

# A rise in HFC-23 emissions from eastern Asia since 2015

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**Abstract.** Trifluoromethane (CHF<sub>3</sub>, HFC-23), one of the most potent greenhouse gases among HFCs, is mainly emitted to the atmosphere as a by-product in the production of the ozone depleting legacy refrigerant and chemical feedstock chlorodifluoromethane (CHClF<sub>2</sub>, HCFC-22). A recent study on atmospheric observation-based global HFC-23 emissions (top-down estimates) showed significant discrepancies over 2014–2017 between the increase in the observation-derived emissions and the 87 % emission reduction expected from capture and destruction processes of HFC-23 at HCFC-22 production facilities implemented by national phase-out plans (bottom-up emission estimates) (Stanley et al., 2020). However, the actual regions responsible for the increased emissions were not identified. Here, we estimate the regional top-down emissions of HFC-23 for eastern Asia based on *in situ* measurements at Gosan, South Korea, and show that the HFC-23 emissions from eastern China have increased from  $5.0 \pm 0.4$  Gg yr<sup>-1</sup> in 2008 to  $9.5 \pm 1.0$  Gg yr<sup>-1</sup> in 2019. The continuous rise since 2015 was contrary to the large emissions reduction reported under the Chinese hydrochlorofluorocarbons production phase-out management plan (HPPMP). The cumulative difference between top-down and bottom-up estimates for 2015–2019 in eastern China was  $\sim 23.7 \pm 3.6$  Gg, which accounts for  $47 \pm 11$  % of the global mismatch. Our analysis based on HCFC-22 production information suggests the HFC-23 emissions rise in eastern China is more likely associated with known HCFC-22 production facilities rather than the existence of unreported, unknown HCFC-22 production, and thus observed discrepancies between top-down and bottom-up emissions could be attributed to unsuccessful factory-level HFC-23 abatement and inaccurate quantification of emission reductions.

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## 1 Introduction

Trifluoromethane (CHF<sub>3</sub>, HFC-23) is mainly emitted as a by-product during the production of chlorodifluoromethane (CHClF<sub>2</sub>, HCFC-22), a legacy ozone depleting refrigerant that is increasingly used as a chemical feedstock mainly for the manufacture of fluoropolymers (UNEP, 2018). HFC-23 has the highest 100-year global warming potential (GWP<sub>100</sub> of 12,690) among hydrofluorocarbons and a long atmospheric lifetime of  $\sim 228$  years (Montzka and Velders, 2018; Newman et al., 2013). HFC-23 is included in the group of greenhouse gases regulated under the 1997 Kyoto Protocol. Participating developed countries (Annex I countries to the United Nations Framework Convention on Climate Change (UNFCCC)) are obligated to report their HFC-23 emissions to the UNFCCC. While developing countries (non-Annex I countries) are not obligated to report their HFC-23 emissions to the UNFCCC, some developing countries (e.g., China, India, South Korea, Mexico, and Argentina) participated in the UNFCCC's Clean Development Mechanism (CDM) between 2003 and 2014, a program to reduce HFC-23 emissions from HCFC-22 production facilities. Participating HCFC-22 facilities were required to report their HFC-23 reductions to the UNFCCC and could trade Certified Emissions Reduction (CER) credits with developed countries. During

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the CDM period, the total HFC-23 emission reductions provided by developing countries were compiled in 474 CDM  
45 monitoring reports (data available at <https://cdm.unfccc.int/>). However, note that not all HCFC-22 production facilities  
participated in the CDM program, and several facilities did not install HFC-23 abatement (destruction) technology (Montreal  
Protocol Technology and Economic Assessment Panel 2017 Assessment: hereinafter TEAP, 2017). Later, the HFC-23 by-  
production and emission associated with the production of HCFC-22 also began to be controlled by the Kigali Amendment to  
the Montreal Protocol on Substances that Deplete the Ozone Layer (MP), adopted in 2016 (Clark and Wagner, 2016), to  
50 preserve the climate benefit achieved by the MP.

The global production of HCFC-22, and thus HFC-23 emissions, are heavily centered around eastern Asia. Notably, China,  
where more than 50 % of the global HCFC-22 has been produced since 2009, can be considered one of the most significant  
HFC-23 emitters in the world (UNEP, 2018; Simmonds et al., 2018). After the end of the CDM in 2014, the Chinese National  
Development and Reform Commission (NDRC) announced its plan for a complete reduction of all HFC-23 emissions from  
55 HCFC-22 production facilities by 2019 (NDRC, 2015). From 2014 to 2016, 13 HFC-23 abatement systems were installed at  
15 HCFC-22 production lines that did not previously participate in the CDM program (TEAP, 2017). TEAP 2021 reported  
that the Chinese total of 32 HCFC-22 production lines in 2015 decreased to 18 lines in 2017, but also that the Chinese HCFC-  
22 production increased in 2017 compared to 2015. China has reported the annual HFC-23 emission reduction under the  
Chinese HCFC Production Phase-out Management Plan (HPPMP). The HFC-23 reductions reported by HCFC-22 production  
60 plants to the HPPMP were 45, 93, 98, and 99 % of total Chinese HFC-23 emissions in 2015, 2016, 2017, and 2018, respectively  
(UNEP, 2018; TEAP, 2021). According to China's HPPMP, the expected HFC-23 emissions from HCFC-22 production in the  
whole of China should have been 7.5, 1.0, 0.3 and 0.3 Gg yr<sup>-1</sup> in 2015, 2016, 2017 and 2018, respectively.

Considering the activities for HFC-23 emissions reduction, we discuss (a) the 2008–2012 “CDM period”, when mitigation  
of HFC-23 emissions had been introduced under the UNFCCC CDM to capture and incinerate HFC-23 co-production; (b) the  
65 2013–2014 “end of CDM period” and (c) the 2015 to 2019 “post-CDM period or HPPMP period” when the Chinese HPPMP  
mitigation activities were underway (after CDM).

A recent top-down study based on atmospheric measurements (Stanley et al., 2020) showed that the global HFC-23 emissions  
had increased and in 2018 were higher than at any point in history. Stanley et al. (2020) also suggested that these results  
contradicted the 87 % emission reduction anticipated over 2014–2017 from HFC-23 capture and destruction activities at  
70 HCFC-22 production facilities under the national phase-out plans initiated following the end of the CDM programs in a few  
developing countries including China (UNEP, 2018; Say et al., 2019). The observed discrepancies between the top-down and  
the inventory-based “bottom-up” global HFC-23 emission estimates implied unexpectedly low abatement rates, unreported  
HCFC-22 production and resulting HFC-23 by-product emissions, and/or unexpected substantial end-use and associated HFC-  
23 emissions. Therefore, these results at the global scale should be supported by investigations at the regional level where  
75 these HFC-23 emission controls have been implemented.

