



TEMPO/GEMS Joint  
Science Team Workshop  
August 26-30, 2024  
Kailua-Kona, Hawaii, USA

NASA National Institute of  
Environmental Research



# Estimation of city-scale $\text{NO}_x$ emission flux using the GEMS dataset

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### **1) The need for estimating $\text{NO}_x$ emissions in a sudden event (e.g., COVID-19)**

Anthropogenic emissions of  $\text{NO}_x$  are primarily generated from combustion activities by power plants and vehicles. Once a sudden event occurs, it is essential to quickly identify and provide information on changes in emission levels in major  $\text{NO}_x$ -emitting cities with dense populations.

### **2) Slow development of bottom-up inventories**

Estimating emissions by assessing the total fuel consumption from each  $\text{NO}_x$ -emitting source is time-consuming. If information on fuel consumption is not available, emission data cannot be obtained. Emissions from unknown anthropogenic sources cannot be identified.

### **3) The existence of geostationary satellite measurement**

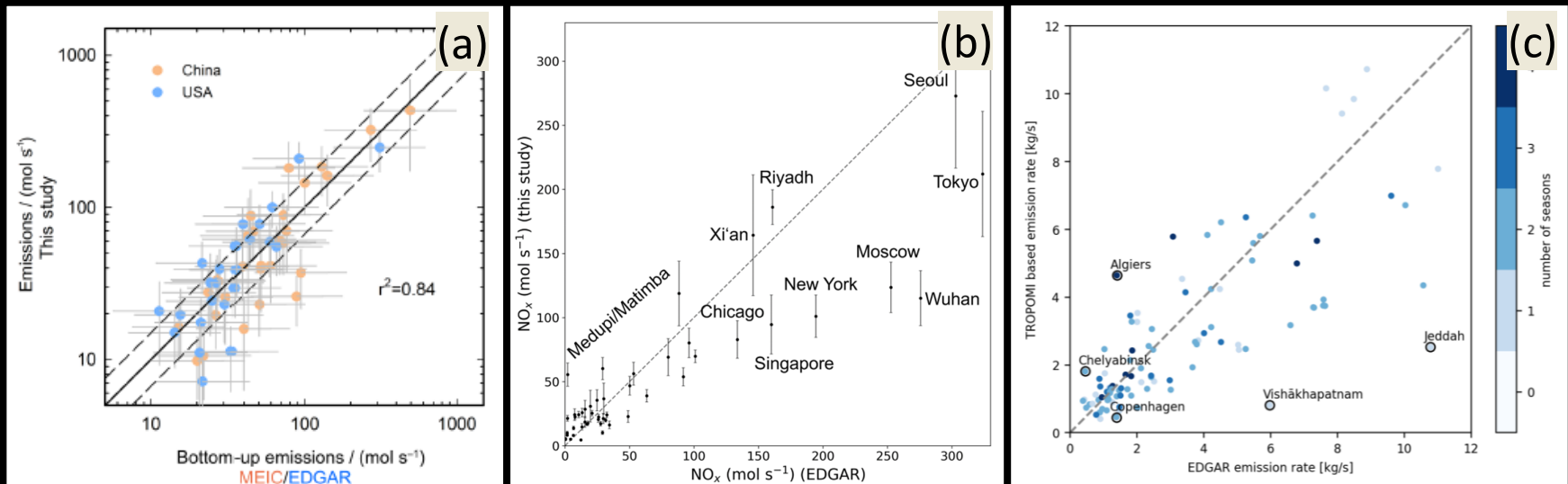
Due to the insufficient data from polar-orbiting satellites, it is challenging to estimate annual/seasonal emissions. But GEMS observes the East Asian region ( $5000 \times 5000 \text{ km}^2$ ) ~6 to 10 times a day. Therefore, when estimating  $\text{NO}_x$  emissions in the East Asian region, the usage of GEMS data is more beneficial as it reduces the uncertainty due to sampling.

