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	O Characterization and Evaluation of AIRS-Based Estimates of the Deuterium Content of	
1) Water Vapor	
1	L	
1	Abstract: Single pixel, tropospheric retrievals of HDO and H ₂ O concentrations are retrieved	
1	from Atmospheric Infrared Sounder (AIRS) radiances using the optimal estimation algorithm	
1	developed for the Aura Tropospheric Emission Spectrometer (TES) project. We evaluate the	Deleted: These retrievals are evaluated against co-located
1	error characteristics and vertical sensitivity of AIRS measurements corresponding to five days of	TES observations taken between 2006 through 2010.
1	5 TES data (or 5 global surveys) during the <u>Northern Hemisphere</u> summers between 2006 and	
1	2010 (~600 co-located comparisons per day). We find that the retrieval characteristics of the	
1	AIRS deuterium content measurements have similar vertical resolution middle-troposphere as	Deleted: and uncertainty in the
1	TES but with slightly less sensitivity in the lower-most troposphere, with a typical degrees-of-	
2	freedom (DOFS) in the tropics of 1.5 and approximately double the uncertainty. The calculated	
2	<u>measurement</u> uncertainty is ~30 per mil (parts per thousand relative to the deuterium	
2	2 composition of ocean water) for a tropospheric average between 750 and 350 hPa, the altitude	
2	region where AIRS is most sensitive. Comparison with the TES data <u>also indicate</u> that the	Deleted: suggest
2	uncertainty of a single target AIRS HDO/H ₂ O measurement <u>is</u> ~30 per mil. Comparison of AIRS	Deleted: calculated and actual
2	and TES data between 30 degrees South and 50 degrees North indicate that the AIRS data is	Deleted: suggest
2	biased low by \sim -2.6 per mil with a latitudinal variation of \sim 7.8 per mil. This latitudinal variation	
2	is consistent with the accuracy of TES data as compared to in situ measurements, suggesting that	
2	both AIRS and TES have similar accuracy.	
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1	Introduction:	-(Formatted: Font: Times, Bold
2	Measurements of the isotopic composition of water can help identify the source of the	~(Formatted: Font: Times
3	water and provide knowledge about its condensation and evaporation history (e.g. Galewsky et	\mathbb{Z}	Formatted: Font: (Default) Times
4	al. and refs therein). Through most of the twentieth century, most isotopic measurements of		Ormatted: Normal, Indent: First line: 0.5", Space Before:) pt, After: 0 pt, Line spacing: 1.5 lines, No bullets or numbering
5	water have been of precipitation (e.g. Craig, 1961). Near global measurements of the isotopic		
6	composition of water vapor became possible with the advent of spectroscopic techniques applied		
7	to in situ measurements (e.g., Noone <i>et al.</i> , 2011) using lasers and for passive ground based and		
8	satellite measurements (e.g. Worden <i>et al.</i> 2006: Frankenberg <i>et al.</i> 2009: Schneider <i>et al.</i> 2012:		
q	Lacour et al. 2012). These data have in turn been used to evaluate the role of convection large		
10	coole dynamics, and symptromenication on the tranical water cycle (a a Worden et al. 2007).		
10	scale dynamics, and evapotranspiration on the tropical water cycle (e.g. worden et al. 2007;	0	
11	Frankenberg et al. 2009; Wright et al. 2017) tropical convection (e.g. Lacour 2018 and refs	Α	Formatted: Font: (Default) Times
12	therein) and the role of plants on global evapotranspiration (Good et al. 2015),		Formatted: Font: (Default) Times
13	In this paper we demonstrate a retrieval algorithm, based upon the Aura TES optimal	A	Deleted: n
14	estimation retrieval algorithm (e.g. Worden <i>et al.</i> 2012) that can provide robustly characterized	1/2	Deleted: Earth Science Data Record (
1-	estimation retrieval algoritam (e.g. worden et al. 2012) that can provide rootsty enabled	λ	Deleted:)
15	measurements of the deuterium content of water vapor (HDO and H_2O) from the AIRS	$\langle \chi \rangle$	Deleted: , as
16	measurements. Our goal is to create a multi-decadal Earth Science Data Record (ESDR) using	$\langle \rangle$	Deleted: Worden <i>et al.</i> 2007a; Frankenberg <i>et al.</i> 2010; Risi
17	the AIRS and TES data; the TES global record spans ~6 years (2005-2010) and the AIRS data	/}	Paleted: Wright et al. 2017
18	span 17+ years starting in 2002. This ESDR could potentially be used for evaluating the	\sum_{k}	Deleted: ; Wright <i>et al.</i> 2017
19	changing water cycle (e.g. Bailey et al., 2017) and its coupling to the carbon cycle (e.g. Zhou et	À	Deleted: The algorithm we use also jointly retrieves methane
20	al., 2014 <u>: Wright et al., 2017</u>).		from the AIRS radiances (e.g. Xiong <i>et al</i> . 2008, 2013) and we will use these methane retrievals for the purpose of
21	We first characterize the vertical resolution and uncertainties for estimates of HDO and		creating a joint AIRS / TES record of CH, and for quantifying lower-tropospheric methane (Worden <i>et al.</i> ,
22	H.O. and their ratio using AIRS radiance observations corresponding to boreal summertime TES		2015) by combining these data with total column methane neasurements (Worden <i>et al.</i> 2013) Worden <i>et al.</i> 2017)
22	global survey's between 2006 through 2010, which is the time period when TES observations		However, the evaluation and validation of these methane
23	ground survey s between 2000 model 2010, when is the time period when TES boservations	$\langle \rangle$	Deleted: H ₂ O
24	sample the (hear) global atmosphere and the calibration approach for TES measurements	Ì	Deleted: and quantify
25	remained the same. We make only these comparisons due to current processing limitations but	/	Deleted: A follow-on paper will compare these AIRS data to
26	expect additional overlap between TES and AIRS data sets in the coming years. We then		aircraft data taken during the NASA ORACLES ObseRvations of Aerosols above CLouds and their
27	compare the AIRS and TES data to evaluate the calculated uncertainties of the AIRS data.	ļ	ntEractionS) campaign.
28		·····(Formatted: Font: Not Bold
29	1)Description of AIRS and TES instruments	1	Deleted: 2)
30		\mathbf{i}	Formatted: Font: Times, Bold
31	The AIRS instrument is a nadir-viewing, scanning infrared spectrometer (Aumann et al.		Numbering Style: 1, 2, 3, + Start at: 1 + Alignment: Left + Aligned at: 0.25" + Indent at: 0.5"
32	2003; Pagano et al., 2003; Irion et al. 2018; DeSouza-Machado et al. 2018) that is onboard the	(Deleted: 3

	1	NASA Aqua satellite and was launched in 2002. AIRS measures the thermal radiance between		
	2	approximately 3-12 microns with a resolving power of approximately 1200. For the 8 micron		
	3	spectral range used for the HDO/HO retrievals, the spectral resolution is ~1 cm ⁺ with a gridding		
	4	of ~0.5 cm ⁺ ; the signal-to-noise (SNR) ranges from ~400 to ~1000 over the 8 micron region for a		
	5	typical tropical scene. A single footprint has a diameter of ~ 15 km in the nadir; with the ~ 1650		
	6	km swath, the AIRS instrument can measure nearly the whole globe in a single day. The Aqua		
	7	satellite is part of the "A-Train" that consists of multiple satellites, including TES, in a sun-		
	8	synchronous orbit at 705 km with an approximately 1:30 pm equator crossing-time.		
	9	The Aura TES instrument is a Fourier Transform Spectrometer that originally was	(Deleted: The A-Train also consists of
1	10	designed to measure the thermal infrared (IR) radiances both in the limb and nadir viewing in		Deleted: t
I	11	order to obtain vertically resolved trace gas profiles of ozone, CO, CH ₄ , HDO and H ₂ O, and) (Deleted: which
	12	several ozone pre-cursors such as ammonia, methanol, and PAN (e.g. Beer et al., 2001; Worden		Deleted: .
	13	et al., 2004; Worden et al. 2006; Luo et al., 2007; Beer et al. 2008; Worden et al., 2012; Payne et	(Deleted: ;
	14	al. 2014). Several of these trace gases, such as CO, CH, and ammonia have also been quantified		
	15	using AIRS radiances (e.g. McMillan et al., 2005; Xiong et al. 2008; Warner et al., 2016). In		
	16	comparison to the AIRS instrument, TES has a spectral resolution of ~0.12 cm ⁺ (apodized) with		
	17	a spectral gridding of 0.06 cm ⁻¹ . The SNR is ~300 in the 8 microns spectral region. The Aura TES		
1	18	instrument, after the summer of 2005, observes one nadir scene every 100 km along the orbit		
	19	path. The effective length of the record is approximately five years, between September 2005		
	20	through November 2009, after which instrument degradation problems resulted in interrupts and		
	21	a decrease in sampling. The AIRS instrument has nearly one thousand times the sampling of TES		
	22	and near continuous operation between 2002 through the present and therefore can be used to		
	23	construct several composition based ESDR's,	(Deleted: Earth System Data Records (
1	24		(Deleted:)
1	25	3) Description of the Radiative Transfer Forward Model		
1	26			
	27	The radiative transfer forward model used for this work is the Optimal Spectral Sampling		
	28	(OSS) fast radiative transfer model (RTM) (Moncet et al., 2015; Moncet et al., 2008). The OSS		
	29	approach is integrated in the operational Cross-Track Infrared Sounder (CrIS, Han et al. 2013)		
	30	processing system (Divarkala et al., 2014) and has also been utilized for trace gas retrievals from	(Deleted: Moncet et al., 2005;
	31	CrIS (e.g. Shephard and Cady-Pereira, 2015). OSS uses a series of approximations tailored to a	< (Deleted: [
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1	specific frequency range and spectral resolution to increase the radiative transfer calculation	
2	performance by approximately a factor of 20-100 relative to a line-by-line calculation	
3	(http://rtweb.aer.com). OSS can be trained to user-defined accuracy relative to the line-by-line	
4	model used for training. Here, the training threshold was set to 20 % of the AIRS noise level.	
5	The line-by-line model used as a reference in the training and to build the absorption coefficient	
6	look-up tables (LUTs) used by the fast RTM is the Line-By-Line Radiative Transfer Model	
7	(LBLRTM) (Clough et al., 2005; Alvarado et al., 2013). The OSS version used in this work is	
8	based on LBLRTM v12.4, using the TES_v2.0 spectroscopic line parameter database. The	
9	TES_v2.0 line parameter database follows the HITRAN 2012 compilation (Rothman et al.,	Deleted: [
10	2013]), with the following exceptions:	
11	• H ₀ positions and intensities are taken from the aer_v_3.4 line parameter database	
12	(http://rtweb.aer.com), closely following the measured and calculated values published in	
13	Coudert et al. (2008).	Deleted: [
14	• CH ₄ includes first order line mixing coefficients (as supplied in the aer_v_3.4 line	Deleted:]
15	parameter database). These were calculated using the approach of Tran et al. (2006),	Deleted: [
16	• CO ₂ line parameters are from the database of Lamouroux et al. (2015), This database	Deleted:]
17	takes most of its line positions, intensities, and lower state energies from the HITRAN	Deleted: [
18	2012 database, but the values for air-broadening half-widths and their temperature	
19	dependences are adjusted from the HITRAN 2012 values to be consistent	
20	throughout the bands, and the air-induced pressure shifts (not given for a majority of	Deleted: pressure shifts
21	transitions in HITRAN 2012) were added. The TES_v2.0 database includes first order	
22	line mixing coefficients (as supplied in the aer_v_3.4.1 line parameter database),	
23	calculated using the software of Lamouroux et al. (2015),	Deleted: [
24	Further information on the AER line parameter databases can be found at http://rtweb.aer.com.	Deleted:]
25	OSS is adapted for use with AIRS radiances using the version 4 AIRS spectral response function	Deleted: a version of the v4
26	(SRF) (Strow et al., 2003) that is interpolated to a uniform grid of 0.004 cm ⁴ centered on the	Deleted: [
27	channel center frequencies. The OSS radiative transfer code provides speedup of 20-100x over	Deleted:]
28	the original TES operational radiation transfer model (Clough <i>et al.</i> , 2006).	
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30 4) Description of <u>the</u> Retrieval Approach

