

Validation of the capability of WRF-Chem model and CAMS to simulate near surface atmospheric CO₂ mixing ratio for the territory of Saint-Petersburg

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Why the monitoring of CO₂ urban emissions is important today?

1. CO₂ – the main anthropogenic greenhouse gas (GHG) - influences the radiation balance of the Earth leading to an increase in tropospheric air temperature
2. CO₂ content in the atmosphere keeps rising due to man-made activity
3. Megacities have essentially determined (~70%) of the anthropogenic CO₂ emissions in the last few decades

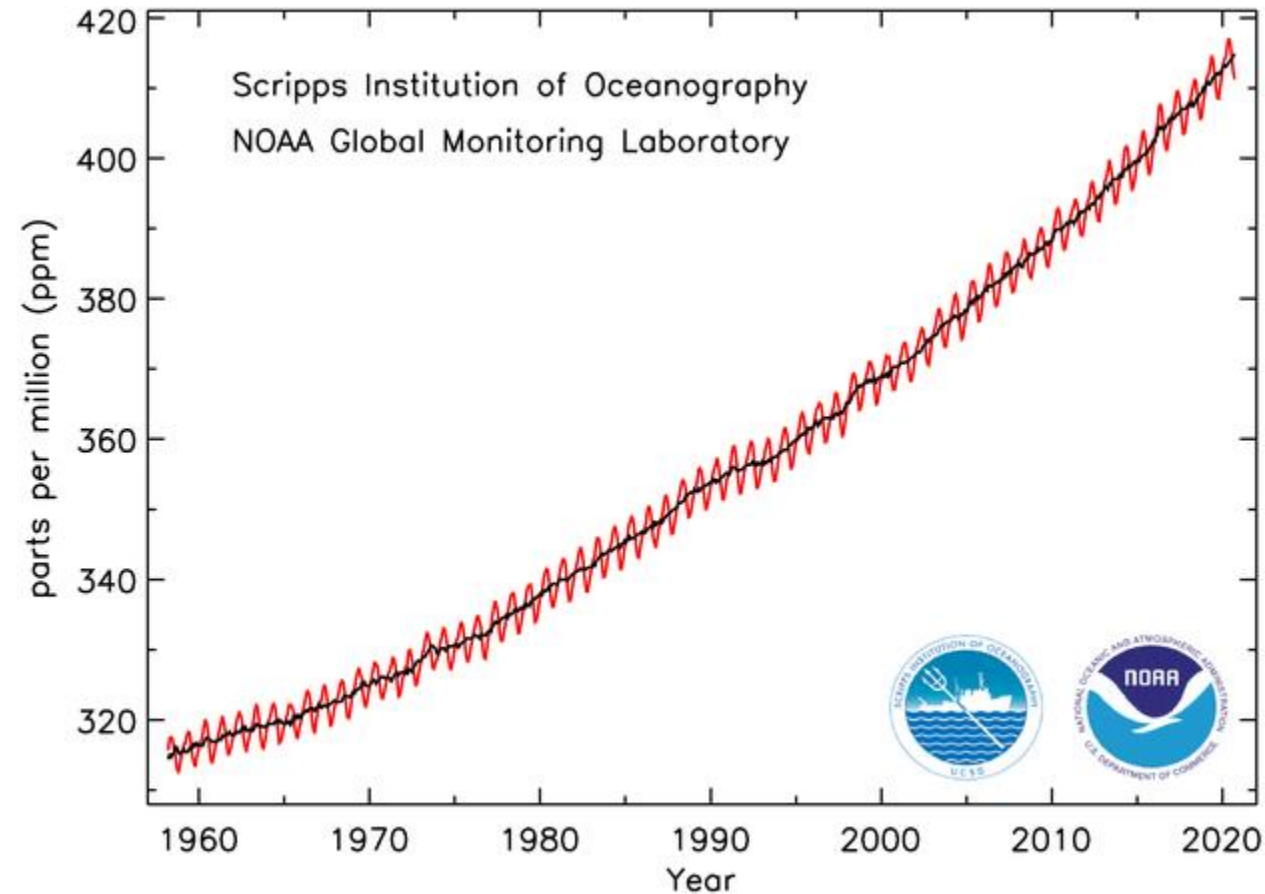


Fig.1 Trend of near-surface atmospheric CO₂ mixing ratio for the period 1958-2020 at Mauna Loa Observatory, Hawaii, USA

Original from <https://www.esrl.noaa.gov/gmd/ccgg/trends/>

How can we estimate CO₂ urban emissions?

