



The Global Methane Budget 2000–2017

Marielle Saunois¹, Ann R. Stavert², Ben Poulter³, Philippe Bousquet¹, Josep G. Canadell², Robert B. Jackson⁴, Peter A. Raymond⁵, Edward J. Dlugokencky⁶, Sander Houweling^{7,8}, Prabir K. Patra^{9,10}, Philippe Ciais¹, Vivek K. Arora¹¹, David Bastviken¹², Peter Bergamaschi¹³, Donald R. Blake¹⁴, Gordon Brailsford¹⁵, Lori Bruhwiler⁶, Kimberly M. Carlson^{16,17}, Mark Carrol⁷⁰, Simona Castaldi^{18,19,20}, Naveen Chandra⁹, Cyril Crevoisier²¹, Patrick M. Crill²², Kristofer Covey²³, Charles L. Curry^{24,71}, Giuseppe Etiope^{25,26}, Christian Frankenberg^{27,28}, Nicola Gedney²⁹, Michaela I. Hegglin³⁰, Lena Höglund-Isaksson³¹, Gustaf Hugelius³², Misa Ishizawa³³, Akihiko Ito³³, Greet Janssens-Maenhout¹³, Katherine M. Jensen³⁴, Fortunat Joos³⁵, Thomas Kleinen³⁶, Paul B. Krummel³⁷, Ray L. Langenfelds³⁷, Goulven G. Laruelle³⁸, Licheng Liu³⁹, Toshinobu Machida³³, Shamil Maksyutov³³, Kyle C. McDonald³⁴, Joe McNorton⁴⁰, Paul A. Miller⁴¹, Joe R. Melton⁴², Isamu Morino³³, Jurek Müller³⁵, Fabiola Murguia-Flores⁴³, Vaishali Naik⁴⁴, Yosuke Niwa^{33,45}, Sergio Noce²⁰, Simon O'Doherty⁴⁶, Robert J. Parker⁴⁷, Changhui Peng⁴⁸, Shushi Peng⁴⁹, Glen P. Peters⁵⁰, Catherine Prigent⁵¹, Ronald Prinn⁵², Michel Ramonet¹, Pierre Regnier³⁸, William J. Riley⁵³, Judith A. Rosentreter⁵⁴, Arjo Segers⁵⁵, Isobel J. Simpson¹⁴, Hao Shi⁵⁶, Steven J. Smith^{57,58}, L. Paul Steele³⁷, Brett F. Thornton²², Hanqin Tian⁵⁶, Yasunori Tohjima⁷², Francesco N. Tubiello⁵⁹, Aki Tsuruta⁶⁰, Nicolas Viovy¹, Apostolos Voulgarakis^{61,62}, Thomas S. Weber⁶³, Michiel van Weele⁶⁴, Guido R. van der Werf⁸, Ray F. Weiss⁶⁵, Doug Worthy⁶⁶, Debra Wunch⁶⁷, Yi Yin^{1,27}, Yukio Yoshida³³, Wenxin Zhang⁴¹, Zhen Zhang⁶⁸, Yuanhong Zhao¹, Bo Zheng¹, Qing Zhu⁵³, Qian Zhu⁶⁹, and Qianlai Zhuang³⁹

¹Laboratoire des Sciences du Climat et de l'Environnement, LSCE-IPSL (CEA-CNRS-UVSQ), Université Paris-Saclay 91191 Gif-sur-Yvette, France

²Global Carbon Project, CSIRO Oceans and Atmosphere, Aspendale, VIC 3195 & Canberra, ACT 2601, Australia

³NASA Goddard Space Flight Center, Biospheric Science Laboratory, Greenbelt, MD 20771, USA

⁴Department of Earth System Science, Woods Institute for the Environment, and Precourt Institute for Energy, Stanford University, Stanford, CA 94305-2210, USA

⁵Yale School of the Environment, Yale University, New Haven, CT 06511, USA

⁶NOAA Global Monitoring Laboratory, 325 Broadway, Boulder, CO 80305, USA

⁷SRON Netherlands Institute for Space Research, Sorbonnelaan 2, 3584 CA Utrecht, the Netherlands

⁸Vrije Universiteit Amsterdam, Department of Earth Sciences, Earth and Climate Cluster, VU Amsterdam, Amsterdam, the Netherlands

⁹Research Institute for Global Change, JAMSTEC, 3173-25 Showa-machi, Kanazawa, Yokohama, 236-0001, Japan

¹⁰Center for Environmental Remote Sensing, Chiba University, Chiba, Japan

¹¹Canadian Centre for Climate Modelling and Analysis, Climate Research Division, Environment and Climate Change Canada, Victoria, BC, V8W 2Y2, Canada

¹²Department of Thematic Studies – Environmental Change, Linköping University, 581 83 Linköping, Sweden

¹³European Commission Joint Research Centre, Via E. Fermi 2749, 21027 Ispra (Va), Italy

¹⁴Department of Chemistry, University of California Irvine, 570 Rowland Hall, Irvine, CA 92697, USA

¹⁵National Institute of Water and Atmospheric Research, 301 Evans Bay Parade, Wellington, New Zealand

¹⁶Department of Environmental Studies, New York University, New York, NY 10003, USA

¹⁷Department of Natural Resources and Environmental Management, University of Hawai'i, Honolulu, HI 96822, USA

- ¹⁸Dipartimento di Scienze Ambientali, Biologiche e Farmaceutiche, Università degli Studi della Campania Luigi Vanvitelli, via Vivaldi 43, 81100 Caserta, Italy
- ¹⁹Department of Landscape Design and Sustainable Ecosystems, RUDN University, Moscow, Russia
- ²⁰Impacts on Agriculture, Forests, and Ecosystem Services Division, Centro Euro-Mediterraneo sui Cambiamenti Climatici, Via Augusto Imperatore 16, 73100 Lecce, Italy
- ²¹Laboratoire de Météorologie Dynamique, LMD-IPSL, Ecole Polytechnique, 91120 Palaiseau, France
- ²²Department of Geological Sciences and Bolin Centre for Climate Research, Stockholm University, Svante Arrhenius väg 8, 106 91 Stockholm, Sweden
- ²³Environmental Studies and Sciences Program, Skidmore College, Saratoga Springs, NY 12866, USA
- ²⁴Pacific Climate Impacts Consortium, University of Victoria, University House 1, P.O. Box 1700 STN CSC Victoria, BC V8W 2Y2, Canada
- ²⁵Istituto Nazionale di Geofisica e Vulcanologia, Sezione Roma 2, via V. Murata 605 00143 Rome, Italy
- ²⁶Faculty of Environmental Science and Engineering, Babes Bolyai University, Cluj-Napoca, Romania
- ²⁷Division of Geological and Planetary Sciences, California Institute of Technology, Pasadena, CA 91125, USA
- ²⁸Jet Propulsion Laboratory, California Institute of Technology, Pasadena, CA 91125, USA
- ²⁹Met Office Hadley Centre, Joint Centre for Hydrometeorological Research, Maclean Building, Wallingford OX10 8BB, UK
- ³⁰Department of Meteorology, University of Reading, Earley Gate, Reading RG6 6BB, UK
- ³¹Air Quality and Greenhouse Gases Program (AIR), International Institute for Applied Systems Analysis (IIASA), 2361 Laxenburg, Austria
- ³²Department of Physical Geography and Bolin Centre for Climate Research, Stockholm University, 106 91 Stockholm, Sweden
- ³³Center for Global Environmental Research, National Institute for Environmental Studies (NIES), Onogawa 16-2, Tsukuba, Ibaraki 305-8506, Japan
- ³⁴Department of Earth and Atmospheric Sciences, City College of New York, City University of New York, New York, NY 10031, USA
- ³⁵Climate and Environmental Physics, Physics Institute and Oeschger Centre for Climate Change Research, University of Bern, Sidlerstr. 5, 3012 Bern, Switzerland
- ³⁶Max Planck Institute for Meteorology, Bundesstr. 53, 20146 Hamburg, Germany
- ³⁷Climate Science Centre, CSIRO Oceans and Atmosphere, Aspendale, Victoria 3195, Australia
- ³⁸Department Geoscience, Environment & Society, Université Libre de Bruxelles, 1050-Brussels, Belgium
- ³⁹Department of Earth, Atmospheric, Planetary Sciences, Department of Agronomy, Purdue University, West Lafayette, IN 47907, USA
- ⁴⁰Research Department, European Centre for Medium-Range Weather Forecasts, Reading, UK
- ⁴¹Department of Physical Geography and Ecosystem Science, Lund University, Sölvegatan 12, 223 62, Lund, Sweden
- ⁴²Climate Research Division, Environment and Climate Change Canada, Victoria, BC, V8W 2Y2, Canada
- ⁴³School of Geographical Sciences, University of Bristol, Bristol, BS8 1SS, UK
- ⁴⁴NOAA/Geophysical Fluid Dynamics Laboratory (GFDL), 201 Forrestal Rd., Princeton, NJ 08540, USA
- ⁴⁵Meteorological Research Institute (MRI), Nagamine 1-1, Tsukuba, Ibaraki 305-0052, Japan
- ⁴⁶School of Chemistry, University of Bristol, Cantock's Close, Clifton, Bristol BS8 1TS, UK
- ⁴⁷National Centre for Earth Observation, University of Leicester, Leicester, LE1 7RH, UK
- ⁴⁸Department of Biology Sciences, Institute of Environment Science, University of Quebec at Montreal, Montreal, QC H3C 3P8, Canada
- ⁴⁹Sino-French Institute for Earth System Science, College of Urban and Environmental Sciences, Peking University, Beijing 100871, China
- ⁵⁰CICERO Center for International Climate Research, Pb. 1129 Blindern, 0318 Oslo, Norway
- ⁵¹Observatoire de Paris, Université PSL, Sorbonne Université, CNRS, LERMA, Paris, France
- ⁵²Department of Earth, Atmospheric and Planetary Sciences, Massachusetts Institute of Technology (MIT), Building 54-1312, Cambridge, MA 02139, USA
- ⁵³Climate and Ecosystem Sciences Division, Lawrence Berkeley National Lab, 1 Cyclotron Road, Berkeley, CA 94720, USA
- ⁵⁴Centre for Coastal Biogeochemistry, School of Environment, Science and Engineering, Southern Cross University, Lismore, NSW 2480, Australia

- ⁵⁵TNO, Dep. of Climate Air & Sustainability, P.O. Box 80015, NL-3508-TA, Utrecht, the Netherlands
- ⁵⁶International Center for Climate and Global Change Research, School of Forestry and Wildlife Sciences, Auburn University, 602 Duncan Drive, Auburn, AL 36849, USA
- ⁵⁷Joint Global Change Research Institute, Pacific Northwest National Lab, College Park, MD 20740, USA
- ⁵⁸Department of Atmospheric and Oceanic Science, University of Maryland, College Park, MD 20740, USA
- ⁵⁹Statistics Division, Food and Agriculture Organization of the United Nations (FAO), Viale delle Terme di Caracalla, 00153 Rome, Italy
- ⁶⁰Finnish Meteorological Institute, P.O. Box 503, 00101, Helsinki, Finland
- ⁶¹Department of Physics, Imperial College London, London SW7 2AZ, UK
- ⁶²School of Environmental Engineering, Technical University of Crete, Chania, Greece
- ⁶³Department of Earth and Environmental Sciences, University of Rochester, Rochester, NY 14627, USA
- ⁶⁴KNMI, P.O. Box 201, 3730 AE, De Bilt, the Netherlands
- ⁶⁵Scripps Institution of Oceanography (SIO), University of California San Diego, La Jolla, CA 92093, USA
- ⁶⁶Environment and Climate Change Canada, 4905, rue Dufferin, Toronto, Canada
- ⁶⁷Department of Physics, University of Toronto, 60 St. George Street, Toronto, Ontario, Canada
- ⁶⁸Department of Geographical Sciences, University of Maryland, College Park, MD 20740, USA
- ⁶⁹College of Hydrology and Water Resources, Hohai University, Nanjing, 210098, China
- ⁷⁰NASA Goddard Space Flight Center, Computational and Information Science and Technology Office, Greenbelt, MD 20771, USA
- ⁷¹School of Earth and Ocean Sciences, University of Victoria, P.O. Box 1700 STN CSC, Victoria, V8W 2Y2 BC, Canada
- ⁷²Center for Environmental Measurement and Analysis, National Institute for Environmental Studies (NIES), Onogawa16-2, Tsukuba, Ibaraki 305-8506, Japan

Correspondence: Marielle Saunois (marielle.saunois@lscce.ipsl.fr)

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Abstract. Understanding and quantifying the global methane (CH₄) budget is important for assessing realistic pathways to mitigate climate change. Atmospheric emissions and concentrations of CH₄ continue to increase, making CH₄ the second most important human-influenced greenhouse gas in terms of climate forcing, after carbon dioxide (CO₂). The relative importance of CH₄ compared to CO₂ depends on its shorter atmospheric lifetime, stronger warming potential, and variations in atmospheric growth rate over the past decade, the causes of which are still debated. Two major challenges in reducing uncertainties in the atmospheric growth rate arise from the variety of geographically overlapping CH₄ sources and from the destruction of CH₄ by short-lived hydroxyl radicals (OH). To address these challenges, we have established a consortium of multidisciplinary scientists under the umbrella of the Global Carbon Project to synthesize and stimulate new research aimed at improving and regularly updating the global methane budget. Following Saunois et al. (2016), we present here the second version of the living review paper dedicated to the decadal methane budget, integrating results of top-down studies (atmospheric observations within an atmospheric inverse-modelling framework) and bottom-up estimates (including process-based models for estimating land surface emissions and atmospheric chemistry, inventories of anthropogenic emissions, and data-driven extrapolations).

For the 2008–2017 decade, global methane emissions are estimated by atmospheric inversions (a top-down approach) to be 576 Tg CH₄ yr⁻¹ (range 550–594, corresponding to the minimum and maximum estimates of the model ensemble). Of this total, 359 Tg CH₄ yr⁻¹ or ~ 60 % is attributed to anthropogenic sources, that is emissions caused by direct human activity (i.e. anthropogenic emissions; range 336–376 Tg CH₄ yr⁻¹ or 50 %–65 %). The mean annual total emission for the new decade (2008–2017) is 29 Tg CH₄ yr⁻¹ larger than our estimate for the previous decade (2000–2009), and 24 Tg CH₄ yr⁻¹ larger than the one reported in the previous budget for 2003–2012 (Saunois et al., 2016). Since 2012, global CH₄ emissions have been tracking the warmest scenarios assessed by the Intergovernmental Panel on Climate Change. Bottom-up methods suggest almost 30 % larger global emissions (737 Tg CH₄ yr⁻¹, range 594–881) than top-down inversion methods. Indeed, bottom-up estimates for natural sources such as natural wetlands, other inland water systems, and geological sources are higher than top-down estimates. The atmospheric constraints on the top-down budget suggest that at least some of these bottom-up emissions are overestimated. The latitudinal distribution of atmospheric observation-based emissions indicates a predominance of tropical emissions (~ 65 % of the global budget, < 30° N) compared to

mid-latitudes ($\sim 30\%$, $30\text{--}60^\circ\text{N}$) and high northern latitudes ($\sim 4\%$, $60\text{--}90^\circ\text{N}$). The most important source of uncertainty in the methane budget is attributable to natural emissions, especially those from wetlands and other inland waters.

Some of our global source estimates are smaller than those in previously published budgets (Saunio et al., 2016; Kirschke et al., 2013). In particular wetland emissions are about $35\text{ Tg CH}_4\text{ yr}^{-1}$ lower due to improved partition wetlands and other inland waters. Emissions from geological sources and wild animals are also found to be smaller by $7\text{ Tg CH}_4\text{ yr}^{-1}$ by $8\text{ Tg CH}_4\text{ yr}^{-1}$, respectively. However, the overall discrepancy between bottom-up and top-down estimates has been reduced by only 5% compared to Saunio et al. (2016), due to a higher estimate of emissions from inland waters, highlighting the need for more detailed research on emissions factors. Priorities for improving the methane budget include (i) a global, high-resolution map of water-saturated soils and inundated areas emitting methane based on a robust classification of different types of emitting habitats; (ii) further development of process-based models for inland-water emissions; (iii) intensification of methane observations at local scales (e.g., FLUXNET- CH_4 measurements) and urban-scale monitoring to constrain bottom-up land surface models, and at regional scales (surface networks and satellites) to constrain atmospheric inversions; (iv) improvements of transport models and the representation of photochemical sinks in top-down inversions; and (v) development of a 3D variational inversion system using isotopic and/or co-emitted species such as ethane to improve source partitioning.

The data presented here can be downloaded from <https://doi.org/10.18160/GCP-CH4-2019> (Saunio et al., 2020) and from the Global Carbon Project.

1 Introduction

The surface dry air mole fraction of atmospheric methane (CH_4) reached 1857 ppb in 2018 (Fig. 1), approximately 2.6 times greater than its estimated pre-industrial equilibrium value in 1750. This increase is attributable in large part to increased anthropogenic emissions arising primarily from agriculture (e.g., livestock production, rice cultivation, biomass burning), fossil fuel production and use, waste disposal, and alterations to natural methane fluxes due to increased atmospheric CO_2 concentrations and climate change (Ciais et al., 2013). Atmospheric CH_4 is a stronger absorber of Earth's emitted thermal infrared radiation than carbon dioxide (CO_2), as assessed by its global warming potential (GWP) relative to CO_2 . For a 100-year time horizon and without considering climate feedbacks $\text{GWP}(\text{CH}_4) = 28$ (IPCC AR5; Myhre et al., 2013). Although global anthropogenic emissions of CH_4 are estimated at around $366\text{ Tg CH}_4\text{ yr}^{-1}$ (Saunio et al., 2016), representing only 3% of the global CO_2 anthropogenic emissions in units of carbon mass flux, the increase in atmospheric CH_4 concentrations has contributed $\sim 23\%$ ($\sim 0.62\text{ W m}^{-2}$) to the additional radiative forcing accumulated in the lower atmosphere since 1750 (Etminan et al., 2016). Changes in other chemical compounds (such as nitrogen oxides, NO_x , or carbon monoxide, CO) also influence the forcing of atmospheric CH_4 through changes to its atmospheric lifetime. From an emission perspective, the total radiative forcing attributable to anthropogenic CH_4 emissions is currently about 0.97 W m^{-2} (Myhre et al., 2013). Emissions of CH_4 contribute to the production of ozone, stratospheric water vapour, and CO_2 and most importantly affect its own life-

time (Myhre et al., 2013; Shindell et al., 2012). CH_4 has a short lifetime in the atmosphere (about 9 years for the year 2010; Prather et al., 2012); hence a stabilization or reduction of CH_4 emissions leads rapidly, in a few decades, to a stabilization or reduction of its atmospheric concentration and therefore its radiative forcing. Reducing CH_4 emissions is therefore recognized as an effective option for rapid climate change mitigation, especially on decadal timescales (Shindell et al., 2012), because of its shorter lifetime than CO_2 .

Of concern, the current anthropogenic methane emissions trajectory is estimated to lie between the two warmest IPCC-AR5 scenarios (Nisbet et al., 2016, 2019), i.e., RCP8.5 and RCP6.0, corresponding to temperature increases above 3°C by the end of this century. This trajectory implies that large reductions of methane emissions are needed to meet the $1.5\text{--}2^\circ\text{C}$ target of the Paris Agreement (Collins et al., 2013; Nisbet et al., 2019). Moreover, CH_4 is a precursor of important air pollutants such as ozone, and, as such, its emissions are covered by two international conventions: the United Nations Framework Convention on Climate Change (UNFCCC) and the Convention on Long-Range Transboundary Air Pollution (CLRTAP), another motivation to reduce its emissions.

Changes in the magnitude and temporal variation (annual to inter-annual) in methane sources and sinks over the past decades are characterized by large uncertainties (Kirschke et al., 2013; Saunio et al., 2017; Turner et al., 2019). Also, the decadal budget suggests relative uncertainties (hereafter reported as min–max ranges) of $20\%\text{--}35\%$ for inventories of anthropogenic emissions in specific sectors (e.g., agriculture, waste, fossil fuels), 50% for biomass burning and natural wetland emissions, and reaching 100% or more for other natural sources (e.g. inland waters, geological sources).

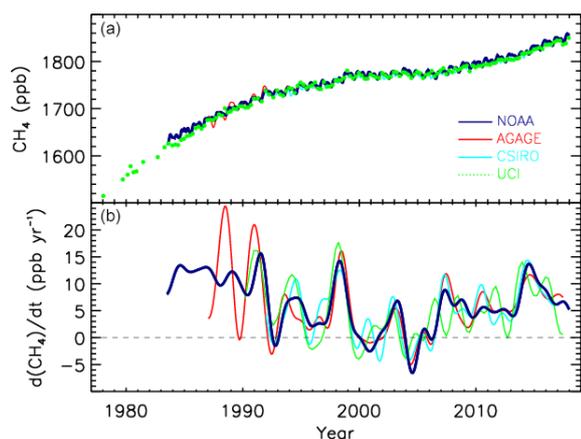


Figure 1. Globally averaged atmospheric CH_4 (ppb) (a) and its annual growth rate G_{ATM} (ppb yr^{-1}) (b) from four measurement programmes, National Oceanic and Atmospheric Administration (NOAA), Advanced Global Atmospheric Gases Experiment (AGAGE), Commonwealth Scientific and Industrial Research Organisation (CSIRO), and University of California, Irvine (UCI). Detailed descriptions of methods are given in the supplementary material of Kirschke et al. (2013).

The uncertainty in the chemical loss of methane by OH, the predominant sink of atmospheric methane, is estimated around 10 % (Prather et al., 2012) to 15 % (from bottom-up approaches in Saunois et al., 2016). This represents, for the top-down methods, the minimum relative uncertainty associated with global methane emissions, as other methane sinks (atomic oxygen and chlorine oxidations, soil uptake) are much smaller and the atmospheric growth rate is well-defined (Dlugokencky et al., 2009). Globally, the contribution of natural CH_4 emissions to total emissions can be quantified by combining lifetime estimates with reconstructed pre-industrial atmospheric methane concentrations from ice cores (e.g. Ehhalt et al., 2001). Regionally, uncertainties in emissions may reach 40 %–60 % (e.g. for South America, Africa, China, and India; see Saunois et al., 2016).

In order to verify future emission reductions, for example to help conduct Paris Agreement's stocktake, sustained and long-term monitoring of the methane cycle is needed to reach more precise estimation of trends, and reduced uncertainties in anthropogenic emissions (Bergamaschi et al., 2018a; Pacala, 2010). Reducing uncertainties in individual methane sources and thus in the overall methane budget is challenging for at least four reasons. Firstly, methane is emitted by a variety of processes, including both natural and anthropogenic sources, point and diffuse sources, and sources associated with three different emission classes (i.e., biogenic, thermogenic, and pyrogenic). These multiple sources and processes require the integration of data from diverse scientific communities. The fact that anthropogenic emissions result from unintentional leakage from fossil fuel production or agriculture further complicates production of accu-

rate bottom-up emission estimates. Secondly, atmospheric methane is removed by chemical reactions in the atmosphere involving radicals (mainly OH) that have very short lifetimes (typically ~ 1 s). The spatial and temporal distributions of OH are highly variable. Although OH can be measured locally, calculating global CH_4 loss through OH measurements would require high-resolution OH measurements (typically half an hour to integrate cloud cover and 1 km spatially to consider OH high reactivity and heterogeneity). As a result, such a calculation is currently possible only through modelling. However, simulated OH concentrations from chemistry–climate models still show uncertain spatio-temporal distribution at regional to global scales (Zhao et al., 2019). Thirdly, only the net methane budget (sources minus sinks) is constrained by precise observations of atmospheric growth rates (Dlugokencky et al., 2009), leaving the sum of sources and the sum of sinks more uncertain. One simplification for CH_4 compared to CO_2 is that the oceanic contribution to the global methane budget is small ($\sim 1\%$ – 3%), making source estimation predominantly a continental problem (USEPA, 2010b). Finally, we lack observations to constrain (1) process models that produce estimates of wetland extent (Kleinen et al., 2012; Stocker et al., 2014) and wetland emissions (Melton et al., 2013; Poulter et al., 2017; Wania et al., 2013), (2) other inland water sources (Bastviken et al., 2011; Wik et al., 2016a), (3) inventories of anthropogenic emissions (Höglund-Isaksson, 2012, 2017; Janssens-Maenhout et al., 2019; USEPA, 2012), and (4) atmospheric inversions, which aim to estimate methane emissions from global to regional scales (Bergamaschi et al., 2013, 2018b; Houweling et al., 2014; Kirschke et al., 2013; Saunois et al., 2016; Spahni et al., 2011; Thompson et al., 2017; Tian et al., 2016).

The global methane budget inferred from atmospheric observations by atmospheric inversions relies on regional constraints from atmospheric sampling networks, which are relatively dense for northern mid-latitudes, with a number of high-precision and high-accuracy surface stations, but are sparser at tropical latitudes and in the Southern Hemisphere (Dlugokencky et al., 2011). Recently the atmospheric observation density has increased in the tropics due to satellite-based platforms that provide column-average methane mixing ratios. Despite continuous improvements in the precision and accuracy of space-based measurements (e.g. Buchwitz et al., 2017), systematic errors greater than several parts per billion on total column observations can still limit the usage of such data to constrain surface emissions (Alexe et al., 2015; Bousquet et al., 2018; Chevallier et al., 2017; Locatelli et al., 2015). The development of robust bias corrections on existing data can help overcome this issue (e.g. Inoue et al., 2016) and satellite-based inversions have been suggested to reduce global and regional flux uncertainties compared to surface-based inversions (e.g. Fraser et al., 2013).

The Global Carbon Project (GCP) seeks to develop a complete picture of the carbon cycle by establishing common, consistent scientific knowledge to support policy de-

bate and actions to mitigate greenhouse gas emissions to the atmosphere (<https://www.globalcarbonproject.org/>, last access: 24 June 2020). The objective of this paper is to analyse and synthesize the current knowledge of the global methane budget, by gathering results of observations and models in order to better understand and quantify the main robust features of this budget and its remaining uncertainties and to make recommendations. We combine results from a large ensemble of bottom-up approaches (e.g., process-based models for natural wetlands, data-driven approaches for other natural sources, inventories of anthropogenic emissions and biomass burning, and atmospheric chemistry models) and top-down approaches (including methane atmospheric observing networks, atmospheric inversions inferring emissions, and sinks from the assimilation of atmospheric observations into models of atmospheric transport and chemistry). The focus of this work is on decadal budgets and on the update of the previous assessment made for the period 2003–2012 to the more recent 2008–2017 decade. More in-depth analysis of trends and year-to-year changes is left to future publications. The regional budget is further discussed in Stavert et al. (2020) and synthesised in Jackson et al., 2020. Our current paper is a living review, published at about 3-year intervals, to provide an update and new synthesis of available observational, statistical, and model data for the overall CH₄ budget and its individual components.

