Simulation of stratospheric water vapor and trends using three reanalyses

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Abstract

The domain-filling, forward trajectory calculation model developed by Schoeberl and Dessler (2011) is extended to the 1979–2010 period. We compare results from NASA’s MERRA, NCEP’s CFSR, and ECMWF’s ERAi reanalyses with HALOE, MLS, and balloon observations. The CFSR based simulation produces a wetter stratosphere than MERRA, and ERAi produces a drier stratosphere than MERRA. We find that ERAi temperatures are cold biased compared to Singapore sondes and MERRA, which explains the ERAi result, and the CFSR grid does not resolve the cold point tropopause, which explains its relatively higher water vapor concentration. The pattern of dehydration locations is also different among the three reanalyses. ERAi dehydration pattern stretches across the Pacific while CFSR and MERRA are concentrate dehydration activity in the West Pacific. CSFR and ERAi also show less dehydration activity in the West Pacific Southern Hemisphere than MERRA. The models’ lower stratospheres tend to be dry at high northern latitudes because of too little methane-derived water appears to be descending from the middle stratosphere. Using the tropical tape recorder signal, we find that MERRA vertical ascent is 15% too weak while ERAi is 30% too strong. The models tend to reproduce the observed weakening of the 100-hPa annual cycle in zonal mean water vapor as it propagates to middle latitudes. Finally, consistent with the observations, the models show less than 0.2 ppm decade$^{-1}$ trends in water vapor both at mid-latitudes and in the tropics.

1 Introduction

The mechanisms responsible for stratospheric dehydration have been studied for more than 60 yr – since the publication of Brewer’s seminal paper (Brewer, 1949). Aside from its possible affect on stratospheric ozone loss processes (Kirk-Davidoff et al., 1999), the concentration of stratospheric water vapor may alter the climate (Forster and Shine, 1999; Solomon et al., 2010). To first order, dehydration of air occurs as
air, rising toward the stratosphere, cools, water vapor saturates, ice forms and then falls out. Dehydration thus primarily depends on the air parcel temperature history, and simulations of the stratospheric water vapor concentration ultimately depend on accurate analyses of temperatures and air parcel movement (e.g. Mote et al., 1996; Fueglistaler et al., 2005, 2009; Liu et al., 2010).

General circulation models have a difficult time reproducing the observed concentration of water vapor in the tropical upper troposphere and lower stratosphere (UTLS) (Gettleman et al., 2010). These difficulties arise from temperature biases, the representation of sub-grid scale phenomena such as convective moistening (Zipser et al., 2006; Schiller et al., 2009; Corti et al., 2008; Tzella and LeGras, 2011) and gravity wave cooling (Jensen and Pfister, 2004). In Schoeberl and Dessler (2011) (hereafter SD2011) we used a Lagrangian forward domain-filling model to simulate water vapor in the stratosphere. In forward domain filling, we continuously release parcels in the upper troposphere so that hundreds of thousands of parcels eventually fill the stratosphere and provide a statistically robust population for analysis. This approach provides a continuous picture of the time evolution of stratospheric constituents and allows us to investigate issues that are difficult to address with traditional back-trajectory approaches or with Eulerian models. For example, in the case of back trajectory or reverse domain fill trajectory calculations, thousands of separate long runs would be required to achieve the same statistical equivalence as a long forward trajectory run. For Eulerian models, the information on parcel history is lost, and standard advection schemes excessively diffuse the strong water vapor gradient at the tropopause.

This paper extends the results of SD2011 to longer periods as well as making use of the recently completed long term reanalyses by NASA, NOAA and ECMWF to produce integrations extending from 1979 to 2010. Each of these models have different biases that impact water vapor. We also compare our simulations to each other and to observations from balloons and satellites.
2 Model and observations

2.1 Reanalysis data sets and model set up

The results shown here are for diabatic trajectory calculations as described in detail in SD2011. SD2011 used both diabatic and kinematic methods and found that the diabatic trajectories produce better results compared with observations so we restrict ourselves to diabatic calculations. The advantage of diabatic trajectories to kinematic trajectories was first noted by Danielson (1961). Subsequently, Schoeberl et al. (2003), Liu et al. (2011), Ploeger et al. (2010, 2011), and others confirmed that even today kinematic trajectories driven by modern assimilation models produce the excessively dispersive vertical velocity field. In MERRA, the vertical velocities are now time averaged to suppress the noise, but even after this suppression the residual enhanced dispersion can affect the water vapor as shown in SD2011. Wohtmann and Rex (2008) suggest use of the thermodynamic equation rather than the continuity equation to produce stratospheric vertical velocities, and this is formally equivalent to using net heating rates as was done here.