1. GHGs Inventories (Bottom-up)

- Based on documented data of potential CO₂ sources (amount of fossil fuel used, number of active cement manufactures, etc.)

Inaccuracies can reach 50% and more!

2. Inverse modelling (Top-down)

Observation data

- In-situ
- Remote

+

Modelling of CO₂ transport
in the atmosphere

- 3-D numerical chemistry transport modelling;
- Lagrangian dispersion models;
- Box models;
- Others

+

A priori information

- CO₂ sources/sinks
- Initial and boundary conditions

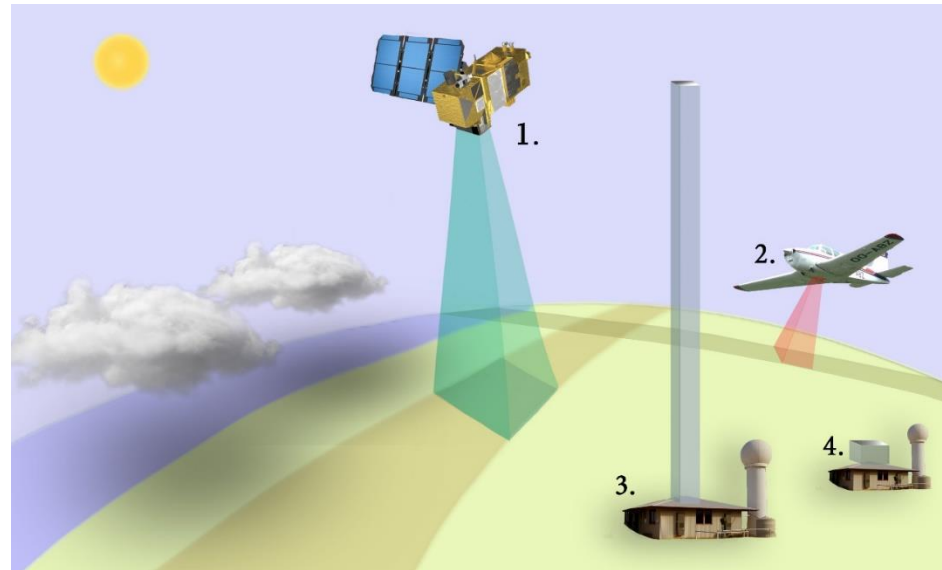


Fig.2 Main methods of atmospheric observations
(1 – satellite, 2 – airplane, 3 – remote ground-based, 4- in-situ)

