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## Improved global wetland carbon isotopic signatures support post-2006 microbial methane emission increase

Youmi Oh <sup>1,2,3✉</sup>, Qianlai Zhuang <sup>1,4✉</sup>, Lisa R. Welp <sup>1,5</sup>, Licheng Liu<sup>1,12</sup>, Xin Lan <sup>2,3</sup>, Sourish Basu <sup>6,7</sup>, Edward J. Dlugokencky<sup>3</sup>, Lori Bruhwiler<sup>3</sup>, John B. Miller <sup>3</sup>, Sylvia E. Michel<sup>8</sup>, Stefan Schwietzke <sup>9</sup>, Pieter Tans<sup>3</sup>, Philippe Ciais<sup>10</sup> & Jeffrey P. Chanton<sup>11</sup>

Atmospheric concentrations of methane, a powerful greenhouse gas, have strongly increased since 2007. Measurements of stable carbon isotopes of methane can constrain emissions if the isotopic compositions are known; however, isotopic compositions of methane emissions from wetlands are poorly constrained despite their importance. Here, we use a process-based biogeochemistry model to calculate the stable carbon isotopic composition of global wetland methane emissions. We estimate a mean global signature of  $-61.3 \pm 0.7\text{‰}$  and find that tropical wetland emissions are enriched by  $\sim 11\text{‰}$  relative to boreal wetlands. Our model shows improved resolution of regional, latitudinal and global variations in isotopic composition of wetland emissions. Atmospheric simulation scenarios with the improved wetland isotopic composition suggest that increases in atmospheric methane since 2007 are attributable to rising microbial emissions. Our findings substantially reduce uncertainty in the stable carbon isotopic composition of methane emissions from wetlands and improve understanding of the global methane budget.

<sup>1</sup>Department of Earth, Atmospheric, and Planetary Sciences, Purdue University, West Lafayette, IN, USA. <sup>2</sup>Cooperative Institute for Research in Environmental Sciences, University of Colorado, Boulder, CO, USA. <sup>3</sup>Global Monitoring Laboratory, NOAA, Boulder, CO, USA. <sup>4</sup>Department of Agronomy, Purdue University, West Lafayette, IN, USA. <sup>5</sup>Purdue Climate Change Research Center, West Lafayette, IN, USA. <sup>6</sup>Global Modeling and Assimilation Office, NASA Goddard Space Flight Center, Greenbelt, MD, USA. <sup>7</sup>Earth System Science Interdisciplinary Center, University of Maryland, College Park, MD, USA. <sup>8</sup>Institute of Arctic and Alpine Research, University of Colorado, Boulder, CO, USA. <sup>9</sup>Environmental Defense Fund, Berlin, Germany. <sup>10</sup>Laboratoire des Sciences du Climat et de l'Environnement, Gif-sur-Yvette, France. <sup>11</sup>Department of Earth, Ocean, and Atmospheric Sciences, Florida State University, Tallahassee, FL, USA. <sup>12</sup>Present address: Department of Bioproducts and Biosystems Engineering, University of Minnesota, St. Paul, MN, USA. ✉email: [youmi.oh@noaa.gov](mailto:youmi.oh@noaa.gov); [qzhuang@purdue.edu](mailto:qzhuang@purdue.edu)

Methane ( $\text{CH}_4$ ) is a powerful greenhouse gas, and its atmospheric abundance (in  $\text{nmol mol}^{-1}$ , abbreviated ppb) has increased by about 170% since the 1750s<sup>1,2</sup>. Unlike the steady increases of atmospheric  $\text{CO}_2$  and  $\text{N}_2\text{O}$ , atmospheric  $\text{CH}_4$  nearly stabilized from 1998 to 2006 and then rapidly increased with a growth rate averaging  $\sim 6$  ppb  $\text{yr}^{-1}$  between 2007 and 2013 and  $\sim 11$  ppb  $\text{yr}^{-1}$  between 2014 and 2021. Since 2007,  $\text{CH}_4$  has increased while its stable carbon isotopic composition ( $\delta^{13}\text{C}\text{-CH}_4$ , Eq. 1) has trended to more negative values, after increasing for 200 years<sup>3–5</sup>. Diagnosing the mechanisms behind these changes continues to generate considerable attention and controversy<sup>6–12</sup>.

Measurements of atmospheric  $\text{CH}_4$  abundance and  $\delta^{13}\text{C}\text{-CH}_4$ , in combination with isotopic signatures of sources and sinks, allow partitioning of  $\text{CH}_4$  budgets into different source categories. This is because isotopic signatures of source categories differ substantially, where the  $\delta^{13}\text{C}\text{-CH}_4$  of microbial sources (mean of  $-61.7$  with variability of  $6.2\text{‰}$ ) is isotopically more depleted than fossil (mean of  $-44.8$  with variability of  $10.7\text{‰}$ ) and biomass burning (mean of  $-26.2$  with variability of  $4.8\text{‰}$ ) sources<sup>9,13</sup>. The destruction of  $\text{CH}_4$ , primarily by reaction with hydroxyl radical ( $\text{OH}$ ), isotopically enriches atmospheric  $\text{CH}_4$  relative to the emission-weighted source signature<sup>7,14,15</sup>. Due to a wide range of  $\delta^{13}\text{C}\text{-CH}_4$  in each source category<sup>13</sup>, spatial and temporal distributions must be known to reduce the uncertainty in source partitioning. Wetlands are the largest single natural  $\text{CH}_4$  source and strongly influence atmospheric  $\delta^{13}\text{C}\text{-CH}_4$  changes<sup>12</sup>, but the spatial and temporal information of wetland  $\delta^{13}\text{C}\text{-CH}_4$  is limited, and often a single uniform value is assumed<sup>15,16</sup>. Studies show that source partitioning in atmospheric modeling is highly sensitive to spatiotemporal understanding of wetland  $\delta^{13}\text{C}\text{-CH}_4$ .

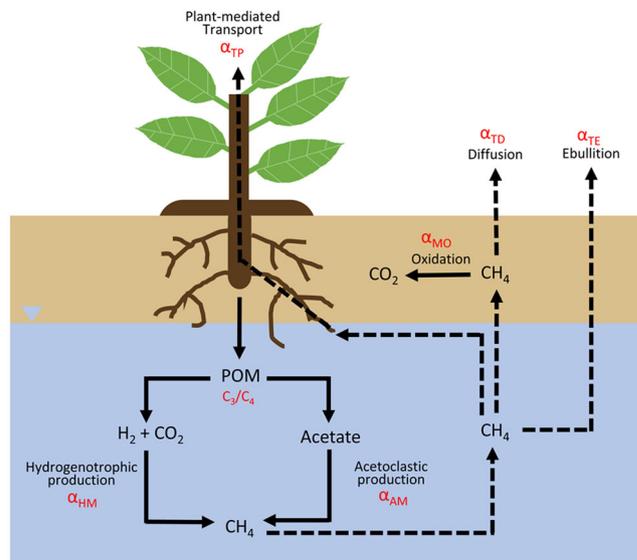
Observations of global wetland  $\delta^{13}\text{C}\text{-CH}_4$  show that  $\text{CH}_4$  emitted from boreal wetlands is isotopically more depleted than  $\text{CH}_4$  emitted from the tropics<sup>17–19</sup>; proposed causes include the abundance of  $\text{C}_4$  plants influencing the  $\delta^{13}\text{C}$  of precursor organic matter (POM) ( $\delta^{13}\text{C}\text{-POM}$ ), differences in  $\text{CH}_4$ -producing archaea (methanogen) communities, and different  $\text{CH}_4$  transport processes<sup>18,20–22</sup>. Ganesan et al.<sup>23</sup> produced a spatially-resolved global wetland  $\delta^{13}\text{C}\text{-CH}_4$  distribution, but their study did not simulate temporal variability and did not fully represent fractionation processes that change based on meteorology, soil and vegetation properties.

Here, we incorporate a carbon isotope module into a biogeochemistry model, the Terrestrial Ecosystem Model (TEM)<sup>24,25</sup> to simulate and mechanistically understand the global wetland  $\delta^{13}\text{C}\text{-CH}_4$  distribution. The model is evaluated using site-level and regional observations. We then use this model to understand the mechanisms behind the spatial and temporal variability of wetland  $\delta^{13}\text{C}\text{-CH}_4$ , and conduct uncertainty and sensitivity tests. Finally, we investigate the effect of new wetland isotope maps on atmospheric  $\delta^{13}\text{C}\text{-CH}_4$  and global  $\text{CH}_4$  emissions by using an atmospheric model and atmospheric observations<sup>5,26</sup>.

## Results

**Modeling wetland  $\delta^{13}\text{C}\text{-CH}_4$  dynamics.** TEM simulates  $\text{CH}_4$  production, oxidation, and transport between soils and the atmosphere (Eqs. 3–10)<sup>24,25,27,28</sup>. A carbon isotope-enabled module is incorporated into TEM, referred to as isoTEM, which explicitly considers carbon isotopic fractionation processes in wetlands (Fig. 1). The isotopic fractionation factor ( $\alpha$ ) for each process is defined in Eq. 2<sup>20</sup>, where  $\alpha$  is larger than 1 when the product is isotopically more depleted than the reactant.

$\delta^{13}\text{C}\text{-POM}$  is determined by the global  $\text{C}_3$  and  $\text{C}_4$  plant distribution (Supplementary Fig. 1)<sup>29</sup>, where  $\text{C}_4$  vegetation is isotopically enriched due to its photosynthetic pathway<sup>30</sup>. We