1		
2	The optimal estimation algorithm used in this analysis for quantifying CH ₄ , HDO, H ₂ O,	
3	temperature, cloud properties, and emissivity is extensively discussed in Worden et al. (2004),	
4	Bowman et al. (2006), and Worden et al (2012). We therefore refer the reader to those papers	
5	for a description of the retrieval algorithm, with a suggestion that they start with the Worden et	
6	al. (2012) paper: however, we will briefly summarize the retrieval approach here. This retrieval	 Deleted: ,
7	algorithm, now called the MUlti-SpEctra, MUlti-SpEcies, MUlti-Sensors (MUSES) algorithm	Moved (insertion) [1]
8	(Worden et al., 2007b; Fu et al., 2013, 2016, 2018; Luo et al., 2013; Worden et al., 2013), can	(Deleted: and
9	use radiances from multiple instruments including TES, CrIS, OMI, OMPS, TROPOMI, and	
10	MLS to quantify geophysical observables that affect the corresponding radiance.	 Deleted: and characterize
11	For the AIRS retrievals discussed here, we simultaneously estimate not just CH. CO	 Deleted: appropriate for
12	HDO and HO but also temperature (surface and atmosphere), emissivity (if over land) and a	Deleted: Briefly, w
12	spectrally varying gray body aloud (a.g. Kulawik et al. 2006 Eldering et al. 2008). As in	(Deleted: e
15	spectrally varying gray body cloud (e.g. Kulawik <i>et al.</i> , 2000, Eldering <i>et al.</i> , 2008). As in	
14	worden <i>et al.</i> (2006) and worden <i>et al.</i> (2012) the constraint matrix used to regularize the HDO	
15	and H ₂ O components of the retrieval includes off-diagonal components that reflect <i>a priori</i>	
16	knowledge about the variability of HDO with respect to H ₂ O in order to ensure that <u>retrieval of</u>	
17	the ratio of HDO to H ₂ O is optimized, as opposed to either HDO or H ₂ O alone. The prior	
18	information used for this covariance is derived from monthly climatologies using the NCAR	
19	Global Climate Model as discussed in Worden et al. (2006). The a priori profile used for the	
20	HDO/H ₂ O ratio is set to be constant over the whole globe, and represents the mean tropical a	
21	priori profile from the NCAR model. However, the H ₂ O a priori profile is allowed to vary by	
22	latitude and is based on re-analysis (Worden et al. 2006); therefore the HDO profile is the mean	
23	tropical profile of the HDO/HO ratio from the NCAR model multiplied by the HO a priori	
24	profile,	 Moved up [1]: This retrieval algorithm, now called the
25	We use single pixel radiances that have not been transformed through "cloud clearing" in	MUlti-SpEctra, MUlti-SpEcies, MUlti-Sensors (MUSES) algorithm (Worden et al., 2007b; Fu et al., 2013, 2016, 2018;
26	order to preserve the original, well characterized radiance noise characteristics for use in our	Luo et al., 2013; Worden et al., 2013), can use radiances from multiple instruments including TES, CrIS, OMI,
27	estimates (Irion et al. 2018; DeSouza-Machado et al. 2018) and because we find that single-pixel	OMPS, TROPOMI, and MLS to quantify and characterize geophysical observables appropriate for the corresponding
28	AIRS radiances have sufficient information about cloud pressure and optical depth to be	radiance.
29	retrieved jointly with the trace gases, as demonstrated empirically through validation of these	
30	AIRS-based composition retrievals with TES retrievals (e.g. Figures 1-4). We assume the noise	
31	in any given pixel is uncorrelated with those from adjacent pixels. However, these correlations	
1		

1	are known to exist (e.g. Pagano et al. 2008) and the impact of ignoring them is that our		
2	calculated uncertainties will be larger than expected and therefore our noise related uncertainty		
3	should be considered a conservative estimate.		
4	A primary difference between the retrieval approach shown in this paper versus the TES		Deleted:
5	methane and HDO retrievals (Worden et al., 2012) and those from previous efforts using AIRS		
6	radiances (e.g. Xiong et al., 2008) is that we retrieve these trace gas profiles using the AIRS		Deleted: s
7	radiances from ~ 8 and ~ 12 microns instead of radiances from the 8 micron region alone in order	(Deleted: es
8	to provide a stronger constraint on atmospheric temperature and hence reduce uncertainty from	(Deleted: to
0	knowledge of temperature on the HDO and HO ratriaval. The 8 micron ration used (-1217 to		
9	$\frac{1}{1015}$		
10	1315 cm ⁴) for these retrievals has the most sensitivity to HDO and H ₂ O whereas the 12 micron		
11	band (~650 to 900 cm ³) is primarily sensitive to temperature and H ₂ O. All channels are used		
12	within this spectra unless flagged as poor during calibration		Deleted: , with the caveat that the 9.6 micron ozone band is
13			Deleted: Adding the AIRS measured radiance at 12 microns
14	5) Characterization of HDO/H ₂ O profiles		is critical for reducing uncertainties in the CH estimate, which is discussed in a subsequent paper.
15		``	1 1 1
16	While H ₂ O is quantified using radiances from both the 12 micron and 8 micron spectral		
17	regions, the primary absorption lines used here to quantify HDO are in the 8 micron region.		
18	There are other HDO (and H _i O) lines available to use from the AIRS radiance but for now we		Deleted: •
19	only use the 8 micron region to ensure consistency between AIRS and TES data. Figure 1 shows		
20	the 8 micron radiance (top panel) and the Jacobian, or sensitivity of the radiance to variations in		
21	the (log) H.O and (log) HDO respectively (middle and bottom panels). These Jacobians are		Deleted: column
22	normalized by the instrument noise. For example, a value of 1 means that it would take a 100%		Deleted: Jacobians for
23	change in the corresponding species to distinguish between two similar radiances (everything		
24	about the observed scene and radiance is the same except for the species of interest) above the	Å	Deleted: with both the H ₂ O and HDO Jacobians indicating
25	noise level. A value of \sim -50 therefore means that only a 2% variation is required (or 1/50).		sufficient sensitivity above the noise level of the radiances to variations of better than 1% in H ₂ O and HDO, because most
26	Figure 2 shows the averaging kernel matrix for the HDO component of the joint retrieval.		values are approximately -100 or better.
27	The averaging kernel describes the response of the estimate or $\log(HDO)$ relative to variations	4	Deleted: partial derivative of the
27	in the two states concernently it can also be used to exclusive the exclusion of the land between the states and the states the state of the state o	L L	Deleted: the
28	in the true state; consequently it can also be used to evaluate the vertical resolution and	Â	Deleted: /H.O
29	sensitivity of the estimate. For example, if HDO varies by 100% at 908 hPa, then the AIRS	(Deleted: ratio
30	estimate would be able to observe about 30% of the variability because the averaging kernel is	λ	Deleted: .
31	approximately 0.3 at that level, The averaging kernel at 908 hPa also depends on the deuterium	\leq	Deleted: In addition,
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1	content at several other pressure levels below and above, indicating that the estimate at 908 hPa		Deleted: ,
2	depends on the deuterium content variations at these other levels. Not shown are the		Deleted: also
3	dependencies of the (log) HDO estimate to those from the (log) HO estimate. These		
4	dependencies between the HDO averaging kernels and with the HO averaging kernels are		
5	accounted for when constructing the HDO/HO ratio; however a residual uncertainty called the		
6	"smoothing" error is imparted when comparing the HDO/HO ratio to independent data; this		
7	smoothing error is part of the error budget shown in Figure 3. As discussed in Worden et al.		Deleted: .
8	(2012) and Schneider et al. (2012), the sensitivity of the estimated HDO/H ₂ O ratio is limited by		
9	the sensitivity of the estimate to HDO. Users of these data should note that this ratio is typically		
10	used with that of H ₂ O in order to better evaluate their joint variation (HDO/H ₂ O, H ₂ O) against		
11	simple mixing and rainfall models (Noone et al. 2011). However, the sensitivity of the radiance		
12	to H ₂ O variations is much stronger than that for HDO, although the altitude region of the HDO		
13	sensitivity typically overlaps with the H _. O sensitivity. Schneider et al. (2012) discusses how to		
14	created HDO/H2O, H2O pairs to mitigate this component of the smoothing error when comparing		
15	these data against the simple models described in Noone et al. (2011). For comparison to more		
16	complex global climate models the user of these data also needs to apply the HDO and H.O.		
17	averaging kernels to the corresponding model fields (e.g. Risi et al., 2012).		
18	Figure 3 (top panel) shows the tropospheric deuterium content (or HDO/H ₂ O ratio)		Deleted:
19	derived from AIRS observations on July 1 2006. Despite the improved computational		
20	performance of the OSS radiative transfer calculation relative to the TES algorithm line-by-line		
21	calculation (Clough et al. 2005), the retrieval is still sufficiently expensive such that we can only		
22	process a sub-set of the AIRS retrievals. Considering the computational cost, for the purpose of		
23	constructing a record we currently only process AIRS retrievals from between 45 degrees South		
24	to 65 degrees North that coincide with the nearest TES observation but with an additional two		
25	observations within 100 km of the TES track over the continents; this ad hoc sampling strategy is		
26	based on experience with previous studies using the TES deuterium and methane measurements.		
27	The traditional notation for this quantity is called "delta-D", or "δ-D" with units of "per mil" or		Deleted: e
28	parts per thousand relative to the Standard Mean Ocean Water (SMOW) deuterium content	\square	Deleted: units
29	which is 3.11x10 ⁺ molecules of HDO per molecule of H ₂ O. The observations shown represent		Deleted: of Deleted: are
30	the deuterium content for the pressures between 750 hPa and 350 hPa, where we find the AIRS)
31	and TES observations have maximal overlap in their vertical resolution.		