Methods: CO₂ in-situ observations in Peterhof

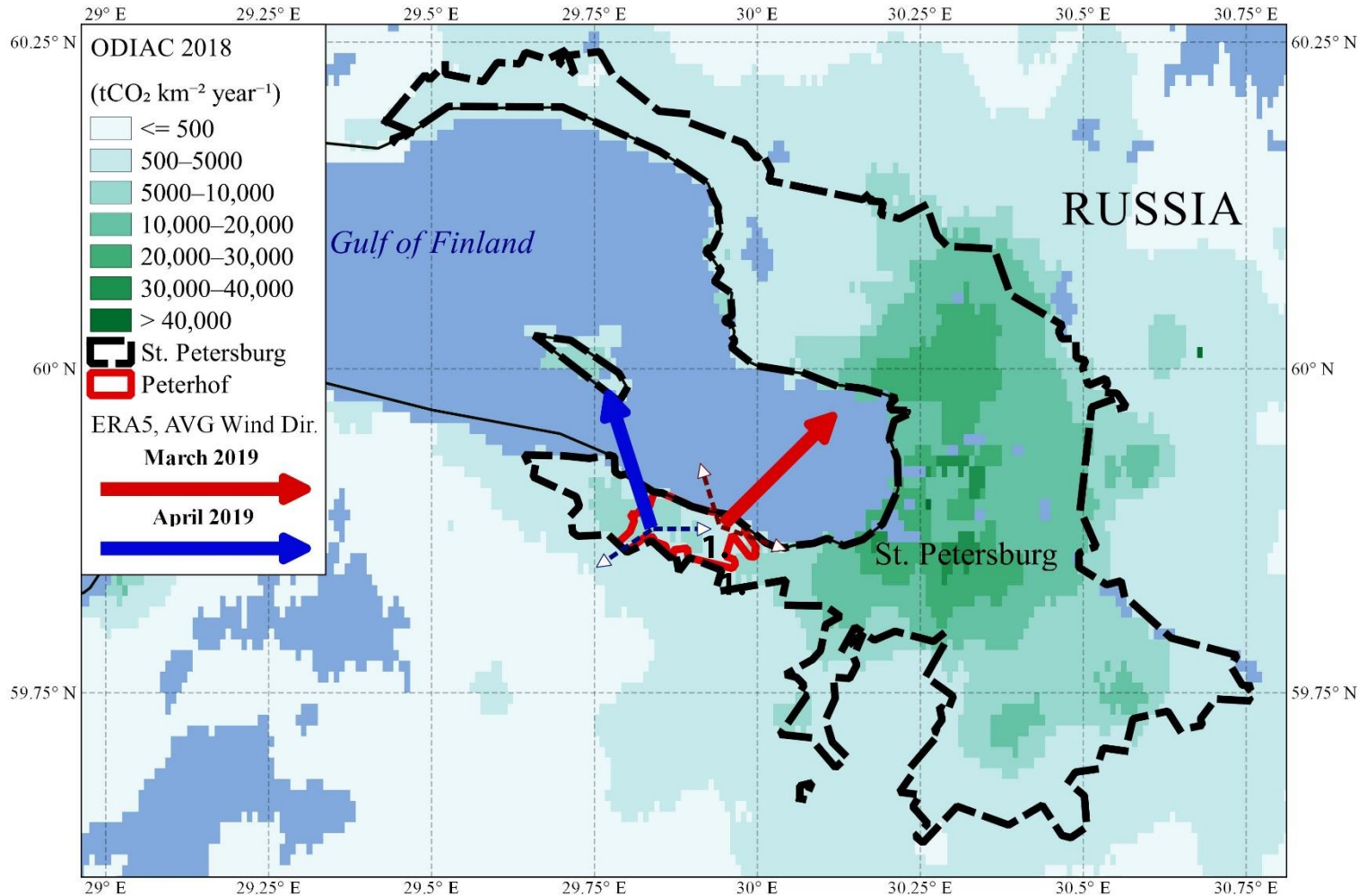


Fig.3 Territory of Saint-Petersburg (black dashed line) and Peterhof (1)

Instrument:

- Los Gatos Research Greenhouse Gas Analyzer (LGR GGA-24-r-EP) at SPbU Faculty of Physics.

Station and observation data:

- Coordinates - 59.88°N, 29.83°E;
- Height – ~ 30 m AMSL (~ 6 m AGL);
- Measurement error - 50-150 ppb depending on accumulation time (100-5 s respectively);
- Data processing – averaging of 15-minutes medians.