Kirschke et al. (2013) were the first to conduct a CH₄ budget synthesis and were followed by Saunois et al. (2016). Kirschke et al. (2013) reported decadal mean CH₄ emissions and sinks from 1980 to 2009 based on bottom-up and top-down approaches. Saunois et al. (2016) reported methane emissions for three time periods: (1) the last calendar decade (2000–2009), (2) the last available decade (2003–2012), and (3) the last available year (2012) at the time. Here, we update reporting methane emissions and sinks for 2000–2009 decade, for the most recent 2008–2017 decade where data are available, and for the year 2017, reducing the time lag between the last reported year and analysis. The methane budget is presented here at global and latitudinal scales, and data can be downloaded from <https://doi.org/10.18160/GCP-CH4-2019> (Saunois et al., 2019).

Five sections follow this introduction. Section 2 presents the methodology used in the budget (units, definitions of source categories and regions, data analysis) and discusses the delay between the period of study of the budget and the release date. Section 3 presents the current knowledge about methane sources and sinks based on the ensemble of bottom-up approaches reported here (models, inventories, data-driven approaches). Section 4 reports atmospheric observations and top-down atmospheric inversions gathered for this paper. Section 5, based on Sects. 3 and 4, provides the updated analysis of the global methane budget by comparing bottom-up and top-down estimates and highlighting differences. Finally, Sect. 6 discusses future developments, miss-

ing components, and the most critical remaining uncertainties based on our update to the global methane budget.

2 Methodology

2.1 Units used

Unless specified, fluxes are expressed in teragrams of CH₄ per year ($1 \text{ Tg CH}_4 \text{ yr}^{-1} = 10^{12} \text{ g CH}_4 \text{ yr}^{-1}$), while atmospheric concentrations are expressed as dry air mole fractions, in parts per billion (ppb), with atmospheric methane annual increases, G_{ATM} , expressed in parts per billion per year. In the tables, we present mean values and ranges for the two decades 2000–2009 and 2008–2017, together with results for the most recent available year (2017). Results obtained from previous syntheses (i.e. Saunois et al., 2016) are also given for the decade 2000–2009. Following Saunois et al. (2016) and considering that the number of studies is often relatively small for many individual source and sink estimates, uncertainties are reported as minimum and maximum values of the available studies, in brackets. In doing so, we acknowledge that we do not consider the uncertainty of the individual estimates, and we express uncertainty as the range of available mean estimates, i.e., differences across measurements and methodologies considered. These minimum and maximum values are those presented in Sect. 2.5 and exclude identified outliers.

The CH₄ emission estimates are provided with up to three digits, for consistency across all budget flux components and to ensure the accuracy of aggregated fluxes. Nonetheless, given the values of the uncertainties in the methane budget, we encourage the reader to consider not more than two digits as significant.

2.2 Period of the budget and availability of data

The bottom-up estimates rely on global anthropogenic inventories, land surface models for wetland emissions, and published literature for other natural sources. The global gridded anthropogenic inventories are updated irregularly, generally every 3 to 5 years. The last reported years of available inventories were 2012, 2014, or 2016 when we started this study. For this budget, in order to cover the reported period (2000–2017), it was necessary to extrapolate some of these datasets as explained in Sect. 3.1.1. The surface land models were run over the full period 2000–2017 using dynamical wetland areas (Sect. 3.2.1).

For the top-down estimates, we use atmospheric inversions covering 2000–2017. The simulations run until mid-2018, but the last year of reported inversion results is 2017, which represents a 3-year lag with the present, a 2-year-shorter lag than for the last release (Saunois et al., 2016). Satellite observations are linked to operational data chains and are generally available days to weeks after the recording of the spectra. Surface observations can lag from months to years because

of the time for flask analyses and data checks in (mostly) non-operational chains. The final 6 months of inversions are generally ignored (spin down) because the estimated fluxes are not constrained by as many observations as the previous periods.

2.3 Definition of regions

Geographically, emissions are reported globally and for three latitudinal bands (90° S–30° N, 30–60° N, 60–90° N, only for gridded products). When extrapolating emission estimates forward in time (see Sect. 3.1.1), and for the regional budget presented by Stavert et al. (2020), a set of 19 regions (oceans and 18 continental regions; see Fig. S1 in the Supplement) were used. As anthropogenic emissions are often reported by country, we define these regions based on a country list (Table S1). This approach was compatible with all top-down and bottom-up approaches considered. The number of regions was chosen to be close to the widely used TransCom inter-comparison map (Gurney et al., 2004) but with subdivisions to separate the contribution from important countries or regions for the methane cycle (China, South Asia, tropical America, tropical Africa, the United States, and Russia). The resulting region definition is the same as used for the GCP N₂O budget (Tian et al., 2019).

2.4 Definition of source categories

Methane is emitted by different processes (i.e., biogenic, thermogenic, or pyrogenic) and can be of anthropogenic or natural origin. Biogenic methane is the final product of the decomposition of organic matter by methanogenic *Archaea* in anaerobic environments, such as water-saturated soils, swamps, rice paddies, marine sediments, landfills, sewage and wastewater treatment facilities, or inside animal digestive systems. Thermogenic methane is formed on geological timescales by the breakdown of buried organic matter due to heat and pressure deep in the Earth's crust. Thermogenic methane reaches the atmosphere through marine and land geological gas seeps. These methane emissions are increased by human activities, for instance the exploitation and distribution of fossil fuels. Pyrogenic methane is produced by the incomplete combustion of biomass and other organic material. Peat fires, biomass burning in deforested or degraded areas, wildfires, and biofuel burning are the largest sources of pyrogenic methane. Methane hydrates, ice-like cages of trapped methane found in continental shelves and slopes and below sub-sea and land permafrost, can be of either biogenic or thermogenic origin. Each of these three process categories has both anthropogenic and natural components.

In the following, we present the different methane sources depending on their anthropogenic or natural origin, which is relevant for climate policy. Here, “natural sources” refer to pre-agricultural emissions even if they are perturbed by anthropogenic climate change, and “anthropogenic

sources” are caused by direct human activities since pre-industrial/pre-agricultural time (3000–2000 BCE; Nakazawa et al., 1993) including agriculture, waste management, and fossil-fuel-related activities. Natural emissions are split between “wetland” and “other natural” emissions (e.g., non-wetland inland waters, wild animals, termites, land geological sources, oceanic geological and biogenic sources, and terrestrial permafrost). Anthropogenic emissions contain “agriculture and waste emissions”, “fossil fuel emissions”, and “biomass and biofuel burning emissions”, assuming that all types of fires cause anthropogenic sources, although they are partly of natural origin (Fig. 6; see also Tables 3 and 6).

Our definition of natural and anthropogenic sources does not correspond exactly to the definition used by the UNFCCC following the IPCC guidelines (IPCC, 2006), where, for pragmatic reasons, all emissions from managed land are reported as anthropogenic, which is not the case here. For instance, we consider all wetlands to be natural emissions, despite some wetlands being managed and their emissions being partly reported in UNFCCC national communications. The human-induced perturbation of climate, atmospheric CO₂, and nitrogen and sulfur deposition may cause changes in the sources we classified as natural. Following our definition, emissions from wetlands, inland water, or thawing permafrost will be accountable in natural emissions, even though we acknowledge that climate change – a human perturbation – may cause increasing emissions from these sources. Methane emissions from reservoirs are considered natural even though reservoirs are human-made, and since the 2019 refinement to the IPCC guidelines (IPCC, 2006, 2019) emissions from reservoirs and other flooded lands are considered anthropogenic by the UNFCCC.

Following Saunois et al. (2016), we report anthropogenic and natural methane emissions for five main source categories for both bottom-up and top-down approaches.

Bottom-up estimates of methane emissions for some processes are derived from process-oriented models (e.g., biogeochemical models for wetlands, models for termites), inventory models (agriculture and waste emissions, fossil fuel emissions, biomass and biofuel burning emissions), satellite-based models (large scale biomass burning), or observation-based upscaling models for other sources (e.g., inland water, geological sources). From these bottom-up approaches, it is possible to provide estimates for more detailed source subcategories inside each main GCP category (see budget in Table 3). However, the total methane emission derived from the sum of independent bottom-up estimates remains unconstrained.

For atmospheric inversions (top-down approach) the situation is different. Atmospheric observations provide a constraint on the global total source and a reasonable constraint on the global sink derived from methyl chloroform (Montzka et al., 2011; Rigby et al., 2017). The inversions reported in this work solve either for a total methane flux (e.g. Pison et al., 2013) or for a limited number of source categories (e.g.

Bergamaschi et al., 2013). In most of the inverse systems the atmospheric oxidant concentrations are prescribed with pre-optimized or scaled OH fields, and thus the atmospheric sink is not solved. The assimilation of CH₄ observations alone, as reported in this synthesis, can help to separate sources with different locations or temporal variations but cannot fully separate individual sources as they often overlap in space and time in some regions. Top-down global and regional methane emissions per source category were obtained directly from gridded optimized fluxes, wherever an inversion had solved for the separate five main GCP categories. Alternatively, if an inversion only solved for total emissions (or for categories other than the main five described above), then the prior contribution of each source category at the spatial resolution of the inversion was scaled by the ratio of the total (or embedding category) optimized flux divided by the total (or embedding category) prior flux (Kirschke et al., 2013). In other words, the prior relative mix of sources at model resolution is kept while updating total emissions with atmospheric observations. The soil uptake was provided separately in order to report total gross surface emissions instead of net fluxes (sources minus soil uptake).

In summary, bottom-up models and inventories are presented for all source processes and for the five main categories defined above globally. Top-down inversions are reported globally and only for the five main emission categories.

2.5 Processing of emission maps and box-plot representation of emission budgets

Common data analysis procedures have been applied to the different bottom-up models, inventories, and atmospheric inversions whenever gridded products exist. Gridded emissions from atmospheric inversions and land surface models for wetland or biomass burning were provided at the monthly scale. Emissions from anthropogenic inventories are usually available as yearly estimates. These monthly or yearly fluxes were provided on a 1° × 1° grid or re-gridded to 1° × 1°, then converted into units of teragrams of methane per grid cell. Inversions with a resolution coarser than 1° were downscaled to 1° by each modelling group. Land fluxes in coastal pixels were reallocated to the neighbouring land pixel according to our 1° land–sea mask, and vice versa for ocean fluxes. Annual and decadal means used for this study were computed from the monthly or yearly gridded 1° × 1° maps.

Budgets are presented as box plots with quartiles (25 %, median, 75 %), outliers, and minimum and maximum values without outliers. Outliers were determined as values below the first quartile minus 3 times the inter-quartile range, or values above the third quartile plus 3 times the inter-quartile range. Mean values reported in the tables are represented as “+” symbols in the corresponding figures.

3 Methane sources and sinks: bottom-up estimates

For each source category, a short description of the relevant processes, original datasets (measurements, models), and related methodology is given. More detailed information can be found in original publication references and in the Supplement of this study.

3.1 Anthropogenic sources

3.1.1 Global inventories gathered

The main bottom-up global inventory datasets covering anthropogenic emissions from all sectors (Table 1) are from the United States Environmental Protection Agency (USEPA, 2012), the Greenhouse gas and Air pollutant Interactions and Synergies (GAINS) model developed by the International Institute for Applied Systems Analysis (IIASA) (Gomez Sanabria et al., 2018; Höglund-Isaksson, 2012, 2017), and the Emissions Database for Global Atmospheric Research (EDGARv3.2.2; Janssens-Maenhout et al., 2019) compiled by the European Commission Joint Research Centre (EC-JRC) and Netherland’s Environmental Assessment Agency (PBL). We also used the Community Emissions Data System for historical emissions (CEDS) (Hoesly et al., 2018) developed for climate modelling and the Food and Agriculture Organization (FAO) dataset emission database (Tubiello, 2019), which only covers emissions from agriculture and land use (including peatland and biomass fires).

These inventory datasets report emissions from fossil fuel production, transmission, and distribution; livestock enteric fermentation; manure management and application; rice cultivation; solid waste; and wastewater. Since the level of detail provided by country and by sector varies among inventories, the data were reconciled into common categories according to Table S2. For example, agricultural and waste burning emissions treated as a separate category in EDGAR, GAINS, and FAO are included in the biofuel sector in the USEPA inventory and in the agricultural sector in CEDS. The GAINS, EDGAR, and FAO estimates of agricultural waste burning were excluded from this analysis (these amounted to 1–3 Tg CH₄ yr⁻¹) in recent decades to prevent any inadvertent overlap with separate estimates of biomass burning emissions (e.g. GFEDv4.1s). In the inventories used here, emissions for a given region/country and a given sector are usually calculated following IPCC methodology (IPCC, 2006), as the product of an activity factor and an emission factor for this activity. An abatement coefficient is used additionally, to account for any regulations implemented to control emissions (see e.g. Höglund-Isaksson et al., 2015). These datasets differ in their assumptions and data used for the calculation; however, they are not completely independent because they follow the same IPCC guidelines (IPCC, 2006), and, at least for agriculture, use the same FAOSTAT activity data. While the USEPA inventory adopts emissions

Table 1. Bottom-up models and inventories for anthropogenic and biomass burning estimates used in this study. ^a Due to its limited sectorial breakdown this dataset was used in Table 3 for the main categories only, replacing CEDS country-based estimates. ^b Extended to 2017 for this study as described in Sect. 3.1.1.

Bottom-up models and inventories	Contribution	Time period (resolution)	Gridded	References
CEDS (country based)	Fossil fuels, agriculture and waste, biofuel	1970–2015 ^b (yearly)	no	Hoesly et al. (2018)
CEDS (gridded) ^a	Fossil fuels, agriculture and waste, biofuel	1970–2014 ^b (monthly)	0.5° × 0.5°	Hoesly et al. (2018)
EDGARv4.2.3	Fossil fuels, agriculture and waste, biofuel	1990–2012 ^b (yearly)	0.1° × 0.1°	Janssens-Maenhout et al. (2019)
IIASA GAINS ECLIP-SEv6	Fossil fuels, agriculture and waste, biofuel	1990–2015 ^b (1990–2015 yearly, > 2015 5-year interval interpolated to yearly)	0.5° × 0.5°	Höglund-Isaksson (2012)
USEPA	Fossil fuels, agriculture and waste, biofuel, biomass burning	1990–2030 (10-year interval, interpolated to yearly)	no	USEPA (2012)
FAO-CH4	Agriculture, biomass burning	1961–2016 ^b 1990–2016 (yearly)	no	Frederici et al. (2015); Tubiello et al. (2013); Tubiello (2019)
FINNv1.5	Biomass burning	2002–2018 (daily)	1 km × 1 km	Wiedinmyer et al. (2011)
GFASv1.3	Biomass burning	2003–2016 (daily)	0.1° × 0.1°	Kaiser et al. (2012)
GFEDv4.1s	Biomass burning	1997–2017 (monthly)	0.25° × 0.25°	Giglio et al. (2013)
QFEDv2.5	Biomass burning	2000–2017 (daily)	0.1° × 0.1°	Darmenov and da Silva (2015)

reported by the countries to the UNFCCC, other inventories (FAOSTAT, EDGAR, and the GAINS model) produce their own estimates using a consistent approach for all countries. These other inventories compile country-specific activity data and emission factor information or, if not available, adopt IPCC default factors (Höglund-Isaksson, 2012; Janssens-Maenhout et al., 2019; Tubiello, 2019). The CEDS takes a different approach starting from pre-existing default emission estimates; for methane, a combination of EDGAR and FAO estimates is used, scaled to match other individual or region-specific inventory values when available. This process maintains the spatial information in the default emission inventories while preserving consistency with country-level data. The FAOSTAT dataset (hereafter FAO-CH₄) was used to provide estimates of methane emissions at the country level but is limited to agriculture (enteric fermentation, manure management, rice cultivation, energy usage, burning of crop residues, and prescribed burning of savannahs) and land use (biomass burning). FAO-CH₄ uses activity data mainly from the FAOSTAT crop and livestock production database, as reported by countries to the FAO (Tubiello et al., 2013), and applies mostly the Tier 1 IPCC methodology for emissions factors (IPCC, 2006), which depend on geographic location

and development status of the country. For manure, the necessary country-scale temperature was obtained from the FAO global agroecological zone database (GAEZv3.0, 2012). Although country emissions are reported annually to the UNFCCC by Annex I countries, and episodically by non-Annex I countries, data gaps of those national inventories do not allow the inclusion of these estimates in this analysis.

In this budget, we use the following versions of these databases (see Table 1):

- EDGARv4.3.2, which provides yearly gridded emissions by sectors from 1970 to 2012 (Janssens-Maenhout et al., 2019);
- GAINS model scenario ECLIPSE v6 (Gomez Sanabria et al., 2018; Höglund-Isaksson, 2012, 2017), which provides both annual sectoral totals by country from 1990 to 2015 and a projection for 2020 (that assumes current emission legislation for the future) and an annual sectoral gridded product from 1990 to 2015;
- USEPA (USEPA, 2012), which provides 5-year sectoral totals by country from 1990 to 2020 (estimates from 2005 onward are a projection), with no gridded distribution available;

- CEDS version 2017-05-18, which provides both gridded monthly and annual country-based emissions by sectors from 1970 to 2014 (Hoesly et al., 2018);
- FAO-CH₄ (database accessed in February 2019, FAO, 2019) containing annual country-level data for the period 1961–2016, for rice, manure, and enteric fermentation and 1990–2016 for burning savannah, crop residue, and non-agricultural biomass burning.

In order to report emissions for the period 2000–2017, we extended and interpolated some of the datasets as explained in Sect. 2.2. The USEPA dataset was linearly interpolated to provide yearly values. The FAO-CH₄ dataset, ending in 2016, was extrapolated to 2017 using a linear fit based on 2014–2016 data. EDGARv4.3.2 was extrapolated to 2017 using the extended FAO-CH₄ emissions for enteric fermentation, manure management, and rice cultivation and using the BP statistical review of fossil fuel production and consumption (BP Statistical Review of World Energy, 2019) for emissions from the coal, oil, and gas sectors. In this extrapolated inventory, called EDGARv4.3.2_{EXT}, methane emissions for year t are set equal to the 2012 (last year) EDGAR emissions ($E_{\text{EDGARv4.3.2}}$) times the ratio between FAO-CH₄ emissions (or BP statistics) of year t ($E_{\text{FAO-CH}_4}(t)$) and FAO-CH₄ emissions (or BP statistics) of 2012 ($E_{\text{FAO-CH}_4}(2012)$). For each emission sector, region-specific emissions of EDGARv4.3.2_{EXT} in year t are estimated following Eq. (1):

$$E_{\text{EDGARv4.3.2ext}}(t) = E_{\text{EDGARv4.3.2}}(2012) \times E_{\text{FAO-CH}_4}(t) / E_{\text{FAO-CH}_4}(2012). \quad (1)$$

Transport, industrial, waste, and biofuel sources were linearly extrapolated in EDGARv4.3.2_{EXT} based on the last 3 years of data while other sources were kept constant at the 2012 level. To allow comparisons through 2017, the CEDS dataset has also been extrapolated in an identical method creating CEDS_{EXT}. However, in contrast to the EDGARv4.3.2 dataset, the CEDS dataset provides only a combined oil and gas sector; hence, we extended this sector using the sum of BP oil and gas emissions. The by-country GAINS dataset was linearly projected by sector for each country using the trend between the historical 2015 and projected 2020 values. These by-country projections were aggregated to the 19 global regions (Sect. 2.3 and Fig. S1) and used to extrapolate the GAINS gridded dataset in a similar manner to that described in Eq. (1). Although we only use the extended inventories, in the following the “EXT” suffix will be dropped for clarity.

3.1.2 Total anthropogenic emissions

In order to avoid double-counting and ensure consistency with each inventory, the range (min–max) and mean values of the total anthropogenic emissions were not calculated as the

sum of the mean and range of the three anthropogenic categories (“agriculture and waste”, “fossil fuels”, and “biomass burning & biofuels”). Instead, we calculated separately the total anthropogenic emissions for each inventory by adding its values for agriculture and waste, fossil fuels, and biofuels with the range of available large-scale biomass burning emissions. This approach was used for the EDGARv4.3.2, CEDS, and GAINS inventories, but we kept the USEPA inventory as originally reported because it includes its own estimates of biomass burning emissions. FAO-CH₄ was only included in the range reported for the agriculture and waste category. For the latter, we calculated the range and mean value as the sum of the mean and range of the three anthropogenic subcategory estimates “enteric fermentation and manure”, “rice”, and “landfills and waste”. The values reported for the upper-level anthropogenic categories (agriculture and waste, fossil fuels, and biomass burning & biofuels) are therefore consistent with the sum of their subcategories, although there might be small percentage differences between the reported total anthropogenic emissions and the sum of the three upper-level categories. This approach provides a more accurate representation of the range of emission estimates, avoiding an artificial expansion of the uncertainty attributable to subtle differences in the definition of sub-sector categorizations between inventories.

Based on the ensemble of databases detailed above, total anthropogenic emissions were 366 [349–393] Tg CH₄ yr^{−1} for the decade 2008–2017 (Table 3, including biomass and biofuel burning) and 334 [321–358] Tg CH₄ yr^{−1} for the decade 2000–2009. Our estimate for the preceding decade is statistically consistent with Saunois et al. (2016) (338 Tg CH₄ yr^{−1} [329–342]) and Kirschke et al. (2013) (331 Tg CH₄ yr^{−1} [304–368]) for the same period. The slightly larger range reported herein with respect to previous estimates is mainly due to a larger range in the biomass burning estimates, as more biomass burning products are included in this update. The range associated with our estimates (~ 10 %–12 %) is smaller than the range reported in Höglund-Isaksson et al. (2015) (~ 20 %), perhaps because they analysed data from a wider range of inventories and projections, plus this study was referenced to one year only (2005) rather than averaged over a decade, as done here.

Figure 2a summarizes global methane emissions of anthropogenic sources (including biomass and biofuel burning) by different datasets between 2000 and 2050. The datasets consistently estimate total anthropogenic emissions of ~ 300 Tg CH₄ yr^{−1} in 2000. The main discrepancy between the inventories is their trend after 2005, with the lowest emissions projected by GAINS and the largest by CEDS. With the U.S. EPA being a projection from 2005 onward, its values and trends deviate from others. For the Sixth Assessment report of the IPCC, seven main Shared Socioeconomic Pathways (SSPs) were defined for future climate projections in the Coupled Model Intercomparison Project Phase 6 (CMIP6) (Gidden et al., 2019; O’Neill et

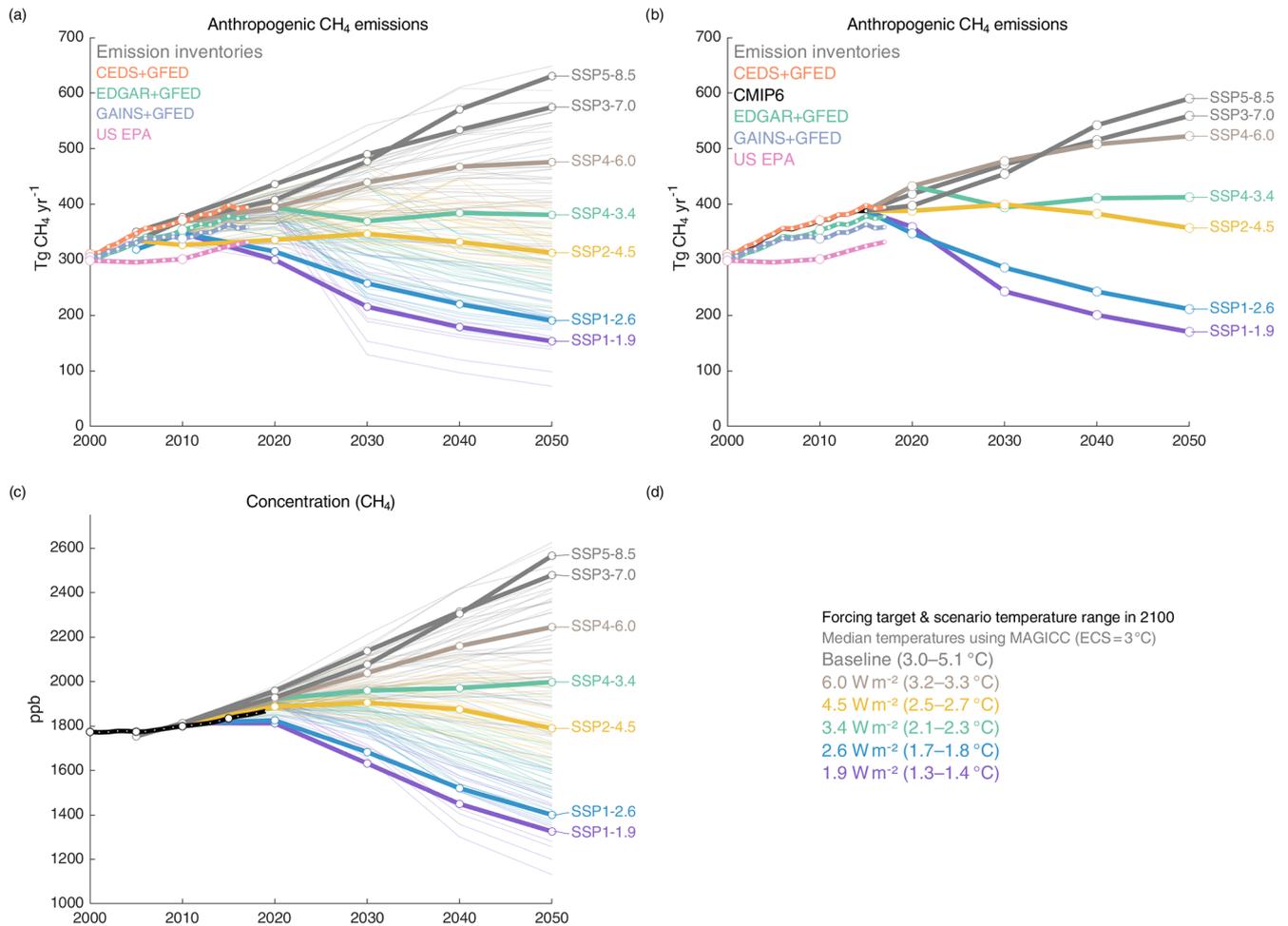


Figure 2. (a, b) Global anthropogenic methane emissions (including biomass burning) from historical inventories and future projections ($\text{Tg CH}_4 \text{ yr}^{-1}$). (a) Inventories and the unharmonized Shared Socioeconomic Pathways (Riahi et al., 2017), with highlighted scenarios representing scenarios assessed in CMIP6 (O’Neill, et al., 2016). (b) The selected scenarios harmonized with historical emissions (CEDS) for CMIP6 activities (Gidden et al., 2019). USEPA and GAINS estimates have been linearly interpolated from the 5-year original products to yearly values. After 2005, USEPA original estimates are projections. (c) Global methane concentrations for NOAA surface site observations (black) and projections based on SSPs (Riahi et al., 2017) with concentrations estimated using MAGICC (Meinshausen et al., 2011).

al., 2016) ranging from 1.9 to 8.5 W m^{-2} radiative forcing by the year 2100 (as shown by the number in the SSP names). The trends in methane emissions from 2010 estimated by current inventories track the pathways with the highest radiative forcing in 2100 (based on the unharmonized scenarios developed by integrated assessment models, Fig. 2a). For the 1970–2015 period, historical emissions used in CMIP6 (Feng et al., 2019) combine anthropogenic emissions from CEDS (Hoesly et al., 2018) and a climatological value from the GFEDv4.1s biomass burning inventory (van Marle et al., 2017). The CEDS anthropogenic emissions estimates, based on EDGARv4.2, are 10–20 Tg higher than the more recent EDGARv4.3.2 (van Marle et al., 2017). Harmonized scenarios used for CMIP6 activities start in 2015 at 388 $\text{Tg CH}_4 \text{ yr}^{-1}$. Since methane emissions continue to track scenarios that assume no or minimal climate policies,

it may indicate that climate policies, when present, have not yet produced sufficient results to change the emissions trajectory substantially (Nisbet et al., 2019). After 2015, the SSPs span a range of possible outcomes, but current emissions appear likely to follow the higher-emission trajectories over the next decade (Fig. 2b). This illustrates the challenge of methane mitigation that lies ahead to help reach the goals of the Paris Agreement. In addition, estimates of methane atmospheric concentrations from the unharmonized scenarios (Riahi et al., 2017) indicate that observations of global methane concentrations fall well within the range of scenarios (Fig. 2c). The methane concentrations are estimated using a simple exponential decay with inferred natural emissions (Meinshausen et al., 2011), and the emergence of any trend between observations and scenarios needs to be confirmed in the following years. In the future, it will be important to

monitor the trends from the year 2015 (the Paris Agreement) estimated in inventories and from atmospheric observations and compare them to various scenarios.