Forward domain-filling works as follows: For each day we initialize a grid of parcels at 360K from ±40° latitude. The injection level is chosen to be above the zero diabatic heating level. The tropical parcels move upward into the stratosphere, filling the stratospheric domain. Parcels move downward at extra tropical latitudes and those moving below 250 hPa are removed – we assume they have re-entered the troposphere. Parcels are initiated with 50 ppmv water vapor and we dehydrate parcels to saturation values when saturation (or supersaturation) occurs. We use the temperature – water vapor saturation relationship described in Murphy and Koop (2005) to determine if parcels have reached saturation. As with most models of this type we assume that the excess water vapor is instantly removed. (We have performed experiments with slower removal as might occur with ice crystal formation and gravitational settling and found that this process has no noticeable impact on our results.) Aside from water vapor we carry methane concentration for each parcel. Methane is oxidized and the resulting...
water is added to the parcel as described in SD2011. Tropospheric methane initial values are increased from 1.54 ppmv in 1979 to 1.8 ppmv in 2010, and the oxidation rate of methane comes from a two-dimensional stratospheric chemistry model (Fleming et al., 2007). We save the location of the final dehydration point (FDP) for each parcel. The FDP has also been referred to in the literature as the Lagrangian dry point (e.g. Liu et al., 2010).

In SD2011, we used the Modern Era Retrospective-Analysis for Research and Applications (MERRA) (Bosilovich et al., 2008; Rienecker et al., 2011). We now include the NOAA’s Climate Forecast System Reanalysis (CFSR, Saha et al., 2010) and ECMWF’s ERA Interim reanalysis (ERAi, Dee et al., 2011). These three reanalyses are described in more detail at http://reanalyses.org/atmosphere/comparison-table. For each run we use identical convective moistening scheme and gravity wave scheme described in SD2011. Our ERAi data set is archived every six hours while the MERRA and CFSR are archived daily. We have recently received a six-hour MERRA data set and our preliminary analysis shows that MERRA six-hour produces a stratosphere as dry as ERAi.

2.2 Observations

Our model water vapor calculations are compared to both satellite and balloon observations. The longest high quality balloon observations are the Boulder data set described by Rosenlof et al. (2008) and more recently by Hurst et al. (2011). The Boulder data set extends back 30 yr.

Long-term satellite water vapor data sets begin with the 1984–2005 SAGE II data (Chiou, et al., 1993), but our examination of that data set suggest that only segments where the retrievals are not contaminated by El Chichon or Pinatubo aerosol are usable. Thus, we begin our satellite data set with the 1993–2005 higher precision UARS Halogen Occultation Experiment (HALOE) measurements (Evans et al., 1998) and continue with Aura Microwave Limb Sounder (MLS) data (Read et al., 2007) from 2005.
to 2010. We offset the HALOE time series to the MLS 2005 average vapor at each latitude and pressure level to the 2005 average MLS data.

### 2.3 Reanalysis tropical temperature differences

As noted above, the most important factor in controlling stratospheric water vapor in the tropical UTLS is temperature. Figure 1a–c compare the 100 hPa temperatures at Singapore [1°14’ N, 103°55’ E] in the Tropical West Pacific (TWP) with the three reanalyses. We chose Singapore because it has the longest record of observations in the TWP, and although 100-hPa level is often just below the cold point, 100 hPa is a standard reporting level for temperature and water vapor data sets. For Fig. 1, the reanalyses are interpolated to the latitude and longitude of Singapore.