In this study, we present top-down HFC-23 emissions in eastern Asia for 2008–2019, including for the first time, post-CDM  
HFC-23 emissions in this region. Our estimates combine long-term, high-frequency *in situ* observations of atmospheric HFC-  
23 concentrations made at a regional site of the Advanced Global Atmospheric Gases Experiment (AGAGE) network in East  
Asia (Gosan station, Jeju Island, Korea, 33.3°N, 126.2°E) for 2008–2019, with a Bayesian atmospheric inverse method based  
80 on a Lagrangian particle dispersion model. To investigate the link of HFC-23 emissions to HCFC-22 production, we also  
analyse bottom-up estimates of HFC-23 emissions derived from reported HCFC-22 productions and emission factors for co-  
productioned rates of HFC-23. The comparison between the regional top-down and bottom-up emissions is further discussed  
in terms of their potential contribution to the unexpected trends in global HFC-23 emissions observed in 2014–2017.

## 2 Method

### 85 2.1 *In situ* Observations at Gosan

High-frequency atmospheric observations were made from 2008 to 2019 at Gosan (33.3°N, 126.2°E, 72 m a.s.l.) with a custom build “Medusa” cryogenic trap and focusing system with Gas Chromatographic and Mass spectrometric (GC-MS) detection (Miller et al., 2008; Arnold et al., 2012). Measurements of ambient air samples, drawn from an air inlet at 72 m high (a.s.l.) every two hours, are bracketed with measurements of a working standard to correct for instrumental drift. The working  
90 standard is calibrated weekly against a secondary calibration tank prepared at the Scripps Institution of Oceanography (SIO). Measurements are reported on SIO calibration scales (e.g., SIO-07 scale for HFC-23 and SIO-05 scale for HCFC-22 and CFC-11 in this study). The “Medusa” system is operated as part of the AGAGE program (Prinn et al., 2018). Precision ( $1-\sigma$ ) derived from repeated analysis ( $n=12$ ) of a working standard of ambient air was better than 1 % (i.e., the precision of HFC-23 < 1 %, HCFC-22 < 0.5 %, CFC-11 < 0.4 %) of background atmospheric concentrations for all compounds. The background  
95 concentrations (i.e., regional baseline) were determined using the AGAGE statistical pollution filtering algorithm (see more details in O’Doherty et al., 2001)

Gosan station is located on a remote cliff-top of the south-western tip of Jeju Island, south of the Korea Peninsula on the boundary between the Pacific Ocean and the Asian continent. The location is ideal for monitoring long-range transport from the surrounding regions with minimal local contamination. In the winter, the station often receives clean air masses  
100 characteristic of the “background” Northern Hemisphere directly from northern Siberia. In the summer, the monsoon brings in southerly oceanic air masses characteristic of “background” in the Southern Hemisphere (Li et al., 2018). During other times, the station samples polluted air masses from the Asian continent, particularly from China and South Korea. Some data gaps in the measurement timeseries occurred in summer and early autumn due to typhoons and heavy rains. In particular, the data gap from September 2016 to mid-April 2017 was due to impact of Super Typhoon Chiba and following reconstruction of  
105 the station (Figure 1 and Figure S6S5).

### 2.2 Bayesian Inverse Modelling of Regional Emissions

#### 2.2.1 Atmospheric Transport Modelling

The FLEXPART model is a widely used Lagrangian particle dispersion model for simulating atmospheric transport and  
110 mixing of trace gases. The model computes trajectories of air parcels from a source to a receptor location to describe atmospheric diffusion and vertical and horizontal transport and mixing processes. This information can then be used to estimate the contribution of surface emissions from different regions to the concentrations measured at a receptor location.

In this study, we use the FLEXPART version 10.4 (Pisso et al., 2019), which models atmospheric transport with meteorological fields from the Climate Forecast System Reanalysis (CFSR) model, at  $0.5^\circ \times 0.5^\circ$  spatial and 1-hour temporal  
115 resolutions (Saha et al., 2010, 2011). 50,000 particles were released from the sampling inlet height at the Gosan station during a 30 min window centered around each measurement time and tracked backwards in time for 20 days from the measurement site throughout the global domain. Footprint sensitivities, which relate the emissions at each  $0.5^\circ \times 0.5^\circ$  grid to measured concentrations at Gosan, are estimated from the model outputs below 100 m in altitude. The total emission sensitivity was used to determine the variable size grids with grid cells ranging in size from  $0.5^\circ \times 0.5^\circ$  to  $12^\circ \times 12^\circ$ , aggregated with larger  
120 grid size with decreasing sensitivity (For more details, refer to Kim et al., 2021). On the basis of a consideration of the sensitivity of the observations to emissions from surrounding model grid cells (Figure S1), we focus our analysis on the region that we denote “eastern Asia ( $20^\circ\text{N}$ – $50^\circ\text{N}$  and  $110^\circ\text{E}$ – $160^\circ\text{E}$ )”. This region contains eastern China, South Korea, North Korea, and western Japan. The boundaries of the nine provinces (Anhui, Beijing, Hebei, Jiangsu, Liaoning, Shandong, Shanghai, Tianjin and Zhejiang) in eastern China are denoted by thin grey lines in Figure S1.

## 125 2.2.2 Bayesian Inversion Framework: FLEXINVERT

The FLEXINVERT+ Bayesian inverse model framework combines measured enhancement concentrations, air transport estimated from FLEXPART, and *a priori* emissions to derive posteriori emissions. The model optimizes the posteriori emissions to minimize the following cost function (Thompson and Stohl, 2014):

$$130 \quad J(p) = 1/2(p-p^0)^T B^{-1}(p-p^0) + 1/2(H(p)-y)^T R^{-1}(H(p)-y) \quad (1)$$

Where  $p$  is the state vector of emissions,  $p^0$  is *a priori* estimate vector,  $y$  is the measured enhancements,  $H$  is the matrices derived from the FLEXPART backward simulations,  $B$  is the covariance matrix of *a priori* emission errors, and  $R$  is the covariance matrix of measurement errors.

135 In the model framework, the “enhancement” concentrations were determined by linearly extrapolating the monthly mean background concentrations and subtracting these extrapolated background concentrations from the measured concentrations corresponding to each measurement time (Kim et al., 2021), and the bihourly enhancements data were used for the inversion without temporal averaging.

*A priori* emissions of HFC-23 for the eastern Asia region and the global were determined as the 2008 *a priori* values of Stohl et al (2010) and Stanley et al (2020), respectively. The *a priori* emissions are 3.2 Gg yr<sup>-1</sup> for eastern China (8.5 Gg yr<sup>-1</sup> for entire China), 0.27 Gg yr<sup>-1</sup> for South Korea, 0.01 Gg yr<sup>-1</sup> for North Korea and 0.03 Gg yr<sup>-1</sup> Western Japan (0.08 Gg yr<sup>-1</sup> for entire Japan) (Stohl et al., 2010). The global *a priori* emission is 13.1 Gg yr<sup>-1</sup> (Stanley et al., 2020). Their magnitudes and spatial distributions were kept constant for the study period (2008–2019), ensuring that resulting trends of posteriori emissions are not biased by any changes in *a priori* inputs and are informed solely by observations. More detailed information on *a priori* distributions, magnitudes, and uncertainties is given in the *SI appendix*.