Methods: CO₂ in-situ observations in Peterhof

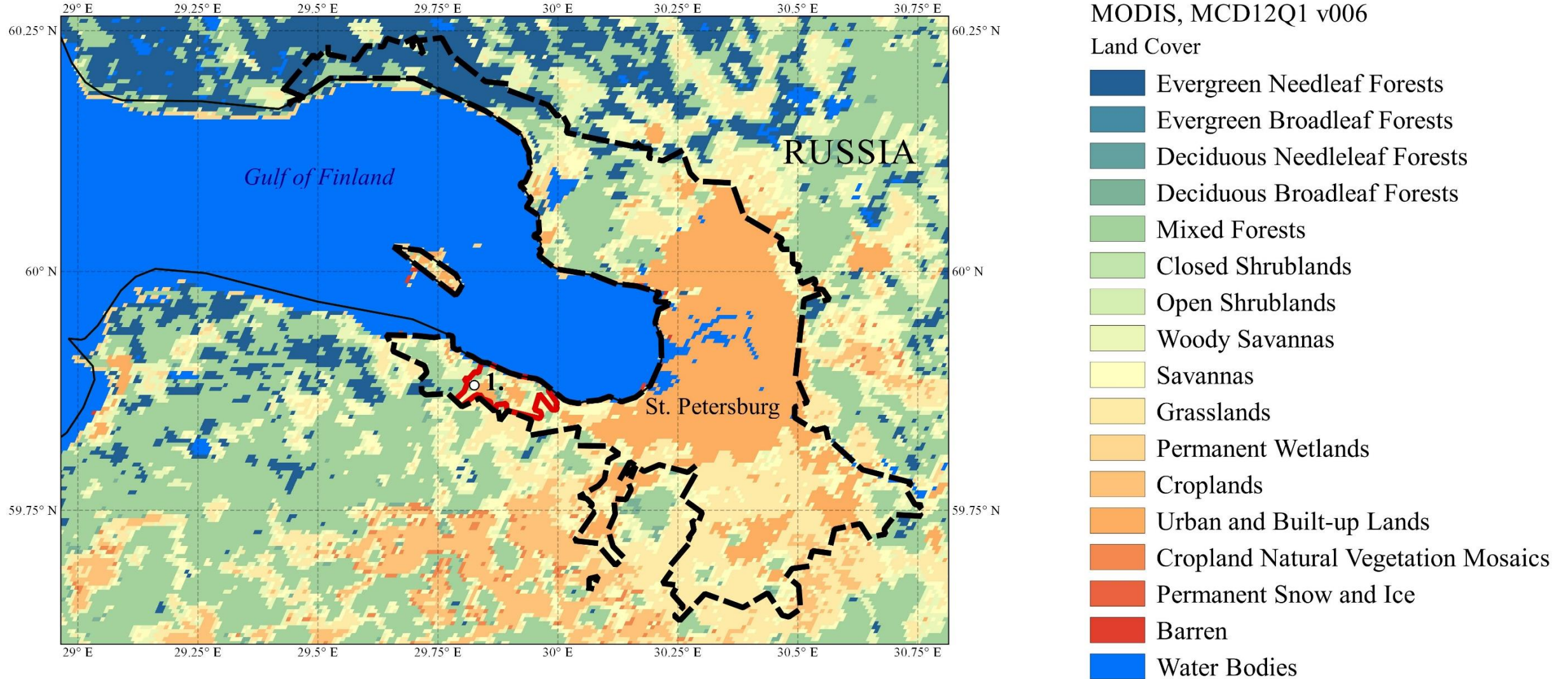


Fig.4 Territory of Saint-Petersburg (black dashed line) and Peterhof (1); the white circle depicts the position of the Peterhof measurement station

Methods: numerical modelling of CO₂ atmospheric transport

Copernicus **A**tmosphere **M**onitoring **S**ervice (CAMS)

- Product type - reanalysis and analysis of the spatio-temporal variation in CO₂ on a global scale;
- Horizontal resolution – 1.9 and 3.8° (reanalysis), 0.15° (analysis);
- Data assimilation – only reanalysis.



Weather **R**esearch and **F**orecasting – **C**hemistry (WRF-Chem)

- Numerical weather prediction and atmospheric chemistry transport model on regional scale;
- Spatial resolution – from tens to ~ 1 km;
- Ability to consider time-varying fluxes of gases.



