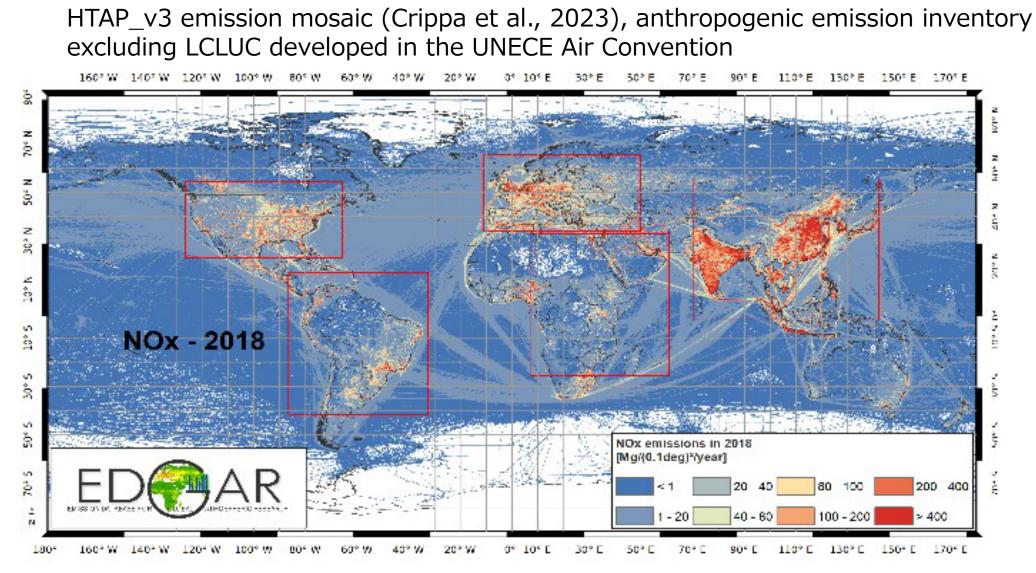
## Long-term trends of anthropogenic emissions in Asia and their validation in Japan

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## Global emission map for NOx in 2018



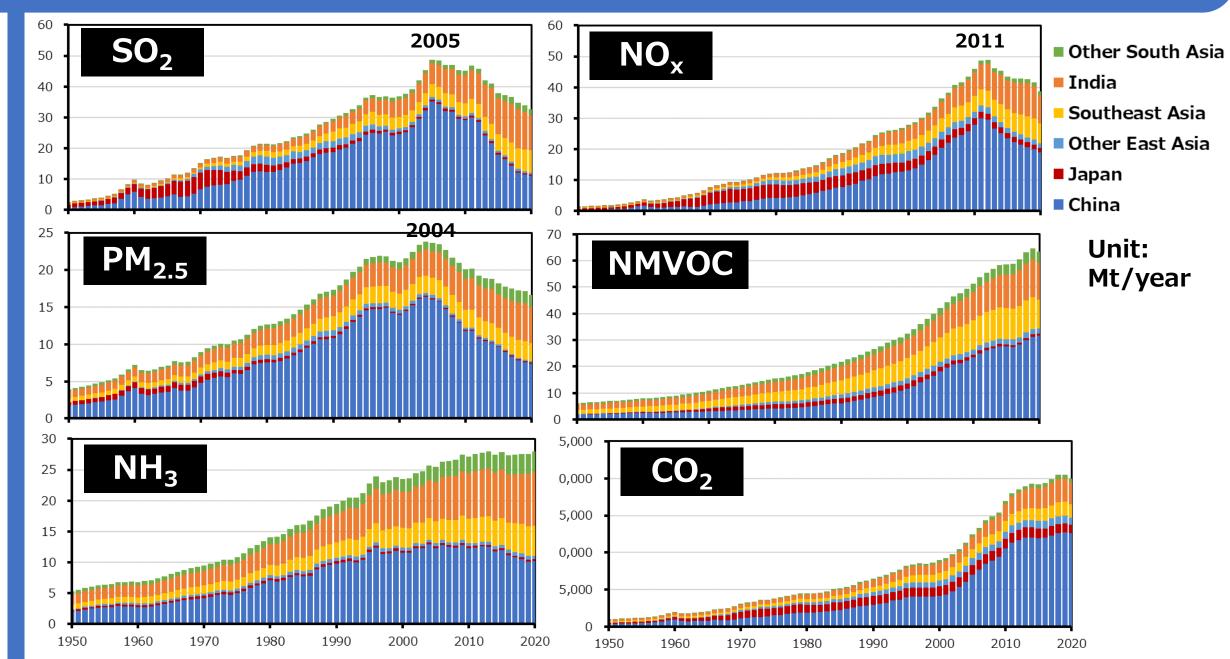
- $\checkmark$  Asia is most polluted region in global
- $\checkmark~$  Asian emissions account for almost half of global