145 The assigned total uncertainty of each enhancement ( $\chi_{Measurement}$ ) was calculated as the square root of the quadratic sum of the three terms:

$$\chi_{Measurement} = \sqrt{(\chi_{Instrument})^2 + (\chi_{Background})^2 + (\chi_{Modelling})^2} \quad (2)$$

150 Where,  $\chi_{instrument}$  is instrumental precision of each compound, based on the repeatability of working standard measurements between sample observations (below 1 % for HFC-23);  $\chi_{Background}$  represents an uncertainty in background concentrations, which was determined as-by a month-to-month extrapolation between one standard deviation values of all the background monthly means and extrapolated similarly to the background concentrations derived at each measurement time over the observation period of 2008–2019 (about 0.2 ppt for the 2008-2019 HFC-23 enhancement average of 2.92 ppt); and  $\chi_{Modelling}$  is the model representation uncertainty that is estimated as about 6 % of our monthly mean background concentrations.

For HCFC-22 emissions estimates, we use the same inverse framework as used for HFC-23 based on high-frequency atmospheric observations from Gosan for 2008-2019. Optimal conditions of the inversion framework and its modelling performance were validated by analyzing CFC-11 emissions for eastern China based on Gosan CFC-11 data for 2008–2019 since the CFC-11 emissions were well-defined in recent studies (Rigby et al., 2019; Park et al., 2021) which used multiple inversion methods. See Supplement for details of the discussion about the inverse emissions estimates for HCFC-22 and CFC-11.

## 2.3 Bottom-up Emissions Estimates

165 Inventory-based bottom-up estimates of HFC-23 emissions can be derived from reports of HCFC-22 production multiplied by yearly emissions factors of co-produced co-production rates of HFC-23 (i.e., the total mass of co-produced and emitted

HFC-23 as a fraction of total HCFC-22 production) (Stanley et al., 2020; TEAP, 2021), and from national emissions reported to the UNFCCC for Annex-I countries. The former is important for non-Annex I countries that do not report emissions to the UNFCCC and refers to the ~~expected-potential~~ emissions when no abatement activities would be applied. To generate unabated bottom-up HFC-23 emissions estimates in eastern China, we used yearly varying ~~emission factors of co-produced co-~~  
170 ~~production rates of~~ HFC-23 in the range of 2.32–2.78 % (Stanley et al., 2020; TEAP, 2017; TEAP, 2021; UNEP, 2017), and the annual HCFC-22 production in eastern China.

For HCFC-22 production, the TEAP 2021 report provides the historical production data of entire China for 2008-2018 including both dispersive and feedstock uses (Figure S2), but not for our inversion domain of eastern China. Detailed information regarding factory-specific HCFC-22 production capacities and plant locations for China was given only for 2015  
175 and 2018 in the TEAP 2017 and TEAP 2021 reports, respectively. Therefore, the production information for eastern China was available only for those two years. To estimate the annual HCFC-22 production for eastern China for other years, we derived yearly fractions of eastern China HCFC-22 production to the China total (green dashed line in Figure S3) using the known 2015 and 2018 fractions in the equation (3). Here, we assume that the eastern China production fractions were correlated exponentially with time because HCFC-22 production rates in developing countries exhibited an exponential growth over time  
180 until mid-2010 (<https://ozone.unep.org/countries>).

$$P_{22}(E. Ch)/P_{22}(Ch) = A * \exp[k * (year - 2008)] \quad (3)$$

Here  $P_{22}(E. Ch)$  and  $P_{22}(Ch)$  are HCFC-22 productions from eastern China and the entire China, respectively.  $A$  is the initial fraction in 2008.  $k$  is a continuous growth rate. They were determined from the known HCFC-22 production fractions for the  
185 years of 2015 and 2018. Then we can calculate eastern China fractions of HCFC-22 productions for other years.

This analysis was based on the assumptions that the eleven HCFC-22 facilities reported to UNEP are all the plants producing HCFC-22 in entire China, and that their production reports are correct. If un-reported and/or newly built plants exist at unknown locations, actual fractions of HCFC-22 productions for eastern China could be changed, but our estimation of the abated HFC-23 emissions for eastern China in Section 3.3 would not be affected because the emissions abatement action under  
190 both the CDM and HPPMP programs must have been taken in the reported facilities. It is also assumed that a varying rate between the 2015 and 2018 production fractions in eastern China can be applied consistently over time. The annual HCFC-22 productions in eastern China calculated from the inferred eastern China fractions (green solid line in Figure S3) accounted for about 60–90 % of the annual totals for China during 2008–2018.

## 195 3 Result and discussion

### 3.1 Atmospheric mole fractions of HFC-23

The time-series of atmospheric HFC-23 concentrations for 2008–2019 at Gosan is shown in Figure 1. The baseline values of the Gosan observations (i.e., background values representing regional clean condition without regional/local pollution events; black dots in Figure 1) are determined using a statistical method developed in AGAGE, which uses a 121-day moving window  
200 to identify positive outliers from a Gaussian distribution that represent pollution events (O’Doherty et al., 2001). The annual averages of HFC-23 baseline concentrations at Gosan increased from  $22.4 \pm 0.4$  ppt in 2008 to  $33.3 \pm 0.5$  ppt in 2019 at a rate of  $1.0 \pm 0.1$  ppt yr<sup>-1</sup>. This rate is consistent with those from AGAGE stations Mace Head (53.3°N, 9.9°W) and Cape Grim (40.7°S, 144.7°E), which are representative of the global background monitoring stations in Northern Hemisphere and Southern Hemisphere, respectively. We observed pollution events (red dots in Figure 1) at Gosan continuously throughout the  
205 record with significant enhancements above background, which were influenced primarily by emissions from eastern Asia

sources. Notably, the pollution signals were significant even during the CDM period (for China, 2006 – early 2014) and later during the Chinese abatement activities under the HPPMP.

