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Precipitation Water Stable Isotopes in the South Tibetan Plateau: Observations and Modeling*

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ABSTRACT

Measurements of precipitation isotopic composition have been conducted on a daily basis for 1 yr at Bomi, in the southeast Tibetan Plateau, an area affected by the interaction of the southwest monsoon, the westerlies, and Tibetan high pressure systems, as well as at Lhasa, situated west of Bomi. The measured isotope signals are analyzed both on an event basis and on a seasonal scale using available meteorological information and airmass trajectories. The processes driving daily and seasonal isotopic variability are investigated using multidecadal climate simulations forced by twentieth-century boundary conditions and conducted with two different isotopic atmospheric general circulation models [the isotopic version of the Laboratoire de Météorologie Dynamique GCM (LMDZiso) and the ECHAM4iso model]. Both models use specific nudging techniques to mimic observed atmospheric circulation fields. The models simulate a wet and cold bias on the Tibetan Plateau together with a dry bias in its southern part. A zoomed LMDZ simulation conducted with ~50-km local spatial resolution dramatically improves the simulation of isotopic compositions of precipitation on the Tibetan Plateau. Simulated water isotope fields are compared with new data and with previous observations, and regional differences in moisture origins are analyzed using back-trajectories. Here, the focus is on relationships between the water isotopes and climate variables on an event and seasonal scale and in terms of spatial and altitudinal isotopic gradients. Enhancing the spatial resolution is crucial for improving the simulation of the precipitation isotopic composition.

1. Introduction

The stable isotopes (18O and D) of meteoric waters have been widely used in hydrological and climatic studies. Since the initiation of the International Atomic Energy Agency/World Meteorological Organization’s (IAEA/WMO) Global Network of Isotopes in Precipitation
(GNIP, information online at http://www-naweb.iaeaea.org/nacflh/GNIP/IHS_GNIP.html) in 1961, several decades’ worth of data are available for a large number of stations, which, combined with theoretical one-dimensional (Ciais and Jouzel 1994) and three-dimensional models (Jouzel et al. 1994), provide improved our understanding of the mechanisms of stable isotopic fractionation in natural processes (Dansgaard 1964; Jouzel and Merlivat 1984; Jouzel et al. 1997). In the late 1980s, large gaps remained in the spatial coverage of GNIP stations, in particular on the Tibetan Plateau. This area acts as a “water tower” for surrounding lowlands; therefore, determining the controls on the isotopic composition of precipitation is crucial to understanding the origin of moisture and has implications for hydrological, tree-ring, and ice-core paleoclimate studies (Thompson et al. 2000; Tian et al. 2003; Yao et al. 2008).

There was thus an urgent need for improvements in our monitoring and understanding of modern precipitation isotopic composition over the Tibetan Plateau region. Since the 1990s, a network of about a dozen isotopic sampling sites has been established on the Tibetan Plateau, and has routinely operated on a monthly time step (Tian et al. 2003). These observations have revealed the regional complexity of geographical and climatic factors driving precipitation isotopic composition (Tian et al. 2003, 2007; Zhang et al. 2005; Kurita and Yamada 2008; Liu et al. 2008; Yu et al. 2008) and the imprint of airmass origins on this composition (Araguas-Araguas et al. 1998; Aizen et al. 2006; Tian et al. 2007).

In parallel with the development of monitoring networks, water stable isotopes have been implemented in atmospheric general circulation models (AGCMs) since the early 1980s (Jouzel et al. 1987; Hoffmann and Heimann 1997; Hoffmann et al. 1998; Mathieu et al. 2002; Noone and Simmonds 2002; Vuille et al. 2005; Schmidt et al. 2007; Lee et al. 2007; Yoshimura et al. 2008; Tindall et al. 2009; Risi et al. 2010), significantly improving our understanding of the processes driving precipitation isotopic composition (Joussame et al. 1984). Modeling water stable isotopes provides an independent evaluation of the modeled water cycle and climate. Moreover, the consistent physical framework of the model enables analysis of the climatic controls on the spatial and temporal precipitation isotopic compositions, on global or regional scales. Based on our present-day capabilities, isotopic AGCMs have also been used to explore the temporal stability of the isotope–climate relationship under varying paleoclimate boundary conditions (e.g., glacial climate, or changing orbital parameters) (Jouzel et al. 2000; Hoffmann et al. 2000; LeGrande and Schmidt 2009). While many of these studies focused on the polar ice caps and the interpretation of polar deep-drilling sites, they have revealed the importance of atmospheric circulation changes in the local relationships between changes in precipitation isotopic composition and climate, including the Asian area (e.g. LeGrande and Schmidt 2009).

For the Tibetan Plateau, comparisons between isotopic AGCMs and data remain a challenge. Earlier studies (Hoffmann et al. 1998; Yoshimura et al. 2003; Vuille et al. 2005) have shown that some models can capture the major observed spatial patterns, but perform better for short-term (daily or several days) isotope variability than on seasonal or interannual time scales. The challenge is first due to the difficulty in realistically capturing the basic features of the atmospheric circulation and precipitation in the area, including the correct airmass advection and local recycling patterns (Kurita and Yamada 2008). Recent studies have shown remarkable improvements in the realism of simulated precipitation (Shi et al. 2008; Nguyen and McGregor 2009). Model–data mismatches can be linked with the complex Tibetan Plateau topography (Chow and Chan 2009). Due to the current spatial resolution of AGCMs, they cannot capture the impacts of subgrid topography on regional circulation. The second challenge for Tibetan Plateau model–data comparisons is the lack of empirical observations on the Tibetan Plateau, which limits our knowledge of spatial and temporal precipitation isotopic composition gradients and, in particular, of the submonthly variability. The monthly mean isotopic composition is indeed produced by the juxtaposition of individual precipitation events. Filling observational gaps is therefore essential to better understand the processes at play and to evaluate the water cycle simulated by the atmospheric models.

The first goal of this paper is to reveal the daily and seasonal characteristics of isotopic compositions in precipitation and understand the links between isotopic compositions in precipitation and the moisture trajectory path to the south of the Tibetan Plateau, in relationship with monsoon dynamics (Sengupta and Sarkar 2006). For this purpose, we present new observations of precipitation isotopic composition from a ~1-yr (October 2007–November 2008) monitoring experiment conducted on an event basis at two locations, Bomi and Lhasa, both in the south area of the Tibetan Plateau. We analyze their relationships with local meteorological parameters. Then these relationships were confirmed by simulated moisture trajectory from the isotopic version of the Laboratoire de Météorologie Dynamique GCM (LMDZiso).

The second goal of this paper is to evaluate the model performance of stable isotopes in precipitation on different time scales along the southern Tibetan Plateau. We use three nudged simulations run with two isotopic AGCMs (LMDZiso at two different spatial resolutions and ECHAM4iso) to analyze the distribution and processes
controlling south Tibetan precipitation isotopic composition. The seasonal, annual, spatial, and altitudinal variations of the precipitation isotopic composition simulated by both models are compared with the monthly observations from three Tibetan stations (Bomi, Lhasa, and Tuotuohe, China) and two nearby stations (Mumbai and Shilong, India) influenced by the Asian monsoon.