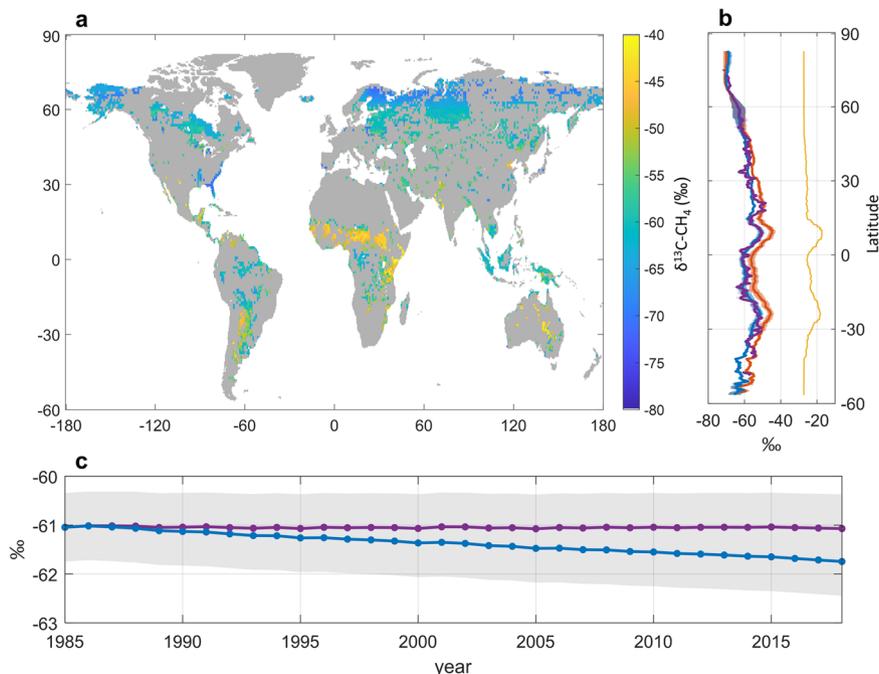


**Fig. 1 Schematic diagram of wetland  $\text{CH}_4$  dynamics and fractionations for isoTEM.**

The model simulates  $\delta^{13}\text{C}$  of precursor organic matter (POM) ( $\delta^{13}\text{C}\text{-POM}$ ),  $\text{CH}_4$  production, oxidation, and transport to the surface.  $\delta^{13}\text{C}\text{-POM}$  is determined by global  $\text{C}_3/\text{C}_4$  plant distribution and long-term trends of atmospheric  $\delta^{13}\text{C}\text{-CO}_2$ .  $\text{CH}_4$  is produced by two pathways, one using  $\text{H}_2$  and  $\text{CO}_2$  and another using acetate, with fractionation factors ( $\alpha$ ) for HMs ( $\alpha_{\text{HM}} \approx 1.030\text{--}1.080$ ) and for AMs ( $\alpha_{\text{AM}} \approx 1.000\text{--}1.040$ ). Produced  $\text{CH}_4$  is partly oxidized by methanotrophs with a fractionation factor  $\alpha_{\text{MO}} \approx 1.015\text{--}1.035$ . Residual produced  $\text{CH}_4$  is emitted to the surface via three processes, plant-mediated transport (TP), diffusion (TD), and ebullition (TE), with different fractionations,  $\alpha_{\text{TP}} \approx 1.000\text{--}1.030$ ,  $\alpha_{\text{TD}} \approx 1.005$ ,  $\alpha_{\text{TE}} \approx 1.000$ , respectively (Supplementary Tables 2–4 and Method “Model development, Model optimization”). Bold and dashed lines in the figure refer to chemical and transport processes, respectively.

incorporated observed long-term trends of atmospheric  $\delta^{13}\text{C}\text{-CO}_2$  into soil  $\delta^{13}\text{C}\text{-POM}$  (Supplementary Fig. 2)<sup>31–33</sup>.  $\text{CH}_4$  is produced from POM in anaerobic soils by two distinct methanogen communities: hydrogenotrophic methanogens (HMs) which use  $\text{H}_2$  and  $\text{CO}_2$  and acetoclastic methanogens (AMs) which use acetate<sup>34</sup>. The fractional contribution of these pathways is important because HMs produce isotopically more depleted  $\text{CH}_4$  compared to AMs ( $\alpha_{\text{HM}}$  and  $\alpha_{\text{AM}}$  in Eq. 12)<sup>19,35</sup>. To quantify the fractional contribution, we used in situ observations from Holmes et al.<sup>19</sup> and conducted a regression analysis between the fractional contribution and main environmental factors, including soil pH, carbon, and latitude (Eq. 11, Supplementary Fig. 3, and Supplementary Table 1). Total produced  $\delta^{13}\text{C}\text{-CH}_4$  is then calculated using a mixing of  $\text{CH}_4$  pools from the two methanogen communities (Eqs. 13–14). The  $\text{CH}_4$  produced is partly oxidized by methanotrophs in aerobic soil layers<sup>36</sup> with  $^{12}\text{CH}_4$  being oxidized preferentially relative to  $^{13}\text{CH}_4$  ( $\alpha_{\text{MO}}$  in Eq. 15). Then, the remaining  $\text{CH}_4$  is emitted to the atmosphere through three processes: plant-mediated transport, diffusion, and ebullition, with fractionation factors of  $\alpha_{\text{TP}}$ ,  $\alpha_{\text{TD}}$ , and  $\alpha_{\text{TE}}$ , respectively (Eq. 16)<sup>20</sup>. We calculated oxidized and emitted  $\delta^{13}\text{C}\text{-CH}_4$  using the ratio of oxidation and transport processes and their fractionation factors (Eqs. 17–22) (Method “Model development”).

We optimized four fractionation factors related to  $\text{CH}_4$  production, oxidation, and plant-mediated transport ( $\alpha_{\text{HM}}$ ,  $\alpha_{\text{AM}}$ ,  $\alpha_{\text{MO}}$ ,  $\alpha_{\text{TP}}$ ) using field observations in boreal ( $50\text{--}90^\circ\text{N}$ ), temperate ( $30\text{--}50^\circ\text{N/S}$ ), and tropical ( $<30^\circ\text{N/S}$ ) wetlands<sup>35,37,38</sup> (Eqs. 12, 15, 16, Supplementary Table 2–4 and Supplementary Figs. 4, 5). We



**Fig. 2** Global distribution of wetland  $\delta^{13}\text{C-CH}_4$  and its latitudinal and long-term gradients simulated by isoTEM. **a** Modeled global wetland  $\delta^{13}\text{C-CH}_4$  for wetland grid cells with static inundation data<sup>49</sup>. **b** Mean latitudinal distribution of  $\delta^{13}\text{C}$  of POM (yellow), produced  $\text{CH}_4$  (red), and  $\text{CH}_4$  emitted to the atmosphere for all grid cells (blue) and flux-weighted grid cells (purple). **c** Long-term trends of global mean wetland  $\delta^{13}\text{C-CH}_4$  with and without incorporating long-term trend in  $\delta^{13}\text{C-POM}$  (blue and purple, respectively). The shaded area in (**b**, **c**) represents one standard deviation determined from 20 ensembles of simulations where the optimized parameters were varied.

set  $\alpha_{\text{TE}}$  to 1.000 and  $\alpha_{\text{TD}}$  to 1.005 based on previous studies<sup>20</sup> since ebullition and diffusion are governed by physical processes. To quantify uncertainties in model simulations, we used 20 ensemble members of optimization. We simulated global wetland  $\text{CH}_4$  fluxes and their isotopic signatures during 1984–2016 at a spatial resolution of  $0.5^\circ$  with a 50-year spin-up to let  $\delta^{13}\text{C-CH}_4$  of carbon pools come to a steady state (Methods “Model optimization, Simulation setup”).

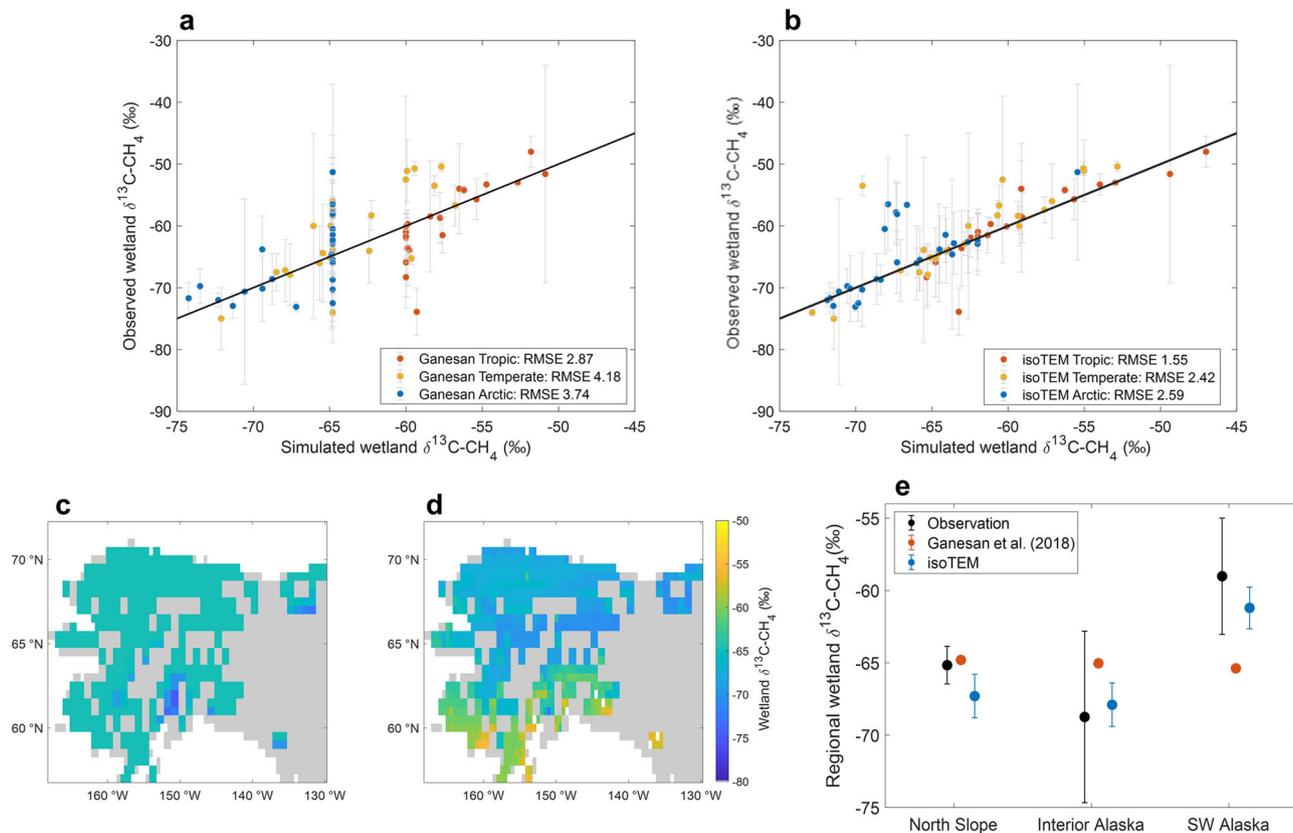
**Simulated wetland  $\delta^{13}\text{C-CH}_4$  and its comparison with observations.** We estimated the mean global wetland source signature to be  $-61.3 \pm 0.7\text{‰}$  during 1984–2016 (Fig. 2a). This value is more enriched than the mean wetland signature of  $-62.3\text{‰}$  in Ganesan et al.<sup>23</sup> but similar to the mean value of  $-61.5\text{‰}$  reported in Sherwood et al.<sup>13</sup> (Supplementary Figs. 8, 9). The latitudinal distribution of  $\delta^{13}\text{C-CH}_4$  ranges from a mean of  $-57 \pm 3\text{‰}$  in the tropics to  $-68 \pm 4\text{‰}$  in boreal regions (Fig. 2b). Our model simulates isotopically depleted global  $\delta^{13}\text{C-CH}_4$  during the summer due to larger emissions from boreal regions (Supplementary Fig. 10) and a long-term trend of  $-0.7 \pm 0.1\text{‰}$  during 1984–2016 (blue line in Fig. 2c) when incorporating the long-term trend in  $\delta^{13}\text{C-POM}$  (Supplementary Fig. 2)

We compared the magnitude and spatial variability of the simulated wetland  $\delta^{13}\text{C-CH}_4$  with site-level observations (Method “Model-data comparison”). We used 70 in situ measurements of global wetland  $\delta^{13}\text{C-CH}_4$  from previous studies after excluding the measurements applied for optimization (Supplementary Data 1, Supplementary Fig. 11)<sup>13,19</sup>. We showed that isoTEM reduced the root mean square error (RMSE) by 40% compared to Ganesan et al.<sup>23</sup> (2.2 vs. 3.6) (Fig. 3a, b). Compared to a static isoTEM map in July, 2016, temporally-varying isoTEM reduced the RMSE slightly (2.2 vs. 2.4) (Supplementary Fig. 12). Ganesan et al.<sup>23</sup> prescribed maximum and minimum values as boundary conditions, resulting in unrealistic clusters of wetland  $\delta^{13}\text{C-CH}_4$

near  $-65\text{‰}$  for boreal and  $-60\text{‰}$  for tropical sites (Fig. 3a and Supplementary Fig. 9).