## Regional Emission inventory in ASia (REAS)

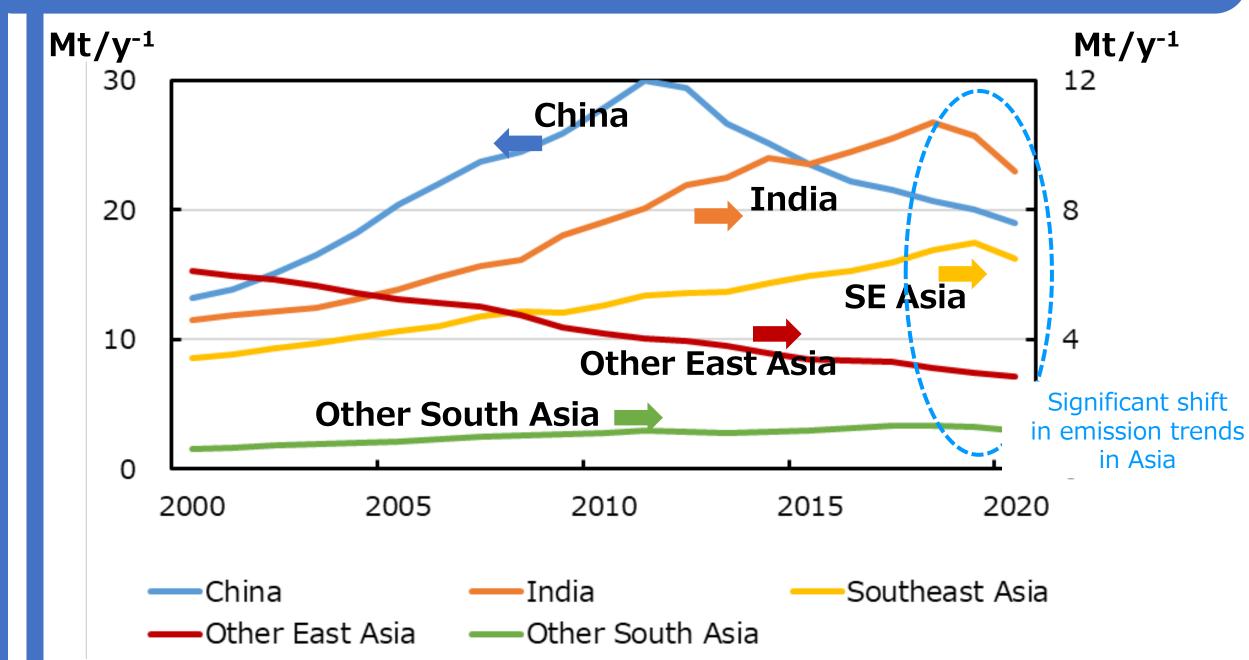
- Anthropogenic, comprehensive, and historical inventory
  Version: 3.2.1 (now updating)
- Country and regional emissions for detailed sources
- Gridded emissions for major sources
- Target Years : 1950-2015 ( ⇒ 2020)
- Target Areas : East, Southeast, and South Asia
- Horizontal Resolution :  $0.25^{\circ} \times 0.25^{\circ}$  ( $\Rightarrow 0.1^{\circ} \times 0.1^{\circ}$ )
- Temporal Resolution : Monthly
- Target Species : SO<sub>2</sub>, NO<sub>x</sub>, CO, NMVOC, PM<sub>10</sub>, PM<sub>2.5</sub>, BC, OC, NH<sub>3</sub>, CO<sub>2</sub> and CH<sub>4</sub>

	SO <sub>2</sub>	NO <sub>x</sub>	CO	PM <sub>10</sub>	PM <sub>2.5</sub>	BC	NMVOC	NH <sub>3</sub>	CO <sub>2</sub>	CH <sub>4</sub>
Combustion									$\bullet$	$\bullet$
Industrial Process	$\bullet$				$\bullet$	$\bullet$		$\bullet$	$\bullet$	$\bullet$
Agriculture								•		
Others										

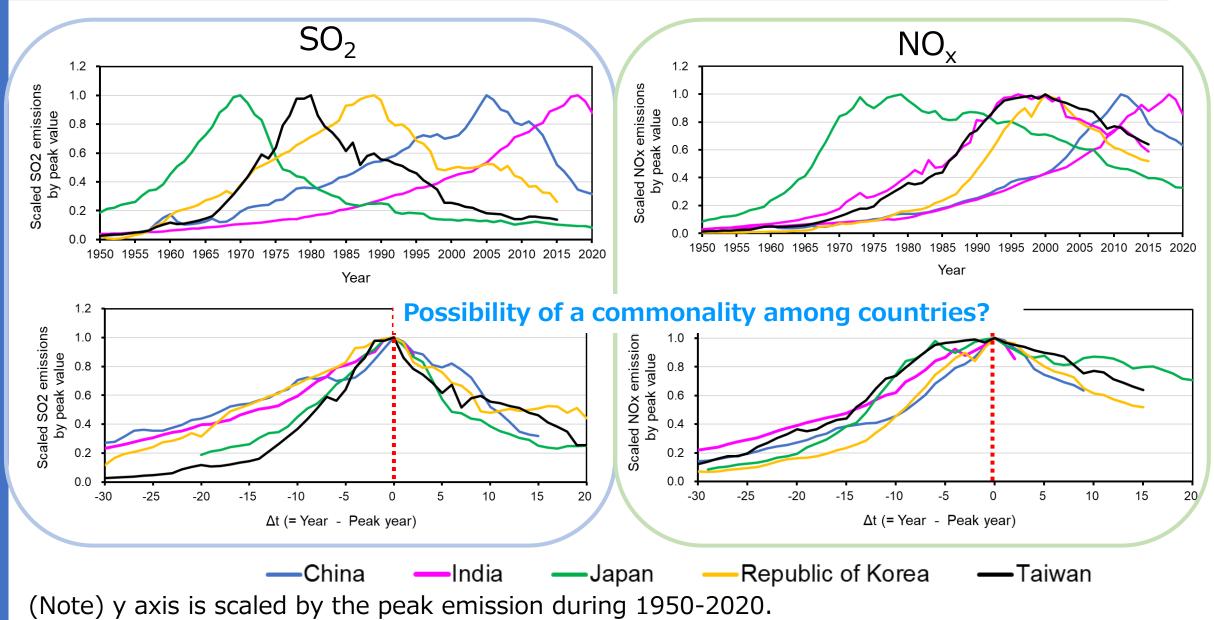
## 4 Temporal variations of emissions in Asia during 1950-2020



