Continuous monitoring of summer surface water vapor isotopic composition above the Greenland Ice Sheet

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Abstract. We present here surface water vapor isotopic measurements conducted from June to August 2010 at the NEEM (North Greenland Eemian Drilling Project) camp, NW Greenland (77.45° N, 51.05° W, 2484 m a.s.l.). Measurements were conducted at 9 different heights from 0.1 m to 13.5 m above the snow surface using two different types of cavity-enhanced near-infrared absorption spectroscopy analyzers. For each instrument specific protocols were developed for calibration and drift corrections. The intercomparison of corrected results from different instruments reveals excellent reproducibility, stability, and precision with a standard deviations of ~0.23 ‰ for δ¹⁸O and ~1.4 ‰ for δD. Diurnal and intraseasonal variations show strong relationships between changes in local surface humidity and water vapor isotopic composition, and with local and synoptic weather conditions. This variability probably results from the interplay between local moisture fluxes, linked with firn–air exchanges, boundary layer dynamics, and large-scale moisture advection. Particularly remarkable are several episodes characterized by high (> 40 ‰) surface water vapor deuterium excess. Air mass back-trajectory calculations from atmospheric analyses and water tagging in the LMDZiso (Laboratory of Meteorology Dynamics Zoom-isotopic) atmospheric model reveal that these events are associated with predominant Arctic air mass origin. The analysis suggests that high deuterium excess levels are a result of strong kinetic fractionation during evaporation at the sea-ice margin.

1 Introduction

Water stable isotopes from Greenland ice cores provide highly resolved, well-dated climate information (e.g., Svensson et al., 2006). The archived climate signal is however an integrated signal of the precipitation isotopic composition, which itself is controlled by variations in moisture origin and condensation history (Sodemann et al., 2008; Masson-Delmotte et al., 2005; Johnsen et al., 1989; Fisher, 1992, 1991). The initial snowfall isotopic composition signal is subject to post-deposition processes linked with wind

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scouring, exchange between top firn layer and atmosphere, and interstitial diffusion (Johnsen, 1977; Johnsen et al., 2000; Simonsen et al., 2011; Kavanaugh and Cuffey, 2003; Fisher, 1992, 1991; Koerner and Fisher, 1990; Fisher, 1990). Classically, the interpretation of ice core data relies on theoretical calculations, using Rayleigh distillation models or atmospheric general circulation models coupled with water stable isotopes (Hoffmann et al., 2000; Joussauyme et al., 1984; Johnsen et al., 1989; Ciais and Jouzel, 1994), to explore the climatic controls on precipitation isotopic composition. However key parameterizations of the water and vapor isotope cycles in these models remain associated with significant uncertainty – from the evaporation in the source region (Merlivat and Jouzel, 1979), through the presence of mixed phase clouds (Ciais and Jouzel, 1994), to the formation of snow crystals under super saturation (Jouzel and Merlivat, 1984). To validate some of the assumptions associated with these parameterizations, models have been compared with global datasets of precipitation isotopic composition and Greenland ice core records (Sjolte et al., 2011). Only very limited datasets of direct Greenland snowfall measurements are available (Grootes and Stuiver, 1997; Steen-Larsen et al., 2011), and, so far, atmospheric models equipped with water stable isotopes have not been validated against these data.

The isotopic composition of atmospheric water vapor represents an intermediate product in the hydrological cycle between the source evaporation and cloud condensation. Combined with precipitation data, measurements of water vapor isotopic composition have therefore the potential of enhancing our knowledge of the processes governing the hydrological cycle. Indeed, large-scale advection, vertical mixing in the atmospheric boundary layer, cloud microphysics, and exchange between the atmosphere and the snow surface do affect water vapor and precipitation isotopic composition. Direct isotopic water vapor measurements in the atmosphere are therefore needed. As the use of satellite remote sensing of water vapor isotopic composition is currently not available for high latitudes (Worden et al., 2006; Frankenber et al., 2009), in situ observations are essential. So far, only few studies have carried out long-term monitoring of the water vapor in the atmosphere (Jacob and Sonntag, 1991; Angert et al., 2008) or analyzed the water vapor isotopic composition above the Greenland Ice Sheet (Grootes and Stuiver, 1997; Steen-Larsen et al., 2011).

We briefly review here existing Arctic water vapor isotope datasets. The earliest water vapor studies showed that the isotopic composition of the surface vapor was strongly affected by changes in air mass origin as well as by the history of precipitation from the air mass prior to collection (Craig and Gordon, 1965; Dansgaard, 1954). However, due to the labor intensity related to collecting the vapor samples using a cold trap, such studies have remained extremely limited especially for the Arctic.

High d-excess (d-excess = δD − 8 × δ18O) levels were observed in moisture originating from the Arctic by Kurita (2011) during a cruise in the Bering Strait in September and October 2008 where 140 samples were collected with a 6-hourly resolution. These high d-excess values were attributed to a strong kinetic fractionation caused by dry polar air masses transported above the open sea near the edge of the sea ice. Similar effect of kinetic fractionation on d-excess was previously observed in the eastern Mediterranean (Gat et al., 2003; Angert et al., 2008). This kinetic fractionation effect arises in the skin layer where molecular diffusion is dominant. The humidity gradient in the skin layer affects the relative fractionation of HD16O and H218O due to their difference in molecular diffusivities (Merlivat and Jouzel, 1979; Fisher, 1991).