# Previous study: Estimation of $\text{NO}_x$ emission using LEO satellites

Several previous studies (e.g., Beirle et al., 2011; Liu et al., 2016) succeeded to estimate city-scale  $\text{NO}_x$  emissions using OMI or TROPOMI satellite data, by **analyzing the differences in  $\text{NO}_2$  distribution between calm and windy condition.**

- (a) Emission estimates for 53 cities in China and the United States using OMI data in the ozone season during 2005-2013. (Liu et al., ACP, 2016)
- (b) Emission estimates for 100 cities worldwide using TROPOMI data during 2018-2020. (Lange et al., ACP, 2022)
- (c) Emission estimates for 100 cities worldwide using TROPOMI data during 2018-2021. (Beirle et al., EGU sphere, 2024)

(x-axis: bottom-up, y-axis: satellite-based estimation)

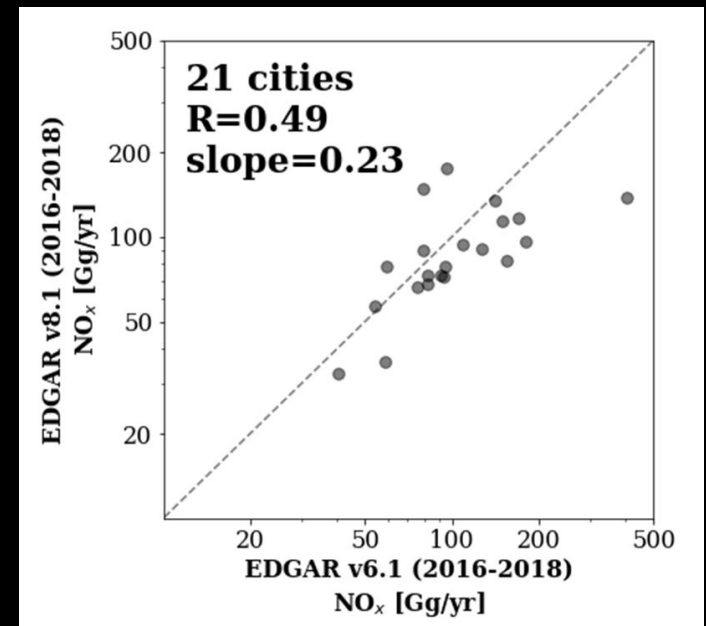


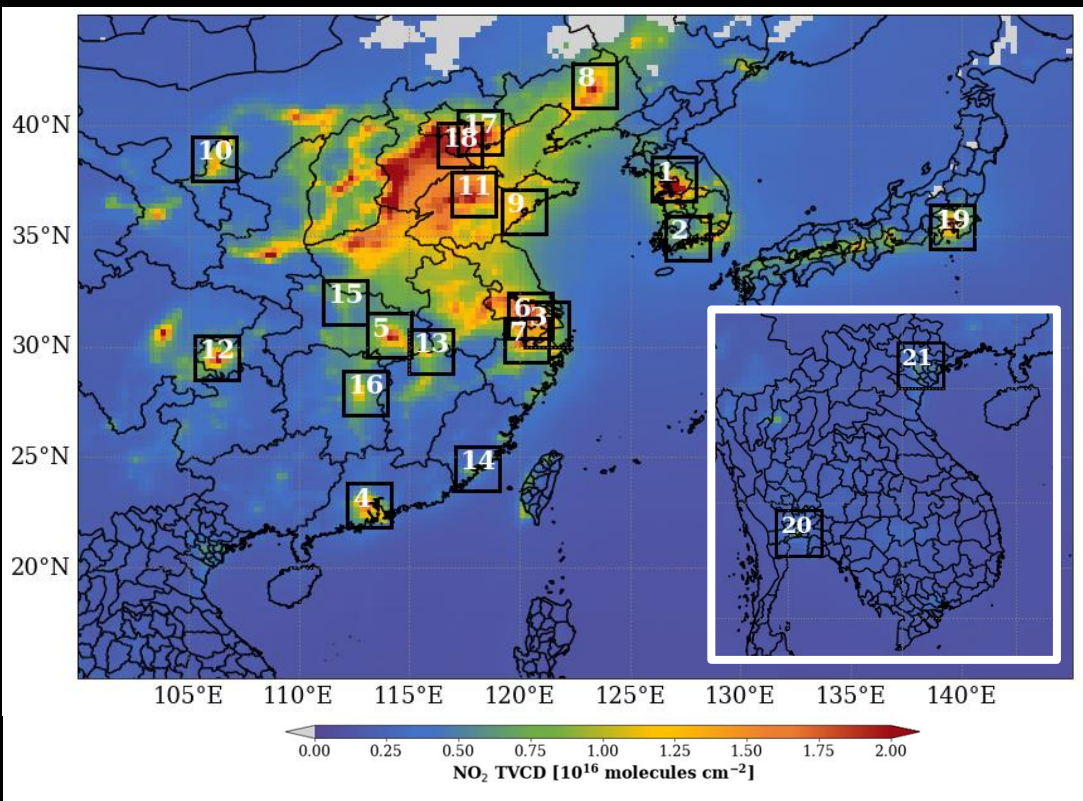
	Satellite	Reanalysis	Bottom-up inventory	
	GEMS L2 v2.0	ECMWF ERA5	EDGAR v6.1	EDGAR v8.1
Period	03.01.2022 – 02.28.2023	03.01.2022 – 02.28.2023	2016-2018	2016-2022
Variable	tropospheric NO <sub>2</sub> VCD (cloud fraction < 0.5, SZA < 70°)	u, v component (averaged at 1000hPa and 975hPa)	Total NO <sub>x</sub> emissions	
Resolution	hourly, 7×8 km <sup>2</sup>	hourly, 0.25°×0.25°	Annual, 0.1°×0.1°	