1	The errors are calculated during the optimal estimation retrieval (Bowman et al. 2007;			
2	Worden et al. 2012) and depend on the expected noise of the AIRS radiances and the parameters			
3	that are co-retrieved with the AIRS HDO/H ₂ O ratio such as temperature, surface emissivity,			
4	clouds, and methane. As noted in Worden et al. (2012) these co-retrieved parameters affect both			
5	the precision and accuracy whereas the noise only affects the precision. The total error (middle			
6	panel) is given in units of per mil and ranges between 25 to 30 per mil, The DOFS, or trace of	(Deleted: across the	globe
7	the averaging kernel, are shown in the bottom panel and indicate that many of the HDO/HO		Deleted: the	
8	retrievals can resolve different parts of the troposphere, at least in the tropics, because (as	(Deleted:	
9	demonstrated in Figure 2) the rows of the averaging kernels are separated between the boundary			
10	layer region (surface to ~750 hPa) and the free-troposphere (~600 to 300 hPa). However, these			
11	observations cannot completely resolve the total variability in these two regions of the			
12	atmosphere because the total DOFS is typically 1.5 or less and for the measurement to be able to	(Deleted: approximat	ely
13	resolve the variability (to within the calculated error) of the two regions there would need to be			
14	at least 2 DOFS.			
15				
16	6) Comparison of AIRS and TES HDO/H ₂ O retrievals			
17				
18	Figure 4 shows a comparison between overlapping AIRS and TES estimates of the			
19	HDO/H ₂ O ratio for June 1 2006. The AIRS and TES measurements effectively overlap in space			
20	and within a few seconds in time as the instruments are in the same orbit. However not all the			
21	comparisons shown in Figure 4 overlap as retrievals may be rejected due to poor quality. We			
22	therefore compare all data that are within 200 km in the free troposphere. We do not expect			
23	substantive error to occur due to spatial mismatch of 2 degrees or less because air parcels in the			
24	free-troposphere have length scales that are several hundred kilometers long (e.g. Worden et al.			
25	2013). The average between approximately 750 hPa and 350 hPa are shown for when the DOFS			
26	are larger than one for this altitude region. There is a slight bias of -2.7 +/- 1.5 per mil between			
27	TES and AIRS as shown in the top panel. The calculated and actual (RMS difference between			
28	AIRS and TES) uncertainties are shown and are approximately 30 per mil, primarily driven by			
29	the uncertainty in the AIRS based estimates as the TES based estimates have an uncertainty of			
30	approximately 15 per mil. Figure 5 shows a direct comparison of the AIRS and TES data. The			
31	correlation is about 0.89 and the one-to-one line (solid line) overlaps this distribution. However			

 distribution in the sensitivity depends on the amount of HDO and hence we should expect differences between the TES and AIRS deuterium measurements for these lower-sensitivity retrievals., A comparison of the AIRS and TES HDO/H₂O ratio for five single global surveys taken between 2006 and 2010 (one global survey per year during boreal summer) is shown in Table 1 and indicates that the overall bias varies between -2.7 to 3.7 per mil, Using all 5 TES global surveys that are summarized in Table 1 we can construct how AIRS and TES compare as a function of latitude as shown in Figure 6 is constructed by averaging the difference between TES and AIRS observations within 5 degree latitudinal bins. The mean bias across latitudes is ~-2.6 per mil. The error bars shown on the differences divided by the square root of the number of co-located observations; as this error bar is a measure of precision for each latitude bin, this comparison demonstrates that there are variations in the comparison that are larger than the 	ited during optimal Worden <i>et al.</i> of the AIRS etrieved with the y, surface d in Worden <i>et al.</i> it both the only affects the
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14 <u>comparison demonstrates that there are variations in the comparison that are larger than the</u>	
15 precision and are therefore related to systematic errors in either the TES data or AIRS data or	
16 both. Variations in these systematic errors can be seen in the Jatitudinal variability, which has an Deleted: The RMS of the	
17 <u>RMS variation of ~7.8 per mil for the different latitude bins but can vary by as much as ~-15 to</u> Deleted: is	
18 ~+15 per mil in the tropics. Typically these variations are due to a combination of uncertainties	
19 in the spectroscopy along with temperature, water vapor, and surface properties; they may also	
20 be due to "smoothing error" which is related to how differences in the vertical resolution affect	
21 the tropospheric average of the deuterium content shown in these figures (e.g. Worden <i>et al.</i>	
22 2004). This 7.8 per mil variation across latitudes is about the same as the reported accuracy of Deleted: which is similar to the mean var	iations shown in
23 the Aura TES delta-d observations that are based on comparisons of TES data with surface and Deleted: the	
24 aircraft measurements (Worden <i>et al.</i> 2011; Herman <i>et al.</i> 2014). We therefore <u>report</u> the <u>current</u> Deleted: take	
25 accuracy of the AIRS data to be ~7.8 per mil. We expect future comparisons between these data	
26 and those from aircraft or revisions to the AIRS retrieval approach will modify this estimate of	
27 the accuracy, Deleted: These comparisons therefore sh	ow that the AIRS
28 estimates of the HDO/H _Q ratio are robus and actual uncertainties are consistent wit	t as the calculated h an accuracy
29	•
30 8) Conclusion	

1	This paper describes the vertical resolution and error characteristics of retrievals of the		
2	deuterium content (or the HDO/H.Q. ratio) of water vapor using AIRS radiances and then	(Deleted: and
3	evaluates the consistency between AIRS and TES retrievals of HDO and H ₂ O. We find that the		Deleted: H _. O
4	AIRS and TES deuterium content for the lower-troposphere (750 - 350 hPa) are consistent, or		
5	within their calculated uncertainties, for the 5 year period in which TES observations span the		
6	globe (2006-2010). We find the total uncertainty for a single AIRS observation is ~30 per mil		
7	with an accuracy of ~7.8 per mil. These uncertainties can be compared to the observed total		
8	variability, which can range from approximately -350 to -50 per mil over the whole globe, as		Deleted: 0
9	observed by the Aura TES data (Worden et al. 2006) and shown in Figure 3 for AIRS data,		Deleted: 0
10	While only five days of comparisons are shown here for the purpose of evaluating the		Deleted: .
11	retrieval approach and error characteristics of these AIRS retrievals, we expect to produce a		
12	record of the AIRS-based deuterium content retrievals from the start of the mission (2002)		
13	through the present. Because of computational limitations, we expect to process data from 45		
14	degrees South to 65 degrees North at approximately four times the sampling of the Aura TES		
15	measurements and with increased sampling (~3x) over the continental regions with the goal of		
16	increasing this sampling once the initial record is completed and as additional resources become		
17	available.		
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20	Acknowledgements		
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22 23 24	The research was carried out at the Jet Propulsion Laboratory, California Institute of Technology, under a contract with the National Aeronautics and Space Administration.		
26			Deleted: 1 Table 1: Comparison between averaged TES and AIRS HDO/H ₂ O ratio (750-350 hPa). The units are in parts per thousand relative to Standard Mean Ocean Water. The last

row shows the average and RMS for the mean differences in the far right column.⁶ Date ... [1]