From samples collected twice per day by Grootes and Stuiver (1997) during the summer field season of 1990 at GISP2 (Greenland Ice Sheet Project 2), central Greenland, very high correlation was found between the variability of δ18O of the snow and frost, and the δ18O of the atmospheric water vapor samples. A similar conclusion was reached based on 29 atmospheric water vapor samples collected morning and evening during the 2008 field season at the North Greenland Eemian Drilling Project (NEEM), NW Greenland (Steen-Larsen et al., 2011). They showed δ18O of the vapor to be lagging variations in precipitation δ18O during snowfall events. Both d-excess measurements of these samples (Steen-Larsen et al., 2011) and 17O-excess measurements conducted on a subset of these samples (Landais et al., 2011) indicated that the surface water vapor predominantly is in isotopic equilibrium with the snow surface. Still, unknown factors exist in relation to formation and post-depositional processes of the isotopic composition of the snow surface. Specifically, the magnitude of the interaction between the snow surface and the water vapor above the snow surface as well as the cloud condensation height governing the relation between temperature and isotopic composition of the precipitation.

Cavity-enhanced absorption spectroscopy has recently allowed continuous in situ measurements of atmospheric water vapor isotopic composition (Kerstel et al., 1999; Lee et al., 2005). In this study, we take advantage of different types of commercial laser instruments recently developed (Crosson et al., 2002; Baer et al., 2002). By deploying two different laser instruments and comparing the independently produced data records we are directly able to evaluate the accuracy and precision. We have implemented specific calibration protocols to correct for instrumental drifts and produce results against international standard waters. To enhance the interpretation of ice core records, we have conducted our surface water vapor monitoring program at the NEEM deep ice core drilling camp, NW Greenland, during the 2010 summer field season.

In Sect. 2, we describe the methods applied (NEEM site characteristics, field setup, calibration protocol, back trajectory analyses and modeling tools). In Sect. 3, we quantify the precision and accuracy of the isotopic vapor measurements,
describe the observed isotopic variability and combine atmospheric modeling with our observations to investigate the large-scale drivers of day-to-day isotopic variability.

2 Material and methods

Throughout the paper, we will use the δ notation (Dansgaard, 1964). The definition of the δ notation is based on the relative composition (R) of the two stable water isotopes \(^{1}H^{2}H^{18}O\) and \(^{2}H^{18}O\) compared to \(^{1}H^{2}H^{18}O\), and is given by

\[
\delta^s = \left( \frac{R_{\text{Sample}}}{R_{\text{V-SMOW}}} - 1 \right) \times 1000 [\%] ,
\]

where \(\delta^s\) represents either \(\delta^{18}O\) or \(\delta^D\), and \(R_{\text{V-SMOW}}\) is the relative composition of the V-SMOW standard, Vienna Standard Mean Ocean Water.

2.1 NEEM site characteristics

An international deep drilling program has been conducted at NEEM, NW Greenland (77.45° N, 51.05° W, 2484 m a.s.l.), from 2007 to 2012, providing climatic and glaciological information back to the last glacial period. Specific monitoring efforts have been dedicated to water stable isotopes in pits, surface snow, precipitation and water vapor (Steen-Larsen et al., 2011). Air temperature and relative humidity (postcorrected with respect to ice) were measured using Campbell Sci. HMP45C (\(±0.1\) K and \(±5\% < 90\%\) RH, relative humidity, and \(±10\% > 90\%\) RH, wind direction and speed using an R.M. Young propeller-type vane (\(±5^\circ\) and \(±0.1\) m s\(^{-1}\)), and station pressure using Vaisala PTB101B (\(±0.1\) mb) (Steffen and Box, 2001; Steffen et al., 1996). The estimated mean summer (JJA) temperature at NEEM is \(∼−11\ ± \ 5^\circ\) C (1σ based on 2006–2011), and the annual mean accumulation rate from 1964 to 2005 is estimated at \(20 \text{ cm a}^{-1}\) (water equiv.), with a large fraction (between 2.5 and 4.5) of precipitation occurring in JJA compared to DJF (Steen-Larsen et al., 2011). July of 2010, which is a particular focus of this paper was characterized by a high and stable AO (Arctic Oscillation) index of \(∼−0.42\) compared to a climatologically mean of \(∼−0.11\ ± 0.43\). The mean of the variability of the AO July index is \(∼0.63 \ ± 0.25\) compared to the 2010 variability of the AO July index of \(∼0.40\).

2.2 Atmospheric modeling

In order to characterize the long-range air mass advection to the NEEM site back-trajectory calculations using the FLEXTRA (FLEXible TRAjectories) model (Stohl and Seibert, 1998; Stohl et al., 1995) have been conducted for the period of the measurement campaign. The advection field is based on 6 hourly ECMWF (European Centre for Medium-Range Weather Forecasts) meteorological analysis data supplemented by 3 h forecast data. Seven-day back-trajectories were initiated at 2500 and 3000 m a.s.l every 6 h over the NEEM site (altitude in the ECMWF model \(∼2380\) m a.s.l). Depending on the meteorological situation, air mass trajectories could differ in shape depending on initial altitude, but at most times the initial flow direction was similar. We therefore only show results from the trajectories started at 2500 m a.s.l.

