface and 850hPa. The method seems to be slightly more sensitive for the second layer in the retrieval. Furthermore, one should take into account that we only have a total DOF of 1 for most observations. Hence one cannot expect a perfect AVK peaking at its own level. We point the referee to the Figures R2.1-R2.4 in our reply to Referee #2.

3. Also, while error estimations of the profile retrievals are presented, it would be helpful to have those numbers for the total column amounts as well.

The estimated errors on the FTIR total column amounts are mentioned in section 2.2, being in the order of 30% for which we point you onward to Damers et al., (2015). In case of CrIS we do not mention a specific percentage in the text, but most total columns have an estimated error in the order of 10 %. This estimate however is on the low side as it does not yet include an estimate for the systematic errors in the retrieval.

Specific comments:

4. L27-35: The abstract should be made more concise. These lines, which include mainly motivation could be skipped.

We removed Line 27-35, and edited the abstract to be more concise. Furthermore we made the following edits:

Line 44: Added “(<1.0x10¹⁶ molecules cm⁻²)”

Line 45: Removed “and the FTIR total columns are smaller than 1.0x10¹⁶ molecules cm⁻²,”

Line 46: Removed “are small with CrIS showing”

Line 47: Added “show”

Line 47: Removed “around +2.4 x 10¹⁵ (standard deviation = ±5.5 x 10¹⁵) molecules cm⁻², which corresponds to a relative difference of ~+50% (std = ±100 %).”

Line 48: Added “The CrIS and FTIR profile comparisons differences are mostly within the range of the estimated retrieval uncertainties single level retrieved profile values showing average difference in the range of ~20 to 40%”

Line 50: Removed “for these comparisons”

Line 51: Added “into the boundary layer that typically peaks at”

Line 51: Removed “to”

Line 52: Added “(~1.5 km)”

Line 52: Removed “and”

Line 52: Removed “retrieved profiles also compare well with the”

Line 53: Added “is”

Line 53: Removed “of”

Line 53: Added “std =”

Line 53 Added “,”

Line 53 Removed “and a”

Line 53: Added “%”

Line 53: Added “std =”

Line 54: Removed “Most of the absolute and relative profile comparison differences are in the range of the estimated retrieval uncertainties. However, t”

Line 56: Added “At the surface, where CrIS typically has lower sensitivity,”

Line 55: Removed “he CrIS retrieval does”

Line 55: Added “it”

Line 56: Added “s” to “tends”

Line 56: Removed “the concentrations in the levels near the surface at”

Line 56: Added “under”

Line 56: Added “conditions, and underestimate under higher atmospheric concentration conditions.”

Line 58: Removed “, most probably due to the detection limit of the instrument, and at higher concentrations shows more of an underestimation of”

We also made a number of small edits to improve the readability of the main text:

Line 25: Edited the email address as the old one is no longer viable (change of institute)

Line 71: Added “,”

Line 97: Added “can”

Line 97: Removed “and”

Line 107: Added “,”

Line 110: Added “,”

Line 192: Changed pseudo-lines to Cross-sections

Line 442: Removed “which”

Line 443: Added “,”

Line 445: Added “,”

Line 462: Removed “.”

Line 463: Removed “Because of”

Line 463: Added “Due to”

Line 608: Removed “and Jacob Siemons (ECCC)”

5. L39, L48-49: ‘compare well’ These are qualitative terms which do not contain much (if any) information content. Please try to avoid those throughout the manuscript and concentrate on quantitative assessments.

Removed qualitative terms throughout the document.

Line 30 (all line statements are the positions within the new document): changed “compare well with” to “have a positive”.

Line 301: changed “The overall agreement is good” to “There is an overall agreement”

Line 391: changed “good” to “high”

Line 400: “removed well”

Line 440: changed “agree quite well” to “show agreement”

Line 473: removed “good”

Line 523: changed ‘agree well with’ to “ have”

Line 526: changed “agree very well” to “are in agreement”

6. LL95-96: ‘However, the uncertainty of the satellite observations is still high due to a lack of validation.’ The reasoning is a bit strange: the uncertainty is not caused by lack of validation but rather the knowledge of the uncertainty.

Edited the sentence to “However, the overall quality of the satellite observations is still highly uncertain due to a lack of validation.”

7. LL245-246: ‘Do note that on average the observations have a DOFS between 0.9 and 1.1.’ Could you please provide a Figure or numbers of the DOF distribution of all measurements entering the comparison.

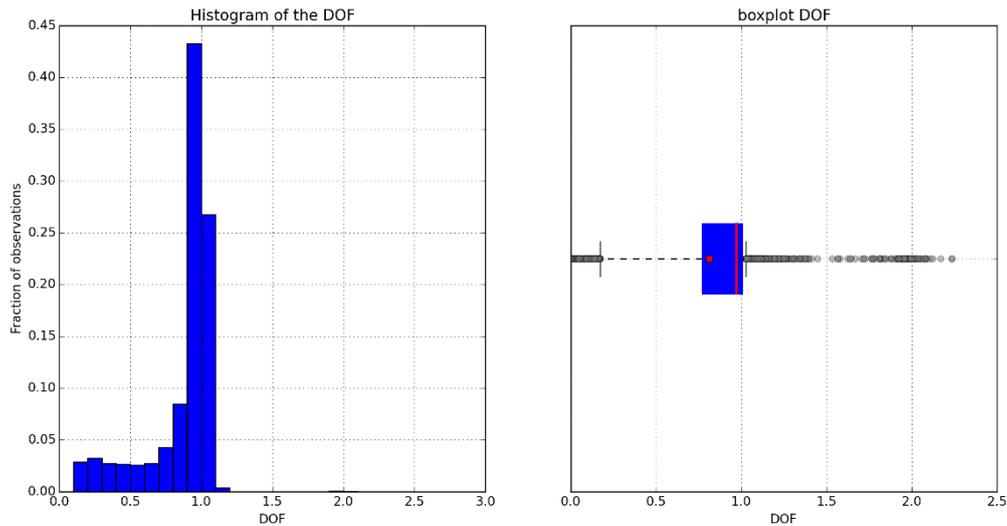


Fig R1.2 Distribution plots of the DOF of all CrIS observations used in this study. The left panel shows the fraction of all observations for each specific DOF range. The right panel shows a boxplot of the same distribution. The box edges are the 25th and 75th percentiles, the red line in the box is the median, the red square is the mean, the whiskers are the 10th and 90th percentiles, and the grey circles are the outlier values outside the whiskers.

Figure R1.2 shows the distribution of the DOF of all measurements. Note that the <0.1 DOF are already removed from this set. $\sim 80\%$ of the observations have a DOF in between 0.9 and 1.1, with a median of almost 1.0.

8. L246: ‘clouds will implicitly be accounted for by the quality control’: What is the effect of a partially cloudy field-of-view?

That’s a good point. Currently, there is a cloud filter in development to exclude clouded scenes in the future. In our case we remove all observations with a DOF of <0.1 which removes most of the clouded scenes (e.g. thick clouds \rightarrow no ammonia observed). Besides a reduction in DOF we do not expect further major impacts as mentioned in the TES-NH₃ retrieval paper (Shephard et al., 2011). As an example to illustrate potential effects (or the lack there of) we give Fig R1.3. The figure shows a MODIS scene for northern Canada with both visible clouds and fire plumes. The bottom panel shows the calculated CrIS surface NH₃ for the same period. As one can observe there are a number of hotspots for NH₃ found for the observations surrounding the fires and of the plumes. The optically thick clouds are filtered out by our artificial cut off. The remaining retrieved concentrations for observations with partially covered and by optically thin clouds do not show any strange patterns or alternating high and low retrieved concentrations.

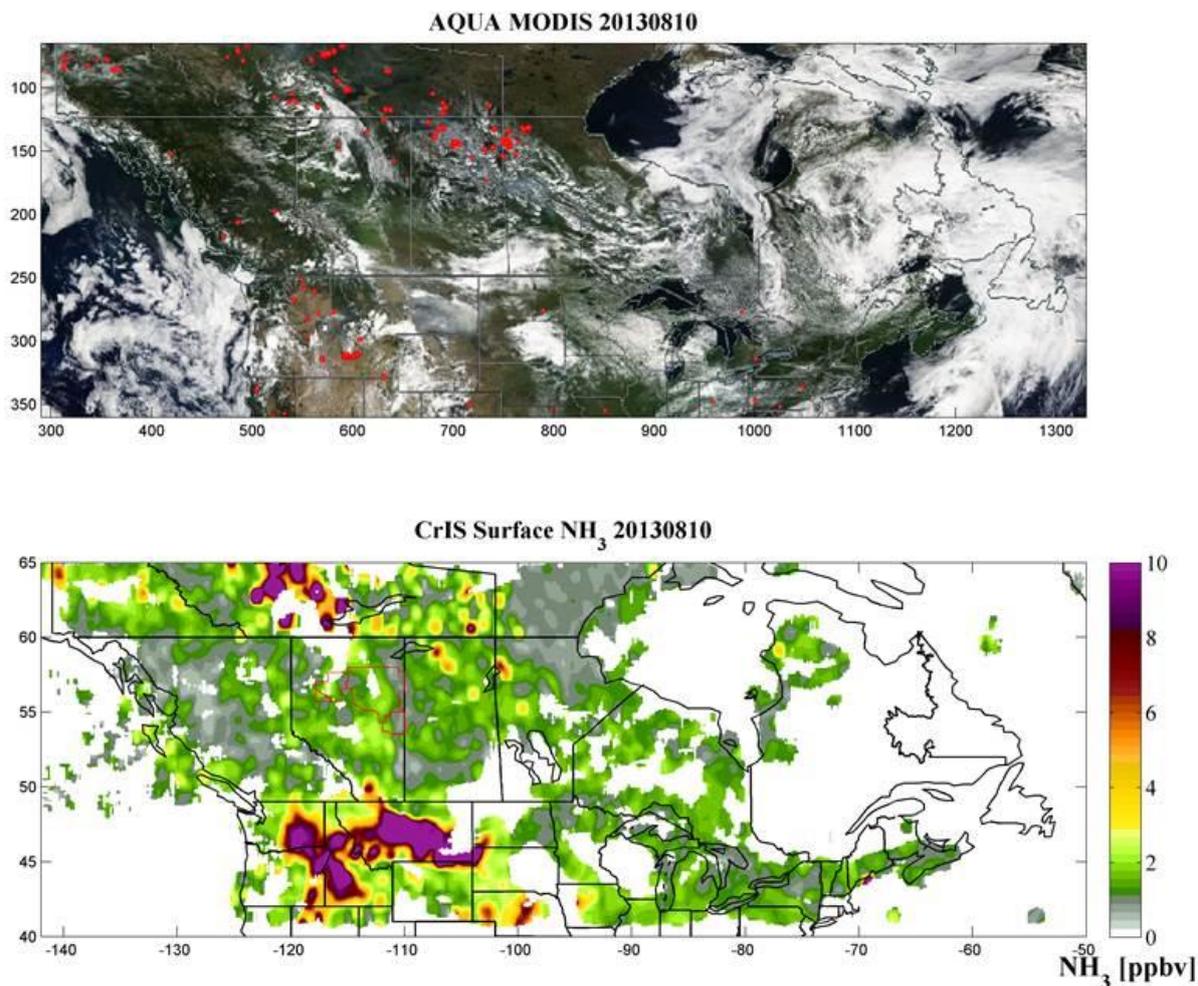


Fig R1.3. Top panel shows the MODIS image over Canada on the 10th of August 2013. Bottom panel shows the retrieved CrIS surface concentrations for the same day.

9. L301: ‘total column comparison’: Has this comparison been performed with or without the application of the FTIR-AK as described in chapter 2.5? Since the FTIR is generally better suited for total column retrievals due to its better sensitivity nearer at the ground (where most of the NH₃ is present), I doubt that the transformation like in Eq(1), L278 is helpful. Here the better instrument (FTIR) should be transformed to the worse (CrIS) to compare with the CrIS total column amounts. I.e. there should be Figures like Figs. 2 and 3 with the raw data and after the transformation as just described.

The total column comparison **has been** performed with the application of the FTIR-AK as described in chapter 2.5. In principle we agree that the better instrument should be transformed to the worse (CrIS). However, we wanted to keep the study comparable to the IASI validation study, and thus apply the AVK in the same manner as done in that study. The IASI product does not produce an averaging kernel and thus we cannot apply the satellite observational operator in both cases. Furthermore, to meet readers who would rather see it the other way around, we added alternative figures transforming the FTIR profiles with the CrIS AVK, which are shown in figures A5, A6 and A7.

10. L318: ‘In Toronto, Bremen and Pasadena there is good agreement’ In case of Pasadena, I would not call the agreement good. Please also avoid this qualitative terms.

Changed the qualitative terms as mentioned in edit number 5.. We also added some lines on the results at Pasadena and Wollongong, see edit number 13 for the full description.

Removed, “Pasadena”,

11. L319: ‘and low bias in the CrIS total columns for intermediate values’ This seems not to be the case for Bremen.

The only outlier for Bremen is the value that is marked as an outlier by the three sigma filter as used in Figure 2. Furthermore the number of observations is too small for any good statistics.

Added “except for the outlying observation in Bremen, which is marked as an outlier by our three sigma filter used for Figure 2.”

12. L322, Fig. 3: Could you also discuss in the text what the reason for the apparently systematic deviations at Wollongong may be.

There are a number of reasons why the Wollongong bias might look higher than the others. The first is the date of observation. The two comparisons with the highest CrIS to FTIR ratio were both made during the end of November in 2012 when there were multiple fires occurring in the surrounding region (GFED4.1s). Possibly the CrIS footprint covers the plumes from the fire, which was not observed by the FTIR due to an (for us) unfavourable wind direction. The remaining comparisons on average show a MD of $\sim 5 \times 10^{15}$, which is similar to our station wide result. Another explanation might be the difference in observed air masses which can be larger for coastal sites (e.g. Wollongong, and essentially Toronto). Depending on the wind direction there is either clean air coming in from above the lake/ocean which will mean there is a reduction in FTIR observed NH₃ while the satellite potentially observes above land. Vice versa observations from the satellite above the ocean/lake can be far lower than the columns observed by the FTIR with a wind direction coming from an inland direction. This heterogeneity is also visible for sites with larger gradients in orography, such as Pasadena and Mexico City.

Line 311: Added “Similarly to Mexico City the comparison also shows an increase in scatter for Pasadena, where the FTIR site is also located on a hill.”

Line 315: Added “In Wollongong, there is less agreement between the instruments. There are two comparisons with large CrIS to FTIR ratios while most of the other comparisons also show a bias for CrIS. For both cases the bias can be explained by the heterogeneity of the ammonia concentrations in the surrounding regions. The two outlying observations were made during the end of November, 2012, which coincides with wild fires in the surrounding region. Furthermore the Wollongong site is located coastally, which will increase the occurrences where one instrument observes clean air from the ocean while the other observes inland air masses.”

13. L331, Fig. 4: ‘show the standard deviation for each value’ Is this the standard deviation of the distribution of the differences or the standard error of the mean difference (i.e. the former divided by sqrt(number of values))? The latter should (also) be shown to detect any significant measurement bias.

Fig 4 showed the standard deviation of the distribution of the differences. As noted in the reply to Referee number 2 we edited the figure to show the 95% confidence interval i.e. $\sim 2x$ standard error.

14. LL379-380: ‘along with the shorter atmospheric path lengths for observations from the ground-based solar-pointing FTIR’ Could you explain, why the FTIR path length is shorter compared to the satellite? Is this always the case?

In principle the atmospheric path length should be more or less similar. The path length of both instruments vary per location of the site, time of day and field of view of the satellite, but the difference should be more or less near zero.

Line 379: Removed sentence

15. LL377-385: As already mentioned, for this discussion the total column operator or the partial column(number density) AK would be interesting. As the FTIR is more sensitive down to the ground level than the satellite where there are highest concentrations of NH₃, the satellite retrieval should be determined by the a-priori there. So the higher column amounts may be produced by higher a-priori values at the ground.

That's a good point but for the fact that the application of the observational operator should reduce the effects of the difference in sensitivity and a-priori choice. Any remaining effect of the a-priori is hard to judge without a repeat of the retrieval. What potentially can be done to further reduce the influence of the a-priori is switching out the a-priori. For a number of examples of the a-priori switch, we point you to figures R2.1 to R2.4 in our reply to Referee number 2.

16. L550: 'improvements to the NH₃ line spectroscopy to reduce the uncertainty coming from this error source' Could you give the information if the CrIS retrieval also uses Hitran2012, like the FTIR?

The CrIS retrieval also uses HITRAN 2012.

Added:

"and uses the HITRAN database (Rothman et al., 2014) for its spectral lines"

Technical comments:

L566, Fig. A1: the arrangement of the panels in the figure is transposed with respect to the description in the caption.

Good catch,

L566 Fig. A1 caption, changed to: "Error profiles for each of the error terms. The left panels show the random errors, the right panels the systematic errors. The top two panels show the error in VMR. The bottom panels show the errors in partial column layers [molecules cm⁻²]. (See Figure A.2 for the same figure but with the errors relative to the final VMR and partial columns per layer)"

L571, Fig. A2: same problem as in Fig. A1. Moreover, the top and the bottom row seem to show identical data.

It is correct that the top and bottom row are showing identical data as the error is initially derived for the VMR value and subsequently applied to the partial columns.

L571 Fig. A2 caption, changed to: "Relative error profiles for each of the error terms. The left panels show the Random errors, right panels the Systematic errors. All four panels show the error in a fraction of the original unit used in Figure A1. (See Figure A.1 for the same figure but with the absolute errors)"

References.