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1	References		
2	Alvarado, M. J., Payne, V. H., Mlawer, E. J., Uymin, G., Shephard, M. W., Cady-Pereira, K. E.,		Formatted: Font: Times
3	Delamere, J. S. and Moncet, J. L .: Performance of the Line-By-Line Radiative Transfer Model		
4	(LBLRTM) for temperature, water vapor, and trace gas retrievals: recent updates evaluated with		
5	IASI case studies, Atmospheric Chemistry and Physics, 13(14), 6687-6711, doi:10.5194/acp-13-		
6	<u>6687-2013-supplement, 2013.</u>		Deleted: Alvarado, M. J., V. H. Payne, E. J. Mlawer, G.
7 8 9 10 11 12	Aumann, H. H., Chahine, M. T., Gautier, C., Goldberg, M. D., Kalnay, E., McMillin, L. M., Revercomb, H., Rosenkranz, P. W., Smith, W. L., Staelin, D. H., Strow, L. L. and Susskind, J.: AIRS/AMSU/HSB on the aqua mission: design, science objectives, data products, and processing systems, IEEE TRANSACTIONS ON GEOSCIENCE AND REMOTE SENSING, 41(2), 253–264, doi:10.1109/TGRS.2002.808356, 2003.		Uymin, M. W. Shephard, K. E. Cady-Pereira, J. Delamere, and J-L. Moncet: Performance of the line-by-line radiative transfer model (LBLRTM) for temperature, water vapor, and trace gas retrievals: Recent updates evaluated with IASI case studies. Atmos. Chem. Phys. Discuss., 13, 79–144, doi: 10.5194/acpd-13-79-2013
13			
14 15 16 17	Bailey, A., Blossey, P. N., Noone, D., Nusbaumer, J. and Wood, R.: Detecting shifts in tropical moisture imbalances with satellite-derived isotope ratios in water vapor, Journal of Geophysical Research-Atmospheres, 122(11), 5763–5779, doi:10.1029/2010JD015197, 2017.		
18	Beer, R., Glavich, T. A. and Rider, D. M.: Tropospheric emission spectrometer for the Earth		
19	Observing System's Aura satellite, Applied optics, 40(15), 2356–2367, 2001.		
20			
21	Beer, R., Shephard, M. W., Kulawik, S. S., Clough, S. A., Eldering, A., Bowman, K. W., Sander,		Formatted: Font: Times
22	S. P., Fisher, B. M., Payne, V. H., Luo, M., Osterman, G. B. and Worden, J. R.: First satellite		
23	observations of lower tropospheric ammonia and methanol, Geophysical Research Letters, 35(9),		
24	<u>L09801, doi:10.1029/2008GL033642, 2008.</u>		Deleted: Beer, R., Shephard, M. W., Kulawik, S. S., Clough, S. A. Eldering, A. Bowman, K. W. Sander, S. P. Eisher, B.
25 26 27 28 29 30	Bowman, K. W., Rodgers, C. D., Kulawik, S. S., Worden, J., Sarkissian, E., Osterman, G., Steck, T., Lou, M., Eldering, A. and Shephard, M.: Tropospheric emission spectrometer: Retrieval method and error analysis, IEEE TRANSACTIONS ON GEOSCIENCE AND REMOTE SENSING, 44(5), 1297–1307, 2006.		S. A., Ederling, A., Bownian, K. W., Sahed, S. T., Hshel, B. M., Payne, V. H., Luo, M., Osterman, G. B. and Worden, J. R.: First satellite observations of lower tropospheric ammonia and methanol, Geophys. Res. Lett, 35(9), L09801, doi:10.1029/2008GL033642, 2008.
30	Clough S A M W Shephard F I Mlawer I S Delamere M I Jacono K Cady-Pereira S		
32	Boukabara and P. D. Brown: Atmospheric radiative transfer modeling: A summary of the AER		
33	codes. J. Quant. Spectroscopic. Radiative Transfer. 91, 233–244. doi:		Deleted: t
34	10.1016/i.josrt.2004.05.058.2005	-	Deleted
35	551 ,		(binnu)
36	Clough, S. A., Shephard, M. W., Worden, J., Brown, P. D., Worden, H. M., Luo, M., Rodgers,		
37	C. D., Rinsland, C. P., Goldman, A. and Brown, L.: Forward model and Jacobians for		
38	tropospheric emission spectrometer retrievals, IEEE TRANSACTIONS ON GEOSCIENCE		
39	AND REMOTE SENSING, 44(5), 1308–1323, 2006.		
40			
41	Craig, H.: Isotopic variations in meteoric waters, Science, 133, 1702–1703, 1961.		Formatted: Font: Times
42			
43	Coudert, L, Wagner, G., Birk, M., Baranov, Y.I., Lafferty, W.J., Flaud, J-M.: The (H ₂ O)-O-16		Field Code Changed
44	molecule: Line position and line intensity analyses up to the second triad, Journal Molecular,		Deleted: .
45	Spectroscopy., 251 , 339-357, 2008		Deleted: .
46			~

DeSouza-Machado, S., Strow, L. L., Tangborn, A., Huang, X., Chen, X., Liu, X., Wu, W. and 1 Yang, Q.: Single-footprint retrievals for AIRS using a fast Two_Slab cloud-representation model 2 3 and the SARTA all-sky infrared radiative transfer algorithm, Atmospheric, Measurement, 4 Techniques, 11(1), 529-550, doi:10.1029/2005GL023211, 2018. 5 Divakarla, M. G., and Coauthors: The CrIMSS EDR algorithm: Characterization, optimization, 6 7 and validation. J. Geophysical Research-Atmospheres, 119, 4953-4977, doi: 10.1002/2013JD020438 (2014) 8 9 Deleted: 10 Eldering, A., Kulawik, S. S., Worden, J., Bowman, K. and Osterman, G.: Implementation of cloud retrievals for TES atmospheric retrievals: 2. Characterization of cloud top pressure and 11 12 effective optical depth retrievals, Journal of Geophysical Research-Atmospheres, 113(D16), D16S37, doi:10.1029/2007JD008858, 2008. 13 14 Frankenberg, C., Yoshimura, K., Warneke, T., Aben, I., Butz, A., Deutscher, N., Griffith, D., 15 Hase, F., Notholt, J., Schneider, M., Schrijver, H. and Rockmann, T.: Dynamic Processes 16 17 Governing Lower-Tropospheric HDO/H2O Ratios as Observed from Space and Ground, 18 Science, 325(5946), 1374-1377, doi:10.1126/science.1173791, 2009. 19 20 Frankenberg, C., Wunch, D., Toon, G., Risi, C., Scheepmaker, R., Lee, J.-E., Wennberg, P. and 21 Worden, J .: Water vapor isotopologue retrievals from high-resolution GOSAT shortwave infrared spectra, Atmospheric Measurement Techniques, 6(2), 263-274, doi:10.5194/amt-6-263-22 23 2013, 2013. 24 25 Fu, D., Kulawik, S. S., Miyazaki, K., Bowman, K. W., Worden, J. R., Eldering, A., Livesey, N. 26 J., Teixeira, J., Irion, F. W., Herman, R. L., Osterman, G. B., Liu, X., Levelt, P. F., Thompson, A. M. and Luo, M.: Retrievals of tropospheric ozone profiles from the synergism of AIRS and 27 28 OMI: methodology and validation, Atmospheric Measurement Techniques, 11(10), 5587-5605, 29 doi:10.5194/amt-11-5587-2018-supplement, 2018. 30 31 Fu, D., Bowman, K. W., Worden, H. M., Natraj, V., Worden, J. R., Yu, S., Veefkind, P., Aben, 32 I., Landgraf, J., Strow, L. and Han, Y.: High-resolution tropospheric carbon monoxide profiles retrieved from CrIS and TROPOMI, Atmospheric Measurement Techniques, 9(6), 2567-2579, 33 34 doi:10.5194/amt-9-2567-2016-supplement, 2016. 35 Fu, D., Worden, J. R., Liu, X., Kulawik, S. S., Bowman, K. W. and Natraj, V.: Characterization 36 37 of ozone profiles derived from Aura TES and OMI radiances, Atmospheric Chemistry and Physics, 13(6), 3445-3462, doi:10.5194/acp-13-3445-2013, 2013. 38 39 40 Galewsky, J., Larsen, H. S., Field, R. D., Worden, J. R., Risi, C. and Schneider, M.: Stable 41 isotopes in atmospheric water vapor and applications to the hydrologic cycle, Rev. Geophys., 42 doi:10.1002/2015RG000512, 2016. 43 44 Good, S. P., Noone, D. and Bowen, G.: Hydrologic connectivity constrains partitioning of global 45 terrestrial water fluxes, Science, 349(6244), 175-177, doi:10.1126/science.aaa5931, 2015. 46

Deleted: .	
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Deleted: .	
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Deleted: Galewsky, J., Larsen, H. S., Field, R. D., Worden, J. R., Risi, C. and Schneider, M .: Stable isotopes in atmospheric water vapor and applications to the hydrologic cycle, Rev. Geophys., doi:10.1002/2015RG000512, 2016.