### 3.2 HFC-23 emissions from eastern Asia derived from atmospheric observations using inverse modelling

210 Unlike the global background monitoring stations, we observed large pollution events at Gosan continuously throughout the record, which has prompted a more detailed analysis of HFC-23 in eastern Asia to find and quantify potential source. Using the measurement enhancements in concentration units above the monthly-mean baseline, the atmospheric transport model (the Lagrangian particle dispersion model FLEXPART), and a regional inverse model framework (FLEXINVERT+), we calculate the qualitative spatial distribution of a posteriori emissions and estimate regional emissions of HFC-23 in eastern Asia  
215 including eastern China, North Korea, South Korea and western Japan (Figure 2 and Table 1). These regions are based on the high sensitivity of the Gosan atmospheric measurements to HFC-23 emissions from the surrounding areas (see Figure S1). The region denoted “eastern China” contains the nine provinces (Anhui, Beijing, Hebei, Jiangsu, Liaoning, Shandong, Shanghai, Tianjin, and Zhejiang) and “western Japan” contains the four regions (Chūgoku, Kansai, Kyūshū & Okinawa and Shikoku) (Rigby et al., 2019; Park et al., 2021; Kim et al., 2021; Western et al., 2022).

220 Our top-down results show that HFC-23 emissions in eastern Asia are increasing overall from  $5.11 \pm 0.36$  Gg yr<sup>-1</sup> in 2008 to  $10.01 \pm 1.04$  Gg yr<sup>-1</sup> in 2019, with emissions from eastern China accounting for 92–99 % of the regional total emissions (Table 1). HFC-23 emissions in eastern China showed an overall upward trend, with an appreciable decrease for 2010–2012 and an increase at the end of the CDM period (until 2014). During the HPPMP period, the eastern China HFC-23 emissions initially dropped (in 2015), but then increased continuously from  $5.7 \pm 0.3$  Gg yr<sup>-1</sup> in 2015 to  $9.5 \pm 1.0$  Gg yr<sup>-1</sup> in 2019. Our results do  
225 not support the significant reduction in HFC-23 emissions that could be expected from the national mitigation activities under the Chinese HPPMP starting after the end of the CDM period. The cumulative total emissions were  $21.9 \pm 2.1$  Gg (equivalent to  $278.1 \pm 26.1$  MtCO<sub>2</sub>-eq) for the CDM period of 2008–2012 vs.  $28.2 \pm 3.6$  Gg (equivalent to  $357.6 \pm 45.5$  MtCO<sub>2</sub>-eq) for the post-CDM period of 2015–2019. The Chinese emissions of HFC-23 presented by previous [top-down](#) studies agree with our estimates from 2008 to 2012, considering their large variability (Figure S4). Until this work, there was no HFC-23  
230 emissions estimate for China and other eastern Asia countries for 2012 or later, including the post-CDM period. For South Korea, HFC-23 emissions increased from 0.0–0.2 Gg yr<sup>-1</sup> (min–max) in the CDM period to 0.1–0.5 Gg yr<sup>-1</sup> (min–max) in the post-CDM period. The distribution map of HFC-23 emissions (Figures 3 and 4) also showed this temporal change in South Korea’s emissions. In 2019, the emissions from western Japan also increased slightly compared to previous years. The emissions from North Korea were near zero within modelling uncertainties. These national emission trends and magnitudes  
235 suggest that the observed increase in eastern Asia emissions was predominantly driven by those from eastern China. The cumulative emissions from eastern Asia for the entire study period 2008–2019 were  $73.8 \pm 9.2$  Gg which is equivalent to  $999.6 \pm 117.3$  MtCO<sub>2</sub>-eq.

The spatial distribution of emissions inferred within our eastern Asia region for the 2008–2019 period is shown in Figure 3 and compared to known HCFC-22 factory locations taken from TEAP reports, company websites and media articles (Table  
240 S1). The high emission flux densities were found predominantly in eastern China, where most of HCFC-22 factories are located. The posteriori emission maps are used to examine how the spatial emission distribution has changed with the HFC-23 mitigation projects. We grouped a posteriori HFC-23 emission maps for the CDM period (2008–2012), the end of the CDM period (2013–2014), and the HPPMP period (2015–2019) (see Figure 4). Over the three periods, the broad patterns in the emission hotspots did not change much, except for the end of the CDM, when we detected relatively stronger emission signals  
245 from eastern China, and a slight increase in the overall emissions. As revealed in the HFC-23 emission time series in Figure 2, the emissions during the HPPMP period were similar to or even higher than those during other periods.

### 3.3 Discrepancy between bottom-up and top-down eastern China estimates

250 The unabated bottom-up HFC-23 emissions in eastern China were estimated using yearly varying ~~emission factors~~  
~~production rates~~ of HFC-23 ~~co-produced~~ in the manufacture of HCFC-22 and the annual HCFC-22 production in eastern China  
(see Section 2.3 for details) and are shown in Figure 5 (red dotted line). HFC-23 emissions for eastern China with abatement  
(red dashed line in Figure 5) were estimated by including emissions reduction expected from the CDM and the HPPMP  
programs reports (UNEP, 2018). The anticipated reductions for eastern China were scaled down by the fractions of HCFC-22  
255 production capacities in eastern China determined by equation (3) in Section 2.3. The CDM abatement emission of HFC-23  
in eastern China was ~8.2 Gg in 2012 and decreased close to zero by the end of 2014. During the HPPMP activities, we  
estimate that the HFC-23 emissions reductions of eastern China by the HPPMP should have been 4.9, 11.0, 13.4 and 12.8 Gg  
yr<sup>-1</sup> in 2015, 2016, 2017 and 2018, respectively, based on reported abatement rates (UNEP 2018; TEAP, 2021). Therefore, the  
deduced bottom-up emissions with abatement for eastern China (red dashed line in Figure 5; note that the 2019 bottom-up  
260 emission was assumed constant after 2018 because the 2019 reduction of HFC-23 was not reported yet) dramatically dropped  
in 2015 and remained very low since 2016, contrary to the increasing emissions shown in our observation-based top-down  
emissions estimates (red solid line and shading in Figure 5).

It is interesting that the abated bottom-up estimates for 2012–2014 were larger than our observation-based emissions,  
although the global emissions in 2013–2014 also showed a very similar anomaly. Possible explanations are: (a) excess  
reduction above the reported abatement capacity might occur and/or (b) branches of HCFC-22 manufacturing facilities might  
265 be installed and operated at other locations outside eastern China, and their production capacities were accounted for and  
recorded as those of main (headquarter) factories located in eastern China (Figure 3). If the latter is the case, then our eastern  
China HCFC-22 production fractions could be overestimated, thereby generating overestimates of no-abatement/abatement  
bottom-up emissions.

Despite these open questions, our observation-based top-down emissions were much larger than those expected from  
270 abatement/reduction activities from 2015, and it is clear that abated emissions in eastern China were supposed to be near zero  
since 2015 under the reduction plan. The total excess emission (i.e., the integrated difference between our top-down emissions  
and the bottom-up abatement emissions) of eastern China for 2015–2019 is  $23.7 \pm 3.6$  Gg (equivalent to around  $301.3 \pm 45.5$   
MtCO<sub>2</sub>-eq) (yellow + red hatched area of Figure 5), which represents ~3 % of the total greenhouse gases emissions of 9,876.5  
Mt-CO<sub>2</sub> for the whole of China for 2019 (Greenhouse Gas Emissions by Country, 2022).

