

**$^2\text{H}/\text{H}$  and  $^{18}\text{O}/^{16}\text{O}$  Non-Equilibrium Fractionation Factors for Ocean Evaporation in the North-West Atlantic Region**

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**Introduction**

In this Supporting Information file are reported all the additional explanations, table and figures regarding methods and results that are not critical for the outcome of the main manuscript but might help providing a clearer picture of the study to the reader. In order, the supporting file includes:

**Text S1, Table S1:** Values and references used to estimate the isotopic composition of the ocean around Bermuda

**Text S2, Figure S1:** Explanation of the Craig Gordon model focusing on non-equilibrium fractionation factors.

**Text S3:** Explanation of the algorithm used to estimate non-equilibrium fractionation factors

**Text S4, Figures S2 – S4:** Effect of data filtering on PDFs of main variables

**Figure S5:** scatterplot of  $k_2$  vs 10-m wind speed

### Text S1: Isotopic composition of the ocean.

No measurements of ocean water isotopic composition near the study site are available for the time period of interest. However, the temporal variability of ocean isotopic composition in the study area is expected to be very low. Several sources have been considered to estimate the most representative composition of ocean water around the study site: gridded dataset, North Atlantic cruises published data as well as from samples collected at the Bermuda Atlantic Time-series Study (BATS) site during 2012. Details for these datasets are briefly reported in **Error! Reference source not found.**  $\delta^{18}\text{O}$ ,  $\delta\text{D}$  and S data from BATS cruises for summer 2012 are fully comparable with Western North Atlantic Ocean data for summer 2015 and 2016 around Bermuda area and with Eastern North Atlantic Ocean data for summer 2012. All  $\delta^{18}\text{O}$  observations reported in **Error! Reference source not found.** are more enriched than in gridded dataset<sup>3</sup>, but still comparable when considering uncertainties. We calculated the isotopic composition of ocean water for this study combining salinity measurements performed during the study period (Hog Reef, Crescent Reef and BATS) and  $\delta$  vs S relationship<sup>5</sup> and BATS 2012 data. This approach gives us an interval for ocean water isotopic composition in the study area (last line in **Error! Reference source not found.**). The average isotopic composition of the ocean in this study is then assumed to be  $\delta^{18}\text{O}_\text{L} = 1.09\text{‰}$  and  $\delta\text{D}_\text{L} = 7.25\text{‰}$  (d-excess = -1.47).

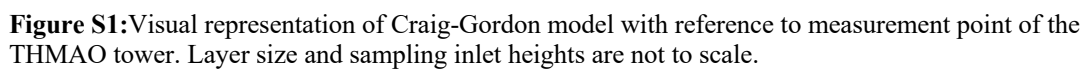
**Table S1:** Isotopic composition and salinity for North Atlantic Ocean.

Source	$\delta^{18}\text{O} \pm 1\text{SD}$ (‰)	$\delta\text{D} \pm 1\text{SD}$ (‰)	d- excess (‰)	Salinity (PSU)	Details
BATS, 2012 (BIOS, 2021)	$1.27 \pm 0.07$ [0.39 S – 13.1‰]	$7.92 \pm 0.81$ [8.64 S – 307.6]	-2.22	36.55	BATS cruises: May – Sept 2012
(Marion Benetti et al., 2014)	$1.24 \pm 0.09$ [0.28 S – 9.0‰]	$8.18 \pm 0.65$ [1.58 S – 50.6‰]	-1.72	37.03	Strasse Cruise: Aug – Sept 2012 Eastern N. Atlantic
(Marion Benetti et al., 2017)	$1.19 \pm 0.12$ [0.32 S – 10.5‰]	$8.03 \pm 0.81$ [2.04 S – 65.8‰]	-1.51	36.22	Rara Cruise: Mar and May 2015 Colibri Cruise: Aug 2016 Leg in Bermuda area
(LeGrande & Schmidt, 2006)	$1.07 \pm 0.15$ [0.55 S – 18.98‰]	-	-	-	Averaged in a 10°x10° centered in Bermuda
Average in Bermuda Area	1.23	7.97	-0.51	36.39	BATS, 2012 Benetti et al., 2017
Estimated from $\delta$ vs S	$1.12^*$ $1.06^\dagger$	$8.29^*$ $6.20^\dagger$	$-0.67^*$ $-2.28^\dagger$	36.32	June – Dec 2013

[ $\delta$  vs S linear relationship].  $\delta$  vs S relationship for N. Atlantic (Marion Benetti et al., 2017):

\*Using  $\delta$  vs S in for gridded dataset (LeGrande & Schmidt, 2006).  $\dagger$ Using  $\delta$  vs S estimated from BIOS (BIOS, 2021).