Furthermore, we compared the spatial variability of simulated wetland  $\delta^{13}\text{C-CH}_4$  with estimated signatures from airborne measurements for three regions in Alaska during 2012–2013 and 2015 using Miller-Tans plots (Fig. 3c–e) (Method “Model data comparison”)<sup>39</sup>. In situ flux observations collected across Alaskan wetlands show an average of  $-65\text{‰}$ , but with a large 9% variance<sup>40</sup>, which could be due to changes in wetland habitat including soil nutrients, pH, and vegetation distribution. The estimated signatures from observation also show that compared with  $\delta^{13}\text{C-CH}_4$  from the North Slope of Alaska ( $-65 \pm 1\text{‰}$ ),  $\delta^{13}\text{C-CH}_4$  from interior Alaska is more depleted ( $-69 \pm 6$ ) and  $\delta^{13}\text{C-CH}_4$  from southwest Alaska is more enriched ( $-59 \pm 4\text{‰}$ ) (Supplementary Fig. 13 and Supplementary Table 5). IsoTEM reproduces the spatial variability ( $-67 \pm 1$ ,  $-68 \pm 1$ , and  $-61 \pm 2\text{‰}$  for North Slope, interior, and southwest Alaska, respectively), whereas Ganesan et al.<sup>23</sup> simulated no spatial variability around  $-65\text{‰}$  (Fig. 3e). IsoTEM simulates the spatial variability as the model optimized parameters for vegetated and non-vegetated sites separately and incorporated meteorology and soil inputs that vary spatially and temporally.

**Mechanistic understanding of spatial and temporal variability of wetland  $\delta^{13}\text{C-CH}_4$ .** We investigated the relative importance of the isotopic fractionation processes that affect the latitudinal gradient of wetland  $\delta^{13}\text{C-CH}_4$  (Fig. 2b and Supplementary Fig. 14). First, compared to the boreal zone,  $\delta^{13}\text{C-POM}$  is enriched in the tropics by  $5 \pm 2\text{‰}$  as  $\text{C}_4$  plants are more prevalent (yellow line in Fig. 2b, Supplementary Figs. 1, 14a). Second, due to a larger fraction of AM in the tropics (Supplementary Fig. 3), the  $\delta^{13}\text{C-CH}_4$  produced by methanogens is enriched by  $12 \pm 3\text{‰}$  (red line in Fig. 2b, Supplementary Fig. 14b). Third,  $\delta^{13}\text{C-CH}_4$  emitted from wetlands is  $6 \pm 4\text{‰}$  more depleted in the tropics due to a larger proportion of plant-mediated transport causing higher



**Fig. 3 Site-level and regional model-data comparison of wetland  $\delta^{13}\text{C-CH}_4$ .** **a, b** Site-level model-data comparison of observations with **(a)** Ganesan et al.<sup>23</sup> and **(b)** temporally-varying isoTEM. **c–e** Regional model-data comparison of simulated wetland  $\delta^{13}\text{C-CH}_4$  in Alaska by **(c)** Ganesan et al.<sup>23</sup> and **(d)** isoTEM, and **(e)** their comparison with observation-based source signatures from NOAA aircraft measurements. The source signature is derived using Miller-Tans plots<sup>39</sup>. All observation data used for site-level comparison are listed in Supplementary Data 1. Error bars for observations in **(a, b, e)** represent one standard deviation of measured/inferred wetland  $\delta^{13}\text{C-CH}_4$ . Error bars for isoTEM in panel **e** represent one standard deviation determined from 20 ensemble simulations where the optimized parameters were varied.

effective transport fractionation ( $\alpha_T$ ) (blue line in Fig. 2b, Eq. 19, Supplementary Figs. 14d, 15, 16). Thus, in our simulation,  $\delta^{13}\text{C-CH}_4$  emitted from tropical wetlands is enriched by  $\sim 11\text{‰}$  compared to boreal wetlands. This difference is strengthened due to the distribution of  $\text{C}_4$  plants ( $+5 \pm 2\text{‰}$ ) and the fractional contribution of differing methanogen communities ( $+12 \pm 3\text{‰}$ ) but weakened due to plant-mediated transport ( $-6 \pm 4\text{‰}$ ).

The long-term decrease in wetland  $\delta^{13}\text{C-CH}_4$  simulated by isoTEM is mostly due to the decrease in atmospheric  $\delta^{13}\text{C-CO}_2$ <sup>32</sup>. The decreasing trend is incorporated into  $\delta^{13}\text{C-POM}$  (Supplementary Fig. 2) and causes the long-term decrease in wetland  $\delta^{13}\text{C-CH}_4$  of  $\sim 0.7\text{‰}$  from 1984 to 2016 (blue line in Fig. 2c)<sup>31</sup>. We conducted a simulation without the decreasing trend in  $\delta^{13}\text{C-POM}$ , which showed that increased temperature caused plant productivity and plant-mediated transport to increase and  $\delta^{13}\text{C-CH}_4$  to decrease by  $\sim 0.1\text{‰}$  during 1984–2016 (purple line in Fig. 2c and Supplementary Fig. 15). This implies that wetland  $\delta^{13}\text{C-CH}_4$  could further change in the future due to decreases in  $\delta^{13}\text{C-POM}$  and increases in plant-mediated transport.

There is no continuous long-term measurements of wetland  $\delta^{13}\text{C-CH}_4$  to verify our simulated long-term trend. Instead, we ran a regression analysis using observations collected from various wetland locations since the early 1980s (Supplementary Data 1) (Method “Uncertainty and sensitivity tests”). The results show that the representation of data increases when adding year as a parameter for the regression analysis (Supplementary Table 6), and the observed data show a long-term decreasing trend with year ( $\sim -0.1\text{‰ year}^{-1}$ ) (Supplementary Fig. 17). More

continuous long-term observations of wetland  $\delta^{13}\text{C-CH}_4$  are necessary to further verify the simulated long-term trends in wetland  $\delta^{13}\text{C-CH}_4$ .

**Uncertainty and sensitivity tests.** The version of TEM that we use for this study explicitly simulates soil  $\text{CO}_2$  and  $\text{CH}_4$  but not soil  $\text{H}_2$  and acetate pools<sup>27</sup>, because the spatial and temporal soil  $\text{H}_2$  and acetate pools are highly uncertain, and it is hard to verify the simulated pool changes with limited observations. On the contrary, the  $\text{CH}_4$  production, oxidation, and transport processes in TEM have been thoroughly validated for global regions from previous studies<sup>24,25,27,41–44</sup>. Therefore, instead of introducing additional uncertainty from explicitly simulating  $\text{H}_2$  and acetate pools that cannot be validated, we applied the observed fraction of different methanogen communities ( $f_{\text{HM}}$ ) based on regression to the total  $\text{CH}_4$  production rates simulated by TEM (Supplementary Fig. 3 and Supplementary Table 1). Thus, in our simulation, the fraction of HM and AM ( $f_{\text{HM}}$ ) changes spatially but not temporally.

To quantify the uncertainty of our regression analysis of  $f_{\text{HM}}$ , we ran additional sensitivity tests by varying the  $f_{\text{HM}}$  based on the uncertainty from Markov Chain Monte Carlo approach (Method “Uncertainty and sensitivity tests” and Supplementary Table 1)<sup>45</sup>. The results show that varying the parameters do not change the wetland  $\delta^{13}\text{C-CH}_4$  substantially ( $<1\%$ ) (Supplementary Table 7). We acknowledge that this simplification would cause uncertainty in our model results, and future studies should explicitly measure

changes in H<sub>2</sub> and acetate concentrations in soils to incorporate the detailed processes into the model.

The simplification of CH<sub>4</sub> production processes may also cause uncertainty in the fractionation as we do not explicitly simulate fractionation processes from POM to CO<sub>2</sub>/acetate and from CO<sub>2</sub>/acetate to CH<sub>4</sub>. However, studies show that fractionation factors of the fermentation (POM to CO<sub>2</sub>) and syntrophy (POM to acetate) processes are minor ( $\alpha \approx 1.00$ )<sup>19,46,47</sup>. There may be additional CO<sub>2</sub> produced by acetoclastic methanogenesis that have large fractionation ( $\alpha \approx 1.05$ ), but the fraction is negligible from observations<sup>19</sup>. Thus, we believe our fractionation factors for HMs and AMs ( $\alpha_{HM}$  and  $\alpha_{AM}$ , respectively) reasonably represent the major fractionation processes of CH<sub>4</sub> production.

Furthermore, to quantify the influence of the uncertainty of our model inputs on simulation results, we varied temperature, precipitation, net primary productivity (NPP), atmospheric CH<sub>4</sub>, and applied transient inundation maps<sup>48</sup> (Method “Uncertainty and sensitivity tests”). The results show that meteorology and substrate inputs alter mean wetland  $\delta^{13}C-CH_4$  by  $\pm 1\%$  (Supplementary Table 7). Our TEM simulations showed that CH<sub>4</sub> fluxes are sensitive to these inputs<sup>27</sup>. However,  $\delta^{13}C-CH_4$  shows small changes because the fractionation is determined by the fraction of CH<sub>4</sub> oxidation and transport processes (Eqs. 21, 22), that are calculated as a function of soil CH<sub>4</sub> production and the resultant CH<sub>4</sub> concentration changes ( $C_M$  in Eqs. 6–10). When CH<sub>4</sub> production increases due to input changes, CH<sub>4</sub> oxidation and transport increase simultaneously, causing minor variation in the fraction of oxidation and transport (Supplementary Fig. 16). Inundation datasets also alter wetland  $\delta^{13}C-CH_4$  by changing the areas where wetland emissions occur ( $\pm 2\%$ ) (Supplementary Table 7 and Supplementary Figs. 6, 7).