**Collocation** : To classify satellite data based on wind direction and wind speed

- Spatial collocation : GEMS data was regridded to a resolution of 0.05°×0.05°, resulting in 25 GEMS data grid cells being included within a single grid cell of the ERA5 data. 25 GEMS data grid cells are averaged for the same day under specific wind direction and wind speed conditions.
- Temporal collocation : we matched the time of GEMS observation with that of ERA5. (e.g. GEMS 03:50 UTC ⇔ ERA5 04 UTC)

## EDGAR (v6.1 versus v8.1)





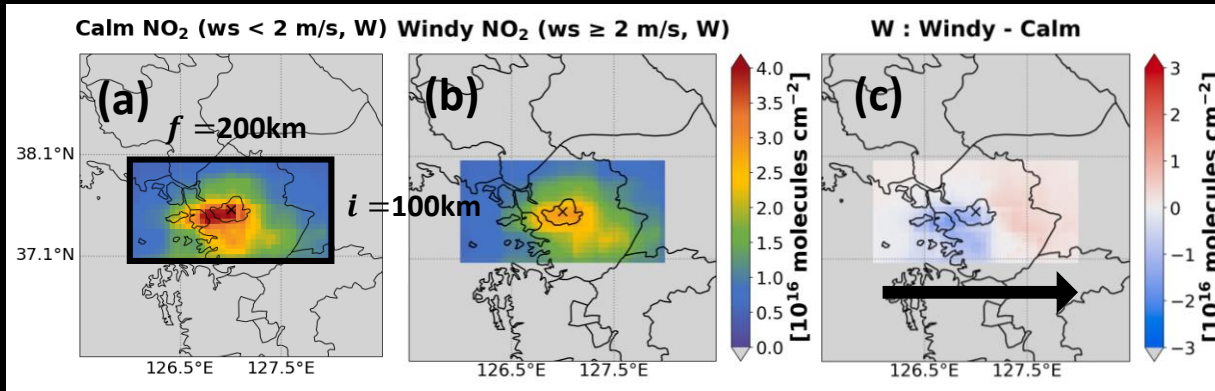
City		
1. Seoul	2. Yeosu	South Korea
3. Shanghai	4. Guangzhou	
5. Wuhan	6. Suzhou	China
7. Hangzhou	8. Shenyang	
9. Qingdao	10. Yinchuan	
11. Jinan	12. Chongqing	
13. Jiujiang	14. Xiamen	
15. Xiangyang	16. Changsha	
17. Tangshan	18. Tianjin	
19. Tokyo		
20. Bangkok		Thailand
21. Hanoi		Vietnam

- 21 cities are selected which have high NO<sub>2</sub> and located in GEMS scan area
- South Korea and Japan : Cities with high NO<sub>2</sub> concentrations are selected.
- China : Cities are selected as study area in previous studies. (Liu et al., 2016; Xue et al., 2022)
- Thailand and Vietnam : The capitals of each country are selected because of significant NO<sub>x</sub> emissions and minimal elevation difference.

(Main reference: Liu et al., 2016)

## 1) Mean tropospheric NO<sub>2</sub> VCD (vertical column density) distribution

- All NO<sub>2</sub> data within the analysis period are categorized by wind direction and wind speed for each grid, and then averaged (Cloud screening: cloud fraction < 30%).