To support the interpretation of the isotopic water vapor data, we use outputs from a general circulation model, LMDZiso (Laboratory of Meteorology Dynamics Zoom-isotopic; Hourdin et al., 2006), enabled with water isotopes (Risi et al., 2010a, b). The model is used with a resolution of 2.5° in latitude, 3.75° in longitude and 19 vertical levels. The model is forced by the monthly sea surface temperature from the NCEP (National Centers for Environmental Prediction) reanalysis (Kalnay et al., 1996; Kanamitsu et al., 2002). Simulations using sea-ice concentration from Nimbus-7 SSMR (Scanning Multichannel Microwave Radiometer) and DMSP SSM/I (Defense Meteorological Satellite Program Special Sensor Microwave/Imager) passive microwave data (Fetterer et al., 2002) led to little change compared to standard forcing used by LMDZiso (not shown). The simulated three-dimensional fields of horizontal winds are nudged towards those of the ECMWF operational analysis, so that the model can capture weather patterns at the daily scale. We use the simulated water vapor isotopic composition from the lowest model level at the NEEM grid point to compare with our observations. To investigate the relationship between the inflow of Arctic moisture and the water vapor isotopic composition, we ran LMDZiso with the water tagging functionality (Risi et al., 2011) to track the water vapor evaporated north of 70° N (Masson-Delmotte et al., 2011). We use the LMDZiso since it has previously been used to understand the isotope variability of the NEEM ice core and was shown to correctly capture interannual variability (\(R^2 = 0.32\) (Steen-Larsen et al., 2011). However the model does also show the same caveats as other GCMs (general circulation models) applied to the Greenland Ice Sheet, having a warm bias and too enriched isotope levels. The super saturation is given by \(S = 1.00 − 0.004\ T\) (Jouzel and Merlivat, 1984). The fractionation coefficient between 0°C and \(−10°C\) is the linear combination of the fractionation coefficient for liquid and for ice. Below \(−10°C\) only the fractionation coefficient for ice is used.

2.3 Laser instruments and field setup

We used two different Picarro CRDS (cavity ring down spectroscopy)-analyzers (hereafter Picarro analyzer) and one LGR inc. ICOS (integrated cavity output spectroscopy)-analyzer (hereafter LGR analyzer). See Table 1 for overview of the different deployment period and calibration periods. Both types of analyzers are based on cavity-enhanced, near-infrared laser absorption spectroscopy (CEAS) techniques: the Picarro analyzer specifically uses cavity-ring down spectroscopy (CRDS) whereas the LGR analyzer uses a later CEAS technique called off-axis integrated-cavity-output
Fig. 1. Illustration of setup with Picarro analyzer connected to small tower and LGR analyzer connected to tall tower.

spectroscopy (OA-ICOS) (Baer et al., 2002). CRDS uses time-based measurements of the exponential decay of light resonating in the optical cavity to quantify the optical loss at different optical wavelengths across a molecular absorption feature (Brand et al., 2009; Crosson et al., 2002). The OA-ICOS is based on measurement of the transmitted intensity through the cavity, and typically averages several hundred continuous sweeps per second through a molecular absorption feature. After each sweep a ring-down measurement is made to verify the baseline absorption. Both techniques use the Beer–Lambert law to calculate the concentrations of each species, with key differences being that CRDS lends itself to smaller cavity volumes due to its on-axis beam geometry while ICOS offers much faster scan speeds and wider dynamical ranges.

We used the LGR water vapor isotope standard source (WVISS) in a dual-inlet mode to calibrate and drift-correct both the Picarro and LGR analyzer outputs. The WVISS consists of a hot chamber (2 L) where water of a known isotopic composition is injected using a nebulizer. The water concentration level can be controlled by changing the airflow (2–10 L min\(^{-1}\)) through the hot chamber. The nebulizer and hot chamber ensure perfect evaporation of the liquid water hence providing an airstream with a known isotopic concentration of the water vapor.

Figure 1 displays our system setup. Two laser analyzers were placed in a heated tent, ~50 m SW of the NEEM camp on the edge of the clean air zone (Steen-Larsen et al., 2011). The temperature inside the tent was observed to have fluctuations of up to 20 °C during a day, in relationship with local weather. To minimize temperature-driven drifts of the analyzers, the Picarro analyzer was placed in a passive temperature regulated box. However, we found that this passive temperature regulation was not sufficient and temperature variations were likely the cause of instrumental instabilities and loss of accuracy (see next sections). The LGR analyzer was placed in an active temperature-regulated box controlling the temperature within 0.2 °C.

Two towers were erected next to each other, with respective heights of 1.5 m (hereafter small tower) and 13.5 m (hereafter tall tower). For the majority of the measuring campaign, the Picarro analyzer measured air samples collected on the small tower, while the LGR analyzer measured air samples on the tall tower. On the small tower air was sampled at 0.1, 0.3, 0.6, 1.0, and 1.5 m above the snow surface, while on the tall tower air was sampled at 1.0, 1.5, 3.5, 7.5, 10.5, and 13.5 m. An automatic valve system shifted between the different heights on the two towers and with regular intervals to the calibration unit in order to correct for drifts. Three-way solenoid valves installed on the sample line of each laser analyzer allowed selecting the inflow of either inlet line air or calibration unit air into each of the laser analyzers (see Fig. 1 for an illustration of the setup). The calibration vapor was introduced into the sample line right before the analyzers in order to minimize time spent on calibration due to memory effect of the system. Also to minimize the complexity of this field-deployed system it was decided not to introduce the calibration vapor at the beginning of the inlets on the tower. It was thereby assumed that besides from memory effects no interactions between the ambient vapor and the inside of the sample tubes occurred. This assumption is supported by no observed discrepancy between the measurements from significantly different length of tubes going to respectively the top and bottom inlet on the tower. Each level was sampled for 15 min of which the first 5 min were disregarded to remove memory effects. Vapor from the calibration unit was measured for 10 min. To decrease the residence time of air in the inlet lines, the sample line air was pumped at approximately 4 L min\(^{-1}\) on the small tower and 6 L min\(^{-1}\) on the tall tower. A second pump flushed the other lines not being sampled with about 1 L min\(^{-1}\) flow for each of the different lines. The inlet lines consisted of 1/4” outer diameter Synflex 1300 and were heated to a maximum temperature of 60 °C using a self-regulating heat trace. At the entrance of inlet lines, we used plastic bottles with perforations in the bottom to prevent sucking snow crystals into our
Table 1. Summary of the different commercial analyzers deployed and the timing of deployment and calibrations.