3.1.3 Fossil fuel production and use

Most anthropogenic methane emissions related to fossil fuels come from the exploitation, transportation, and usage of coal, oil, and natural gas. Additional emissions reported in this category include small industrial contributions such as production of chemicals and metals, fossil fuel fires (e.g., underground coal mine fires and the Kuwait oil and gas fires), and transport (road and non-road transport). Methane emissions from the oil industry (e.g. refining) and production of charcoal are estimated to be a few teragrams of methane per year only and are included in the transformation industry sector in the inventory. Fossil fuel fires are included in the subcategory “oil & gas”. Emissions from industries and road and non-road transport are reported apart from the two main subcategories oil & gas and “coal mining”, contrary to Saunio et al. (2016); each of these amounts to about 5 Tg CH₄ yr⁻¹ (Table 3). The large range (0–12 Tg CH₄ yr⁻¹) is attributable to difficulties in allocating some sectors to these sub-sectors consistently among the different inventories (see Table S2). The spatial distribution of methane emissions from fossil fuels is presented in Fig. 3 based on the mean gridded maps provided by CEDS, EDGARv4.3.2, and GAINS for the 2008–2017 decade; the USEPA lacks a gridded product.

Global mean emissions from fossil-fuel-related activities, other industries, and transport are estimated from the four global inventories (Table 1) to be of 128 [113–154] Tg CH₄ yr⁻¹ for the 2008–2017 decade (Table 3), but with large differences in the rate of change during this period across inventories. The sector accounts on average for 35 % (range 30 %–42 %) of total global anthropogenic emissions.

Coal mining

During mining, methane is emitted primarily from ventilation shafts, where large volumes of air are pumped into the mine to keep the CH₄ mixing ratio below 0.5 % to avoid accidental ignition, and from dewatering operations. In countries of the Organization for Economic Co-operation and Development (OECD), methane released from ventilation shafts is in principle used as fuel, but in many countries, it is still emitted into the atmosphere or flared, despite the efforts for coal mine recovery under the UNFCCC Clean Development Mechanisms (<http://cdm.unfccc.int>, last access: 29 June 2020). Methane also leaks occur during post-mining handling, processing, and transportation. Some CH₄ is released from coal waste piles and abandoned mines; while emissions from these sources were believed to be low (IPCC, 2000), recent work has estimated these to be 22 billion m³ (compared with 103 billion m³ from functioning coal mines)

in 2010 with emissions projected to increase into the future (Kholod et al., 2020).

In 2017, almost 40 % (IEA, 2019b) of the world’s electricity was still produced from coal. This contribution grew in the 2000s at the rate of several per cent per year, driven by Asian economic growth where large reserves exist, but global coal consumption has declined since 2014. In 2018, the top 10 largest coal producing nations accounted for ~90 % of total world methane emissions for coal mining; among them, the top three producers (China, United States, and India) produced almost two-thirds (64 %) of the world’s coal (IEA, 2019a).

Global estimates of CH₄ emissions from coal mining show a large range of 29–61 Tg CH₄ yr⁻¹ for 2008–2017, in part due to the lack of comprehensive data from all major producing countries. The highest value of the range comes from the CEDS inventory while the lowest comes from the USEPA. CEDS seems to have overestimated coal mining emissions from China by almost a factor of 2, most likely due to its dependence on the EDGARv4.2 emission inventory. As highlighted by Saunio et al. (2016), a county-based inventory of Chinese methane emissions also confirms the overestimate of about +38 % with total anthropogenic emissions estimated at 43 ± 6 Tg CH₄ yr⁻¹ (Peng et al., 2016). The EDGARv4.2 inventory follows the IPCC guidelines and uses a European averaged emission factor for CH₄ from coal production to substitute missing data for China, which appear to be overestimated by a factor of approximately 2. These differences highlight significant errors resulting from the use of emission factors, and applying Tier 1 approaches for coal mine emissions is not sufficiently accurate as stated by the IPCC guidelines. The newly released version of EDGARv4.3.2 used here has revised China coal methane emission factors downwards and distributed them to more than 80 times more coal mining locations in China. Coal mining emission factors depend strongly on the type of coal extraction (underground mining emits up to 10 times more than surface mining), geological underground structure (region-specific), history (basin uplift), and quality of the coal (brown coal emits more than hard coal). Finally, coal mining is the main source explaining the differences between inventories globally (Fig. 2).

For the 2008–2017 decade, methane emissions from coal mining represent 33 % of total fossil-fuel-related emissions of methane (42 Tg CH₄ yr⁻¹, range of 29–61). An additional very small source corresponds to fossil fuel fires (mostly underground coal fires, ~0.15 Tg yr⁻¹ in 2012, EDGARv4.3.2).

Oil and natural gas systems

This subcategory includes emissions from both conventional and shale oil and gas exploitation. Natural gas is comprised primarily of methane, so both fugitive and planned emissions during the drilling of wells in gas fields, extraction, transportation, storage, gas distribution, end use, and incomplete

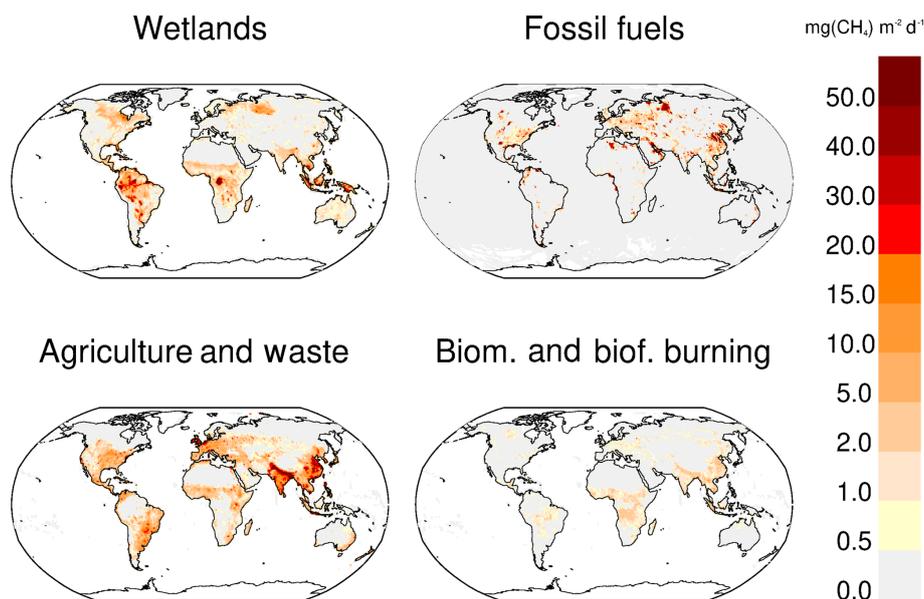


Figure 3. Methane emissions from four source categories: natural wetlands (excluding lakes, ponds, and rivers), biomass and biofuel burning, agriculture and waste, and fossil fuels for the 2008–2017 decade ($\text{mg CH}_4 \text{ m}^{-2} \text{ d}^{-1}$). The wetland emission map represents the mean daily emission average over the 13 biogeochemical models listed in Table 2 and over the 2008–2017 decade. Fossil fuel and agriculture and waste emission maps are derived from the mean estimates of gridded CEDS, EDGARv4.3.2, and GAINS models. The biomass and biofuel burning map results from the mean of the biomass burning inventories listed in Table 1 added to the mean of the biofuel estimate from CEDS, EDGARv4.3.2, and GAINS models.

combustion of gas flares emit methane (Lamb et al., 2015; Shorter et al., 1996). Persistent fugitive emissions (e.g., due to leaky valves and compressors) should be distinguished from intermittent emissions due to maintenance (e.g. purging and draining of pipes). During transportation, fugitive emissions can occur in oil tankers, fuel trucks, and gas transmission pipelines, attributable to corrosion, manufacturing, and welding faults. According to Lelieveld et al. (2005), CH_4 fugitive emissions from gas pipelines should be relatively low; however distribution networks in older cities may have higher rates, especially those with cast-iron and unprotected steel pipelines (Phillips et al., 2013). Measurement campaigns in cities within the United States and Europe revealed that significant emissions occur in specific locations (e.g. storage facilities, city gates, well and pipeline pressurization–depressurization points) along the distribution networks (e.g. Jackson et al., 2014a; McKain et al., 2015; Wunch et al., 2016). However, methane emissions vary significantly from one city to another depending, in part, on the age of city infrastructure and the quality of its maintenance, making urban emissions difficult to scale up. In many facilities, such as gas and oil fields, refineries, and offshore platforms, venting of natural gas is now replaced by flaring with almost complete conversion to CO_2 ; these two processes are usually considered together in inventories of oil and gas industries. Also, single-point failure of natural gas infrastructure can leak methane at a high rate for months, such as at the Aliso Canyon blowout in the Los Angeles,

CA, basin (Conley et al., 2016) or the recent shale gas well blowout in Ohio (Pandey et al., 2019), thus hampering emission control strategies. Production of natural gas from the exploitation of hitherto unproductive rock formations, especially shale, began in the 1970s in the United States on an experimental or small-scale basis, and then, from the early 2000s, exploitation started at large commercial scale. The shale gas contribution to total dry natural gas production in the United States reached 62% in 2017, growing rapidly from 40% in 2012, with only small volumes produced before 2005 (EIA, 2019). The possibly larger emission factors from the shale gas compared to the conventional ones have been widely debated (e.g. Cathles et al., 2012; Howarth, 2019; Lewan, 2020). However, the latest studies tend to infer similar emission factors in a narrow range of 1%–3% (Alvarez et al., 2018; Peischl et al., 2015; Zavala-Araiza et al., 2015), different from the widely spread rates of 3%–17% from previous studies (e.g. Caulton et al., 2014; Schneising et al., 2014).

Methane emissions from oil and natural gas systems vary greatly in different global inventories (72 to 97 Tg yr^{-1} in 2017, Table 3). The inventories generally rely on the same sources and magnitudes for activity data, with the derived differences therefore resulting primarily from different methodologies and parameters used, including emission factors. Those factors are country- or even site-specific, and the few field measurements available often combine oil and gas activities (Brandt et al., 2014) and remain largely unknown

for most major oil- and gas-producing countries. Depending on the country, the reported emission factors may vary by 2 orders of magnitude for oil production and by 1 order of magnitude for gas production (Table S5.1 of Höglund-Isaksson, 2017). The GAINS estimate of methane emissions from oil production, for instance, is twice as high as EDGARv4.3.2. For natural gas, the uncertainty is of a similar order of magnitude. During oil extraction, natural gas generated can be either recovered (re-injected or utilized as an energy source) or not recovered (flared or vented to the atmosphere). The recovery rates vary from one country to another (being much higher in the United States, Europe, and Canada than elsewhere) and from one type of oil to another: flaring is less common for heavy oil wells than for conventional ones (Höglund-Isaksson et al., 2015). Considering recovery rates could lead to 2-times-higher methane emissions accounting for country-specific rates of generation and recovery of associated gas than when using default values (Höglund-Isaksson, 2012). This difference in methodology explains, in part, why GAINS estimates are higher than those of EDGARv4.3.2.

Most studies (Alvarez et al., 2018; Brandt et al., 2014; Jackson et al., 2014b; Karion et al., 2013; Moore et al., 2014; Olivier and Janssens-Maenhout, 2014; Pétron et al., 2014; Zavala-Araiza et al., 2015), albeit not all (Allen et al., 2013; Cathles et al., 2012; Peischl et al., 2015), suggest that methane emissions from oil and gas industry are underestimated by inventories and agencies, including the USEPA. Zavala-Araiza et al. (2015) showed that a few high-emitting facilities, i.e., super-emitters, neglected in the inventories, dominated US emissions. These high-emitting points, located on the conventional part of the facility, could be avoided through better operating conditions and repair of malfunctions. As US production increases, absolute methane emissions almost certainly increase. US crude oil production also doubled over the last decade and natural gas production rose more than 50 % (EIA, 2019). However, global implications of the rapidly growing shale gas activity in the United States remain to be determined precisely.

For the 2008–2017 decade, methane emissions from upstream and downstream oil and natural gas sectors are estimated to represent about 63 % of total fossil CH₄ emissions (80 Tg CH₄ yr⁻¹, range of 68–92 Tg CH₄ yr⁻¹, Table 3), with a lower uncertainty range than for coal emissions for most countries.

3.1.4 Agriculture and waste sectors

This main category includes methane emissions related to livestock production (i.e., enteric fermentation in ruminant animals and manure management), rice cultivation, landfills, and wastewater handling. Of these, globally and in most countries, livestock is by far the largest source of CH₄, followed by waste handling and rice cultivation. Conversely, field burning of agricultural residues is a minor source of

CH₄ reported in emission inventories. The spatial distribution of methane emissions from agriculture and waste handling is presented in Fig. 3 based on the mean gridded maps provided by CEDS, EDGARv4.3.2, and GAINS over the 2008–2017 decade.

Global emissions from agriculture and waste for the period 2008–2017 are estimated to be 206 Tg CH₄ yr⁻¹ (range 191–223, Table 3), representing 56 % of total anthropogenic emissions.

Livestock: enteric fermentation and manure management

Domestic ruminants such as cattle, buffalo, sheep, goats, and camels emit methane as a by-product of the anaerobic microbial activity in their digestive systems (Johnson et al., 2002). The very stable temperatures (about 39 °C) and pH (6.5–6.8) values within the rumen of domestic ruminants, along with a constant plant matter flow from grazing (cattle graze many hours per day), allow methanogenic *Archaea* residing within the rumen to produce methane. Methane is released from the rumen mainly through the mouth of multi-stomached ruminants (eructation, ~ 87 % of emissions) or absorbed in the blood system. The methane produced in the intestines and partially transmitted through the rectum is only ~ 13 %.

The total number of livestock continues to grow steadily. There are currently (2017) about 1.5 billion cattle globally, 1 billion sheep, and nearly as many goats (<http://www.fao.org/faostat/en/#data/GE>, last access: 29 June 2020). Livestock numbers are linearly related to CH₄ emissions in inventories using the Tier 1 IPCC approach such as FAOSTAT. In practice, some non-linearity may arise due to dependencies of emissions on total weight of the animals and their diet, which are better captured by Tier 2 and higher approaches. Cattle, due to their large population, large individual size, and particular digestive characteristics, account for the majority of enteric fermentation CH₄ emissions from livestock worldwide (Tubiello, 2019), particularly in intensive agricultural systems in wealthier and emerging economies, including the United States (USEPA, 2016). Methane emissions from enteric fermentation also vary from one country to another as cattle may experience diverse living conditions that vary spatially and temporally, especially in the tropics (Chang et al., 2019).

Anaerobic conditions often characterize manure decomposition in a variety of manure management systems globally (e.g., liquid/slurry treated in lagoons, ponds, tanks, or pits), with the volatile solids in manure producing CH₄. In contrast, when manure is handled as a solid (e.g., in stacks or dry lots) or deposited on pasture, range, or paddock lands, it tends to decompose aerobically and to produce little or no CH₄. However aerobic decomposition of manure tends to produce nitrous oxide (N₂O), which has a larger warming impact than CH₄. Ambient temperature, moisture, energy contents of the feed, manure composition, and manure storage or residency time affect the amount of CH₄ produced. Despite these com-

plexities, most global datasets used herein apply a simplified IPCC Tier 1 approach, where amounts of manure treated depend on animal numbers and simplified climatic conditions by country.

Global methane emissions from enteric fermentation and manure management are estimated in the range of 99–115 Tg CH₄ yr⁻¹, for the year 2010, in the GAINS model and CEDS, USEPA, FAO-CH₄, and EDGARv4.3.2 inventories. These values are slightly higher than the IPCC Tier 2 estimate of Dangal et al. (2017) (95.7 Tg CH₄ yr⁻¹ for 2010) and the IPCC Tier 3 estimates of Herrero et al. (2013) (83.2 Tg CH₄ yr⁻¹ for 2000), but in agreement with the recent IPCC Tier 2 estimate of Chang et al. (2019) (99 ± 12 Tg CH₄ yr⁻¹ for 2012).

For the period 2008–2017, we estimated total emissions of 111 [106–116] Tg CH₄ yr⁻¹ for enteric fermentation and manure management, about one-third of total global anthropogenic emissions.

Rice cultivation

Most of the world's rice is grown in flooded paddy fields (Baicich, 2013). The water management systems, particularly flooding, used to cultivate rice are one of the most important factors influencing CH₄ emissions and one of the most promising approaches for CH₄ emission mitigation: periodic drainage and aeration not only cause existing soil CH₄ to oxidize, but also inhibit further CH₄ production in soils (Simpson et al., 1995; USEPA, 2016; Zhang, 2016). Upland rice fields are not typically flooded and therefore are not a significant source of CH₄. Other factors that influence CH₄ emissions from flooded rice fields include fertilization practices (i.e. the use of urea and organic fertilizers), soil temperature, soil type (texture and aggregated size), rice variety, and cultivation practices (e.g., tillage, seeding, and weeding practices) (Conrad et al., 2000; Kai et al., 2011; USEPA, 2011; Yan et al., 2009). For instance, methane emissions from rice paddies increase with organic amendments (Cai et al., 1997) but can be mitigated by applying other types of fertilizers (mineral, composts, biogas residues) or using wet seeding (Wassmann et al., 2000).

The geographical distribution of rice emissions has been assessed by global (e.g. Janssens-Maenhout et al., 2019; Tubiello, 2019; USEPA, 2012) and regional (e.g. Castelán-Ortega et al., 2014; Chen et al., 2013; Chen and Prinn, 2006; Peng et al., 2016; Yan et al., 2009; Zhang and Chen, 2014) inventories or land surface models (Li et al., 2005; Pathak et al., 2005; Ren et al., 2011; Spahni et al., 2011; Tian et al., 2010, 2011; Zhang, 2016). The emissions show a seasonal cycle, peaking in the summer months in the extra-tropics associated with monsoons and land management. Similar to emissions from livestock, emissions from rice paddies are influenced not only by extent of rice field area (analogous to livestock numbers), but also by changes in the productivity of plants (Jiang et al., 2017) as these alter the CH₄ emis-

sion factor used in inventories. Nonetheless, the inventories considered herein are largely based on IPCC Tier 1 methods, which largely scale with cultivated areas but include region-specific emission factors.

The largest emissions from rice cultivation are found in Asia, accounting for 30 % to 50 % of global emissions (Fig. 3). The decrease in CH₄ emissions from rice cultivation over recent decades is confirmed in most inventories, because of the decrease in rice cultivation area, changes in agricultural practices, and a northward shift of rice cultivation since the 1970s, as in China (e.g. Chen et al., 2013).

Based on the global inventories considered in this study, global methane emissions from rice paddies are estimated to be 30 [25–38] Tg CH₄ yr⁻¹ for the 2008–2017 decade (Table 3), or about 8 % of total global anthropogenic emissions of methane. These estimates are consistent with the 29 Tg CH₄ yr⁻¹ estimated for the year 2000 by Carlson et al. (2017).

Waste management

This sector includes emissions from managed and non-managed landfills (solid waste disposal on land), and wastewater handling, where all kinds of waste are deposited. Methane production from waste depends on the pH, moisture, and temperature of the material. The optimum pH for methane emission is between 6.8 and 7.4 (Thorneloe et al., 2000). The development of carboxylic acids leads to low pH, which limits methane emissions. Food or organic waste, leaves, and grass clippings ferment quite easily, while wood and wood products generally ferment slowly, and cellulose and lignin even more slowly (USEPA, 2010a).

Waste management was responsible for about 11 % of total global anthropogenic methane emissions in 2000 (Kirschke et al., 2013). A recent assessment of methane emissions in the United States found landfills to account for almost 26 % of total US anthropogenic methane emissions in 2014, the largest contribution of any single CH₄ source in the United States (USEPA, 2016). In Europe, gas control has been mandatory on all landfills since 2009, following the ambitious objective raised in the EU Landfill Directive (1999) to reduce landfilling of biodegradable waste to 65 % below the 1990 level by 2016. This mitigation is attempted through source separation and treatment of separated biodegradable waste in composts, bio-digesters, and paper recycling.

Wastewater from domestic and industrial sources is treated in municipal sewage treatment facilities and private effluent treatment plants. The principal factor in determining the CH₄ generation potential of wastewater is the amount of degradable organic material in the wastewater. Wastewater with high organic content is treated anaerobically, which leads to increased emissions (André et al., 2014). Excessive and rapid urban development worldwide, especially in Asia and Africa, could enhance methane emissions from waste unless

adequate mitigation policies are designed and implemented rapidly.

The GAINS model and CEDS and EDGAR inventories give robust emission estimates from solid waste in the range of 29–41 Tg CH₄ yr⁻¹ for the year 2005 and more uncertain wastewater emissions in the range 14–33 Tg CH₄ yr⁻¹.

In our study, the global emission of methane from waste management is estimated in the range of 60–69 Tg CH₄ yr⁻¹ for the 2008–2017 period with a mean value of 65 Tg CH₄ yr⁻¹, about 12 % of total global anthropogenic emissions.

3.1.5 Biomass and biofuel burning

This category includes methane emissions from biomass burning in forests, savannahs, grasslands, peats, agricultural residues, and the burning of biofuels in the residential sector (stoves, boilers, fireplaces). Biomass and biofuel burning emits methane under incomplete combustion conditions (i.e., when oxygen availability is insufficient for complete combustion), for example in charcoal manufacturing and smouldering fires. The amount of methane emitted during the burning of biomass depends primarily on the amount of biomass, burning conditions, and the specific material burned.

In this study, we use large-scale biomass burning (forest, savannah, grassland, and peat fires) from five biomass burning inventories (described below) and the biofuel burning contribution from anthropogenic emission inventories (EDGARv4.3.2, CEDS, GAINS, and USEPA). The spatial distribution of emissions from the burning of biomass and biofuel over the 2008–2017 decade is presented in Fig. 3 based on data listed in Table 1.

At the global scale, during the period of 2008–2017, biomass and biofuel burning generated methane emissions of 30 [26–40] Tg CH₄ yr⁻¹ (Table 3), of which 30 %–50 % is from biofuel burning.

Biomass burning

Fire is an important disturbance event in terrestrial ecosystems globally (van der Werf et al., 2010) and can be of either natural (typically ~ 10 % of fires, ignited by lightning strikes or started accidentally) or anthropogenic origin (~ 90 %, human-initiated fires) (USEPA, 2010b, chap. 9.1). Anthropogenic fires are concentrated in the tropics and subtropics, where forests, savannahs, and grasslands may be burned to clear land for agricultural purposes or to maintain pastures and rangelands. Small fires associated with agricultural activity, such as field burning and agricultural waste burning, are often not well detected by remote sensing methods and are instead estimated based on cultivated area.

Emission rates of biomass burning vary with biomass loading (depending on the biomes) at the location of the fire, the efficiency of the fire (depending on the vegetation type), the fire type (smoldering or flaming) and emission factor

(mass of the considered species / mass of biomass burned). Depending on the approach, these parameters can be derived using satellite data and/or a biogeochemical model, or through simpler IPCC default approaches.

In this study, we use five products to estimate biomass burning emissions. The Global Fire Emission Database (GFED) is the most widely used global biomass burning emission dataset and provides estimates from 1997. Here, we use GFEDv4.1s (van der Werf et al., 2017), based on the Carnegie–Ames–Stanford approach (CASA) biogeochemical model and satellite-derived estimates of burned area (from the MODerate resolution Imaging Sensor, MODIS), fire activity, and plant productivity. GFEDv4.1s (with small fires) is available at a 0.25° resolution and on a daily basis from 1997 to 2017. One characteristic of the GFEDv4.1s burned area is that small fires are better accounted for compared to GFEDv4.1 (Randerson et al., 2012), increasing carbon emissions by approximately 35 % at the global scale.

The Quick Fire Emissions Dataset (QFED) is calculated using the fire radiative power (FRP) approach, in which the thermal energy emitted by active fires (detected by MODIS) is converted to an estimate of methane flux using biome-specific emissions factors and a unique method of accounting for cloud cover. Further information related to this method and the derivation of the biome specific emission factors can be found in Darmanov and da Silva (Darmanov and da Silva, 2015). Here we use the historical QFEDv2.5 product available daily on a 0.1 × 0.1 grid for 2000 to 2017.

The Fire Inventory from NCAR (FINN; Wiedinmyer et al., 2011) provides daily, 1 km resolution estimates of gas and particle emissions from open burning of biomass (including wildfire, agricultural fires, and prescribed burning) over the globe for the period 2002–2018. FINNv1.5 uses MODIS satellite observations for active fires, land cover, and vegetation density.

We use v1.3 of the Global Fire Assimilation System (GFAS; Kaiser et al., 2012), which calculates emissions of biomass burning by assimilating fire radiative power (FRP) observations from MODIS at a daily frequency and 0.5° resolution and is available for 2000–2016.

The FAO-CH₄ yearly biomass burning emissions are based on the most recent MODIS 6 burned-area products, coupled with a pixel-level (500 m) implementation of the IPCC Tier 1 approach, and are available from 1990 to 2016 (Table 1).

The differences in emission estimates for biomass burning arise from specific geographical and meteorological conditions and fuel composition, which strongly impact combustion completeness and emission factors. The latter vary greatly according to fire type, ranging from 2.2 g CH₄ kg⁻¹ dry matter burned for savannah and grassland fires up to 21 g CH₄ kg⁻¹ dry matter burned for peat fires (van der Werf et al., 2010).