Shephard, M. W., Cady-Pereira, K. E., Luo, M., Henze, D. K., Pinder, R. W., Walker, J. T., Rinsland, C. P., Bash, J. O., Zhu, L., Payne, V. H., and Clarisse, L.: TES ammonia retrieval strategy and global observations of the spatial and seasonal variability of ammonia, *Atmos. Chem. Phys.*, 11, 10743-10763, doi:10.5194/acp-11-10743-2011, 2011.

van der Werf, G. R., Randerson, J. T., Giglio, L., van Leeuwen, T. T., Chen, Y., Rogers, B. M., Mu, M., van Marle, M. J. E., Morton, D. C., Collatz, G. J., Yokelson, R. J., and Kasibhatla, P. S.: Global fire emissions estimates during 1997–2015, *Earth System Science Data Discussions*, doi:10.5194/essd-2016-62, in review, 2017.

Reply to comments, Referee #2

We would like to thank Referee #2 for his/her time, constructive and helpful comments and suggestions. Throughout the document (R1 stands for reviewer 1, and figures named R1.x point to the x'th figure in "Reply to comments, Referee #1)

General Comments:

- 1. One particular issue that could hamper the interpretation of the results is the potentially limited information content captured by the CrIS retrievals. The current DOFS cut off is taken at >0.1, which entails that some measurements are/could be heavily dominated by the a-priori. The authors allude that particularly measurements with low NH₃ concentrations could be effected. One way to at least give some information on this is to replot Fig2, whereby each measurement is coloured related to its (average) DOFS.**

The **artificial DOFS cut off** was chosen to remove observations without information. We added colouring to the scatterplot indicating the average DOFS for each CrIS observation, with the colour bar ranging from 0.1 to more than 1.1. See the reply to the comments of ref #1 for a distribution of the DOFS for all observations used in this study.

Added "The colouring on the scatter indicates the mean DOFS of each the CrIS coincident data." to the caption of Figure 2.

- 2. Another way to test whether observed differences between CrIS and FTIR are driven by differences in a-priori rather than the actual retrieval, is to (prior to mapping CrIS to FTIR –see eq(1)) conform the CrIS retrieval to the FTIR a-priori as in Rodgers (2000): $x(\text{CrIS}, \text{ftir apriori corrected}) = x(\text{CrIS}) + [A(\text{CrIS}) - I] * [\text{apriori}(\text{CrIS}) - \text{apriori}(\text{ftir})]$**

That is a great question and it is something we have tried before submitting. A posteriori switching out the CrIS a-priori for the FTIR a-priori essentially brings us closer to what we want, validating just the observations without an effect from the a-priori. The problem however is the non-linearity in the retrievals (Kulawik et al., 2008). The resulting retrieved profile is not always near/comparable to the initial a-priori shape and amplitude (which in itself shows that the initial choice of a-priori does not greatly influence the retrieval), which makes an a posteriori switch of the a-priori profile troublesome. Without a repeat of the CrIS retrievals it is not possible to distinguish between the effect of the a priori on the retrieval and the effect of the a posteriori switch. The study by Kulawik et al., (2008) showed that the effect does not have to be major as long as the a priori is representative of the final retrieved profile. This however is not always the case in our retrievals. To illustrate we give a number of examples. Fig R2.1 to R2.4 show the effect for a range of atmospheric ammonia concentrations at a variety of sites, starting with situations with medium to large concentrations for Pasadena (Fig R2.1, which is Figure 5 in the main manuscript), Bremen (Fig R2.2) and Toronto (Fig R2.3). Furthermore we added a figure showing the situation when there is not much ammonia i.e. Wollongong (Fig R2.4). In the case of Pasadena the difference is only small with a few percent change in the concentrations of the individual layers. For Bremen the difference is larger near the surface, corresponding to the higher concentration levels. Especially near the surface the concentration change is high following the abrupt difference in the a priori shape, which does not have the sharp peak like the CrIS a priori. In the case of Toronto the effect is in the order of 10%, although concentrations around 750hPa, turn negative. In the Wollongong example the relative difference is large, as the retrieved concentrations are low. The large difference shows that a change of 1-2 ppb in the a priori shape and amplitude cannot be seen as a small enough difference to permit an a posteriori change.

FTIR:CrIS retrieved profiles: 20130709: Pasadena

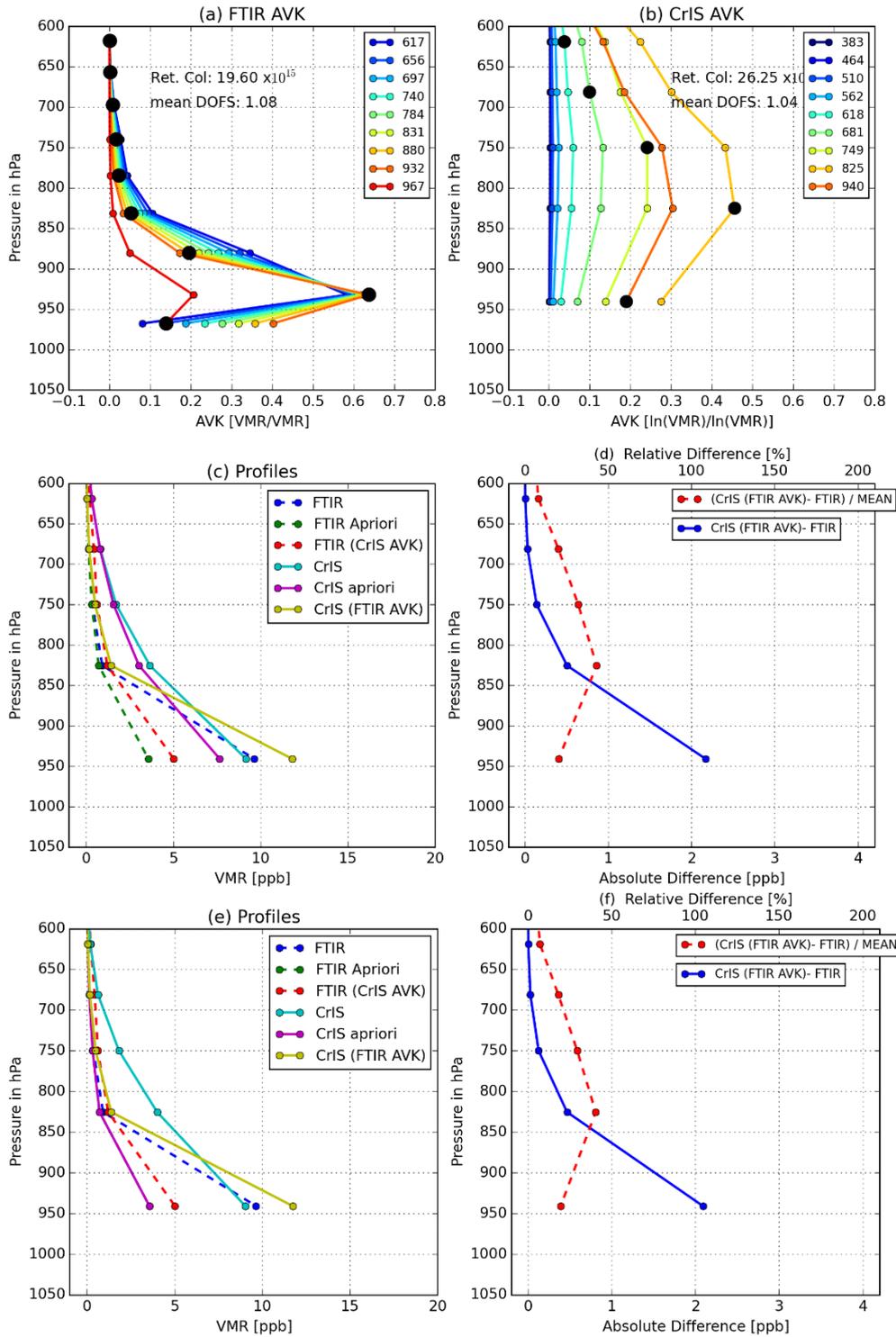


Fig R2.1. Example of the effect of switching out the CrIS a priori for the FTIR a priori to the CrIS Retrieved profile, for an FTIR profile matched with a CrIS profile measured around the **Pasadena** site. For the full figure caption see Figure 5 in the main manuscript. Panel (e) shows is a copy of panel (c) but now with the CrIS a priori switched out for the FTIR a priori. Similarly (f) shows the same as panel (d) after switching out the CrIS a priori for the FTIR a priori.

FTIR:CrIS retrieved profiles: 20130802: Bremen

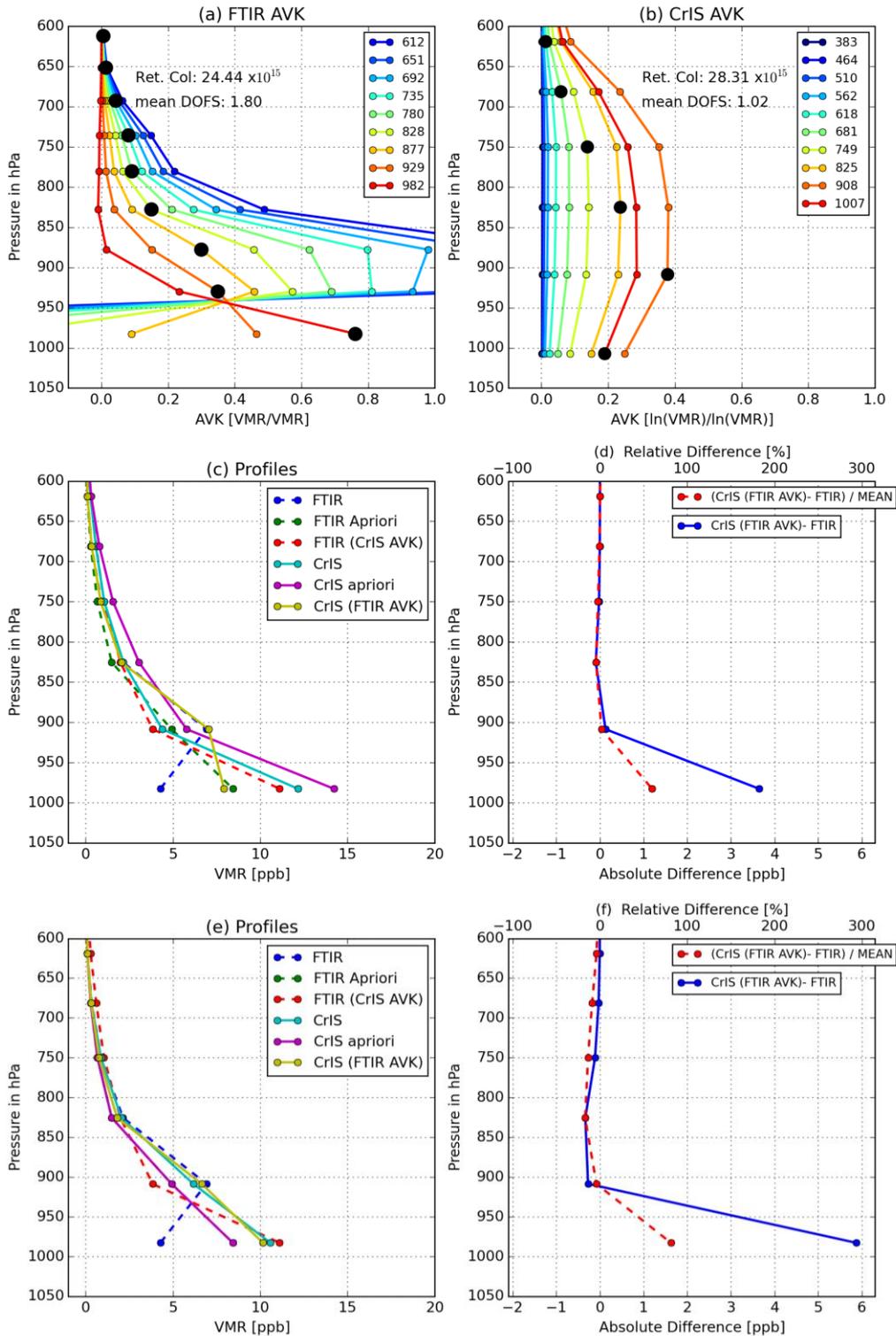


Fig R2.2. Example of the effect of switching out the CrIS a priori for the FTIR a priori to the CrIS Retrieved profile, for an FTIR profile matched with a CrIS profile measured around the **Bremen** site. For the full figure caption see Figure 5 in the main manuscript. Panel (e) shows is a copy of panel (c) but now with the CrIS a priori switched out for the FTIR a priori. Similarly (f) shows the same as panel (d) after switching out the CrIS a priori for the FTIR a priori.

FTIR:CrIS retrieved profiles: 20120515: Toronto

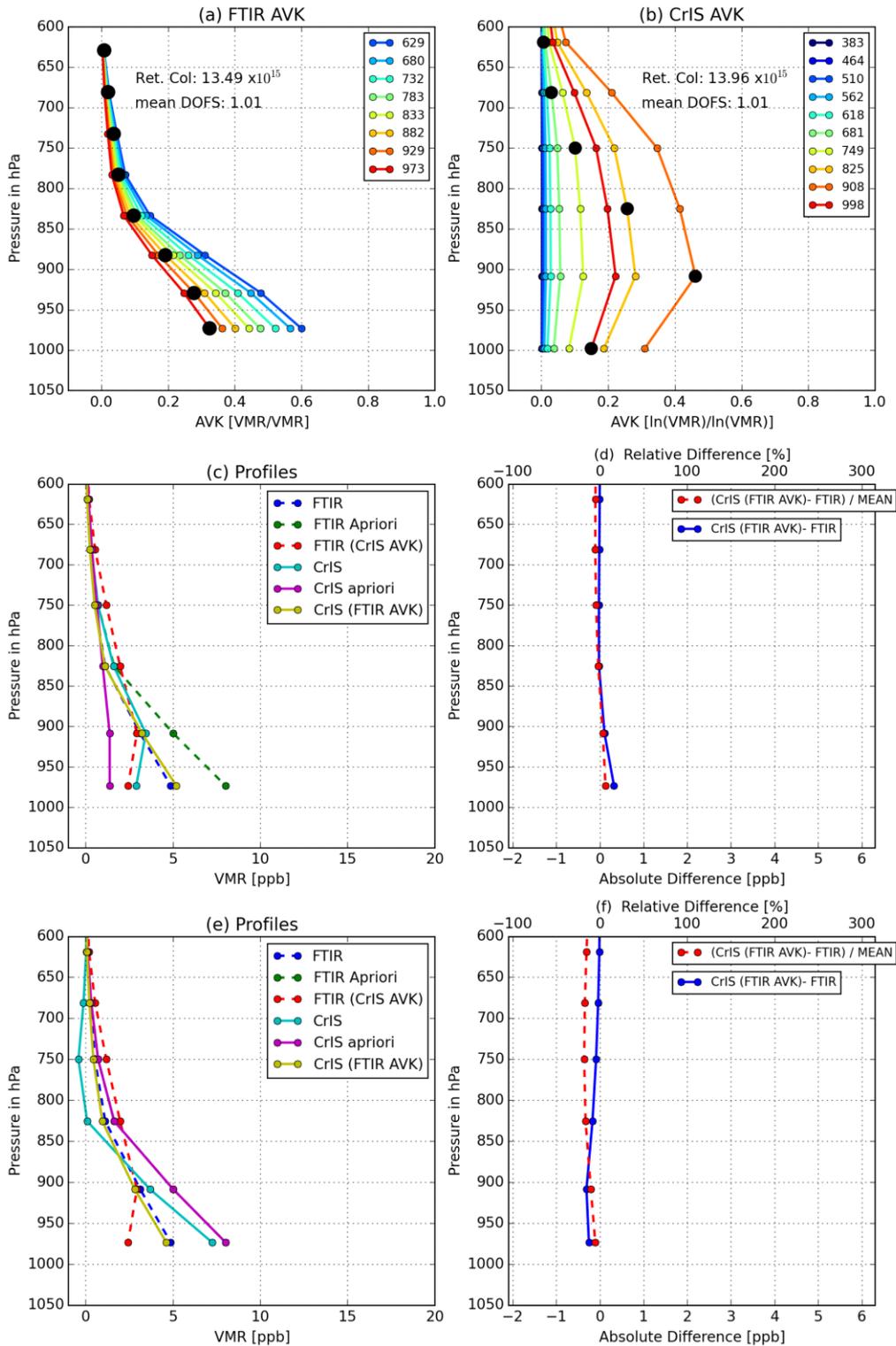


Fig R2.3. Example of the effect of switching out the CrIS a priori for the FTIR a priori to the CrIS Retrieved profile, for an FTIR profile matched with a CrIS profile measured around the **Toronto** site. For the full figure caption see Figure 5 in the main manuscript. Panel (e) shows is a copy of panel (c) but now with the CrIS a priori switched out for the FTIR a priori. Similarly (f) shows the same as panel (d) after switching out the CrIS a priori for the FTIR a priori.