### Implication for atmospheric modeling and global CH<sub>4</sub> budget.

We constructed four scenarios with different wetland emissions and isotopic signature maps as inputs for TM5 atmospheric modeling during 1984–2016 to understand the impacts of spatially- and temporally-resolved wetland  $\delta^{13}C-CH_4$  (Table 1). Scenario A uses a globally uniform value of wetland  $\delta^{13}C-CH_4$ ; Scenario B uses a temporally static but spatially variable wetland isotope map from Ganesan et al.<sup>23</sup>; and Scenario C uses spatially- and temporally-resolved maps from isoTEM. We used the same wetland fluxes<sup>27</sup> with a static inundation map<sup>49</sup> for Scenarios A–C that applied a step increase in fluxes in 2007 and 2014 by hypothesizing that microbial wetland emissions are the dominant driver of the post-2006 atmospheric CH<sub>4</sub> increase<sup>9,26,50</sup> (46 Tgyr<sup>-1</sup> increase in total 2016 emissions across the global wetlands compared to the

averaged total emissions in 1999–2006) (Supplementary Fig. 19). However, since other studies have suggested an increase in fossil emission as a dominant driver for post-2006 CH<sub>4</sub> increases<sup>12</sup>, we created scenario D that uses isoTEM wetland isotope maps with increases in both microbial and fossil emissions since 2007 (Table 1).

For Scenarios A–D, we adjusted global mean fossil and ruminant fluxes simultaneously to satisfy the long-term average mass balance of atmospheric CH<sub>4</sub> (Fig. 4a) and  $\delta^{13}C-CH_4$  (Method “Forward modeling using TM5 atmospheric model”), as done by Lan et al.<sup>26</sup>. These adjustments bring the long-term global average  $\delta^{13}C-CH_4$  from simulation to the observed atmospheric levels without changing the post-2006 trends in simulated  $\delta^{13}C-CH_4$ <sup>9,26</sup>. After adjustments, global mean fossil fluxes in scenarios A–D are between 170 and 190 Tgyr<sup>-1</sup> (Supplementary Fig. 19), within the uncertainty range in Schwietzke et al.<sup>9</sup>. For all other fluxes, their isotopic signatures, and CH<sub>4</sub> sinks that include OH, Cl, and O(<sup>1</sup>D)<sup>14,51,52</sup>, we used the same setup in our model as in Lan et al.<sup>26</sup> (Supplementary Table 8). We compared simulated CH<sub>4</sub> and  $\delta^{13}C-CH_4$  with observations from NOAA/INSTAAR global flask-air measurements (Supplementary Table 10)<sup>2,5</sup>.

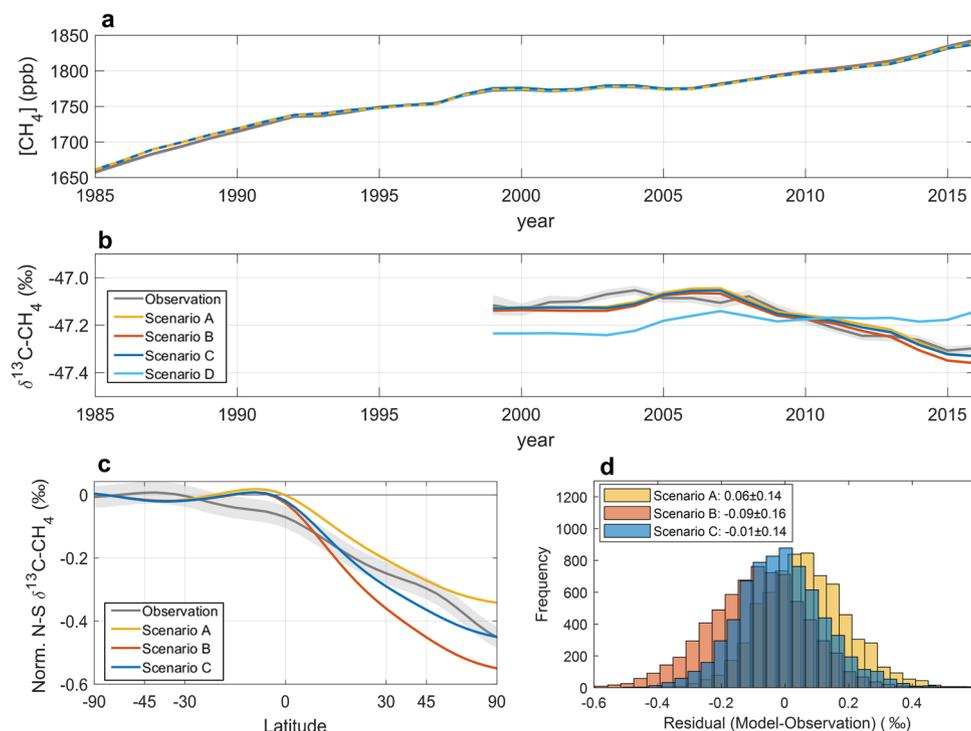
The atmospheric simulation showed that Scenarios A–C follow the observed  $\delta^{13}C-CH_4$  trend reasonably closely (Fig. 4b). However, Scenario D, which hypothesizes a post-2006 increase in microbial and fossil fluxes, does not follow the decreasing trend in global mean  $\delta^{13}C-CH_4$ . As pointed out earlier<sup>8,9,26,50</sup>, the magnitude of the  $\delta^{13}C-CH_4$  decrease suggests that the increase in microbial emissions dominates fossil emissions in the post-2006 global CH<sub>4</sub> increase. We also confirmed a dominant increase in post-2006 microbial emissions, even though the long-term decrease in wetland  $\delta^{13}C-CH_4$  of  $\sim 0.7\%$  allow for a larger fossil emission increase. An additional simulation of Scenario C without including the long-term decrease in wetland  $\delta^{13}C-CH_4$  shows differences of  $\sim 0.1\%$  in simulated atmospheric  $\delta^{13}C-CH_4$  in 2016 compared with model results with long-term wetland  $\delta^{13}C-CH_4$  trend (Supplementary Fig. 23). This difference can accommodate more post-2006 emission increases from isotopically enriched fossil sources for Scenario C.

We differentiated Scenarios A–C by comparing their simulated latitudinal gradients of atmospheric  $\delta^{13}C-CH_4$  with observations (Fig. 4c and Supplementary Fig. 20). The observed mean latitudinal gradient during 1998–2016 shows more negative  $\delta^{13}C-CH_4$  at northern high latitudes compared to the Southern Hemisphere by  $0.45 \pm 0.05\%$  (Supplementary Table 9), resulting from the dominance of northern emissions combined with the subsequent fractionation by reaction with OH during transport to the Southern Hemisphere<sup>17</sup>. Scenario C, which uses IsoTEM maps,

**Table 1 Setup of TM5 atmospheric modeling for Scenarios A–D.**

Scenario	Wetland isotope map	Assumption of post-2006 CH <sub>4</sub> increase	Global mass balance of CH <sub>4</sub> and $\delta^{13}C-CH_4$ <sup>a</sup>
A: Uniform w/Microbial Increase	One uniform value ( $-62.3\%$ , a mean signature of Ganesan et al. <sup>23</sup> )	Wetland emission increase (46 TgCH <sub>4</sub> yr <sup>-1</sup> increase from 1999–2006 to 2016)	Yes
B: Ganesan w/Microbial Increase	One spatial map from Ganesan et al. <sup>23</sup> (mean of $-62.3\%$ )		
C: isoTEM w/Microbial Increase	Spatio-temporally-resolved maps from isoTEM (mean of $-61.3\%$ )		
D: isoTEM w/Microbial + Fossil Increase	(this study)	Wetland (60%) + fossil (40%) emission increase <sup>12</sup> (28 TgCH <sub>4</sub> yr <sup>-1</sup> increase from wetland, 18 TgCH <sub>4</sub> yr <sup>-1</sup> increase from fossil, from 1999–2006 to 2016)	

<sup>a</sup>Using a global mass balance model from previous studies<sup>9,26</sup>, the long-term mean fossil and ruminant fluxes were adjusted from EDGAR 4.3.2 inventory to match the observed atmospheric growth rate of CH<sub>4</sub> during 1984–2016 and the 1998–2016 mean of  $\delta^{13}C-CH_4$ . By conducting the mass balance for all scenarios, we intended to reduce the spin-up time for atmospheric  $\delta^{13}C-CH_4$  to be stabilized and compare all scenarios fairly (Method “Forward modeling using TM5 atmospheric model”).



**Fig. 4** Observed and simulated atmospheric CH<sub>4</sub> and δ<sup>13</sup>C-CH<sub>4</sub> from TM5 atmospheric modeling. **a**, **b** Model-data comparison of long-term trend of (a) atmospheric CH<sub>4</sub> from 1985 to 2016 (in ppb) and (b) δ<sup>13</sup>C-CH<sub>4</sub> from 1999 to 2016 (in ‰) by observation (gray) and simulations from Scenario A (yellow), B (red), C (blue), and D (skyblue). **c** Model-data comparison of normalized north-south gradient of atmospheric δ<sup>13</sup>C-CH<sub>4</sub> for Scenario A (yellow), B (red), and C (blue) in 2012. The north-south δ<sup>13</sup>C-CH<sub>4</sub> was calculated by zonally-averaging the surface δ<sup>13</sup>C-CH<sub>4</sub> and normalized based on the mean δ<sup>13</sup>C-CH<sub>4</sub> at 60–90 °S. The normalized north-south δ<sup>13</sup>C-CH<sub>4</sub> for other years is in Supplementary Fig. 20 and Supplementary Table 9. **d** Histogram of the difference between simulated and observed δ<sup>13</sup>C-CH<sub>4</sub> for Scenario A (yellow), B (red), and C (blue) for 6 measurement sites located in the northern hemisphere. The histogram plots for all measurement sites are in Supplementary Fig. 22. Information about Scenarios A–D is in Table 1.

best reproduces the observed north-south gradient (0.48‰); Scenarios A and B under- and over-estimate the gradient by ~0.1‰ (0.37‰, and 0.59‰, respectively). The difference is also clear when comparing simulated atmospheric δ<sup>13</sup>C-CH<sub>4</sub> of Scenarios A–C at 10 measurement sites (Supplementary Figs. 21, 22 and Supplementary Table 10). The simulated and observed atmospheric δ<sup>13</sup>C-CH<sub>4</sub> differ the most at Northern Hemispheric sites, where Scenario C best reproduces the atmospheric δ<sup>13</sup>C-CH<sub>4</sub> data, but Scenario A and Scenario B simulate more negative and positive δ<sup>13</sup>C-CH<sub>4</sub>, respectively (Fig. 4d).

The difference in north-south gradient of atmospheric δ<sup>13</sup>C-CH<sub>4</sub> between scenarios in Fig. 4c has an implication on regional partitioning of sources. Our sensitivity test of atmospheric modeling showed that all scenarios with transient inundation data<sup>48</sup> (Scenarios E–G) underestimated the north-south δ<sup>13</sup>C-CH<sub>4</sub> gradient (0.27 ± 0.06‰) compared with observations (0.45 ± 0.05‰) (Method “Forward modeling using TM5 atmospheric model”, Supplementary Table 11, Supplementary Figs. 26–30). Thus, we ran an additional Scenario H that increased emissions from boreal wetlands by 2.5 times over the original transient data (Supplementary Fig. 26 and Supplementary Table 11), which increased the north-south gradient by ~0.1‰ and improved the match with the observed north-south δ<sup>13</sup>C-CH<sub>4</sub> gradient (0.39‰) (Supplementary Figs. 29, 30).