In section 2, we describe the study area, the observational data used here, and the AGCM models enabled with water stable isotopes. In section 3, we discuss the relationships between the observed stable isotopic compositions and the meteorological data. In section 4, we compare observations with simulations from LMDZiso and ECHAMiso.

2. Material and methods

a. Study area

Located on the southeastern margin of the Tibetan Plateau, at the junction of the Nyainqentanglha and Himalaya Mountains, the Bomi area (29°21′–30°40′N, 94°00′–96°30′E) is a crucial region for studying moisture transport processes and the interaction of major circulation systems (southwest monsoon, westerly flow, and the Tibetan high pressure system; see Fig. 1).

Moving along the Brahmaputra River, the southern moisture flow reaches this area with the onset of the Indian monsoon, resulting in abundant rainfall. At Bomi, located in the valley of the Parlung Zangbo River (29°50′N, 95°47′E, 2730 m MSL), meteorological data show that the annual mean precipitation is ~835 mm (from 1961 to 2003) with evaporation reaching ~700 mm (from 1980 to 2003). Monthly mean precipitation amounts are higher than 50 mm from March to October and usually peak in June. The annual mean relative humidity is about 71%, reaching up to 80% in September. The annual mean temperature is 8.6°C, ranging from 0°C in January to 16.5°C in July (Fig. 2a). During our sampling year (from October 2007 to November 2008), the spring and summer (March–September) precipitation amount accounts for ~80% of the annual rainfall, similar to the long-term average precipitation percentage at Bomi recorded at the meteorological station. The average temperature (9.9°C) and relative humidity (72%) are also close to the long-term averages. We can therefore consider that this single year of precipitation isotopic composition observations represents a “normal monsoon” year.

Lhasa is located in a vast valley along the Brahmaputra River, to the west of Bomi by 570 km. The annual mean precipitation amount at the Lhasa meteorological station (29°42′N, 91°08′E, 3650 m MSL) is ~620 mm, and the area is dominated by monsoon precipitation, with about 85% of the total annual precipitation occurring as rainfall from May to September. The annual mean temperature is 7.5°C and the annual mean relative humidity is about 47%, showing that Lhasa is drier than Bomi outside of July–August (Fig. 2b). The interannual variabilities in monthly temperature and relative humidity are higher at Lhasa than are those at Bomi; while Bomi shows a larger variability of annual precipitation. During our sampling year (October 2007–November 2008), the rainfall from May to September accounts for 90% of the total annual precipitation; similar to the long-term average percentage of precipitation occurring during the monsoon period at Lhasa. At Lhasa, the precipitation isotopic composition has been monitored on an event basis since the 1990s (Tian et al. 2003). The annual weighted mean δ18O in precipitation calculated from our event-based sampling (see next section) (~16.9‰) is close to the multiyear weighted mean values (~17.4‰). Thus, our observation year is representative of a normal monsoon year and the isotopic composition of precipitation from Lhasa and Bomi is comparable to comparisons with other long-term datasets. These results will be discussed in section 3.

Here, we also refer to monthly mean observations available from different precipitation collection stations on and around the Tibetan Plateau: two stations in the IAEA’s GNIP network (Mumbai and Shilong) (data available from 1961 to 1978, downloaded from the Web site http://www-naweb.iaea.org/napc/ih/GNIP/IHS_GNIP.html) and Lhasa and Tuotuohe, meteorological stations (our observations; data available from 1996 to 1999; see Fig. 1).
Figure 2 shows the mean seasonal variations of meteorological conditions and observed monthly mean $\delta^{18}O$ in precipitation at these five stations. Mumbai (10 m MSL) is our southernmost site, with an annual mean temperature of 27.1°C and an annual precipitation amount of ~1985 mm, which is mostly (90%) provided by wet season precipitation (June–September). The climate at Shilong, located northward and at higher altitude (1598 m MSL), is colder (annual mean temperature of 14.7°C) and is also controlled by the summer monsoon, with a mean annual precipitation amount of ~2155 mm. As previously explained, Bomi and Lhasa are located in the valley near the Brahmaputra where monsoon moisture is transported toward the southern Tibetan Plateau (Lin and Wu 1990). Tuotuohe with the highest altitude (4533 m) is our northernmost, coldest (annual mean temperature of ~2°C), and driest (~260 mm yr$^{-1}$) site, as it is outside of the monsoon influence and is under the influence of continental air masses. The observed monthly mean weighted $\delta^{18}O$ in precipitation decreases from south to north, except for at Tuotuohe (Fig. 2c). In the southern Tibetan Plateau, the amount effect dominates the variation of precipitation $\delta^{18}O$ and the heavy isotopes undergo a gradual depletion along the northward monsoon transport. In the northern Tibetan Plateau, a different moisture origin is found and the variations in precipitation $\delta^{18}O$ are controlled by temperature (Tian et al. 2003, 2007). These monthly mean records will be discussed in detail in section 4 for comparison with isotopic simulations.

b. Precipitation sampling and data acquirement

Precipitation samples were collected on an event basis from October 2007 to November 2008 at the Bomi and Lhasa meteorological stations. Rainfall samples were immediately sealed in plastic bottles to prevent evaporation. Solid precipitation samples were collected and melted at room temperature, and then preserved as rainfall samples. All of the samples were stored frozen until analysis.

The isotopic composition of the 113 precipitation samples at Bomi was measured at the Laboratoire des Sciences du Climat et de l’Environnement, France. The $\delta^{18}O$ amounts were measured by equilibration on a MAT-252 mass spectrometer, with an analytical precision of 0.05‰, and $\deltaD$ was measured by reduction on uranium, with an analytical precision of 0.5‰. The $\delta^{18}O$ measurements of the 90 precipitation samples at Lhasa were performed with a MAT-253 mass spectrometer with GasBench-II (Thermo Scientific, Waltham Massachusetts) and the $\deltaD$ of 156 precipitation samples were measured by inductively coupled plasma mass spectrometry (ICP-MS; X-7 series, Thermo Scientific) at the Key Laboratory of the Chinese Academy of Sciences (CAS) for Tibetan Environment and Land Surface Processes, Institute of Tibetan Plateau Research, with a precision of 0.05‰ for $\delta^{18}O$ and 0.5‰ for $\deltaD$. 