FTIR:CrIS retrieved profiles: 20120511: Wollongong

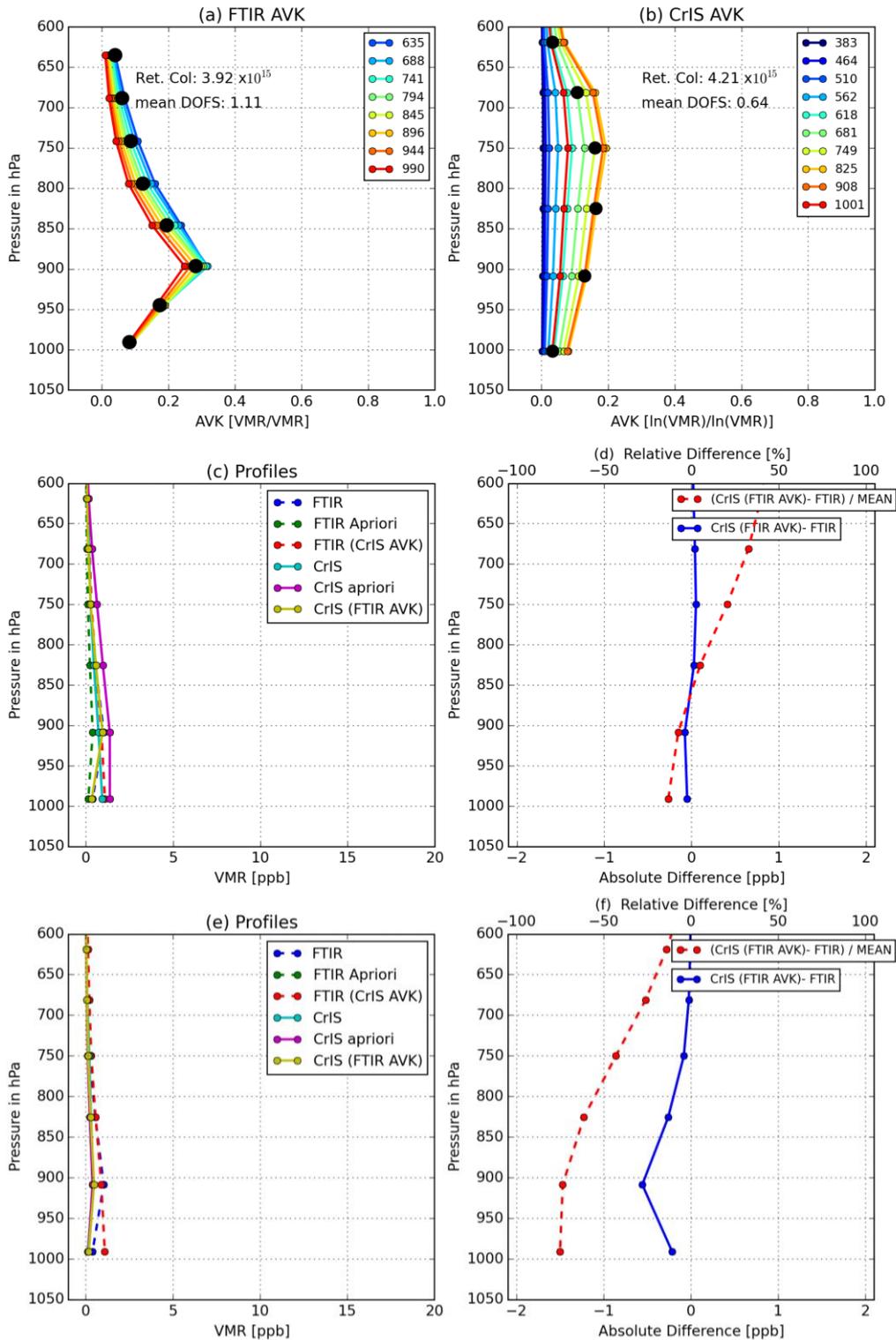


Fig R2.4. Example of the effect of switching out the CrIS a priori for the FTIR a priori to the CrIS Retrieved profile, for an FTIR profile matched with a CrIS profile measured around the **Wollongong** site. For the full figure caption see Figure 5 in the main manuscript. Panel (e) shows is a copy of panel (c) but now with the CrIS a priori switched out for the FTIR a priori. Similarly (f) shows the same as panel (d) after switching out the CrIS a priori for the FTIR a priori.

3. In any case, the authors need to look deeper into the possible effects of the DOFS on the bias.

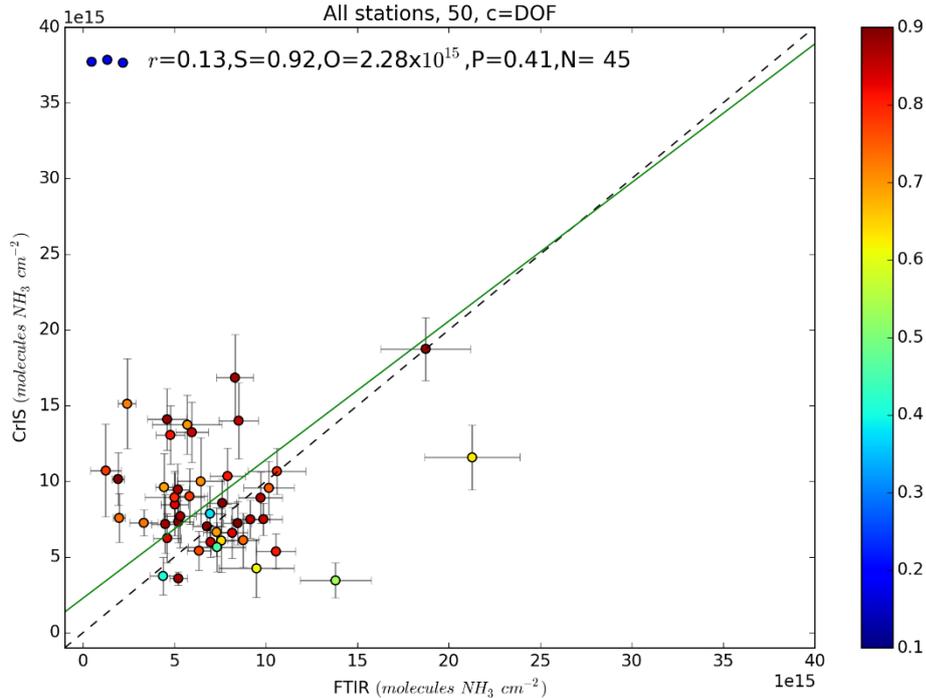


Fig R2.5. Correlation between the FTIR and CrIS total columns using the coincident data from all measurement sites for all CrIS observations with a mean DOF < 0.9. The horizontal and vertical bars show the total error on each FTIR and CrIS observation. The colouring on the scatter indicates the mean DOF of each the CrIS coincident data. The trend line shows the results of the regression analysis.

We looked into the effects of the DOF, but decided to not put any further emphasis on it in the main manuscript. Essentially most (>80%) of the observations have a DOF between 0.9 – 1.1. From the remaining 20 % the larger number are above >0.7 leaving a small set of observations (~10%) with a DOF <0.7. Figure R2.5 shows a scatterplot similar to Figure 2 in the main manuscript but now with only observations with a DOF <0.9. As one can see there is no clear relation visible between the amount of scatter and the DOF. About 20% of the observations (N=45 in total) used for Figure 2 have a DOF < 0.9. Table R1 shows the mean difference and mean relative difference for the observations with a DOF<0.9. For comparability we added the full set of observations to the table, coloured in red. The observations with small DOFs usually are observations with a relatively low ammonia concentration. This makes that we can only really compare the lower range of total columns to our earlier results as there are only 6 observations with a total column > 10 x 10¹⁵ molecules cm⁻². The observations smaller than > 10 x 10¹⁵ molecules cm⁻² have a MD of 2.6 with a std of 4.1 which is comparable to the complete set with 3.3 (std = 4.1). Similarly the MRD is 33.0 % (std = 54.9%) which is also in the same range as the original set's 30.2% (std = 38.0%).

Table R1. Results of the total column comparisons of the FTIR to CrIS for observations with a $\text{DOF} < 0.9$. N is the number of averaged total columns, MD is the mean difference [10^{15} molecules cm^{-2}], MRD is the mean relative difference [frac, in %]. Value for the complete set used in Table 3 and Fig 4 are given in red. Take note that the combined value N does not add up with all the separate sites as observations have been included for FTIR total columns $> 5 \times 10^{15}$ molecules cm^{-2} .

Retrieval	Column range total in molecules cm^{-2}	N	MD in 10^{15} (1σ)	MRD in % (1σ)	FTIR mean in 10^{15} (1σ)
CrIS-NH3	$< 10.0 \times 10^{15}$	39	2.6 (4.1)	33.0 (54.9)	6.1 (2.2)
CrIS-NH3	$\geq 10.0 \times 10^{15}$	6	-4.3 (4.4)	-41.6 (44.3)	14.2 (4.3)
CrIS-NH3	$< 10.0 \times 10^{15}$	93	3.3 (4.1)	30.2 (38.0)	7.5 (1.5)
CrIS-NH3	$\geq 10.0 \times 10^{15}$	109	0.4 (5.3)	-1.39 (34.4)	16.7 (8.5)

A few outlying values are observed in Fig R2.5 which show more of a dependency to location than to DOF , as illustrated in Fig R2.6. All of the larger retrieved total columns (both FTIR and CrIS, $> 10 \times 10^{15}$ molecules cm^{-2}) are observations from the Toronto measurement site. As noted in the main manuscript the Toronto results are influenced by the local conditions, which increase the heterogeneity in the region. The site is located within the city, further away from the main sources surrounding the city. Furthermore Toronto is located at the edge of Lake Ontario, which increases the differences as for days with wind originating from the south one can expect clean air observed by the FTIR, where the satellite observes the emitted ammonia of the sources outside the city. Similarly for conditions with wind from the north one can expect

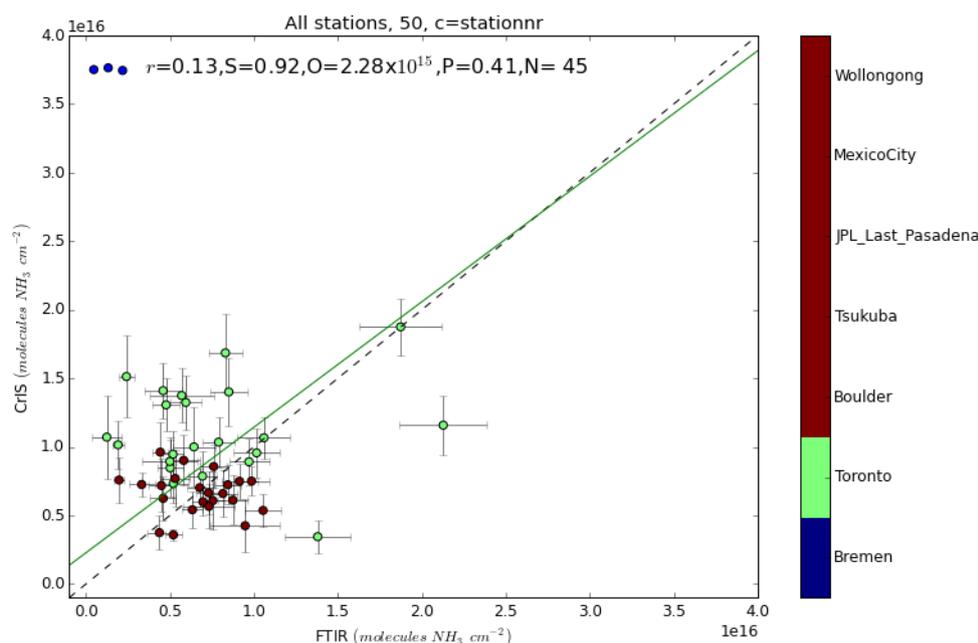


Figure R2.6. Correlation between the FTIR and CrIS total columns using the coincident data from all measurement sites for all CrIS observations with a mean $\text{DOF} < 0.9$. The horizontal and vertical bars show the total error on each FTIR and CrIS observation. The colouring on the scatter indicates the FTIR measurement site. The trend line shows the results of the regression analysis.

4. A second general comment is the error analysis which could be improved. The document either misses a general statement that all presented errors correspond with the 1-sigma standard deviation or it sometimes needs to be more specific when it uses the term ‘error’ as it sometimes relates to the standard deviation on the bias and sometimes on the bias itself. That said, 1-sigma standard deviations often tell little with regards to the statistical significance of an observed difference. For instance one claim made by the authors is that in the 0.5-1.0 $e16$ bin CrIS is significantly higher. This is likely to be true but from the article alone I cannot verify this. In Figure 4 the observed binned biases are shown with their standard deviations. A much better metric to

show statistical significance would be the 95 or 99% confidence interval on the mean. This goes for all metrics where statistical significance is claimed or investigated.

Some parts are indeed confusing. We cleaned up the text and added a few sentences to clarify what error we are talking about. We define two types of errors, 1. Estimated errors: FTIR & CrIS retrieval & prior knowledge and 2. the actual errors i.e. the mean and mean relative differences that come out of the comparison. In the case of Figure 4, we edited the figure to include the 95 % confidence level as the number of observations were not included. Throughout the text we left it at standard deviation. Optionally one could calculate it directly from the standard deviation and the number of observations.

Line 296: edited caption, “total estimated error”

Line 305: removed “significantly”

Line 363, 365: added , N = 229)

Line 365, 364: added “std =” in between brackets”

Line 403: added “estimated” to “estimated error”

Line 460: changed “bias” to “actual error”

Line 538: added “estimated”

Line 527, 529, 530, 531: added “std =” in between brackets

Line 582: changed “bias” to “actual error”

Caption Fig 4. Added “The number of observations in each set is shown in the bottom panel.”

Specific comments:

5. L155: A list of the dominant interfering species would be useful here

L155: Changed “i.e. interfering species” to “(i.e. major interfering species such as H₂O, CO₂, and O₃)”

6. L165: A representation of the used collocation area would be useful in this figure

Adjusted Figure 1 to include 2 circles with radii of 25 and 50 km.

Added a second sentence to the caption of figure 1: “The two circles show the collocation area when for radii of 25 and 50 km.”

7. L305: show the error (be more specific= standard deviation is better)

L305: Added “total” to indicate it is not a standard deviation of the observations. The bars indicate the mean total error of the combined observations.

8. L319: Pasadena looks worse at elevated values

L319: Removed Pasadena.

9. L320: in, and low bias (“in” is obsolete)

L320: removed “in”

10. L465: red diamond -> red square

Changed “diamond” to “square in all bar plot figures.

11. L468: Summary of the errors. . . Could be interpreted as the uncertainty on the biases, not the actual absolute and relative bias

Caption figure 7: Changed “summary of the errors” to “Summary of the absolute and relative bias”.

L580: Caption Fig. A5. Similarly changed to “Summary of the absolute and relative bias”

Author comment

We made a number of small edits to the text to improve the readability of the manuscript. The changes made are the following;

Added a figure to the appendix to improve comparability of the results of the IASI and CrIS retrievals.

Line 366: added “To put the results of this study into perspective of the IASI-LUT and IASI-NN products we added Figure A1 to the Appendix, which shows the total column comparison for both products.”

Inserted caption:

“**Figure A1.** Correlation between the FTIR and the IASI-LUT (left, blue) and IASI-NN (right, red) total columns using the coincident data from all measurement sites. The horizontal and vertical bars show the total estimated error on each FTIR and CrIS observation. A three sigma outlier filter was applied to the IASI-LUT dataset and the same observations were removed from the IASI-NN set. Contrary to the earlier study by Dammers et al., (2016a) no thermal contrast filter was applied to the dataset.”

We changed the numbering of the other appendix figures to match the new order.

Shortened and slightly edited the abstract for readability

Line 44: Added “($<1.0 \times 10^{16}$ molecules cm^{-2})”

Line 45: Removed “and the FTIR total columns are smaller than 1.0×10^{16} molecules cm^{-2} ,”

Line 46: Removed “are small with CrIS showing”

Line 47: Added “show”

Line 47: Removed “around $+2.4 \times 10^{15}$ (standard deviation = $\pm 5.5 \times 10^{15}$) molecules cm^{-2} , which corresponds to a relative difference of $\sim +50\%$ (std = $\pm 100\%$).”

Line 48: Added “The CrIS and FTIR profile comparisons differences are mostly within the range of the estimated retrieval uncertainties single level retrieved profile values showing average difference in the range of ~ 20 to 40% ”

Line 50: Removed “for these comparisons”

Line 51: Added “into the boundary layer that typically peaks at”

Line 51: Removed “to”

Line 52: Added “(~ 1.5 km)”

Line 52: Removed “and”

Line 52: Removed “retrieved profiles also compare well with the”

Line 53: Added “is”

Line 53: Removed “of”

Line 53: Added “std =”

Line 53: Added “,”

Line 53: Removed “and a”

Line 53: Added “%”

Line 53: Added “std =”

Line 54: Removed “Most of the absolute and relative profile comparison differences are in the range of the estimated retrieval uncertainties. However, t”

Line 56: Added “At the surface, where CrIS typically has lower sensitivity,”

Line 55: Removed “he CrIS retrieval does”

Line 55: Added “it”

Line 56: Added “s” to “tends”

Line 56: Removed “the concentrations in the levels near the surface at”

Line 56: Added “under”

Line 56: Added “conditions, and underestimate under higher atmospheric concentration conditions.”