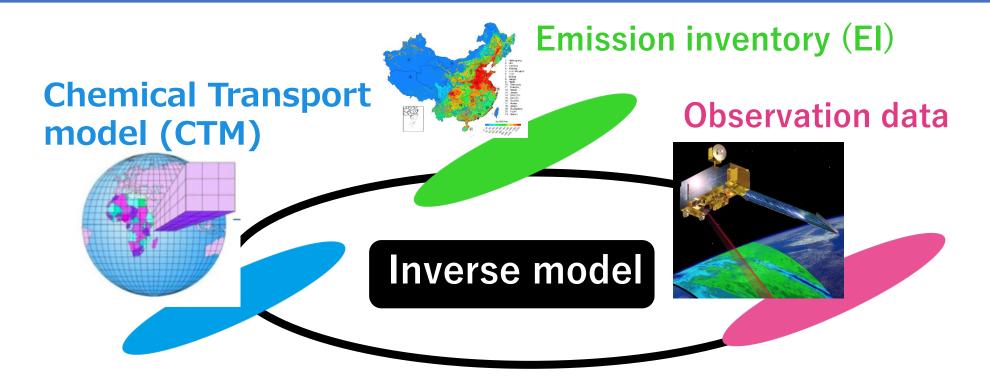
#### **Temporal variations of NOx emissions after 2000**



### Similarity of SO<sub>2</sub> and NOx emission changes in China, India, South Korea, Taiwan and Japan



#### Inverse modeling using satellite data

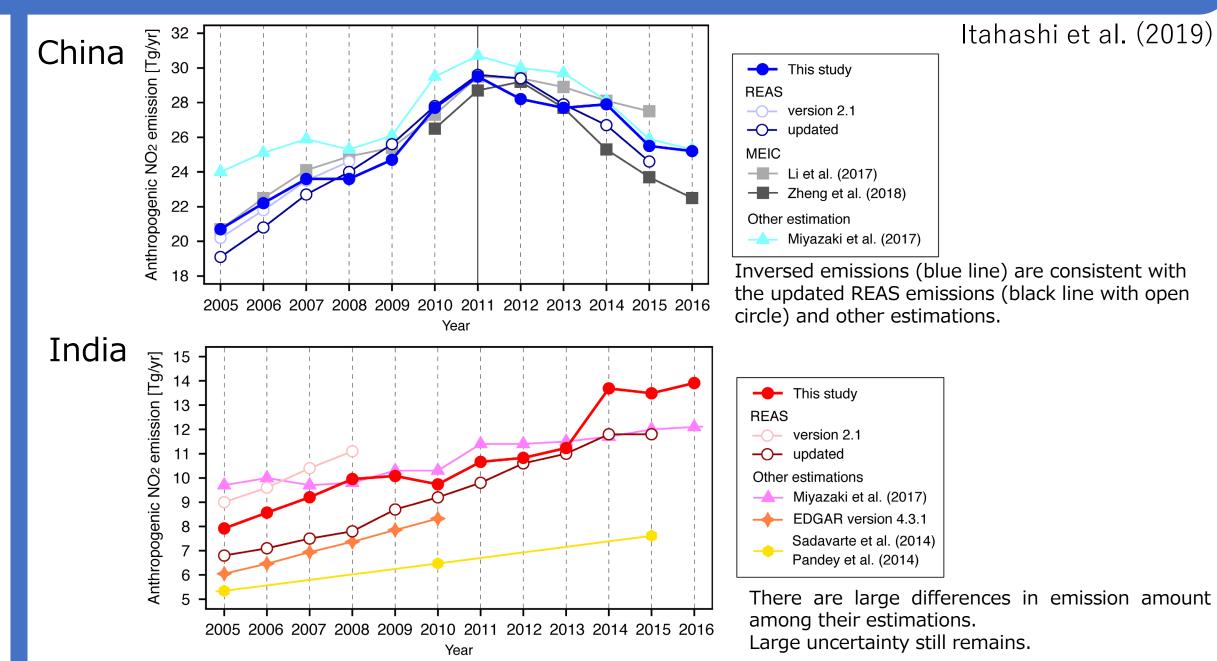


Inverse modeling integrates EI (a priori data), CTM and observation data to complement (optimize) emissions.