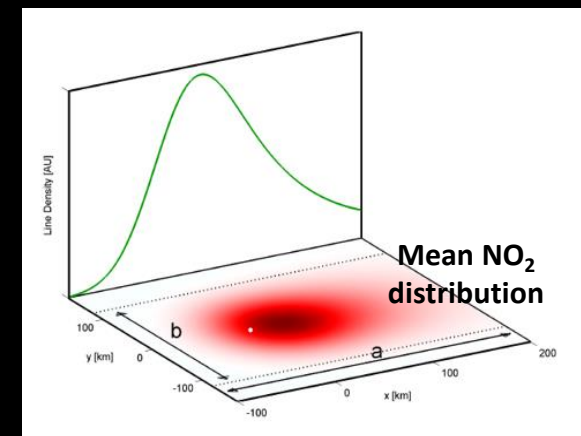


<b>wd</b>	N, S, E, W, NE, SW, NW, SE
<b>ws</b>	Calm wind / Windy condition (threshold : 2 m/s)

- (a) : calm wind condition
- (b) : windy condition
- (c) : (b)-(a)

## 2) NO<sub>2</sub> line density

- A curve calculated by integrating the mean NO<sub>2</sub> distribution along a direction (b) perpendicular to a specific wind direction (a).
- NO<sub>2</sub> line density is obtained under each wind direction and wind speed condition.
- A peak appears at the location where the emission source is located.

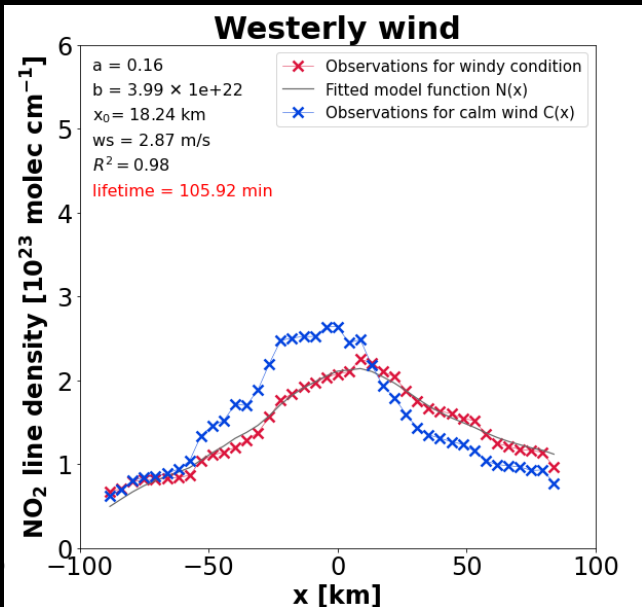


Beirle et al. (2011)



## 3) Effective NO<sub>x</sub> lifetime

- NO<sub>2</sub> line density calculated under calm wind conditions and windy conditions is used.
- A nonlinear least squares fit is performed by substituting the NO<sub>2</sub> line density under **calm wind conditions for C(x)** and parameters (a, b, x<sub>0</sub>) in the model function N(x).
- When the **fitted N(x)** takes the shape of the line density under windy conditions, a, b, and x<sub>0</sub> are fitted and provide information about the NO<sub>2</sub> distribution that changed by the wind.
- Specifically, x<sub>0</sub> represents the distance the plume has been transported (e-folding distance), and by substituting it into eq. (3), the effective lifetime (τ) can be calculated.



$$N(x) = a \times (e \otimes C)(x) + b \quad \text{----- (1)}$$

$$e(x) = \exp\left(-\frac{x-X}{x_0}\right) \text{ for } x \geq X, \quad 0 \text{ otherwise} \quad \text{--- (2)}$$

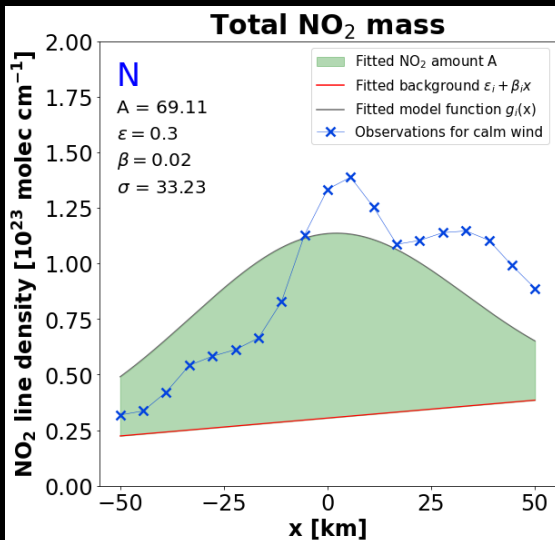
$$\tau = \frac{x_0}{w}, \quad w = w_{windy} - w_{calm} \quad \text{----- (3)}$$

a : scaling factor    b : offset    x<sub>0</sub> : e-folding distance    w : mean wind speed

⇒ Transport, chemical decay, and spatial smoothing are not considered separately. So τ is not exactly same with lifetime in the real atmosphere.