Methods: WRF-Chem modelling

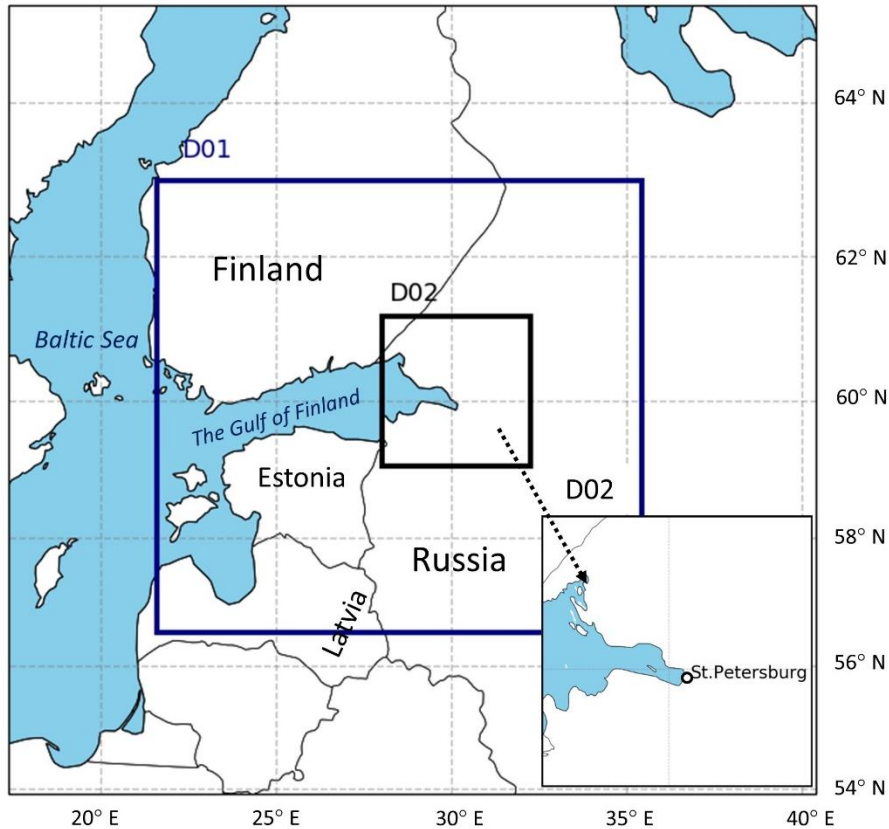


Fig. 5. Modelling domains

Table 1. The main characteristics of the WRF-Chem runs

No of WRF-Chem Model Run		1a	1b	2a	2b	3a	3b
Horizontal resolution		D01—9 km, D02—3 km					
Vertical resolution		39 hybrid vertical layers (up to 50 hPa)					
Initial and boundary conditions	Meteorology	GFS ANL (0.5°, 3 h)					
	Atmospheric CO ₂ mixing ratio	CAMS Global analysis of CO ₂ (0.15°, 6 h)					
Length of simulation		March 2019	April 2019	March 2019	April 2019	March 2019	April 2019
CO ₂ sources and sinks	Anthropogenic emissions (1)	ODIAC 2018, diurnal temporal variation				ODIAC 2018, no temporal variation	
	Biogenic fluxes (2)	VPRM, temporal variation—3 h		No biogenic fluxes		No biogenic fluxes	

1. **Open-source Data Inventory for Anthropogenic CO₂ (ODIAC)** – CO₂ anthropogenic emissions inventory of high spatial resolution (~1 km) (Oda & Maksyutov, 2015).

2. **Vegetation Photosynthesis and Respiration Model (VPRM)** – model of CO₂ biogenic fluxes (Mahadevan et al., 2008).