Formatted: Font: Times

Deleted: Eldering, A., S. S. Kulawik, J. Worden, K. Bowman, and G. Osterman (2008), Implementation of cloud retrievals for TES atmospheric retrievals: 2. Characterization of cloud top pressure and effective optical depth retrievals, J. Geophys. Res, 113(D16), D16S37, doi:10.1029/2007JD008858.

Formatted: Font: Times

Deleted: Frankenberg, C., Wunch, D., Toon, G., Risi, C., Scheepmaker, R., Lee, J.-E., Wennberg, P. and Worden, J .: Water vapor isotopologue retrievals from high-resolution GOSAT shortwave infrared spectra, Atmos. Meas. Tech., 6(2), 263-274, doi:10.5194/amt-6-263-2013, 2013.

Formatted: Font: Times

Deleted: Fu, D., Kulawik, S. S., Miyazaki, K., Bowman, K. W., Worden, J. R., Eldering, A., Livesey, N. J., Teixeira, J., Irion, F. W., Herman, R. L., Osterman, G. B., Liu, X., Levelt, P. F., Thompson, A. M., and Luo, M.: Retrievals of Tropospheric Ozone Profiles from the Synergic Observation of AIRS and OMI: Methodology and Validation, Atmos. Meas. Tech. Discuss., https://doi.org/10.5194/amt-2018-138, in review, 2018.

Formatted: Font: Times

Deleted: Fu, D., Bowman, K. W., Worden, H. M., Natraj, V., Worden, J. R., Yu, S., Veefkind, P., Aben, I., Landgraf, J., Strow, L., and Han, Y.: High-resolution tropospheric carbon monoxide profiles retrieved from CrIS and TROPOMI, Atmos. Meas. Tech., 9, 2567-2579, https://doi.org/10.5194/amt-9-2567-2016, 2016.

Formatted: Font: Times

Deleted: Fu, D., Worden, J. R., Liu, X., Kulawik, S. S., Bowman, K. W., and Natraj, V .: Characterization of ozone profiles derived from Aura TES and OMI radiances, Atmos. Chem. Phys., 13, 3445-3462, https://doi.org/10.5194/acp-13-3445-2013, 2013.

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1	Han V Bayaraamh H. Cramp M. Cu D. Jahrson D. Maanay D. Saatt D. Straw J.		Esumettadi Eauti Timor
1 2	Ringham G. Borg J. Chen Y. DeSlover D. Forlin M. Hagan D. Jin Y. Kruteson P.		rormatted; ront: 11mes
2	Motteler H Predina I Suwinski I Taylor I Tohin D Tremblay D Wang C Wang I		
4	Wang I and Zavvalov V Suomi NPP CrIS measurements sensor data record algorithm		
5	calibration and validation activities and record data quality. Journal of Geophysical Research		
6	Atmospheres 118(22) 12 734–12 748 doi:10 1002/2013ID020457 2013		Deleted: Han X et al : Suomi NPP CrIS measurements
7	<u>Autospheres</u> , 110(22), 12, 151–12, 10, doi:10.1002/2015/D020157, 2015		sensor data record algorithm, calibration and validation
8	Herman R.I. Cherry J.E. Young J. Welker J.M. Noone, D. Kulawik S.S. and Worden	$\langle \cdot \rangle$	activities, and record data quality. J. Geophys. Res. Atmos.,
9	L. Aircraft validation of Aura Tropospheric Emission Spectrometer retrievals of HDO / H2O	$\langle \rangle \langle$	118, 12 734–12 748, doi:10.1002/2013JD020344
10	Atmospheric Measurement Techniques 120 3127-3138 doi:10.5194/amt-7-3127-2014 2014	/ 1	Formatted: Font: Times
11		\mathcal{N}	Formatted: Font: Times
12	Irion F. W. Kahn B. H. Schreier, M. M. Fetzer, E. J. Fishbein, E. Fu, D. Kalmus, P. Wilson	WV	Formatted: Font: Times, 12 pt
13	R. C., Wong, S. and Yue, O.: Single-footprint retrievals of temperature, water vapor and cloud	$\langle \rangle$	Formatted: Font: Times
14	properties from AIRS. Atmospheric Measurement Techniques, 11(2), 971–995.	$\left(\right)$	Deleted: Herman, R. L., J. E. Cherry, J. Young, J. M.
15	doi:10.1117/12.615244.2018.		Welker, D. Noone, S. S. Kulawik, and J. Worden (2014),
16	·· · · · · · · · · · · · · · · · ·	$\langle \rangle \rangle$	Spectrometer retrievals of HDO / H ₂ O, <i>Atmos. Meas. Tech.</i> ,
17	Kulawik, S. S., Worden, J., Eldering, A., Bowman, K., Gunson, M., Osterman, G. B., Zhang, L.,	/	120, 3127–3138, doi:10.5194/amt-7-3127-2014.
18	Clough, S. A., Shephard, M. W. and Beer, R.: Implementation of cloud retrievals for		Formatted: Font: 12 pt
19	Tropospheric Emission Spectrometer (TES) atmospheric retrievals: part 1. Description and		Formatted: Font: Times
20	characterization of errors on trace gas retrievals, Journal of Geophysical Research-Atmospheres,	N N	Deleted: Irion, F. W., Kahn, B. H., Schreier, M. M., Fetzer,
21	111, D24204, doi:10.1029/2005JD006733, 2006.	N.	E. J., Fishbein, E., Fu, D., Kalmus, P., Wilson, R. C., Wong,
22		11	S. and Yue, Q.: Single-footprint retrievals of temperature,
23	Kulawik, S. S., Bowman, K. W., Luo, M., Rodgers, C. D. and Jourdain, L.: Impact of		Tech., 11(2), 971–995, doi:10.1117/12.615244, 2018.
24	nonlinearity on changing the a priori of trace gas profile estimates from the Tropospheric	- //	Formatted: Font: Times
25	Emission Spectrometer (TES), Atmospheric Chemistry Physics, 8, 3081–3092, 2008.	/	Deleted: I. Geophys. Res
26			Ditetta: J. Geophys. Res
27	Kulawik, S. S., Worden, J. R., Wofsy, S. C., Biraud, S. C., Nassar, R., Jones, D. B. A., Olsen, E.		
28	T., Jiménez, R., Park, S., Santoni, G. W., Daube, B. C., Pittman, J. V., Stephens, B. B., Kort, E.		
29	A., Osterman, G. B.TES team: Comparison of improved Aura Tropospheric Emission		
30	Spectrometer CO ₂ with HIPPO and SGP aircraft profile measurements, Atmospheric Chemistry		
31	and Physics, 13(6), 3205–3225, doi:10.5194/acp-13-3205-2013, 2013.		
32			
33	Lacour, J. L., Risi, C., clarisse, L., Bony, S., hurtmans, D., Clerbaux, C. and Coheur, PF.: Mid-		Formatted: Font: Times
34	tropospheric δD observations from IASI/MetOp at high spatial and temporal resolution,		
35	Atmospheric Chemistry and Physics, 12(22), 10817–10832, doi:10.5194/acp-12-10817-2012,		
36	<u>2012.</u>		
37			
38	Lacour, JL., Risi, C., Worden, J., Clerbaux, C. and Coheur, PF.: Importance of depth and		Formatted: Font: Times
39	intensity of convection on the isotopic composition of water vapor as seen from IASI and TES δ		
40	D observations, Earth and Planetary Science Letters, 481, 387–394,		
41	<u>doi:10.1016/j.epsl.2017.10.048, 2018.</u>		
42		1	Deleted: .
43	Lamouroux, J., Régalia, L., Thomas, X., Vander Auwera, J., Gamache, R. R., Hartmann, J-M.:	1/	Deleted: t.
44	CO2 line-mixing database and software update and its tests in the 2.1 μ m and 4.3 μ m regions,		(Deleted: .
45	Journal Quantitative Spectroscopy Radiative Transfer, 151, 88-96, 2015	<u> </u>	Deleted: .
46			

Deleted:

1	Luo, M., Rinsland, C. P., Rodgers, C. D., Logan, J. A., Worden, H., Kulawik, S., Eldering, A.,	
2	Goldman, A., Shephard, M. W., Gunson, M. and Lampel, M.: Comparison of carbon monoxide	
3	measurements by TES and MOPITT: Influence of a priori data and instrument characteristics on	
4	nadir atmospheric species retrievals, Journal of Geophysical Research-Atmospheres, 112(D9),	Formatted: Font: Times
5	D09303, doi:10.1029/2006JD007663, 2007.	Deleted: J. Geophys. Res
6		Deleted: ¶
7	Moncet, J-L., G. Uymin, A. E. Lipton, and H. E. Snell: Infrared radiance modeling by optimal	Moncet, J-L. et al: Algorithm theoretical basis document for
8	spectral sampling. J. Atmos. Sci., 65, 3917–3934, doi: 10.1175/2008JAS2711.1, 2008	Environmental Data Records (EDR), version 4.2, AER Tech.
9		Doc. P1187-TR-I-08, 298 pp. [Available online at]
10	Moncet, J-L., Uymin, G., Liang, P. and Lipton, A.E: Fast and accurate radiative transfer in the	http://npp.gsfc.nasa.gov/sciencedocuments/2013-01/474-
11	thermal regime by simultaneous optimal spectral sampling over all channels. Journal of the	00050_KevAbasenne.pdi.j, 2005j
12	Atmospheric Sciences, vol 72, 2622-2641, doi: 0.1175/JAS-D-14-0190.1, 2015	
13	Noone D. Galewsky I. Sharp 7 D. Worden I. Barnes I. Baer D. Bailey A. Brown D. P.	Deleted: Noone D et al. (2011). Properties of air mass
14	Christensen I. Crosson F. Dong F. Hurley, I.V. Johnson I. R. Strong M. Toohey, D.	mixing and humidity in the subtropics from
15	Van Pelt, A. and Wright, I. S.: Properties of air mass mixing and humidity in the subtropics from	measurements of the D/H isotope ratio of water vapor at
16	measurements of the D/H isotope ratio of water vapor at the Mauna Loa Observatory, Journal	the Mauna Loa Observatory, J. Geophys. Res, 116(D22), D22113, doi:10.1029/2011JD015773.
17	Geophysical Research-Atmospheres, 116(D22), D22113, doi:10.1029/2011JD015773, 2011.	Formatted: Font: 12 pt
10		Formatted: Font: Times
18	Pagano, I. S., Aumann, H. H., Hagan, D. E., and Overoye, K.: Prelaunch and in-flight	Formatted: Font: 12 pt, Not Bold
20	Geoscience and Remote Sensing 41 265-273 2003	Formatted: Font: Times, 12 pt, Not Bold
20	Coscience and Remote Sensing, 41, 203–213, 2005.	Deleted: IEEE T. Geosci. Remote
21	Pagano, T. S., Aumann, H. H., Schindler, R., Elliott, D., Broberg, S., Overoye, K., and Weiler, M. H.:	Formatted: Font: 12 pt, Not Bold
22	Absolute radiometric calibration accuracy of the Atmospheric Infrared Sounder (AIRS), in: Proc.	Formatted: Pattern: Clear (White)
23	<u>SPIE, vol. 7081, doi:10.1117/12.795445, 2008.</u>	Formatted: Font: Times New Roman
24	Daura V.H. Alvarada M.I. Cadu Daraira K.E. Wardan J.D. Kulawik S.S. and Ficabar F.	
24	Y : Satallite observations of peroxyacetyl nitrate from the Auro Tronospheric Emission	
25	Spectrometer Atmospheric Measurement Techniques 7(11) 3737 3740 doi:10.5104/amt 7	Formettad: Font: Times
27	3737-2014 2014	Deleted: Atmas Mass Task
28	5757 2011, 2011.	Deleted: Aunos. Meas. Tech.
29	Rodgers, C. D. and Connor, B. L.: Intercomparison of remote sounding instruments. Journal of	
30	Geophysical Research-Atmospheres, 108, 4116, doi:10.1029/2002JD002299, 2003.	
31		
32	Risi, C., Noone, D., Worden, J., Frankenberg, C., Stiller, G., Kiefer, M., Funke, B., Walker, K.,	
33	Bernath, P., Schneider, M., Wunch, D., Sherlock, V., Deutscher, N., Griffith, D., Wennberg, P.	
34	O., Strong, K., Smale, D., Mahieu, E., Barthlott, S., Hase, F., García, O., Notholt, J., Warneke,	
35	T., Toon, G., Sayres, D., Bony, S., Lee, J., Brown, D., Uemura, R. and Sturm, C.: Process-	
36	evaluation of tropospheric humidity simulated by general circulation models using water vapor	
37	isotopologues: 1. Comparison between models and observations, Journal of Geophysical	Formatted: Font: Times
38	Research-Atmospheres, 117(D5), D05303, doi:10.1029/2011JD016621, 2012.	Deleted: J. Geophys. Res
39		
40	Risi, C., Noone, D., Frankenberg, C. and Worden, J.: Role of continental recycling in	Formatted: Font: Times
41	intraseasonal variations of continental moisture as deduced from model simulations and water	

1 2 3	vapor isotopic measurements, Water Resour. Res., 49(7), 4136–4156, doi:10.1002/wrcr.20312, 2013.		
4 5 6	Rothman, L. S., I. E. Gordon, Y. Babikov et al., "The HITRAN 2012 Molecular Spectroscopic Database", <i>Journal of Quantitative Spectroscopy and Radiative Transfer</i> 130 , 4-50 (2013)		
7 8 9	Schneider, M., Barthlott, S., Hase, F., González, Y., Yoshimura, K., García, O. E., Sepúlveda, E., Gomez-Pelaez, A., Gisi, M., Kohlhepp, R., Dohe, S., Blumenstock, T., Wiegele, A., Christner, F. Strong K. Weaver, D. Palm M. Deutscher, N. M. Warneke, T. Notholt, J. Leieune, B.		
10	Demoulin, P., Jones, N., Griffith, D. W. T., Smale, D. and Robinson, J.: Ground-based remote		
11	sensing of tropospheric water vapour isotopologues within the project MUSICA, Atmospheric		Formatted: Font: Times
12	Measurement Techniques, 5(12), 3007–3027, doi:10.5194/amt-5-3007-2012, 2012.		Deleted: Atmos. Meas. Tech.
13			
14	Shephard, M. W., Worden, H. M., Cady-Pereira, K. E., Lampel, M., Luo, M., Bowman, K. W.,		
15	Sarkissian, E., Beer, R., Rider, D. M., Tobin, D. C., Revercomb, H. E., Fisher, B. M., Tremblay,		
16	D., Clough, S. A., Osterman, G. B. and Gunson, M.: Tropospheric Emission Spectrometer nadir		
17	spectral radiance comparisons, Journal of Geophysical Research-Atmospheres, 113(D15),		Deleted: J. Geophys. Res
18	D15S05, doi:10.1029/2007JD008856, 2008.		Formatted: Font: Times
19			
20	Shephard, M. W. and Cady-Pereira, K. E.: Cross-track Infrared Sounder (CrIS) satellite		Formatted: Font: Times
21	observations of tropospheric ammonia, Atmospheric Measurement Techniques, 8(3), 1323-1336,		
22	doi:10.5194/acpd-15-4823-2015, 2015,	*****	Deleted: Shephard, M. W., and <u>K. Cady-Pereira Cross-track</u>
23	Strow, L. L., Hannon, S., Weiler, M., Overoye, K., Gaiser, S. L., Aumann, H. H.: Prelaunch		Infrared Sounder (CrIS) satellite observations of tropospheric ammonia Atmos Meas Tech 8 1323-1336
24	spectral calibration of the atmospheric infrared sounder (AIRS), IEEE Transactions Geoscience		doi:10.5194/amt-8-1323-2015, 2015.
25	and Remote Sensing, 41(2), 274-286, doi: <u>10.1109/TGRS.2002.808245</u> , 2003		Deleted: .
26		$\langle \rangle$	Deleted: .
27	Tran, H., <u>Flaud</u> J-M, Gabard, T., Hase, F., Von Clarmann, T., Camy-Peyret, C., Payan, S.,	111	Deleted
28	Hartmann, J-M.: Model, software, and database for line-mixing effects in the n, and n, bands of	())	Deleted: 10 1100/ECDS 2002 808245
29	CH_4 and tests using laboratory and planetary measurements. I. N_2 (and air) broadenings and the	$\langle \rangle$	Deleted: 10.1109/10K3.2002.606245,
30	Earth atmosphere, Journal Quantitative, Spectroscopy, Radiative, Transfer, 101, 284-303, 2000	~ 1	Formatted: Font: Times
21	Worden I Kulawik S. Shenard M. Claugh S. Worden H. Bowman K. and Coldman A.	111	Deleted: Flaud,
22	Predicted errors of tropospheric emission spectrometer padir retrievals from spectral window		Deleted: .
34	selection Journal of Geophysical Research-Atmospheres 109(D9) D09308	() (Deleted: .
35	doi:10.1029/2004ID004522_2004	////	Deleted:
36	<u>aantonobsizoo isboo isb</u>		Deleted: t
37	Worden, J., Bowman, K., Noone, D., Beer, R., Clough, S., Eldering, A., Fisher, B., Goldman, A.,	//l	Deleted:
38	Gunson, M., Herman, R., Kulawik, S. S., Lampel, M., Luo, M., Osterman, G., Rinsland, C.,	$\langle \rangle \rangle$	Deleted: f.
39	Rodgers, C., Sander, S., Shephard, M. and Worden, H.: Tropospheric Emission Spectrometer	//	Formatted: Font: Times
40	observations of the tropospheric HDO/H 2O ratio: Estimation approach and characterization,	/	Deleted: Worden I Kulawik S Shepard M Clough S
41 42	Journal of Geophysical Research-Atmospheres, 111(D16), doi:10.1029/2005JD006606, 2006.		Worden, H., Bowman, K. and Goldman, A.: Predicted errors of tropospheric emission spectrometer nadir retrievals from spectral window selection, J. Geophys. Res. 109(D9).
43	worden, J., Noone, D., Bowman, K., Beer, R., Eldering, A., Fisher, B., Gunson, M., Goldman,	11	D09308, doi:10.1029/2004JD004522, 2004.
44	A., Herman, K., Kulawik, S. S., Lampel, M., Osterman, G., Kinsland, C., Rodgers, C., Sander,		Formatted: Font: Times
45	5., Snephard, M., webster, C. K. and Worden, H.: Importance of rain evaporation and	/	Deleted: J. Geophys. Res