The CG model can be parametrized into a five layers process, as shown in Figure S1, with an atmospheric component based on the Langmuir linear-resistance model for evaporation where transport of water molecules can be defined as a resistance to atmospheric transport. Starting from the bottom, the first layer is well-mixed ocean water. The second layer is the interface between water and air. Here, humidity is at saturation ( $h = 1$ ) and equilibrium fractionation is assumed to be reached. The third layer is a diffusion-controlled layer where different water isotopologues are characterized by different diffusivities in air. The fourth layer is a turbulence-controlled layer. In this fourth layer, no isotopic fractionation occurs because all water isotopologues of water are characterized by identical mixing behavior. The fifth layer is the “Free atmosphere”, where humidity and isotopologues profiles become less pronounced with height. The actual size of the layers 2-3-4 depends on location, time of the day and atmospheric conditions. However, it is expected that layers 2 and 3 are extremely thin and very hard to resolve with current measurement techniques (Madsen et al., 2019). Measurements above the ocean surface, like in this study, predominantly take place in the fifth or within the fourth layer (top and bottom inlet in Figure S1) making the abovementioned assumption “*The mixing process in the gradient measurement space is fully turbulent and does not introduce any fractionation*” reasonable.



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$$k = \frac{\left(\frac{D}{D_i}\right)^n - 1}{\left(\frac{D}{D_i}\right)^n + \frac{\rho_T}{\rho_M}} \quad (\text{eq. S1})$$

where the ratio  $D/D_i$  is the ratio between diffusivities of the rare isotopologue of water  $i$  and the common isotopologue of water; the exponent  $n$  depends on the wind regime and is equal to 2/3 and to 1/2 for smooth and rough wind regimes;  $\rho_T$  and  $\rho_M$  are the resistances relative to turbulent and molecular transport in air, respectively. In the Brutsaert evaporation model, the ratio of resistances can be estimated for smooth and rough surfaces with equations S2 and S3 (Brutsaert, 1965):

$$\frac{\rho_T}{\rho_M} = \frac{1}{\chi} \frac{\ln\left(\frac{u_* z}{30\nu}\right)}{13.6\left(\frac{\nu}{D}\right)^n} \quad (\text{eq. S2})$$

$$\frac{\rho_T}{\rho_M} = \frac{1}{\chi} \frac{\ln\left(\frac{z}{z_0}\right) - 5}{7.3 R_{es}^{1/4} \left(\frac{\nu}{D}\right)^n} \quad (\text{eq. S3})$$

where  $\chi$  is Von Karman constant;  $u_*$  is the friction velocity [ $\text{m s}^{-1}$ ];  $\nu$  is the kinematic air viscosity [ $\text{kg m}^{-1} \text{s}^{-1}$ ];  $z$  is the height above the water surface [ $\text{m}$ ];  $z_0$  is the roughness length [ $\text{m}$ ];  $R_{es}$  is the surface roughness Reynolds number [-]. Therefore, the fractionation factor  $k$  can be directly calculated as a function of wind speed  $u$  [ $\text{m s}^{-1}$ ] at a reference level (e.g. 10 m) because:

$$u_* = \frac{\chi u}{\ln\left(\frac{z}{z_0}\right)} \quad (\text{eq. S4})$$

$$R_{es} = \frac{u_* z_0}{\nu} \quad (\text{eq. S5})$$

It has been proposed that equation S2 is valid when  $R_{es} < 1$  and equation S3 when  $R_{es} > 1$  (Merlivat, 1978). This variation of transport regime produces a discontinuity in the  $k$  vs wind speed relationship. Such discontinuity for 10m wind speed is  $\approx 6 \text{ m s}^{-1}$ .