The Aura Microwave Limb Sounder (MLS) measured 100 hPa winter TWP lower stratospheric water vapor concentration interpolated to the location of Singapore is ∼3 ppmv (see Read et al., 2007 for a discussion of these measurements). In this region, CH₄ oxidation does not contribute to the water vapor concentration so we may assume that water vapor is totally controlled by dehydration/hydration processes. Figure 1a shows the 1979–2010 probability distribution function (PDF) of the Singapore sonde temperatures with the three reanalyses. MERRA shows the best agreement although the mean of both CFSR and MERRA are quite close. ERAi, shows a bimodal distribution with the larger peak colder than the observations. The fluctuations in the Singapore temperatures have ∼0.7 correlation with all the reanalyses. CFSR, MERRA and ERAi are also compared at http://gmao.gsfc.nasa.gov/ref/merra/atlas/atlas.php and it is evident that ERAi is about ∼1 K colder than MERRA or CFSR for the boreal winter.

In Fig. 1b we plot the mean temperature of the three reanalyses and Singapore sondes and the saturation water vapor mixing ratio. Also plotted are the mean temperatures minus one standard deviation and the resultant water vapor mixing ratios compared to MLS. The figure shows how sensitive water vapor saturation mixing ratio is to the temperatures, ∼0.5 ppmv deg⁻¹ at these pressures. Thus even a small bias
in reanalysis temperatures can produce a significant shift in water vapor concentration (e.g. Randel et al., 2004).

2.4 Model parameters

As discussed in SD2011 we have three free parameters we can adjust to control water vapor. These parameters are: (1) supersaturation, which is frequently observed to be >100% near the tropical tropopause (Jensen et al., 2005; Kramer et al., 2009) – increasing the supersaturation will increase the water vapor concentration; (2) the magnitude of high frequency gravity wave temperature fluctuations (Jensen and Pfister, 2004) – gravity waves reduce the water vapor by increasing the probability of parcels encountering colder temperatures; (3) the convective overshooting and ice injection – this will increase the water vapor concentration on the average (e.g. Wright et al., 2011). To parameterize convective ice injection we use the scheme developed by Dessler et al. (2007).

In SD2011 we found that increasing the supersaturation level (say to 104%) increases stratospheric water vapor by about 0.1–0.25 ppmv. Likewise the gravity wave parameterization scheme reduces water vapor concentration by an equivalent amount. Convective lofting of ice (as implemented in SD2011) increases the overall water vapor concentration by about 0.5 ppmv. Although our parameterization schemes for gravity waves, supersaturation and convective lofting are loosely based upon observations, there are insufficient constraints to further isolate the roles of the individual schemes. In other words, we can simulate the observed stratospheric water vapor by increasing the supersaturation from 105 % to 120 % and increasing the gravity wave amplitude by a factor of 2. The two changes cancel each other yielding the same result and yet both changes are within the uncertainty of existing observations.

In addition to the uncertainty in the parameterization schemes, the analyses differ in their absolute temperature values near the tropopause which can have a significant impact as seen in Fig. 1b. For example, using the CFSR simulations produce a stratosphere that is ~0.5 ppmv wetter than MERRA. The main reason is that the vertical
resolution of the temperature field in the TTL is coarser in CFSR (levels are about \(\sim 2\) km apart between 16 and 21 km) than in MERRA (\(\sim 1\)-km resolution). Linearly interpolating the CFSR temperatures across the tropopause produces a statistically warmer cold point compared to MERRA even though MERRA and CFSR 100-hPa temperatures are nearly the same.

3 Results

Before we address the long-term integrations and trends in Sect. 3.2, we first revisit some of the basic results from the SD2011 model. This first section can be viewed as an expansion of SD2011 to show the differences between CFSR and ERAi along side MERRA.

3.1 Analysis comparisons

3.1.1 Mean age

The main difference between the MERRA and CFSR data sets is the spatial resolution and altitude of the top boundary. In Fig. 2 we show the mean age-of-air from trajectory calculations using the three analyses schemes, CFSR, MERRA and ERAi (hereafter we refer to the trajectory integrations using the name of the analysis). The MERRA and ERAi results show a well-defined narrow tropical pipe region (see Plumb, 2002) compared to the CFSR. The result is that much younger air is found in the CFSR lower stratospheric tropics and CFSR age-of-air is younger at extra-tropical latitudes as well.