CAMS vs in-situ measurements of CO₂

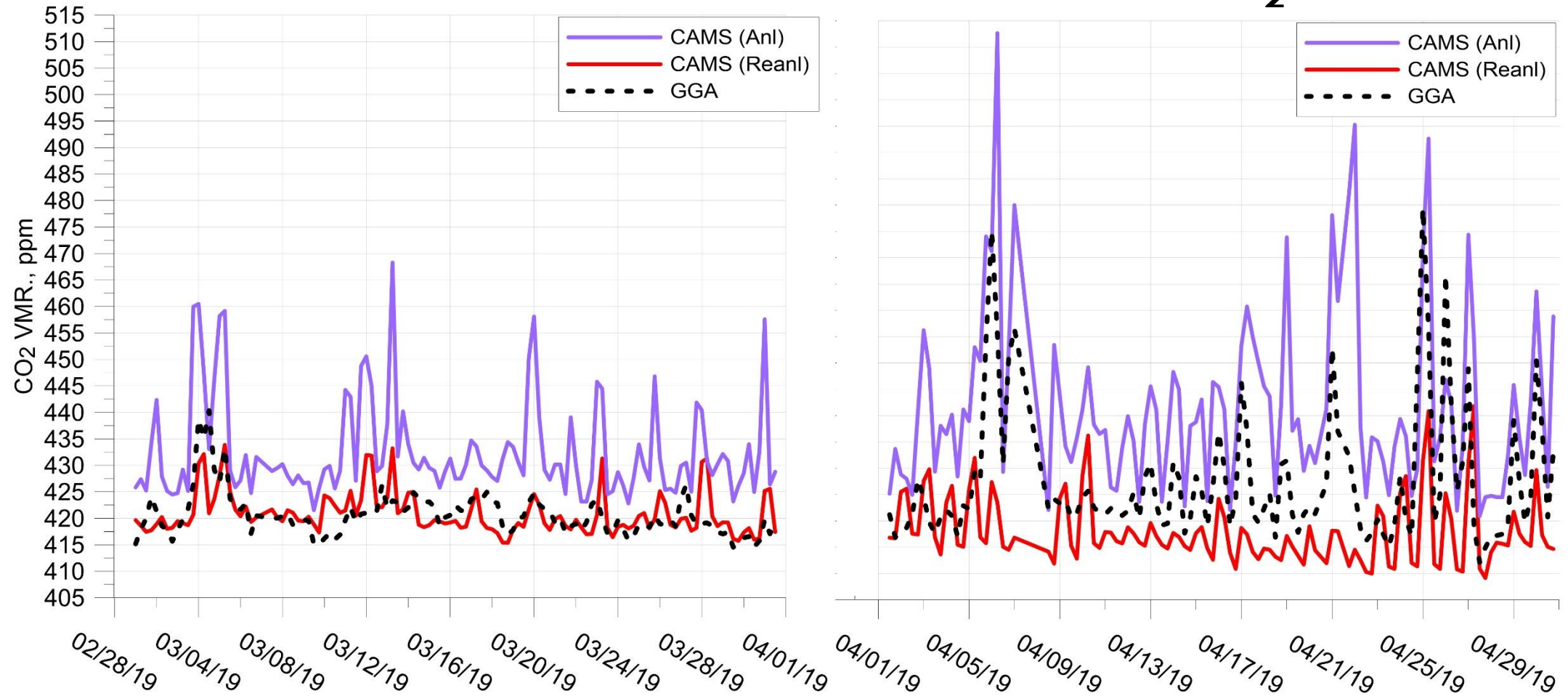


Fig.6 Temporal variation (6 h) in the surface atmospheric CO₂ mixing ratio according to the CAMS and in-situ observation data in Peterhof in March (left) and April (right) 2019

CAMS vs in-situ measurements of CO₂

Table 2. Statistical characteristics of the difference between the CAMS and observation data for Peterhof in March and April 2019; Reanl: reanalysis; Anl: analysis

Period	March 2019		April 2019	
	GGA–CAMS (Reanl)	GGA–CAMS (Anl)	GGA–CAMS (Reanl)	GGA–CAMS (Anl)
M, ppm	–0.1	–11.8	9.4	–14.3
RMSD, ppm	4.2	14.4	15.1	19.0
R	0.46 ± 0.16	0.52 ± 0.15	0.37 ± 0.18	0.69 ± 0.14