FIG. 2. Mean seasonal cycle of meteorological conditions at five stations (red represents Mumbai, green represents Shilong, black represents Bomi, blue represents Lhasa, and magenta represents Tuotuohe): (a) monthly mean temperature (°C), (b) precipitation amount (mm month$^{-1}$), and (c) observed $\delta^{18}O$ (‰) in precipitation. Vertical error bars display the interannual monthly standard deviations. Purple lines are reference horizontal lines. The monthly mean data are available at Mumbai from 1961 to 1976, at Shilong from 1966 to 1978, and at Lhasa and Tuotuohe from 1996 to 1999.
δD. All of the measured results were calibrated with respect to Vienna standard mean ocean water (VSMOW). Interlaboratory δ18O comparisons were conducted with 53 samples and demonstrate the lack of systematic biases (the average of the difference is 0.02‰ and the standard deviation is 0.07‰), making possible the comparison of the spatial and temporal variations of δ18O in precipitation at Bomi and Lhasa.

The air temperatures and precipitation amounts were recorded for each precipitation event at both stations, through the average of the observed temperature before and after the precipitation event, and through the total precipitation amount for each event. Monthly meteorological data are also available from 1961 to 2003 (see section 2a).

c. AGCM simulations

Two isotopic GCMs (LMDZiso and ECHAMiso) are used in this paper to evaluate the performance of simulations and explore processes that control the isotopic composition of precipitation in India and the Tibetan Plateau.

LMDZ, developed at the Laboratoire de Météorologie Dynamique (LMD), is the atmospheric component of the L’Institut Pierre-Simon Laplace (IPSL) model. The model resolution is shown in Table 1. The dynamical equations are discretized on a latitude–longitude grid. Water in its vapor and condensed forms is advected by the Van Leer advection scheme (Van Leer 1977), which is a monotonic second-order finite-volume scheme. The physical parameterization scheme is described in detail in Hourdin et al. (2006). It includes the Emanuel convective parameterization (Emanuel 1991; Emanuel and Zivkovic-Rothman 1999) coupled to the (Bony and Emanuel 2001) cloud scheme. Risi et al. (2010) describe in detail the implementation of water stable isotopes in LMDZiso and the performances of the model for present and past climates. The representation of isotopic processes in the convection scheme is described in Bony et al. (2008). In this paper, we use three different simulations conducted with LMDZiso. The first is a climatological simulation run from 1979 to 2007 using the Atmospheric Model Intercomparison Project (AMIP) protocol (Gates 1992). In this simulation monthly interannually varying SSTs and sea ice, as well as a fixed CO2 value (348 ppm), are prescribed (Risi et al. 2010). The second run is a “nudged” simulation. Using the same boundary conditions as in the first run, the horizontal wind fields are nudged to the European Centre for Medium-Range Weather Forecasts (ECMWF) reanalysis results (Uppala et al. 2005). To do so, the simulated winds are relaxed in each grid box and at all levels toward the reanalyzed winds with a relaxation time of 1 h (Coindreau et al. 2007; Xin et al. 2008; Risi et al. 2010). In this case, the simulated large-scale wind field is consistent with the observations, making possible the comparison of simulated and observed daily or monthly meteorological conditions and precipitation isotopic compositions. The third run is an additional nudged simulation, as in the second, was performed using the zoom functionality of the LMDZ model. The grid was stretched so that the horizontal resolution was in the range of 50–60 km in a domain extending between 0° and 55°N and 60° and 130°E. This domain is larger than that in Fig. 1, so that each of our sites of interest is far enough from the zoom borders. This additional zoomed simulation was run over 3 yr (2005–07) to test the impact of spatial resolution on LMDZ outputs (see section 4). As in the second run, horizontal winds were nudged to the ECMWF reanalyses with relaxation times of 1 and 4 h outside and inside the zoom, respectively.

Here, we compare LMDZiso results with the results of another AGCM previously used for the monsoon area (Hoffmann and Heimann 1997). The ECHAM model has 19 vertical layers, from surface pressure up to a pressure level of 30 hPa, with a horizontal resolution of 2.8° × 2.8° (time step of 24 min; see Table 1). The model’s specific humidity and optional tracers are transported using a semi-Lagrangian advection scheme (Rasch and Williamson 1990). This model is documented in detail in Roeckner et al. (1996). The water isotope physics was implemented in a way similar to that in the Goddard Institute for Space Studies (GISS) AGCM (Jouzel et al. 1987), which is described in Hoffmann et al. (1998). It should be pointed out that there are four main differences between LMDZiso and ECHAMiso. First, LMDZ is a grid-point model whereas ECHAM is a spectral model, though it is combined with a semi-Lagrangian transport. Second, the ECHAM has a higher global horizontal resolution than LMDZ in the standard resolution, but a coarser one than LMDZ in the zoomed configuration. Third, LMDZiso explicitly calculates the evolution of raindrops composition as they evaporate into, and re-equilibrate with, the

<table>
<thead>
<tr>
<th>Model</th>
<th>Simulation date</th>
<th>Data type</th>
<th>Data No.</th>
<th>Model resolution</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard LMDZ</td>
<td>1 Jan 1979–31 Dec 2007</td>
<td>Daily</td>
<td>10 592 × 5</td>
<td>2.5° lat × 3.75° lon, 19 vertical layers</td>
</tr>
<tr>
<td>Zoomed LMDZ</td>
<td>1 Jan 2005–31 Dec 2007</td>
<td>Daily</td>
<td>1095 × 5</td>
<td>50–60 km, 19 vertical layers</td>
</tr>
<tr>
<td>ECHAM</td>
<td>Jan 1970–Dec 2001</td>
<td>Monthly</td>
<td>384 × 5</td>
<td>2.8° lat × 2.8° lon, 19 vertical layers</td>
</tr>
</tbody>
</table>
surrounding vapor following Stewart (1975), as detailed in Bony et al. (2008). In contrast, ECHAM prescribes a constant percentage of raindrops that re-equilibrates depending on two cloud types. Fourth, the evolution of both the rain and vapor compositions throughout the evaporation process is considered in LMDZiso, contrary to ECHAM (as detailed in Bony et al. 2008).

As for LMDZiso, we use here two different simulations: 1) A “free” 1970–2001 ECHAM run using monthly SST, sea ice, and greenhouse gas concentration datasets (Hurrell and Trenberth 1999) and 2) a nudged simulation where the 3D horizontal winds are also nudged toward the results of the 40-yr ECMWF Re-Analysis (ERA40; Uppala et al. 2005).