3.1.2 Dehydration patterns

In Fig. 3, we show the dehydration point locations from the three reanalyses. To create this figure we generate a normalized PDF of dehydration points using a 9° longitude
by 2° latitude grid. We consider only dehydration events for which the trajectory experienced no further dehydration events for at least a year to make sure that they are FDPs. The dehydration patterns are similar, including the large zone of dehydration over South America, although MERRA shows higher concentration of FDPs in the Southern Hemisphere compared to CFSR and ERAi. This difference arises from MERRA’s enhanced convection over the southern part of the TWP, which is believed to be due to a problem with assimilation of the ATOVS radiances (S. Pawson, personal communication, 2011). The enhanced convection produces a colder tropopause and more frequent FDPs. MERRA also shows lower numbers of FDPs over Southeast Asia relative to the other reanalyses. CFSR and ERAi show similar patterns of dehydration except that ERAi stretches the dehydration zone across the Pacific along the ITCZ. Both ERAi and CFSR also show more dehydration occurring over SE Asia than MERRA.

The appearance of FDPs over Antarctica is due to the very cold temperatures within the Antarctic vortex. Air that enters in the tropics is further dehydrated over Antarctica, which contributes to the overall dry bias in southern hemispheric water vapor compared to the north (SD2011). This dehydration occurs during the Southern Hemisphere winter.

Figure 4 shows the mid-winter water vapor measured by MLS and computed using the three reanalyses. Each plot also shows the zonal mean temperature for each reanalysis. Table 1 shows global average water vapor from 18–28 km.

The MERRA simulation shows the best agreement to MLS while CFSR is relatively wetter and ERAi is relatively drier. The differences are not great. Note that we have adjusted the gravity wave, supersaturation and convection scheme to improve MERRA’s agreement with MLS. The other reanalyses were run with the same free parameter settings as MERRA – this means that it would be possible to tune the parameters to bring the other reanalyses into agreement with MLS. The tropopause cold region is clearly smaller in CFSR and that explains the wetter stratosphere, while ERAi has a slightly larger cold zone producing a drier stratosphere. The zone of tropical dry air between 22 and 26 km centered near 25 km in MLS data is the previous winter tape recorder
signal. In MERRA, the center of this zone is roughly at 22.5 km while in CFSR it is at 25 km and in ERAi it is at ~25.5 km. Thus MERRA has a slower tropical upward transport relative to the two other reanalyses while ERAi is slightly faster. We will address this observation more quantitatively when we discuss Fig. 8, below.

Figure 5 shows a comparison between the various model zonal mean annual, boreal winter (DJF), and boreal summer (JJA) 100-hPa water vapor concentration and MLS water vapor along with the mean age. In the Southern Hemisphere, Antarctic dehydration forces water vapor values downward for all the models, and this is also seen in the MLS observations. In the Northern Hemisphere, MLS water vapor generally increases toward the North Pole while MERRA and CFSR shows a decrease. ERAi on the other hand shows an increase from its low tropical values. If we look at the winter case (Fig. 5b) MERRA and MLS agree in the tropics but MERRA shows much less water at boreal latitudes. This difference is also apparent in summer. Overall the difference between tropical water vapor amounts and boreal amounts is less than the difference between MLS tropical and boreal values, the exception is ERAi in winter. This result suggests that insufficient methane-derived water might be descending from the stratosphere in the models – in other words, with the exception of ERAi, the high latitude descent is too weak. However, we note that the northern high-latitude mean age for MERRA is the oldest suggesting that descent in MERRA is more robust than ERAi and CFSR. Clearly, dynamics does not provide the whole explanation of the high latitude water vapor deficit.

Figure 6 shows an overall comparison between the models run with and without including methane oxidation. To make the comparisons cleaner we set the saturation to 100% and turn off the gravity wave and convection schemes. Because more methane is oxidized the longer the parcel spends in the stratosphere, water enhancement due to methane tends shown in Fig. 4 follows the age contours shown in Fig. 2. Without CH₄, (center column of Fig. 6) water vapor concentration is more featureless since there are few stratospheric sinks for water. The exception, already noted, is Antarctic dehydration zone (see SD2011). The difference plots between H₂O with CH₄ oxidation
and H₂O without CH₄ oxidation are shown in the third column of Fig. 6 and clearly delineate CH₄’s critical role in providing water at higher latitudes and also helps explains the differences between the models away from the tropics. For example, methane is providing much less water at high latitudes in CFSR than in either MERRA or ERAi. This is because CFSR’s tropical stratosphere is too ventilated so CH₄ moves to higher latitudes and descends out of the model before it can be oxidized. This is consistent with CFSR’s young air age bias (Fig. 2). On the other hand, CH₄ supplied water is significantly enhancing high latitudes in the ERAi case. This result suggests that the ERAi overturning circulation is stronger than MERRA or CFSR – and this suggestion is quantified in the next section.