## Discussion

The atmospheric CH<sub>4</sub> burden has grown rapidly since 2007, and the largest annual increase since NOAA began measurements in 1983 was observed in 2021<sup>53,54</sup>. Since 2019, δ<sup>13</sup>C-CH<sub>4</sub> decreased more steeply<sup>55</sup>, suggesting a further increase in microbial emissions

as this and other studies suggest<sup>8,9,26,50</sup>. The microbial sources include anthropogenic emissions from ruminants, agriculture, and waste, and natural emissions from wetlands and other aquatic ecosystems<sup>12</sup>. Our simulation with increase in wetland emissions can reproduce the observed post-2006 δ<sup>13</sup>C-CH<sub>4</sub> decrease (Fig. 4), and our additional sensitivity test with increase in anthropogenic microbial emissions also tracks the post-2006 δ<sup>13</sup>C-CH<sub>4</sub> decrease (Supplementary Figs. 24, 25). However, the scenario with emission increase from both microbial and fossil sources did not reproduce the decreasing trend in atmospheric δ<sup>13</sup>C-CH<sub>4</sub> (Scenario D in Fig. 4). Other atmospheric studies that use atmospheric δ<sup>13</sup>C-CH<sub>4</sub> observations also showed that fossil emission increase is not a dominant reason of recent CH<sub>4</sub> increase<sup>26,56</sup>.

Atmospheric δ<sup>13</sup>C-CH<sub>4</sub> measurements have not been widely used to inform global methane budget because of uncertainty and spatiotemporal variation in source signatures, specifically citing limitation in wetland source signatures<sup>11</sup>. In this study, we mechanistically explain the spatiotemporal variations of wetland δ<sup>13</sup>C-CH<sub>4</sub> and validate the simulation using regional, latitudinal, and global measurements, which substantially reduce the uncertainty in δ<sup>13</sup>C-CH<sub>4</sub> source signatures (Fig. 3). The small decreasing trend in wetland δ<sup>13</sup>C-CH<sub>4</sub> allow for more fossil emission increase in our estimate, but cannot change the conclusion that fossil emission increases are not the dominant driver for post-2006 global CH<sub>4</sub> increases.

This study considers wetland δ<sup>13</sup>C-CH<sub>4</sub> during the historical period only, but the future changes in wetland δ<sup>13</sup>C-CH<sub>4</sub> will depend on multiple factors. First, our simulation shows that changes in δ<sup>13</sup>C-POM affect wetland δ<sup>13</sup>C-CH<sub>4</sub> as SOC is mostly derived from new carbon from vegetation. The simulated active layer depth from a previous study<sup>57</sup> shows that the active layer

depth had a minor change during our simulation period (mean of <0.1 m) (Supplementary Fig. 18). However, the usage of old stored carbon in Arctic permafrost may play an important role as a substrate for methanogens in the future<sup>58</sup>. Also, studies found the importance of microbial fossil CH<sub>4</sub> emissions from Arctic regions in the future<sup>59,60</sup>. The emissions are partially included as geologic seep emissions in our atmospheric modeling simulation (Supplementary Fig. 19 and Supplementary Table 8), and we also considered microbial fossil emissions with depleted  $\delta^{13}\text{C}$ -CH<sub>4</sub> in our total fossil emission estimates<sup>26</sup>. Lastly, our simulation shows that the increase in NPP cause more plant-mediated transport. This effect will be more important in the future as plant functional types and plant growth change due to temperature increase.

There are several aspects of the model that could be improved. First, our optimization of fractionation factors was based on limited observations; additional long-term measurements of wetland  $\delta^{13}\text{C}$ -CH<sub>4</sub> would reduce the uncertainty. Second, the fractional contribution of two methanogen communities (HMs and AMs) changes spatially but not temporally in the model. We need a better understanding of temporal changes in methanogen communities especially following permafrost thaw and disturbance<sup>35</sup>, and explicitly measure changes in H<sub>2</sub> and acetate concentrations in soils to incorporate detailed CH<sub>4</sub> production processes into the model. Third, various vertical methanogenic and non-methanogenic processes change  $\delta^{13}\text{C}$  of CH<sub>4</sub> and CO<sub>2</sub>, the vertical CO<sub>2</sub>/CH<sub>4</sub> ratios, and thus  $\delta^{13}\text{C}$ -CH<sub>4</sub> emitted from wetlands, since CO<sub>2</sub> is a substrate for HM<sup>61,62</sup>. We need to identify detailed vertical subsurface processes by conducting manipulation experiments using isotopic labeling analysis and inhibitor techniques to include those fractionation processes in future modeling studies<sup>63</sup>. Fourth, current wetland models do not simulate large CH<sub>4</sub> emissions and  $\delta^{13}\text{C}$ -CH<sub>4</sub> from tropical tree stems and aquatic sources properly<sup>64–66</sup>. More measurements from these sources are crucial to improve the estimate of natural CH<sub>4</sub> emission and  $\delta^{13}\text{C}$ -CH<sub>4</sub> changes.

## Conclusion

To the best of our knowledge, this study is the first to use a biogeochemistry model to mechanistically explain and reduce the uncertainty in global wetland  $\delta^{13}\text{C}$ -CH<sub>4</sub>. IsoTEM explains the latitudinal gradient of wetland  $\delta^{13}\text{C}$ -CH<sub>4</sub> that is increased by the distribution of C<sub>3</sub>/C<sub>4</sub> plants and methanogen community type but decreased by plant-mediated transport. The long-term trends of the simulated wetland  $\delta^{13}\text{C}$ -CH<sub>4</sub> is controlled by  $\delta^{13}\text{C}$ -POM and plant-mediated transport. Our results suggest that rising microbial emissions is the dominant driver for the post-2006 global CH<sub>4</sub> increase and the concurrent decrease in atmospheric  $\delta^{13}\text{C}$ -CH<sub>4</sub>, and the isoTEM spatial distribution of wetland  $\delta^{13}\text{C}$ -CH<sub>4</sub> better reproduces the observed atmospheric  $\delta^{13}\text{C}$ -CH<sub>4</sub> latitudinal gradient.

## Methods

**Model development.** We incorporated a carbon isotope module of methane (CH<sub>4</sub>) into an existing process-based biogeochemistry model, the TEM (Fig. 1). The stable carbon isotope in delta notation ( $\delta$ ) describes the ratio of the heavy isotope to the light isotope in the sample ( $R_{\text{sam}} = (^{13}\text{C}/^{12}\text{C})_{\text{sam}}$ ) relative to a known standard ratio,  $R_{\text{std}}$ , which is Vienna Pee Dee Belemnite (VPDB) for carbon<sup>20</sup> (Eq. 1). The deviation of this ratio-of-ratios from one is multiplied by 1000 to express isotope variations in parts per thousand (‰, permil). To express isotopic fractionation for the reaction  $A \rightarrow B$ , we used a fractionation factor ( $\alpha$ ) defined in Eq. 2<sup>20</sup>, where reactant A is in the numerator and product B is in the denominator. If  $\alpha$  is larger than 1, the  $\delta^{13}\text{C}$  of product is isotopically more depleted in the heavy isotope than the  $\delta^{13}\text{C}$  of reactant, and if  $\alpha$  is smaller than 1, the  $\delta^{13}\text{C}$  of product is more enriched in <sup>13</sup>C than the  $\delta^{13}\text{C}$  of reactant.

$$\delta^{13}\text{C} = (R_{\text{sam}}/R_{\text{std}}) - 1 \quad (1)$$

$$\alpha = \frac{R_A}{R_B} = \left( \frac{\delta^{13}\text{C}_A}{1000} + 1 \right) / \left( \frac{\delta^{13}\text{C}_B}{1000} + 1 \right) \quad (2)$$

**Terrestrial Ecosystem Model (TEM).** TEM is a commonly used biogeochemistry model and its CH<sub>4</sub>, soil, thermal, and hydrological dynamics have been evaluated in previous studies<sup>24,28,41–44</sup>. The CH<sub>4</sub> dynamics module of TEM simulates CH<sub>4</sub> production, oxidation, and three transport processes—diffusion, ebullition, and plant-mediated transport—between soil and atmosphere. Please refer to the details of TEM in Oh et al.<sup>25</sup> and Liu et al.<sup>27</sup>.

In TEM wetland model, changes in CH<sub>4</sub> concentrations ( $C_M$ ) at depth  $z$  and time  $t$  ( $\partial C_M(z,t)/\partial t$ ) are governed by Eq. 3, where  $M_p(z,t)$ ,  $M_o(z,t)$ ,  $R_p(z,t)$ , and  $R_E(z,t)$  are CH<sub>4</sub> production, oxidation, plant-mediated transport, and ebullition rates, respectively, and  $\partial F_D(z,t)/\partial z$  represents flux divergence from gaseous and aqueous diffusion. CH<sub>4</sub> is produced by methanogens in anaerobic soils ( $M_p$ ) and is calculated by multiplying maximum potential production rate ( $M_{CO}$ ) and limiting functions of substrate, soil temperature, pH, and redox potentials ( $S_{OM}$ ,  $M_{ST}$ ,  $pH$ , and  $R_x$ , respectively) (Eq. 4). For this study, we assume that substrates for methanogens are mainly from soil organic carbon (SOC) derived from vegetation (Net Primary Productivity, NPP), where  $NPP(mon)$  is monthly NPP (gC m<sup>-2</sup> month<sup>-1</sup>),  $NPP_{MAX}$  is ecosystem-specific maximum monthly NPP, and  $f(C_{DIS}(z))$  describes the relative availability of organic carbon substrate at depth  $z$  (Eq. 5). The substrate availability changes depending on atmospheric CO<sub>2</sub>, meteorology, and soil properties<sup>67</sup>.

$$\frac{\partial C_M(z,t)}{\partial t} = M_p(z,t) - M_o(z,t) - \frac{\partial F_D(z,t)}{\partial z} - R_p(z,t) - R_E(z,t) \quad (3)$$

$$M_{P,TEM}(z,t) = M_{GO}f(S_{OM}(z,t))f(M_{ST}(z,t))f(pH(z,t))f(R_x(z,t)) \quad (4)$$

$$f(S_{OM}(z,t)) = \left( 1 + \frac{NPP(mon)}{NPP_{max}} \right) f(C_{DIS}(z)) \quad (5)$$