In this study, based on the five aforementioned products, biomass burning emissions are estimated at 17 Tg CH₄ yr⁻¹

[14–26] for 2008–2017, representing about 5 % of total global anthropogenic methane emissions.

Biofuel burning

Biomass that is used to produce energy for domestic, industrial, commercial, or transportation purposes is hereafter called biofuel burning. A largely dominant fraction of methane emissions from biofuels comes from domestic cooking or heating in stoves, boilers, and fireplaces, mostly in open cooking fires where wood, charcoal, agricultural residues, or animal dung are burned. It is estimated that more than 2 billion people, mostly in developing countries, use solid biofuels to cook and heat their homes daily (André et al., 2014), and yet methane emissions from biofuel combustion have received relatively little attention. Biofuel burning estimates are gathered from the CEDS, USEPA, GAINS, and EDGAR inventories. Due to the sectoral breakdown of the EDGAR and CEDS inventories, the biofuel component of the budget has been estimated as equivalent to the “RCO – Energy for buildings” sector as defined in Worden et al. (2017) and Hoesly et al. (2018) (see Table S2). This is equivalent to the sum of the IPCC 1A4a_Commercial-institutional, 1A4b_Residential, 1A4c_Agriculture-forestry-fishing, and 1A5_Other-unspecified reporting categories. This definition is consistent with that used in Saunois et al. (2016) and Kirschke et al. (2013). While this sector incorporates biofuel use, it also includes the use of other combustible materials (e.g. coal or gas) for small-scale heat and electricity generation within residential and commercial premises. Data provided by the GAINS inventory suggest that this approach may overestimate biofuel emissions by between 5 % and 50 %.

In our study, biofuel burning is estimated to contribute $12 \text{ Tg CH}_4 \text{ yr}^{-1}$ [10–14] to the global methane budget, about 3 % of total global anthropogenic methane emissions for 2008–2017.

3.1.6 Other anthropogenic sources (not explicitly included in this study)

Other anthropogenic sources not included in this study are related to agriculture and land use management. In particular, increases in global palm oil production have led to the clearing of natural peat forests, reducing natural peatland area and associated natural CH_4 emissions. While studies have long suggested that CH_4 emissions from peatland drainage ditches are likely to be significant (e.g. Minkinen and Laine, 2006), CH_4 emissions related to palm oil plantations have yet to be properly quantified. Taylor et al. (2014) have quantified global palm oil wastewater treatment fluxes to be $4 \pm 32 \text{ Tg CH}_4 \text{ yr}^{-1}$ for 2010–2013. This currently represents a small and highly uncertain source of methane but one potentially growing in the future.

3.2 Natural sources

Natural methane sources include vegetated wetland emissions and inland water systems (lakes, small ponds, rivers), land geological sources (gas–oil seeps, mud volcanoes, microseepage, geothermal manifestations, and volcanoes), wild animals, termites, thawing terrestrial and marine permafrost, and oceanic sources (biogenic, geological, and hydrate). In water-saturated or flooded ecosystems, the decomposition of organic matter gradually depletes most of the oxygen in the soil, resulting in anaerobic conditions and methane production. Once produced, methane can reach the atmosphere through a combination of three processes: (1) diffusive loss of dissolved CH_4 across the air–water boundary; (2) ebullition flux from sediments, and (3) flux mediated by emergent aquatic macrophytes and terrestrial plants (plant transport). On its way to the atmosphere, in the soil or water columns, methane can be partly or completely oxidized by a group of bacteria called methanotrophs, which use methane as their only source of energy and carbon (USEPA, 2010b). Concurrently, methane from the atmosphere can diffuse into the soil column and be oxidized (see Sect. 3.3.4 on soil uptake).

3.2.1 Wetlands

Wetlands are generally defined as ecosystems in which soils or peats are water saturated or where surface inundation (permanent or not) dominates the soil biogeochemistry and determines the ecosystem species composition (USEPA, 2010b). In order to refine such overly broad definition for methane emissions, we define wetlands as ecosystems with inundated or saturated soils or peats where anaerobic conditions lead to methane production (Matthews and Fung, 1987; USEPA, 2010b). Brackish water emissions are discussed separately in Sect. 3.2.6. Our definition of wetlands includes peatlands (bogs and fens), mineral soil wetlands (swamps and marshes), and seasonal or permanent floodplains. It excludes exposed water surfaces without emergent macrophytes, such as lakes, rivers, estuaries, ponds, and reservoirs (addressed in the next section), as well as rice agriculture (see Sect. 3.1.4, rice cultivation paragraph) and wastewater ponds. It also excludes coastal vegetated ecosystems (mangroves, seagrasses, salt marshes) with salinities usually > 0.5 psu (see Sect. 3.2.6). Even with this definition, some wetlands could be considered anthropogenic systems, being affected by human land use changes such as impoundments, drainage, or restoration (Woodward et al., 2012). In the following we retain the generic denomination “wetlands” for natural and human-influenced wetlands, as discussed in Sect. 2.2.

The three most important factors influencing methane production in wetlands are the spatial and temporal extent of anoxia (linked to water saturation), temperature, and substrate availability (Valentine et al., 1994; Wania et al., 2010; Whalen, 2005).

Land surface models estimate CH₄ emissions through a series of processes, including CH₄ production, oxidation, and transport. The models are then forced with inputs accounting for changing environmental factors (Melton et al., 2013; Poulter et al., 2017; Tian et al., 2010; Wania et al., 2013; Xu et al., 2010). Methane emissions from wetlands are computed as the product of an emission flux density and a methane-producing area or surface extent (see Supplement; Bohn et al., 2015; Melton et al., 2013). Wetland extent appears to be a primary contributor to uncertainties in the absolute flux of methane emissions from wetlands, with meteorological response the main source of uncertainty for seasonal and inter-annual variability (Bohn et al., 2015; Desai et al., 2015; Poulter et al., 2017).

In this work, 13 land surface models computing net CH₄ emissions (Table 2) were run under a common protocol with a 30-year spin-up (1901–1930) followed by a simulation through the end of 2017 forced by CRU-JRA reconstructed climate fields (Harris, 2019). Of the 13 models, 10 previously contributed to Saunois et al. (2016), three models were new to this release (JSBACH, LPJ-GUESS, and TEM-MDM) (Table S3). Atmospheric CO₂ influencing wetland net primary production (NPP) was also prescribed in the models. In all models, the same remote-sensing-based wetland area and dynamics dataset called WAD2M (Wetland Area Dynamics for Methane Modeling) was prescribed. WAD2M provides year-to-year varying monthly global wetland areas over 2000–2017, partly addressing known issues, such as separation between wetlands and other inland waters (Poulter et al., 2017). WAD2M combines microwave remote sensing data from Schroeder et al. (2015) with various regional inventory datasets to develop a monthly global wetland area dataset, which will be further presented in the near future by Zhang et al. (2020). Non-vegetated wetland inland waters (i.e., lakes, rivers, and ponds) were subtracted using the Global Surface Water dataset of Pekel et al. (2016), assuming that permanent waters were those that were present > 50 % of the time within a 32-year observing period. Then, wetland inventories for the tropics (Gumbrecht et al., 2017), high latitudes (Hugelius et al., 2014; Widhalm et al., 2015), and temperate regions (Lehner and Döll, 2004) were used to set the long-term annual mean wetland area, to which a seasonal cycle of fractional surface water was added using data from the Surface Water Microwave Product Series version 3.2 (SWAMPS) (Jensen and McDonald, 2019; Schroeder et al., 2015). Rice agriculture was removed using the MIRCA2000 dataset from circa 2000, as a fixed distribution. The combined remote sensing and inventory WAD2M product leads to a maximum wetland area of 14.9 M km² during the peak season (8.4 M km² on annual average, with a range of 8.0 to 8.9 M km² from 2000 to 2017, about 5.5 % of the global land surface). The largest wetland areas in WAD2M are in Amazonia, the Congo Basin, and the Western Siberian Lowlands, which in previous studies were underestimated by inventories (Bohn et al., 2015).

The average emission map from wetlands for 2008–2017 built from the 13 models is plotted in Fig. 3. The zones with the largest emissions are the Amazon basin, equatorial Africa and Asia, Canada, western Siberia, eastern India, and Bangladesh. Regions where methane emissions are robustly inferred (defined as regions where mean flux is larger than the standard deviation of the models) represent 61 % of the total methane flux due to natural wetlands. This contribution is 80 % lower than found in Saunois et al. (2016) probably due to the different ensemble of models gathered here and the more stringent exclusion of inland waters. The main primary emission zones are consistent between models, which is clearly favoured by the prescribed common wetland extent. However, the different sensitivities of the models to temperature, vapour pressure, precipitation, and radiation can generate substantially different patterns, such as in India. Some secondary (in magnitude) emission zones are also consistently inferred between models: Scandinavia, continental Europe, eastern Siberia, central United States, and tropical Africa.

The resulting global flux range for natural wetland emissions is 101–179 Tg CH₄ yr⁻¹ for the 2000–2017 period, with an average of 148 Tg CH₄ yr⁻¹ and a one-sigma standard deviation of 25 Tg CH₄ yr⁻¹. For the last decade, 2008–2017, the average ensemble emissions were 149 Tg CH₄ yr⁻¹ with a range of 102–182 (Table 3). Using a prognostic set of simulations, where models used their own internal approach to estimate wetland area and dynamics, the average ensemble emissions were 161 Tg CH₄ yr⁻¹ with a range of 125–218 for the 2008–2017 period. The greater range of uncertainty from prognostic area models is due to unconstrained wetland area, but generally the magnitude and inter-annual variability agree between diagnostic and prognostic area approaches. Wetland emissions represent about 20 % of the total (natural plus anthropogenic) methane sources estimated by bottom-up approaches. The large range in the estimates of wetland CH₄ emissions results from difficulties in defining wetland CH₄-producing areas as well as in parameterizing terrestrial anaerobic conditions that drive sources and the oxidative conditions leading to sinks (Melton et al., 2013; Poulter et al., 2017; Wania et al., 2013). The ensemble mean emission using diagnostic wetland extent in the models is lower by ~ 35 Tg CH₄ yr⁻¹ than the one previously reported (see Table 3, for 2000–2009 with comparison to Saunois et al., 2016). This difference results from a reduction in double-counting due to (i) decreased wetland area in WAD2M, especially for high-latitude regions where inland waters, i.e., lakes, small ponds, and lakes, were removed, and (ii) to some extent, an improved removal of rice agriculture area using the MIRCA2000 database.

For the last decade, 2008–2017, the average ensemble emissions were 149 Tg CH₄ yr⁻¹ with a range of 102–182.

Table 2. Biogeochemical models that computed wetland emissions used in this study. Runs were performed for the whole period 2000–2017. Models run with prognostic (using their own calculation of wetland areas) and/or diagnostic (using WAD2M) wetland surface areas (see Sect. 3.2.1).

Model	Institution	Prognostic	Diagnostic	References
CLASS-CTEM	Environment and Climate Change Canada	y	y	Arora et al. (2018); Melton and Arora (2016)
DLEM	Auburn University	n	y	Tian et al. (2010, 2015)
ELM	Lawrence Berkeley National Laboratory	y	y	Riley et al. (2011)
JSBACH	MPI	n	y	Kleinen et al. (2020)
JULES	UKMO	y	y	Hayman et al. (2014)
LPJ-GUESS	Lund University	n	y	McGuire et al. (2012)
LPJ-MPI	MPI	n	y	Kleinen et al. (2012)
LPJ-WSL	NASA GSFC	y	y	Zhang et al. (2016)
LPX-Bern	University of Bern	y	y	Spahni et al. (2011)
ORCHIDEE	LSCE	y	y	Ringeval et al. (2011)
TEM-MDM	Purdue University	n	y	Zhuang et al. (2004)
TRIPLEX_GHG	UQAM	n	y	Zhu et al. (2014, 2015)
VISIT	NIES	y	y	Ito and Inatomi (2012)

3.2.2 Other inland water systems (lakes, ponds, reservoirs, streams, rivers)

This category includes methane emissions from freshwater systems (lakes, ponds, reservoirs, streams, and rivers). To date, very few process-based models exist for these fluxes, relying on data-driven approaches and extrapolations. Meta-data analyses are hampered for methane due to a mix of methodological approaches, which capture different components of emissions, and different scales in space and time, depending on method and time of deployment and data processing (Stanley et al., 2016). Altogether, this inconsistency in the data collection makes detailed modelling of fluxes highly uncertain. For many lakes, particularly smaller shallower lakes and ponds, it is established that ebullition and plant fluxes (in lakes with substantial emergent macrophyte communities) can make up a substantial contribution to fluxes, potentially accounting for 50 % to more than 90 % of the flux from these water bodies. While contributions from ebullition appear lower from rivers, there are currently insufficient measurements from these systems to determine its role (Crawford et al., 2014; Stanley et al., 2016). Ebullition fluxes are very challenging to measure, due to the high degree of spatio-temporal variability with very high fluxes occurring in parts of an ecosystem over the time frames of seconds followed by long periods without ebullition.

Freshwater methane fluxes from streams and rivers were first estimated to be $1.5 \text{ Tg CH}_4 \text{ yr}^{-1}$ (Bastviken et al., 2011). However, this study had measurements from only 21 sites globally. More recently, Stanley et al. (2016) compiled a dataset of 385 sites and estimated a diffusive emission of $27 \text{ Tg CH}_4 \text{ yr}^{-1}$ (5th–95th percentiles: $0.01\text{--}160 \text{ Tg CH}_4 \text{ yr}^{-1}$). Detailed regional studies in the tropics and temperate watersheds (Borges et al., 2015; Campeau and del Giorgio, 2014) support a flux in the range of

$27 \text{ Tg CH}_4 \text{ yr}^{-1}$ as opposed to the initial $\sim 1.5 \text{ Tg CH}_4 \text{ yr}^{-1}$. However, the low number of measurements, the lack of clarity on ebullition fluxes, and the large degree of variance in measurements have precluded an accurate spatial representation of stream and river methane fluxes. Canals and ditches have recently been highlighted as high areal emitters (e.g. Stanley et al., 2016), and their contribution to large-scale emission is typically included in estimates for overall running waters so far. No new global estimates have been published since Stanley et al. (2016) and Saunio et al. (2016). As a result, here we use the same estimate for stream and rivers as in Saunio et al. (2016): $27 \text{ Tg CH}_4 \text{ yr}^{-1}$.

Methane emissions from lakes were first estimated to be $1\text{--}20 \text{ Tg CH}_4 \text{ yr}^{-1}$ based on measurements in two systems (Great Fresh Creek, Maryland, and Lake Erie; Ehhalt, 1974). A subsequent global emission estimate was $11\text{--}55 \text{ Tg CH}_4 \text{ yr}^{-1}$ based on measurements from three Arctic lakes and a few temperate and tropical systems (Smith and Lewis, 1992) and $8\text{--}48 \text{ Tg CH}_4 \text{ yr}^{-1}$ using extended data from different latitudes (73 lakes; Bastviken et al., 2004). Based on data from 421 lakes and ponds, Bastviken et al. (2011) updated their values to $71.6 \text{ Tg CH}_4 \text{ yr}^{-1}$, including emissions from non-saline lakes and ponds. High-latitude lakes have received a large amount of attention in the last decade. They include both postglacial and thermokarst lakes (small water bodies formed when peat over melting permafrost collapses), the latter having larger emissions per square metre but smaller regional emissions than the former because of their smaller areal extent (Wik et al., 2016b). Water body depth, sediment type, and ecoclimatic region are the key factors explaining variation in methane fluxes from lakes (Wik et al., 2016b). Small artificial water bodies (ponds) have a high surface-area-to-volume ratio and shallow depth, and they are likely to be a notable source of methane, at least

Table 3. Global methane emissions by source type ($\text{Tg CH}_4 \text{ yr}^{-1}$) from Saunio et al. (2016) (left column pair) and for this work using bottom-up and top-down approaches. Because top-down models cannot fully separate individual processes, only five categories of emissions are provided (see text). Uncertainties are reported as the [min–max] range of reported studies. Differences of $1 \text{ Tg CH}_4 \text{ yr}^{-1}$ in the totals can occur due to rounding errors.

Period of time	Saunio et al. (2016)		This work					
	2000–2009		2000–2009		2008–2017		2017	
Approaches	Bottom-up	Top-down	Bottom-up	Top-down	Bottom-up	Top-down	Bottom-up	Top-down
Natural sources								
Wetlands	183 [151–222]	166 [125–204]	147 [102–179]	180 [153–196]	149 [102–182]	181 [159–200]	145 [100–183]	194 [155–217]
Other natural sources	199 [104–297]	68 [21–130]	222 [143–306]	35 [21–47]	222 [143–306]	37 [21–50]	222 [143–306]	39 [21–50]
Other land sources	185 [99–272]		209 [134–284]					
Freshwaters ^a	122 [60–180]		159 [117–212]					
Geological (onshore)	40 [30–56]		38 [13–53]					
Wild animals	10 [5–15]		2 [1–3]					
Termites	9 [3–15]		9 [3–15]					
Wildfires	3 [1–5]			^k				
Permafrost soils (direct)	1 [0–1]		1 [0–1]					
Vegetation		^j		^j				
Oceanic sources	14 [5–25]		13 [9–22]					
Geological (offshore)	12 [5–20]		7 [5–12]					
Biogenic open and coastal ^b	2 [0–5]		6 [4–10]					
Total natural sources	382 [255–519]	234 [194–292]	369 [245–485]	215 [176–243]	371 [245–488]	218 [183–248]	367 [243–489]	232 [194–267]
Anthropogenic sources								
Agriculture and waste	190 [174–201]	183 [112–241]	192 [178–206]	202 [198–219]	206 [191–223]	217 [207–240]	213 [198–232]	227 [205–246]
Enteric ferm. & manure	103 [95–109] ^c		104 [93–109]		111 [106–116]		115 [110–121]	
Landfills & waste	57 [51–61] ^c		60 [55–63]		65 [60–69]		68 [64–71]	
Rice cultivation	29 [23–35] ^c		28 [23–34]		30 [25–38]		30 [24–40]	
Fossil fuels	112 [107–126]	101 [77–126]	110 [94–129]	101 [71–151]	128 [113–154]	111 [81–131]	135 [121–164]	108 [91–121]
Coal mining	36 [24–43] ^c		32 [24–42]		42 [29–61]		44 [31–63]	
Oil & gas	76 [64–85] ^{c,f}		73 [60–85]		80 [68–92]		84 [72–97]	
Industry	–		2 [0–6]		3 [0–7]		3 [0–8]	
Transport	–		4 [1–11]		4 [1–12]		4 [1–13]	
Biomass & biof. burn.	30 [26–34]	35 [16–53]	31 [26–46]	29 [23–35]	30 [26–40]	30 [22–36]	29 [24–38]	28 [25–32]
Biomass burning	18 [15–20]		19 [15–32]		17 [14–26]		16 [11–24]	
Biofuel burning	12 [9–14]		12 [9–14]		12 [10–14]		13 [10–14]	
Total anthropogenic sources ^g	338 [329–342]	319 [255–357]	334 [321–358]	332 [312–347]	366 [349–393]	359 [336–376]	380 [359–407]	364 [340–381]
Sinks								
Total chemical loss	604 [483–738]	514 ^e	595 [489–749]	505 ^h [459–516]	595 [489–749]	518 ^h [474–532]	595 [489–749]	531 ^h [502–540]
Tropospheric OH	528 [454–617]		553 [476–677]					
Stratospheric loss	51 [16–84]		31 [12–37]					
Tropospheric Cl	25 [13–37]		11 [1–35]					
Soil uptake	28 [9–47]	32 [27–38]	30 [11–49]	34 [27–41]	30 [11–49]	38 [27–45]	30 [11–49]	40 [37–47]
Total sinks	632 [592–785]	546 ^d	625 [500–798]	540 [486–556]	625 [500–798]	556 [501–574]	625 [500–798]	571 [540–585]
Sources – sink imbalance								
Total sources	719 [583–861]	552 [535–566]	703 [566–842]	547 [524–560]	737 [594–881]	576 [550–594]	747 [602–896]	596 [572–614]
Total sinks	632 [592–785]	546 ^d	625 [500–798]	540 [486–556]	625 [500–798]	556 [501–574]	625 [500–798]	571 [540–585]
Imbalance		6 ^d	78	3 [–10–38] ^h	112	13 [0–49] ^h	120	12 [0–41] ^h
Atmospheric growth ⁱ		6.0 [4.9–6.6]		5.8 [4.9–6.6]		18.2 [17.3–19.0]		16.8 [14.0–19.5]

^a Freshwater includes lakes, ponds, reservoirs, streams, and rivers. ^b Includes flux from hydrates considered at 0 for this study, includes estuaries. ^c For IIASA inventory the breakdown of agriculture and waste (rice, enteric fermentation & manure, landfills & waste) and fossil fuel (coal, oil, gas & industry) sources used the same ratios as the mean of the EDGAR and USEPA inventories in Saunio et al. (2016). ^d Total sink was deduced from global mass balance and not directly computed in Saunio et al. (2016). ^e Computed as the difference of global sink and soil uptake in Saunio et al. (2016). ^f Industry and transport emissions were included in the oil & gas category in Saunio et al. (2016). ^g Total anthropogenic emissions are based on estimates of a full anthropogenic inventory and not on the sum of the “agriculture and waste”, “fossil fuels”, and “biofuel and biomass burning” categories (see Sect. 3.1.2). ^h Some inversions did not provide the chemical sink. These values are derived from a subset of the inversion ensemble. ⁱ Atmospheric growth is given in the same unit ($\text{Tg CH}_4 \text{ yr}^{-1}$), based on the conversion factor of $2.75 \text{ Tg CH}_4 \text{ ppb}^{-1}$ given by Prather et al. (2012) and the atmospheric growth rates provided in the text in parts per billion per year. ^j Uncertain but likely small for upland forest and aerobic emissions, potentially large for forested wetland, but likely included elsewhere. ^k We stop reporting this value to avoid potential double-counting with satellite-based products of biomass burning (see Sect. 3.1.5).

at the regional scale (Grinham et al., 2018; Ollivier et al., 2019). These studies found that emissions varied by pond type (for example: livestock rearing farm dams vs. cropping farm dams vs. urban ponds vs. weirs). A rough estimate of the global impact of this emission source is globally significant, between 3 and 8 Tg CH₄ yr⁻¹ (calculated using the mean emission rates from Grinham et al., 2018, and Ollivier et al., 2019) and an estimate of global farm impoundment surface area of 77 000 km² (Downing et al., 2006). This rough estimate does emphasize the potential significance of these sources, although double-counting with current uncertain estimates from natural inland water systems is possible (Thornton et al., 2016a).

A regional estimate for latitudes above 50° north (Wik et al., 2016b) estimated lake and pond methane emissions to be 16.5 Tg CH₄ yr⁻¹ (compared to 13.4 Tg CH₄ yr⁻¹ in Bastviken et al. (2011), above 54° N). Tan et al. (2016) used atmospheric inversion approaches and estimated that the current pan-Arctic (north of 60° N) lakes emit 2.4–14.2 Tg CH₄ yr⁻¹, while a process-based lake biogeochemistry model (bLake4Me) estimated the emissions at 11.9 [7.1–17.3] Tg CH₄ yr⁻¹ (Tan and Zhuang, 2015). These numbers for northern or Arctic lakes need to be considered with regard to the latitudinal area encompassed which differs among studies (Thornton et al., 2016a). Saunois et al. (2016) estimates for emissions from natural lakes and ponds were based on those from Bastviken et al. (2011), using the emissions from the northern high latitudes above 50° N from Wik et al. (2016b), leading to a rounded mean value of 75 Tg CH₄ yr⁻¹. Based on the bLake4Me gridded map from Tan and Zhuang (2015), we calculate lake and pond emissions of 5.2 Tg CH₄ yr⁻¹ above 66° N, close to the 6.8 Tg CH₄ yr⁻¹ found by Bastviken et al. (2011). Averaging these two values for the emissions above 66° N and combining with Bastviken et al. (2011) estimates south of 66° N (64.8 Tg CH₄ yr⁻¹) leads to a rounded mean global estimate of 71 Tg CH₄ yr⁻¹ close to that of Bastviken et al. (2011) (71.6 Tg CH₄ yr⁻¹ at the global scale).

On top of emissions pathways described for inland waters, reservoirs have specific ones including degassing of CH₄ from turbines (hydropower reservoirs only) and elevated diffusive emissions in rivers downstream of the reservoir – these latter emissions are enhanced if the water outlet comes from anoxic CH₄-rich hypolimnion waters in the reservoir (Bastviken et al., 2004; Guérin et al., 2006, 2016). In Saunois et al. (2016), methane emissions from reservoirs were estimated to be 20 Tg CH₄ yr⁻¹ using Bastviken et al. (2011), which was based on data from 32 systems. A more recent and extensive review estimated total reservoir emissions to be 18 Tg CH₄ yr⁻¹ (95 % confidence interval 12–30 Tg CH₄ yr⁻¹; $n = 75$, Deemer et al., 2016) and is used to revise our estimate in this study.

Combining emissions from lakes and ponds from Bastviken et al. (2011) (71.6 Tg CH₄ yr⁻¹) with the recent estimate of Deemer et al. (2016) for reservoirs and the streams

and river estimates from Stanley et al. (2016) leads to total inland freshwater emissions of 117 Tg CH₄ yr⁻¹. Recently, using a new up-scaling approach based on size weighting productivity and chlorophyll *a*, DelSontro et al. (2018) provided combined lake and reservoir estimates of 104 (5th–95th percentiles: 67–165), 149 (5th–95th percentiles: 95–236), and 185 (5th–95th percentiles: 119–295) Tg CH₄ yr⁻¹, using the lake size distributions from Downing et al. (2006), Messenger et al. (2016), and Verpoeter et al. (2014), respectively. These estimates are higher (by 10 %, 57 %, and almost 100 %, respectively) than previously reported in Saunois et al. (2016) (i.e., 95 Tg CH₄ yr⁻¹ for lakes, ponds, and reservoirs).

Previously, Kirschke et al. (2013) reported a range of 8–73 Tg CH₄ yr⁻¹ for this ensemble of emissions and Saunois et al. (2016) a mean value of 122 Tg CH₄ yr⁻¹ (75 Tg CH₄ yr⁻¹ for lakes and ponds, adding 20 Tg CH₄ yr⁻¹ for reservoirs (Bastviken et al., 2011) and 27 Tg CH₄ yr⁻¹ for streams and rivers; Stanley et al., 2016). This mean value reported by Saunois et al. (2016) was based on a single set of estimates, to which a 50 % uncertainty was associated as a range (60–180 Tg CH₄ yr⁻¹). Here the new estimates of DelSontro et al. (2018) lead to a mean estimate of all inland freshwaters at 159 Tg CH₄ yr⁻¹ associated with the range 117–212 Tg CH₄ yr⁻¹ that reflects the minimum and maximum values of the available studies (see Methodology, Sect. 2). However, it should be noted that this range does not consider the uncertainty of individual studies. Importantly, these current estimates do not include the smallest size class of lakes or ephemeral streams, resulting in a possible misallocation of freshwater fluxes to wetland ecosystems in spite of the attempts to discount open-water emissions from the wetland estimate (see above). The present data indicate that lakes or natural ponds, flooded land–reservoirs, and streams–rivers account for 70 %, 13 %, and 17 % of the average inland water fluxes, respectively (given the large uncertainty, the percentages should be seen as approximate relative magnitudes only). The anthropogenic part of the inland water fluxes is best constrained for larger reservoirs but remains less clear for other human-made flooded land. It should be noted that issues regarding spatio-temporal variability are not considered in consistent ways at present (Natchimuthu et al., 2015; Wik et al., 2016a). Given the inconsistencies in the areal flux data and in area estimates, the aim to make frequent updates of the methane emissions is presently not possible for inland water emissions. Even more than for other emission categories, differences in inland water flux values used to estimate emissions, as well as how the data were processed, are more likely to represent differences between data, rather than reflecting real temporal trends in the environment.