### Text S3: Algorithm for choosing the best $k$ value

Non-equilibrium fractionation factor  $k$  is estimated from a direct comparison between observed and modeled isotopic composition of evaporation flux. For a given flux observation  $i$ , it is possible to calculate  $m$  different values of flux composition with CG model by varying the kinetic fractionation factors within a certain range. Then, applying the same procedure to all the  $n$ -observations available, it is possible to obtain a  $n \times m$  matrix that can be compared to observed flux in the following way:

$$\overline{DF}_{(n \times m)} = (\overline{CG} - \overline{\delta}_E)^{\circ 2} \quad (\text{eq. S6})$$

where the matrix  $\overline{\delta}_E$  is composed by the column vector of observed flux isotopic composition concatenated  $m$  times. Then, the differences between modeled and observed values are squared (Hadamard power) and the minimum difference for each observation can be calculated as follows:

$$\overline{min}_{n \times 1} = \min_{j \in J} (\overline{DF}), \quad \forall i \in I \quad (\text{eq. S7})$$

being  $I [I, \dots, i, \dots, n]$  and  $J [J, \dots, j, \dots, m]$  the indices of the rows and columns of the matrix  $\overline{DF}$ , respectively. Similarly, the vector of minimum differences can be concatenated  $m$  times to obtain the matrix  $\overline{MIN}_{n \times m}$  that can be compared with  $\overline{DF}$  to obtain a weighting matrix  $\overline{BM}$ :

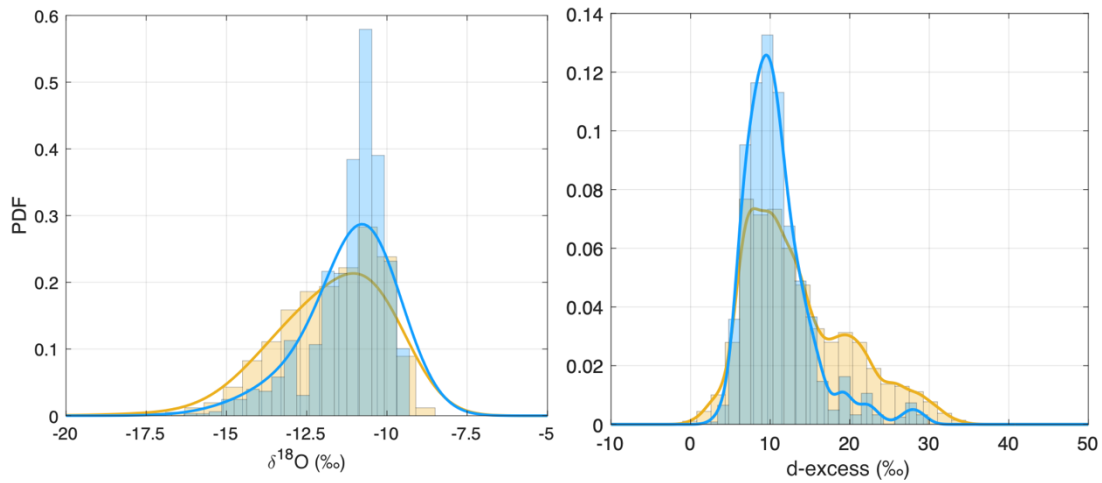
$$\overline{BM}_{(n \times m)} = \overline{DF} == \overline{MIN} \quad (\text{eq. S8})$$

where the double equal in equation S8 is used to distinguish Boolean equality comparison from variable assignation. The  $\overline{BM}$  matrix identify the elements of  $\overline{CG}$  that best match the observed isotopic composition of evaporation flux by minimization of the differences and, as a direct consequence, the best kinetic fractionation value for each observation ( $\overline{BK}_{n \times 1}$ ). The mean kinetic fractionation value  $k$  then can be estimated with a weighted average using the error on flux composition ( $\sigma E_i$ ) as the weight:

$$k = \sum_{i=1}^n w_i * BK_i, \quad w_i = \frac{\frac{|\sigma E_i|}{\delta E_i}}{\sum_{i=1}^n \frac{|\sigma E_i|}{\delta E_i}} \quad (\text{eq13})$$

#### Text S4: Impact of data filtering on observations PDFs

After data filtering, top inlet  $\delta^{18}\text{O}$  data distribution becomes less affected by depleted values maintaining approximately unchanged its mean and median values (mean = median  $\approx -11\text{‰}$  prior and after data filtering with a standard deviation of  $1\text{‰}$ ). The change in distribution shape is more appreciable for d-excess, where data filtering reduced the magnitude of secondary modes (see Figure S1 for  $\delta^{18}\text{O}$  and d-excess).