3.2 Long-term integrations

In this section we describe the results of long-term integration of the model from 1979–2010 and compare the results with observations.

3.2.1 Tape recorder simulation

Figure 7 compares the tape recorder signal in the simulations to observations from a combined HALOE (1993–2005) (Evans et al., 1997) and MLS (2005–2010) time series. To combine the observations, the HALOE and MLS data have been interpolated onto a regular bi-monthly time grid and the mean difference was then removed at each level. Figure 7 shows that model does a good job reproducing the tape recorder up to about 30 km where the lack of parcels creates a noisy signal. Note that the descending QBO’s secondary circulation creates “kinks” in the tape signal around 24 km that is also reproduced in the models (Plumb and Bell, 1982; Punge et al., 2009, SD2011).

In order to better compare the observations with the models, we have computed the correlation between the observations and the model as a function of the lag between the models and measurements. This shifting allows us to quantify the phase lag/lead between the model tape recorder and observations. The results are shown in Fig. 8.
The MERRA tape recorder tends to lag the observations as altitude increases producing an ~2 month lag at higher altitudes while ERAi tends to lead the observations by about 3 months at higher altitude. The average vertical velocity in the lower tropical stratosphere region is between 0.02 and 0.03 cm s\(^{-1}\) (Schoeberl et al., 2008), and thus this result suggests that MERRA tropical upwelling circulation is about 15% too weak while ERAi is about 30% too strong.

3.2.2 Lower stratosphere/upper troposphere

In this section, we analyze the lower stratospheric/upper troposphere results and compare them to MLS observations. Figure 9 shows the time series of observations and model results at 100 hPa. The largest component of water vapor variability is the annual cycle. As part of the annual cycle, winter dry zones propagate to the extra-tropics, arriving at the poles about six months later (Randel et al., 2004). The water vapor biases compared to MLS are also quite evident. For MERRA, the extra tropics are too dry while CFSR is wet. Recall that MERRA’s tropical water vapor simulation is quite close to MLS observations. As indicated above, the extra tropical dry bias is caused by insufficient water descending from higher altitudes, where it was produced by methane oxidation as shown in Fig. 6. The importance of methane-derived water to high latitudes is further illustrated in the ERAi simulation at the bottom of Fig. 9. Note the ERAi polar maximum in water vapor that appears in midwinter detached from the tropical summer wet zones as a result of the polar descent of stratospheric water that has been enhanced by methane photolysis.

The annual cycle in 100-hPa model water vapor (Fig. 9) is complicated by the differences in the zonal means. By adjusting the zonal means for each model to the MLS zonal mean we can produce a clearer picture of the annual cycle in water vapor. The results are shown in Fig. 10. It is quite clear that CFSR’s annual tropical cycle is too large compared to observations while MERRA and ERAi’s annual cycle is about right. On the other hand, MERRA and ERAi tend to isolate the tropics from the extra-tropics.
too much with the dry high latitude summer period arriving later than observed in those reanalyses.

To quantify the annual cycle amplitude differences further we have performed a regression analysis on the observed and model water vapor fields. This approach is described in Stolarski et al. (2006) where the data are fit to the annual cycles, QBO, solar cycle, ENSO, volcanic eruptions, etc. As a consistency check on the trends, we also have performed a simple linear fit to the data and found that they agree. The annual cycle component is listed in Table 2.

As noted above the annual cycle is relatively larger in CFSR both in the tropics and extra tropics, but the extra-tropical annual cycle is relatively weaker in ERAi and MERRA. The extra-tropical annual cycle is smaller than in the Boulder data compared to both the models and HALOE + MLS observations. This is not the result of comparing a zonal mean value to the local value – similar differences show up when local comparisons are done.