## 4) NO<sub>x</sub> total mass

- To minimize interference caused by the advection, only the NO<sub>2</sub> line density under calm wind conditions is used.
- The NO<sub>2</sub> line density is calculated for a 40×100 km<sup>2</sup> area in 21 cities, and line densities under 8 wind directions are fitted by the EMG (Exponentially Modified Gaussian) function from eq. (4).
- Fitted A means NO<sub>2</sub> total mass excepting background NO<sub>2</sub> ( $\epsilon_i + \beta_i x$ ). (detailed in next page)
- Scaling NO<sub>2</sub> to NO<sub>x</sub> : To calculate NO<sub>x</sub> total mass, fitted A is multiplied by [NO<sub>x</sub>]/[NO<sub>2</sub>] ratio (=1.32).
- NO<sub>x</sub> emission rate (NO<sub>x</sub> amount per unit time) = NO<sub>x</sub> total mass / lifetime ( $\tau$ )**

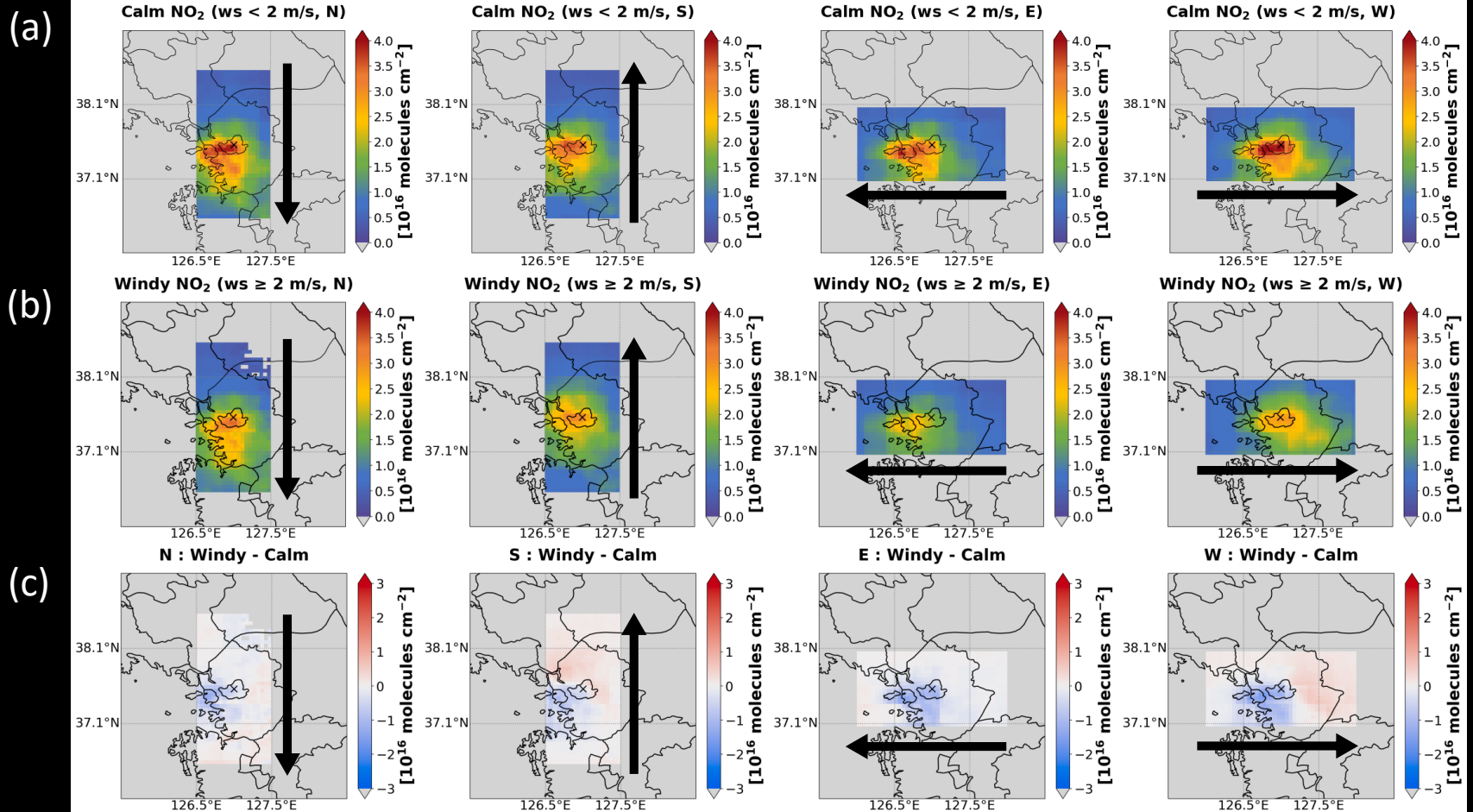


$$g_i(x) = A \times \frac{1}{\sqrt{2\pi}\sigma_i} \exp\left(-\frac{(x - X)^2}{2\sigma_i^2}\right) + \epsilon_i + \beta_i x \quad \text{----- (4)}$$

- # Blue line : NO<sub>2</sub> line density in the calm wind condition
- # Red line : background NO<sub>2</sub> ( $\epsilon_i + \beta_i x$ ) - Result of linear regression using NO<sub>2</sub> line density calculated up to the 5th percentile
- # Gray line : fitted  $g_i(x)$
- # Green area : fitted A = NO<sub>2</sub> total mass



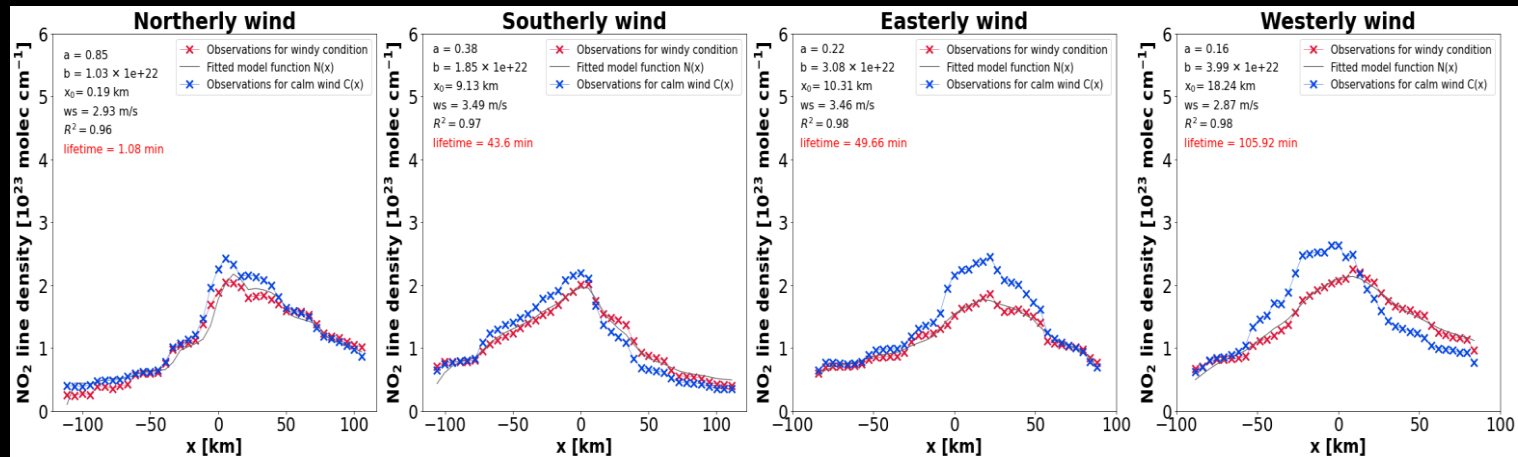
# Results : Mean NO<sub>2</sub> distribution