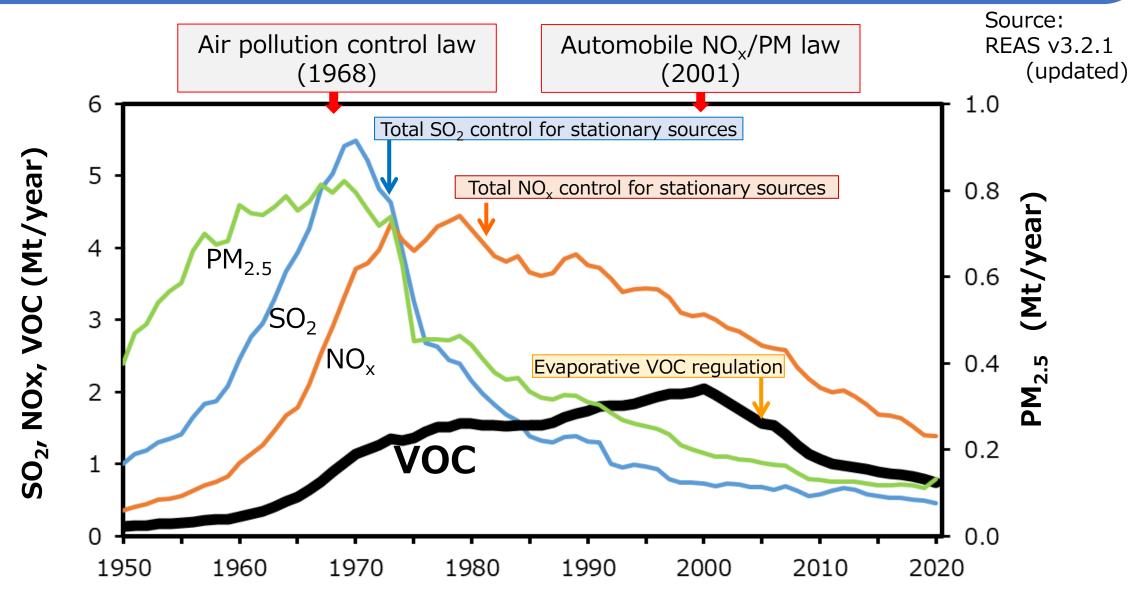
## = Inverse estimation of emissions

Bottom-up approach: Estimate emissions from statistical data Top-down approach : Estimate emissions from observations and CTM

#### 8 Temporal variation of annual NOx emissions over China and India



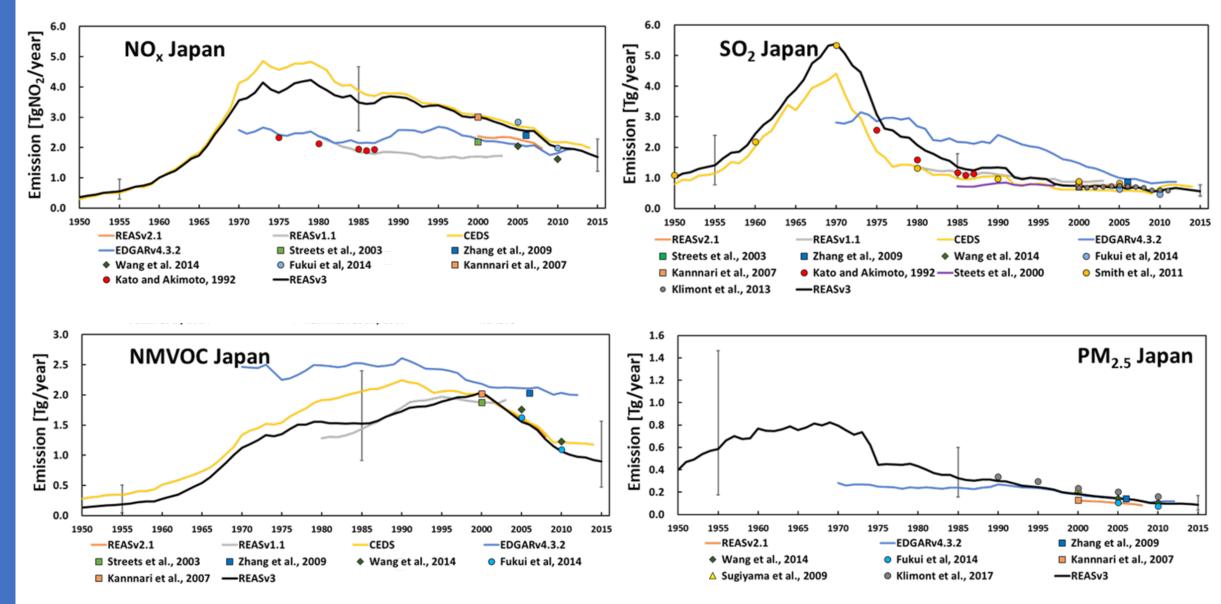
#### Historical emissions in Japan during 1950-2020 (Japanese case as good example of effects of emission control)



The peak year depends on the balance of emission control implementation and changes of activities related to emissions .