<table>
<thead>
<tr>
<th>Instrument</th>
<th>Serial number</th>
<th>Deployment period in 2010</th>
<th>Days of humidity-isotope response calibration</th>
<th>Days of VSMOW calibration</th>
<th>Frequency of drift correction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Picarro inc. L1102-i</td>
<td>HBDS-48</td>
<td>Day ~ 144–154</td>
<td>Day 140</td>
<td>Day 144</td>
<td>~ 1.5 h for days 144–147 and ~ 12 h for days 148–154</td>
</tr>
<tr>
<td>Picarro inc. L1102-i</td>
<td>HBDS-12</td>
<td>Day ~ 160–210</td>
<td>Day 159</td>
<td>Days 160, 171, and 178</td>
<td>~ 12 h for full period of deployment</td>
</tr>
<tr>
<td>LGR inc. 908-0004-0003</td>
<td>10-0037</td>
<td>Day ~ 144–204</td>
<td>Days 140 and 159</td>
<td>Days 144, 160, 171, and 178</td>
<td>~ 1.5 h for full period of deployment</td>
</tr>
</tbody>
</table>

system. Direct observations showed that the temperature inside the clear plastic bottles was about 10 °C warmer than the ambient air due to solar heating, thereby preventing condensation. All lines were insulated with either Rockwool™ or polyethylene pipe insulation with a minimum wall thickness of ~ 2.5 cm.

2.4 Calibration protocols

In order to allow the comparison of laser water-vapor measurements with other data (for example with snow samples measured by IRMS or with outputs from isotopic models), it is crucial to produce isotopic-composition data against a known isotopic standard. A thorough characterization of the instrumental system is necessary to detect the variability caused by the analyzer (e.g., due to temperature or humidity responses) from the true atmospheric isotopic signals. The characterization of each individual isotopic analyzer follows in many ways the protocols classically established for IRMS instruments.

We refer the reader to the supplementary material for details but to sum up our calibration protocol, 6 steps must be followed:

1. The humidity-isotope response function must be determined for each instrument. If the field campaign is conducted over an extended period of time, the humidity-isotope response function needs to be determined repeatedly due to, e.g., temperature and pressure caused drifts.

2. Using known-isotopic vapor standards, a linear function for transferring the instrument isotopic values to the V-SMOW – SLAP scale must be determined.

   To check for stability, this function should be re-determined during extended field campaigns.

3. Raw measurements must be transferred to a reference humidity level using the humidity-isotope response function determined for each instrument.

4. The humidity-corrected measurements must be transferred from the instrument isootope scale to a V-SMOW – SLAP scale.

5. The final calibrated isotopic water vapor data are obtained by drift correcting the V-SMOW – SLAP calibrated measurements using a known vapor standard measured at regular intervals (depending on the drift of the instrument).

6. Possible poor performance can be flagged if measurements of the known vapor standard used for drift corrections are significantly different from the trend of the drift.

3 Results and discussions

3.1 Instrument comparison and overall performance

Instrument-specific humidity and drift corrections have been applied for each analyzer. A comparison of the corrected results obtained with the two different instruments can therefore provide a stringent assessment of the validity of our calibration protocols.

Figure 2 shows the δD, δ18O, and d-excess measured at 1 m above the snow surface by the LGR analyzer and the Picarro analyzer (HBDS 48 for days ~ 144–154 and HBDS 12 for days ~ 160–210). From day ~ 144 to ~ 147, both the LGR and Picarro analyzer (HBDS 48) were installed to measure the same sampled air on the tall tower. During this period, both instruments were drift corrected every ~ 1.5 h as explained above and in the supplementary material.

From day ~ 148 and through the rest of the campaign, the LGR and Picarro analyzers (HBDS 48 for days 148–154 and HBDS 12 for days 160–210) were measuring air sampled on separate lines. During this time period, the LGR analyzer is drift corrected every ~ 1.5 h as in the beginning, while the Picarro analyzer is drift corrected every ~ 12 h. We use the period from day ~ 144–147 to compare the performance of the LGR analyzer and Picarro (HBDS 48) analyzer when both
Fig. 2. Meteorological measurements shown in the top four panels. From the top – wind direction in true north, wind speed in m s\(^{-1}\), relative humidity in %, and temperature in °C. Green bars indicate dates with snow showers. Fifth panel shows the humidity in ppmv as measured by the LGR analyzer. The three lower panels show, from below, δ\(_D\), δ\(_{18}O\), and d-excess in ‰ calibrated on the V-SMOW – SLAP scale. Red line shows data from the Picarro analyzers (HBDS 48 for days ∼144–154 and HBDS 12 for days ∼160–210) and blue line shows data from the LGR analyzer. Black bars in the bottom of the figure indicate the period where calibration was working optimally.

Instruments are drift corrected every ∼1.5 h. To compare the performance of the Picarro (HBDS 12) analyzer when drift corrected every ∼12 h, we use the period from ∼160–180. This period was selected because both instruments and calibration unit were optimally working.

For the first inter-calibration period, the inter-instrument difference in the δ\(_D\) data is 0.2 ± 1.4 ‰ (standard deviation calculated from 191 points, averaged over 10 min). Similarly, the difference in δ\(_{18}O\) has a mean value of 0.02 ± 0.23 ‰ (N = 191). Due to the small mean of the differences in δ\(_D\) and δ\(_{18}O\), we conclude that we have an optimal humidity-isotope response calibration and V-SMOW calibration for both machines. Based on the standard deviation of the differences we conclude that with ∼1.5 h calibration we have a precision on the LGR and Picarro analyzer (HBDS 48) of ±1.4 ‰ for δ\(_D\) and ±0.23 ‰ for δ\(_{18}O\).