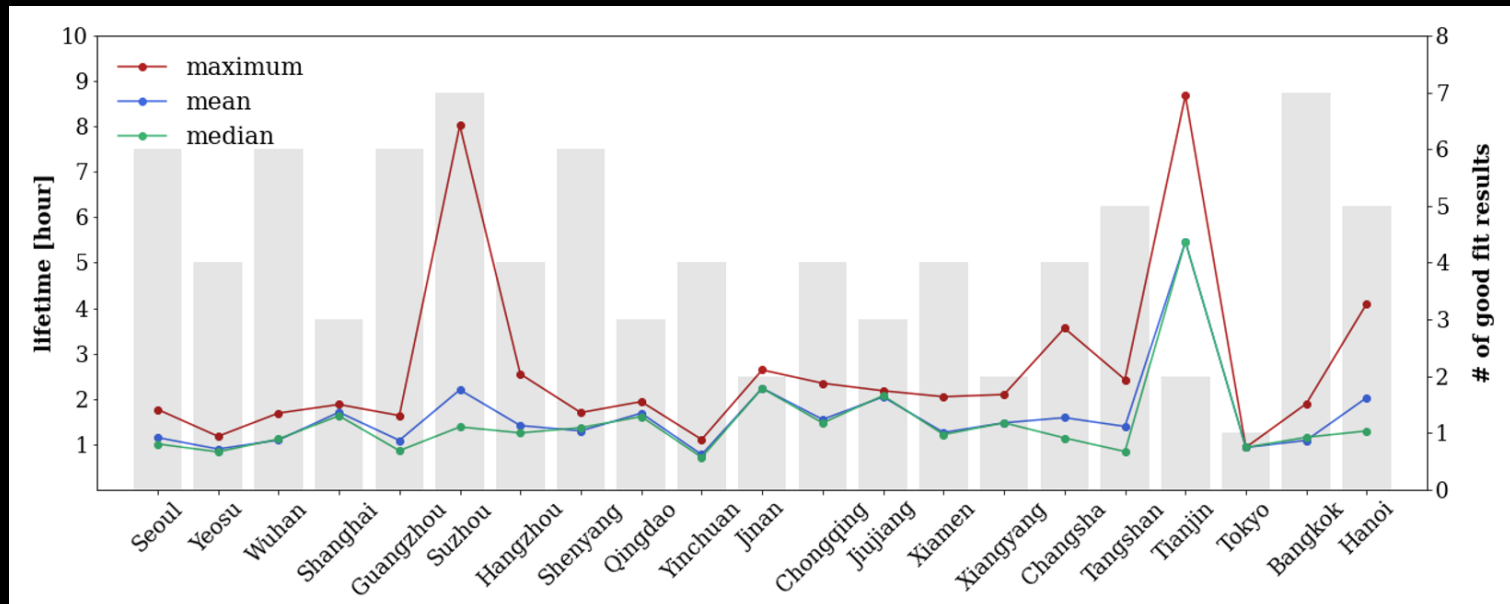
- (a) : The result under calm wind condition for N, S, E, W wind.
- (b) : The result under windy condition for N, S, E, W wind.
- (c) : (b)-(a). According to (c), the movement of the NO<sub>2</sub> plume in the troposphere can be observed following the wind direction (black arrows).
- (a) and (b) are used to calculate line densities.

# Results : effective $\text{NO}_x$ lifetime

## Fitting result under N, S, E, W wind in Seoul, 2022

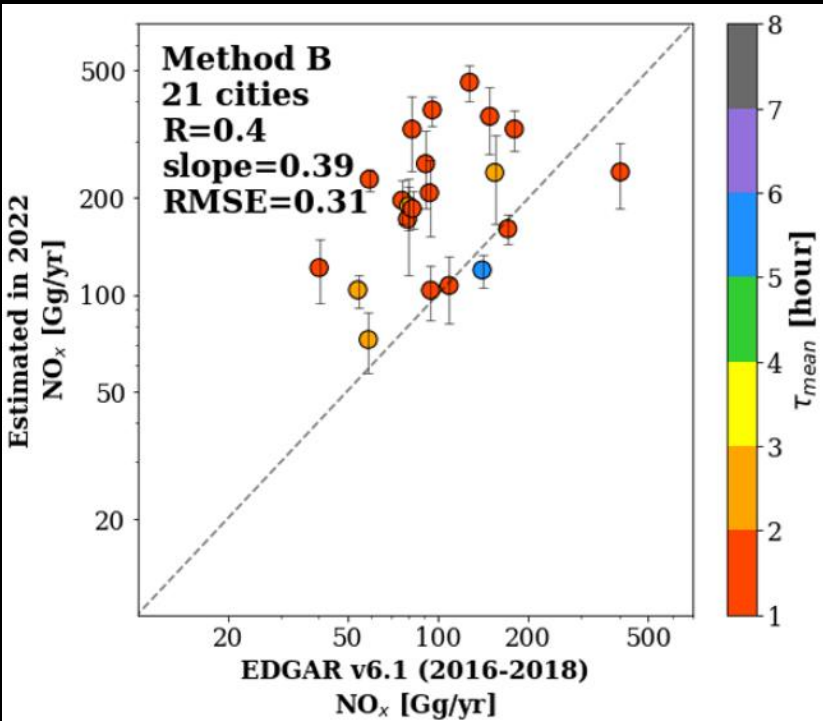


Estimated effective e-folding lifetime (Maximum, mean, and median among total 8 windy cases)