The produced CH<sub>4</sub> is partly oxidized by methanotrophs and is calculated by the multiplying the maximum potential oxidation rate ( $O_{MAX}$ ) and limiting functions of CH<sub>4</sub> concentration, soil temperature, soil moisture, redox potential, nitrogen deposition, diffusion limited by high soil moisture, and oxygen concentration ( $C_M$ ,  $T_{SOIL}$ ,  $E_{SM}$ ,  $R_{OX}$ ,  $N_{DP}$ ,  $D_{MS}$ , and  $C_{O2}$  respectively) (Eq. 6). We use Michaelis-Menten kinetics with  $k_{CH4,LAM}$  of 5  $\mu\text{M}$  for the CH<sub>4</sub> limitation (Eq. 7).

$$M_{O,TEM}(z,t) = O_{MAX}f(C_M(z,t))f(T_{SOIL}(z,t))f(E_{SM}(z,t))f(R_{OX}(z,t))f(N_{DP}(z,t))f(D_{MS}(z,t))f(C_{O2}(z,t)) \quad (6)$$

$$f(C_M(z,t)) = \frac{C_M(z,t)}{k_{CH4,LAM} + C_M(z,t)} \quad (7)$$

The remaining CH<sub>4</sub> is emitted to the surface with three different transport processes. First, gaseous and aqueous diffusion ( $F_D$ ) occur due to concentration gradients of CH<sub>4</sub> ( $\partial C_M(z,t)/\partial t$ ) (Eq. 8). The molecular diffusion coefficient ( $D$ ) in different soil layers depends on soil texture and soil moisture. Ebullition ( $R_E$ ) occurs when CH<sub>4</sub> bubble forms with  $C_M$  greater than  $\mu\text{mol L}^{-1}$ , and is calculated with a constant rate of  $K_e$  (1.0 h<sup>-1</sup>) (Eq. 9). Plant-mediated transport ( $R_p$ ) occurs for plants that function as a direct conduit for CH<sub>4</sub> to the atmosphere, and is functions of rate constant of 0.01 h<sup>-1</sup>, vegetation type, root density, vegetation growth, and soil CH<sub>4</sub> concentrations ( $K_p$ ,  $TR_{veg}$ ,  $f_{ROOT}$ ,  $f_{GROW}$ , and  $C_M$ , respectively) (Eq. 10)<sup>68</sup>.  $R_p$  depends on ecosystem-specific plant functional types and increases in a warmer soil due to the increase in vegetation growth. In TEM model, the soil profile was divided into 1-cm layers, and soil temperature, moisture, and CH<sub>4</sub> dynamics of TEM were simulated at an hourly time step<sup>24,27</sup>.

$$F_D(z,t) = -D(z) \frac{\partial C_M(z,t)}{\partial t} \quad (8)$$

$$R_E(z,t) = K_e f(C_M(z,t)) \quad (9)$$

$$R_p(z,t) = K_p TR_{veg} f_{ROOT}(z) f_{GROW}(t) C_M(z,t) \quad (10)$$

**Methane stable carbon isotope module in TEM (isoTEM).** IsoTEM explicitly considers carbon isotopic fractionation processes for precursor organic matter (POM) and CH<sub>4</sub> during production, oxidation, and transport process. The  $\delta^{13}\text{C}$  of POM ( $\delta^{13}\text{C}$ -POM) is determined by the global C<sub>3</sub> and C<sub>4</sub> vegetation distribution<sup>29</sup> and is set to  $-27\text{‰}$  and  $-13\text{‰}$  for C<sub>3</sub>- and C<sub>4</sub>-only vegetation areas, respectively. The  $\delta^{13}\text{C}$ -POM for areas with mixed C<sub>3</sub> and C<sub>4</sub> vegetation is determined by the proportion of each type of photosynthetic pathway (Supplementary Fig. 1). We also incorporated long-term trends of atmospheric  $\delta^{13}\text{C}$ -CO<sub>2</sub> into soil  $\delta^{13}\text{C}$ -POM changes. Atmospheric  $\delta^{13}\text{C}$ -CO<sub>2</sub> became depleted in <sup>13</sup>C by  $\approx 2\text{‰}$  during 1951–2016<sup>5,33</sup>, and this signal is transferred to photosynthates and POM for CH<sub>4</sub> emissions in wetlands<sup>69</sup>. We incorporated this trend with a 6-year carbon residence time between photosynthesis and CH<sub>4</sub> emission in wetlands (Supplementary Fig. 2)<sup>31</sup>.

The CH<sub>4</sub> is then produced in anaerobic soils by two distinct methanogen communities: hydrogenotrophic methanogens (HMs) use H<sub>2</sub> and CO<sub>2</sub> and

acetoclastic methanogens (AMs) use acetate ( $\text{CH}_3\text{COO}^-$ ) for  $\text{CH}_4$  production<sup>34</sup>. Both mechanisms produce equimolar amounts of  $\text{CO}_2$  and  $\text{CH}_4$  from cellulose-like substrates. Using in situ observations from Holmes et al.<sup>19</sup> the fractional contribution of the two methanogen communities is calculated based on a multiple regression analysis with the main environmental factors (Eq. 11). From the principal component analysis, Holmes et al.<sup>19</sup> found a combination of environmental parameters including pH, vegetation type, soil organic carbon (SOC), and latitude are correlated with the dominant methanogenic pathway. The regression results show that fractional contribution of HMs ( $f_{\text{HM}}$ ) is positively correlated with latitude with a steep increase at 60°N (slope of 0.11 and 5.19 for latitudes below and above 60°N, respectively), and negatively correlated with pH (slope of -9.23) and SOC (slope of -0.7) ( $R^2$  of 0.41,  $p < 0.001$ ) (Eq. 11, Supplementary Table 1, and Supplementary Fig. 3).

$$f_{\text{HM}} = \begin{cases} a_1 \times \text{lat} + b \times \text{pH} + c \times \text{SOC} + d & \dots \text{for latitude} < \text{latitude}_{\text{step}} \\ a_1 \times \text{lat} + a_2 \times (\text{latitude} - \text{latitude}_{\text{step}}) + b \times \text{pH} + c \times \text{SOC} + d & \dots \text{for latitude} > \text{latitude}_{\text{step}} \end{cases} \quad (11)$$

The  $\delta^{13}\text{C}\text{-CH}_4$  produced by HMs and AMs more negative than the  $\delta^{13}\text{C}\text{-POM}$ , with the fractionation factors for HMs ( $\alpha_{\text{HM}} \approx 1.030\text{--}1.080$  and for AMs ( $\alpha_{\text{AM}} \approx 1.000\text{--}1.040$ ) (Eq. 12). The produced  $\delta^{13}\text{C}\text{-CH}_4$  is calculated using a binary mixing of  $\text{CH}_4$  pools from the two methanogen communities (Eqs. 13, 14).

$$\alpha_{\text{HM}} = \frac{1000 + \delta^{13}\text{C}_{\text{POM}}}{1000 + \delta^{13}\text{CH}_{4,\text{prod,HM}}}, \alpha_{\text{AM}} = \frac{1000 + \delta^{13}\text{C}_{\text{POM}}}{1000 + \delta^{13}\text{CH}_{4,\text{prod,AM}}} \quad (12)$$

$$\delta^{13}\text{CH}_{4,\text{prod,HM}} = \delta^{13}\text{C}_{\text{POM}} - 1000 \times \ln(\alpha_{\text{HM}}), \delta^{13}\text{CH}_{4,\text{prod,AM}} = \delta^{13}\text{C}_{\text{POM}} - 1000 \times \ln(\alpha_{\text{AM}}) \quad (13)$$

$$\delta^{13}\text{CH}_{4,\text{prod}} = f_{\text{HM}} \times \delta^{13}\text{CH}_{4,\text{prod,HM}} + (1 - f_{\text{HM}}) \times \delta^{13}\text{CH}_{4,\text{prod,AM}} \quad (14)$$

The produced  $\text{CH}_4$  is partly oxidized by methanotrophs in aerobic soils, which prefer  $^{12}\text{CH}_4$ , thus  $\alpha$  for  $\text{CH}_4$  oxidation ( $\alpha_{\text{MO}} \approx 1.015\text{--}1.035$ ) (Eq. 15). Then, the produced  $\text{CH}_4$  is transported to the atmosphere through three processes, plant-mediated transport, diffusion, and ebullition, with different fractionation factors  $\alpha_{\text{TP}} \approx 1.000\text{--}1.030$ ,  $\alpha_{\text{TE}} \approx 1.000\text{--}1.010$ ,  $\alpha_{\text{TD}} \approx 1.000\text{--}1.005$ , respectively<sup>20</sup> (Eq. 16).

$$\alpha_{\text{MO}} = \frac{1000 + \delta^{13}\text{CH}_{4,\text{prod}}}{1000 + \delta^{13}\text{CH}_{4,\text{oxid}}} \quad (15)$$

$$\alpha_{\text{TP}} = \frac{1000 + \delta^{13}\text{CH}_{4,\text{prod}}}{1000 + \delta^{13}\text{CH}_{4,\text{TP}}}, \alpha_{\text{TE}} = \frac{1000 + \delta^{13}\text{CH}_{4,\text{prod}}}{1000 + \delta^{13}\text{CH}_{4,\text{TE}}}, \alpha_{\text{TD}} = \frac{1000 + \delta^{13}\text{CH}_{4,\text{prod}}}{1000 + \delta^{13}\text{CH}_{4,\text{TD}}} \quad (16)$$

We calculated the oxidized and transported  $\delta^{13}\text{C}\text{-CH}_4$  based on “open system equations” at steady state to consider residual enriched  $\text{CH}_4$  after oxidation and transport processes<sup>70–73</sup>. We approximated that  $\text{CH}_4$  produced in the entire vertical soil column is either oxidized or transported in each hourly time step (Eq. 17). In Eqs. 17, 18,  $M_p(z,t)$ ,  $M_o(z,t)$ ,  $R_p(z,t)$ , and  $R_E(z,t)$  represent  $\text{CH}_4$  production, oxidation, plant-mediated transport, and ebullition rates, respectively, and  $\partial F_D(z,t)/\partial z$  represents flux divergence due to gaseous and aqueous diffusion for each soil layer  $z$  and time  $t$ . For simplicity, we defined effective transport fractionation,  $\alpha_T$ , by flux-weighting the proportions of fractionation factors of three transport processes in Eq. 19. The isotopic difference between oxidation and transport processes can be described by a fractionation factor,  $\alpha_{\text{T/MO}}$ , in Eq. 20. Given these conditions, isotopic signatures for oxidation and transport to the atmosphere (emission) can be written in Eqs. 21, 22. For more details, refer to Hayes<sup>74</sup>.