Line 58: Removed “, most probably due to the detection limit of the instrument, and at higher concentrations shows more of an underestimation of”

Further small edits readability in main text

Line 25: Edited the email address as the old one is no longer viable (change of institute)
Line 71: Added “,”
Line 97: Added “can”
Line 97: Removed “and”
Line 107: Added “,”
Line 110: Added “,”
Line 192: Changed pseudo-lines to Cross-sections
Line 442: Removed “which”
Line 443: Added “,”
Line 445: Added “,”
Line 462: Removed “.”
Line 463: Removed “Because of”
Line 463: Added “Due to”
Line 608: Removed “and Jacob Siemons (ECCC)”

Validation of the CrIS Fast Physical NH₃ Retrieval with ground-based FTIR

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Abstract

Global reactive nitrogen emissions into the air have increased to unprecedented levels. Limiting the loss of reactive nitrogen into the environment is one of the major challenges for humankind. At the current levels ammonia (NH₃) is a threat to both the environment and human health. However, relatively little is known about the total nitrogen budget and distribution around the world, due in part to the sparseness of observations over most of the globe. Recent advances in the capabilities of measuring NH₃ with satellite instruments have improved the situation with sensors such as the Infrared Atmospheric Sounding Interferometer (IASI) and the Cross Track Infrared Sounder (CrIS) making twice daily observations with global coverage. However, these require validation to be truly useful, and one of the main challenges in the validation of the satellite NH₃ profile and total column data products is the scarcity of measurements that can be directly compared. Presented here is the validation of the CrIS Fast Physical Retrieval (CFPR) NH₃ column and profile measurements using ground-based Fourier Transform Infrared (FTIR) observations. We use the total columns and profiles from seven FTIR sites in the Network for the Detection of Atmospheric Composition Change (NDACC) to validate the satellite data products. The overall FTIR and CrIS total columns compare well with have a positive correlation of $r = 0.77$ ($N=218$) with very little bias (a slope of 1.02). Binning the comparisons by total column amounts, for concentrations larger than 1.0×10^{16} molecules cm⁻², i.e. ranging from moderate to polluted conditions, the relative difference is on average $\sim 0 - 5\%$ with a standard deviation of 25-50%, which is comparable to the estimated retrieval uncertainties in both CrIS and the FTIR. For

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43 the smallest total column range ($<1.0 \times 10^{16}$ molecules cm^{-2}) where there are a large number of observations at or
44 near the CrIS noise level (detection limit) ~~and the FTIR total columns are smaller than 1.0×10^{16} molecules cm^{-2} ,~~
45 the absolute differences between CrIS and the FTIR total columns ~~are small with CrIS showing~~ a slight
46 positive column bias ~~around $+2.4 \times 10^{15}$ (standard deviation = 5.5×10^{15}) molecules cm^{-2} , which corresponds to a~~
47 ~~relative difference of $\sim +50\%$ (std = 100%).~~ The CrIS and FTIR profile comparisons differences are mostly within
48 ~~the range of the estimated retrieval uncertainties single level retrieved profile values, showing average differences~~
49 ~~in the range of ~ 20 to 40% .~~ The CrIS retrievals ~~for these comparisons~~ typically show good vertical sensitivity
50 down ~~into the boundary layer that typically peaks at ~ 850 hPa (~ 1.5 km), and at this level~~ ~~the the retrieved~~
51 ~~profiles also compare well with the~~ median absolute difference ~~of~~ 0.87 (std = ± 0.08) ppb ~~and a,~~ corresponding
52 ~~to a~~ median relative difference of 39% (std = $\pm 2\%$)%. Most of the absolute and relative profile comparison
53 differences are in the range of the estimated retrieval uncertainties. ~~However, the CrIS retrieval does~~ ~~At the~~
54 ~~surface, where CrIS typically has lower sensitivity, it tends to overestimate the concentrations in the levels near~~
55 ~~the surface at under low concentrations conditions, and underestimate under higher atmospheric concentration~~
56 ~~conditions, most probably due to the detection limit of the instrument, and at higher concentrations shows more~~
57 ~~of an underestimation of the concentrations in these lower levels.~~

58 1. Introduction

59

60 The disruption of the nitrogen cycle by the human creation of reactive nitrogen has created one of the major
61 challenges for humankind (Rockström et al., 2009). Global reactive nitrogen emissions into the air have
62 increased to unsurpassed levels (Fowler et al., 2013) and are currently estimated to be four times larger than pre-
63 industrial levels (Holland et al., 1999). As a consequence the deposition of atmospheric reactive nitrogen has
64 increased causing ecosystems and species loss (Rodhe et al 2002; Dentener et al., 2006; Bobbink et al., 2010).
65 Ammonia (NH₃) as fertilizer is essential for agricultural production and is one of the most important reactive
66 nitrogen species in the biosphere. NH₃ emission, atmospheric transport, and atmospheric deposition are major
67 causes of eutrophication and acidification of soils and water in semi-natural environments (Erisman et al., 2008,
68 2011). Through reactions with sulphuric acid and nitric acid, ammonium nitrate and ammonium sulphate are
69 formed, which embody up to 50% of the mass of fine mode particulate matter (PM_{2.5}) (Seinfeld and Pandis.,
70 1988; Schaap et al., 2004). PM_{2.5} has been associated with various health impacts (Pope et al., 2002; 2009). At
71 the same time, atmospheric aerosols impact global climate directly through their radiative forcing effect and
72 indirectly through the formation of clouds (Adams et al., 2001; Myhre et al., 2013). By fertilizing ecosystems,
73 deposition of NH₃ and other reactive nitrogen compounds also plays a key role in the sequestration of carbon
74 dioxide (Oren et al., 2001).

75

76 Despite the significance and impact of NH₃ on the environment and climate, its global distribution and budget
77 are still relatively uncertain (Erisman et al., 2007; Clarisse et al., 2009; Sutton et al., 2013). One of the reasons is
78 that in-situ measuring of atmospheric NH₃ at ambient levels is complex due to the sticky nature and reactivity of
79 the molecule, leading to large uncertainties and/or sampling artefacts with the currently used measuring
80 techniques (von Bobruzki et al., 2010; Puchalski et al., 2011). Measurements are also very sparse. Currently,
81 observations of NH₃ are mostly available in north-western Europe and central North America, supplemented by
82 a small number of observations made in China (Van Damme et al., 2015b). Furthermore, there is a lack of
83 detailed information on its vertical distribution as only a few dedicated airborne measurements are available
84 (Nowak et al., 2007, 2010; Leen et al., 2013, Whitburn et al., 2015, Shephard et al., 2015). The atmospheric
85 lifetime of NH₃ is rather short, ranging from hours to a few days. In summary, global emission estimates have
86 large uncertainties. Estimates of regional emissions attributed to source types different from the main regions
87 are even more uncertain due to a lack of process knowledge and atmospheric levels (Reis et al., 2009).

88

89 Over the last decade the developments of satellite observations of NH₃ from instruments such as the Cross-track
90 Infrared Sounder (CrIS, Shephard and Cady-Pereira, 2015), the Infrared Atmospheric Sounding Interferometer
91 (IASI, Clarisse et al., 2009; Coheur et al., 2009; Van Damme et al., 2014a), the Atmospheric Infrared Sounder
92 (AIRS, Warner et al., 2016), and the Tropospheric Emission Spectrometer (TES, Beer et al., 2008; Shephard
93 et al., 2011) show potential to improve our understanding of the NH₃ distribution. Recent studies show the global
94 distribution of NH₃ measured at a twice daily scale (Van Damme et al., 2014a, Van Damme et al., 2015a) and
95 can reveal seasonal cycles and distributions for regions where measurements were unavailable until now.

96 Comparisons of these observations to surface observations and model simulations, show underestimations of the
97 modelled NH₃ concentration levels, pointing to underestimated regional and national emissions (Clarisse et al.,

98 2009; Shephard et al., 2011; Heald et al., 2012; Nowak et al., 2012; Zhu et al., 2013; Van Damme et al., 2014b;
99 Lonsdale et al., 2016; Schiferl et al., 2014, 2016; Zondlo et al., 2016). However, the ~~overall uncertainty quality~~
100 of the satellite observations is still ~~highly uncertain~~ due to a lack of validation. The few validation studies
101 showed a limited vertical, spatial and or temporal coverage of surface observations to do a proper uncertainty
102 analysis (Van Damme et al., 2015b; Shephard et al., 2015; Sun et al., 2015). A recent study by Dammers et al.
103 (2016a) explored the use of Fourier transform infrared (FTIR-NH₃, Dammers et al., 2015) observations to
104 evaluate the uncertainty of the IASI-NH₃ total column product. The study showed the good performance of the
105 IASI-LUT (Look up table, LUT, Van Damme et al., 2014a) retrieval with a high correlation ($r \sim 0.8$), but
106 indicated an underestimation of around 30% due to potential assumptions of the shape of the vertical profile
107 (Whitburn et al., 2016, IASI-NN (Neural Network, NN)), uncertainty in spectral line parameters and
108 assumptions on the distributions of interfering species. The study showed the potential of using FTIR
109 observations to validate satellite observations of NH₃, but also stressed the challenges of validating retrievals
110 that do not provide the vertical measurement sensitivity, such as the IASI-LUT retrieval. Since no IASI satellite
111 averaging kernels are provided for each retrieval, and thus no information is available on the vertical sensitivity
112 and/or vertical distribution of each separate observation, it is hard to determine the cause of the discrepancies
113 between both observations.

114

115 The new CrIS Fast Physical Retrieval (Shephard and Cady-Pereira, 2015) uses an optimal estimation retrieval
116 approach that provides the information content and the vertical sensitivity (derived from the averaging kernels,
117 for more details see Shephard and Cady-Pereira, 2015), and robust and straightforward retrieval error estimates
118 based on retrieval input parameters. The quality of the retrieval has so far not been thoroughly examined against
119 other observations. Shephard and Cady-Pereira (2015) used Observing System Simulation Experiment (OSSE)
120 studies to evaluate the initial performance of the CrIS NH₃ retrieval, and report a small positive retrieval bias of
121 6% with a standard deviation of $\pm 20\%$ (ranging from ± 12 to $\pm 30\%$ over the vertical profile). Note that no
122 potential systematic errors were included in these OSSE simulations. Their study also shows good qualitative
123 comparisons with the Tropospheric Emission Spectrometer (TES) satellite (Shephard et al., 2011) and the
124 ground-level in situ Quantum Cascade-Laser (QCL) observations (Miller et al., 2014) for a case study over the
125 Central Valley in CA, USA, during the DISCOVER-AQ campaign. However, currently there has not been an
126 extensive validation of the CrIS NH₃ retrievals using direct comparisons against vertical profile observations. In
127 this study we will provide both direct comparisons of the CrIS retrieved profiles against ground-based FTIR
128 observations, and comparisons of CrIS total column values against the FTIR and IASI.

129

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130 **2. Methods**

131 **2.1 The CrIS Fast Physical Retrieval**

132 CrIS was launched in late October 2011 on board the Suomi NPP platform. CrIS follows a sun-synchronous
133 orbit with a daytime overpass time at 13:30 local time (ascending) and a night time equator overpass at 1:30.
134 The instrument scans along a 2200 km swath using a 3 x 3 array of circular shaped pixels with a diameter of 14
135 km at nadir for each pixel, becoming larger ovals away from nadir. In this study we use the NH₃ retrieval as
136 described by Shephard and Cady-Pereira (2015). The retrieval is based on an optimal estimation approach
137 (Rodgers, 2000) that minimizes the differences between CrIS spectral radiances and simulated forward model
138 radiances computed from the Optimal Spectral Sampling (OSS) OSS-CrIS (Moncet et al., 2008), which is built
139 from the well-validated Line-By-Line Radiative Transfer Model (LBLRTM) (Clough et al., 2005; Shephard et
140 al., 2009; Alvarado et al., 2013) and uses the HITRAN database (Rothman et al., 2014) for its spectral lines. The
141 fast computational speed of OSS facilitates the operational production of CrIS retrieved (Level 2) products using
142 an optimal estimation retrieval approach (Moncet et al., 2005). The CrIS OSS radiative transfer forward model
143 computes the spectrum for the full CrIS LW band, at the CrIS spectral resolution of 0.625 cm⁻¹ (Tobin, 2012),
144 thus the complete NH₃ spectral band (near 10 μm) is available for the retrievals. However, only a small number
145 of micro windows are selected for the CrIS retrievals to both maximize the information content and minimize
146 the influence of errors. Worden et al., (2004) provides an example of a robust spectral region selection process
147 that takes into consideration both the estimated errors (i.e. instrument noise, spectroscopy errors, interfering
148 species, etc.) and the associated information content in order to select the optimal spectral regions for the
149 retrieval. The a-priori profiles selection for the optimal estimation retrievals follows the Tropospheric Emission
150 Spectrometer (TES) retrieval algorithm (Shephard et al., 2011); Based on the relative NH₃ signal in the spectra
151 the a-priori is selected from one of three possible profiles representing unpolluted, moderate, and polluted
152 conditions. The initial guess profiles are also selected from these three potential profiles.

153
154 An advantage of using an optimal estimation retrieval approach is that averaging kernels (sensitivity to the true
155 state) and the estimated errors of the retrieved parameter are computed in a robust and straight-forward manner
156 (for more details see Shephard and Cady-Pereira, 2015). The total satellite retrieved parameter error is
157 expressed as the sum of the smoothing error (due to unresolved fine structure in the profile), the measurement
158 error (random instrument noise in the radiance spectrum propagated to the retrieval parameter), and systematic
159 errors from uncertainties in the non-retrieved forward model parameters and cross-state errors propagated from
160 retrieval-to-retrieval (i.e. major interfering species such as H₂O, CO₂, and O₃) (Worden et al., 2004). As of yet
161 we have not included error estimates for the systematic errors. The CrIS smoothing error is computed, but since
162 in these FTIR comparison results we apply the FTIR observational operator (which accounts for the smoothing
163 error), the smoothing error contribution is not included in the CrIS errors reported in the comparisons. Thus,

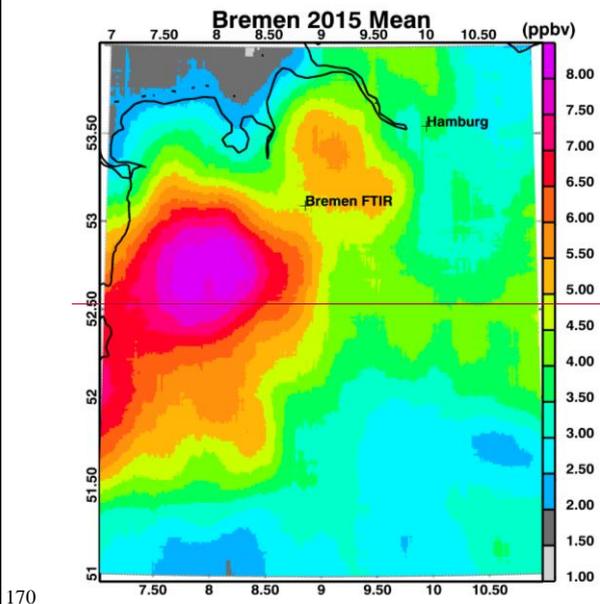
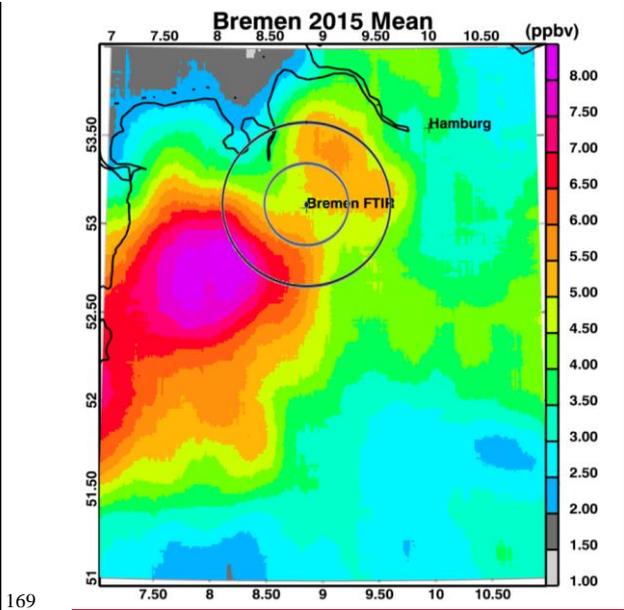
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164 only the measurement errors are reported for observations used here; these errors can thus be considered the
165 lower limit on the total estimated CrIS retrieval error.

166 Figure 1 shows an example of CrIS NH_3 observations surrounding one of the ground-based FTIR instruments.
167 This is a composite map of all days in Bremen with observations in 2015. This figure shows the wide spread
168 elevated amounts of NH_3 across north-western Germany as observed by CrIS.



171 **Figure 1.** Annual mean of the CrIS retrieved NH_3 surface VMR values around the Bremen FTIR site for 2015.

172 The two circles show the collocation area when for radii of 25 and 50 km.

173 Since the goal of this analysis is to evaluate the CrIS retrievals that provide information beyond the a-priori, we

174 only performed comparisons when the CrIS spectrum presents a NH_3 signal. We also focused our efforts on

175 FTIR stations that have FTIR observations with total columns larger than 5×10^{15} molecules cm^{-2} (~1-2 ppb
 176 surface VMR). This restriction does mean that a number of sites of the FTIR-NH₃ dataset will not be used. For
 177 comparability of this study to the results of the IASI-LUT evaluation in an earlier study by Dammers et al.,
 178 (2016a) we include a short paragraph on the performance of the IASI-LUT and the more recent IASI-NN
 179 product when applying similar constraints.

180 2.2 FTIR-NH₃ retrieval

181 The FTIR-NH₃ product used in this study is similar to the set described in Dammers et al. (2016a) and is based
 182 on the retrieval methodology described by Dammers et al. (2015). The retrieval methodology uses two spectral
 183 micro-windows whose spectral width depends on the NH₃ background concentration determined for the
 184 observation stations and location (wider window for stations with background concentrations less than one ppb).
 185 NH₃ is retrieved by fitting the spectral lines in the two micro-windows MW1 [930.32-931.32 cm^{-1} or wide:
 186 929.40-931.40 cm^{-1}] and MW2 [962.70-970.00 cm^{-1} or wide: 962.10-970.00 cm^{-1} . An optimal estimation
 187 approach (Rodgers et al., 2000) is used, implemented in the SFIT4 algorithm (Pougatchev et al., 1995; Hase et
 188 al., 2004, 2006). There are a number of species that can interfere to some extent in both windows, with the
 189 major species being H₂O, CO₂ and O₃ and the minor species N₂O, HNO₃, CFC-12, and SF₆. The HITRAN 2012
 190 database (Rothman et al., 2014) is used for the spectral lines. A further set of spectroscopic line parameter
 191 adjustments are added for CO₂ taken from the ATMOS database (Brown et al., 1996) as well as a set of pseudo-
 192 lines for the broad absorptions by the CFC-12 and SF₆ molecules (created by NASA-JPL, G.C. Toon,
 193 <http://mark4sun.jpl.nasa.gov/pseudo.html>). The NH₃ a-priori profiles are based on balloon measurements (Toon
 194 et al., 1999) and refitted to match the local surface concentrations (depending on the station either measured or
 195 estimated by model results). For the interfering species a-priori profiles we use the Whole Atmosphere
 196 Community Climate Model (WACCM, Chang et al., 2008, v3548). The estimated errors in the FTIR-NH₃
 197 retrievals are in the order of ~30% (Dammers et al., 2015) with the uncertainties in the NH₃ line spectroscopy
 198 being the most important contributor. Based on the data requirements in section 2.1, a set of seven stations is
 199 used (Table 1). For all sites except Wollongong in Australia we use the basic narrow spectral windows. For
 200 Wollongong the wide spectral windows are used. For a more detailed description of each of the stations see the
 201 publications listed in Table 1 or Dammers et al. (2016a).