3. Observed characteristics of stable isotopes in precipitation in the south Tibetan Plateau

In this section, we first show the observed isotopic composition of precipitation at Bomi from 2007 to 2008 and its relationships with local meteorological data. We then compare the temporal distributions and seasonal cycles of the observations at Bomi and Lhasa to explore the different moisture origins.

a. Isotopic compositions of precipitation at Bomi and links with local weather variables

Figure 3 presents the temporal variations in the relative humidity, air temperature, precipitation amount, and precipitation $\delta^{18}$O of individual precipitation events collected at Bomi from October 2007 to November 2008 (see methods in section 2a). The $\delta^{18}$O values show significant seasonality and, at the event basis, some links with local weather parameters. During the dry winter, the most depleted $\delta^{18}$O values occur within the largest storm precipitation events. From November 2007 to May 2008, $\delta^{18}$O values vary between 5.5 and $-22.8\%_{\text{v}}$ and are positively correlated with temperature ($R = 0.58$, $n = 41$), in agreement with the isotopic temperature effect (Dansgaard 1964). From June to October, $\delta^{18}$O is progressively depleted, varying from $-0.5\%_{\text{v}}$ to $-29.1\%_{\text{v}}$. During this summer–autumn period, the lowest $\delta^{18}$O values correspond to the highest precipitation amounts, with a correlation coefficient of $-0.56$ ($n = 72$) between $\delta^{18}$O and precipitation amount. Notably, in March and April, marked by 200 mm of total precipitation, $\delta^{18}$O values show a positive correlation with temperature ($R = 0.47$, $n = 23$), a negative correlation with relative humidity ($R = -0.56$, $n = 23$), and no correlation with precipitation amount. In October, $\delta^{18}$O values are lowest and show no linear correlation with temperature or relative humidity but the strongest inverse correlation with precipitation ($R = -0.72$, $n = 14$).

All of these correlation coefficients are significant at a level exceeding 99% and demonstrate that, at the event basis, different meteorological factors affect the variations in Bomi precipitation $\delta^{18}$O in different seasons.

Additional deuterium measurements enable discussion of the local meteoric water line and the deuterium excess variability. The local meteoric water line (LMWL) at Bomi is calculated as $\delta D = 7.85 \delta^{18}O + 4.27$ ($R^2 = 0.98$) (Fig. 3b). The observed slope at Bomi is close to the
slope of 8 for the global meteoric water line (Yurtsever and Gat 1981) and suggests a limited impact of non-equilibrium processes on precipitation. This value is also very close to the slope of LMWL at Lhasa (7.9), as well as other observed values reported by Tian et al. (2001). The intercept of 4.27 is rather low, compared to values of 10 in the global precipitation data, ~8.0 at Mumbai and Shilong (Gupta et al. 2005), and 6.2 in LMWL at Lhasa (Tian et al. 2001). This low intercept reflects high relative humidity at Bomi’s moisture source (Clark and Fritz 1997).

The d excess (d = δD − 8δ^{18}O; Dansgaard 1964) of precipitation events varies between −8.9‰ to 22.3‰. The annual precipitation weighted mean d excess (9.8‰) is close to the d excess of precipitation at Kolkata (Sengupta and Sarkar 2006). For the whole dataset, d excess is negatively correlated with Bomi temperature, with a correlation coefficient of −0.51 (slope = −0.75‰ °C⁻¹, n = 109, p < 0.0001) and positively correlated with the precipitation amount (R = 0.42, slope = 0.50‰ mm⁻¹, n = 109, p < 0.0001). There is no significant correlation between the d excess of events and temperature, precipitation amount, and relative humidity in winter, spring, and autumn. In summer, the d excess of events shows a positive relationship with precipitation amount (R = 0.49, slope = 0.77‰ mm⁻¹, p < 0.0001) and does not show any correlation with local relative humidity. Seasonal shifts in deuterium excess reflect the inflow of monsoon moisture (Fig. S1 in the supplemental material), with May–September levels below 10‰.

The seasonal variations appear more clearly from monthly precipitation weighted data (Fig. 4). The monthly δ^{18}O increases with temperature from February to May, and shows an inverse correlation with precipitation amount from June to October. Apparently, the links between weather and δ^{18}O are therefore consistent from the event base to the monthly scale at Bomi: precipitation δ^{18}O appears to be mostly driven by changes in the local precipitation amount during the monsoon season and by local temperature changes during the rest of the year. Since d excess is affected by the relative humidity, temperature, and wind speed during primary evaporation (Merlivat and Jouzel 1979), it is expected to reflect different moisture levels during the monsoon season and nonmonsoon seasons in this region (Tian et al. 2007; Xu et al. 2009). It has poor correlation with latitude (Rozanski et al. 1992). Starting from high winter–spring values (~11‰), the monthly precipitation-weighted d excess abruptly decreases in May to reach a low ~5‰ level that persists until September, close to the d excess of surface water near the Bay of Bengal coast in eastern India (Gupta et al. 2005). After September, d excess increases while there is no clear shift in δ^{18}O.

The seasonal cycle of the Bomi precipitation isotopic composition can be interpreted as resulting from different air masses: during late autumn–spring, an inflow of continental (the westerlies and local convection; see Fig. S2 in the supplemental material) air masses is marked by a temperature control of δ^{18}O and high d-excess values; during summer and early autumn, the monsoon flow is marked by a precipitation effect of δ^{18}O, a progressive depletion of δ^{18}O, and low d-excess values. In May and September–October, the isotopic data suggest a simultaneous occurrence of continental and monsoon moisture transport systems.

b. Comparison of precipitation δ^{18}O at Bomi and Lhasa

The main monsoonal moisture flow is transported along the Brahmaputra River toward the southern Tibetan Plateau (Gao et al. 1985; Lin and Wu 1990; Tian et al. 2003, 2007). Here, we compare the precipitation δ^{18}O
sampled in parallel at these two locations on an event basis to estimate the difference in moisture transport.

At each station, the number of precipitation samples (Table 2) depends on the precipitation frequency and the seasonal precipitation distribution. At Lhasa, precipitation events primarily take place in summer (June–September, JJAS), while Bomi precipitation events are more evenly distributed, mostly from March to October. Although Lhasa and Bomi are located at the same latitude and within 570-km distance, Lhasa received 128.5 mm less precipitation during our sampling period (Table 2, Fig. 5), consistent with the climatological averages (see section 2a). Two main reasons can be invoked to explain this difference: 1) earlier onset and later retreat of monsoon flow at Bomi than at Lhasa and 2) stronger continental recycling and convection near Bomi (with higher relative humidity), providing larger precipitation amounts.

At both stations, the seasonal distribution of precipitation \( \delta^{18}O \) is characterized by peak \( \delta^{18}O \) values in spring (March–May, MAM) and minimum summer \( \delta^{18}O \) values (JJAS) (Fig. 5a). During our observation period, the summer precipitation amount is more than 100 mm higher at Lhasa compared to Bomi, with 57% of the total summer precipitation amounts at Lhasa and 68% at Bomi occurring on the same days. This suggests the influence of common large-scale processes. During summer, \( \delta^{18}O \) values are systematically 5.3‰ more depleted at Lhasa than those of simultaneous precipitation events sampled at Bomi. This could be linked to the orographic uplift of monsoon flow, resulting in deeper convection at Lhasa, located about 1000 m higher than Bomi.