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### 3.4 Comparison of HFC-23 emissions in eastern Asia to global HFC-23 emissions

Our estimated HFC-23 emissions from eastern Asia (Figure S5) accounted for  $49 \pm 11$  % of the global top-down emissions  
between 2008–2018 taken from Stanley et al (2020). Since our results show HFC-23 emissions from eastern China account  
for most of the total emissions for eastern Asia (92–99 %), we focus here on how eastern China has influenced global emissions  
280 trends (as shown in Figure 5).

The expected reduction in HFC-23 emissions from eastern China for 2015–2017 accounts for 78–82 % of the reported global  
HFC-23 emission reductions. The bottom-up emissions with abatement for eastern China demonstrate a similar temporal trend  
shown in the corresponding global values. The cumulative total excess emissions in eastern China of  $7.9 \pm 1.8$  Gg for 2015–  
2017 (top-down minus bottom-up; yellow hatched area of Figure 5) account for 47 % of the global cumulative excess emissions  
285 (gray hatched area of Figure 5). Given that the 2015–2019 excess emission in eastern China is  $23.7 \pm 3.6$  Gg, with an  
assumption that 47 % of the global excess in eastern China would be maintained for 2018–2019, the global cumulative excess  
emissions are expected to reach to  $50.5 \pm 7.6$  Gg (around  $641.1 \pm 96.8$  MtCO<sub>2</sub>-eq) for 2015–2019. This means that if the HFC-

23 emissions reduction reported by China and other countries had actually taken place, unknown HCFC-22 production of as much as  $2079 \pm 314$  Gg for 2015–2019 would be required to explain the global excess HFC-23 emissions of  $50.5 \pm 7.6$  Gg. Approximately 400 Gg per year of ~~unknown~~-HCFC-22 production with unabated HFC-23 emissions was produced during 2015–2019, which is more than two-thirds of the total HCFC-22 production per year in entire China (Figure S3). The eastern China excess HFC-23 emission of  $23.7 \pm 3.6$  Gg should result in unknown HCFC-22 production of  $977 \pm 147$  Gg, which is equivalent to 136–184 % of the total quantity of HCFC-22 produced in eastern China or 118–159 % of the total production in A5 parties reported for 2018.

On the other hand, while re-considering the eastern China HFC-23 emissions in terms of HCFC-22 production, we estimated HCFC-22 emissions from eastern China by using the HCFC-22 measurement data obtained at Gosan for 2008–2019 (Figure S6-S5 and Figure S7S6(a)), and then inferred the entire China emissions of HCFC-22 from the fraction of the population of eastern China. These observation-based China emissions of HCFC-22 were consistent not only with previous top-down studies but also with the inventory-based HCFC-22 emissions as shown in Figure S7S6(b), thereby suggesting that the bottom-up emissions for China were relatively well-defined and thus the associated HCFC-22 production records were reasonably accurate. Therefore, using the fraction of the HCFC-22 production in eastern China against the global total production (TEAP, 2021; Figure S3), we calculated the eastern China HFC-23 emissions by scaling down the global top-down HFC-23 emissions given in Stanley et al. (2020). The HCFC-22 production-scaled HFC-23 emissions (gray lines in Figure 6) agree well with our observation-inferred emissions for eastern China even during the HPPMP period, suggesting that recent increases in both global and regional HFC-23 emissions is predominantly associated with known HCFC-22 production facilities.

#### 4 Conclusions

Our top-down inversion estimates for HFC-23 emissions in eastern Asia for 2008–2019 were based on twelve-year, high-frequency in situ observations of atmospheric HFC-23 concentrations made at Gosan.

The first published post-CDM HFC-23 emissions in this region show that our observation-derived emissions were much larger than the bottom-up estimates that were expected to be close to zero after 2015 due to Chinese abatement activities under the HPPMP. Several considerations suggest that observed discrepancies between the top-down and bottom-up HFC-23 emissions estimates are less likely to originate from the existence of unreported, unknown HCFC-22 production. First, the spatial distribution of our derived HFC-23 emissions is well correlated to the locations of the known HCFC-22 plants. Second, our cumulative excess emissions of  $23.7 \pm 3.6$  Gg from eastern China between 2015 and 2019 would require unknown HCFC-22 production of as much as  $977 \pm 147$  Gg, which is equivalent to 136–184 % of the total quantity of HCFC-22 produced for 2018 in eastern China. Such large unknown production seems highly unlikely, given that our observation-based HCFC-22 emissions in China were consistent with the inventory-based HCFC-22 emissions estimates thereby suggesting that the bottom-up emissions for China were relatively well-constrained and thus the associated HCFC-22 production records were also reasonably accurate. Third, when we inferred the eastern China HFC-23 emissions by scaling down the global top-down HFC-23 emissions given in Stanley et al. (2020) based on the fraction of the HCFC-22 production in eastern China to the global total, the HCFC-22 production-scaled HFC-23 emissions were consistent with our observation-derived emissions for eastern China even during the HPPMP period. This suggests that both global and regional HFC-23 emissions, which were independently determined, are primarily associated with known HCFC-22 production facilities. On the basis of this reasoning, the discrepancies between top-down vs. bottom-up emissions of HFC-23 in eastern China are most likely due to unsuccessful reduction processes of HFC-23 at factory level and inaccurate quantification of emission reductions.

In addition, our results show that the HFC-23 emissions from eastern Asia explain a substantial fraction of the post-CDM rise in global emissions and are probably due to the unabated emissions in the manufacture of HCFC-22. Thus, they underscore actions to identify responsible industrial practices and organize more effective regulatory controls in order to address this

330 critical issue and prevent unwanted emissions at both regional and global scales. Other regions including the rest of China, which are not well detected by the current observations, may also have contributed to the unexpected trends in HFC-23. Further investigation will be required to better understand which mechanisms are responsible for the recent increase in global HFC-23 emissions in different regions.

#### 335 **Data availability**

Data from Gosan, Mace Head and Cape Grim stations are available from the AGAGE website (<http://agage.mit.edu/data/agage-data>) or upon request by contacting SP, SOD, and PBK.

#### **Author contribution**

340 HP, SG, YK and SP carried out the Gosan measurements. HP analyzed the data with SP and performed inverse modelling with contributions from JK, RLT and HC. JM, PKS, CMH and RFW supported the calibration, precision, and data quality of the long-term observations at Gosan. SOD, PJF, PGS and PBK provided the *in situ* measurement data from the global baseline stations. KMS reviewed the manuscript. RFW and RGP supported the AGAGE network. HP and SP wrote the manuscript with contributions from all co-authors.

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**Competing interests** The authors declare no competing interests.