During the second period, where the LGR and Picarro analyzer (HBDS 12) were measuring on separate lines, we compare the measurements carried out at 1.0 m by the LGR analyzer with the closest timewise measurement carried out by the Picarro analyzer at any of the heights 0.3, 0.6, 1.0, or 1.5 m. The measurements are drift corrected and transferred to the V-SMOW – SLAP scale. The difference between the LGR and Picarro analyzers is 0.7 ± 5.4 ‰ for δ\(_D\) (N = 231) and 0.14 ± 0.9 ‰ for δ\(_{18}O\) (N = 231). We have no reason to believe that the accuracy of the LGR analyzer and correction protocol have changed. The difference between the LGR and Picarro analyzers is a composite of the precision of the LGR and of the Picarro analyzer, plus the uncertainty introduced by the increased time between calibrations, and the noise caused by the different sampling lines (up to 0.7 m height difference). Using the standard deviation of the distribution of differences we conservatively estimate the precision of the Picarro analyzer (HBDS 12) with 12 h drift correction to be ±5.4 ‰ for δ\(_D\) and 0.9 ‰ for δ\(_{18}O\). Table 2 summarizes these findings on the precision and accuracy of our measurements. Using this, the uncertainty on the d-excess of the LGR and Picarro (HBDS 48) analyzers, when averaged over a 10 min time window, will be 2.3 ‰ when using 1.5 h drift correction, while when using 12 h drift correction on the Picarro (HBDS 12) analyzer the uncertainty of the d-excess becomes ∼9 ‰. We attribute this large increase in uncertainty to the large temperature variations, which the Picarro analyzers were subjected to, while the LGR analyzer was placed in a temperature-controlled box. It should also be noted that the Picarro analyzers are operated outside the humidity range specified by the manufacturer.
Table 2. Direct comparison between LGR analyzer and Picarro analyzers (HBDS 12 and HBDS 48). The measurements are averaged over a period of 10 min.

<table>
<thead>
<tr>
<th>Difference</th>
<th>$\delta D$ Mean</th>
<th>$\delta D$ Standard deviation</th>
<th>$\delta^{18}O$ Mean</th>
<th>$\delta^{18}O$ Standard deviation</th>
<th>$N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>LGR – Picarro (HBDS 48) analyzer. 1.5 h drift correction for both. Days $\sim$ 144 to $\sim$ 147</td>
<td>0.2‰</td>
<td>1.4‰</td>
<td>0.02‰</td>
<td>0.23‰</td>
<td>191</td>
</tr>
<tr>
<td>Measurement on same inlets</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>LGR – Picarro (HBDS 12) analyzer. 12 h drift correction for LGR analyzer. Days $\sim$ 160 to $\sim$ 180</td>
<td>0.7‰</td>
<td>5.4‰</td>
<td>0.14‰</td>
<td>0.90‰</td>
<td>231</td>
</tr>
<tr>
<td>Measurement on different inlets</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

3.2 $\delta D$ vs. $\delta^{18}O$

Figure 2 shows the water vapor isotopic composition measured at 1 m above the snow surface. The Picarro analyzers (red solid line) measured this first on the tall tower (days $\sim$ 144–147) and then on the small tower (days $\sim$ 148–154 and $\sim$ 160–210) while the LGR analyzer measured this on the tall tower (blue solid line). The black bars in the bottom of the figure illustrate the time intervals when the calibration routine was working optimally (Sect. 2.5). Unfortunately, during the second half of the field campaign, the stability of the calibration unit was affected by an undetected loose wire in the WVISS. Based on analysis of the signal, we find that the largest confidence should be placed in the signal from the LGR analyzer, except the last 5 days of the campaign.

We notice in Fig. 2 several events of very high d-excess levels in the water vapor. In order to obtain an objective identification of high d-excess events, we perform a cluster analysis using an expectation-maximization algorithm for Gaussian mixture models (Moon, 1996). This analysis shows a distinct cluster with a mean ($\mu$) of $\sim 23$ ‰ and standard deviation ($\sigma$) of $\sim 5$ ‰, and a second cluster centered around $\mu \approx 37$ ‰ with $\sigma \approx 8$ ‰. We subjectively define normal d-excess as values being within $\pm 0.5 \sigma$ of the $\mu$ of the main cluster and high d-excess as being larger than the $\mu + 3\sigma$ of the main cluster.

Figure 3 shows the observation by the LGR analyzer for $\delta D$ vs. $\delta^{18}O$ at 1 m above the snow surface. The red dots indicate data obtained during periods of high d-excess as defined above while blue dots indicate data obtained during periods of normal d-excess. Black dots are data that do not fall into any of the two categories. Red crosses represent mean daily outputs from the LMDZiso model (Risi et al., 2010a, b). These model outputs are also presented in Fig. 6.

We first compare the $\delta D$ vs. $\delta^{18}O$ slope of our water vapor monitoring data. A best linear fit through all measurements (black solid line) reveals a slope of $6.46 \pm 0.07$ and intercept of $-32.6 \pm 3.0$ ‰. The slope is comparable to the best fit through the cold trap vapor measurements presented in Steen-Larsen et al. (2011), where the slope was found to be $6.89 \pm 0.15$ and intercept $-17.8 \pm 0.6$ ‰. We consider this as support for an only negligible uncorrected bias in our measurements presented here. For the high d-excess data only, we obtain a significantly different slope and intercept, respectively $7.44 \pm 0.17$ and $21.2 \pm 7.4$ ‰ (red solid line). Removing the set of measurements corresponding to high d-excess as defined above, a best fit through the remaining...
measurements (green solid line) reveals a slope and intercept of respectively 7.21 ± 0.06 and −5.3 ± 2.5. We conclude that (i) the continuous measurement data are consistent with the earlier cold trap data available for NEEM water vapor; (ii) that the δD vs. δ18O slope is systematically lower in the NEEM vapor than in the NEEM precipitation, estimated to be ∼7.6 (Steen-Larsen et al., 2011); (iii) that the high versus normal d-excess events correspond to different δD vs. δ18O relationships and therefore different distillation/source histories.