$$\sum_z M_p(z,t) = \sum_z M_o(z,t) + \sum_z \frac{\partial F_D(z,t)}{\partial z} + \sum_z R_p(z,t) + \sum_z R_E(z,t) \quad (17)$$

$$f_{\text{ox}} = \frac{\sum_z M_o(z,t)}{\sum_z M_p(z,t)}, f_{\text{TP}} = \frac{\sum_z R_p(z,t)}{\sum_z M_p(z,t)}, f_{\text{TE}} = \frac{\sum_z R_E(z,t)}{\sum_z M_p(z,t)}, f_{\text{TD}} = \frac{\sum_z \frac{\partial F_D(z,t)}{\partial z}}{\sum_z M_p(z,t)} \quad (18)$$

$$\alpha_T = \frac{(f_{\text{TP}}\alpha_{\text{TP}} + f_{\text{TE}}\alpha_{\text{TE}} + f_{\text{TD}}\alpha_{\text{TD}})}{f_{\text{TP}} + f_{\text{TE}} + f_{\text{TD}}} \quad (19)$$

$$\alpha_{\text{T/MO}} = \frac{\alpha_{\text{MO}}}{\alpha_T} = \epsilon_{\text{T/MO}} + 1 \quad (20)$$

$$\delta^{13}\text{CH}_{4,\text{oxid}} = \frac{\delta^{13}\text{CH}_{4,\text{prod}} - (1 - f_{\text{ox}})\epsilon_{\text{T/MO}}}{\alpha_{\text{T/MO}}(1 - f_{\text{ox}}) + f_{\text{ox}}} \quad (21)$$

$$\delta^{13}\text{CH}_{4,\text{emitted}} = \frac{\alpha_{\text{T/MO}}\delta^{13}\text{CH}_{4,\text{prod}} + f_{\text{ox}}\epsilon_{\text{T/MO}}}{\alpha_{\text{T/MO}}(1 - f_{\text{ox}}) + f_{\text{ox}}} \quad (22)$$

**Model optimization.** We optimized 4 fractionation factors,  $\alpha_{\text{HM}}$ ,  $\alpha_{\text{AM}}$ ,  $\alpha_{\text{MO}}$ , and  $\alpha_{\text{TP}}$ , using in situ observations for six wetland ecosystem types (Eqs. 12, 15, 16). Since the fractionation factors for ebullition and diffusion are governed by physical processes, we set them as constants based on literature ( $\alpha_{\text{TE}} = 1.000$ ,  $\alpha_{\text{TD}} = 1.005$ )<sup>20</sup>. The wetland ecosystems are divided into forested and non-forested wetlands for boreal (50–90°N), temperate (30–50°N/S), and tropical (<30°N/S) regions. To optimize parameters, we collected observation data from six sites representing each ecosystem (Supplementary Tables 2–4)<sup>35,37,38</sup>. For tropical wetlands, we used observation data from Burke et al.<sup>38,75</sup>. For forested wetlands, we used data from “Willow Marsh Trail” station, a swamp wetland dominated by hardwoods and *Lemnaceae*. For non-forested wetlands, we used data from “St. Petersburg” site where Sawgrass is the dominant vegetation. For temperate wetlands, we used data from Kelly et al.<sup>37</sup>. For forested wetlands, we used data from “S2 Bog” where is entirely forested with *Picea mariana*. For non-forested wetlands, we used data from “Junction Fen” where is treeless and dominated by *Carex oligosperma*. For Arctic wetlands, we used data from McCalley et al.<sup>35</sup>. For forested wetlands, we could not find  $\delta^{13}\text{C}\text{-CH}_4$  data from the well-drained “Palsa” occupied by woody plants, mosses, and ericaceous. Thus, we used  $\delta^{13}\text{C}\text{-CH}_4$  data from “Sphagnum” site that is in the transition between the Palsa and Eriophorum sites, and showed similar  $\text{CH}_4$  fluxes as the “Palsa” site. For non-forested wetlands, we used data from the “Eriophorum” site.

Besides the observed meteorology from field sites, we also used CRU time-series version 4.01 to fill missing meteorological inputs<sup>76</sup>. We then used the Shuffled Complex Evolution Approach in R language (SCE-UA-R) to minimize the difference between simulated and observed  $\delta^{13}\text{C}\text{-CH}_4$ <sup>77</sup>. For each site, 20 ensembles were run using SCE-UA-R with 10,000 maximum loops per parameter ensemble, and all of them reached steady state before the end of the loops. Our optimization results show that isoTEM captures the magnitude and seasonality of observed soil  $\text{CH}_4$  fluxes and  $\delta^{13}\text{C}\text{-CH}_4$  (Supplementary Fig. 4).

**Simulation setup.** To estimate spatially- and temporally-varying  $\delta^{13}\text{C}\text{-CH}_4$  from global wetlands, we used spatially explicit data of land cover, soil pH and textures, meteorology and leaf area index (LAI)<sup>24,27</sup>. Land cover, soil pH and textures were used to assign vegetation-specific and texture-specific parameters to a grid cell<sup>78–80</sup>. Meteorological inputs were derived from historical air temperature, precipitation, vapor pressure, and cloudiness from gridded CRU time-series version 4.01<sup>76</sup>. We used monthly LAI derived from satellite imagery<sup>81</sup> to prescribe LAI for each  $0.5^\circ \times 0.5^\circ$  grid cell. All other parameters except fractionation factors were set the same as in Liu et al.<sup>27</sup>. We simulated global wetland  $\text{CH}_4$  fluxes and their isotopic ratios between 1984 and 2016 at a spatial resolution of  $0.5^\circ \times 0.5^\circ$  with a 50-year spin-up to let the carbon isotopic composition of carbon pools come to a steady state.

Because various wetland inundation data exist<sup>82</sup>, we first assumed that every global land grid cell can potentially be saturated, thus this product can be used with any wetland inundation data in future studies. To fill the grid cells without wetland types, we set forested and non-forested wetlands based on global vegetation types (Supplementary Fig. 5). In our analyses, simulated ecosystem-specific  $\delta^{13}\text{C}\text{-CH}_4$  from wetlands was flux weighted for each grid cell, based on  $\text{CH}_4$  emissions simulated by TEM defined over the static inundation data from Matthews and Fung (Supplementary Fig. 6a)<sup>49</sup>.

#### Model-data comparison

**Site level.** We compared our model results with previously published data from 58 in situ measurements compiled by Holmes et al.<sup>19</sup> and 66 in situ measurements by Sherwood et al.<sup>13</sup>. Holmes et al.<sup>19</sup> compiled latitude, fraction of HM and AM, pH, vegetation, and  $\delta^{13}\text{C}\text{-CH}_4$  to understand factors affecting the methanogenic pathway in global wetlands. The wetland database of Sherwood et al.<sup>13</sup> includes literature reference, latitude, wetland types, and measurement methods. After combining overlapped data of Holmes et al.<sup>19</sup> and Sherwood et al.<sup>13</sup> and excluding data that we used for our model optimization<sup>35,37,38</sup>, 70 sites remained for site-level validation (Supplementary Fig. 11 and Supplementary Data 1). Due to a possible mismatch of soil and vegetation properties, and wetland distribution of grid cells between model and observation, we compared observed  $\delta^{13}\text{C}\text{-CH}_4$  with simulated  $\delta^{13}\text{C}\text{-CH}_4$  of the sampling year within two adjacent grid cells ( $1^\circ \times 1^\circ$ ) of the observation.

**Regional level.** We used aircraft air samples from 3 regions in Alaska from the Carbon in Arctic Reservoirs Vulnerability Experiment (CARVE)<sup>83,84</sup>. From 2012 to 2015, CARVE collected airborne measurements of atmospheric chemical components and relevant land surface parameters in the Alaskan Arctic to provide insights into Arctic carbon cycling. During the flights, flask-air samples were collected then sent to NOAA GML for measurements of 50 trace gases including  $\text{CO}_2$ ,  $\text{CH}_4$ ,  $\text{CO}$ ,  $\text{OCS}$ ,  $\text{NMHCs}$ , and then sent to INSTAAR for and the isotopic composition of  $\text{CO}_2$  and  $\text{CH}_4$ . After excluding airborne data with flags, there are 1476 measurements during the sampling period.

In situ flux observations collected across Alaskan wetlands show an average of  $-65\%$  but a large  $9\%$  variation, due to the complex vegetation and soil properties<sup>40</sup>. To compare the spatial variability of wetland  $\delta^{13}\text{C}\text{-CH}_4$ , we divided the Alaskan continent into three regions: North Slope, interior, and southwest Alaska based on latitude ( $62\text{--}68^\circ\text{N}$ ,  $52\text{--}62^\circ\text{N}$  and  $140\text{--}155^\circ\text{W}$ , and  $52\text{--}62^\circ\text{N}$  and  $155\text{--}170^\circ\text{W}$  for North Slope, interior, and southwest Alaska, respectively). We used Miller-Tans plots to identify the source signatures of  $\delta^{13}\text{C}\text{-CH}_4$  from wetlands using the airborne measurements<sup>39</sup>. To identify wetland isotopic signatures, we removed measurements that may have effects from fossil fuel emission ( $\text{C}_3\text{H}_8 < 300$  ppt), biomass burning ( $\text{CO} < 300$  ppb), and transport influence (Altitude  $< 1500$  m), and we set the background altitude to  $> 5000$  m. After plotting the data, 2014 was excluded due to limited data and small  $R^2$  (Supplementary Table 5).

### Uncertainty and sensitivity tests

**Long-term trends in wetland  $\delta^{13}\text{C}\text{-CH}_4$  from observations.** We considered latitude, pH, and soil carbon as key parameters that determine variability of wetland  $\delta^{13}\text{C}\text{-CH}_4$  to run a linear regression using the site-level observations collected from global wetlands since the early 1980s (Supplementary Data 1). We added year as additional parameter for the linear regression and see if it improves the fit with data. The regression results show that wetland  $\delta^{13}\text{C}\text{-CH}_4$  is negatively correlated with year, latitude, and SOC (slope of  $-0.11$ ,  $-0.10$ , and  $-0.20$ , respectively), and positively correlated with pH (slope of  $2.21$ ) ( $R^2$  of  $0.30$ ,  $p < 0.001$ ) (Eq. 23, Supplementary Fig. 17, and Supplementary Table 6). The regression without year as a parameter showed smaller coefficient ( $R^2$  of  $0.25$ ,  $p < 0.001$ ).

$$\delta^{13}\text{C} - \text{CH}_4 = a \times \text{lat} + b \times \text{pH} + c \times \text{SOC} + d \times \text{year} + e \quad (23)$$

**Markov Chain Monte Carlo for the fraction of HM ( $f_{\text{HM}}$ ).** We used a Markov Chain Monte Carlo (MCMC) approach for parameter uncertainty estimation for  $f_{\text{HM}}$ . MCMC is a method for estimating the posterior probability density function for a set of parameters, given priors on those parameters and a set of observations<sup>45</sup>. We used independent, uniform prior probability density functions for each parameter in Supplementary Table 1. Thirty-nine data points from Holmes et al.<sup>19</sup> were used to constrain the model. Gaussian errors were assumed. We generated a Markov chain with 100,000 elements to estimate the joint posterior probability density functions. The chain converged after about 10,000 elements. We used the posterior probability density function to estimate the uncertainty of parameter (Supplementary Table 1).