Figure 5b displays the relationships between summer precipitation \( \delta^{18}O \) and precipitation amount at the monthly scale and at the event basis (Fig. 5c). On a monthly scale, the isotopic depletions at Bomi and Lhasa (\(-9.06\)‰ and \(-8.07\)‰ from June to August) are comparable despite a factor of 2 in the precipitation amounts at these 2 locations; being much higher at Lhasa. There is a general inverse correlation between monthly mean \( \delta^{18}O \) and precipitation amount at Lhasa. In particular, the difference in the relationship between precipitation \( \delta^{18}O \) and precipitation amount at these two stations in June and July may indicate more complicated moisture origins at Bomi. At the event scale, the “amount effect” is much stronger at Lhasa than at Bomi, both in terms of correlation coefficient and in terms of the \( \delta^{18}O \)-precipitation slope; precipitation \( \delta^{18}O \) does not show an obvious correlation with daily temperature during our observation period at Lhasa. In the summer (JJAS), the precipitation \( \delta^{18}O \) has a remarkable negative correlation with precipitation amount, with a coefficient of \(-0.56 (n = 62, p < 0.0001) \) which is better than that of Bomi (\( R = -0.26, n = 57, p < 0.05 \)), while the slope of the \( \delta^{18}O \) precipitation is \(-0.51 \) at Lhasa, lower than that at Bomi (\(-0.34 \)). The annual mean \( \delta^{18}O \) excess at Lhasa (7.9‰) is

<table>
<thead>
<tr>
<th>Stations</th>
<th>Obs periods</th>
<th>Alt (m)</th>
<th>No. of samples</th>
<th>Total precipitation (mm)</th>
<th>No. of summer events</th>
<th>Precipitation amount in summer (mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bomi</td>
<td>21 Oct 2007 to 22 Nov 2008</td>
<td>2730</td>
<td>113</td>
<td>652.5</td>
<td>57</td>
<td>292.6</td>
</tr>
<tr>
<td>Lhasa</td>
<td>9 Oct 2007 to 4 Nov 2008</td>
<td>3658</td>
<td>90</td>
<td>524</td>
<td>62</td>
<td>427.3</td>
</tr>
</tbody>
</table>

Fig. 5. (a) Comparison of daily \( \delta^{18}O \) in precipitation and precipitation amounts observed from October 2007 to November 2008 at Bomi (black) and Lhasa (gray). Relationships between summer precipitation \( \delta^{18}O \) and precipitation amounts (b) on a monthly scale and (c) on an event basis.
about $2^{\%}$ lower than that at Bomi. These differences may be explained by different moisture sources. Lhasa’s moisture is mostly provided by monsoon flow, while Bomi receives moisture provided by different moisture advection paths (the westerlies, local recycling, and monsoon flow) resulting in a subdued precipitation amount effect. Long-term monthly observations available for Lhasa reveal that interannual precipitation $\delta^{18}$O levels are also driven by the amount effect (Tian et al. 2003).

4. Comparison of observations with simulations from LMDZiso and ECHAMiso models

The GNIP network initiated in 1961 provides long-term monthly data for a number of sites in the world. Through monitoring stations initiated in the 1990s, a multiyear dataset is available for some sites on the Tibetan Plateau. Here, we use three of these Tibetan sites (Lhasa, Bomi, and Tuotuohe) and two GNIP sites located in India (Mumbai and Shilong) with multiyear records (Fig. 1), previously described in section 2, to compare the observed seasonal patterns with LMDZiso and ECHAMiso model outputs. For each model, we also compare two types of simulations, the free versus nudged simulations. The averaged data over the closest grid box compared with the observed data from monitoring stations. Note that the scale mismatch may contribute to the difference between observations and simulations.

a. Spatial distribution of annual mean isotopes in precipitation in India and on the Tibetan Plateau

Risi et al. (2010) have documented the validity of global simulations from LMDZiso at different scales. Figure 6 shows maps of simulated annual mean $\delta^{18}$O and d excess in precipitation in India and along the Tibetan Plateau. The nudged ECHAMiso and LMDZiso results do capture the spatial pattern of precipitation $\delta^{18}$O and the increasing northward isotopic depletion (Table 3), although there are mismatches in some regions. The standard LMDZiso (hereafter LMDZiso$\text{ctrl}$) and ECHAM simulate annual mean $\delta^{18}$O values close to the observations below 2000 and above 4500 m, but they fail to capture the strongly depleted $\delta^{18}$O values between 2000 and 4500 m (e.g., Bomi and Lhasa). This may be due to too a weak history of distillation upstream resulting from the underestimated precipitation on the foothills. Satisfactorily, the zoomed LMDZiso (hereafter LMDZiso$\text{hr}$) successfully simulated the lower $\delta^{18}$O along the Himalayas and the Tibetan Plateau as the most depleted $\delta^{18}$O region, matching better with the observations. This demonstrates that the improved spatial resolution is crucial for realistic simulations. Thanks to the high-resolution topography representation, the air masses undergo more distillation before arriving on the Tibetan Plateau (discussed in detail in section 4b). Thus, the zoomed annual mean $\delta^{18}$O values are lower than the standard simulations, with a depletion of $\sim 5^{\%}$ and match the observed $\delta^{18}$O in precipitation.

Simulations for both models show the “temperature effect,” amount effect (Risi et al. 2010), and “altitude effect” used to reconstruct the history of the Himalayan–Tibetan Plateau uplift, which is critical to understand past Asian monsoon evolutionary patterns (Ramstein et al. 1997). Precipitation isotopic data collected at these five stations suggest a $\delta^{18}$O versus elevation gradient of $-0.3^{\%} (100 \text{ m})^{-1}$ ($R^2 = 0.48$; see Fig. 7), in agreement with previous Tibetan Plateau analyses (Hou et al. 2003; Tian et al. 2008). The overall simulated $\delta^{18}$O–altitude relationships are rather consistent with the observations, with mean slopes of $-0.2^{\%} (100 \text{ m})^{-1}$ for all simulations. In winter, the observed $\delta^{18}$O–altitude relationship is significant in this region with a slope of $-0.6^{\%} (100 \text{ m})^{-1}$, and all simulations underestimate this $\delta^{18}$O–altitude relationships with slopes of $-0.3^{\%} (100 \text{ m})^{-1}$ (even lower for LMDZiso$\text{hr}$ simulation). In summer, all simulations again underestimate the $\delta^{18}$O–altitude relationship with slopes of $-0.1^{\%} (100 \text{ m})^{-1}$ from LMDZiso$\text{ctrl}$, $-0.15^{\%} (100 \text{ m})^{-1}$ for LMDZiso$\text{hr}$, and $-0.2^{\%} (100 \text{ m})^{-1}$ from ECHAM, all of which are less than the slope of $-0.3^{\%} (100 \text{ m})^{-1}$ obtained from observations. At the locations of the Tibetan Plateau monitoring stations, models have a higher elevation on average (because they cannot resolve the subgrid valley topography) and strongly overestimate the local precipitation amount (Table 3); it is therefore surprising that the models capture reasonably well the overall elevation gradient. Finally, the model results overestimate the elevation effect for the high-altitude sites (between 2000 and 5000 m) with large simulated slopes $[-0.5^{\%} (100 \text{ m})^{-1}]$. While clearly improving the seasonal cycle of precipitation isotopic composition in the Tibetan Plateau, the zoomed simulation does not capture the right amplitude of the altitude effects on d excess or $\delta^{18}$O, especially in winter. The fact that observed slopes differ at low and high elevations, and that models underestimate elevation impacts at middle elevations but overestimate these elevation impacts for high-elevation locations, clearly brings into question the use of isotopic AGCMs to refine paleo-elevation reconstructions.