202 **Table 1.** The location, longitudinal and latitudinal position, altitude above sea level, and type of instrument for
 203 each of the FTIR sites used in this study. In addition, a reference is given to a detailed site description, when
 204 available.

Station	Lon (degrees)	Lat (degrees)	Altitude (m.a.s.l)	FTIR instrument	Reference
Bremen, Germany	8.85E	53.10N	27	Bruker 125 HR	Velazco et al., 2007
Toronto, Canada	79.60W	43.66N	174	ABB Bomem DA8	Wiacek et al., 2007 Lutsch et al., 2016
Boulder, United States	105.26W	39.99N	1634	Bruker 120 HR	
Pasadena, United States	118.17W	34.20N	350	MkIV_JPL	
Mexico City, Mexico	99.18W	19.33N	2260	Bruker Vertex 80	Bezanilla et al., 2014
Wollongong, Australia	150.88E	34.41S	30	Bruker 125 HR	
Lauder, New Zealand	169.68E	45.04S	370	Bruker 120 HR	Morgenstern et al., 2012

206 2.3 IASI-NH₃

207 The CrIS retrieval will also be compared with corresponding IASI/FTIR retrievals using results from a previous
208 study by Dammers et al. (2016a). Both the IASI-LUT (Van Damme et al., 2014a) and the IASI-NN (Neural
209 Networks, Whitburn et al., 2016) retrievals from observations by the IASI instrument aboard MetOp-A will be
210 used. A short description of both IASI retrievals is provided here, for a more in-depth description, see the
211 respective publications by Van Damme et al. (2014a) and Whitburn et al. (2016). The IASI instrument on board
212 the MetOp-A platform is in a sun-synchronous orbit and has a daytime overpass at around 9:30 local solar time
213 and a night time overpass at around 21:30. The instrument has a circular footprint of about 12 km diameter for
214 nadir viewing angles with of nadir observations along a swath of 2100 km. Both IASI retrievals are based on the
215 calculation of a dimensionless spectral index called the Hyperspectral Range Index (HRI) (Van Damme et al.,
216 2014a). The HRI is representative of the amount of NH₃ in the measured column. The IASI-LUT retrieval
217 makes a direct conversion of the HRI to a total column density with the use of a look-up-table (LUT). The LUT
218 is created using a large number of simulations for a wide range of atmospheric conditions which links the
219 Thermal Contrast (TC, the difference between the air temperature at 1.5 km altitude and the temperature of the
220 Earth surface) and the HRI to a NH₃ total column density. The retrieval includes a retrieval error based on the
221 uncertainties in the initial HRI and TC parameters. The more recent IASI-NN retrieval (Whitburn et al., 2016)
222 follows similar steps but it makes use of a neural network. The neural network combines the complete
223 temperature, humidity and pressure profiles for a better representation of the state of the atmosphere. At the
224 same time the retrieval error estimate is improved by including error terms for the uncertainty in the profile
225 shape, and the full temperature and water vapour profiles. The IASI-NN version uses the fixed profiles that were
226 described by Van Damme et al., (2014) but allows for the use of third party profiles to improve the
227 representation of the NH₃ atmospheric profile. The IASI-LUT and IASI-NN retrievals have both been
228 previously compared with FTIR observations (Dammers et al., 2016a, Dammers et al., 2016b). They compared
229 reasonably well with correlations around $r=0.8$ for a set of FTIR stations, with an underestimation of around
230 30% that depends slightly on the magnitude of total column amounts, with the IASI-NN performing slightly
231 better.

232

233 **2.4 Data criteria & quality**

234 NH₃ concentrations show large variations both in space and time as the result of the large heterogeneity in
235 emission strengths due to spatially variable sources and drivers such as meteorology and land use (Sutton et al.,
236 2013). This high variability poses challenges in matching ground-based point observations made by FTIR
237 observations with CrIS downward-looking satellite measurements which have a 14-km nadir footprint. For the
238 pairing of the measurement data we apply data selection criteria similar to that described in Dammers et al.
239 (2016a) and summarized in Table 2. To minimize the impact of the heterogeneity of the sources, we choose a
240 maximum of 50 km between the centre points of the CrIS observations and the FTIR site location. To diminish
241 the effect of temporal differences between the FTIR and CrIS observations a maximum time difference of 90
242 minutes is used. Topographical effects are reduced by choosing a maximum altitude difference of 300 m at any
243 point between the FTIR site location and the centre point of the satellite pixel location. The altitude differences
244 are calculated using the Space Shuttle Radar Topography Mission Global product at 3 arc-second resolution
245 (SRTMGL3, Farr et al., 2007). To ensure the data quality of CrIS-NH₃ retrieval for Version 1.0, a small number
246 of outliers with a maximum retrieved concentration above 200 ppb (at any point in the profile) were removed

247 from the comparison dataset. While potentially a surface NH₃ value of 200 ppb (and above) would be possible
 248 (i.e. downwind of forest fires), it is highly unlikely to occur over the entire footprint of the satellite instrument.
 249 Moreover, after inspecting these data points, they seem to be affected by numerical issues in the fitting
 250 procedure (possibly due to interfering species). As we are interested in validating the CrIS observational
 251 information (not just a-priori information), we only select comparisons that contain some information from the
 252 satellite (degrees of freedom for signal (DOFS) ≥ 0.1). Do note that on average the observations have a DOFS
 253 between 0.9 and 1.1. The DOFS > 0.1 filter only removes some of the outliers at the lower end. No explicit filter
 254 is applied to account for clouds; however, clouds will implicitly be accounted for by the quality control as CrIS
 255 will not measure a NH₃ signal (e.g. DOFS < 0.1) below optically thick clouds (e.g. cloud optical depth $> \sim 1$). In
 256 addition, the CrIS observations are matched with FTIR observations taken only during clear-sky conditions,
 257 which mostly eliminates influence from cloud cover. Finally, the high signal to noise ratios (SNR) of the CrIS
 258 instrument, allows it to retrieve NH₃ from a thermal contrast approaching 0 K during daytime observations
 259 (Clarisse et al., 2010). Given this, we decided not to apply a thermal contrast filter to the CrIS data. No
 260 additional filters are applied to the FTIR observations beyond the clear-sky requirement.

261
 262 For both IASI retrievals, we use the same observation selection criteria as described in Dammers et al. (2016a).
 263 The set of criteria is similar to those used here for the CrIS observations. Observations from both IASI retrievals
 264 are matched using the overpass time, and longitudinal and latitudinal positions. For comparability with CrIS a
 265 spatial difference limit of 50 km limit was used, instead of the 25 km spatial limit used in the previous study.
 266 Furthermore we apply the thermal contrast ($> 12\text{K}$, difference between the temperatures at 1.5 km and the
 267 surface) and Earth skin temperature criteria to the IASI observations to match the previous study.

268
 269 **Table 2. Coincidence criteria and quality flags applied to the satellite and FTIR data. The third through**
 270 **fifth columns show the number of observations remaining after each subsequent data criteria step and**
 271 **the number of possible combinations between the CrIS and FTIR observations. The first set of numbers**
 272 **indicate the number of CrIS observations within a $1^\circ \times 1^\circ$ degree square surrounding the FTIR site.**

Filter	Data Criteria	Nr. Obs.		Combinations
		FTIR	CrIS	
CrIS		15661	25855	
Temporal sampling difference	Max 90 min	1576	13959	112179
Spatial sampling difference	Max 50 km	1514	3134	22869
Elevation difference	Max 300 m	1505	1642	9713
Quality flag	DOFS ≥ 0.1	1433	1453	8579

273
 274 **2.5 Observational Operator Application**

275
 276 To account for the vertical sensitivity and the influence of the a-priori profiles of both retrievals we apply the
 277 observational operator (averaging kernel and a-priori of the retrieval) of the FTIR retrieval to the CrIS retrieved
 278 profiles. The CrIS observations are matched to each individual FTIR observation in time and space following
 279 the matching criteria. The FTIR averaging kernels, a-priori profiles, and retrieved profiles are first mapped to
 280 the CrIS pressure levels (fixed pressure grid, layers are made smaller or cut off for observations above elevation

281 to fit the fixed pressure grid). Following Rodgers and Connor (2003) and Calisesi et al. (2005) this results in the
 282 mapped FTIR averaging kernel, A_{ftir}^{mapped} , the mapped FTIR apriori, $x_{ftir}^{mapped,apriori}$, and the mapped FTIR
 283 retrieved profile, x_{ftir}^{mapped} . Then we apply the FTIR observational operator to the CrIS observations using eq.
 284 (1).

$$285 \hat{x}_{CrIS} = x_{ftir}^{mapped,apriori} + A_{ftir}^{mapped}(x_{CrIS} - x_{ftir}^{mapped,apriori}) \quad (1)$$

$$286 \widehat{\Delta x}_{abs} = \hat{x}_{CrIS} - x_{ftir}^{mapped} \quad (2)$$

$$287 \widehat{\Delta x}_{rel} = (\hat{x}_{CrIS} - x_{ftir}^{mapped}) / (0.5 x_{ftir}^{mapped} + 0.5 \hat{x}_{CrIS}) \quad (3)$$

288 where $x_{ftir}^{apriori}$ is the FTIR a-priori profile, x_{ftir}^{mapped} is the interpolated FTIR profile, A_{ftir}^{mapped} is the FTIR
 289 averaging kernel, and \hat{x}_{CrIS} is the smoothed CrIS profile.

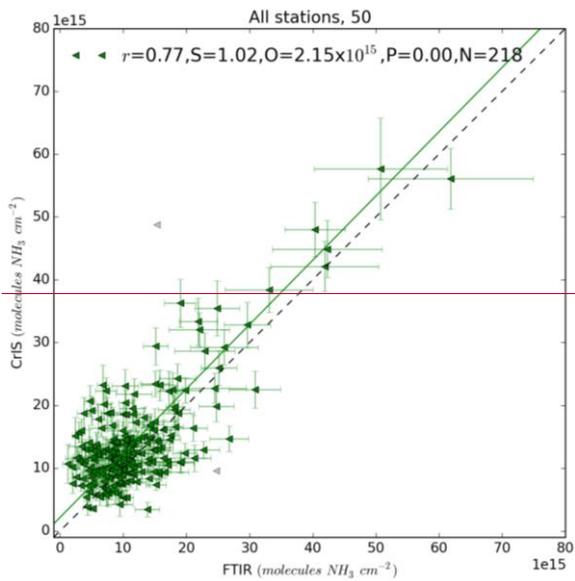
290
 291 The CrIS smoothed profile \hat{x}_{CrIS} calculated from equation (1) provides an estimate of the FTIR retrieval applied
 292 to the CrIS satellite profile. Next we evaluate both total column and profile measurements.

293 For the first validation step, following Dammer et al. (2016a), who evaluated the IASI-LUT (Van Damme et
 294 al., 2014a) product, we sum the individual profile (\hat{x}_{CrIS}) to obtain a column total to compare to the FTIR total
 295 columns. This step gives the opportunity to evaluate the CrIS retrieval in a similar manner as was done with the
 296 IASI-LUT retrieval. If multiple FTIR observations match a single CrIS overpass we also average those together
 297 into a single value as well as each matching averaged CrIS observation. Therefore, it is possible to have multiple
 298 FTIR observations, each with multiple CrIS observations all averaged into a single matching representative
 299 observation. For the profile comparison this averaging is not performed to keep as much detail available as
 300 possible. An important point to make is that this approach assumes that the FTIR retrieval gives a better
 301 representation of the truth. While this may be true, the FTIR retrieval will not match the truth completely. For
 302 readability we assume that the FTIR retrieval indeed gives a better representation of the truth, and in the next
 303 sections will describe the case in which we apply the FTIR observational operator to the CrIS values. For the
 304 tenacious reader we included a similar set of results in the appendix, using the CrIS observational operator
 305 instead of the FTIR observational operator, as the assumption of the FTIR being truth is not exactly right.

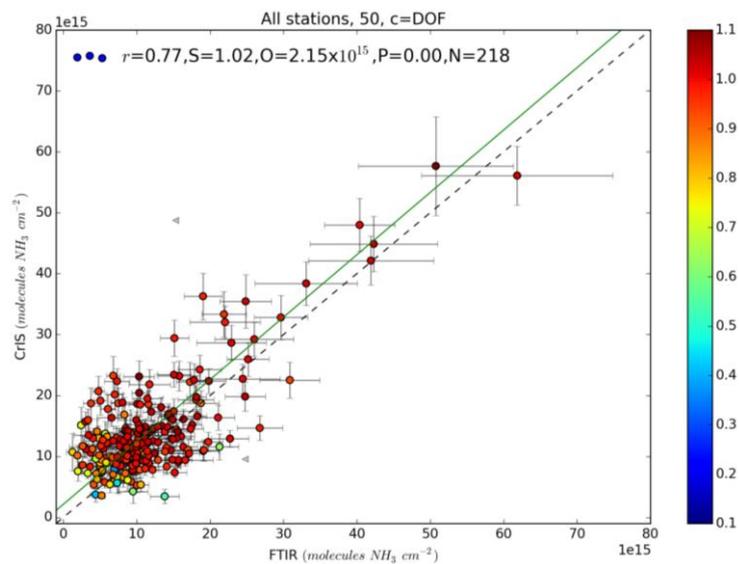
306

307 3. Results and discussion

308 3.1 Total column comparison



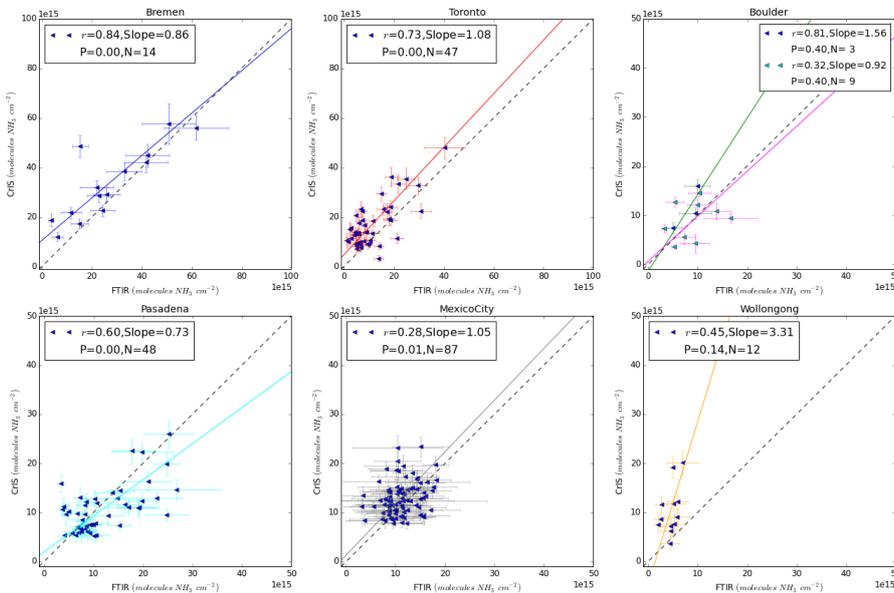
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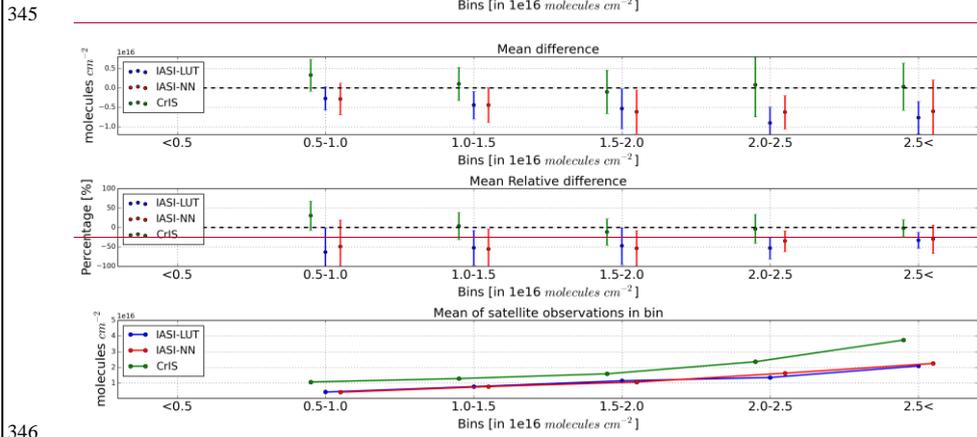
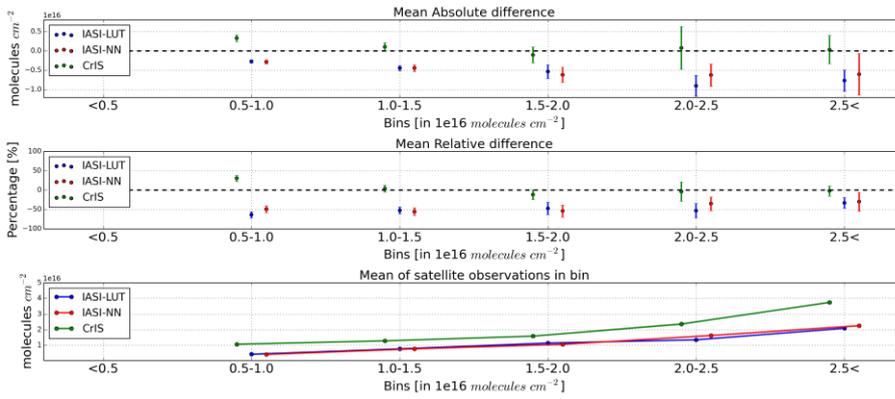
311 **Figure 2.** Correlation between the FTIR and CrIS total columns using the coincident data from all measurement
 312 sites. The horizontal and vertical bars show the total estimated error on each FTIR and CrIS observation. The
 313 colouring on the scatter indicates the mean DOF of each the CrIS coincident data. The trend line shows the
 314 results of the regression analysis.
 315