**Sensitivity test with meteorological and substrate inputs,  $f_{\text{HM}}$ , and inundation.** We conducted 8 sensitivity tests of meteorology and substrate inputs. Specifically, we altered air temperature by  $\pm 3^\circ\text{C}$ , precipitation by  $\pm 30\%$ , and atmospheric  $\text{CH}_4$  abundance by  $\pm 30\%$ , and NPP by  $\pm 30\%$ , uniformly for each grid cell, while maintaining all other variables at their default isoTEM values. We also varied parameters for  $f_{\text{HM}}$  based on the uncertainty range from MCMC (Supplementary Table 1). We further varied a wetland inundation using satellite-driven Surface Water Microwave Product Series-Global Lakes and Wetlands Database (SWAMPS-GLWD)<sup>48</sup>.

### Forward modeling using TM5 atmospheric model

**Global mass balance for bottom-up inventory.** We adjusted global long-term mean fossil fluxes to match the simulated growth rate of  $\text{CH}_4$  during 1984–2016 and the 1998–2016 mean of  $\delta^{13}\text{C}\text{-CH}_4$  with observation (Table 1 and Supplementary Table 11). Lan et al.<sup>26</sup> showed that there is an offset of simulated global mean  $\delta^{13}\text{C}\text{-CH}_4$  when using EDGAR 4.3.2 inventory as the inventory underestimates fossil fluxes. To remove the offset and compare our scenarios fairly, we adjusted fossil fluxes between 170 and 190  $\text{TgCH}_4\text{yr}^{-1}$  (Supplementary Fig. 19), within the uncertainty range in Schwietzke et al.<sup>9</sup>. To satisfy the global mass balance, we ran one box model that included  $\text{CH}_4$  sources of biogenic, fossil and biomass/biofuel emissions, with corresponding isotopic signatures, and  $\text{CH}_4$  sinks due to reaction with OH, Cl, and  $\text{O}(^1\text{D})$  and soil bacteria, all with different fractionation factor. When we increased or decreased fossil fluxes, we accordingly decreased or increased ruminant flux, respectively, so the total annual  $\text{CH}_4$  fluxes followed the observed atmospheric  $\text{CH}_4$  growth rate, and the long-term mean total emission was set to  $536\text{--}538 \text{ TgCH}_4\text{yr}^{-1}$  during 1984–2016. For more details on the setup and equations for global mass balance, refer to Lan et al.<sup>26</sup>.

**Data sources for  $\text{CH}_4$  emissions and its isotopic source signatures.** We used the bottom-up inventory constructed by Lan et al.<sup>26</sup> (Supplementary Table 8). In specific, for  $\text{CH}_4$  emissions, we used GFED 4.1 s for biomass burning for 1997–2016<sup>85</sup> and annual emissions from the Reanalysis of Tropospheric chemical composition project before 1997, and the EDGAR 4.3.2 inventory for other anthropogenic emissions for 1984–2016<sup>86</sup>. For emissions from geological seeps, we used gridded emission from Etiope et al.<sup>87</sup>. Emission estimates from wild animals and termites were adopted from Bergamaschi et al.<sup>88</sup>. For  $\delta^{13}\text{C}\text{-CH}_4$  source signature, fossil fuel source signature data were based on the global  $\delta^{13}\text{C}\text{-CH}_4$  source signature inventory 2020<sup>89</sup>, where the data were categorized by coal gas, conventional gas, and shale gas. Biomass burning, biofuel burning, ruminant, and wild animal  $\delta^{13}\text{C}\text{-CH}_4$  data were based on the global maps of  $\text{C}_3/\text{C}_4$  distribution<sup>29</sup>. The geological seeps  $\delta^{13}\text{C}\text{-CH}_4$  data were from Etiope et al.<sup>87</sup>.

**TM5 atmospheric modeling of  $\text{CH}_4$  and  $\delta^{13}\text{C}\text{-CH}_4$ .** Atmospheric  $\text{CH}_4$  mole fractions and  $\delta^{13}\text{C}\text{-CH}_4$  were simulated from 1984 to 2016 by coupling the surface fluxes and isotope source signatures from the bottom-up inventory with the TM5 tracer transport model driven by ECMWF ERA Interim meteorology with the 4DVAR branch of the TM5 model<sup>90,91</sup>. TM5 was run globally at  $6^\circ \times 4^\circ$  over 25 vertical sigma-pressure hybrid levels, for total  $\text{CH}_4$  and  $^{13}\text{C}\text{-CH}_4$ . For each source type,  $^{13}\text{C}\text{-CH}_4$  fluxes were derived from total  $\text{CH}_4$  fluxes and source-specific isotope source signatures. We spun up our model during 1984–1999 and selected 2000–2016 to compare with atmospheric observations to ensure our spin-up period was sufficient for equilibration of atmospheric  $\delta^{13}\text{C}\text{-CH}_4$  inter-hemispheric gradient<sup>26,92</sup>. As per Lan et al.<sup>26</sup> we applied tropospheric Cl sink of Hossaini et al.<sup>51</sup> and the OH field from Spivakovsky et al.<sup>14</sup> with a fractionation factor of  $-3.9\%$ . The  $\text{CH}_4$  sinks varied spatially and seasonally but did not change interannually. For more details on setup for TM5 modeling, refer to Lan et al.<sup>26</sup>.

**Atmospheric  $\text{CH}_4$  and  $\delta^{13}\text{C}\text{-CH}_4$  measurement.** Observational data of atmospheric  $\text{CH}_4$  and  $\delta^{13}\text{C}\text{-CH}_4$  used to evaluate model results are from flask-air measurements from NOAA's Global Greenhouse Gas Reference Network<sup>26,54</sup>. The flask-air samples was analyzed for  $\delta^{13}\text{C}\text{-CH}_4$  at the Institute of Arctic and Alpine Research (INSTAAR), University of Colorado, Boulder. Gas chromatography-Isotope-ratio mass spectrometry (GC-IRMS) is used for  $\delta^{13}\text{C}\text{-CH}_4$  analysis<sup>5</sup>. The  $\delta^{13}\text{C}\text{-CH}_4$  in air measurements are referenced against the Vienna Pee Dee Belemnite (VPDB) standard (Eq. 1). A subset of the observation sites predominantly influenced by well-mixed background air is used to construct a Marine Boundary Layer (MBL) zonally averaged surface using methods developed by Masarie and Tans (1995)<sup>93</sup>, to represent the observational-based global long-term trend and north-south gradient. This includes 31 sites with  $\text{CH}_4$  measurements during study period of 1984–2016 and 10 of which with  $\delta^{13}\text{C}\text{-CH}_4$  measurements starting in 1998 (Supplementary Fig. 21 and Supplementary Table 10). More details on the MBL data products and uncertainties can be found at <https://www.esrl.noaa.gov/gmd/ccgg/mbl/mbl.html>. For model-observation comparisons, model results from the same set of MBL sites are sampled, and the same calculation methods are applied to model results and observations for global long-term and north-south gradient. The north-south gradient was calculated as the difference of atmospheric  $\delta^{13}\text{C}\text{-CH}_4$  between  $60\text{--}90^\circ\text{S}$  and  $60\text{--}90^\circ\text{N}$ .

**Atmospheric modeling with transient inundation data for Scenarios E-H.** Since we used static wetland inundation data<sup>49</sup> for our default Scenarios A–D, we used transient wetland inundation data from Poulter et al.<sup>48</sup> and ran TM5 atmospheric model (Supplementary Figs. 26–30 and Supplementary Table 11). Same as Scenarios A–C, we constructed Scenarios E–G with different wetland isotopic signature maps as inputs for TM5 atmospheric modeling in 1984–2016. In specific, the first uses a globally uniform wetland  $\delta^{13}\text{C}\text{-CH}_4$  of  $-62.3\%$ , the mean wetland signature from Ganesan et al.<sup>23</sup> (referred to as Scenario E), the other uses a static wetland isotope spatial map from Ganesan et al.<sup>23</sup> (referred to as Scenario F), and the last used spatially- and temporally-resolved maps from isoTEM (referred to as Scenario G).

The wetland fluxes for Scenarios E–G are based on Liu et al.<sup>27</sup> and transient inundation<sup>48</sup> but applied an increase in fluxes after 2006 by hypothesizing that the microbial wetland emission is a dominant driver of post-2006 atmospheric  $\text{CH}_4$  increase (Supplementary Fig. 26), same as Scenarios A–C. We also conducted the global mass balance by adjusting global long-term mean fossil fluxes between 160 and 180  $\text{TgCH}_4\text{yr}^{-1}$  for Scenarios E–G to match the simulated growth rate of  $\text{CH}_4$  during 1984–2016 and the 1998–2016 mean of annual  $\delta^{13}\text{C}\text{-CH}_4$  with observations.

Scenarios E–G reproduced the observed global  $\text{CH}_4$  growth rate during 1984–2016 and the global long-term mean  $\delta^{13}\text{C}\text{-CH}_4$  with observation during 1998–2016 (Supplementary Fig. 28), as we set the fluxes based on the mass balance. However, Scenarios E–G with transient inundation data underestimated the north-south  $\delta^{13}\text{C}\text{-CH}_4$  gradient ( $0.27 \pm 0.06\%$ ) compared with observations ( $0.45 \pm 0.05\%$ ) (Supplementary Fig. 29). Thus, we ran an additional scenario H that increased emissions from boreal wetlands by 2.5 times over the original transient data (Supplementary Fig. 26 and Supplementary Table 11), which improved the match with the observed north-south  $\delta^{13}\text{C}\text{-CH}_4$  gradient ( $0.39\%$ ) (Supplementary Fig. 29). The site-level comparison with atmospheric  $\delta^{13}\text{C}\text{-CH}_4$  from 10 observation sites also confirmed that Scenario H more closely reproduced the observation (Supplementary Fig. 30). This implies that the transient inundation data from Poulter et al.<sup>48</sup> may need more wetland emissions from boreal regions as found in static inundation data<sup>49</sup> (Supplementary Fig. 6) and other satellite-derived inundation data<sup>94</sup>.

### Data availability

Supplementary Data 1 is available at: [https://figshare.com/articles/dataset/Supplementary\\_Data\\_1\\_of\\_Oh\\_et\\_al\\_2022/19929965](https://figshare.com/articles/dataset/Supplementary_Data_1_of_Oh_et_al_2022/19929965). The stable carbon isotopic composition of wetland emissions is available at: <https://doi.org/10.25925/9s6n-g811>.

### Code availability

The code is also archived and available at: <https://doi.org/10.15138/cem6-ka15>.

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### Author contributions

Y.O., Q.Z., and X.L. conceived the study. Y.O., Q.Z., L.L., and L.R.W. built the model. E.J.D., S.E.M., J.B.M., S.S., and P.C. provided unpublished or raw data. Y.O. conducted model runs. S.B., L.B., P.T., and J.P.C., and all other authors contributed to data interpretation and preparation of paper text.

### Competing interests

The authors declare no competing interests.

### Additional information

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**Correspondence** and requests for materials should be addressed to Youmi Oh or Qianlai Zhuang.

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