#### **Uncertainties of Japanese emissions**

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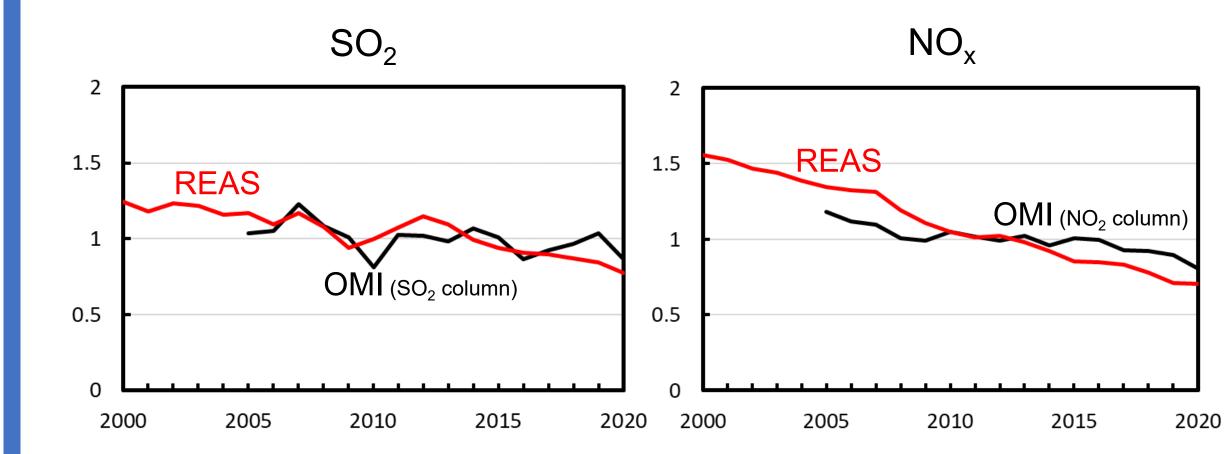
Kurokawa and Ohara (2020)

## **11 Topics on validation of emission trends in Japan**

## Historical variations of air pollutants emissions over Japan (using OMI column density and AQ monitoring data)

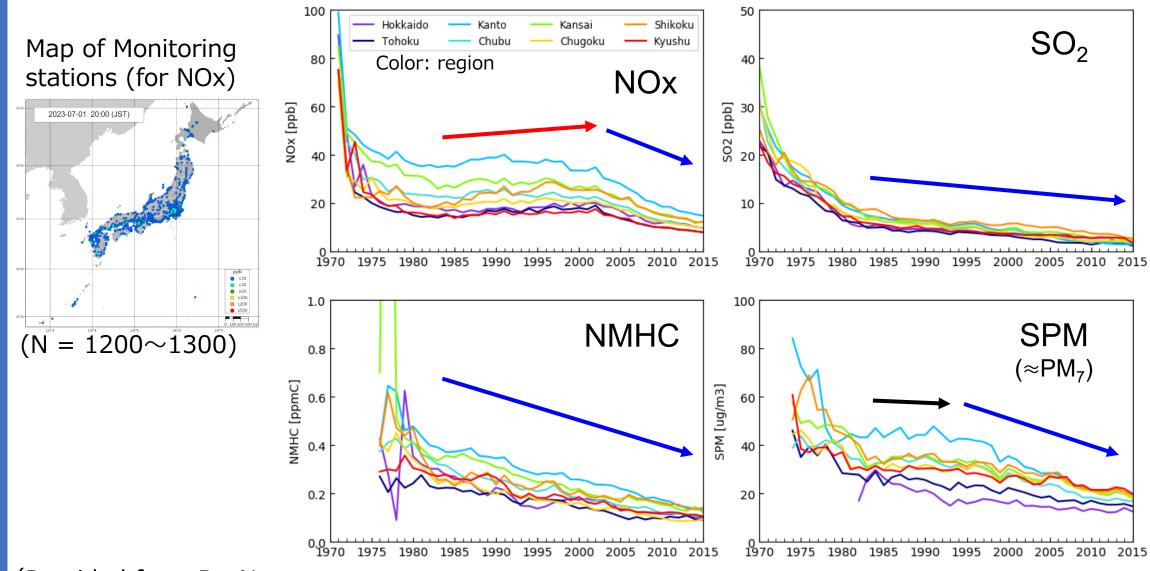
## 2. Trend of CO<sub>2</sub> emissions in urban areas (using observed data in CO<sub>2</sub> monitoring stations)