316 The total columns are averaged as explained in Section 2.4 to show a direct comparison of FTIR measurements
 317 with CrIS observations in Figure 2. A three sigma outlier filter was applied to calculate the regression statistics.
 318 The filtered outliers are displayed in grey, and may be caused by low information content (DOFS) and terrain
 319 characteristics. For the regression we used the reduced major axis regression (Bevington and Robinson, 1992),
 320 accounting for possible errors both in the x and y values. ~~There is an overall agreement is good~~ with a
 321 correlation of $r = 0.77$ ($P < 0.01$, $N = 218$) and a slope of 1.02 (± 0.05). At the lower range of values the CrIS
 322 column totals are ~~significantly~~ higher than the observed FTIR values. Possibly the CrIS retrieval overestimates
 323 due to the low sensitivity to low concentrations. Without the sensitivity the retrieval will find a value more
 324 closely to the a-priori, which may be too high. Figure 3 shows the comparisons at each station. When the
 325 comparisons are broken down by station (Figure 3), the correlation varies from site to site, from a minimum of
 326 0.28 in Mexico City (possibly due to retrieval errors associated with the highly irregular terrain) to a maximum
 327 of 0.84 in Bremen. Similarly to Mexico City the comparison also shows an increase in scatter for Pasadena,
 328 where the FTIR site is also located on a hill. In Toronto, ~~and Bremen and Pasadena~~ there is good agreement
 329 when NH_3 is elevated ($> 20 \times 10^{15}$ molecules cm^{-2}) ~~in~~, and low bias in the CrIS total columns for intermediate
 330 values (between 10 and 20×10^{15} molecules cm^{-2}) except for the outlying observation in Bremen, which is
 331 marked as an outlier by our three sigma filter used for Figure 2. In Wollongong, there is less agreement between
 332 the instruments. There are two comparisons with large CrIS to FTIR ratios while most of the other comparisons
 333 also show a bias for CrIS. For both cases the bias can be explained by the heterogeneity of the ammonia
 334 concentrations in the surrounding regions. The two outlying observations were made during the end of
 335 November, 2012, which coincides with wild fires in the surrounding region. Furthermore the Wollongong site is
 336 located coastally, which will increase the occurrences where one instrument observes clean air from the ocean
 337 while the other observes inland air masses.
 338



339

340 **Figure 3.** FTIR vs CrIS comparison scatter plots showing the correlations for each of the individual stations,
 341 with estimates error plotted for each value. The trend lines show the individual regression results. Note the
 342 different ranges on the x and y axis. The results for the Boulder (green line) and Lauder (pink line) sites are
 343 shown in the same panel.
 344



346 **Figure 4.** Plots of the mean absolute and relative differences between CrIS and IASI, as a function of NH₃ total
 347 column. Observations are separated into bins of total columns. The upper panel shows the mean absolute
 348 difference (MD). The middle panel shows the mean relative difference. The bars in these top two panels show
 349 the ~~standard deviation~~ 95 % confidence interval for each value. The bottom panel shows the mean of the
 350 observations in each bin. The number of observations in each set is shown in the bottom panel.
 351

352
 353 The mean absolute (MD) and relative difference (MRD) are calculated following equation 4 and equation 5;

354
$$MRD = \frac{1}{N} \sum_{i=1}^N \frac{(CrIS\ column_i - FTIR\ column_i) \times 100}{0.5 * FTIR\ column_i + 0.5 * CrIS\ column_i} \quad (4)$$

355
$$MD = \frac{1}{N} \sum_{i=1}^N (CrIS\ column_i - FTIR\ column_i) \quad (5)$$

356 with N being the number of observations.

357
 358 **Table 3.** Results of the total column comparisons of the FTIR to CrIS, FTIR to IASI-LUT and FTIR to IASI-
 359 NN. N is the number of averaged total columns, MD is the mean difference [10^{15} molecules cm^{-2}], MRD is the
 360 mean relative difference [frac, in %]. Take note that the combined value N does not add up with all the separate
 361 sites as observations have been included for FTIR total columns $> 5 \times 10^{15}$ molecules cm^{-2} .

Retrieval	Column total range in molecules cm^{-2}	N	MD in 10^{15} (1σ)	MRD in % (1σ)	FTIR mean in 10^{15} (1σ)
CrIS-NH ₃	$< 10.0 \times 10^{15}$	93	3.3 (4.1)	30.2 (38.0)	7.5 (1.5)
CrIS-NH ₃	$\geq 10.0 \times 10^{15}$	109	0.4 (5.3)	-1.39 (34.4)	16.7 (8.5)
IASI-LUT	$< 10.0 \times 10^{15}$	229	-2.7 (3.0)	-63.6 (62.6)	7.1 (1.4)
IASI-LUT	$\geq 10.0 \times 10^{15}$	156	-5.1 (4.2)	-50.2 (43.6)	14.8 (6.7)
IASI-NN	$< 10.0 \times 10^{15}$	212	-2.2 (3.6)	-57.0 (68.7)	7.1 (1.4)
IASI-NN	$\geq 10.0 \times 10^{15}$	156	-5.0 (5.1)	-52.5 (49.7)	14.8 (6.7)

362
 363 We evaluate the data by subdividing the comparisons over a set of total column bins as a function of the FTIR
 364 total column value of each individual observation. The bins (with a range of 5×10^{15} to 25×10^{15} molecules cm^{-2}
 365 with iterations steps of 5×10^{15} molecules cm^{-2}) give a better representation of the performance of the retrieval
 366 as it shows the influence of the retrieval as a function of magnitude of the total column densities. The results of
 367 these total column comparisons are presented in Figure 4. Table 3 summarizes the results for each of the FTIR
 368 to satellite column comparisons into two total column bins, which splits the comparisons between smaller and
 369 larger than 10×10^{15} molecules cm^{-2} . A few combinations of the IASI-NN and FTIR retrievals have a small
 370 denominator value that causes problems in the calculation of the MRD. A three sigma outlier filter based on the
 371 relative difference is applied to remove these outliers ($< 10 \times 10^{15}$ molecules cm^{-2} , only the IASI-NN set). The
 372 statistical values are not given separately by site because of the low number of matching observations for a
 373 number of the sites.

374
 375 The CrIS/FTIR comparison results show a large positive difference in both the absolute (MD) and relative
 376 (MRD) for the smallest bin, (5.0 - 10.0×10^{15} molecules cm^{-2}). The rest of the CrIS/FTIR comparison bins with
 377 NH₃ values $> 10.0 \times 10^{15}$ agree very well with a nearly constant bias (MD) around zero, and a standard
 378 deviation of the order of 5.0×10^{15} that slightly dips below zero in the middle bin. The standard deviation over
 379 these bins is also more or less constant, and the weak dependence on the number of observations in each bin
 380 indicates that most of the effect is coming from the random error on the observations. The relative difference
 381 becomes systematically smaller with increasing column total amounts, and tend towards zero with a standard
 382 deviation ~ 25 - 50% , which is on the order of the reported estimated errors of the FTIR retrieval (Dammers et al,
 383 2015).

384
 385 For a comparison against previous reported satellite results, we included both the IASI-LUT (Van Damme et al.,
 386 2014a) and the IASI-NN (Whitburn et al., 2016) comparisons against the FTIR observations. [To put the results](#)
 387 [of this study into perspective of the IASI-LUT and IASI-NN products we added Figure A1 to the Appendix,](#)
 388 [which shows the total column comparison for both products.](#) Both IASI products show similar differences as a
 389 function of NH₃ column bins, which is somewhat different from the CrIS/FTIR comparison results. The absolute
 390 difference (MD) is mostly negative with the smallest factor for the smallest total column bin, with a difference
 391 around -2.5×10^{15} ([std = \$\pm 3.0 \times 10^{15}\$, N = 229](#)) molecules cm^{-2} that slowly increases as a function of the total

392 column. However, the relative difference (MRD) is at its maximum for the smaller bin with a difference of the
393 order -50% ($\text{std} = \pm 50\%$, $N=229$) which decreases to $\sim -10-25\%$ ($\text{std} = \pm 25\%$) with increasing bin value. For
394 both the IASI-NN and IASI-LUT retrievals we find an underestimation of the total columns, which originates
395 mostly from a large systematic error in combination with more randomly distributed error sources such as the
396 instrument noise and interfering species, which is similar to results reported earlier for IASI-LUT (Dammers et
397 al., 2016b).

398
399 A number of factors, besides the earlier reported FTIR uncertainties, can explain the differences between the
400 FTIR and CrIS measurements. The small positive bias found for CrIS points to a small systematic error. The
401 higher SNR, from both the low radiometric noise and high spectral resolution, ~~along with the shorter~~
402 ~~atmospheric path lengths for observations from the ground-based solar pointing FTIR instrument,~~ enables it to
403 resolve smaller gradients in the retrieved spectra, which potentially can provide greater vertical information and
404 detect smaller column amounts (lower detection limit). This could explain the larger MRD and MD CrIS
405 differences at the lower end of the total column range. However, a number of standalone tests with the FTIR
406 retrieval showed only a minor increase in the total column following a decrease in spectral resolution, which
407 indicates that the spectral resolution itself is not enough to explain the difference.

408

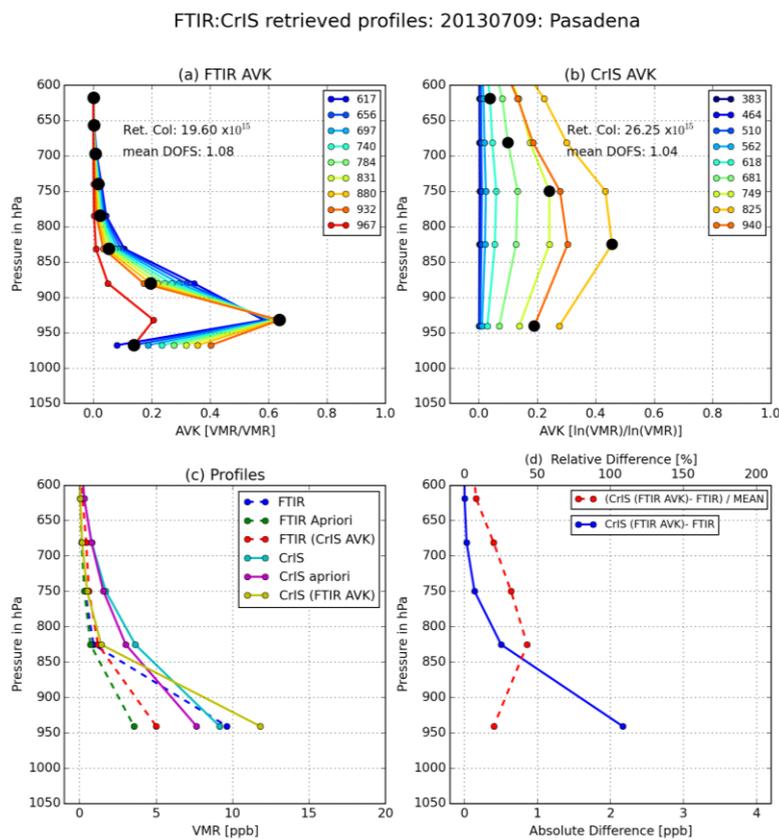
409 3.2 Profile Comparison

410 The CrIS satellite and FTIR retrieved profiles are matched using the criteria specified above in Table 2 and
411 compared. It is possible for a CrIS observation to be included multiple times in the comparison as there can be
412 more than one FTIR observation per day, and /or, the possibility of multiple satellite overpasses that match a
413 single FTIR observation.

414 A representative profile example

415 An example of the profile information contained in a representative CrIS and FTIR profile is shown in Fig. 5.
416 Although the vertical sensitivity and distribution of NH_3 differs per station this is a fairly representative. The
417 FTIR usually has a somewhat larger DOFS in the order of 1.0-2.0, mostly depending on the concentration of
418 NH_3 , compared to the CrIS total of ~ 1 DOFS. Figure 5a shows an unsmoothed FTIR averaging kernel [vmr
419 $^{-1}$] of a typical FTIR observation. The averaging kernel (AVK) peaks between the surface and ~ 850 hPa, which is
420 typical for most observations. In specific cases with plumes overpassing the site, the averaging kernel peak is at
421 a higher altitude matching the location of the NH_3 plume. The CrIS averaging kernel (Fig. 5b) usually has a
422 maximum somewhere in between 680-850 hPa depending on the local conditions. This particular observation
423 has a maximum near the surface, an indication of a day with ~~good-high~~ thermal contrast. Both the FTIR and
424 CrIS concentration profiles have a maximum at the surface with a continuous decrease that mostly matches the
425 a-priori profile in shape following the low DOFS. This is visible for layers at the lower pressures (higher
426 altitudes) where the FTIR and CrIS a-priori and retrieved volume mixing ratios become similar and near zero.
427 The absolute difference between the FTIR and CrIS profiles can be calculated by applying the FTIR
428 observational operator to the CrIS profile, as we described in section 2.5. The largest absolute difference (Fig.
429 5d) is found at the surface, which is also generally where the largest absolute NH_3 values occur. The FTIR
430 smoothed relative difference (red, striped line) peaks at the pressure where the sensitivity of the CrIS retrieval is
431 highest ($\sim 55\%$), which goes down to $\sim 20-30\%$ for the higher altitude and surface pressure layers. Overall the

432 retrievals agree well with most of the difference explained by the estimated errors of the individual retrievals.
 433 For an illustration of the systematic and random errors on the FTIR and CrIS profiles shown in Fig 5, see the
 434 figures in the appendix: for the FTIR error profile see Fig. A24 (absolute error) and A32 (relative error) and for
 435 the CrIS measurement error profile see Fig. A43. Please note that we only show the diagonal error covariance
 436 values for each of the errors, which is common practice. The total column of our example profile is $\sim 20 \times 10^{15}$
 437 molecules cm^{-2} which is a slightly larger value than average. The total random error is $< 10\%$ for each of the
 438 layers, mostly dominated by the measurement error, which is somewhat smaller than average (Dammers et al.,
 439 2015) following the larger NH_3 VMR. A similar value is found for the CrIS measurement error with most layers
 440 showing an error $< 10\%$. The FTIR systematic error is around $\sim 10\%$ near the surface and grows to a larger 40%
 441 for the layers between 900 – 750 hPa. The error is mostly due to the errors in the NH_3 spectroscopy (Dammers
 442 et al., 2015). The shape of the relative difference between the FTIR and CrIS closely follows the shape
 443 systematic error on the FTIR profile pointing to that error as the main cause of difference.



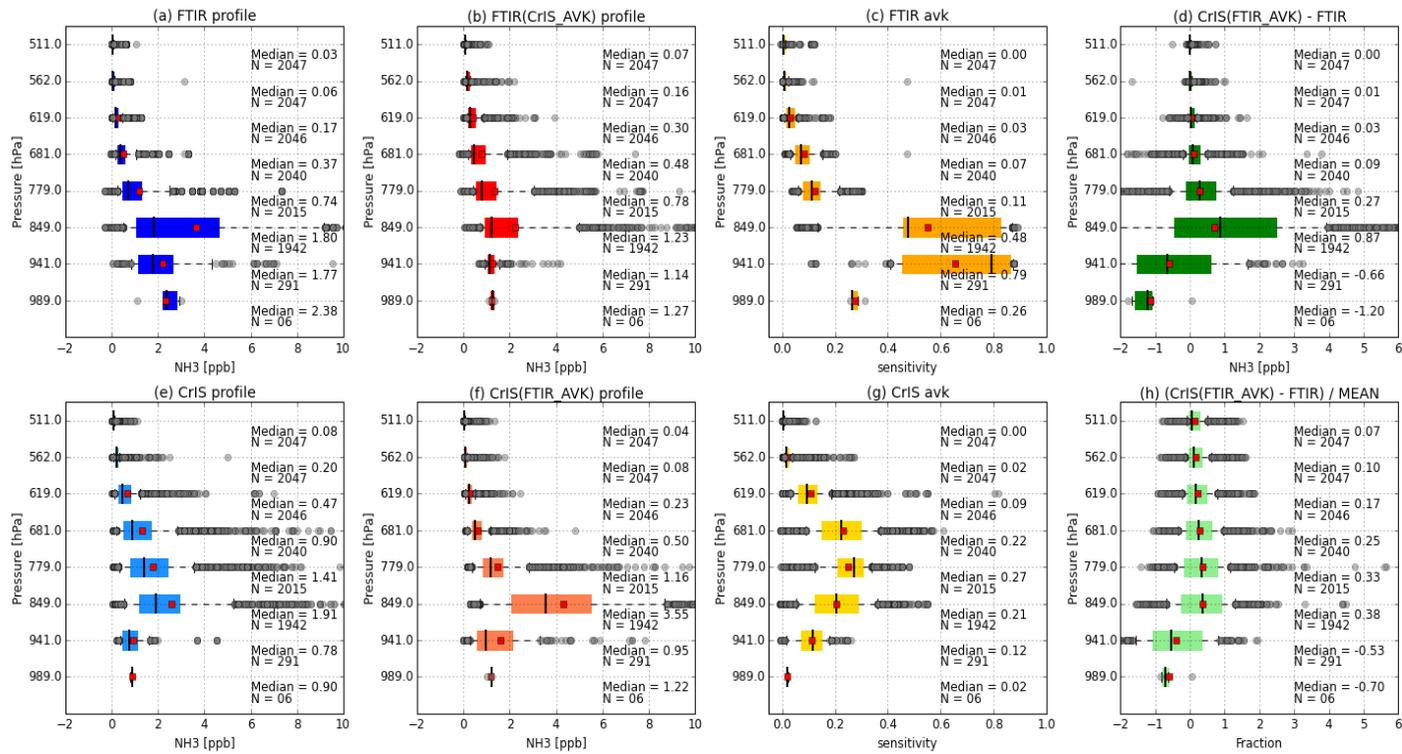
444
 445 **Figure 5.** Example of the NH_3 profile comparison for an FTIR profile matched with a CrIS profile measured
 446 around the Pasadena site. With (a) the FTIR averaging kernel, (b) the CrIS averaging kernel. For both averaging
 447 kernels the black dots show the matrices diagonal values. Panel (c) shows the retrieved profiles of both FTIR

448 (blue) and CrIS (cyan) with the FTIR values mapped to the CrIS pressure layers. Also shown are the FTIR a-
449 priori (green), the CrIS a-priori (purple), the CrIS retrieved profile smoothed with the FTIR averaging kernel
450 [CrIS (FTIR AVK)] (yellow) and the FTIR profile smoothed with the CrIS averaging kernel [FTIR (CrIS
451 AVK)](red). In panel (d), the blue line is the absolute difference between the FTIR profile (blue, panel (c)) and
452 the CrIS profile smoothed with the FTIR averaging kernel (Yellow, panel (c)) with the red line the
453 corresponding relative difference.