CAMS Reanalysis – observation assimilation

CAMS Analysis – **no** observation assimilation

WRF-Chem vs in-situ measurements of CO₂

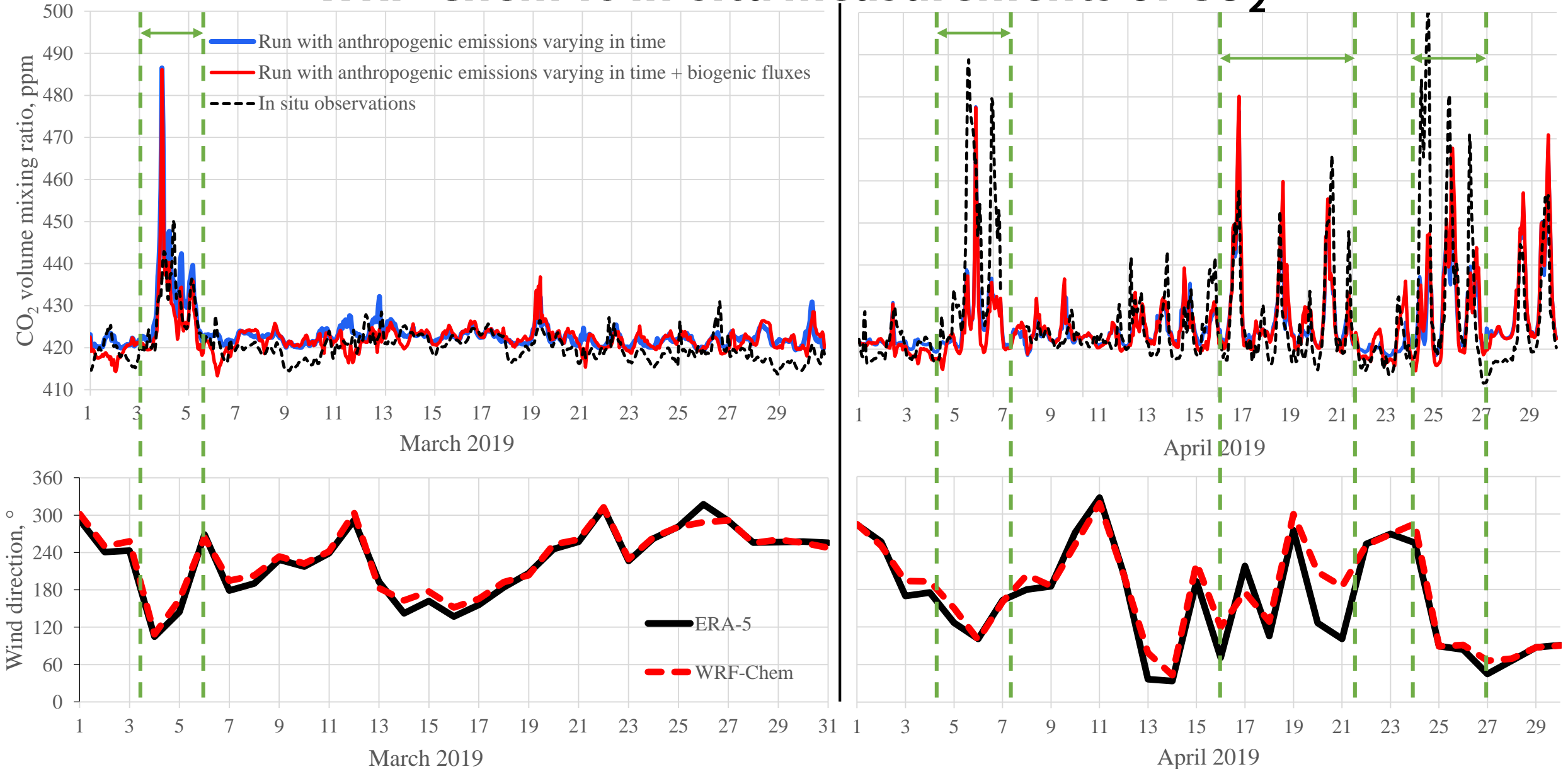


Fig.7 Temporal variation (1 h) in the surface atmospheric CO₂ mixing ratio according to the WRF-Chem runs and in-situ observation data (top) and daily average wind direction in near-surface level according to ERA5 and WRF-Chem data (bottom) in Peterhof in March (left) and April (right) 2019

WRF-Chem vs in-situ measurements of CO₂

Table 3. Statistical characteristics of the difference between the WRF-Chem and observation data for Peterhof in March-April 2019; t.v. - temporal variation, const. - time constant

Period	March 2019			April 2019			March–April 2019		
GGA– WRF- Chem	t.v. Anth + Bio	t.v. Anth	Const. Anth	t.v. Anth + Bio	t.v. Anth	Const. Anth	t.v. Anth + Bio	t.v. Anth	Const. Anth
M, ppm	–1.7	–2.7	–2.7	1.3	1.2	1.0	–0.3	–0.8	–0.9
RMSD, ppm	4.7	4.6	4.6	11.5	11.4	11.6	8.7	8.6	8.8
R	0.55 ± 0.06	0.69 ± 0.05	0.70 ± 0.05	0.60 ± 0.06	0.60 ± 0.06	0.58 ± 0.06	0.61 ± 0.04	0.62 ± 0.04	0.61 ± 0.04

Statistics of the three WRF-Chem runs modelled data look pretty much the same!

WRF-Chem vs in-situ measurements of CO₂

Monthly average **diurnal variation** shows the **distinguishable differences** between the **three WRF-Chem runs**.