The spatial distribution of the annual mean d excess in precipitation is not as well simulated as $\delta^{18}$O in all of the models. Both LMDZiso$\text{ctrl}$ and ECHAM overestimate the d excess in precipitation around the Himalayas and show a stable d-excess spatial distribution, but LMDZiso$\text{hr}$ underestimates the d excess in this region while also strongly overestimating the d excess at Mumbai (Fig. 6). ECHAM reproduces the d-excess altitude relationship
FIG. 6. Spatial variations of $\delta^{18}$O and d excess in precipitation from observations and three simulations (ECHAM4iso, standard LMDZiso, and zoomed LMDZiso) in study region. Annual mean data are displayed.
more reasonably with a slope of 0.24\%\textsubscript{o}(100 m\textsuperscript{-1}) \((R = 0.92)\) while the observed d-altitude slope is 0.12\%\textsubscript{o}(100 m\textsuperscript{-1}) \((R = 0.78)\). The standard and zoomed d excesses fail to reproduce this relationship.

b. Seasonal cycles of stable isotopes in precipitation in India and on the Tibetan Plateau

Figure 8 shows the simulated seasonal cycles of temperature, precipitation amount, \(\delta^{18}O\), and d excess in precipitation over five GNIP stations (including our new measurements for Bomi, Lhasa, and Tuotuohe). We can therefore compare the modeled results with the observations (Figs. 2 and 8). From south to north, both models reproduce the seasonal patterns of temperature to some extent, despite systematic biases. The nudged LMDZiso (hereafter LMDZiso\textsubscript{ctrl}) simulation is 6°C too warm during the spring at Shilong, but underestimates Tibetan Plateau temperatures by up to 10°C. The comparison between the free AMIP simulation and the nudged simulation reveals a slight improvement in the seasonal cycle of temperature with the nudging at Mumbai and Shilong, which is consistent with the results of Yoshimura et al. (2008). In contrast, the nudging deepens the model cold bias from LMDZiso for the Tibetan Plateau. This means that the LMDZiso model biases are not linked with the large-scale atmospheric dynamics and altitude difference (Table 3) but to model parameterizations, which can include, for instance, land surface properties (albedo and latent and sensible heat fluxes), boundary layer characteristics, and/or cloud properties (Kripalani et al. 2007a; Kripalani et al. 2007b). ECHAM performs relatively well for temperatures at Mumbai and Shilong, but also shows a 7°C systematic cold bias for the Tibetan Plateau. With ECHAM, the nudging slightly enhances the cold winter–spring bias at Lhasa, but induces much warmer temperatures in Tuotuohe, in better agreement with the observations. For this location, driving the model toward the observed large-scale wind fields clearly improves the realism of the simulated climate.

Regarding the precipitation seasonal cycle, nudged simulations of both models capture the pattern of the precipitation seasonal cycle, but strongly underestimate the precipitation amounts (by 60%–70% for LMDZiso

Table 3. Summary of observed–modeled isotope and meteorological data from the standard LMDZ model for five stations.

<table>
<thead>
<tr>
<th>Station</th>
<th>Latitude (°N)</th>
<th>Longitude (°E)</th>
<th>Altitude (m)</th>
<th>Obs Period</th>
<th>LMDZiso\textsubscript{ctrl} Period</th>
<th>ECHAM Period</th>
<th>Annual Temperature (°C), Obs</th>
<th>Simulated by LMDZiso\textsubscript{ctrl}</th>
<th>Simulated by ECHAM</th>
<th>Annual Rainfall (mm), Obs</th>
<th>Simulated by LMDZiso\textsubscript{ctrl}</th>
<th>Simulated by ECHAM</th>
<th>Weighted (\delta^{18}O) (‰), Obs</th>
<th>Simulated by LMDZiso\textsubscript{ctrl}</th>
<th>Simulated by ECHAM</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mumbai</td>
<td>18°54’N</td>
<td>72°49’E</td>
<td>10 (R = 0.92)</td>
<td>1960–78</td>
<td>1979–2007</td>
<td>1970–2001</td>
<td>27.1</td>
<td>26.3</td>
<td>29.4</td>
<td>1984.8</td>
<td>453.6</td>
<td>-1.51</td>
<td>-1.12</td>
<td>-1.558</td>
<td></td>
</tr>
<tr>
<td>Bomi</td>
<td>29°51’N</td>
<td>95°47’E</td>
<td>2730 (R = 0.78)</td>
<td>2008</td>
<td>1979–2007</td>
<td>1970–2001</td>
<td>12.2</td>
<td>0.1</td>
<td>2.1</td>
<td>826.4</td>
<td>1288.4</td>
<td>-11.81</td>
<td>-6.55</td>
<td>-9.659</td>
<td></td>
</tr>
<tr>
<td>Lhasa</td>
<td>29°42’N</td>
<td>91°08’E</td>
<td>3658 (R = 0.78)</td>
<td>1996–2004</td>
<td>1979–2007</td>
<td>1970–2001</td>
<td>7.7</td>
<td>4.5</td>
<td>-1.7</td>
<td>627.3</td>
<td>2083.8</td>
<td>-4.5</td>
<td>-9.65</td>
<td>-10.6024</td>
<td></td>
</tr>
</tbody>
</table>

FIG. 7. Annual mean (filled circles), summer (black triangles), and winter (white triangles) variations of precipitation-weighted \(\delta^{18}O\) for observations (black), LMDZiso\textsubscript{ctrl} (red), LMDZiso\textsubscript{hr} (green), and ECHAM (blue) as a function of site elevation. Note that, due to the model grids, elevations for the geographical locations of Bomi, Lhasa, and Tuotuohe are systematically higher in the model “averaged topography” than the actual elevation of the valleys where the observations are conducted. The lower elevation for the coarser-resolution simulations (in apparent better agreement with the data) reflects the underestimated average elevation of the Tibetan plateau at coarse resolution.
and 0% – 75% for ECHAM, respectively) in India, while both models overestimate the precipitation amounts on the Tibetan Plateau. For LMDZ, the nudging clearly improves the hydrological cycle compared to the free simulation, which is even drier at Shilong and wetter in Tibet. For ECHAM, the nudging enhances the model wet bias in summer for Bomi and Lhasa. Overestimating the precipitation over high-elevation regions and underestimating it over nearby plains is a widespread bias in GCMs (Codron and Sadourny 2002).