 continental convection in the tropical water cycle, Nature, 445(71) doi:10.1038/nature05508, 2007a. 	27), 528–532,
3 4 Worden I. D. Neene, I. Geleweltz, A. Beiley, K. Beyrman, D. Bi	Deleted:
4 worden, J., D. Noone, J. Galewsky, A. Daney, K. Bowman, D. Bi 5 Los and M. Strong (2011). Estimate of bigs in Auro TES HDO/H	Owil, J. Hurley, S. Kulawik, J.
5 Lee, and W. Subing (2011), Estimate of bias in Auta TES HDO/II	Pointes from comparison
6 of TES and in situ HDO/H2O measurements at the Mauna Loa ob	Servalory, Almospheric Formatted: Font: 12 pt
7 Chemistry and Physics, 11(9), 4491–4503, doi:10.5194/acp-11-44	91-2011.
 Worden, J., Kulawik, S., Frankenberg, C., Payne, V., Bowman, K Lee, JE. and Noone, D.: Profiles of CH4, HDO, H2O, and N2O 	, Cady-Peirara, K., Wecht, K., vith improved lower
11 tropospheric vertical resolution from Aura TES radiances, Atmos	heric Measurement (Formatted: Font: Times
12 <u>Techniques</u> , 5(2), 397–411, doi:10.5194/amt-5-397-2012, 2012.	Deleted: Atmos. Meas. Tech.
13	
14 Worden, J., Wecht, K., Frankenberg, C., Alvarado, M., Bowman,	K., Kort, E., Kulawik, S., Lee, Formatted: Font: Times
15 <u>M., Payne, V. and Worden, H.: CH4 and CO distributions over tro</u>	pical fires during October
16 2006 observed by the Aura TES satellite instrument and modeled	by GEOS-Chem, Atmospheric
$\frac{17}{10} = \frac{1}{1000} \frac{1}{100$	<u>79-2015, 2015.</u>
10 19 Worden I.B. Turner A.I. Bloom A. Kulawik S.S. Liu I.I.	e M. Weidner R. Bowman
20 K Frankenberg C Parker R and Pavne V H : Quantifying lov	ver tronospheric methane
concentrations using GOSAT near-IR and TES thermal IR measure	ements Atmospheric Formatted: Font: Times
22 Measurement Techniques , 8(8), 3433–3445, doi:10.5194/amt-8-3	433-2015, 2015. Deleted: Atmos Meas Tech
23	Detecti Athlos Model Ton.
24 Wright, J. S., Fu, R., Worden, J. R., Chakraborty, S., Clinton, N.,	Risi, C., sun, Y. and Yin, L.:
25 Rainforest-initiated wet season onset over the southern Amazon, H	proceedings of the National
26 academy of Sciences, 1–6, doi:10.1073/pnas.1621516114/-/DCSu	pplemental, 2017.
27	Formatted: Font: Times
28 Xiong, X., Barnet, C., Maddy, E., Sweeney, C., Liu, X., Zhou, L.	and Goldberg, M.:
29 <u>Characterization and validation of methane products from the Atn</u>	iospheric Infrared Sounder
30 (AIRS), Journal Geophysical Research-Biogeosciences, 113, doi:	<u>U.1029/2007/JG000500, 2008.</u> Lin J. Zhou, and M. Goldberg (2008). Characterization and
31 22 - Viene V. Demet C. Meddy F. Wefey S. C. Chen I. A. Kewi	validation of methane products from the Atmospheric
 Along, A., Barnet, C., Maduy, E., Wolsy, S. C., Chen, L. A., Karl Detection of methane depletion associated with stratespheric intru 	Infrared Sounder (AIRS), J. Geophys. Res, 113(null),
sounder (AIRS) Geophysical Research Letters 40(10) 2455–245	9 doi:10.1002/gtl 50476
35 2013	Formatted: Font: Times
36	Deleted: Xiong, X., C. Barnet, E. Maddy, S. C. Wofsy, L. A Chen A Karion and C Sweeney (2013) Detection of
Xiong, X., Barnet, C., Maddy, E., Sweeney, C., Liu, X., Zhou, L.	and Goldberg, M.: methane depletion associated with stratospheric intrusion
38 Characterization and validation of methane products from the Atn	by atmospheric infrared sounder (AIRS), Geophys. Res.
39 (AIRS), J. Geophys. Res, 113(null), G00A01, doi:10.1029/2007JC	2000500, 2008.
40	Jeieteu;Page Dreak
41 Zhou, L., Tian, Y., Myneni, R. B., Ciais, P., Saatchi, S., Liu, Y. Y	., Piao, S., Chen, H., Vermote, Formatted: Font: Times
42 E. F., Song, C. and Hwang, T.: Widespread decline of Congo rain	Forest greenness in the past
decade, Nature, 508(7498), 86–90, doi:10.1038/nature13265, 2014	ł.
44	Formatted: Font: Times
45	Deleted: ¶

1 <u>Table 1: Comparison between averaged TES and AIRS HDO/H₂O ratio (750-350 hPa). The units</u>

2 are in parts per thousand relative to Standard Mean Ocean Water. The second column shows the

- 3 expected RMS based on the uncertainties of the TES and AIRS data. The third column shows the
- 4 actual RMS difference between TES and AIRS. The last column shows the mean difference

Date	Expected RMS	Actual RMS	Mean (TES-AIRS)
	(per mil / SMOW)	(per mil / SMOW)	(per mil / SMOW)
<u>2006-06-01</u>	<u>31.1</u>	<u>30.6</u>	<u>-2.7 +/- 1.5</u>
2007-06-02	<u>30.0</u>	<u>31.9</u>	<u>-0.6 +/- 1.5</u>
<u>2008-06-02</u>	<u>31.5</u>	<u>29.3</u>	<u>0.5 +/- 1.4</u>
2009-07-06	<u>31.6</u>	<u>27.1</u>	<u>0.7 +/- 1.4</u>
2010-06-02	<u>31.6</u>	<u>28.2</u>	<u>3.7 +/- 1.2</u>

5 6

(Moved (insertion) [2]

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Figure 1: (top) AIRS radiance at approximately 8 microns for a typical tropical scene. (middle) The total column (log) Jacobian for H₂O normalized by the AIRS NESR. (bottom) The total column (log) Jacobian for HDO normalized by the AIRS NESR.







Figure 2: The rows of the averaging kernel matrix for the HDO retrieval corresponding to the radiance shown in Figure 1. The different colors and symbols are <u>used</u> to indicate the pressure levels corresponding to each row of the averaging kernel matrix.



from AIRS radiance measurements. (middle) The total error for the measurements in the top panel (also in units of per mil relative to SMOW). (bottom) The DOFS for the retrieval.









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I would like to greatly thank both reviewers for their detailed review and comments. Reviewing and 1 2 editing these papers is quite a bit of (effectively voluntary) work and both reviews really went into 3 great detail on fixing the presentation. With respect to the comment on "minimalism", I actually was 4 striving for minimalism in this paper but apparently overshot my goal! 🙂. For example, I did not want to write (yet) another paper full of the optimal estimation description in all of its gory, (or 5 6 glory?) equation detail but instead report on the basic notions..... that the AIRS radiances can be 7 used to generation global deuterium content retrievals and that their vertical resolution is about 8 the same as TES but with slightly poorer uncertainties, and that we will soon produce a long record 9 of the data that spans most of the globe and can hopefully produce scientific awesomeness. 10 Again, thanks for the review. Below are the comments and my responses. 11 12 **Response to Reviewer 1** 13 General Comment: This study presents the application of an existing retrieval methodology of 14 HDO/H2O vertical profiles originally applied on TES, on AIRS thermal infrared measurements. The 15 authors briefly remind the retrieval methodology, describe the error and sensitivity, and show a 16 comparison with co-located TES retrievals. In my view, this is a welcome study as the capabilities of 17 AIRS sensors for HDO/H2O ratio retrievals were unknown/not tested, and the sampling characteristics of AIRS offer great potential for isotopes related studies. The manuscript is short 18 19 and generally convincing but the presentation is too minimalist and should be improved. Some 20 discussions on previous improvements in characterizing HDO/H2O-H2O pairs retrieval is missing. I 21 list a few comments which should be easily resolved by the authors. 22 **Specific Comments:** 23 Introduction: A short introduction on water isotopes, their usefulness and a description on what are the remote sensing capabilities to observe HDO/H2O ratios in the free 24 25 troposphere would be useful to strengthen the importance of this work and to smooth the 26 feeling of reading a purely technical report. 27 **Response**: I added a paragraph at the front that describes a bit of history on water isotope measurements, and how these vapor based measurements have helped address global 28 29 water / carbon questions. 30 • P2, Line 19: estimates of HDO/H2O ratios and not HDO **Response**: Added "and their ratio". We actually do retrieve HDO and H₂O seaparately even 31 32 if the retrieval setup optimizes the ratio. 33 • P2, Line 20: Why only summertime TES global survey's? Do you mean boreal summertime? 34 Response: added boreal and added a statement about current limited processing

35 <u>capabilities</u>

1 2 3	•	P2. Line 23: "We then compare the AIRS and TES data to evaluate and quantify the calculated uncertainties of the AIRS data" - To evaluate and quantify the calculated uncertainties sound a little odd. This needs to be rephrased.
4		Response: removed "and quantify"
5 6	•	This paper is relatively short and yet there is a lot of statements about futures publications (P2, L17-18:P2, L23-24:P5, L29 – P6,L8). Some of them could be removed.
7 8		Response : Removed most of these references where appropriate and modified some of the language about the utility of the 12 micron band for constraining atmospheric temperature.
9 10	•	<u>P3, L8: There is a redundancy here of the statement that TES is part of the A-Train, it was</u> just said in the previous sentence.
11		Response removed
12 13 14	•	P5, L9: "This retrieval algorithm can use radiances () to quantify and characterize geophysical observables appropriate for the corresponding radiance." – What is an appropriate geophysical observable? To retrieve different geophysical parameters?
15 16 17		Response : changed "appropriate" to "that affect". I think this wording is appropriate but terse. I can also add another line such as (e.g. the ozone concentrations affect radiances in the 9.6 micron ozone band) but that seems too wordy.
18 19	•	<u>P5, L16-17: "in order to ensure that [the retrieval of] the ratio is optimized, as opposed ()"</u> [missing]
20		Response fixed
21 22	•	<u>P5, L29 – P6,L8: All this part describes the importance of including the 12 microns</u> radiances for the methane retrieval. That is not interesting in the frame of this paper.
23		Response (fixed in above response, hopefully 🕲).
24 25	•	P6, L17-19: Jacobians have not be defined. What does the -50 treshold represent? How is it calculated?
26 27		Response: I added the definition for a Jacobian and changed the language around basically 2% is equivalent to $1/501\% = 1/100$ etc.
28 29 30	•	P6, L22: "() partial derivative of the estimate relative to [partial derivative] of the true state". Or maybe in a language more accessible to potential users not familiar with optimal estimation: "the response of the retrieved state to perturbations of the true state"
31		Response: changed