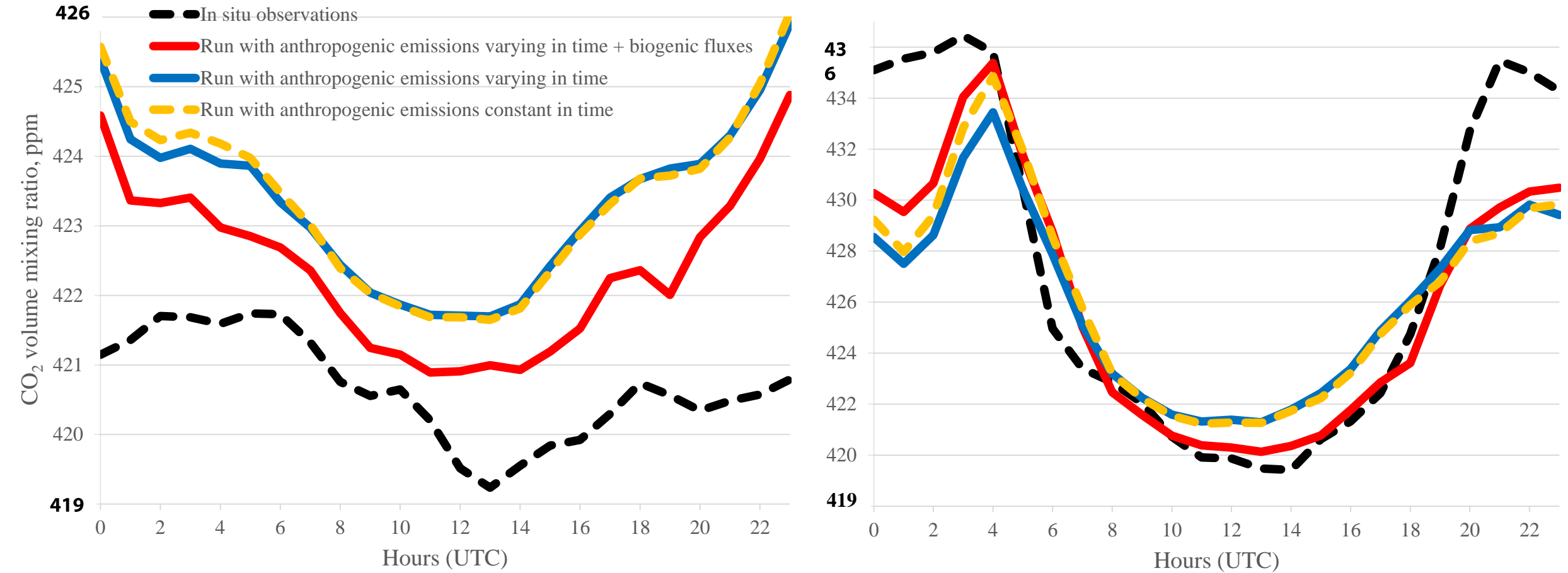


Fig.8 Monthly-averaged diurnal variation of near-surface CO₂ mixing ratio according to the WRF-Chem runs and in situ observations in Peterhof in March (a) and April (b) 2019.

WRF-Chem run with **anthropogenic** and **biogenic CO₂ fluxes** had **the best fit** with the **observations**.

Conclusions

1. Differences between the CAMS data and in-situ observations of CO₂ surface-level mixing ratio in Peterhof in March and April 2019 varied from 4.3 to 19.0 ppm depending on CAMS product type and meteorological conditions. The data may be used in inverse modelling of CO₂ anthropogenic emissions on the scales larger than Saint-Petersburg territory;
2. The WRF-Chem adequately simulated the temporal variation in the near-surface CO₂ mixing ratio on a high spatial resolution (3 km) in Peterhof (Saint Petersburg) in March and April 2019 - differences and correlation coefficients were in range 4.6-11.6 ppm and 0.6-0.7 respectively depending on meteorological conditions;
3. Using the biogenic CO₂ fluxes made WRF-Chem data to fit the in-situ observations a bit better;
4. According to the WRF-Chem modelling the main factors determining the near-surface CO₂ mixing ratio in Peterhof in March and April 2019 were the wind in the surface layer and the anthropogenic CO₂ emissions from the Saint Petersburg urban area;
5. To investigate whether the WRF-Chem model is suitable for the inverse modelling of the CO₂ anthropogenic emissions from the territory of Saint Petersburg, the analysis of modelled CO₂ total column is needed.

Reference

Nerobellov G, Timofeyev Y, Smyshlyaev S, Foka S, Mammarella I, Virolainen Y. Validation of WRF-Chem Model and CAMS Performance in Estimating Near-Surface Atmospheric CO₂ Mixing Ratio in the Area of Saint Petersburg (Russia). *Atmosphere*. 2021; 12(3):387. <https://doi.org/10.3390/atmos12030387>

Thank you!