454 **All paired data**

455 In Fig. 6 all the individual site comparisons were merged. The Mexico City site was left out of this figure
456 because of the large number of observations in combination with a difference in pressure grid due to the high
457 altitude of the city ~~which~~-obscured the overall analysis and biased the results towards the results of one station.
458 Similar to the single profile example, the FTIR profile peaks near the surface for most observations, slowly
459 going towards zero with decreasing pressure. Compared to the representative profile example a number of
460 differences emerge. A number of FTIR observations peak further above the surface and are shown as outliers,
461 which drag the mean further away from the median values. The combined CrIS profile in Fig. 6 shows a similar
462 behaviour, although for the lowest pressure layer it has a lower median and mean compared to the layer above.
463 The difference between Fig. 5 and Fig. 6e derives mostly from the number of observations used in the boxplot,
464 many with weak sensitivity at the surface. Similar to the single profile example in Fig. 5, the FTIR averaging
465 kernels in Fig. 6c on average peak near or just above the surface (with the diagonal elements of the AVK's
466 shown in the figure). The sensitivity varies a great deal between the observations as shown by the large spread
467 of the individual layers. The CrIS averaging kernels (Fig. 6g) usually peak in the boundary layer around the 779
468 hPa layer with the 2 surrounding layers having somewhat similar values. The instrument is less sensitive to the
469 surface layer as is demonstrated by the large decrease in the AVK near the surface, but this varies depending on
470 the local conditions. We find the largest absolute differences in the lower three layers, as was seen in the
471 example in Fig. 5, although the differences decrease downwards rather than increase. The relative difference
472 shows a similar shape to Fig 5. Overall both retrievals ~~agree quite well~~show agreement. The relative differences
473 in the single level retrieved profile values in Fig. 6h show an average difference in the range of ~20 to 40% with
474 the 25th and 75th percentiles at around 60-80%, which partially follows from our large range of concentrations.
475 The absolute difference shows an average difference in the range of -0.66 to 0.87 ppb around the peak
476 sensitivity levels of the CrIS observations (681 to 849 hPa). The lower number of surface observations follow
477 from the fact that only the Bremen site is located at an altitude low enough for the CrIS retrieval to provide a
478 result at this pressure level-. ~~Because of~~Due to this difference in retrieval layering, the remaining 227
479 observations mostly follow from matching observations in Bremen, which is located in a region of significant
480 NH₃ emissions.

481



482

483 **Figure 6.** Profile comparison for all stations combined. Observations are combined following pressure “bins”, i.e. the midpoints of the CrIS pressure grid.

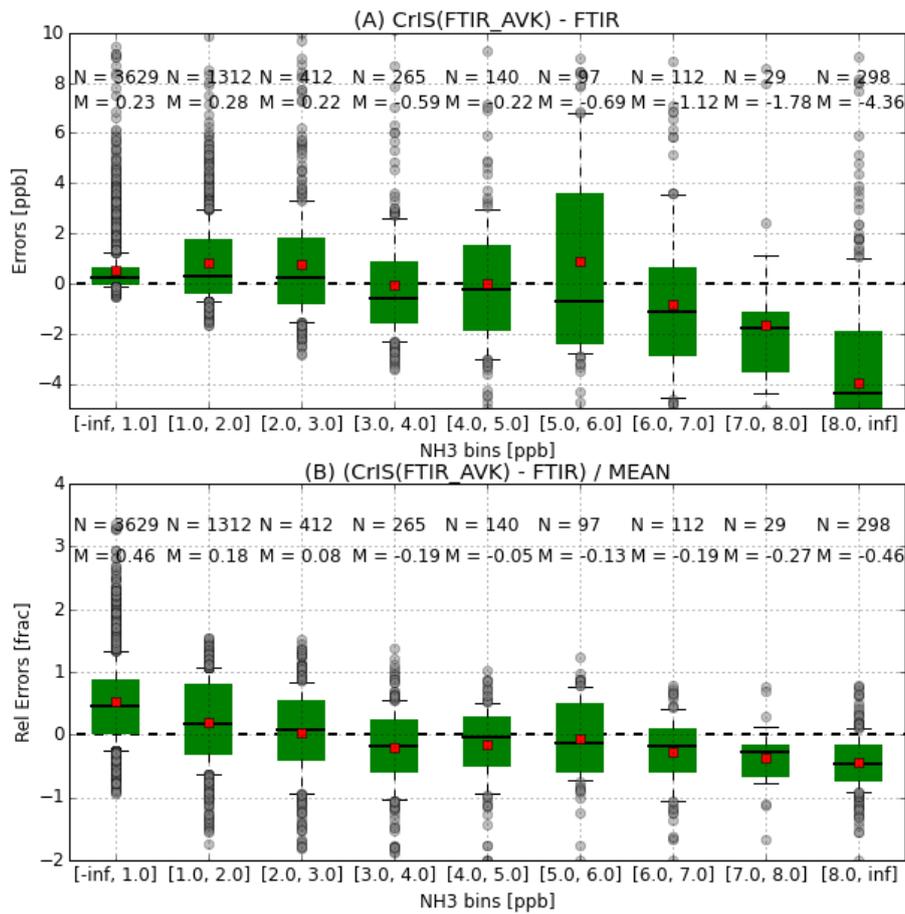
484 Subplot (a) shows the mean profiles of the FTIR (blue), (b) the profiles of FTIR with the CrIS averaging kernel applied to it (red), (c) the FTIR averaging

485 kernel diagonal values, and (d) shows the absolute difference [VMR] between profiles (f) and (a). The second row shows the CrIS mean profile in (e), (f) the

486 profiles of CrIS with the FTIR averaging kernel applied, (g) the CrIS averaging kernel diagonal values, (h) the relative difference [Fraction] between the

487 profiles in (f) and (a). Each of the boxes edges are the 25th and 75th percentiles, the black lines in each box is the median, the red diamond-square is the mean,

488 the whiskers are the 10th and 90th percentiles, and the grey circles are the outlier values outside the whiskers.



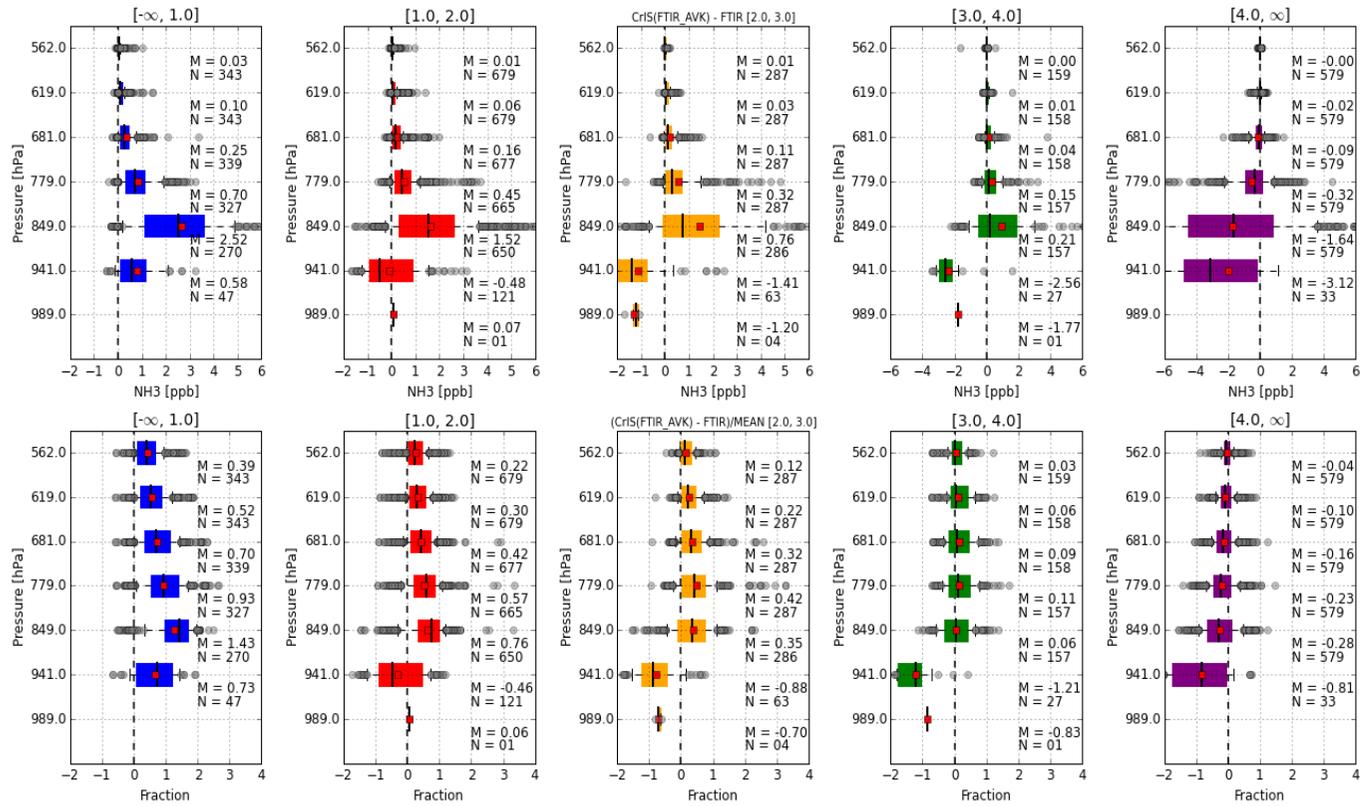
489
 490 **Figure 7.** Summary of the errors-absolute and relative actual error as a function of the VMR of NH₃ in the
 491 individual FTIR layers. The box edges are the 25th and 75th percentiles, the black line in the box is the median,
 492 the red square diamond is the mean, the whiskers are the 10th and 90th percentiles, and the grey circles are the
 493 outlier values outside the whiskers. Only observations with a pressure greater than 650 hPa are used. The top
 494 panel shows the absolute difference for each VMR bin, the bottom panel shows the relative difference for each
 495 VMR bin.

496
 497 The switch between negative and positive values in the absolute difference (see Fig. 6d), occurs in the two
 498 lowest layers dominated by the Bremen observations and provides insight into the relation between absolute
 499 differences as function of retrieved concentration. Fig. 7 shows a summary of the differences as a function of the
 500 individual NH₃ VMR layer amounts. As seen before in the column comparison, e.g. Fig 2 and 4, the CrIS
 501 retrieval gives larger total columns than the FTIR retrieval for the small values of VMR. For increasing VMRs,
 502 this slowly tends to a negative absolute difference with a relative difference in the range of 20-30%. However,

503 note that the number of compared values in these high VMR bins are by far lower than in the first three bins
504 leading to relatively less effect in the total column and merged VMR figures (Figs. 2 and 6) from these high
505 VMR bins. We now combine the results of Figs. 6 and 7 into Figure 8 to create a set of subplots showing the
506 difference between both retrieved profiles as a function of the maximum VMR of each retrieved FTIR profile.
507 For the layers with pressure less than 681 hPa we generally find ~~good~~ agreement, which is expected but not very
508 meaningful, since there is not much NH₃ (and thus sensitivity) in these layers and any differences are smoothed
509 out by the application of the observational operator. The relative differences for these layers all lie around ~0-
510 20%. For the lowest two VMR bins we find again that CrIS gives larger results than the FTIR, around the CrIS
511 sensitivity peak in the layer centred around 849 hPa, and to a lesser extent in the layer below. At these VMR
512 levels (< 2 ppb) the NH₃ signal approaches the spectral noise of the CrIS measurement, making the retrievals
513 more uncertain. The switch lies around 2-3 ppb where the difference in the SNR between the instruments
514 becomes less of an issue. Also easily observed is the relation between the concentration and the absolute and
515 relative differences. This can be explained by the difference in sensitivity of the instruments, and the
516 measurement noise of both instruments. For the largest VMR bin [> 4.0 ppb] we find that CrIS is biased for the
517 four lowest layers. Differences are largest in the surface layer where only a few observations are available,
518 almost all from the Bremen site. Most of these CrIS observations have a peak satellite sensitivity at a higher
519 altitude than the FTIR. Assuming that most of the NH₃ can be found directly near the surface, with the
520 concentration dropping off with a sharp gradient as a function of altitude, it is likely that these concentrations
521 are not directly observed by the satellite but are observed by the FTIR instruments. This difference in sensitivity
522 should be at least partially removed by the application of the observational operator but not completely, due to
523 the intrinsic differences between both retrievals. The CrIS retrieval uses one of three available a-priori profiles,
524 which is chosen following a selection based on the strength of NH₃ signature in the spectra. The three a-priori
525 profiles (unpolluted, moderately polluted and polluted) are different in both shape and concentrations. Out of the
526 entire set of 2047 combinations used in Fig. 8, only six are of the not polluted a priori category. About 1/3 of the
527 remaining observations use the polluted a-priori, which has a sharper peak near the surface (see Fig. 5c),
528 compared to the moderately polluted profile, which is used by 2/3s of the CrIS retrievals shown in this work.
529 Based on the results as a function of retrieved VMR (as measured with the FTIR so not a perfect restriction), it
530 is possible that the sharper peak at the surface as well as the low a-priori concentrations are restricting the
531 retrieval. The dependence of the differences on VMR can also possibly follow from uncertainties in the line
532 spectroscopy. In the lower troposphere there is a large gradient in pressure and temperature and the impact of
533 any uncertainty in the line spectroscopy is greatly enhanced. Even for a day with large thermal contrast and NH₃
534 concentrations (e.g. Fig 5.), the difference between both the CrIS and FTIR retrievals was dominated by the line
535 spectroscopy. This effect is further enhanced by the higher spectral resolution and reduced instrument noise of
536 the FTIR instrument, which potentially makes it more able to resolve the line shapes.

537

538



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542

543

Figure 8. Summary of differences as a function of maximum volume mixing ratio (VMR). The maximum VMR of each FTIR profiles is used for the classification. Absolute (Top row) and relative profile differences (bottom row) following the FTIR and CrIS (FTIR AVK applied) profiles. Observations are following pressure layers, i.e. the midpoints of the CrIS pressure grid. The box edges are the 25th and 75th percentiles, the black line in the box is the median, the red square diamond is the mean, the whiskers are the 10th and 90th percentiles, and the grey circles are the outlier values outside the whiskers.

544 To summarise, the overall differences between both retrievals are quite small, except for the lowest layers in the
545 NH₃ profile where CrIS has less sensitivity. The differences mostly follow the errors as estimated by the FTIR
546 retrieval and further effort should focus on the estimated errors and uncertainties. A way to improve the
547 validation would be to add a third set of measurements with a better capability to vertically resolve NH₃
548 concentrations from the surface up to ~750 hPa (i.e. the first 2500 m). One way to do this properly is probably
549 by using airplane observations that could measure a spiral around the FTIR path coinciding with a CrIS
550 overpass. The addition of the third set of observations would improve our capabilities to validate the satellite
551 and FTIR retrievals and point out which retrieval specifically is causing the absolute and relative differences at
552 each of the altitudes.

553

554 4. Conclusions

555

556 Here we presented the first validation of the CrIS-NH₃ product using ground-based FTIR-NH₃ observations. The
557 total column comparison shows that both retrievals agree well with have a correlation of $R = 0.77$ ($P < 0.01$, $N =$
558 218) and almost no bias with an overall slope of 1.02 ($std = \pm 0.05$). For the individual stations we find varying
559 levels of agreement mostly limited by the small range of NH₃ total columns. For FTIR total columns $> 10 \times 10^{15}$
560 molecules cm⁻² the CrIS and FTIR observations agree very well are in agreement with only a small bias of 0.4
561 ($std = \pm 5.3$) $\times 10^{15}$ molecules cm⁻², and a relative difference 4.57 ($std = \pm 35.8$) %. In the smaller total column
562 range the CrIS retrieval shows a positive bias with larger relative differences 49.0 ($std = \pm 62.6$) % that mostly
563 seems to follow from observations near the CrIS detection limit. The results of the comparison between the
564 FTIR and the IASI-NN and IASI-LUT retrievals, are comparable to those found in earlier studies. Both IASI
565 products showed smaller total column values compared to the FTIR, with a MRD ~-35- -40%. On average, the
566 CrIS retrieval has one piece of information, while the FTIR retrieval shows a bit more vertical information with
567 DOFS in the range of 1-2. The NH₃ profile comparison shows similar results, with a small mean negative
568 difference between the CrIS and FTIR profiles for the surface layer and a positive difference for the layers
569 above the surface layer. The relative and absolute differences in the retrieved profiles can be explained by the
570 estimated errors of the individual retrievals. Two causes of uncertainty stand out with the NH₃ line spectroscopy
571 being the biggest factor, showing errors of up to 40% in the profile example. The second factor is the signal-to-
572 noise ratio of both instruments which depends on the VMR: under large NH₃ concentrations, the FTIR
573 uncertainty in the signal is in the range of 10%; for measurements with small NH₃ concentrations this greatly
574 increases. Future work should focus on improvements to the NH₃ line spectroscopy to reduce the uncertainty
575 coming from this error source. Furthermore an increased effort is needed to acquire coincident measurements
576 with the FTIR instruments during satellite overpasses as a dedicated validation effort will greatly enhance the
577 number of available observations. Furthermore, a third type of observations measuring the vertical distribution
578 of NH₃ could be used to compare with both the FTIR and CrIS retrievals and further constrain the differences.
579 These observations could be provided by an airborne instrument flying spirals around an FTIR site during a
580 satellite overpass.