#### 12 Comparison of trends of REAS inventory and OMI over Japan



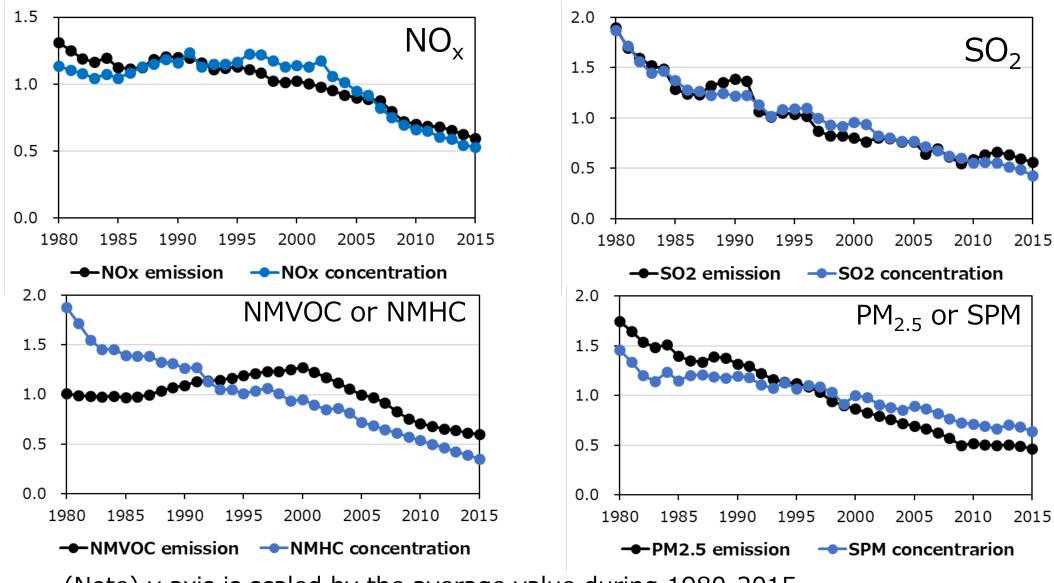
(Note) y axis is scaled by the average value during 2005-2020.

## 13 Long-term variations of annul mean concentrations observed at monitoring stations across Japan



(Provided from Dr. Natsumi Kohno)

## 14 Comparison of emissions and ambient concentrations of SO<sub>2</sub>, NO<sub>x</sub>, NMVOC (NMHC) and PM<sub>2.5</sub> (SPM)



(Note) y axis is scaled by the average value during 1980-2015.

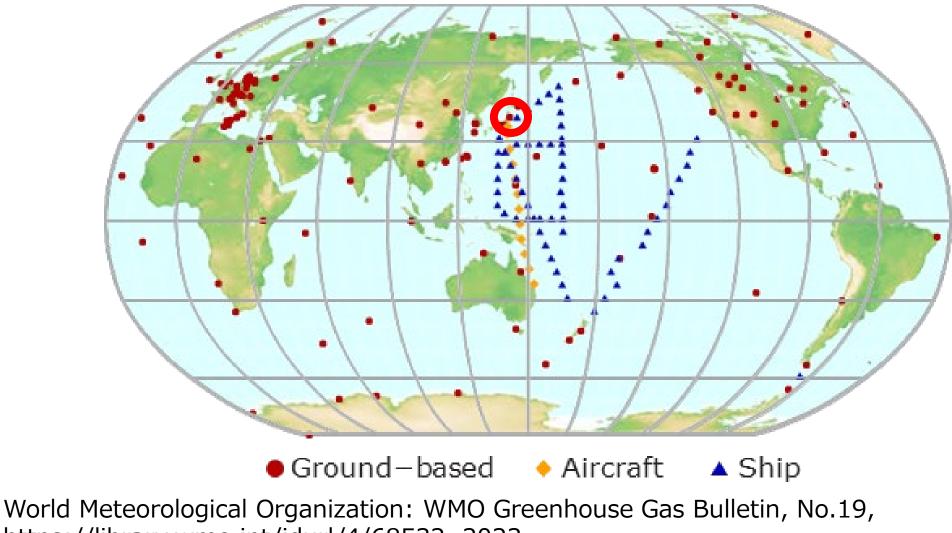
- The urban emissions of GHG were estimated to be about 70% of the global share in 2020 and continue to increase [Lwasa et al., 2022].
- However, the emission inventories for anthropogenic sources in urban area are considered to have large uncertainties [Gurney et al., 2021].
- Therefore, it is becoming a science-based information for tracking and verifying CO<sub>2</sub> emission reduction in urban area based on observing system as well as modeling and analysis methods.
- However, there is no research that has verified the long-term trend in CO<sub>2</sub> emissions in cities over a period of more than ten years.