The temperature in the zoomed simulation is not remarkably improved at Mumbai and Shilong, but it is slightly closer to the observations at the Tibetan stations, compared with simulations from LMDZiso for 31 yr (1970–2000), and from ECHAM for 3 yr (2005–07), and from ECHAM for 3 yr (1970–2000). The lack of Mumbai data arises from a shortage of sampled precipitation and data in the IAEA database, as well as in model outputs.

FIG. 8. Nudged multiyear monthly averages of climatology and stable isotopic compositions of precipitation (from left to right: air temperature, precipitation amount, $\delta^{18}$O, and $d$ excess in precipitation) observed (black) from south to north and compared with simulations from LMDZiso (red, standard simulation; green, zoomed simulation) and ECHAM (blue). Solid lines show the nudged simulations, while dashed lines show the free simulations. Note that, for readability, different vertical-axes scales have been used for $\delta^{18}$O and $d$ excess. We used the standard simulations from LMDZiso for 29 yr (1979–2007), zoomed simulations from LMDZiso for 3 yr (2005–07), and from ECHAM for 31 yr (1970–2000). The lack of Mumbai data arises from a shortage of sampled precipitation and data in the IAEA database, as well as in model outputs.
that from LMDZisoctrl ($R = 0.50, p < 0.0001$). The model performs better for precipitation on the Tibetan Plateau than for India. The zoomed model simulates more annual precipitation at Mumbai and Shilong. Although there are differences between the simulated and observed variations of precipitation during spring and summer, zoomed simulations can realistically reproduce the seasonal precipitation patterns at Bomi and Lhasa. At Tuotuohe, the precipitation simulated by LMDZ is almost coincident with the observations.

To summarize, both models exhibit a summer dry bias for India together with a year-round cold and wet bias for the Tibetan Plateau. Part of the model–data mismatches may arise from the model resolutions and the difficulties involved with capturing the complex influences of the south Tibetan topography, but the comparison between the nudged and free simulations reveals that they cannot be explained by large-scale dynamics and, therefore, point to the model physical parameterizations. Consistent with earlier model intercomparisons, ECHAM seems to perform better than LMDZ regarding monsoon precipitation (Kripalani et al. 2007a,b).

The performances of LMDZ and ECHAM in simulating stable isotopes in precipitation are also scrutinized (Fig. 8). Despite temperature and precipitation biases, LMDZ and ECHAM do capture the seasonal pattern of precipitation $\delta^{18}O$ to some extent. At Mumbai, both models show $\delta^{18}O$ seasonal cycles consistent with the observations, with minimum values in late summer–early autumn. At Shilong, the ECHAM model captures correctly the seasonal variations of $\delta^{18}O$ in precipitation, but slightly underestimates the isotopic depletion observed in September. For this location, LMDZisoctrl shows too flat a seasonal cycle and underestimates the summer depletion by about 5‰: this was expected from the model dry bias for this area. On the Tibetan Plateau (Bomi and Lhasa), LMDZisoctrl shows rather stable isotopic values all year round and fails to capture the observed seasonality. Thanks to a better simulation of the monsoon precipitation on the Tibetan Plateau, ECHAM does reasonably capture the summer’s depleted $\delta^{18}O$ values, which are affected by the amount effect, but, similarly to LMDZctrl it overestimates the winter $\delta^{18}O$. Finally, both models capture correctly the pattern of $\delta^{18}O$ at Tuotuohe, but again exhibit too high isotopic values in winter and not enough depletion in late summer (August–September). The nudged seasonal cycles of $d$ excess are similar for both models at Shilong, Bomi, and Lhasa, albeit with different simulated $d$-excess levels in October and in November. Compared with the observations, both models capture the seasonal pattern and amplitude of $d$ excess at Mumbai and Lhasa, but again fail to reproduce the seasonal $d$-excess cycle at other stations. At Tuotuohe, the ECHAM simulates significantly higher $d$ excess ($3\%$ ~ $10\%$) than that of LMDZisoctrl all year round, while the simulated $\delta^{18}O$ values are similar.

For ECHAM, the nudged simulation clearly improves the realism of the simulated precipitation $\delta^{18}O$, especially on the Tibetan Plateau (Figs. 8k to 8o). Both the free and standard LMDZ simulations produce too flat precipitation $\delta^{18}O$ seasonal cycles, and the nudging even reduces the magnitude of the simulated seasonal cycle. The nudging however improves the simulation of $d$ excess for both models, especially at Mumbai and Shilong.

The LMDZisoctrl simulation dramatically improves the forecasts of precipitation $\delta^{18}O$ and $d$-excess patterns, especially on the Tibetan Plateau. On the Tibetan Plateau, the zoomed simulations of precipitation $\delta^{18}O$ are very close to the observations, with a correlation coefficient of 0.86 ($p < 0.0001$). In fact, the zoomed simulation induces a very large precipitation amount at Shilong, which enhances depletion, leading to low precipitation $\delta^{18}O$ on the Tibetan Plateau in agreement with the observations. The fact that $\delta^{18}O$ is significantly improved in LMDZisoctrl compared to LMDZisoctrl, while temperature and precipitation are not, confirms that the isotopic composition of water constitutes an added value to investigate the water cycle compared to meteorological measurements only, since it reflects the dehydration history and processes along trajectories that are not readily identifiable in meteorological data (Worden et al. 2007; Galewsky and Hurley 2010). However, the zoomed simulation does not improve the seasonal cycle of $d$ excess, which is controlled by factors other than $\delta^{18}O$.

The performance of the models in capturing the seasonality of precipitation $\delta^{18}O$ and $d$ excess is obviously affected by model biases for temperature and precipitation, albeit in a complex manner. Given their coarse spatial resolutions, models do not capture intense storms associated with strong isotopic depletion, and despite the nudging with large-scale observed wind fields, may not represent correctly the seasonal shifts in airmass trajectories. Compared with other Tibetan stations, the models perform better for precipitation isotopic composition where the monsoon influence is weakest. This indicates that the simulation of the monsoon is pivotal to the accurate simulation of the seasonal variations for both models.

c. Comparison with daily observations at Bomi and Lhasa

Because the nudged ECHAM simulation does not cover 2008, we have to restrict this model–data comparison
to LMDZiso. The LMDZiso nudged simulation offers the possibility of comparing the modeled day-to-day variability with the Bomi and Lhasa data between January and December 2008 (Fig. 9). At the event scale, LMDZisoctrl strongly overestimates the magnitude of the Bomi and Lhasa precipitation amounts and simulates systematically too high $\delta^{18}O$ values, with very weak seasonal amplitude. The model does not capture the high spring $\delta^{18}O$ values and the magnitude of summer depletion. LMDZisohr dramatically improves the event simulation of precipitation and precipitation $\delta^{18}O$ with correlation coefficients of 0.3 and 0.71 ($p < 0.0001$), respectively, which are much higher than those from LMDZisoctrl (0.09 for precipitation amount and 0.34 for precipitation $\delta^{18}O$, $p < 0.0001$). The zoomed simulation reproduces the observed seasonal magnitudes with spring high $\delta^{18}O$ and summer low $\delta^{18}O$ values (Figs. 9b and 9f). Although the zoomed model overestimates the precipitation amount in spring and autumn, the simulation reproduced heavy precipitation events at Bomi; while the magnitude of the heavy precipitation is not reproduced and LMDZisohr simulates too much precipitation during the spring and autumn. The zoomed model still cannot capture the exact seasonal amplitude of temperature, underestimating its range by more than 1°C (Figs. 9a and 9e). The zoomed $d$ excess does not capture the pattern of observed $d$-excess variation, especially in spring, and underestimates the magnitude of the $d$-excess fluctuations.