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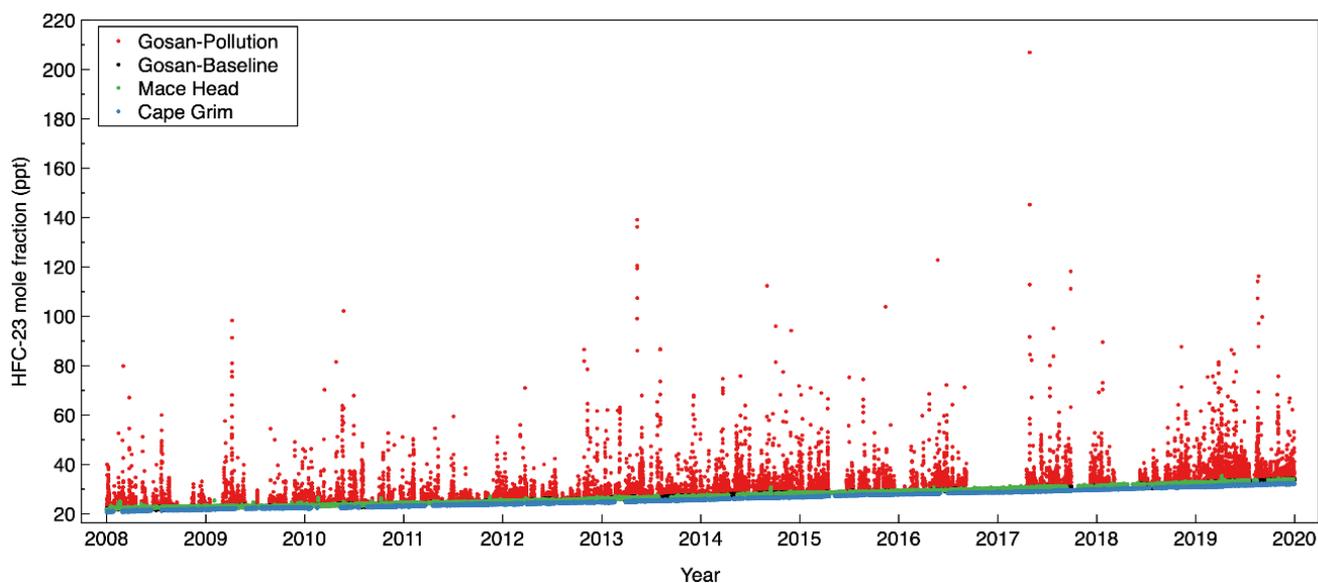
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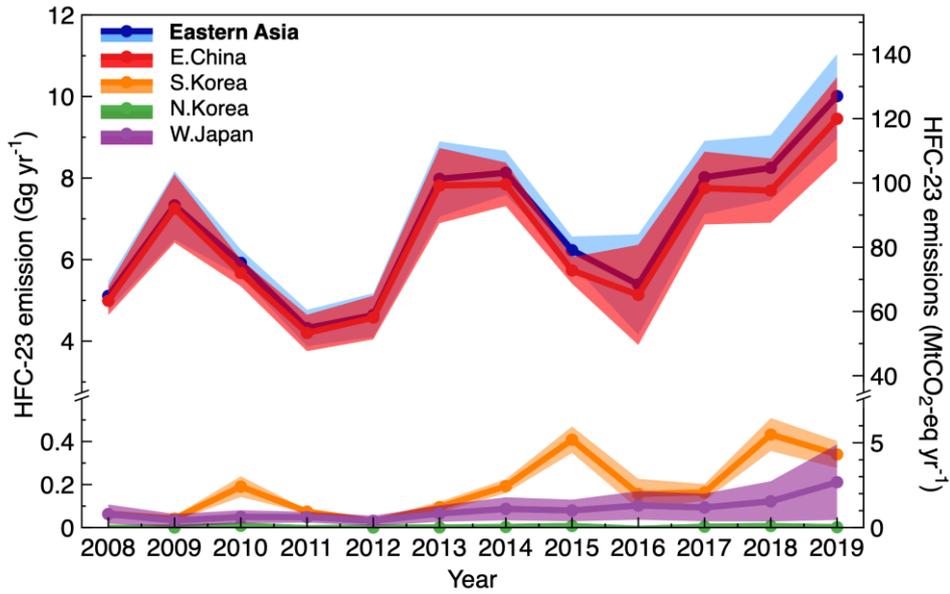
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450 **Figure 1: Atmospheric HFC-23 concentrations observed from 2008 to 2019 at the Gosan station (GSN, 33.3° N, 126.2° E). Pollution events denoted by red dots were identified as significant enhancements in concentrations from background levels shown in black. Some data gaps in the Gosan measurement timeseries occurred due to instrumental system downtime caused mostly by typhoons and heavy rains. The green and blue dots represent all observations of atmospheric HFC-23 at Mace Head (53.3°N, 9.9°W) and Cape Grim (40.7°S, 144.7°E), respectively, for comparison.**