We now compare the outputs of the LMDZiso model with our observations. First, we note that the simulated water vapor δD vs. δ18O-relationship (7.48 ± 0.07) shows a slope significantly larger than observations at NEEM. This higher slope is consistent with the fact that LMDZiso produces too-low d-excess levels in both summer and annual mean precipitation at NEEM (Steen-Larsen et al., 2011). Moreover, LMDZiso produces very similar slopes in the vapor (7.48) and in the summer precipitation (∼7.6, in good agreement with the observed slope of ∼7.6 at NEEM; Steen-Larsen et al., 2011). Assuming that the isotopic vapor measurements are representative for the summer vapor at NEEM, we conclude that the LMDZiso does not correctly resolve the processes accounting for the observed different slopes in water vapor and snowfall. This mismatch may be linked with the physical parameterization of LMDZiso related either to the formation of snow crystals, with the representation of atmosphere–snow surface interactions, boundary layer dynamics, or physical parameterization of the evaporation in the source region. Table 3 summarizes the slope and intercept for observed and modeled water vapor and precipitation at NEEM.

### 3.3 Diurnal signal

The isotopic and meteorological data displayed in Fig. 2 depict a diurnal cycle in both humidity and the isotopic composition with a respective peak-to-peak variability of ∼1800 ppmv and ∼36‰ (δD). A weak diurnal cycle at the limit of detection might be observed in the d-excess from the LGR analyzer with a peak-to-peak variability of ∼6‰ in phase with the isotopic (δ18O and δD) variations. Several periods characterized by abnormally multiday high or low mean isotopic levels have a null or very weak diurnal signal. Some of these intervals are marked by very high d-excess levels, above ∼40‰.

The period from day 180 to 190 is marked by particularly strong diurnal variability (Figs. 2, 4). This is also seen in the 1 m, diurnal surface-air temperature variability, which shows on average a peak-to-peak variability of 10°C. In order to further characterize this variability, we analyze the vertical gradients in humidity and water vapor isotopic composition. During this period, the humidity is on average ∼350 ppmv higher at 13.5 m compared to the 1.0 m height during the evening (00:00 UTC, Universal Time Coordinated; 22:00 LT, local time) when the surface cools down. A reverse effect is seen during the morning (12:00–15:00 UTC, 10:00–13:00 LT) when a ∼200 ppmv excess of humidity exists near the surface. During the evening (marked by a humidity deficit at the surface), water vapor δD is ∼4.5‰ enriched at 13.5 m height compared to the 1.0 m level; a reversed difference of ∼4.0‰ δD is observed during the morning. The diurnal variation of the gradient with height in both humidity and isotopic composition is an indication for the snow surface acting successively as a moisture sink (evening) and source (midday) when stable weather conditions exist. This implies that the surface water vapor could be in isotopic equilibrium with the snow surface during part of the day. Steen-Larsen et al. (2011) compared the isotopic compositions of cryogenically collected vapor samples and the isotopic composition of sampled precipitation and reached the same conclusion. The cold trap measurements however did not have sufficient temporal resolution (maximum 4 daily samples) to allow a characterization of this diurnal cycle. The lack of monitoring of the boundary layer’s vertical structure at NEEM prevents us from further investigating the relationships between boundary layer dynamics and surface water vapor isotopic composition. We briefly note that a series of unresolved questions exist: the magnitude of the firm–air fluxes on a diurnal scale; the influence on the diurnal variability of the water vapor isotope by boundary layer dynamics; and the link between surface water vapor and condensation water vapor forming snow crystals in the clouds.

### 3.4 d-excess signal

We qualitatively compare the d-excess record shown in Fig. 2 with the meteorological and isotopic observation in order to understand if local mechanisms or advection are responsible for producing high d-excess events. First, we note that after the first period with high d-excess (day ∼154–156), d-excess appears to be independent of variations in local surface humidity, temperature, or isotopic composition. We therefore do not find any local, qualitative explanation for the high d-excess events.

Figure 5 shows a comparison of 5-day backward trajectories for high d-excess periods (red) and normal d-excess periods (blue). The back-trajectories corresponding to normal d-excess levels are relatively homogeneous, originating southwards of NEEM (very few from the east or west of NEEM). By contrast, most of the back-trajectories corresponding to high d-excess periods have a westward origin; some of them also have east and south-east origins. The shorter isotopic record presented by Steen-Larsen et al. (2011) also shows a period of high d-excess, which corresponds to air masses originating to the west of Greenland.

The d-excess of the water vapor in an air parcel depends on several things: the condition during the evaporation in the source region, the cumulated amount of condensation from source to sink, the cleanliness of the air, the history
of the super saturation in the cloud, and the temperature at which snow crystals starts to form. The last process is only expected to affect d-excess at very low condensation temperatures (<−25°C; Jouzel and Merlivat, 1984; Hoffmann et al., 1998). During the periods of high d-excess observed at NEEM, no exceptional cold-surface-temperatures were observed (Fig. 2); we assume that cloud temperatures and surface temperatures are related. Furthermore, back trajectories do not show high-elevation transportation paths for the air masses corresponding to high d-excess days. Since the back trajectories are only 5 days back we do not have any information about origin before this period. The absence of any systematic anticorrelation between d-excess and δD rules out a dominant effect of condensation. This leads to the hypothesis that high d-excess events are associated with moisture origin, possibly preserved from evaporation conditions.