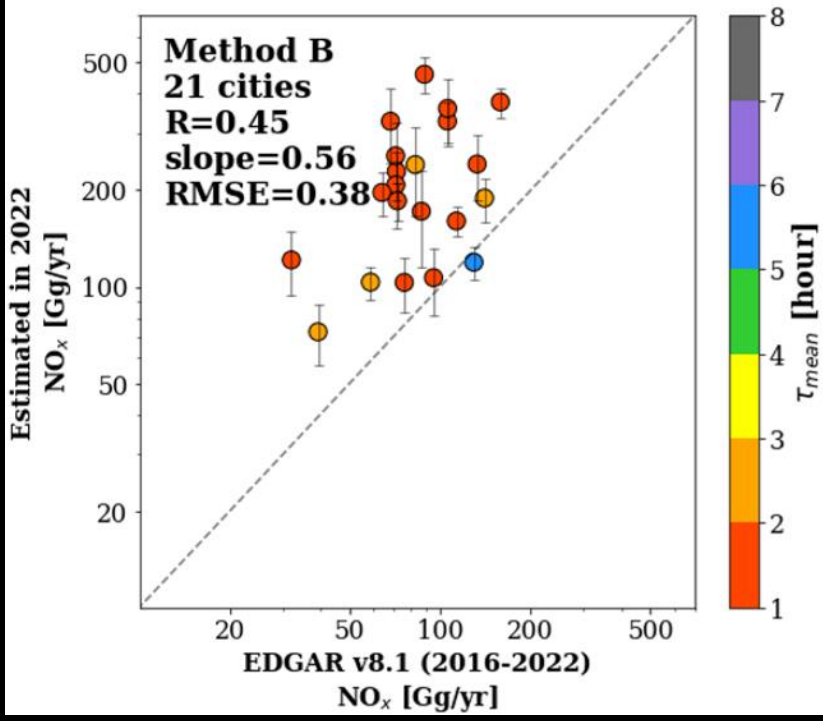


# Results : NO<sub>x</sub> emission rate (compared with EDGAR)

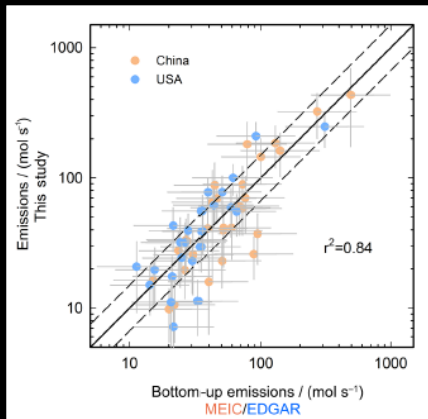
### With EDGAR v6.1 (mean 2016-2018)



### With EDGAR v8.1 (mean 2016-2022)



- Positive correlation is found between the EDGAR and our GEMS-based estimation.
- GEMS-based NO<sub>x</sub> emission shows the overestimation. This will be redeemed better when new GEMS NO<sub>2</sub> product is applied.



### Reducing the GEMS (and TEMPO) NO<sub>2</sub> vs. Enhancing the TROPOMI NO<sub>2</sub>

- I am not able to say which one is better, which one is not. (Qualitatively all products are good to use for the air quality diagnosis. But quantitatively?)
- In the context of methodology used in this study,
  - GEMS NO<sub>2</sub> VCD - based estimated NO<sub>2</sub> emission is **larger than** EDGAR NO<sub>2</sub> emission (my work)
  - TROPOMI NO<sub>2</sub> VCD - based estimated NO<sub>2</sub> emission is **comparable to** EDGAR NO<sub>2</sub> emission (e.g., Liu et al., 2016)
- And new version (improvement assumed) of EDGAR (ver 8.1) shows **lower** emission than old version (ver 6.1) (At least in the range of this study. Need to see more).

Thanks for your attention