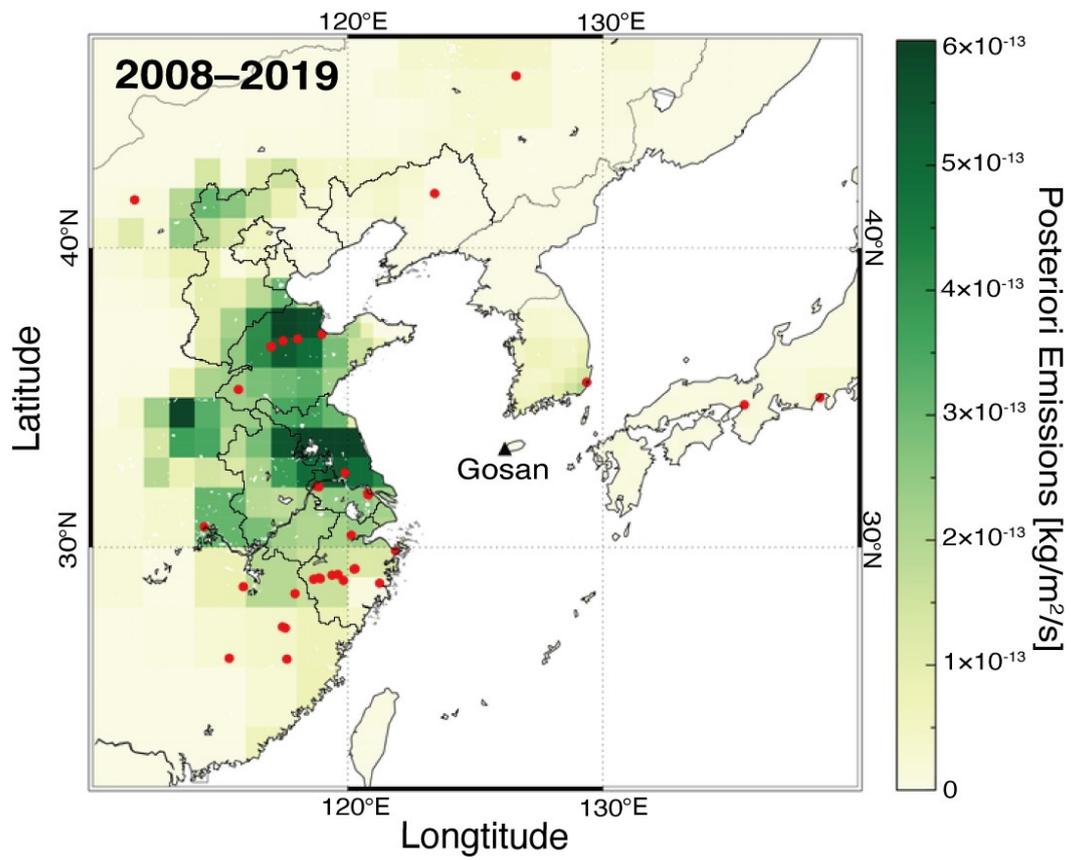


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Figure 2: HFC-23 emissions estimate in eastern Asia (blue) using the FLEXINVERT+ inverse method and in-situ HFC-23 measurements at Gosan, Jeju Island, South Korea. Top-down emissions estimates are shown for eastern China (red), South Korea (orange), North Korea (green), and western Japan (purple) for the years 2008–2019. Each line and the shading denote the annual averages of the a posteriori emission mean of the results from 18 different a priori settings and the 2-σ uncertainty, respectively.

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Right-hand axis is scaled by the 100-yr GWP for HFC-23 to show emission magnitudes in MtCO<sub>2</sub>-eq yr<sup>-1</sup>.



**Figure 3:** A posteriori HFC-23 emissions distribution in eastern Asia for 2008–2019. The red circles indicate locations of known HCFC-22 production plants. The Gosan measurement station is marked with a black triangle.

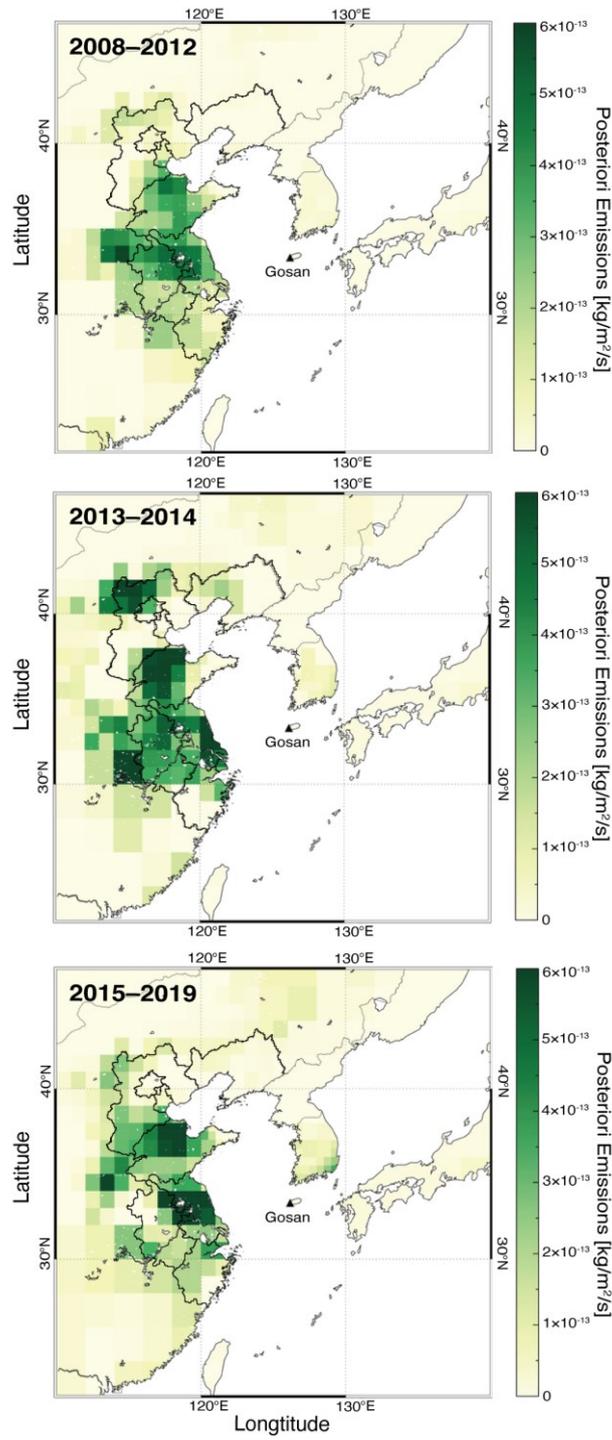
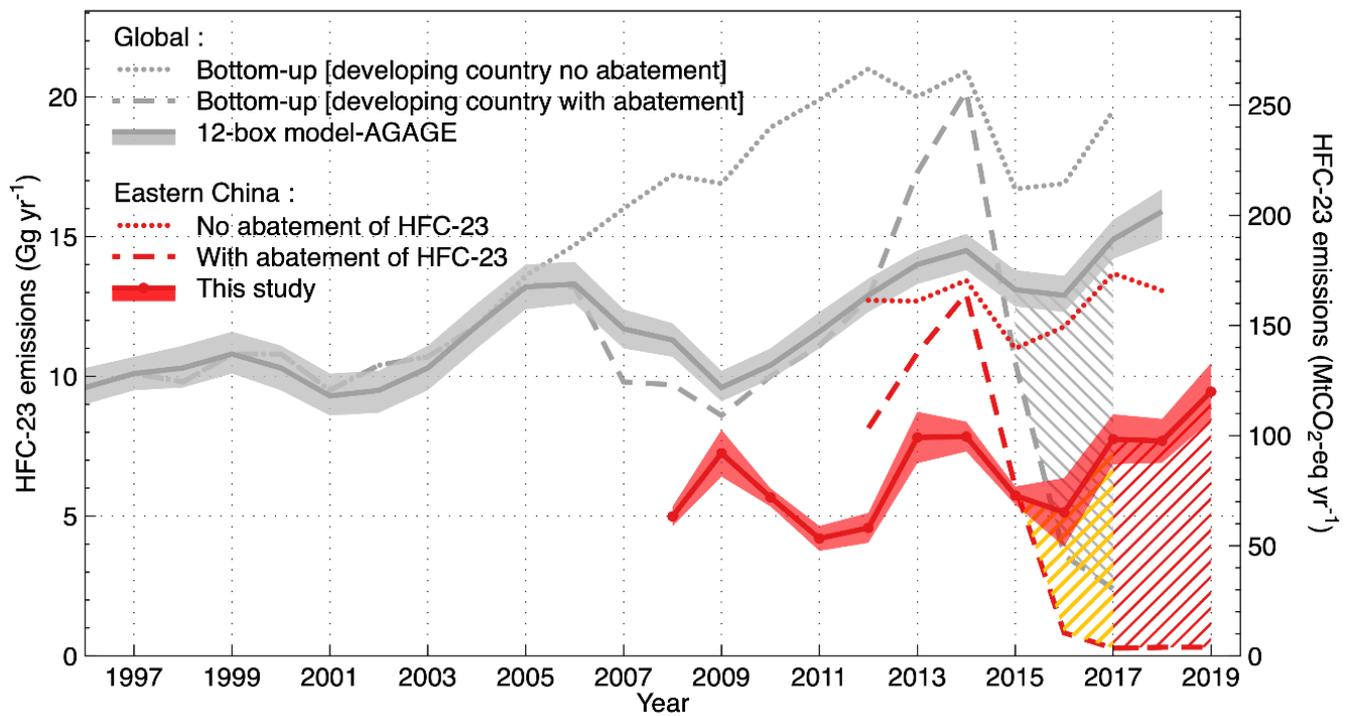
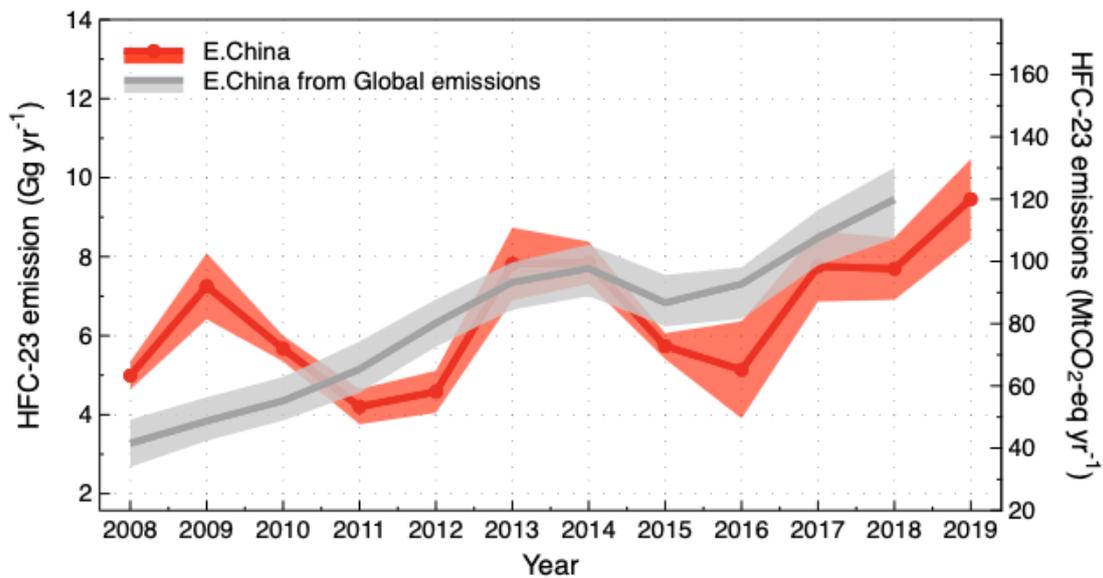