The process of creating high d-excess in the atmospheric water vapor through strong evaporation from the sea surface was observed previously by Gat et al. (2003) for the eastern Mediterranean and by Kurita (2011) for the Bering Strait. The high d-excess in the air masses originating west of Greenland can be explained similarly. Figure 5 illustrates the Arctic sea-ice extent during May and August of 2010. The westerly-trajectories associated with high d-excess (Fig. 5) have all crossed the Baffin Bay area, southwards of the sea-ice cover, and open water between the Arctic sea ice and the North American continent. As the humidity-depleted cold polar air masses cross the sea-ice margin, strong evaporation from the open-water bodies will result in kinetic effects leading to high d-excess in the water vapor. This hypothesis is supported by the observation of high d-excess values in the surface water vapor near the sea-ice margin (Kurita, 2011).

No such simple explanation can be provided for high d-excess episodes when air masses are originating from the east or southeast of NEEM. These trajectories are however marked by shorter transportation distances (Fig. 5). We suggest that a large fraction of the moisture transported by these trajectories is provided by evaporation at the Arctic sea-ice margin areas east of Greenland.

To support our hypothesis that high d-excess values are associated with Arctic origin, where strong kinetic fractionation occurs at evaporation, we use LMDZiso water tagging diagnostics (Risi et al., 2010b) shown in Fig. 6. This diagnostics complements the investigation of air mass origins using back-trajectories by tracking water vapor transportation and accounting for the effect of air mass mixing.

GCMs have some uncertainties in their representation of sub-grid physical processes, but they are more robust in their simulation of large-scale dynamics. The consistency of the LMDZiso simulation with large-scale dynamics is warranted by the wind nudging towards the ECMWF operational analysis, which were shown to accurately capture Arctic circulation (e.g., Jakobson et al., 2012). As a result, LMDZiso correctly captures the observed patterns and magnitudes of NEEM mean daily surface-humidity variability (the correlation coefficient and ratio of standard deviation are respectively 0.70 and 0.56) as shown in Fig. 6.

Table 3. Overview of slope and intersect for δD vs. δ18O for observed and modeled water vapor and precipitation.

<table>
<thead>
<tr>
<th></th>
<th>Cold trap measurements (Steen-Larsen et al., 2011)</th>
<th>All vapor measurements (this work)</th>
<th>High d-excess measurements (this work)</th>
<th>Excluding high d-excess measurements (this work)</th>
<th>Precipitation (Steen-Larsen et al., 2011)</th>
<th>LMDZiso vapor simulation this period</th>
<th>LMDZiso summer precipitation</th>
</tr>
</thead>
<tbody>
<tr>
<td>δD vs. δ18O slope</td>
<td>6.89 ± 0.15</td>
<td>6.46 ± 0.07</td>
<td>7.44 ± 0.17</td>
<td>7.21 ± 0.06</td>
<td>7.57 ± 0.14</td>
<td>7.48 ± 0.07</td>
<td>~ 7.6</td>
</tr>
<tr>
<td>δD vs. δ18O intersect</td>
<td>~17.8 ± 6.0</td>
<td>~32.6 ± 3.0</td>
<td>21.2 ± 7.4</td>
<td>~5.3 ± 2.5</td>
<td>0.2 ± 2.8</td>
<td>~5.4 ± 2.3</td>
<td>NN</td>
</tr>
</tbody>
</table>

Fig. 4. The mean diurnal cycle on days with clear sky and calm weather. Stacked based on diurnal cycles between days ~180 and ~190. Error bars indicate the standard deviation on the mean. Solid lines represent measurements at ~1 m above snow surface. Dashed lines represent measurements at ~13 m above snow surface. Blue lines represent humidity, red lines represent δD, and black line represents temperature.
The proportion of the moisture originating from the Arctic as diagnosed from water tagging appears to co-vary with the observed NEEM d-excess (correlation coefficient of 0.30). Days with high d-excess around days 170, 198 and 203 correspond to maxima in the fraction of NEEM moisture coming from the Arctic. This confirms our result from back-trajectory analysis, and supports our hypothesis of the link between high d-excess and Arctic origin.

The model–data comparison further informs on the causes of the observed δD and d-excess variability. δD variations are reasonably well captured by LMDZiso (correlation coefficient of 0.67 and ratio of standard deviation of 0.51), and so are d-excess variations before mid-June (correlation coefficient of 0.46 and ratio of standard deviation of 0.27). This suggests that δD and d-excess variability before mid-June are mainly controlled by large-scale dynamics. In contrast, LMDZiso is unable to simulate any significant d-excess variations during mid-summer, while large variations are observed (Fig. 6). The variance for this period of the observed d-excess is ~67‰ while the variance for the modeled d-excess is ~1‰. We suspect that this model caveat could be linked to subgrid model physics, especially the difficulty to correctly simulate relative humidity at the surface of the Arctic Ocean. This was indeed demonstrated by the poor performance of LMDZiso against relative-humidity data from the SHEBA (Surface Heat Budget of the Arctic Ocean) campaign in the Arctic ocean (Persson et al., 2002). We cannot exclude that the lack in simulating the high d-excess variability is caused by a wrong parameterization of the processes coupled to the transport of moisture or snow crystal formation in the modeled hydrological cycle. It is outside the scope of this paper to study these processes in more details. However we plan to undertake a detailed study using several isotope-enabled GCMs and water vapor isotope measurements from subsequent seasons to investigate especially the parameterization of moisture transport from subtropics to Arctic regions.