 To verify the estimated reduction of CO<sub>2</sub> emissions in the urban area of Tokyo Metropolitan Area (TMA), using approximately 20 years of CO<sub>2</sub> concentration monitoring data from 2002 to 2020

 To demonstrates that long-term continuous and high-precision observations of CO<sub>2</sub> concentration can scientifically track and verify the effectiveness of CO<sub>2</sub> emission reduction policies and actions in the region

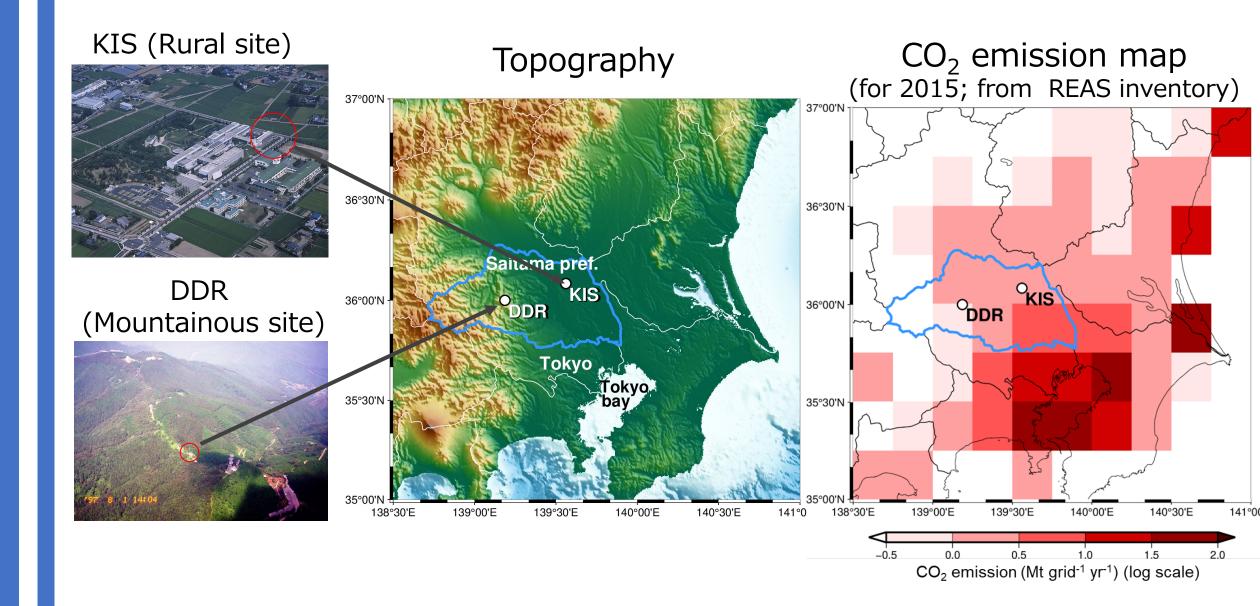
## GAW global network for CO<sub>2</sub> in the last decade

Two ground-based stations are operated by our institute and are a part of a global network of 184 stations which are regularly reported to the World Data Centre for Greenhouse Gases (WDCGG) of WMO.

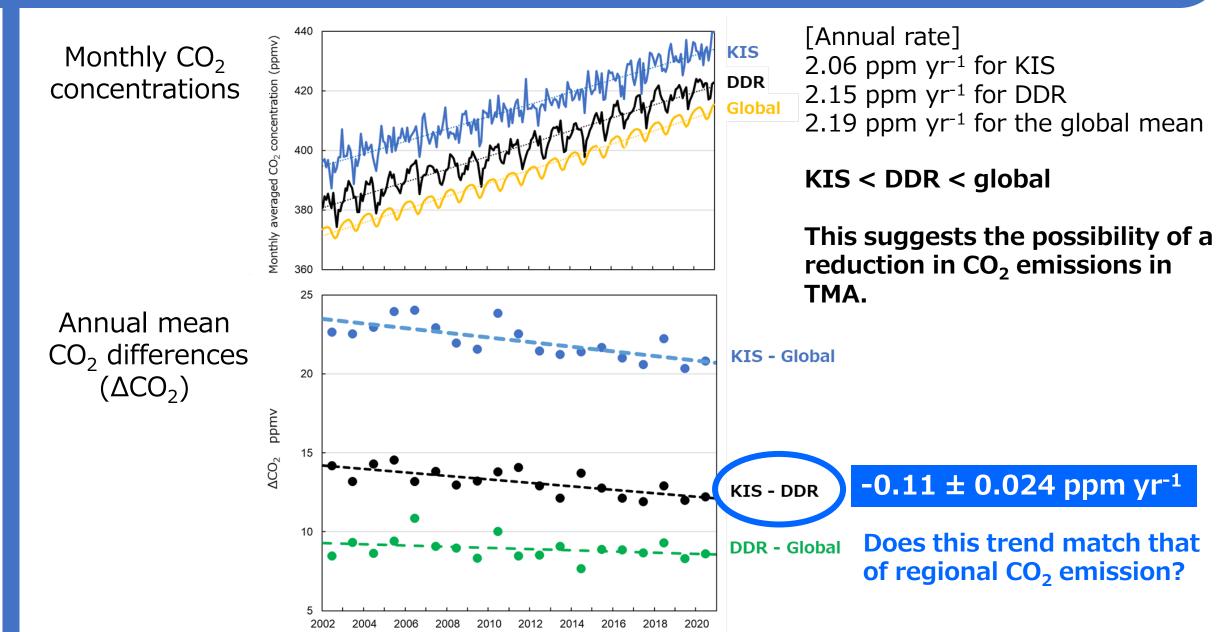


https://library.wmo.int/idurl/4/68532, 2023.