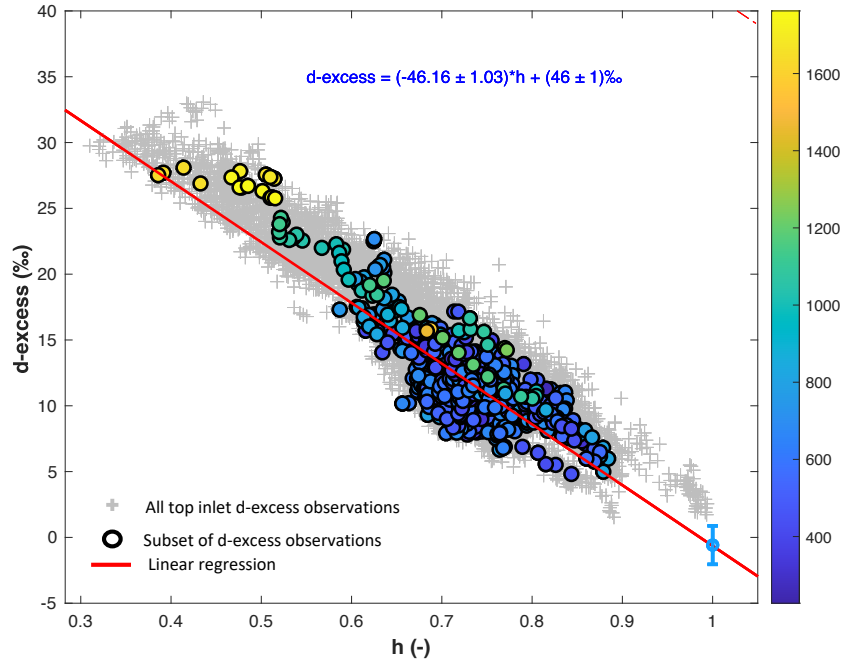


**Figure S2:** Distribution of water vapor isotopic composition prior (orange) and after data filtering (blue) for Top inlet. A continuous kernel density function was estimated with bandwidth =  $1\text{‰}$ .

Presence of several modes in d-excess can be attributed to concurrence of different process involved in water vapor d-excess signal such as change in moisture source area, rain droplets evaporation and increased influence of entrainment. After data filtering, d-excess mean  $\pm 1$  standard deviation lowered from  $15 \pm 6 \text{‰}$  to  $12 \pm 4 \text{‰}$ .

Correlation between water vapor d-excess and h (RH relative to SST) is very high in Bermuda ( $R = -0.91$ , for all top inlet observations). However, dataset reduction lowered the linear regression

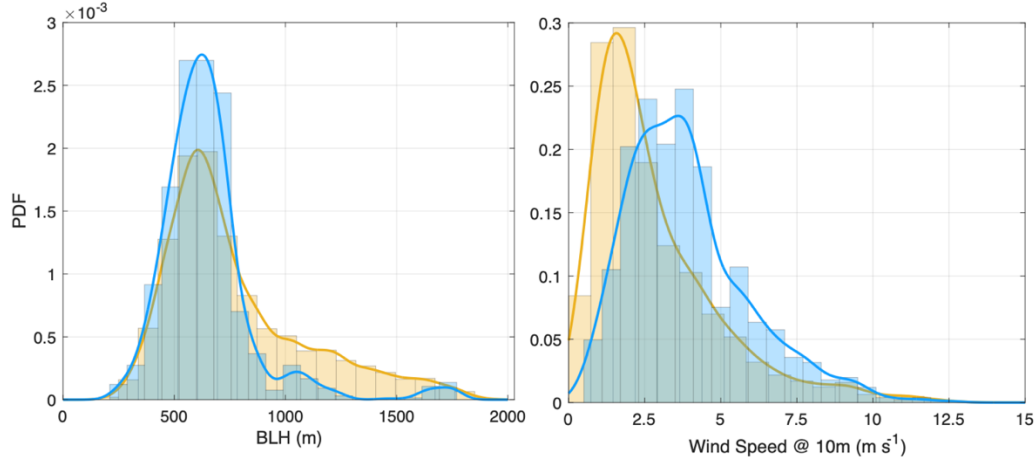
coefficients significantly, from -48 ‰ to -46 ‰ and from 48 ‰ to 46 ‰, for slope and intercept, respectively (Figure S3).