The lack of model $\delta^{18}O$ depletion in LMDZisoctrl during summer is unlikely to be linked with interannual variability, which is very small in LMDZ outputs, and is further supported by the comparison of LMDZ with the longer Lhasa records. The lack of summer–autumn depletion and $d$-excess seasonality can be linked to different processes, such as the model condensation processes, or the model moisture source and transport.

5. Conclusions

The local climate, together with the origin and transport of moisture and precipitation processes, are crucial factors that affect the spatial and temporal variabilities of stable isotopes in precipitation. In this paper, we have shown new results from two stations along the southeastern Tibetan Plateau, at Bomi and Lhasa, where precipitation sampling was achieved on an event basis. The links with local precipitation and temperature have been analyzed at the event and on a monthly basis and these show rather consistent results. The complex seasonal cycle of precipitation $\delta^{18}O$ in this area cannot be understood without taking into account airmass trajectories, which we did using backward trajectories calculated.
from the LMDZ model. Episodic winter precipitation events mostly arise from the westerlies. During the summer, the monsoon produces the lowest $\delta^{18}O$ values both at Bomi and Lhasa. In spring, different moisture transports result in differences in precipitation $\delta^{18}O$ at Bomi and Lhasa. Similarities between Bomi and Lhasa are observed in summer and point to the advection of monsoon flow from Bomi to Lhasa along the Brahmaputra River valley, a regional flow pattern that cannot be resolved by low-resolution AGCMs.

Our study has shown that the amount effect is observed with the monsoon’s onset, which relates high precipitation amounts with strongly depleted $\delta^{18}O$, and explained the links between shifts in moisture origin and southern Tibetan $\delta^{18}O$ values, in particular the highest $\delta^{18}O$ levels in spring and the most depleted levels in summer. Further quantitative understand could be investigated in simulations including the tagging of moisture sources (Delaygue et al. 2000; Masson-Delmotte et al. 2008).

The available observations now enable systematic comparisons with isotopic AGCMs regarding the isotopic variability along precipitation events, spatial patterns, and seasonal cycles. Both LMDZisoctrl and ECHAMiso share common biases in this complex Tibetan Plateau region, notably deficient precipitation at low latitudes and too-wet and too-cold conditions on the Tibetan Plateau. They generally simulate stronger correlations of $\delta^{18}O$ with precipitation and temperature than are observed [something that has been observed before; Hoffmann et al. (2005)], and underestimate the variability of $\delta^{18}O$ at seasonal and event scales. In particular, LMDZisoctrl underestimates the seasonal amplitude of $\delta^{18}O$ in the southern Tibetan Plateau, probably in relationship to the upstream processes and a dominant westerly flow; while generally improving the simulated precipitation amount, the nudging enhances the model cold bias. Nudging ECHAM amplifies the difference in the precipitation amounts at Bomi and Lhasa, but it is effective at producing realistic simulated $\delta^{18}O$ results in precipitation in our study region. In winter, both models overestimate the effects of temperature and precipitation on $\delta^{18}O$. In summer, both models underestimated the amount effects, especially at Mumbai and Shilong, where they are too dry. In general, ECHAM performs better than LMDZiso on the regional climate and precipitation isotopic composition. The model bias however remains significant, typically less than 4 per mille for ECHAM but reaching 10 per mille for LMDZiso. Although ECHAM is able to simulate reasonably well the amount effect and the spatial distribution of the precipitation $\delta^{18}O$, it still fails to capture the southern Tibetan plateau $\delta^{18}O$ seasonal cycle and winter values.

The LMDZisoctrl version makes simulations more realistic thanks to an improved representation of regional topography. Although it is too wet at Shilong, the Tibetan precipitation is much closer to the observations. The simulation of daily and monthly precipitation $\delta^{18}O$ is more reasonable and shows realistic seasonal variations. The improved spatial resolution is therefore crucial for correct simulations of Tibetan Plateau precipitation and $\delta^{18}O$.

The overall simulated $\delta^{18}O$ versus elevation gradient is within the range of observations, although it is underestimated in winter and at low elevations. Both LMDZctrl and ECHAM show wet and cold biases between 2000 and 4500 m. At high altitudes, models simulate steeper-elevation $\delta^{18}O$ than is observed. Because of changing $\delta^{18}O$-elevation slopes with altitude, we are suspicious of the reliability of estimating the paleo-elevation based on the oxygen isotopic composition of precipitation.

This study shows the potential of Tibetan Plateau precipitation isotopic composition databases for testing the realism of the water cycle simulated by isotopic AGCMs and the importance of spatial resolution for realistic simulations. There is a strong need to maintain long-term observations at the monthly time frame, to improve the spatial coverage of the datasets, and to develop event-based studies that are necessary to better understand the exact processes at play, including, in the future, the parallel monitoring of water vapor isotopic composition. Our comparison with two AGCMs shows very large systematic biases in the AGCMs regarding both the regional climate and the properties of water stable isotopes in precipitation. The complex topographic features add to the difficulties in correctly simulating moisture advection from the lowlands to the Tibetan Plateau. Second-order isotopic data such as deuterium excess and oxygen 17 are expected to bring further constraints on the water cycle, moisture origin, local recycling, and convective processes. However, models that can correctly capture these second-order tracers remain a future research effort due to the large biases in second-order isotopic data, which are probably linked to land surface, convection, and cloud representations. Nudging does not resolve these parameterization problems but helps to build a framework within which the synoptic-scale atmospheric dynamics is realistic.

While isotopic AGCMs have previously been used to assess the stability of the isotope–climate relationships at polar locations, there is still much to be done before they can be used with confidence for improving paleoclimate reconstructions conducted with water stable isotopes in the Tibetan Plateau area or for providing constraints on paleo-elevation. Systematic comparisons between isotopic AGCMs should be expanded to other
models than the two AGCMs explored here. The major model–data mismatches shown here are also strong limitations for confidently forecasting the regional impacts of increasing greenhouse gas concentrations. This is particularly frustrating when considering the societal importance of the southern Tibetan Plateau and its glaciers, which serve as a “water tower” for some of the world’s largest rivers and provide water to many surrounding countries.

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