Figure 4: Maps of inferred HFC-23 emissions for the CDM period (top), CDM exit period (middle), and HPPMP period (bottom).



475 Figure 5: Observation-based HFC-23 emissions from eastern China versus top-down and bottom-up global HFC-23 emissions  
 480 derived from AGAGE global data (Simmonds et al., 2018; Stanley et al., 2020). No abatement and abatement bottom-up emissions  
 of eastern China are denoted by red dots line and red dashed line, respectively. No abatement emissions of HFC-23 are determined  
 by HCFC-22 production amounts multiplied by yearly ~~emissions factors of co-produced co-production rates of~~ HFC-23 (TEAP,  
 2021). HFC-23 emissions estimated with abatement include emissions reduction expected from the CDM monitoring reports  
 (<https://cdm.unfccc.int/Projects/registered.html>) and the HPPMP (UNEP, 2018). The hatched areas represent the total cumulative  
 excess emissions for eastern China (yellow from 2015 to 2017, red from 2017 to 2019) and for the globe (gray between 2015-2017).



485 Figure 6: HFC-23 emissions for eastern China inferred by downscaling the global top-down emissions estimates (Stanley et al., 2020) based on annual fractions of the eastern China HCFC-22 production to the global total production. The HCFC-22 production-scaled estimates of HFC-23 emissions (gray line) are in a good agreement with our observation-based emissions (red line).

**Table 1: Top-down emissions of HFC-23 for eastern Asia (eastern China, North Korea, South Korea, western Japan)**HFC-23 (Gg yr<sup>-1</sup>)

Year	Eastern Asia			Eastern China			North Korea			South Korea			Western Japan		
	mean	max	min	mean	max	min	mean	max	min	mean	max	min	mean	max	min
2008	5.11	5.47	4.76	4.99	5.34	4.64	0.00	0.00	-0.01	0.06	0.08	0.04	0.06	0.11	0.02
2009	7.33	8.17	6.49	7.25	8.09	6.41	0.00	0.01	0.00	0.04	0.05	0.03	0.03	0.06	0.00
2010	5.92	6.25	5.59	5.67	6.00	5.35	0.01	0.01	0.00	0.19	0.24	0.14	0.05	0.08	0.02
2011	4.32	4.77	3.87	4.20	4.65	3.76	0.00	0.01	-0.01	0.07	0.08	0.07	0.05	0.07	0.02
2012	4.64	5.18	4.11	4.58	5.11	4.05	0.00	0.00	0.00	0.03	0.04	0.02	0.03	0.06	0.01
2013	7.98	8.90	7.05	7.81	8.74	6.89	0.00	0.01	0.00	0.10	0.12	0.07	0.07	0.10	0.03
2014	8.13	8.67	7.59	7.85	8.38	7.31	0.00	0.01	0.00	0.19	0.22	0.16	0.09	0.14	0.03
2015	6.23	6.57	5.89	5.73	6.06	5.41	0.01	0.01	0.00	0.41	0.47	0.35	0.08	0.13	0.03
2016	5.39	6.62	4.16	5.14	6.36	3.91	-0.01	0.01	-0.02	0.16	0.23	0.09	0.10	0.17	0.04
2017	8.02	8.91	7.12	7.75	8.65	6.86	0.00	0.01	0.00	0.16	0.20	0.13	0.09	0.16	0.03
2018	8.25	9.05	7.46	7.69	8.48	6.91	0.01	0.01	0.00	0.43	0.51	0.36	0.12	0.22	0.03
2019	10.01	11.05	8.97	9.45	10.48	8.43	0.00	0.01	-0.01	0.34	0.40	0.28	0.21	0.39	0.04