On annual timescales Steen-Larsen et al. (2011) reported, for a shallow ice core drilled at the NEEM site, a positive correlation ($R^2 = 0.2$) between winter Arctic Oscillation-anomaly and annual d-excess. Positive AO and NAO (North Atlantic Oscillation) configurations, associated with respectively stronger high-latitude westerlies (Thompson and Wallace, 1998) and dominant East Greenland moisture sources (Sodemann et al., 2008), may favor the transport of high d-excess moisture to NEEM. It remains a speculation in this paper, but maybe part of the d-excess signal observed in ice cores drilled at NEEM need to be attributed to inter-annual variations in relative moisture arriving at NEEM originating from Arctic moisture sources.

4 Conclusions and perspectives

We have demonstrated the ability of laser analyzers to produce reliable measurements of Greenland surface water isotopic composition. We have proposed calibration protocols
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Fig. 6. Comparison between daily mean GCM model outputs from LMDZiso with water tagging and observed daily mean humidity and isotopic composition from 1 m above the snow surface by the LGR analyzer.

to account for humidity and drift effects, and to deliver measurements against V-SMOW references. The systematic comparison of different CRDS analyzers allows us to demonstrate the reliability of our corrections and to estimate the overall analytical precision. When run with frequent drift corrections (1.5 h frequency), we obtain inter-instrument standard deviations of 1.4 ‰ and 0.23 ‰ for δD and δ18O, respectively. When drift correcting the Picarro analyzer only every ~ 12 h, the dispersion increases by a factor of 3–4 to 5.4 ‰ and 0.9 ‰ for δD and δ18O, respectively. We speculate that this significant increase in uncertainty is caused by large temperature variations in the tent and thus around the Picarro analyzer, which did not benefit from the same setup as the LGR analyzer (temperature regulation, frequency of calibrations).

We have shown that water vapor isotope measurements are indeed feasible at low humidity concentrations but corrections are needed to remove biases. Especially for field deployments of laser analyzers either frequent drift correction is needed or significant temperature stability of the analyzer must be applied. Having two independent analyzers allow for detection of issues with the calibration.

Our data reveal a diurnal variability in surface humidity and water vapor isotopic composition (at the limit of detection for d-excess) under stable weather conditions. This diurnal variability is characterized by strongly correlated humidity and isotopic composition gradients from 0.1 to 13.5 m height. We interpret this as indications of firm–air moisture fluxes, which likely depend on the upper firm heat budget and on the atmospheric boundary layer dynamics. Further investigations of this diurnal cycle are expected to shed light on post-depositional processes affecting the isotopic composition of the snow, with large implications for an improved understanding of the isotopic diffusion in the ice core records. The isotopic data should in the future be combined with local atmospheric and surface mass-balance modeling, and may be helpful to assess the magnitude of the sublimation and condensation fluxes.

Our data also depict another mode of isotopic variability; this mode is marked by changes in the daily mean values of the humidity and isotopic composition at the synoptic scale. These synoptic events are characterized by a lack of diurnal cycle in temperature and isotopic composition, and for the majority of the events by slightly higher daily mean temperatures. An indication of larger-than-averaged wind speeds coming from the west can also be observed in a majority of the cases. The lack of diurnal cycles in the isotopic composition is probably related to cloudy conditions, inhibiting diurnal variations in surface temperature. Further investigations are needed to understand the cause of these fluctuations, i.e., snow surface–air exchanges upstream, or changes in air mass origins, or exchanges with snowfall.

Particularly remarkable are periods of vapor with very high d-excess. Back trajectories relate these high d-excess events with air originating from the west and east of Greenland. Water tagging in one single atmospheric model confirms the relationship between high d-excess events and Arctic moisture sources as well. Future work will address this using a suite of models with different physical parameterization as well as by combining precipitation and vapor measurements. In line with earlier studies, we attribute the high d-excess events to large kinetic effects occurring at the sea-ice margin as dry air from over the sea ice is advected over the open water. It appears that the LMDZiso model may not correctly resolve these subgrid processes, and cannot capture the observed high d-excess events at NEEM.

The relationships between the fraction of Arctic moisture sources and NEEM d-excess needs to be further investigated. Our study is restricted to a few summer events. Similar processes may relate moisture sources formed at the sea-ice margin and high d-excess during winter. We speculate that the observed northward spatial d-excess gradient (Steen-Larsen et al., 2011; Masson-Delmotte et al., 2005) could partly be explained by a northward increase in the fraction of Greenland precipitation provided by Arctic moisture
sources. Furthermore, our results have implications for past temperature estimates from ice core data. We indeed suggest that past d-excess variations preserved in ice cores drilled at NEEM may inform on variations in Arctic moisture transported to NEEM.

Our new dataset demonstrates the value of continuous water vapor isotopic measurements for enhancing our understanding of the hydrological cycle of the Arctic, and assessing the ability of atmospheric general-circulation models to capture moisture origins. In a warmer Arctic, with a reduced sea-ice cover, changes in moisture sources indeed appear to be strongly model-dependent (Masson-Delmotte et al., 2011; Sime et al., 2013). Only detailed process studies, enabled by continuous monitoring, can help to reduce these uncertainties. Such monitoring efforts should be extended over longer timescales, in order to address seasonal variations as well as interannual variations and their relationship with weather patterns, on timescales relevant for an improved interpretation of ice core records.

Supplementary material related to this article is available online at: http://www.atmos-chem-phys.net/13/4815/2013/acp-13-4815-2013-supplement.pdf.


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