We have also used the regression model to examine long-term trends in the observations and model results after the annual, ENSO, volcanic, QBO, etc. perturbations are removed. Table 3 summarizes our estimates of the net change in water vapor over two periods, 1982–2010 (Boulder) and 1993–2010 (HALOE + MLS) using both observations. The net change is computed by multiplying the linear trend by the number of years of data. We also show model net changes for the same period. We use a ±5° latitude window for the zonal mean HALOE + MLS observations. Note that the long-term trends for HALOE + MLS and Boulder data sets are not consistent at 40° N over the same period. This inconsistency may be a result of the interference from the Pinatubo aerosol in HALOE retrievals at the beginning of the analysis period. Our data shows a clear decrease in HALOE water at 40° N from 1993–1995, which is at odds with the Boulder data (Hurst et al., 2011). We also note that the models report different trend magnitudes and even different signs for both periods.
4 Summary and conclusions

This paper gives the results of long-term integrations of the forward trajectory model described in SD2011 to compute water vapor. In SD2011, we used MERRA reanalyses; in this paper, we extend analysis period back to 1979 and include the NOAA CFSR and ECMWF ERAi reanalyses as well.

CFSR and ERAi dehydration patterns show less dehydration occurring in the TWP Southern Hemisphere than MERRA, but the other two models also show dehydration taking place over South America as well as East Asia. CFSR tropical tropopause temperatures are somewhat warmer than MERRA – which has higher resolution at the tropopause – leading to ∼3/4 ppmv more water vapor at 100 hPa in the former. ERAi, on the other hand, is cold biased relative to the other reanalyses, producing 3/4 ppmv lower water vapor at 100 hPa than MERRA. Methane produced water is critical in explaining the increase in water vapor with latitude toward the North Pole. In the Southern Hemisphere, Antarctic dehydration causes a decrease in water with latitude toward the South Pole. Extra tropical water vapor is controlled by the stratospheric overturning circulation and changes in that circulation may be evident in future measurements of stratospheric water.

The models do a reasonable job of reproducing the tropical tape recorder and mid-latitude water vapor observations. We find that the ERAi water vapor signal is moving upward too fast in the 17–22 km region, suggesting about 30% too high a vertical velocity in that region. CFSR water vapor signals move upward at about the right speed, and MERRA vertical velocities are too low by about 15%.

The model annual cycles in water vapor are weaker than observed at Boulder but consistent with HALOE + MLS time series. The models and the combined HALOE + MLS observations show no significant long-term trend in tropical 100-hPa water vapor; however, outside of the tropical lower stratosphere, the models cannot agree on the magnitude of the trends nor its sign. One possible explanation is that
there is too little high quality observational data to constrain the model TTL biases in the early periods of the long-term integrations.

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References


Table 1. Water vapor comparisons.

<table>
<thead>
<tr>
<th>Reanalysis</th>
<th>MERRA</th>
<th>CFSR</th>
<th>ERAi</th>
<th>Observed MLS</th>
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<tr>
<td>Global average water vapor 18–28 km</td>
<td>4.54</td>
<td>4.89</td>
<td>4.13</td>
<td>4.53</td>
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</table>
Table 2. Average and Annual Cycle Components for Water Vapor.

<table>
<thead>
<tr>
<th>Data Set</th>
<th>Equator (100 hPa)</th>
<th>40° N (100 hPa)</th>
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<tr>
<td></td>
<td>Average</td>
<td>Annual Cycle</td>
<td>Average</td>
<td>Annual Cycle</td>
</tr>
<tr>
<td>HALOE &amp; MLS(^1)</td>
<td>3.81</td>
<td>1.03</td>
<td>4.53</td>
<td>0.42</td>
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<tr>
<td>Boulder</td>
<td>N/A</td>
<td>N/A</td>
<td>4.17</td>
<td>0.64(^2)</td>
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<td>MERRA</td>
<td>3.75</td>
<td>1.25</td>
<td>3.54</td>
<td>0.34</td>
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<tr>
<td>CFSR</td>
<td>4.42</td>
<td>1.49</td>
<td>4.47</td>
<td>0.63</td>
</tr>
<tr>
<td>ERAi</td>
<td>3.08</td>
<td>0.95</td>
<td>3.46</td>
<td>0.27</td>
</tr>
</tbody>
</table>

\(^1\) HALOE values increased by mean difference between HALOE and MLS; HALOE 1993–2005, MLS 2001–2010.