**Figure S3:** Relationship between d-excess, h and BLH. Colored circles represent reduced dataset observations from the top inlet observations after data reduction. D-excess of water vapor in isotopic equilibrium with ocean water is  $-0.8 \pm 0.7\text{‰}$ , estimated with ocean water composition  $\delta^{18}\text{O}_L = 1.09 \pm 0.03\text{‰}$ ,  $\delta\text{D}_L = 7.25 \pm 1.05\text{‰}$  and mean SST = 25.9°C. Linear regression was calculated for reduced data ( $R^2 = 0.70$ ).

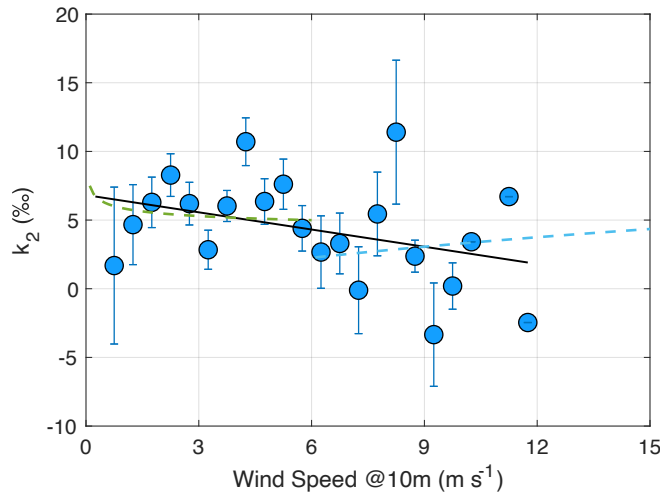
The linear regression models between h and d-excess is able to reproduce the estimated d-excess of water vapor in isotopic equilibrium with ocean water. The predicted d-excess at  $h=1$  resulted -0.31‰ and -0.34‰ for full and filtered dataset, respectively. Previous long-term water vapor observations in Bermuda showed that variability on regression parameters are linked to season and to wind direction (Steen-Larsen et al., 2015). Moreover, as recently pointed out in (Benetti et al., 2018), mixing with upper tropospheric air and planetary boundary layer thickness play an important role in water vapor d-excess variability in the North Atlantic. Bermuda d-excess is highly correlated with BLH ( $R=0.74$ ) and the correlation is still high after data reduction ( $R=0.63$ ).

Typical fully developed boundary layer height is 800 m at 15:00 (LST) which tends to increase from November and later on (1100 m, centered between 12:00 and 15:00 LST). However, the number of observations characterized by large BLH values ( $> 1100$  m) was significantly reduced after data filtering (Figure S4), from ~17% to ~4%.



**Figure S4:** Distribution of BLH and WS datasets prior (orange) and after data reduction (blue). A continuous kernel density function was estimated with bandwidth = 50 m and  $0.5 \text{ ms}^{-1}$  for BLH and WS, respectively. WS was corrected to 10m height.

The main mode, the mean, and the median of BLH does not change significantly after data filtering (from 607 m, 777 m, 679 m to 623 m, 646 m, 620 m for mode, mean and median, respectively). Bottom and top inlet heights are within 1/10 of BLH in 80% of the cases. Therefore, most of the observations can be assumed to be performed within the surface layer, assuming the surface layer height to be roughly the bottom 10% of the BLH (Geernaert, 2003). On the contrary, data filtering significantly affected the wind speed distribution at the study site, largely reducing the number of observations characterized by low wind speed. The distribution shape after data reduction cannot still be considered of the normal-type but still of the Weibull-type. However, skewness reduced from 1.6 to 0.96 with mode, mean and median of WS changed from  $1.6 \text{ ms}^{-1}$ ,  $2.8 \text{ ms}^{-1}$ ,  $2.2 \text{ ms}^{-1}$  to  $3.6 \text{ ms}^{-1}$ ,  $4.0 \text{ ms}^{-1}$ ,  $3.7 \text{ ms}^{-1}$ . Therefore, the main consequence of data reduction from the perspective of d-excess sensitivity to ocean surface condition is a larger impact of shallow mixing with lesser influence on large marine boundary layer development and on low wind speed conditions.



**Figure S5:** Mean  $k_2$  values obtained for each wind speed class. Dashed lines represent the non-equilibrium fractionation factor parametrization as a function of WS at 10m height. Solid black line represents a linear fit  $k_2 = (-0.4 \pm 0.3) * WS + (7 \pm 2) \text{ ‰}$  ( $R^2=0.12$ ) in the wind speed interval  $0 - 10 \text{ m s}^{-1}$ .

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