The improvement in quantifying inland water fluxes is highly dependent on the availability of more accurate assessments of their surface area. For streams and rivers, the 355 000 km² used in Bastviken et al. (2011) was re-evaluated to 540 000 km² by Stanley et al. (2016) due to a new surface area estimate from Raymond et al. (2013). Regarding lakes

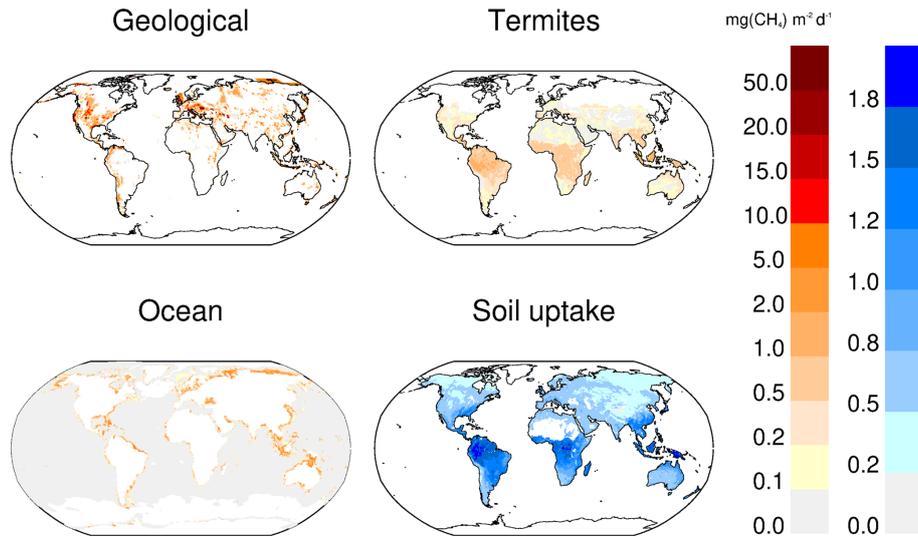


Figure 4. Methane emissions ($\text{mg CH}_4 \text{ m}^{-2} \text{ d}^{-1}$) from three natural sources (left colour scale): geological (Etiopie et al., 2019), termites (this study), and oceans (Weber et al., 2019). Methane uptake in soils ($\text{mg CH}_4 \text{ m}^{-2} \text{ d}^{-1}$) presented in positive units (right colour scale) and based on Murguía-Flores et al. (2018).

and reservoirs, the three current inventories (Downing et al., 2006; Messenger et al., 2016; Verpoorter et al., 2014) show typical differences of a factor of 2 to 5 by size class. Also, it was noted that small ponds, which were not included in either Downing et al. (2006) or Verpoorter et al. (2014), have a diffusive flux higher than any other size class of lakes (Holgerson and Raymond, 2016). Further analysis, and possibly more refined process-based models, is still necessary and urgent to evaluate these global up-scaled estimates against region-specific approaches such as in Wik et al. (2016a) for the northern high-latitude lakes.

In this budget, we report a mean value of $159 \text{ Tg CH}_4 \text{ yr}^{-1}$ from freshwater systems (lakes, ponds, reservoirs, streams, and rivers), with a range of $117\text{--}212 \text{ Tg CH}_4 \text{ yr}^{-1}$. This range shows the minimum and maximum estimates but excludes the uncertainty from each single estimate, which is expected to be large.

3.2.3 Onshore and offshore geological sources

Significant amounts of methane, produced within the Earth's crust, naturally migrate to the atmosphere through tectonic faults and fractured rocks. Major emissions are related to hydrocarbon production in sedimentary basins (microbial and thermogenic methane), through continuous or episodic exhalations from onshore and shallow marine hydrocarbon seeps and through diffuse soil microseepage (Etiopie, 2015). Specifically, five source categories have been considered. Four are onshore sources: gas–oil seeps, mud volcanoes, diffuse microseepage, and geothermal manifestations including volcanoes. One source is offshore: submarine seepage, which may include the same types of gas manifestations occurring on land. Etiopie et al. (2019) have produced the first

gridded maps of geological methane emissions and their isotopic signature for these five categories, with a global total of $37.4 \text{ Tg CH}_4 \text{ yr}^{-1}$ (reproduced in Fig. 4). According to them, the grid maps do not represent, however, the actual global geological CH_4 emission because the datasets used for the spatial gridding (developed for modelling purposes) were not complete or did not contain the information necessary for improving all previous estimates. Combining the best estimates for the five categories of geological sources (from grid maps or from previous statistical and process-based models), the breakdown by category reveals that onshore microseepage dominates ($24 \text{ Tg CH}_4 \text{ yr}^{-1}$), with the other categories having similar smaller contributions: as average values, $4.7 \text{ Tg CH}_4 \text{ yr}^{-1}$ for geothermal manifestations, about $7 \text{ Tg CH}_4 \text{ yr}^{-1}$ for submarine seepage, and $9.6 \text{ Tg CH}_4 \text{ yr}^{-1}$ for onshore seeps and mud volcanoes. These values lead to a global bottom-up geological emission mean of $45 [27\text{--}63] \text{ Tg CH}_4 \text{ yr}^{-1}$ (Etiopie and Schwietzke, 2019).

While all bottom-up and some top-down estimates, following different and independent techniques from different authors, consistently suggest a global geo- CH_4 emission of the order of $40\text{--}50 \text{ Tg yr}^{-1}$, the radiocarbon ($^{14}\text{C}\text{--CH}_4$) data in ice cores reported by Hmiel et al. (2020) appear to lower the estimate, with a minimum of about $1.6 \text{ Tg CH}_4 \text{ yr}^{-1}$ and a maximum estimated value of $5.4 \text{ Tg CH}_4 \text{ yr}^{-1}$ (95 %) for the pre-industrial period. The discrepancy between Hmiel et al. (2020) and all other estimates continues to feed the debate. Eastern Siberian Arctic Shelf (ESAS) emissions have been estimated at $\sim 3 \text{ Tg CH}_4 \text{ yr}^{-1}$ based on current atmospheric surface observations (Thornton et al., 2020), corresponding to the same order of magnitude of the estimate from Hmiel et al. (2020) for global geological emissions. How-

ever, ESAS emissions are likely from both thermogenic and biogenic origins (e.g. Berchet et al., 2020). More investigation and confrontation between top-down and bottom-up results are needed to reduce this discrepancy.

Waiting for further investigation on this topic, we decided to keep the best estimates from Etiope and Shwietzke (2019) for the mean values and associate them with the lowest estimates reported in Etiope et al. (2019). Thus, we report a total global geological emission of 45 [18–63] Tg CH₄ yr⁻¹, with a breakdown between offshore emissions of 7 [5–10] Tg CH₄ yr⁻¹ and onshore emissions of 38 [13–53] Tg CH₄ yr⁻¹. The updated bottom-up estimate is slightly lower than the previous budget mostly due to a reduction of estimated emissions of onshore and offshore seeps (see Sect. 3.2.6 for more offshore contribution explanations).

3.2.4 Termites

Termites are an infraorder of insects (Isoptera), which occur predominantly in the tropical and subtropical latitudes (Abe et al., 2000). CH₄ is released during the anaerobic decomposition of plant biomass in their gut (Sanderson, 1996). The uncertainty related to this CH₄ source is very high as CH₄ emissions from termites in different ecosystem types can vary and are driven by a range of factors, while the number of field measurements, both of termite biomass and emissions, is relatively low (Kirschke et al., 2013).

In Kirschke et al. (2013) (see their Supplement), a re-analysis of CH₄ emissions from termites at the global scale was proposed. Their CH₄ emissions per unit of area were estimated as the product of termite biomass, termite CH₄ emissions per unit of termite mass, and a scalar factor expressing the effect of land use and cover change; the latter two terms were estimated from published literature re-analysis. For tropical climates, termite biomass was estimated by a simple regression model representing its dependence on gross primary productivity (GPP), whereas for forest and grassland ecosystems of the warm temperate climates and for shrublands of the Mediterranean subclimate, termite biomass was estimated from data reported by Sanderson (1996). The CH₄ emission factor per unit of termite biomass (g_{termite}) was estimated as 2.8 mg CH₄ (g_{termite})⁻¹ h⁻¹ for tropical ecosystems and Mediterranean shrublands (Kirschke et al., 2013) and 1.7 mg CH₄ (g_{termite})⁻¹ h⁻¹ for temperate forests and grasslands (Fraser et al., 1986). Emissions were scaled up and annual CH₄ fluxes were computed for the three periods 1982–1989, 1990–1999, and 2000–2007 representative of the 1980s, 1990s, and 2000s, respectively.

The re-analysis of termite emissions proposed in Saunois et al. (2016) maintained the same approach, but the data were calculated using climate zoning (following the Köppen–Geiger classification) applied to updated climate datasets by Santini and di Paola (2015) and were adapted to consider different combinations of termite biomass per unit area and CH₄ emission factor per unit of termite biomass.

Here, this analysis is extended to cover the periods 2000–2007 and 2010–2016. This latest estimate follows the approach outlined above for Saunois et al. (2016). However, in order to extend the analysis to 2016, an alternative MODIS-based measure of GPP from Zhang et al. (2017), rather than from Jung et al. (2009), and Jung et al. (2011) was used to estimate termite biomass. To have coherent datasets of GPP and land use, the latter variable, previously derived from Ramankutty and Foley (1999), was substituted for MODIS maps (Channan et al., 2014; Friedl et al., 2010). These new estimates covered 2000–2007 and 2010–2016 using 2002 and 2012 MODIS data as an average reference year for each period, respectively.

Termite CH₄ emissions show only little inter-annual and inter-decadal variability (0.1 Tg CH₄ yr⁻¹), whereas there is strong regional variability, with tropical South America and Africa being the main sources (23 % and 28 % of the total emissions, respectively) due to the extent of their natural forest and savannah ecosystems (Fig. 4). Changing the GPP and land use dataset sources had only a minimal impact on the 2000–2007 global termite flux, increasing it from 8.7 Tg CH₄ yr⁻¹ as found in the first two re-analyses (Kirschke et al., 2013; Saunois et al., 2016) to 9.9 Tg CH₄ yr⁻¹ (present data), well within the estimated uncertainty (8.7 ± 3.1 Tg CH₄ yr⁻¹). However, it had a noticeable effect on the spatial distribution of the flux (Fig. S2). The most obvious of these changes is a halving of the Southeast Asian flux, aligned with shifts in the underlying GPP product. Previous studies (Mercado et al., 2009; Zhang et al., 2017) had linked these GPP shifts to a methodological issue with light-use efficiency that drove an underestimation of evergreen broadleaf and evergreen needleleaf forest GPP, biomes which are prevalent in the tropics. This value is close to the average estimate derived from previous up-scaling studies, which report values spanning from 2 to 22 Tg CH₄ yr⁻¹ (Ciais et al., 2013).

In this study, we report a decadal value of 9 Tg CH₄ yr⁻¹ (range [3–15] Tg CH₄ yr⁻¹, Table 3).

3.2.5 Wild animals

Wild ruminants emit methane through the microbial fermentation process occurring in their rumen, similarly to domesticated livestock species (USEPA, 2010b). Using a total animal population of 100–500 million, Crutzen et al. (1986) estimated the global emissions of CH₄ from wild ruminants to be in the range of 2–6 Tg CH₄ yr⁻¹. More recently, Pérez-Barbería (2017) lowered this estimate to 1.1–2.7 Tg CH₄ yr⁻¹ using a total animal population estimate of 214 million (range of 210–219), arguing that the maximum number of animals (500 million) used in Crutzen et al. (1986) was poorly justified. Moreover Pérez-Barbería (2017) also stated that the value of 15 Tg CH₄ yr⁻¹ found in the last IPCC reports is much higher than their estimate because this value comes from an extrapolation of Crutzen's work for the

last glacial maximum when the population of wild animals was much larger, as originally proposed by Chappellaz et al. (1993).

Based on these findings, the range adopted in this updated methane budget is 2 [1–3] Tg CH₄ yr⁻¹ (Table 3).

3.2.6 Oceanic sources

Oceanic sources comprise coastal ocean and open ocean methane release. Possible sources of oceanic CH₄ include (1) production from marine (bare and vegetated) sediments or thawing sub-sea permafrost; (2) in situ production in the water column, especially in the coastal ocean because of submarine groundwater discharge (USEPA, 2010b); (3) leaks from geological marine seepage (see also Sect. 3.2.3); and (4) emission from the destabilization of marine hydrates. Once at the seabed, methane can be transported through the water column by diffusion in a dissolved form (especially in the upwelling zones), or by ebullition (gas bubbles, e.g. from geological marine seeps), for instance, in shallow waters of continental shelves. In coastal vegetated habitats methane can also be transported to the atmosphere through the *aerenchyma* of emergent aquatic plants (Ramachandran et al., 2004).

The most common biogenic ocean emission value found in the literature is 10 Tg CH₄ yr⁻¹ (Rhee et al., 2009b). It appears that most studies rely on the work of Ehhalt (1974), where the value was estimated on the basis of the measurements done by Swinnerton and co-workers (Lamontagne et al., 1973; Swinnerton and Linnenbom, 1967) for the open ocean, combined with purely speculated emissions from the continental shelf. Based on basin-wide observations using updated methodologies, three studies found estimates ranging from 0.2 to 3 Tg CH₄ yr⁻¹ (Bates et al., 1996; Conrad and Seiler, 1988; Rhee et al., 2009b), associated with supersaturations of surface waters that are an order of magnitude smaller than previously estimated, both for the open ocean (saturation anomaly ~ 0.04 ; see Rhee et al., 2009a, Eq. 4) and for the continental shelf (saturation anomaly ~ 0.2). In their synthesis, indirectly referring to the original observations from Lambert and Schmidt (1993), Wuebbles and Hayhoe (2002) use a value of 5 Tg CH₄ yr⁻¹. Proposed explanations for discrepancies regarding sea-to-air methane emissions in the open ocean rely on experimental biases in the former studies of Swinnerton and Linnenbom (Rhee et al., 2009b). This may explain why the Bange et al. (1994) compilation cites a global source of 11–18 Tg CH₄ yr⁻¹ with a dominant contribution of coastal regions. Here, we report a range of 0–5 Tg CH₄ yr⁻¹, with a mean value of 2 Tg CH₄ yr⁻¹ for biogenic emissions from open and coastal ocean (excluding estuaries).

Biogenic emissions from brackish waters (estuaries, coastal wetlands) were not reported in the previous budget (Saunois et al., 2016). Methane emissions from estuaries were originally estimated by Bange et al. (1994), Upstill-

Goddard et al. (2000), and Middelburg et al. (2002) to be comprised between 1 and –3 Tg CH₄ yr⁻¹. This range was later revised upwards by Borges and Abril (2011) to about 7 Tg CH₄ yr⁻¹ based on a methodology distinguishing between different estuarine types and accounting for the contribution of tidal flats, marshes, and mangroves, for a total of 39 systems and a global “inner” estuarine surface area of 1.1×10^6 km² (Laruelle et al., 2013). The same methodology as in Laruelle et al. (2013) has been applied here to the same systems using an expanded database of local and regional measurements (72 systems) and suggests however that global estuarine CH₄ emissions were overestimated and may actually not surpass 3–3.5 Tg CH₄ yr⁻¹. Despite this overall reduction, the specific contribution of sediment and water emissions from mangrove ecosystems is however higher and contributes < 0.1 to 1.7 Tg CH₄ yr⁻¹ globally (Rosentreter et al., 2018). This estuarine estimate does not include the uncertain contribution from large river plumes protruding onto the shelves. Their surface area reaches about 3.7×10^6 km² (Kang et al., 2013) but because of significantly lower CH₄ concentration (e.g. Osudar et al., 2015; Zhang et al., 2008) than in inner estuaries, the outgassing associated with these plumes likely does not exceed 1–2 Tg CH₄ yr⁻¹. Seagrass meadows are also not included, although they might release < 0.1 to 2.7 Tg CH₄ yr⁻¹ (Garcias-Bonet and Duarte, 2017). These methane emissions from vegetated coastal ecosystems can partially offset (Rosentreter et al., 2018) their “blue carbon” sink (e.g. Mcleod et al., 2011; Nellemann et al., 2009). Note that the latter two contributions might partly overlap with oceanic (open and coastal) sources estimates. The total (inner and outer) estuarine emission flux, which is based on only about 80 systems is thus in the range 4–5 Tg CH₄ yr⁻¹ (including marshes and mangrove). High uncertainties in coastal ocean emission estimates can be reduced by better defining the various coastal ecosystem types and their boundaries to avoid double-counting (e.g. estuaries, brackish wetlands, freshwater wetlands), updating the surface area of each of these coastal systems, and better quantifying methane emission rates in each ecosystem type.

As a result, here we report a range of 4–10 Tg CH₄ yr⁻¹ for emissions from coastal and open ocean (including estuaries), with a mean value of 6 Tg CH₄ yr⁻¹.

The production of methane at the seabed is known to be significant. For instance, marine seepages emit up to 65 Tg CH₄ yr⁻¹ globally at seabed level (USEPA, 2010b). What is uncertain is the flux of oceanic methane reaching the atmosphere. For example, bubble plumes of CH₄ from the seabed have been observed in the water column, but not detected in the Arctic atmosphere (Fisher et al., 2011; Westbrook et al., 2009). There are several barriers preventing methane from being expelled to the atmosphere (James et al., 2016). From below the seafloor to the sea surface, gas hydrates and permafrost serve as a barrier to fluid and gas migration towards the seafloor; microbial activity around the seafloor can strongly oxidize methane releases or production;

further oxidation occurs in the water column; the oceanic pycnocline acts as a physical barrier towards the surface waters, including efficient dissolution of bubbles; and finally, surface oceans are aerobic and contribute to the oxidation of dissolved methane. However, surface waters can be more supersaturated than the underlying deeper waters, leading to a methane paradox (Sasakawa et al., 2008). Possible explanations involve (i) upwelling in areas with surface mixed layers covered by sea ice (Damm et al., 2015), (ii) the release of methane by the degradation of dissolved organic matter phosphonates in aerobic conditions (Repeta et al., 2016), (iii) methane production by marine algae (Lenhart et al., 2016), or (iv) methane production within the anoxic centre of sinking particles (Sasakawa et al., 2008), but more work is still needed to be conclusive about this apparent paradox.

For geological emissions, the most used value has long been $20 \text{ Tg CH}_4 \text{ yr}^{-1}$, relying on expert knowledge and literature synthesis proposed in a workshop reported in Kvenvolden et al. (2001); the authors of this study recognize that this was a first estimation and needs revision. Since then, oceanographic campaigns have been organized, especially to sample bubbling areas of active seafloor gas seep bubbling. For instance, Shakhova et al. (2010, 2014) infer $8\text{--}17 \text{ Tg CH}_4 \text{ yr}^{-1}$ in emissions just for the Eastern Siberian Arctic Shelf (ESAS), based on the extrapolation of numerous but local measurements, and possibly related to thawing sub-seabed permafrost (Shakhova et al., 2015). Because of the highly heterogeneous distribution of dissolved CH_4 in coastal regions, where bubbles can most easily reach the atmosphere, extrapolation of in situ local measurements to the global scale can be hazardous and lead to biased global estimates. Indeed, using very precise and accurate continuous land shore-based atmospheric methane observations in the Arctic region, Berchet et al. (2016) found a range of emissions for ESAS of $\sim 2.5 \text{ Tg CH}_4 \text{ yr}^{-1}$ (range [0–5]), 4–8 times lower than Shakhova's estimates. Such a reduction in ESAS emission estimate has also been inferred from oceanic observations by Thornton et al. (2016b) with a maximum sea–air CH_4 flux of $2.9 \text{ Tg CH}_4 \text{ yr}^{-1}$ for this region. Etiope et al. (2019) suggested a minimum global total submarine seepage emission of $3.9 \text{ Tg CH}_4 \text{ yr}^{-1}$ simply summing published regional emission estimates for 15 areas for identified emission areas (above $7 \text{ Tg CH}_4 \text{ yr}^{-1}$ when extrapolated to include non-measured areas). These recent results, based on different approaches, suggest that the current estimate of $20 \text{ Tg CH}_4 \text{ yr}^{-1}$ is too large and needs revision.

Therefore, as discussed in Sect. 3.2.2, we report here a reduced range of $5\text{--}10 \text{ Tg CH}_4 \text{ yr}^{-1}$ for marine geological emissions compared to the previous budget, with a mean value of $7 \text{ Tg CH}_4 \text{ yr}^{-1}$.

Among the different origins of oceanic methane, hydrates have attracted a lot of attention. Methane hydrates (or clathrates) are ice-like crystals formed under specific temperature and pressure conditions (Milkov, 2005). Methane hydrates can be either of biogenic origin (formed in situ at

depth in the sediment by microbial activity) or of thermogenic origin (non-biogenic gas migrated from deeper sediments and trapped due to pressure–temperature conditions or due to some capping geological structure such as marine permafrost). The total stock of marine methane hydrates is large but uncertain, with global estimates ranging from hundreds to thousands of Pg CH_4 (Klauda and Sandler, 2005; Wallmann et al., 2012).

Concerning more specifically atmospheric emissions from marine hydrates, Etiope (2015) points out that current estimates of methane air–sea flux from hydrates ($2\text{--}10 \text{ Tg CH}_4 \text{ yr}^{-1}$ in Ciais et al., 2013, or Kirschke et al., 2013) originate from the hypothetical values of Cicerone and Oremland (1988). No experimental data or estimation procedures have been explicitly described along the chain of references since then (Denman et al., 2007; IPCC, 2001; Kirschke et al., 2013; Lelieveld et al., 1998). It was estimated that $\sim 473 \text{ Tg CH}_4$ has been released in the water column over 100 years (Kretschmer et al., 2015). Those few teragrams per year become negligible once consumption in the water column has been accounted for. While events such as submarine slumps may trigger local releases of considerable amounts of methane from hydrates that may reach the atmosphere (Etiope, 2015; Paull et al., 2002), on a global scale, present-day atmospheric methane emissions from hydrates do not appear to be a significant source to the atmosphere, and at least formally, we should consider $0 (< 0.1) \text{ Tg CH}_4 \text{ yr}^{-1}$ emissions.

Combination (biogenic and geological) of open and coastal oceanic emissions. Summing biogenic, geological and hydrate emissions from open and coastal ocean (excluding estuaries) leads to a total of $9 \text{ Tg CH}_4 \text{ yr}^{-1}$ (range 5–17). A recent work (Weber et al., 2019) suggests a new robust estimate of the climatological oceanic flux: the diffusive flux was estimated as $2\text{--}6 \text{ Tg CH}_4 \text{ yr}^{-1}$ and the ebullitive flux as $2\text{--}11 \text{ Tg CH}_4 \text{ yr}^{-1}$, giving a total (open and coastal) oceanic flux estimate of $6\text{--}15 \text{ Tg CH}_4 \text{ yr}^{-1}$ (90% confidence interval) when the probability distributions for the two pathways are combined. Distribution of open and coastal oceanic fluxes from Weber et al. (2019) is shown in Fig. 4. This more robust estimate took benefit from synthesis of in situ measurements of atmospheric and surface water methane concentrations and of bubbling areas, and of the development of process-based models for oceanic methane emissions. Another recent estimate based on the biogeochemistry model PlankTOM10 (Le Quéré et al., 2016) calculates an open and coastal ocean methane flux (excluding estuaries) of $8 [-13/+19] \text{ Tg CH}_4 \text{ yr}^{-1}$ (Buitenhuis et al., 2020), with a coastal contribution of 44%. Our estimate ($9 [5\text{--}17] \text{ Tg CH}_4 \text{ yr}^{-1}$) agrees well with the estimates of $6\text{--}15 \text{ Tg CH}_4 \text{ yr}^{-1}$ by Weber et al. (2019) and $8 \text{ Tg CH}_4 \text{ yr}^{-1}$ (Buitenhuis et al., 2020).

Methane emissions from brackish water were not estimated in Saunois et al. (2016) and an additional $4 \text{ Tg CH}_4 \text{ yr}^{-1}$ is reported in this budget. As a result, includ-

ing estuaries in the oceanic budget, we report a range of 9–22 Tg CH₄ yr⁻¹, with a mean value of 13 Tg CH₄ yr⁻¹, leading to similar total oceanic emissions despite a reduced estimate in geological offshore emissions compared to Saunois et al. (2016).

3.2.7 Terrestrial permafrost and hydrates

Permafrost is defined as frozen soil, sediment, or rock having temperatures at or below 0 °C for at least 2 consecutive years (Harris et al., 1988). The total extent of permafrost in the Northern Hemisphere is about 14 million km² or 15 % of the exposed land surface (Obu et al., 2019). As the climate warms, large areas of permafrost are also warming, and if soil temperatures pass 0 °C, thawing of the permafrost occurs. Permafrost thaw is most pronounced in southern and spatially isolated permafrost zones, but it also occurs in northern continuous permafrost (Obu et al., 2019). Thaw occurs either as a gradual, often widespread, deepening of the active layer or as more rapid localized thaw associated with loss of massive ground ice (thermokarst) (Schuur et al., 2015). A total of 1035 ± 150 Pg of carbon can be found in the upper 3 m of permafrost regions, or ~ 1300 (1100–1500) Pg C for all permafrost (Hugelius et al., 2014).

The thawing permafrost can generate direct and indirect methane emissions. Direct methane emissions rely on the release of methane contained in the thawing permafrost. This flux to the atmosphere is small and estimated to be a maximum of 1 Tg CH₄ yr⁻¹ at present (USEPA, 2010b). Indirect methane emissions are probably more important. They rely on (1) methanogenesis induced when the organic matter contained in thawing permafrost is released, (2) the associated changes in land surface hydrology possibly enhancing methane production (McCalley et al., 2014), and (3) the formation of more thermokarst lakes from erosion and soil collapse. Such methane production is probably already significant today and is likely to become more important in the future associated with climate change and strong positive feedback from thawing permafrost (Schuur et al., 2015). However, indirect methane emissions from permafrost thawing are difficult to estimate at present, with very few data to refer to, and in any case they largely overlap with wetland and freshwater emissions occurring above or around thawing areas. For instance, based on lake and soil measurements Walter Anthony et al. (2016) found that methane emissions (~ 4 Tg CH₄ yr⁻¹) from thermokarst lakes that have expanded over the past 60 years were directly proportional to the mass of soil carbon inputs to the lakes from the erosion of thawing permafrost.