\(^2\) Boulder balloon data over same period averaged using MLS retrieval kernels. Satellite data is zonal mean while balloon data is only over Boulder.
**Table 3.** Net change in water vapor from models and observations in ppmv over indicated period.

<table>
<thead>
<tr>
<th>Latitude/Pressure</th>
<th>MERRA</th>
<th>CFSR</th>
<th>ERAi</th>
<th>Observations</th>
</tr>
</thead>
<tbody>
<tr>
<td>0° N/100 hPa</td>
<td>0.14±0.05</td>
<td>0.05±0.06</td>
<td>0.16±0.02</td>
<td>−0.57±0.06</td>
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<tr>
<td>0° N/30 hPa</td>
<td>−0.18±0.03</td>
<td>−0.55±0.05</td>
<td>0.12±0.02</td>
<td>−0.23±0.02</td>
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<tr>
<td>40° N/100 hPa</td>
<td>0.12±0.02</td>
<td>−0.2±0.03</td>
<td>0.13±0.01</td>
<td>−0.36±0.04</td>
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<tr>
<td>40° N/30 hPa</td>
<td>−0.4±0.017</td>
<td>−0.6±0.03</td>
<td>0.11±0.01</td>
<td>−0.45±0.03</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Latitude/Pressure</th>
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<th>CFSR</th>
<th>ERAi</th>
<th>Observations</th>
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<tbody>
<tr>
<td>40° N/100 hPa</td>
<td>−0.2±0.02</td>
<td>−0.65±0.03</td>
<td>0.42±0.01</td>
<td>−0.12±0.2</td>
</tr>
<tr>
<td>40° N/30 hPa</td>
<td>−0.13±0.01</td>
<td>−0.67±0.02</td>
<td>0.45±0.01</td>
<td>0.46±0.18</td>
</tr>
</tbody>
</table>

HALOE + MLS data analyzed within ±5° window of target latitude. Net changes include 1 sigma uncertainty in trend calculation.
Fig. 1. (a) PDF of 100 hPa winter (DJF) temperatures at Singapore and for the three reanalyses (with units of number of observations per K). (b) Winter temperatures and saturation water vapor concentration using Murphy and Koop (2005) for observations from Singapore and the three reanalyses at Singapore. Solid lines, mean temperatures from 1979–2010; dashed lines, one standard deviation below the mean. MLS (2005–2010) is zonal average of 100 hPa observations for the months of DJF.
Fig. 2. Mean ages (years) for MERRA (a), CFSR (b), ERAi (c) computed in December 2009 after 30 yr integration.
Fig. 3. Density of final dehydration locations from the three reanalyses, (a) MERRA, (b) CFSR, and (c) ERAi.
Fig. 4. Water vapor mixing ratio from MLS (a), MERRA (b), CFSR (c), and ERAi (d). Black contours show zonal mean temperature.
Fig. 5. Zonal (a) water vapor at 100 hPa all seasons (b) mean age. Zonal mean water vapor boreal winter (c) boreal summer (d), MLS, solid, MERRA, dotted, CFSR, dash, ERAi dash-dot.
Fig. 6. Comparison of time mean (August 2004–December 2009) water vapor. Left column, full simulation including methane oxidation; middle column, without methane oxidation; right column shows difference. These integrations do not include gravity wave parameterization, convective adjustment, and saturation is set to 100%.
Fig. 7. Water vapor anomalies at the equator from 1993 to 2010. Top, HALOE and MLS combined (HALOE before 2005 – dark line). Top-middle MERRA, bottom middle CFSR, bottom, ERAi.
Fig. 8. Correlation of MERRA, CFSR and ERAi results with observations after phase shifting the observations by months indicated. Dashed line shows the peak correlation.
Fig. 9. Times series of 100 hPa water vapor for the three reanalyses compared to MLS as in Fig. 4.
Fig. 10. Same as Fig. 9 but the water vapor fields are adjusted to the MLS zonal mean.