581

582 **5. Data availability**

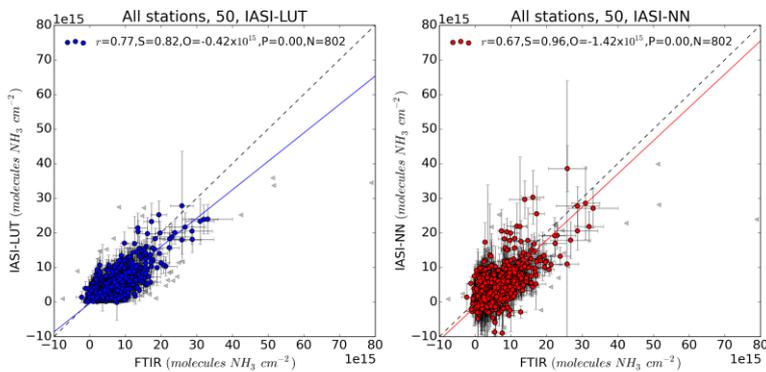
583 FTIR-NH₃ data (Dammers et al., 2015) can be made available on request (M. Palm, Institut für Umweltphysik,
584 University of Bremen, Bremen, Germany). The CrIS-FRP-NH₃ science grade (non-operational) data products
585 used in this study can be made available on request (M. W. Shephard, Environment and Climate Change
586 Canada, Toronto, Ontario, Canada). The IASI-NH₃ product is freely available at <http://www.pole-ether.fr/etherTypo/index.php?id=1700&L=1> (Van Damme et al., 2015a).

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Appendix A.



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Figure A1. Correlation between the FTIR and the IASI-LUT (left, blue) and IASI-NN (right, red) total columns using the coincident data from all measurement sites. The horizontal and vertical bars show the total estimated error on each FTIR and CrIS observation. A three sigma outlier filter was applied to the IASI-LUT dataset and the same observations were removed from the IASI-NN set. Contrary to the earlier study by Dammers et al., (2016a) no thermal contrast filter was applied to the dataset.

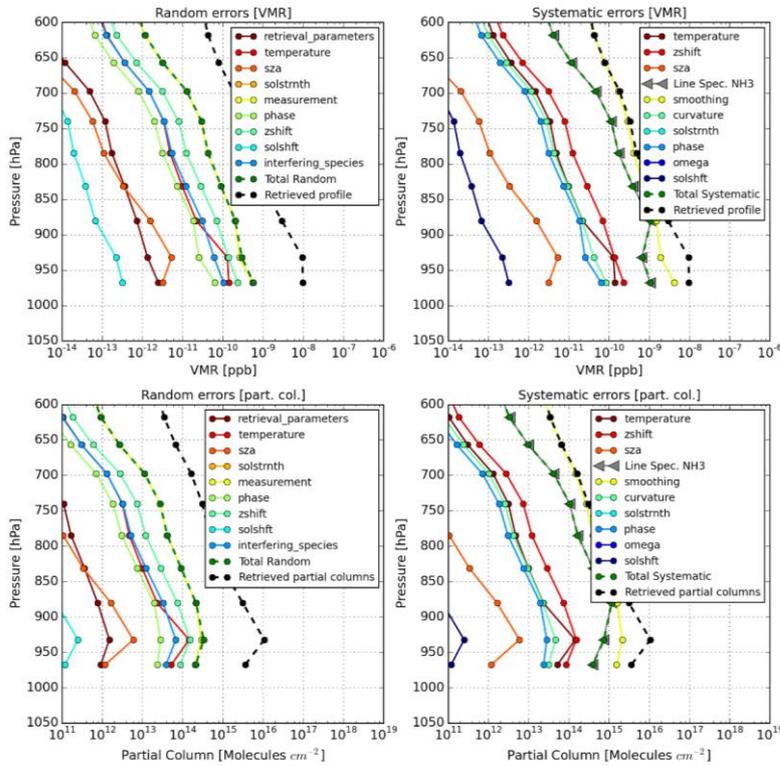
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FTIR Error summary: 20130709: Pasadena



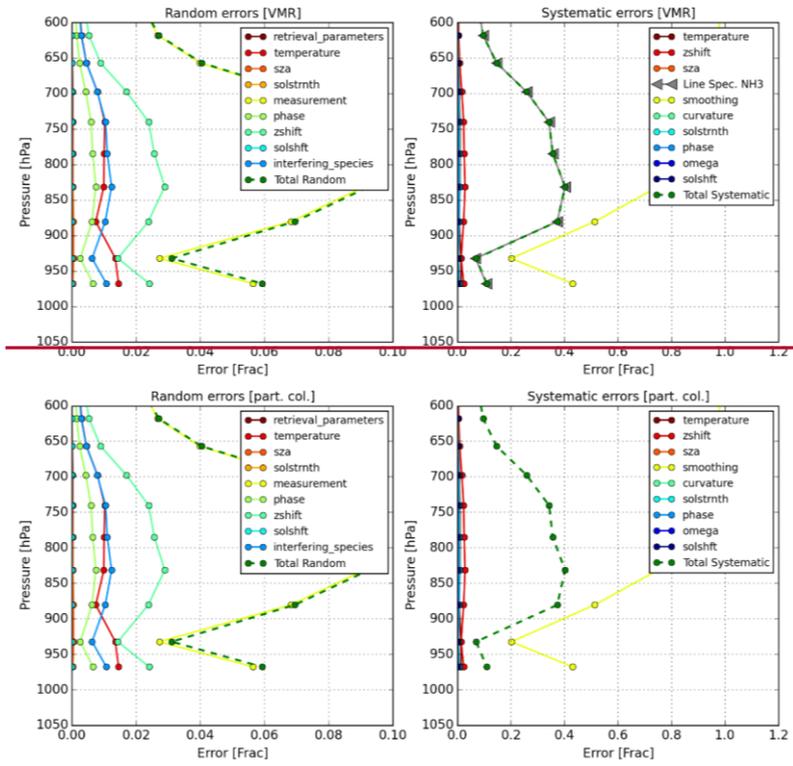
597 7.

598 **Figure A1A2.** Error profiles for each of the error terms. **Top-**The left panels show the **random-Random** errors,
 599 **bottom-the right** panels the **systematic-Systematic** errors. **Left-The top** two panels show the error in VMR. **Right**
 600 **The bottom** panels show the errors in partial column layers [molecules cm⁻²]. (See Figure A-23 for the same
 601 figure but with the errors relative to the final VMR and partial columns per layer)

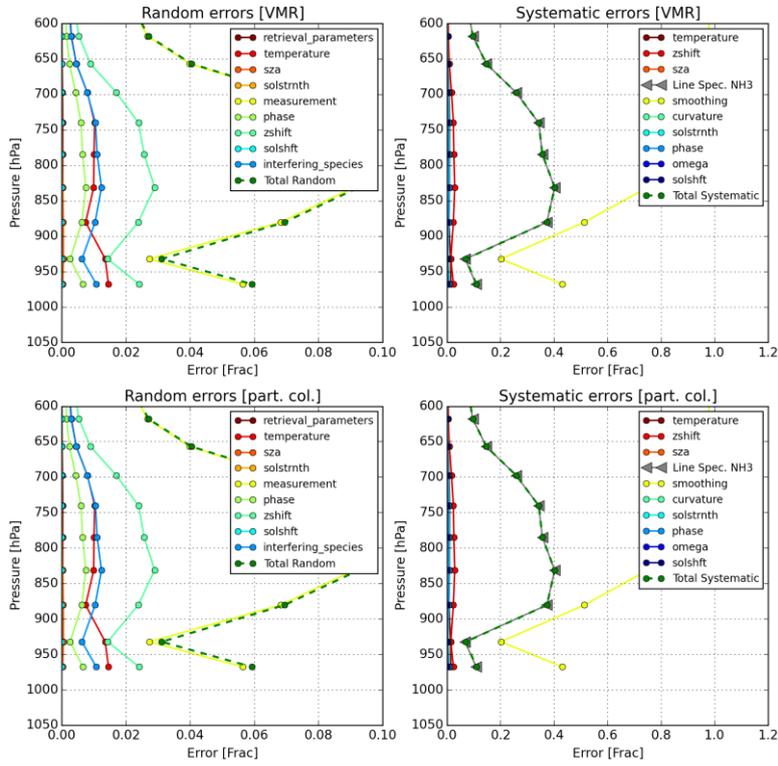
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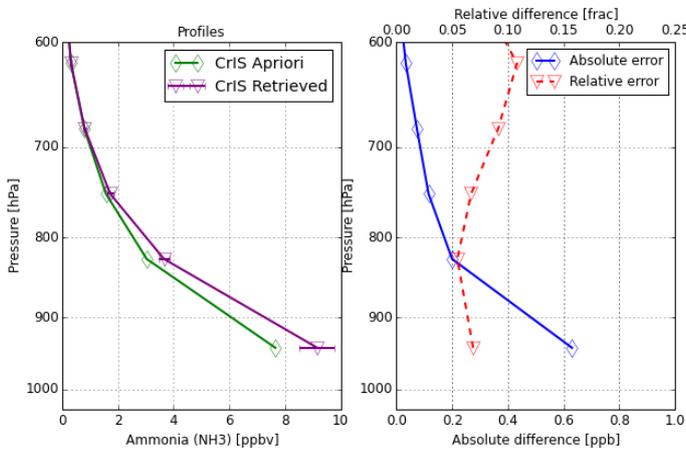


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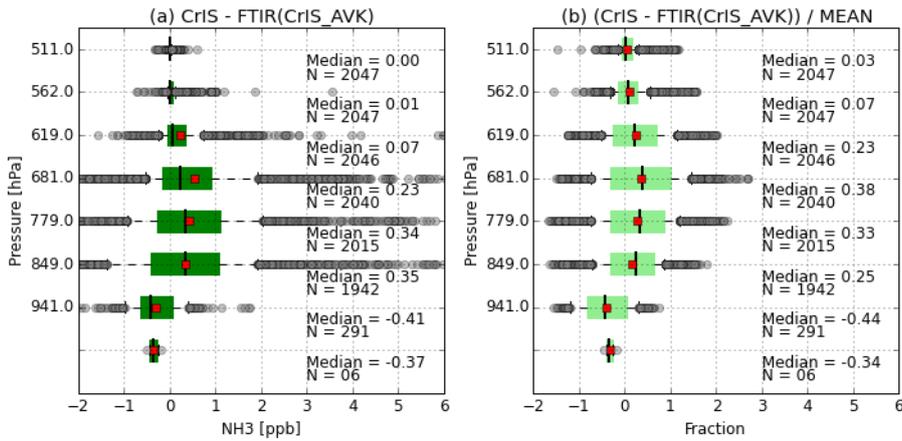


603
 604 **Figure A2A3.** Relative error profiles for each of the error terms. Top-The left panels show the Random errors,
 605 bottom-right panels the Systematic errors. Left-All four panels show the error in a fraction of the original unit
 606 used in Figure A2two panels show the error in VMR. Right panels show the errors in partial column layers
 607 [molecules cm⁻²]. (See Figure A-1-2 for the same figure but with the absolute errors)

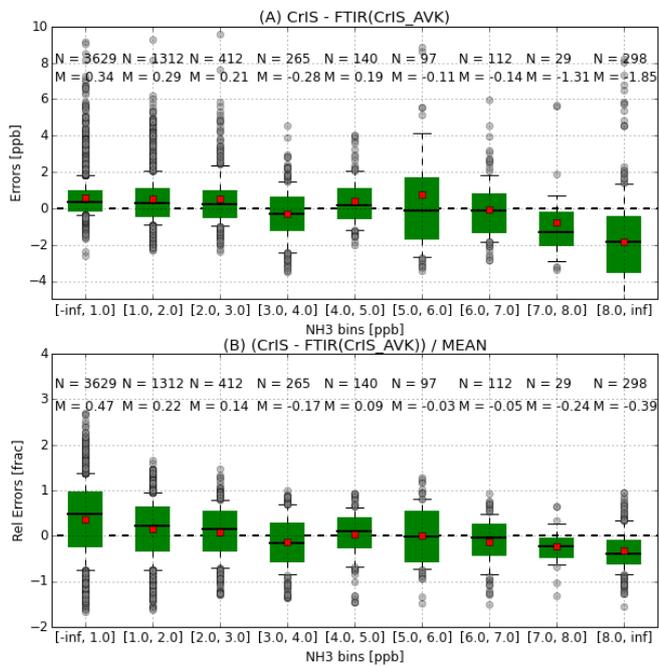
608



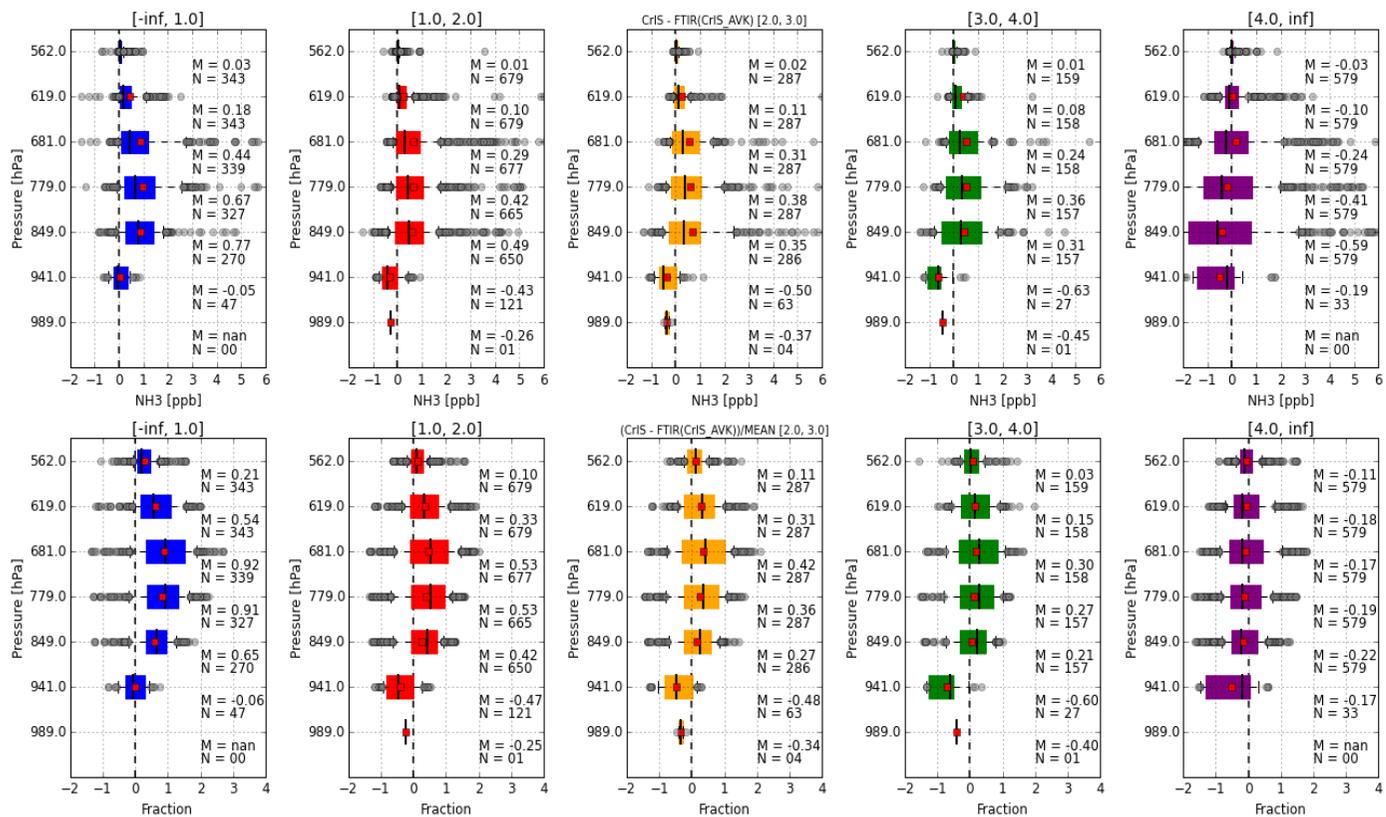
609
 610 **Figure A43.** CrIS-NH₃ relative and absolute error profile. The left plot shows the retrieved and a-priori profiles
 611 similar to the profiles shown in Figure 5c. The right panel shows the measurement error on the CrIS retrieved
 612 profile, with the blue line the absolute value and red line the value relative to the retrieved profile.
 613



614
 615 **Fig A54.** Profile comparison for all stations combined. Observations are combined following pressure “bins”,
 616 i.e. the midpoints of the CrIS pressure grid. Panel (a) shows the absolute difference [VMR] between profiles (f)
 617 and (a). Panel (b) shows the relative difference [Fraction] between the profiles in (Fig 6e) and (Fig 6b). Each
 618 of the boxes edges are the 25th and 75th percentiles, the black lines in each box is the median, the red **square**
 619 **diamond** is the mean, the whiskers are the 10th and 90th percentiles, and the grey circles are the outlier values
 620 outside the whiskers.
 621



622
 623 **Fig A56. Summary of the absolute and relative actual error** Summary of the errors as a function of the VMR of
 624 NH₃ in the individual FTIR layers. The box edges are the 25th and 75th percentiles, the black line in the box is
 625 the median, the red square diamond is the mean, the whiskers are the 10th and 90th percentiles, and the grey
 626 circles are the outlier values outside the whiskers. Only observations with a pressure greater than 650 hPa are
 627 used. The top panel shows the absolute difference for each VMR bin, the bottom panel shows the relative
 628 difference for each VMR bin.
 629
 630
 631



632
 633 **Fig A67.** Summary of actual errors as a function of VMR. The maximum VMR of each FTIR profiles is used for the classification. Absolute (Top row) and
 634 relative profile differences (bottom row) following the FTIR (CrIS AVK applied) and CrIS profiles. Observations are following pressure layers, i.e. the
 635 midpoints of the CrIS pressure grid. The box edges are the 25th and 75th percentiles, the black line in the box is the median, the red square diamond is the mean,
 636 the whiskers are the 10th and 90th percentiles, and the grey circles are the outlier values outside the whiskers.

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638

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657

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