2	•	averaging kernels are for HDO.
3 4 5		Response thanks for pointing this one out I adjusted the language accordingly and added a sentence about how the information about HDO/H ₂ O is limited by HDO; I also added a reference.
6 7 8 9 10 11	•	P6, L28-29: Schneider et al., 2012 proposed an a posteriori methodology to characterize the joint retrieval of H2O and HDO. The method allows to transform the products obtained in the log(H2O).log(HDO) space into a proxy state log(H2O).&D which is very useful for characterization. Moreover, the HDO/H2O ratio product is often used in pair with H2O it is therefore important to discuss the differences of sensitivity of H2O and HDO/H2O ratios. This is missing here.
12 13 14		Response I added language on how the averaging kernels for H ₂ O are different than that for HDO and that Schneider et al discusses an approach to use these data with simple models while accounting for the different sensitivities.
15 16 17	•	P7. L13-L15: There are a lot of measurements within the tropics with DOFS between 0.5 and 1 so I wouldn't generalize this situation to the whole tropics. This might be valid only for the averaging kernels shown.
18		Response: adjusted the language now we say "many" which implies "not all"
19	•	P8, L6->L11: All this part would better fit in the error characterization part
20		Response: moved
21 22	•	Comparisons of AIRS and TES retrievals - In order to be really convincing, this part needs to be completed.
23		- Would it be possible to show a scatter plot of AIRS versus TES?
24	•	- What is the correlation between AIRS and TES retrievals?
25 26 27 28		Response : I am not that convinced this is a meaningful figure as it is the difference between AIRS and TES that can be used to determine if the AIRS data is (relatively) well characterized. Having said that I have included it here in case readers find it useful. The correlation for this day is 0.89.
29 30	•	- Because this kind of product is used in pairs with humidity retrievals it is also interesting to show that both sounders show the same humidity- δD information and not only δD .
31 32 33 34		Response : I dont think this comparison is of use to this specific paper as it shows that the pairs generated by AIRS are similar to those from TES but slightly different, as expected because the sensitivity and errors are different. For this reason I would prefer not to show here, but will show in subsequent papers when we start looking at the science!

1 <u>•</u> 2 3 4 5 <u>•</u> 6 7	 I didn't understand the error assessment reasoning. The mean bias across latitude is -2.6 permil, later on the authors assess the RMS to be 7.8 permil then the authors say the accuracy is 7.8 permil. Is this a mistake or do I miss something? The language between accuracy and precision should be clarified. What about the latitudinal variations of the bias which are greater (-15 to 15 permil) than the mean standard error? It looks like there is a latitudinal bias, could it be caused by some dependence on temperature or humidity content?
8 9 10 11	Response: I attempted to clean up the language here, hopefully it's a bit more clear! I also added language that the latitudinal variations are typically due to uncertainties in temperature, water vapor, and spectroscopy, as well as differences in the vertical resolution.
12 •	- Could you plot the data in Figure 5 until 40°S as in the previous figure?
13	Response: Done!
14 • 15 16 17 18	The conclusions could be more developed. One of the interest of this paper lies in the development of a HDO retrieval methodology from AIRS data which was unknown and opens great perspectives for users interested in such measurements. In this context, a word on the future plans of the authors on processing more AIRS data, or not, would be interesting.
19	Response: Added a paragraph on current and future plans with respect to building an ESDR.
20 •	P9, L8: Please reference the natural variability of δD
21	Response: Added statement about Figure 3 and cited a TES paper.
22	Technical corrections
23	• Abstract, L17: Northern instead of N; • P1, L28: a verb is missing (fixed)
24	• L29, degrees (fixed)
25 • 26 • 27 • 28 • 29 •	P4 . L30 : Description of Retrieval Approach -> Description of the retrieval approach (fixed) P5. L29 : (e.g. Figures 1-4). (fixed) P7. L4: add degrees to latitude (fixed) P7. L8: use the delta Greek notation δ (this is stylistic, I have added "or δ" instead) Figure 4: A legend is missing, what is TES and what is AIRS? (fixed)

1 <u>Reviewer 2</u>

2

3 <u>General Comment</u>

4 As this study is targeting the preparation of a new Earth Science Data Record covering AIRS

- 5 HDO/H2O observations, it clearly has a high scientific significance. The manuscript itself is clear
- and concise, but I would agree with Reviewer #1 that the presentation is indeed somewhat
 "minimalistic" and could be extended and improved. Please carefully follow suggestions and
- 8 comments provided by Reviewer #1 and those listed below so that the paper can be published
- 9 <u>soon.</u>

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10 Specific comments

- 11 p3, 14: The AIRS swath width is 1650 km (Aumann et al., 2003) rather than 1250 km.
- 12 <u>Response: (fixed)</u>
 - p5, 125-26: Although the AIRS noise is characterized well for individual channels, in other work I noticed noise can be spectrally correlated between neighboring channels, which is due to the 1-D linear detector arrays of AIRS sharing the same electric module (Pagano et al., 2008). This may be too specific to discuss in your paper; I just wondered if you considered this?

18 <u>Response: We have not explicitly addressed this issue and in fact the same is true with the TES</u>

<u>data because the data are apodized but the calculated errors assume the noise is random (or not apodized)</u>. On the other hand the apodization is accounted for in the TES retrievals when we

21 calculate the forward model radiance. In contrast with the TES data where we can account for

the apodiziation it is not clear how to account for these noise correlations in the AIRS data.

23 Instead we added a statement that the noise is assumed to be random as we are unable to account

24 for correlations between channels. The effect of this assumption is that the calculated errors will

- 25 be too large and is therefore a conservative estimate of the uncertainties.
- 26 Pagano, T. S., Aumann, H. H., Schindler, R., Elliott, D., Broberg, S., Overove, K., and Weiler, M. H.:
- Absolute radiometric calibration accuracy of the Atmospheric Infrared Sounder (AIRS), in: Proc.
 SPIE, vol. 7081, doi:10.1117/12.795445, 2008.
- p7, 120-29: Are the HDO retrieval results correlated with the simultaneous H2O retrievals? Does the AVK matrix show any correlations between these retrieval variables?

31 <u>Response: Yes! These correlations are addressed 1) in model comparisons by applying the</u>

- 32 averaging kernels for both HDO and H₂O to the model (e.g. Risi et al. 2013) or 2) by calculating
- 33 the resulting error (e.g. Worden *et al.* 2006) and including that in the error budget, or mitigating

34 <u>further by 3) projecting to HDO-HO pairs as discussed by the first reviewer. I have discussed the</u>

35 pairing approach based on the first reviewers comments and will add the Risi reference as well.

1 2	Fig. 2: Maybe show also the integral of the AVKs, to indicate the amount of measurement information in the retrieval results? (Fix "are used to indicate" in the caption.)
3 4	Response: Fixed caption. The integral of the AVKs (or trace actually) is shown in the bottom panel of Figure 3 and discussed in the text.
5	Fig. 4: A legend/definition of the colors used for the plot on the top seems to be missing.
6	Response: Fixed!
7 8 9 10	<u>Technical corrections</u> <u>p1, l17: N> Northern (also in other places) (fixed)</u> <u>p1, l22:reduced spectral resolution of AIRS (for clarity?) (removed sentence as it was confusing</u> <u>to have in the abstract)</u>
11 12	p1, l24: Suggest to remove reference (Worden et al., 2004) from abstract. p1, l27-28: Please fix incomplete sentence. (see above)
13 14 15	p1.129: Add degree symbols to "30 S and 50 N" (also in other places)? I have added the word "degrees" instead.
16 17 18	p2. 12: The copyright statement "All rights reserved." is not allowed in the given form, I think, please see https://www.atmospheric-measurement-techniques.net/about/licence and copyright.html for details.
19 20	Response: I have to use this copyright for JPL during the submission phase. Once / if the paper is accepted I put in another form where JPL puts in a modification to the Copernicus agreement.
21	p2. l4: "Introduction:" -> "Introduction" p3. l11: PAN> PAN
22	fixed
23	p3.122: Earth Science Data Records (ESDR's) -> ESDR's (acronym was already introduced)
24	fixed
25	p4, l25: "a version of the v4 AIRS" -> "version 4 of the AIRS" (?) (fixed)
26	<u>p5,16: will only briefly summarize (?) (fixed)</u>
27	p5, 19-10: Not sure if "appropriate for the corresponding radiance." is a good phrase here?
28	(changed, see comment from reviewer 1)
29	p6, 14-6: Remove redundant sentence. (fixed)

- p6, l31: Change date format to "1 July 2016" (also in other places); I think this is a US Versus Europe
 date thing [©]. Can I keep as is? Its like asking me to drive in the left lane [©].
- 4 <u>p7, l2: we can only (?) (fixed)</u>
- 5 6 <u>p8,114: indicate s (fixed)</u>

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Microsoft Office User