Here, we choose to report only the direct emission range of 0–1 Tg CH₄ yr⁻¹, keeping in mind that current wetland, thermokarst lakes, and other freshwater methane emissions already likely include a significant indirect contribution originating from thawing permafrost.

3.2.8 Vegetation

Three distinct pathways for the production and emission of methane by living vegetation are considered here (see Covey and Megonigal, 2019, for an extensive review). Firstly, plants produce methane through an abiotic photochemical process induced by stress (Keppler et al., 2006). This pathway was initially criticized (e.g. Dueck et al., 2007; Nisbet et al., 2009), and although numerous studies have since confirmed aerobic emissions from plants and better resolved its physical drivers (Fraser et al., 2015), global estimates still vary by 2 orders of magnitude (Liu et al., 2015). This plant source has not been confirmed in-field however, and although the potential implication for the global methane budget remains unclear, emissions from this source are certainly much smaller than originally estimated in Keppler et al. (2006) (Bloom et al., 2010; Fraser et al., 2015). Second, and of clearer significance, plants act as “straws”, drawing up and releasing microbially produced methane from anoxic soils (Cicerone and Shetter, 1981; Rice et al., 2010). For instance, in the forested wetlands of Amazonia, tree stems are the dominant ecosystem flux pathway for soil-produced methane; therefore, including stem emissions in ecosystem budgets can reconcile regional bottom-up and top-down estimates (Pangala et al., 2017). Third, the stems of both living trees (Covey et al., 2012) and dead wood (Covey et al., 2016) provide an environment suitable for microbial methanogenesis. Static chambers demonstrate locally significant through-bark flux from both soil (Pangala et al., 2013, 2015) and tree-stem-based methanogens (Pitz and Megonigal, 2017; Wang et al., 2016). A recent synthesis indicates stem CH₄ emissions significantly increase the source strength of forested wetlands and modestly decrease the sink strength of upland forests (Covey and Megonigal, 2019). The scientific activity covering CH₄ emissions in forested ecosystems reveals a far more complex story than previously thought, with an interplay of, productive–consumptive, aerobic–anaerobic, biotic–abiotic processes occurring between upland–wetland soils, trees, and atmosphere. Understanding the complex processes that regulate CH₄ source–sink dynamics in forests and estimating their contribution to the global methane budget requires cross-disciplinary research, more observations, and new models that can overcome the classical binary classifications of wetland versus upland forest and of emitting versus uptaking soils (Barba et al., 2019; Covey and Megonigal, 2019). Although we recognize these emissions are potentially large (particularly tree transport from inundated soil), global estimates for each of these pathways remain highly uncertain and/or are currently ascribed here to other flux category sources (e.g. inland waters, wetlands, upland soils).

3.3 Methane sinks and lifetime

Methane is the most abundant reactive trace gas in the troposphere and its reactivity is important to both tropospheric

and stratospheric chemistry. The main atmospheric sink of methane ($\sim 90\%$ of the total sink mechanism) is oxidation by the hydroxyl radical (OH), mostly in the troposphere (Ehhalt, 1974). Other losses are by photochemistry in the stratosphere (reactions with chlorine atoms, Cl, and excited atomic oxygen, $O(^1D)$), oxidation in soils (Curry, 2007; Dutaour and Verchot, 2007), and photochemistry in the marine boundary layer (reaction with Cl; Allan et al., 2007; Thornton et al., 2010). Uncertainties in the total sink of methane as estimated by atmospheric chemistry models are on the order of 20%–40% (Saunois et al., 2016). It is much less (10%–20%) when using atmospheric proxy methods (e.g. methyl chloroform; see below) as in atmospheric inversions (Saunois et al., 2016). In the present release of the global methane budget, we estimate bottom-up methane chemical sinks and lifetime mainly based on global model results from the Chemistry Climate Model Initiative (CCMI) (Morgenstern et al., 2017).

3.3.1 Tropospheric OH oxidation

OH radicals are produced following the photolysis of ozone (O_3) in the presence of water vapour. OH is destroyed by reactions with CO, CH_4 , and non-methane volatile organic compounds.

Following the Atmospheric Chemistry and Climate Model Intercomparison Project (ACCMIP), which studied the long-term changes in atmospheric composition between 1850 and 2100 (Lamarque et al., 2013), a new series of experiments was conducted by several chemistry–climate models and chemistry-transport models participating in the Chemistry-Climate Model Initiative (CCMI) (Morgenstern et al., 2017). Mass-weighted OH tropospheric concentrations do not directly represent methane loss, as the spatial and vertical distributions of OH affect this loss, through, in particular, the temperature dependency and the distribution of methane (e.g. Zhao et al., 2019). However, estimating OH concentrations and spatial and vertical distributions is a key step in estimating methane loss through OH. Over the period 2000–2010, the multi-model mean (11 models) global mass-weighted OH tropospheric concentration was $11.7 \pm 1.0 \times 10^5$ molecules cm^{-3} (range 9.9 – 14.4×10^5 molecules cm^{-3} ; Zhao et al., 2019) consistent with the previous estimates from ACCMIP ($11.7 \pm 1.0 \times 10^5$ molecules cm^{-3} , with a range of 10.3 – 13.4×10^5 molecules cm^{-3} , Voulgarakis et al., 2013, for the year 2000) and the estimates of Prather et al. (2012) at $11.2 \pm 1.3 \times 10^5$ molecules cm^{-3} . Nicely et al. (2017) attribute the differences in OH simulated by different chemistry-transport models to, in decreasing order of importance, different chemical mechanisms, various treatments of the photolysis rate of ozone, and modelled ozone and carbon monoxide. Besides the uncertainty on global OH concentrations, there is an uncertainty in the spatial and temporal distribution of OH. Models often simulate higher OH in the Northern Hemisphere leading to a NH / SH OH ratio

greater than 1 (Naik et al., 2013; Zhao et al., 2019). However, there is evidence for parity in inter-hemispheric OH concentrations (Patra et al., 2014), which needs to be confirmed by other observational and model-derived estimates.

OH concentrations and their changes can be sensitive to climate variability (Dlugokencky et al., 1996; Holmes et al., 2013; Turner et al., 2018), biomass burning (Voulgarakis et al., 2015) and anthropogenic activities. For instance, the increase in the oxidizing capacity of the troposphere in South and East Asia associated with increasing NO_x emissions (Mijling et al., 2013) and decreasing CO emissions (Yin et al., 2015) possibly enhances CH_4 oxidation and therefore limits the atmospheric impact of increasing emissions (Dalsøren et al., 2009). Despite such large regional changes, the global mean OH concentration was suggested to have changed only slightly over the past 150 years (Naik et al., 2013). This is due to the compensating effects of the concurrent increases in positive influences on OH (water vapour, tropospheric ozone, nitrogen oxides (NO_x) emissions, and UV radiation due to decreasing stratospheric ozone) and of OH sinks (methane burden, carbon monoxide and non-methane volatile organic compound emissions and burden). CCMI models show OH inter-annual variability ranging from 0.4% to 1.8% (Zhao et al., 2019) over 2000–2010, lower than the value deduced from methyl chloroform measurements (proxy, top-down approach). However, these simulations consider meteorology variability but not emission inter-annual variability (e.g., from biomass burning) and thus are expected to simulate lower OH inter-annual variability than in reality. Using an empirical model constrained by global observations of ozone, water vapour, methane, and temperature as well as the simulated effects of changing NO_x emissions and tropical expansion, Nicely et al. (2017) found an inter-annual variability in OH of about 1.3%–1.6% between 1980 and 2015, in agreement with the methyl chloroform proxy (Montzka et al., 2011).

We report here a climatological range for the tropospheric loss of methane by OH oxidation of 553 [476–677] Tg CH_4 yr^{-1} derived from the seven models that contributed to CCMI for the total tropospheric loss of methane by OH oxidation over the period 2000–2009 (tropopause height at 200 hPa), which is slightly higher than the one from the ACCMIP models (528 [454–617] Tg CH_4 yr^{-1} reported in Kirschke et al. (2013) and Saunois et al. (2016).

3.3.2 Stratospheric loss

In the stratosphere, CH_4 is lost through reactions with excited atomic oxygen $O(^1D)$, atomic chlorine (Cl), atomic fluorine (F), and OH (Brasseur and Solomon, 2005; le Texier et al., 1988). Uncertainties in the chemical loss of stratospheric methane are large, due to uncertain inter-annual variability in stratospheric transport as well as its chemical interactions and feedbacks with stratospheric ozone (Portmann et al., 2012). In particular, the fraction of stratospheric loss

due to the different oxidants is still uncertain, with possibly 20 %–35 % due to halons, about 25 % due to $O(^1D)$ mostly in the high stratosphere, and the rest due to stratospheric OH (McCarthy et al., 2003).

In this study, seven chemistry–climate models from the CCMI project (Table S4) are used to provide estimates of methane chemical loss, including reactions with OH, $O(^1D)$, and Cl; CH_4 photolysis is also included but occurs only above the stratosphere. Considering a 200 hPa tropopause height, the CCMI models suggest an estimate of 31 [12–37] Tg CH_4 yr⁻¹ for the methane stratospheric sink for the period 2000–2010 (Table S4). The 20 Tg difference compared to the mean value reported by Kirschke et al. (2013) and Saunois et al. (2016) for the same period (51 [16–84] Tg CH_4 yr⁻¹) is probably due to the plausible double-counting of $O(^1D)$ and Cl oxidations in our previous calculation, as the chemistry–climate models usually report the total chemical loss of methane (not OH oxidation only).

We report here a climatological range of 12–37 Tg CH_4 yr⁻¹ associated with a mean value of 31 Tg CH_4 yr⁻¹.

3.3.3 Tropospheric reaction with Cl

Halogen atoms can also contribute to the oxidation of methane in the troposphere. Allan et al. (2005) measured mixing ratios of methane and $\delta^{13}C-CH_4$ at two stations in the Southern Hemisphere from 1991 to 2003, and they found that the apparent kinetic isotope effect (KIE) of the atmospheric methane sink was significantly larger than that explained by OH alone. A seasonally varying sink due to atomic chlorine (Cl) in the marine boundary layer of between 13 and 37 Tg CH_4 yr⁻¹ was proposed as the explanatory mechanism (Allan et al., 2007; Platt et al., 2004). This sink was estimated to occur mainly over coastal and marine regions, where NaCl from evaporated droplets of seawater reacts with NO_2 to eventually form Cl_2 , which then UV-dissociates to Cl. However significant production of nitryl chloride ($ClNO_2$) at continental sites has been recently reported (Riedel et al., 2014) and suggests the broader presence of Cl, which in turn would expand the significance of the Cl sink in the troposphere. Recently, using a chemistry-transport model, Hossaini et al. (2016) suggest a chlorine sink in the lower range of Allan et al. (2007), ~ 12 –13 Tg CH_4 yr⁻¹ (about 2.5 % of the tropospheric sink). They also estimate that $ClNO_2$ yields a 1 Tg yr⁻¹ sink of methane. Another modelling study (Wang et al., 2019b) produced a more comprehensive analysis of global tropospheric chlorine chemistry and found a chlorine sink of 5 Tg yr⁻¹, representing only 1 % of the total methane tropospheric sink. Both the KIE approach and chemistry-transport model simulations carry uncertainties (extrapolations based on only a few sites and use of indirect measurements, for the former; missing sources, coarse resolution, underestimation of some anthropogenic sources for the latter). However, Gromov et

al. (2018) found that chlorine can contribute only 0.23 % to the tropospheric sink of methane (about 1 Tg CH_4 yr⁻¹) in order to balance the global $^{13}C(CO)$ budget.

Awaiting further work to better assess the magnitude of the chlorine sink in the methane budget, we suggest a lower estimate but a larger range than in Saunois et al. (2016) and used the following climatological value for the 2000s: 11 [1–35] Tg CH_4 yr⁻¹.

3.3.4 Soil uptake

Unsaturated oxic soils are sinks of atmospheric methane due to the presence of methanotrophic bacteria, which consume methane as a source of energy. Dutaur and Verchot (2007) conducted a comprehensive meta-analysis of field measurements of CH_4 uptake spanning a variety of ecosystems. Extrapolating to the global scale, they reported a range of 36 ± 23 Tg CH_4 yr⁻¹, but they also showed that stratifying the results by climatic zone, ecosystem, and soil type led to a narrower range (and lower mean estimate) of 22 ± 12 Tg CH_4 yr⁻¹. Modelling studies, employing meteorological data as external forcing, have also produced a considerable range of estimates. Using a soil depth-averaged formulation based on Fick's law with parameterizations for diffusion and biological oxidation of CH_4 , Ridgwell et al. (1999) estimated the global sink strength at 38 Tg CH_4 yr⁻¹, with a range of 20–51 Tg CH_4 yr⁻¹ reflecting the model structural uncertainty in the base oxidation parameter. Curry (2007) improved on the latter by employing an exact solution of the one-dimensional diffusion-reaction equation in the near-surface soil layer (i.e., exponential decrease in CH_4 concentration below the surface), a land surface hydrology model, and calibration of the oxidation rate to field measurements. This resulted in a global estimate of 28 Tg CH_4 yr⁻¹ (9–47 Tg CH_4 yr⁻¹), the result reported by Zhuang et al. (2013), Kirschke et al. (2013), and Saunois et al. (2016). Ito and Inatomi (2012) used an ensemble methodology to explore the variation in estimates produced by these parameterizations and others, which spanned the range 25–35 Tg CH_4 yr⁻¹. Murguia-Flores et al. (2018) further refined the Curry (2007) model's structural and parametric representations of key drivers of soil methanotrophy, demonstrating good agreement with the observed latitudinal distribution of soil uptake (Dutaur and Verchot, 2007). Their model simulated a methane soil sink of 32 Tg CH_4 yr⁻¹ for the period 2000–2017 (Fig. 4), compared to 38 and 29 Tg CH_4 yr⁻¹ using the Ridgwell et al. (1999) and Curry (2007) parameterizations, respectively, under the same meteorological forcing. As part of a more comprehensive model accounting for a range of methane sources and sinks, Tian et al. (2010, 2015, 2016) computed vertically averaged CH_4 soil uptake including the additional mechanisms of aqueous diffusion and plant-mediated (*aerenchyma*) transport, arriving at the estimate 30 ± 19 Tg CH_4 yr⁻¹ (Tian et al., 2016). The still more comprehensive biogeochemical model of Riley et al. (2011)

included vertically resolved representations of the same processes considered by Tian et al. (2016), in addition to grid cell fractional inundation and, importantly, the joint limitation of uptake by both CH₄ and O₂ availability in the soil column. Riley et al. (2011) estimated a global CH₄ soil sink of 31 Tg CH₄ yr⁻¹ with a structural uncertainty of 15–38 Tg CH₄ yr⁻¹ (a higher upper limit resulted from an elevated gas diffusivity to mimic convective transport; as this is not usually considered, we adopt the lower upper bound associated with no limitation of uptake at low soil moisture). A model of this degree of complexity is required to explicitly simulate situations where the soil water content increases enough to inhibit the diffusion of oxygen, and the soil becomes a methane source (Lohila et al., 2016). This transition can be rapid, thus creating areas (for example, seasonal wetlands) that can be either a source or a sink of methane depending on the season.

The previous Curry (2007) estimate can be revised upward based on subsequent work and the increase in CH₄ concentration since that time, which gives a central estimate of 30.1 Tg CH₄ yr⁻¹. Considering structural uncertainty in the various models' assumptions and parameters, we report here the median and range of Tian et al. (2016): 30 [11–49] Tg CH₄ yr⁻¹ for the periods 2000–2009 and 2008–2017.

3.3.5 CH₄ lifetime

The atmospheric lifetime of a given gas in steady state may be defined as the global atmospheric burden (Tg) divided by the total sink (Tg yr⁻¹) (IPCC, 2001). Global models provide an estimate of the loss of the gas due to individual sinks, which can then be used to derive lifetime due to a specific sink. For example, methane's tropospheric lifetime is determined as global atmospheric methane burden divided by the loss from OH oxidation in the troposphere, sometimes called “chemical lifetime”. Methane total lifetime corresponds to the global burden divided by the total loss including tropospheric loss from OH oxidation, stratospheric chemistry, and soil uptake. The CCMI models (described in Morgenstein et al., 2017) estimate the tropospheric methane lifetime at about 9 years (average over years 2000–2009), with a range of 7.2–10.1 years (see Table S4). While this range agrees with previous values found in ACCMIP (9.3 [7.1–10.6] years; Voulgarakis et al., 2013), the mean value reported here is lower than previously reported, probably due to a smaller and different ensemble of climate models. Adding 30 Tg to account for the soil uptake to the total chemical loss of the CCMI models, we derive a total methane lifetime of 7.8 years (average over 2000–2009 with a range of 6.5–8.8 years). These updated model estimates of total methane lifetime agree with the previous estimates from ACCMIP (8.2 [6.4–9.2] years for the year 2000; Voulgarakis et al., 2013). Reducing the large spread in methane lifetime (between models and between models and observation-based estimates) would (1) bring an

improved constraint on global total methane emissions and (2) ensure an accurate forecast of future climate.

4 Atmospheric observations and top-down inversions

4.1 Atmospheric observations

Systematic atmospheric CH₄ observations began in 1978 (Blake et al., 1982) with infrequent measurements from discrete air samples collected in the Pacific at a range of latitudes from 67° N to 53° S. Because most of these air samples were from well-mixed oceanic air masses and the measurement technique was precise and accurate, they were sufficient to establish an increasing trend and the first indication of the latitudinal gradient of methane. Spatial and temporal coverage was greatly improved soon after (Blake and Rowland, 1986) with the addition of the Earth System Research Laboratory from US National Oceanic and Atmospheric Administration (NOAA/ESRL) flask network (Steele et al., 1987, Fig. 1) and of the Advanced Global Atmospheric Gases Experiment (AGAGE) (Cunnold et al., 2002; Prinn et al., 2000), the Commonwealth Scientific and Industrial Research Organisation (CSIRO; Francey et al., 1999), the University of California Irvine (UCI; Simpson et al., 2012), and in situ and flask measurements from regional networks, such as the ICOS (Integrated Carbon Observation System) network in Europe (INGOS, 2018; ICOS-RI, 2019; <https://www.icos-ri.eu/>, last access: 29 June 2020). The combined datasets provide the longest time series of globally averaged CH₄ abundance. Since the early 2000s, CH₄ column-averaged mole fractions have been retrieved through passive remote sensing from space (Buchwitz et al., 2005a, b; Butz et al., 2011; Crevoisier et al., 2009; Frankenberg et al., 2005; Hu et al., 2018). Ground-based Fourier transform infrared (FTIR) measurements at fixed locations also provide time-resolved methane column observations during daylight hours and a validation dataset against which to evaluate the satellite measurements such as the TCCON network (e.g. Pollard et al., 2017; Wunch et al., 2011) or the Network for Detection of Atmospheric Composition Change (NDACC) (e.g. Bader et al., 2017).

In this budget, in situ observations from the different networks were used in the top-down atmospheric inversions to estimate methane sources and sinks over the period 2000–2017. Satellite observations from the TANSO/FTS instrument on board the satellite GOSAT were used to estimate methane sources and sinks over the period 2009–2017. Other atmospheric data (FTIR, airborne measurements, AirCore, isotopic measurements, etc.) have been used for validation by some groups but not specifically in this study. However, further information is provided in the Supplement, and a more comprehensive validation of the inversions is planned to use some of these data.

4.1.1 In situ CH₄ observations and atmospheric growth rate at the surface

We use globally averaged CH₄ mole fractions at the Earth's surface from the four observational networks (NOAA/ESRL, AGAGE, CSIRO, and UCI). The data are archived at the World Data Centre for Greenhouse Gases (WDCGG) of the WMO Global Atmospheric Watch (WMO GAW) programme, including measurements from other sites that are not operated as part of the four networks. The CH₄ in situ monitoring network has grown significantly over the last decade due to the emergence of laser diode spectrometers which are robust and accurate enough to allow deployments with minimal maintenance, enabling the development of denser networks in developed countries (Stanley et al., 2018; Yver Kwok et al., 2015) and new stations in remote environments (Bian et al., 2015; Nisbet et al., 2019).

The networks differ in their sampling strategies, including the frequency of observations, spatial distribution, and methods of calculating globally averaged CH₄ mole fractions. Details are given in the Supplement of Kirschke et al. (2013). The global average values of CH₄ concentrations presented in Fig. 1 are computed using long time series measurements through gas chromatography with flame ionization detection (GC–FID), although chromatographic schemes vary among the labs. Because GC–FID is a relative measurement method, the instrument response must be calibrated against standards. The current WMO reference scale, maintained by NOAA/ESRL, WMO-X2004A (Dlugokencky et al., 2005), was updated in July 2015. NOAA and CSIRO global means are on this scale. AGAGE uses an independent standard scale maintained by Tohoku University (Aoki et al., 1992), but direct comparisons of standards and indirect comparisons of atmospheric measurements show that differences are below 5 ppb (Tans and Zwellberg, 2014; Vardag et al., 2014). UCI uses another independent scale that was established in 1978 and is traceable to NIST (Flores et al., 2015; Simpson et al., 2012) but has not been included in standard exchanges with other networks so differences with the other networks cannot be quantitatively defined. Additional experimental details are presented in the supplementary material from Kirschke et al. (2013) and references therein.

In Fig. 1, (a) globally averaged CH₄ and (b) its growth rate (derivative of the deseasonalized trend curve) through to 2017 are plotted for the four measurement programmes using a procedure of signal decomposition described in Thoning et al. (1989). We define the annual G_{ATM} as the increase in the atmospheric concentrations from 1 January in one year to 1 January in the next year. Agreement among the four networks is good for the global growth rate, especially since ~1990. The large differences observed mainly before 1990 probably reflect the different spatial coverage of each network. The long-term behaviour of globally averaged atmospheric CH₄ shows a decreasing but positive growth rate (defined as the derivative of the deseasonalized mixing ratio)

from the early 1980s through 1998, a near stabilization of CH₄ concentrations from 1999 to 2006, and a renewed period with positive persistent growth rates since 2007, slightly larger after 2014. When a constant atmospheric lifetime is assumed, the decreasing growth rate from 1983 through 2006 may imply that atmospheric CH₄ was approaching steady state, with no trend in emissions. The NOAA global mean CH₄ concentration was fitted with a function that describes the approach to a first-order steady state (ss index): $[\text{CH}_4](t) = [\text{CH}_4]_{\text{ss}} - ([\text{CH}_4]_{\text{ss}} - [\text{CH}_4]_0)e^{-t/\tau}$; solving for the lifetime, τ , gives 9.3 years, which is very close to current literature values (e.g. Prather et al., 2012, 9.1 ± 0.9 years). Such an approach includes uncertainties, especially due to the strong assumption of no trend in emissions and sinks, which does not agree with some studies explaining the stabilization period by decreasing emissions associated with increasing sink (e.g. Bousquet et al., 2006). However, this value seems consistent albeit higher than the chemistry–climate estimates (8.2 years; see Sect. 3.3.5).

From 1999 to 2006, the annual increase in atmospheric CH₄ was remarkably small at 0.6 ± 0.1 ppb yr⁻¹. Since 2006, the atmospheric growth rate has recovered to a level similar to that of the mid-1990s (~5 ppb yr⁻¹), or even to that of the 1980s for 2014 and 2015 (> 10 ppb yr⁻¹). On decadal timescales, the annual increase is on average 2.1 ± 0.3 ppb yr⁻¹ for 2000–2009, 6.6 ± 0.3 ppb yr⁻¹ for 2008–2017, and 6.1 ± 1.0 ppb yr⁻¹ for the year 2017.

4.1.2 Satellite data of column-average CH₄

In this budget, we use satellite data from the JAXA satellite Greenhouse Gases Observing SATellite (GOSAT) launched in January 2009 (Butz et al., 2011; Morino et al., 2011) containing the TANSO-FTS instrument, which observes in the shortwave infrared (SWIR). Different retrievals of methane based on TANSO-FTS GOSAT products are made available to the community: from NIES (Yoshida et al., 2013), from SRON (Schepers et al., 2012), and from the University of Leicester (Parker et al., 2011). The three retrievals are used by the top-down systems (Tables 4 and S6). Although GOSAT retrievals still show significant unexplained biases and limited sampling in cloud-covered regions and in the high-latitude winter, it represents an important improvement compared to the first satellite measuring methane from space, SCIAMACHY (Scanning Imaging Absorption spectrometer for Atmospheric Cartography) for both random and systematic observation errors (see Table S2 of Buchwitz et al., 2017).

Atmospheric inversions based on SCIAMACHY and GOSAT CH₄ retrievals were reported in Saunio et al. (2017). Here, only inversions using GOSAT retrievals are used.

4.2 Top-down inversions used in the budget

An atmospheric inversion is the optimal combination of atmospheric observations, of a model of atmospheric transport and chemistry, of a prior estimate of methane sources and sinks, and of their uncertainties, in order to provide improved estimates of the sources and sinks, and their uncertainty. The theoretical principle of methane inversions is detailed in the Supplement (Sect. S2) and an overview of the different methods applied to methane is presented in Houweling et al. (2017).

We consider here an ensemble of inversions gathering various chemistry-transport models, differing in vertical and horizontal resolutions, meteorological forcing, advection and convection schemes, and boundary layer mixing. Including these different systems is a conservative approach that allows us to cover different potential uncertainties of the inversion, among them model transport, set-up issues, and prior dependency. General characteristics of the inversion systems are provided in Table 4. Further details can be found in the referenced papers and in the Supplement. Each group was asked to provide gridded flux estimates for the period 2000–2017, using either surface or satellite data, but no additional constraints were imposed so that each group could use their preferred inversion set-up. A set of prior emission distributions was built from the most recent inventories or model-based estimates (see Supplement), but its use was not mandatory (Table S6). This approach corresponds to a flux assessment but not to a model inter-comparison as the protocol was not too stringent. Estimating posterior uncertainty is time and computer resource consuming, especially for the 4D-Var approaches and Monte Carlo methods. Posterior uncertainties have been provided by only two groups and are found to be lower than the ensemble spread. Indeed, chemistry-transport models differ in inter-hemispheric transport, stratospheric methane profiles, and OH distribution, which are not fully considered in the individual posterior uncertainty. As a result, we do not use the posterior uncertainties provided by these two groups but report the minimum–maximum range among the different top-down approaches.