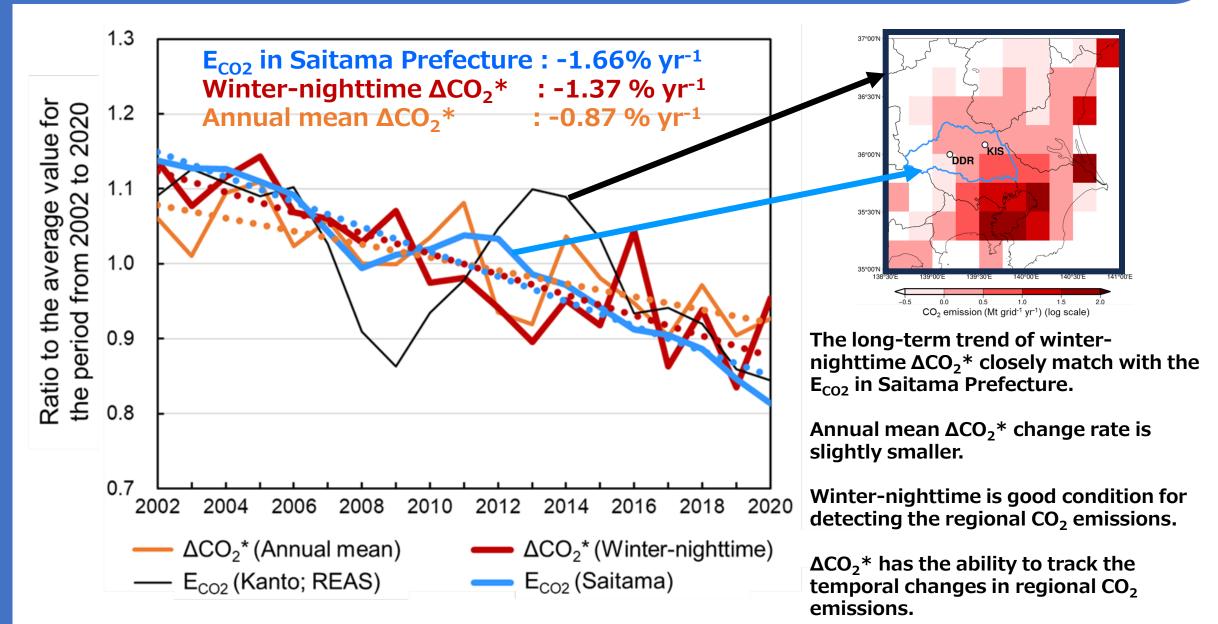
#### **Monitoring stations**



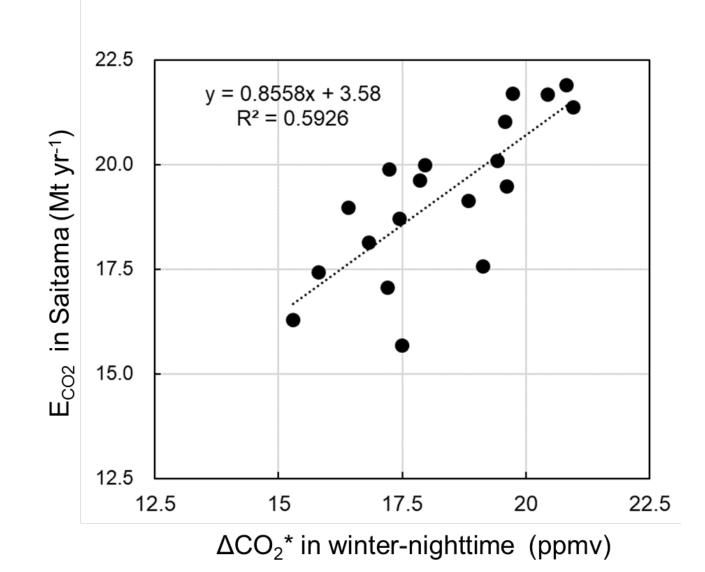
#### Temporal variations of CO<sub>2</sub> concentrations observed in two stations (DDR and KIS) and global mean



#### **Comparison with the annual average and winter-nighttime average** 20 of $\Delta CO_2^*$ (=KIS-DDR) and the annual emissions of $CO_2$ (E<sub>CO2</sub>)



# Relationships between $\Delta CO_2^*$ (winter-nighttime) and $E_{CO2}$ in Saitama prefecture



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It was the first time to demonstrate the long-term reduction of CO<sub>2</sub> emissions in mega cities by observations!

This study demonstrates that the long-term continuous observations of CO<sub>2</sub> concentrations in multiple stations can scientifically track and verify the effectiveness of emission reduction policies and actions in the region.

Furthermore, it can be verified through the observation of CO<sub>2</sub> concentration that the long-term trend in anthropogenic CO<sub>2</sub> emissions inventory from local government (Saitama prefecture) has high certainty.

To quantitatively understand, analyses using chemical transport models and inverse modeling are considered to be useful.

## 23 Important challenges for emission inventory in Asia

 Developing a methodology to reduce the uncertainty of EI using a top-down approach

 Developing national EIs in each country and capacity building in Southeast and South Asia

Future prediction of EIs under decarbonization scenarios

## **Questions and Interests**

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