Nine atmospheric inversion systems using global Eulerian transport models were used in this study compared to eight in Saunois et al. (2016). Each inversion system provided one or several simulations, including sensitivity tests varying the assimilated observations (surface or satellite) or the inversion set-up. This represents a total of 22 inversion runs with different time coverage: generally 2000–2017 for surface-based observations and 2010–2017 for GOSAT-based inversions (Tables 4 and S6). In poorly observed regions, top-down surface inversions may rely on the prior estimates and bring little or no additional information to constrain (often) spatially overlapping emissions (e.g. in India, China). Also, we recall that many top-down systems solve for the total fluxes at the surface only or for some categories that may differ from the GCP categories. When multiple sensitivity tests

were performed the mean of this ensemble was used not to over-weight one particular inverse system. It should also be noticed that some satellite-based inversions are in fact combined satellite and surface inversions as they use satellite retrievals and surface measurements simultaneously (Alexe et al., 2015; Bergamaschi et al., 2013; Houweling et al., 2014). Nevertheless, these inversions are still referred to as satellite-based inversions.

Each group provided gridded monthly maps of emissions for both their prior and posterior total and for sources per category (see the categories Sect. 2.3). Results are reported in Sect. 5. Atmospheric sinks from the top-down approaches have been provided for this budget and are compared with the values reported in Kirschke et al. (2013). Not all inverse systems report their chemical sink; as a result, the global mass imbalance for the top-down budget is derived as the difference between total sources and total sinks for each model when both fluxes were reported.

5 Methane budget: top-down and bottom-up comparison

5.1 Global methane budget

5.1.1 Global total methane emissions

At the global scale, the total emissions inferred by the ensemble of 22 inversions is $576 \text{ Tg CH}_4 \text{ yr}^{-1}$ [550–594] for the 2008–2017 decade (Table 3), with the highest ensemble mean emission of $596 \text{ Tg CH}_4 \text{ yr}^{-1}$ [572–614] for 2017. Global emissions for 2000–2009 ($547 \text{ Tg CH}_4 \text{ yr}^{-1}$) are consistent with Saunois et al. (2016) and the range for global emissions; $524\text{--}560 \text{ Tg CH}_4 \text{ yr}^{-1}$ is in line with the range in Saunois et al. (2016) (535–569), although the ensemble of inverse systems contributing to this budget is different than for Saunois et al. (2016). Indeed, only six inverse systems of the nine examined here (Table S7) contributed previously to the Saunois et al. (2016) budget. The range reported gives the minimum and maximum values among studies and does not reflect the individual full uncertainties. Also, most of the top-down models use the same OH distribution from the TRANSCOM experiment (Patra et al., 2011), leading to a global budget that is quite constrained, probably explaining the rather low range (10 %) compared to bottom-up estimates (see below).

The estimates made via the bottom-up approaches considered here are quite different from the top-down results, with global emissions almost 30 % larger, at $737 \text{ Tg CH}_4 \text{ yr}^{-1}$ [594–881] for 2008–2017 (Table 3). Moreover, the range estimated using bottom-up approaches does not overlap with that of the top-down estimates. The bottom-up estimates are given by the sum of individual anthropogenic and natural processes, without any constraint on the total. For the period 2000–2009, the discrepancy between bottom-up and top-down was 30 % of the top-down estimates in

Table 4. Top-down studies used in our analysis, with their contribution to the decadal and yearly estimates noted. For decadal means, top-down studies have to provide at least 8 years of data over the decade to contribute to the estimate.

Model	Institution	Observation used	Time period	Number of inversions	2000–2009	2008–2017	2017	References
CarbonTracker Europe CH ₄	FMI	Surface stations	2000–2017	1	y	y	y	Tsuruta et al. (2017)
CarbonTracker Europe CH ₄	FMI	GOSAT NIES L2 v2.72	2010–2017	1	n	y	y	Tsuruta et al. (2017)
GELCA	NIES	Surface stations	2000–2015	1	y	y	n	Ishizawa et al. (2016)
LMDz-PYVAR	LSCE/CEA	Surface stations	2010–2016	2	n	y	n	Yin et al. (2015)
LMDz-PYVAR	LSCE/CEA	GOSAT Leicester v7.2	2010–2016	4	n	y	n	Yin et al. (2015)
LMDz-PYVAR	LSCE/CEA	GOSAT Leicester v7.2	2010–2017	2	n	y	y	Zheng et al. (2018b, a)
MIROC4-ACTM	JAMSTEC	Surface stations	2000–2016	1	y	y	n	Patra et al. (2016, 2018)
NICAM-TM	NIES	Surface stations	2000–2017	1	y	y	y	Niwa et al. (2017a, b)
NIES-TM-FLEXPART	NIES	Surface stations	2000–2017	1	y	y	y	Maksyutov et al. (2020); Wang et al. (2019a)
NIES-TM-FLEXPART	NIES	GOSAT NIES L2 v2.72	2010–2017	1	n	y	y	Maksyutov et al. (2020); Wang et al. (2019a)
TM5-CAMS	TNO/VU	Surface stations	2000–2017	1	y	y	y	Bergamaschi et al. (2010, 2013); Pandey et al. (2016); Segers and Houwelling (2018)
TM5-CAMS	TNO/VU	GOSAT ESA/CCI v2.3.8 (combined with surface observations)	2010–2017	1	n	y	y	Bergamaschi et al. (2010, 2013); Pandey et al. (2016); Segers and Houwelling (2018)
TM5-4DVAR	EC-JRC	Surface stations	2000–2017	2	y	y	y	Bergamaschi et al. (2013, 2018)
TM5-4DVAR	EC-JRC	GOSAT OCPR v7.2 (combined with surface observations)	2010–2017	2	n	y	y	Bergamaschi et al. (2013, 2018)
TOMCAT	Uni. of Leeds	Surface stations	2003–2015	1	n	y	n	McNorton et al. (2018)

Saunois et al. (2016) ($167 \text{ Tg CH}_4 \text{ yr}^{-1}$); this has been reduced only slightly (now $156 \text{ Tg CH}_4 \text{ yr}^{-1}$ for the same 2000–2009 period). This reduction is due to (1) a better agreement in the anthropogenic emissions (top-down and bottom-up difference reducing from 19 to $2 \text{ Tg CH}_4 \text{ yr}^{-1}$), (2) a reduction in the estimates of some natural sources other than wetlands based on recent literature (by $7 \text{ Tg CH}_4 \text{ yr}^{-1}$ from geological sources, by $8 \text{ Tg CH}_4 \text{ yr}^{-1}$ from wild animals, and by $3 \text{ Tg CH}_4 \text{ yr}^{-1}$ from allocation of wildfires to biomass & biofuel burning; see Table 3), and (3) a reduction of $35 \text{ Tg CH}_4 \text{ yr}^{-1}$ in the bottom-up estimates of wetland emissions by models when excluding lakes and paddies as wetlands (see Sect. 5.1.2 below). These reductions ($-70 \text{ Tg CH}_4 \text{ yr}^{-1}$) in the bottom-up budget are partially offset by revised freshwater emissions with higher values ($+37 \text{ Tg CH}_4 \text{ yr}^{-1}$) resulting from the integration of a recent study on lake, pond, and reservoir emissions (DelSontro et al., 2018; see Sect. 3.2.2) and the integration of estuary emissions in this budget ($+4 \text{ Tg CH}_4 \text{ yr}^{-1}$). Overall, the uncertainty range of some natural emissions has decreased in this study compared to Kirschke et al. (2013) and Saunois et al. (2016), for example for oceans, termites, wild animals, and geological sources. However, the uncertainty in

the global budget remains high because of the large range reported for emissions from freshwater systems. Still, as noted in Kirschke et al. (2013), such large global emissions from the bottom-up approaches are not consistent with top-down estimates that rely on OH burden constrained by methyl chloroform atmospheric observations, and are very likely overestimated. This overestimation likely results from errors related to up-scaling local measurements and double-counting of some natural sources (e.g. wetlands, other inland water systems; see Sect. 5.1.2).

5.1.2 Global methane emissions per source category

The global methane emissions from natural and anthropogenic sources (see Sect. 2.3) for 2008–2017 are presented in Figs. 5 and 6 and Table 3. Top-down estimates attribute about 60% of total emissions to anthropogenic activities (range of 55%–70%) and 40% to natural emissions. As natural emissions estimated from bottom-up approaches are much larger, the anthropogenic versus natural emission ratio is nearly 1, not consistent with ice core data. A current predominant role of anthropogenic sources of methane emissions is consistent with and strongly supported by avail-

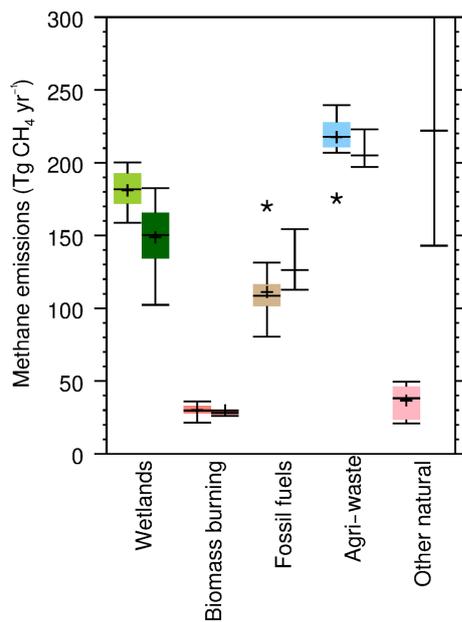


Figure 5. Methane global emissions from the five broad categories (see Sect. 2.3) for the 2008–2017 decade for top-down inversion models (left light coloured box plots, Tg CH₄ yr⁻¹) and for bottom-up models and inventories (right dark coloured box plots). Median value and first and third quartiles are presented in the boxes. The whiskers represent the minimum and maximum values when suspected outliers are removed (see Sect. 2.2). Suspected outliers are marked with stars when they exist. Bottom-up quartiles are not available for bottom-up estimates, except for wetland emissions. Mean values are represented with “+” symbols; these are the values reported in Table 3.

able ice core and atmospheric methane records. These data indicate that atmospheric methane varied around 700 ppb during the last millennium before increasing by a factor of 2.6 to ~1800 ppb since pre-industrial times. Accounting for the decrease in mean lifetime over the industrial period, Prather et al. (2012) estimated from these data a total source of 554 ± 56 Tg CH₄ in 2010 of which about 64 % (352 ± 45 Tg CH₄) was of anthropogenic origin, consistent with the range in our top-down estimates.

For wetlands in 2008–2017, the top-down and bottom-up estimates of 181 Tg CH₄ yr⁻¹ (range 159–200) and 149 Tg CH₄ yr⁻¹ (range 102–182), respectively, are statistically consistent. Bottom-up mean wetland emissions for the 2000–2009 period are smaller in this study than those of Saunois et al. (2016). Conversely, the current 2000–2009 mean top-down wetland estimates are larger than those of Saunois et al. (2016) (Table 3). The reduction in wetland emissions from bottom-up models is related to an updated wetland extent dataset (WAD2M; see Sect. 3.2.1). Top-down wetland emissions estimates are higher on average but the range is reduced by 50 % compared to Saunois et al. (2016) for 2000–2009. In the bottom-up estimates, the amplitude of the range of emissions of 102–179 is similar

to that in Saunois et al. (2016) (151–222 for 2000–2009), and narrowed by a third compared to the previous estimates from Melton et al. (2013) (141–264) and from Kirschke et al. (2013) (177–284). Here and in Saunois et al. (2016), the land surface models were forced with the same wetland extent and climate forcing (see Sect. 3.2.1) contrary to Melton et al. (2013) and Kirschke et al. (2013). This suggests that differences in wetland extent explain about a third (30 %–40 %) of the former range of the emission estimates of global natural wetlands. The remaining range is due to differences in model structures and parameters. Bottom-up and top-down estimates for wetland emissions differ more in this study (~ 30 Tg yr⁻¹ for the mean) than in Saunois et al. (2016) (~ 17 Tg yr⁻¹), due to reduced estimates from the bottom-up models and increased estimates from the top-down models. Natural emissions from freshwater systems are not included in the prior fluxes entering the top-down approaches. However, emissions from these non-wetland systems may be accounted for in the posterior estimates of the top-down models, as these two sources are close and probably overlap at the rather coarse resolution of the top-down models. In the top-down budget, natural wetlands represent 30 % on average of the total methane emissions but only 22 % in the bottom-up budget (because of higher total emissions inferred). Neither bottom-up nor top-down approaches included in this study point to significant changes in wetland emissions between the two decades 2000–2009 and 2008–2017 at the global scale.

For other natural emissions, the discrepancy between top-down and bottom-up budgets is the largest for the natural emission total, which is 371 Tg CH₄ yr⁻¹ [245–488] for bottom-up and only 218 Tg CH₄ yr⁻¹ [183–248] for top-down over the 2008–2017 decade. This discrepancy comes from the estimates in “other natural” emissions (freshwater systems, geological sources, termites, oceans, and permafrost). Indeed, for the 2008–2017 decade, top-down inversions infer non-wetland emissions of 37 Tg CH₄ yr⁻¹ [21–50], whereas the sum of the individual bottom-up emissions is 222 Tg CH₄ yr⁻¹ [143–306]. Atmospheric inversions infer about the same amount over the decade 2000–2009 as over 2008–2017, which is almost half of the value reported in Saunois et al. (2016) (68 [21–130] Tg CH₄ yr⁻¹). This reduction is due to either (1) a more consistent way of considering other natural emissions in the various inverse systems or (2) a difference in the ensemble of top-down inversions reported here. It is worth noting that lacking gridded products to use in their prior scenarios, most of the top-down models include only ocean and termite emissions in their prior scenarios. Some of them now include geological sources, but none include freshwater or permafrost emissions in their prior fluxes and thus in their posterior estimates. Regarding the bottom-up budget, the two main contributors to the larger bottom-up total are freshwaters (~ 75 %) and geological emissions (~ 15 %), both of which have large uncertainties and lack of spatially explicit representation with gridded products available to date, for freshwaters for exam-

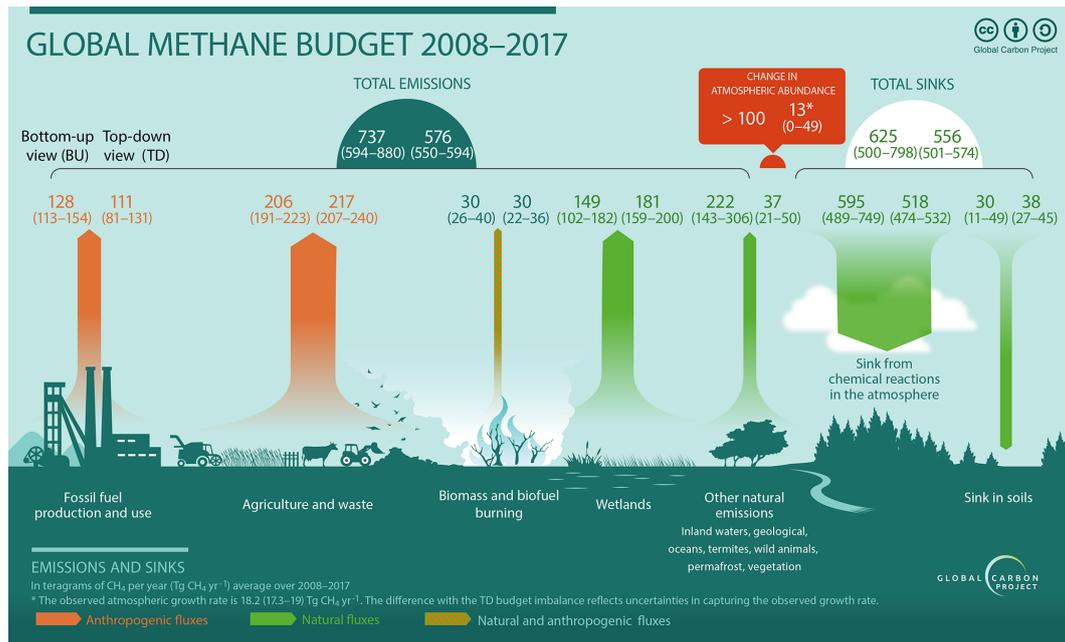


Figure 6. Global methane budget for the 2008–2017 decade. Both bottom-up (left) and top-down (right) estimates (Tg CH₄ yr⁻¹) are provided for each emission and sink category, as well as for total emissions and total sinks. Biomass and biofuel burning emissions are depicted here as both natural and anthropogenic emissions while they are fully included in anthropogenic emissions in the budget tables and text (Sect. 3.1.5).

ple. Because of the discrepancy, the category other natural represents 7 % of total emissions in the top-down budget but up to 25 % in the bottom-up budget.

Geological emissions are associated with relatively large uncertainties, and marine seepage emissions are still widely debated (Thornton et al., 2020). However, summing up all bottom-up fossil-CH₄-related sources (including the anthropogenic emissions) leads to a total of 173 Tg CH₄ yr⁻¹ [131–219] in 2008–2017, which is about 30 % of the top-down global methane emissions and 23 % of the bottom-up total global estimate. These results agree with the value inferred from ¹⁴C atmospheric isotopic analyses of 30 % contribution of fossil CH₄ to global emissions (Etiope et al., 2008; Lassey et al., 2007b). This total fossil fuel emissions from bottom-up approaches agree well with the ¹³C-based estimate of Schwietzke et al. (2016) at 192 ± 32 Tg CH₄ yr⁻¹. Uncertainties on bottom-up estimates of natural emissions lead to probably overestimated total methane emissions resulting in a lower contribution compared to Lassey et al. (2007b). All non-geological and non-freshwater land source categories (wild animals, termites, permafrost) have been estimated to be lower than in Kirschke et al. (2013) and Saunois et al. (2016) and contribute only 13 Tg CH₄ yr⁻¹ [4–19] to global emissions. From a top-down point of view, the sum of all the natural sources is more robust than the partitioning between wetlands and other natural sources. Better constraining the partitioning of methane emissions between wetlands and freshwater systems, including emissions from thawing permafrost,

may be the key to reconcile the top-down and bottom-up budget on natural sources. Also, including all known spatio-temporal distributions of natural emissions in top-down prior fluxes would be a step forward to consistently compare natural versus anthropogenic total emissions between top-down and bottom-up approaches.

Total anthropogenic emissions for the period 2008–2017 were assessed to be statistically consistent between top-down (359 Tg CH₄ yr⁻¹, range 336–376) and bottom-up approaches (366 Tg CH₄ yr⁻¹, range 349–393). The partitioning of anthropogenic emissions between agriculture and waste, fossil fuel extraction and use, and biomass and biofuel burning also shows good consistency between top-down and bottom-up approaches, though top-down approaches suggest less fossil fuel and more agriculture and waste emissions than bottom-up estimates (Table 3 and Figs. 5 and 6). For 2008–2017, agriculture and waste contributed an estimated 217 Tg CH₄ yr⁻¹ [207–240] for the top-down budget and 206 Tg CH₄ yr⁻¹ [191–223] for the bottom-up budget. Fossil fuel emissions contributed 111 Tg CH₄ yr⁻¹ [81–131] for the top-down budget and 128 Tg CH₄ yr⁻¹ [113–154] for the bottom-up budget. Biomass and biofuel burning contributed 30 Tg CH₄ yr⁻¹ [22–36] for the top-down budget and 30 Tg CH₄ yr⁻¹ [26–40] for the bottom-up budget. Biofuel methane emissions rely on very few estimates currently (Wuebbles and Hayhoe, 2002). Although biofuel is a small source globally (~ 12 Tg CH₄ yr⁻¹), more estimates are needed to allow a proper uncertainty assessment. Overall

for top-down inversions the global fraction of total emissions for the different source categories is 38 % for agriculture and waste, 19 % for fossil fuels, and 5 % for biomass and bio-fuel burning. With the exception of biofuel emissions, the uncertainty associated with global anthropogenic emissions appears to be smaller than that of natural sources but with an asymmetric uncertainty distribution (mean significantly different than median). The relative agreement between top-down and bottom-up approaches may indicate the limited capability of the inversion to separate the emissions and should therefore be treated with caution. Indeed, in poorly observed regions, top-down inversions rely on the prior estimates and bring little or no additional information to constrain (often) spatially overlapping emissions (e.g. in India, China). Also, as many top-down systems solve for the total fluxes at the surface or for some categories that may differ from the GCP categories, their posterior partitioning relies on the prior ratio between categories that are prescribed using bottom-up inventories.

5.1.3 Global budget of total methane sinks

For top-down estimates, the CH₄ chemical removal from the atmosphere is estimated at 518 Tg CH₄ yr⁻¹ over the period 2008–2017, with an uncertainty of about ±5 % (range 474–532 Tg CH₄ yr⁻¹). All the inverse models account for CH₄ oxidation by OH and O(¹D), and some include stratospheric chlorine oxidation (Table S6). In addition, most of the top-down models use OH distributions from the TRANSCOM experiment (Patra et al., 2011), probably explaining the rather low range of estimates compared to bottom-up estimates (see below). Differences between transport models affect the chemical removal of CH₄, leading to different chemical loss rates, even with the same OH distribution. However, uncertainties in the OH distribution and magnitude (Zhao et al., 2019) are not considered in our study, while it could contribute to a significant change in the chemical sink and then in the derived posterior emissions through the inverse process (Zhao et al., 2020). The chemical sink represents more than 90 % of the total sink, the rest being attributable to soil uptake (38 [27–45] Tg CH₄ yr⁻¹). Half of the top-down models use the climatological soil uptake magnitude (37–38 Tg CH₄ yr⁻¹) and distribution from Ridgwell et al. (1999), while half of the models use an estimate from the biogeochemical model VISIT (Ito and Inatomi, 2012), which calculates varying uptake between 31 and 38 Tg CH₄ yr⁻¹ over the 2000–2017 period. These sink estimates used as the prior estimate in the inversions are generally higher than the mean estimate of the soil sink calculated by bottom-up models (30 Tg CH₄ yr⁻¹, Sect. 3.3.4).

For the bottom-up estimates, the total chemical loss for the 2000s reported here is 595 Tg CH₄ yr⁻¹ with an uncertainty of 22 % (~ 130 Tg CH₄ yr⁻¹). Differences in chemical schemes (especially in the stratosphere) and in the volatile

organic compound treatment probably explain most of the discrepancies among models (Zhao et al., 2019).

5.2 Latitudinal methane budget

5.2.1 Latitudinal budget of total methane emissions

The latitudinal breakdown of emissions inferred from atmospheric inversions reveals a dominance of tropical emissions at 368 Tg CH₄ yr⁻¹ [337–399], representing 64 % of the global total (Table 5). A total of 32 % of the emissions are from the mid-latitudes (186 Tg CH₄ yr⁻¹ [166–204]) and 4 % are from high latitudes (above 60° N). The ranges around the mean latitudinal emissions are larger than for the global methane sources. While the top-down uncertainty is about ±5 % at the global scale, it increases to ±10 % for the tropics and the northern mid-latitudes to more than ±25 % in the northern high latitudes (for 2008–2017, Table 5). Both top-down and bottom-up approaches consistently show that methane decadal emissions have increased by about 20 Tg CH₄ yr⁻¹ in the tropics and by 7–18 Tg CH₄ yr⁻¹ in the northern mid-latitudes between 2000–2009 and 2008–2017, but not in the northern high latitudes.

Over 2010–2017, at the global scale, satellite-based inversions infer almost identical emissions to ground-based inversions (difference of 3 [0–7] Tg CH₄ yr⁻¹), when consistently comparing surface versus satellite-based inversions for each system. This difference is much lower than the range derived between the different systems (range of 20 Tg CH₄ yr⁻¹ using surface- or satellite-based inversions). This result reflects that differences in atmospheric transport among the systems probably have more impact than the types of observations assimilated on the estimated global emissions. In Saunois et al. (2016), satellite-based inversions reported 12 Tg higher global methane emissions compared to surface-based inversions. Differences in the ensemble, the use of only GOSAT data and the treatment of satellite data within each system compared to Saunois et al. (2016) explain the contrasting results.

As expected, the regional distributions of inferred emissions differ depending on the nature of the observations used (satellite or surface). The largest differences (satellite-based minus surface-based inversions) are observed over the tropical region, between -13 and +26 Tg CH₄ yr⁻¹ (90° S to 30° N), and the northern mid-latitudes (between -20 and +15 Tg CH₄ yr⁻¹). Satellite data provide stronger constraints on fluxes in tropical regions than surface data, due to a much larger spatial coverage. It is therefore not surprising that differences between these two types of observations are found in the tropical band, and consequently in the northern mid-latitudes to balance total emissions, thus affecting the north–south gradient of emissions. However, the regional patterns of these differences are not consistent through the different inverse systems. Indeed, some systems found higher emissions in the tropics when using GOSAT instead of sur-

Table 5. Global and latitudinal total methane emissions ($\text{Tg CH}_4 \text{ yr}^{-1}$), as decadal means (2000–2009 and 2008–2017) and for the year 2017, for this work using bottom-up and top-down approaches. Global emissions for 2000–2009 are also compared with Saunois et al. (2016) and Kirschke et al. (2013) for top-down and bottom-up approaches. Latitudinal total emissions for 2000–2009 are compared with Saunois et al. (2016) for top-down studies only. Uncertainties are reported as the [min–max] range. Differences of $1 \text{ Tg CH}_4 \text{ yr}^{-1}$ in the totals can occur due to rounding errors.

Period	2000–2009		2008–2017		2017	
	Bottom-up	Top-down	Bottom-up	Top-down	Bottom-up	Top-down
Global						
This work	703 [566–842]	547 [524–560]	737 [594–881]	576 [550–594]	747 [602–896]	596 [572–614]
Saunois et al. (2016)	719 [583–861]	552 [535–566]	–	–	–	–
Kirschke et al. (2013)	678 [542–852]	553 [526–569]	–	–	–	–
90° S–30° N						
This work	408 [322–532]	346 [320–379]	430 [338–547]	368 [337–399]	434 [343–568]	383 [351–405]
Saunois et al. (2016)	–	356 [334–381]	–	–	–	–
30–60° N						
This work	252 [202–342]	178 [159–199]	267 [218–349]	186 [166–204]	272 [223–351]	188 [171–209]
Saunois et al. (2016)	–	176 [159–195]	–	–	–	–
60–90° N						
This work	42 [28–70]	23 [17–32]	43 [26–72]	22 [17–29]	40 [24–70]	24 [21–28]
Saunois et al. (2016)	–	20 [15–25]	–	–	–	–

face observations, while others found the opposite. This difference between the systems may depend on whether or not a bias correction is applied to the satellite data based on surface observations, and also on the modelled horizontal and vertical model transports, in the troposphere and in the stratosphere.

5.2.2 Latitudinal methane emissions per source category

The analysis of the latitudinal methane budget per source category (Fig. 7) can be performed for both bottom-up and top-down approaches but with limitations. On the bottom-up side, some natural emissions are not (yet) available at regional scale (mainly inland waters). Therefore, for freshwater emissions, we applied the latitudinal distribution of Bastviken et al. (2011) to the global reported value. Further details are provided in the Supplement to explain how the different bottom-up sources were handled. On the top-down side, as already noted, the partitioning of emissions per source category has to be considered with caution. Indeed, using only atmospheric methane observations to constrain methane emissions makes this partitioning largely dependent on prior emissions. However, differences in spatial patterns and seasonality of emissions can be utilized to constrain emissions from different categories by atmospheric methane observations (for those inversions solving for different sources categories, see Sect. 2.3).

Agriculture and waste are the largest sources of methane emissions in the tropics ($130 [121–137] \text{ Tg CH}_4 \text{ yr}^{-1}$ for the bottom-up budget and $139 [127–157] \text{ Tg CH}_4 \text{ yr}^{-1}$ for the top-down budget, about 38 % of total methane emissions in this region). However, wetland emissions are nearly as large with $116 [71–146] \text{ Tg CH}_4 \text{ yr}^{-1}$ for the bottom-up budget and $135 [116–155] \text{ Tg CH}_4 \text{ yr}^{-1}$ for the top-down budget. One top-down model suggests lower emissions from agriculture and waste compared to the ensemble but suggests higher emissions from fossil fuel: this recalls the necessary caution when discussing sectorial partitioning when using top-down inversions. Anthropogenic emissions dominate in the northern mid-latitudes, with the highest contribution from agriculture and waste emissions (42 % of total emissions), closely followed by fossil fuel emissions (31 % of total emissions). Boreal regions are largely dominated by wetland emissions (60 % of total emissions).

The uncertainty for wetland emissions is larger in the bottom-up models than in the top-down models, while uncertainty in anthropogenic emissions is larger in the top-down models than in the inventories. The large uncertainty in tropical wetland emissions (65 %) results from large regional differences between the bottom-up land surface models. Although they are using the same wetland extent, their responses in terms of flux density show different sensitivities to temperature, water vapour pressure, precipitation, and radiation.

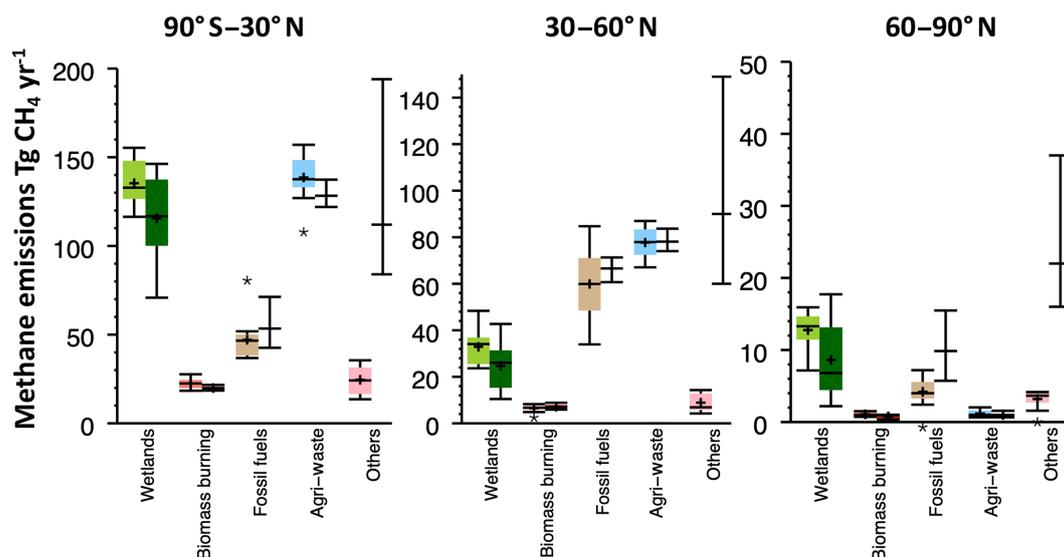


Figure 7. Methane latitudinal emissions from the five broad categories (see Sect. 2.3) for the 2008–2017 decade for top-down inversions models (left light-coloured box plots, $\text{Tg CH}_4 \text{ yr}^{-1}$) and for bottom-up models and inventories (right dark-coloured box plots). Median value and first and third quartiles are presented in the boxes. The whiskers represent the minimum and maximum values when suspected outliers are removed (see Sect. 2.2). Suspected outliers are marked with stars as shown. Bottom-up quartiles are not available for bottom-up estimates, except wetland emissions. Mean values are represented with “+” symbols; these are the values reported in Table 6.

Table 6. Latitudinal methane emissions in teragrams of CH_4 per year for the last decade 2008–2017, based on top-down and bottom-up approaches. Uncertainties are reported as the [min–max] range of reported studies. Differences of $1 \text{ Tg CH}_4 \text{ yr}^{-1}$ in the totals can occur due to rounding errors. Bottom-up anthropogenic estimates are based only on the gridded products from EDGARv4.3.2, GAINS, and CEDS.

Latitudinal band	90° S–30° N		30–60° N		60–90° N	
	Bottom-up	Top-down	Bottom-up	Top-down	Bottom-up	Top-down
Natural sources	228 [155–340]	160 [130–189]	115 [70–192]	42 [29–54]	31 [18–55]	16 [11–20]
Natural wetland	116 [71–146]	135 [116–155]	25 [10–43]	33 [24–48]	9 [2–18]	13 [7–16]
Other natural	112 [84–194]	25 [14–36]	90 [60–149]	9 [4–14]	22 [16–37]	3 [2–4]
Anthropogenic sources	202 [183–217]	208 [186–229]	152 [148–157]	144 [117–170]	12 [8–8]	6 [2–10]
Agriculture & waste	130 [121–137]	139 [127–157]	80 [77–84]	78 [67–87]	1 [1–1]	1 [1–2]
Fossil fuels	53 [43–71]	47 [37–52]	67 [61–71]	60 [34–85]	10 [6–15]	4 [2–7]
Biomass & biofuel burning	20 [18–22]	22 [18–28]	7 [6–9]	6 [5–8]	1 [0–1]	1 [1–1]
Sum of sources	430 [338–557]	368 [337–399]	267 [218–349]	186 [166–204]	43 [26–72]	22 [17–29]

More regional discussions were developed in Saunois et al. (2016) and have been updated in Stavert et al. (2020).

6 Future developments, missing elements, and remaining uncertainties

In this budget, uncertainties on sources and sinks estimated by bottom-up or top-down approaches have been highlighted as well as discrepancies between the two budgets. Limitations of the different approaches have also been highlighted. Four shortcomings of the methane budget were already identified in Kirschke et al. (2013) and Saunois et al. (2016). Although progress has been made, they are still relevant, and actions are needed. However, these actions fall into differ-

ent timescales and parties. In the following, we revisit the four shortcomings, or axis of research, of the current methane budget: how each weakness has been corrected since Saunois et al. (2016), followed by a list of recommendations, from higher to lower priority, associated with the involved parties.

1. *Towards reducing the high uncertainty in the amount of methane emitted by wetland and inland water systems and reduced double-counting issue.*

The remaining large uncertainties strongly suggest the need to develop more studies integrating the different systems (wetlands, ponds, lakes, reservoirs, streams, rivers, estuaries, and marine systems), to avoid double-counting issues, to associate proper emissions with each category, but

also to account for lateral fluxes. Since Saunio et al. (2016), several workshops (e.g. Turner et al., 2019) and publications (e.g. Knox et al., 2019; Thornton et al., 2016a) contributed to implement previous recommendations and strategies to reduce uncertainties of methane emissions due to wetlands and other freshwater systems. One achievement is the reduced estimate (by $\sim 20\%$, i.e. $35 \text{ Tg CH}_4 \text{ yr}^{-1}$) of the global wetland emissions, due to a refined wetland extent analysis and modifications of land surface model calibration.

Methodology changes that could be integrated into the next methane budget releases include

- calibrating land surface models independently from top-down estimates,
- evaluating land surface models against in situ observations such as FLUXNET-CH₄ (Knox et al., 2019), and
- using different wetland extent products to infer wetland emissions (e.g. WAD2M, GIEMS-2; Prigent et al., 2020).

Next steps, in the short term, for modelling, can be addressed by the land biogeochemistry community.

- Finalize a global high-resolution (typically tens of metres) classification of saturated soils and inundated surfaces based on satellite data (visible and microwave), surface inventories, and expert knowledge. This improved area distribution will prevent double-counting between wetlands and other freshwater systems, when used by land surface models.
- Finalize ongoing efforts to develop process-based modelling approaches to estimate freshwater methane emissions, including lateral fluxes, and avoiding upscaling issues, as recently done by Maavara et al. (2019) for N₂O.
- Use the collected flux measurements within the FLUXNET-CH₄ activity (Knox et al., 2019) to provide global flux maps based on machine learning approaches (Peltola et al., 2019).

Over the long run, developing measurement systems will help to improve estimates of wetland and inland water sources, and further reduce uncertainties.

- More systematic measurements from sites reflecting the diverse lake morphologies will allow us to better understand the short-term biological control on ebullition variability, which remains poorly known (Wik et al., 2014, 2016a).
- Extending monitoring of methane fluxes year round from the different natural sources (wetlands, freshwaters) complemented with environmental meta-data (e.g., soil temperature and moisture, vegetation types, water

temperature, acidity, nutrient concentrations, NPP, soil carbon density) will allow us to enrich the FLUXNET-CH₄ observation dataset and to better constrain methane fluxes and their isotopic signatures in land surface models (Glagolev et al., 2011; Turetsky et al., 2014; Ganesan et al., 2019).

2. Towards a better assessment of uncertainties for global methane sinks in top-down and bottom-up budgets.

The inverse systems used here have the same caveats as described in Saunio et al. (2016) (same OH field, same kind of proxy method to optimize it), leading to quite constrained atmospheric sink and therefore total global methane sources. Although we have used a state-of-the-art ensemble of chemistry-transport models (CTMs) and climate-chemistry models (CCMs) simulations from the CCMI (Chemistry-Climate Model Initiative, Morgenstern et al., 2017), the uncertainty of derived CH₄ chemical loss from the chemistry-climate models remains at the same (large) level compared to the previous intercomparison project ACCMIP (Lamarque et al., 2013). Nicely et al. (2017) found that the main cause of the large differences in the CTM representation of CH₄ lifetime is variations in the chemical mechanisms implemented in the models. Using the ensemble of CTMs and CCMs from the CCMI experiment, Zhao et al. (2019) quantified the range of CH₄ loss induced by the ensemble of OH fields to be equivalent up to about half of the discrepancies between CH₄ observations and simulations as forced by the current anthropogenic inventories. These results emphasize the need to first assess, and then improve, atmospheric transport and chemistry models, especially vertically, and to integrate robust representation of OH fields in atmospheric models.

Methodology changes that could be integrated into the next methane budget include

- integrating sensitivity tests on the prior fluxes (use of updated fluxes for natural sources, soil uptake) and
- integrating sensitivity tests on chemical sinks (different OH fields, including inter-annual variability).

Next steps, in the short term, could include developments by the atmospheric modelling community.

- Assess the impact of using updated and varying soil uptake estimates, especially considering a warmer climate (Ni and Groffman, 2018). Indeed, for top-down models resolving for the net flux of CH₄ at the surface integrating a larger estimate of soil uptake would allow larger emissions and then reduce the uncertainty with the bottom-up estimates of total CH₄ sources.
- Further study the reactivity of the air parcels in the chemistry-climate models and define new diagnostics to assess modelled CH₄ lifetimes.

- Develop robust representation of 3D OH fields to be used in the inverse models: based on chemistry–climate models and using correction from measurements, on multispecies assimilating systems (e.g. Gaubert et al., 2017; Miyazaki et al., 2015), or on a simple parametrization applied at grid scales.
- Integrate the aforementioned different potential OH chemical fields, also including inter-annual variability, to assess the impact on the methane budget following Zhao et al. (2020).

Over the long run, other parameters should be (better) integrated into top-down approaches, among them

- the magnitude of the CH₄ loss through oxidation by tropospheric chlorine, a process debated in the recent literature. More modelling (Thanwerdas et al., 2019) and instrumental studies should be devoted to reducing the uncertainty of this potential additional sink before integrating it in top-down models.

3. Towards a better partitioning of methane sources and sinks by region and process using top-down models.

In this work, we report inversions assimilating satellite data from GOSAT, which bring more constraints than provided by surface stations alone, especially over tropical continents. However, we found that satellite- and surface-based inversions and the different inversions systems do not consistently infer the same regional flux distribution.

Methodology changes that could be integrated into the next methane budget releases include the following.

- Integrate GOSAT and GOSAT-2 (launched in October 2018, with expected improved precision and accuracy, JAXA, 2019) for the satellite inversion.
- Investigate the reasons for the regional differences derived by the inverse systems based on the model evaluation and a more detailed questionnaire for the modellers on the treatment of satellite data (bias correction) and stratospheric profiles.

Next steps, in the short term, could integrate developments to be made by the top-down community.

- Evaluate the benefits of using new satellite missions with high spatial resolution and “imaging capabilities” (Crisp et al., 2018) at the global scale, such as the TROPOMI instrument on Sentinel 5P, launched in October 2017 (Hu et al., 2018).
- Integrate the newly available updated gridded products for the different natural sources of CH₄ in their prior fluxes to reach a full spatial description of sources and sinks and to be able to better compare the top-down budget with the bottom-up budget.

- Release more regular updates and intercomparison of emission inventories in order to improve prior scenarios of inverse studies and reduce the need for extending them beyond their available coverage.

- Develop a 4D variational inversion system using isotopic and/or co-emitted species in the top-down budget. Indeed methane isotopes can provide additional constraints to partition the different CH₄ sources and sinks, if isotopic signatures can be better known spatially and temporally (Ganesan et al., 2018). Radiocarbon can help for fossil and non-fossil emissions (Lassey et al., 2007b, a; Petrenko et al., 2017), ¹³CH₄ and CH₃D for biogenic–pyrogenic–thermogenic emissions and OH loss (Röckmann et al., 2011), and emerging clumped isotope measurements for biogenic–thermogenic emissions (Stolper et al., 2014) and OH loss (Haghnegahdar et al., 2017). Also, carbon monoxide (e.g. Fortems-Cheiney et al., 2011) can provide useful constraints for biomass burning emissions and ethane for fugitive emissions (e.g. Simpson et al., 2012; Turner et al., 2019).

- Improve the availability of in situ data for the scientific community, especially ones covering poorly documented regions such as China (Fang et al., 2015), India (Lin et al., 2015; Tiwari and Kumar, 2012) and Siberia (Sasakawa et al., 2010; Winderlich et al., 2010), which have not been included so far in international databases.

Over the long run, integrating more measurements and regional studies will help to improve the top-down systems, and further reduce the uncertainties.

- Integrate global data from future satellite instruments with intrinsic low bias, such as active lidar techniques with MERLIN (Ehret et al., 2017), that are promising to overcome issues of systematic errors (Bousquet et al., 2018) and should provide measurements over the Arctic, contrary to the existing and planned passive missions.
- Extend the CH₄ surface networks to poorly observed regions (e.g. the tropics, China, India, high latitudes) and to the vertical dimension: regular aircraft campaigns (e.g. Paris et al., 2010; Sweeney et al., 2015), AirCore campaigns (e.g. Andersen et al., 2018; Membrive et al., 2017), and TCCON observations (e.g. Wunch et al., 2011, 2019). These observations are still critical to complement satellite data that do not observe well in cloudy regions and at high latitudes and also to evaluate and eventually correct satellite biases (Buchwitz et al., 2017).
- Extend and develop continuous isotopic measurements of methane using laser-based instruments to help partition methane sources and to be integrated in 4D variational isotopic inversions.

- Develop regional components of the CH₄ budget to improve global totals by feeding them with regional top-down and bottom-up approaches; for example, regional inversions using regional measurements and high-resolution models, such as the INGOS project (Bergamaschi et al., 2018b; INGOS, 2018) or the VERIFY project (<https://verify.lsce.ipsl.fr>, last access: 29 June 2020) with the European ICOS network (ICOS-RI, 2019, <https://www.icos-ri.eu/home>, last access: 29 June 2020). The RECCAP-2 project should also provide a scientific framework to further refine greenhouse gas budgets, including methane, at regional scales (<http://cci.esa.int/reccap2>, last access: 29 June 2020).
4. *Towards reducing uncertainties in the modelling of atmospheric transport in the models used in the top-down budget.*

The TRANSCOM experiment synthesized in Patra et al. (2011) showed a large sensitivity of the representation of atmospheric transport to methane concentrations in the atmosphere. In particular, the modelled CH₄ budget appeared to depend strongly on the troposphere–stratosphere exchange rate and thus on the model vertical grid structure and circulation in the lower stratosphere. Also, regional changes in the methane budget depend on the characteristics of the atmospheric transport models used in the inversion (Bruhwiler et al., 2017; Locatelli et al., 2015). This axis of research is demanding important development from the atmospheric modelling community. Waiting for future improvements (finer horizontal and vertical resolutions, more accurate physical parameterization, increase in computing resources, etc.) and assessing atmospheric transport error and the impact on the top-down budget remain crucial.

Methodology changes that could be integrated into the next methane budget releases include

- evaluation of the inversions provided against independent measurements such as regular aircraft campaigns (e.g. Paris et al., 2010; Sweeney et al., 2015), AirCore campaigns (e.g. Andersen et al., 2018; Membrive et al., 2017), and TCCON observations (e.g. Wunch et al., 2011, 2019) and use of this evaluation to weight the different models used in the methane budget.

Next steps, in the short term, could include some development to be addressed by the top-down community to reduce atmospheric transport errors:

- developing further methodologies to extract stratospheric partial column abundances from observations such as TCCON data (Saad et al., 2014; Wang et al., 2014), AirCore, or even ACE-FTS or MIPAS satellite data and using them to replace erroneous simulated stratospheric profiles.

In the long run, developments within atmospheric transport models such as the implementation of hybrid vertical

coordinates (Patra et al., 2018) or of a hexagonal-icosahedric grid with finer resolution (Dubos et al., 2015; Niwa et al., 2017a) and improvements in the simulated boundary layer dynamics are promising to reduce atmospheric transport errors.

7 Data availability

The data presented here are made available in the belief that their dissemination will lead to greater understanding and new scientific insights into the methane budget and changes to it and help to reduce its uncertainties. The free availability of the data does not constitute permission for publication of the data. For research projects, if the data used are essential to the work to be published, or if the conclusion or results largely depend on the data, co-authorship should be considered. Full contact details and information on how to cite the data are given in the accompanying database.

The accompanying database includes one Excel file organized in the following spreadsheets and two NetCDF files defining the regions used to extend the anthropogenic inventories.

The file `Global_Methane_Budget_2000–2017_v2.0.xlsx` includes (1) a summary, (2) the methane observed mixing ratio and growth rate from the four global networks (NOAA, AGAGE, CSIRO and UCI), (3) the evolution of global anthropogenic methane emissions (including biomass burning emissions) used to produce Fig. 2, (4) the global and latitudinal budgets over 2000–2009 based on bottom-up approaches, (5) the global and latitudinal budgets over 2000–2009 based on top-down approaches, (6) the global and latitudinal budgets over 2008–2017 based on bottom-up approaches, (7) the global and latitudinal budgets over 2008–2017 based on top-down approaches, (8) the global and latitudinal budgets for the year 2017 based on bottom-up approaches, (9) the global and latitudinal budgets for the year 2017 based on top-down approaches, and (10) the list of contributors to contact for further information on specific data.

This database is available from ICOS (<https://doi.org/10.18160/GCP-CH4-2019>, Saunois et al., 2020) and the Global Carbon Project (<http://www.globalcarbonproject.org>, last access: 29 June 2020).

8 Conclusions

We have built a global methane budget by using and synthesizing a large ensemble of new and published methods and results using a consistent and transparent approach, including atmospheric observations and inversions (top-down models), process-based models for land surface emissions and atmospheric chemistry, and inventories of anthropogenic emissions (bottom-up models and inventories). For the 2008–2017 decade, global CH₄ emissions are 576 Tg CH₄ yr^{−1} (range of 550–594 Tg CH₄ yr^{−1}), as esti-

mated by top-down inversions. About 60 % of global emissions are anthropogenic (range of 50 %–70 %). Bottom-up models and inventories suggest much larger global emissions (737 Tg CH₄ yr⁻¹ [594–881]) mostly because of larger and more uncertain natural emissions from inland water systems, natural wetlands, and geological leaks and some likely unresolved double-counting of these sources. It is also likely that some of the individual bottom-up emission estimates are too high, leading to larger global emissions from the bottom-up perspective than the atmospheric constraints suggest.

The latitudinal breakdown inferred from top-down approaches reveals a dominant role of tropical emissions (~ 64 %) compared to middle (~ 32 %) and high (~ 4 %) northern latitude (above 60° N) emissions.

Our results, including an extended set of atmospheric inversions, are compared with the previous budget syntheses of Kirschke et al. (2013) and Saunois et al. (2016) and show overall good consistency when comparing the same decade (2000–2009) at the global and latitudinal scales, although estimation methods and reported studies have evolved between the three budgets. While a comparison of top-down emissions estimates determined with and without satellite data agrees well globally, they differ significantly at the latitudinal scale. Most worryingly, these differences were not even consistent in sign, with some models showing notable increases in a given latitudinal flux and others decreases. This suggests that while the inclusion of satellite data may, in the future, significantly increase our ability to attribute fluxes regionally, this is not currently the case due to their existing inherent biases along with the inconsistent application of methods to account for these biases and also differences in model transport, especially in the stratosphere (see recommendations in Sect. 6).

Among the different uncertainties raised in Kirschke et al. (2013), Saunois et al. (2016) estimated that 30 %–40 % of the large range associated with modelled wetland emissions in Kirschke et al. (2013) was due to the estimation of wetland extent. Here, wetland emissions are 35 Tg CH₄ yr⁻¹ smaller than previous estimates due to a refinement of wetland extent. The magnitude and uncertainty of all other natural sources have been revised and updated, leading to smaller emission estimates for oceans, geological sources, and wild animals but higher emission estimates associated with a larger range for non-wetland freshwater systems. This result places a number one priority on reducing uncertainties in emissions from inland water systems by better quantifying the emission factors of each contributing subsystem (streams, rivers, lakes, ponds) and reducing both uncertain up-scaling and likely double-counting with wetland emissions. As a second priority, the uncertainty on the chemical loss of methane needs to be better assessed in both the top-down and the bottom-up budgets. Our work also suggests the need for more interactions among groups developing emission inventories in order to clarify the definition of the sectoral breakdown

in inventories. Such an approach would allow easier comparisons at the subcategory scale.

Building on the improvement of the points detailed in Sect. 6, our aim is to continually update this budget synthesis as a living review paper regularly (~ every 2–3 years). Each update will produce a more recent decadal CH₄ budget, highlight changes in emissions and trends, and incorporate newly available data and model improvements.

In addition to the decadal CH₄ budget presented in this paper, and following former studies (e.g. Bousquet et al., 2006), trends and year-to-year changes in the methane cycle continue to be thoroughly discussed in the recent literature (e.g. Nisbet et al., 2019; Turner et al., 2019). After almost a decade of stagnation in the late 1990s and early 2000s (Dlugokencky et al., 2011; Nisbet et al., 2016), a sustained atmospheric growth rate of more than +5 ppb yr⁻¹ has been observed since 2007, with a further acceleration after 2014 (Nisbet et al., 2019) and several years with a two-digit atmospheric growth as in the 1980s. To date, no consensus has yet been reached in explaining the CH₄ trend since 2007. A likely explanatory scenario, already introduced in Saunois et al. (2017) and further investigated by some other studies since then, includes, by decreasing order of certainty, a positive contribution from microbial and fossil sources (e.g. Nisbet et al., 2019; Schwietzke et al., 2016), a negative contribution from biomass burning emissions before 2014 (Giglio et al., 2013; Worden et al., 2017), a downward revision of Chinese emissions, a negligible role of Arctic emission changes (e.g. Nisbet et al., 2019; Saunois et al., 2017), and a tropical dominance of the increasing emissions (e.g. Saunois et al., 2017). Changes in atmospheric OH concentrations, the largest methane sink, may have contributed to the recent increase in methane concentrations (e.g. Dalsøren et al., 2016; Holmes et al., 2013; McNorton et al., 2016, 2018; Morgenstern et al., 2018; Rigby et al., 2017; Turner et al., 2017), but considerable uncertainty in OH inter-annual variability and trends needs to be further investigated. The challenging and sustained increase in atmospheric CH₄ during the past decade still needs additional research to be fully understood (Nisbet et al., 2019; Turner et al., 2019). The GCP will continue to take its part in analysing and synthesizing recent changes in the global to regional methane cycle based on the ensemble of top-down and bottom-up studies gathered for the budget analysis presented here.

Appendix A

Table A1. Funding supporting the production of the various components of the global methane budget in addition to the authors' supporting institutions (see also Acknowledgements).

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Table A1. Continued.

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NASA: grants NAG5-12669, NNX07AE89G, NNX11AF17G, and NNX16AC98G to MIT	MIT theory and inverse modelling
Swedish National Infrastructure for Computing (SNIC) at the Lund University Centre for Scientific and Technical Computing (Lunarc), project no. 2017/1-423 – Aurora resource	LPJ-GUESS simulations
Support for atmospheric observations	
Australian Antarctic Division	CSIRO flask network
Australian Institute of Marine Science	CSIRO flask network
Bureau of Meteorology (Australia)	Cape Grim AGAGE, CSIRO flask network
Commonwealth Scientific and Industrial Research Organisation (CSIRO, Australia)	Cape Grim AGAGE, CSIRO flask network
Department of the Environment and Energy (DoEE, Australia)	Cape Grim AGAGE
Meteorological Service of Canada	CSIRO flask network
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National Oceanic and Atmospheric Administration (NOAA, United States) contract RA133R15CN0008 to the University of Bristol	Barbados
NOAA United States	CSIRO flask network
Refrigerant Reclaim Australia	Cape Grim AGAGE
UK Department for Business, Energy & Industrial Strategy (BEIS) contract TRN1537/06/2018 to the University of Bristol	Mace Head
ALICE High Performance Computing Facility at the University of Leicester	GOSAT retrievals, Robert Parker
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MS, AS, BP, PB, PeC, and RJ coordinated the global budget. MS, AS, BP, PB, PeC, RJ, SH, PP, and PCi contributed to the update of the full text and all coauthors appended comments. MS, ED, and GP produced the figures. VA, NG, AI, FJ, TK, LL, KMcD, PM, JMe, JMu, CP, SP, WR, HS, HT, WZ, ZZ, QinZ, QiuZ, and QiaZ performed surface land model simulations to compute wetland emissions. DB, MC, PC, SC, KC, GE, GH, KMJ, GL, SN, CP, PRa, Pre, BT, NV, and TW provided datasets useful for natural emission estimates and/or contributed to text on bottom-up natural emissions. LHI, GJM, FT, GvW, and KMC provided anthropogenic datasets and contributed to the text for this section. PP, BP, NC, MI, SM, JMcN, YN, AS, AT, YY, and BZ performed atmospheric inversions to compute top-down methane emission estimates. DRB, GB, CCr, CF, PK, RL, TM, IM, SO'D, RJP, RP, MR, IJS, PS, YT, RFW, DWo, DWu, and YYo are PIs of atmospheric observations used in top-down inversions and/or contributed the text describing atmospheric methane observations. YZ, MvW, AV, VN, and MIH contributed to the chemical sink section by providing datasets, processing data, and/or contributed to the text. FMF and CCu provided data for the soil sink and contributed to the text of this section.

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