

Supplementary Material

Manuscript title: Methane emissions in boreal forest fire regions: Assessment of five biomass biomass-burning emission inventories based on carbon sensing satellites

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S1 Detrended Fluctuation Analysis calculation steps and formulas

Principle:

Data trend removal refers to subtracting an optimal (least squares) fitting line, plane, or surface from the data, so that the mean of the data after trend removal is zero [1].

General steps:

1. Calculate the cumulative deviation of a time series $x(t)$ with a length of N , $t=1, 2, \dots, N$, and Calculate its cumulative deviation and convert it into a new sequence:

$$y(t) = \sum_{i=1}^t [x(i) - \bar{x}] \quad (1)$$

In the equation, \bar{x} is the average of the time series: $\bar{x} = \frac{1}{N} \sum_{t=1}^N x(t)$

2. Divide $y(t)$ into m non-overlapping intervals of equal length n , where n is the interval length, i.e., the time scale, and m is the number of intervals (or windows), which is the integer part of N/n .

3. Use the least squares method to linearly fit the local trend $y_n(t)$ for each segment of the sequence.

4. Remove the local trend of each interval for $y(t)$ and calculate the root mean square of the new sequence:

$$F(n) = \sqrt{\frac{1}{N} \sum_{t=1}^N [y(t) - y_n(t)]^2} \quad (2)$$

5. Change the size of the window length and repeat steps (2), (3), and (4).

[1] Xu, Z.; Wan, J.W.; Li, G.; Su, F. Target Detection within Sea Clutter Based on Multifractal Detrended Fluctuation Analysis. In Proceedings of the SMART TECHNOLOGIES FOR COMMUNICATION, 2012; pp. 259-262.

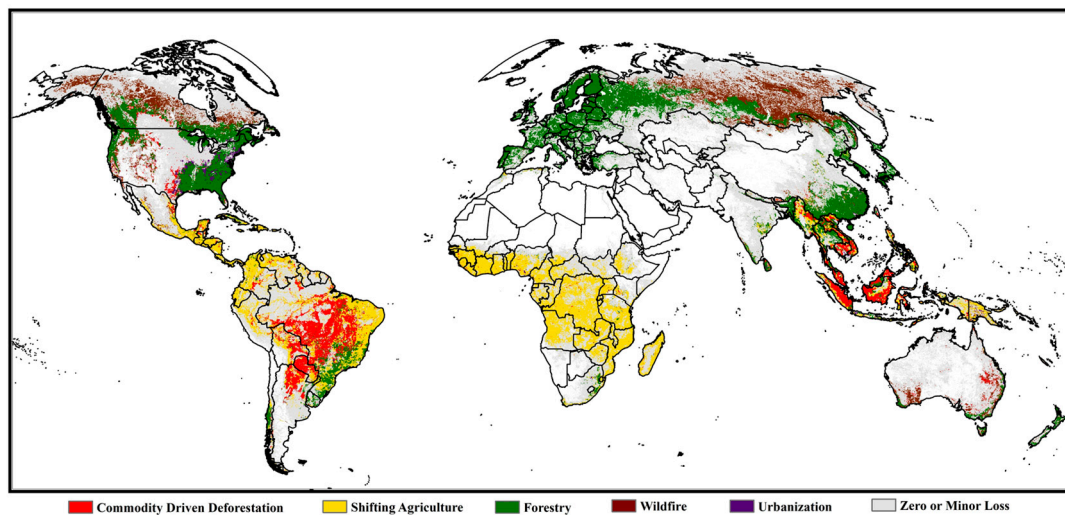


Figure. S1 Global forest disturbance drivers map.

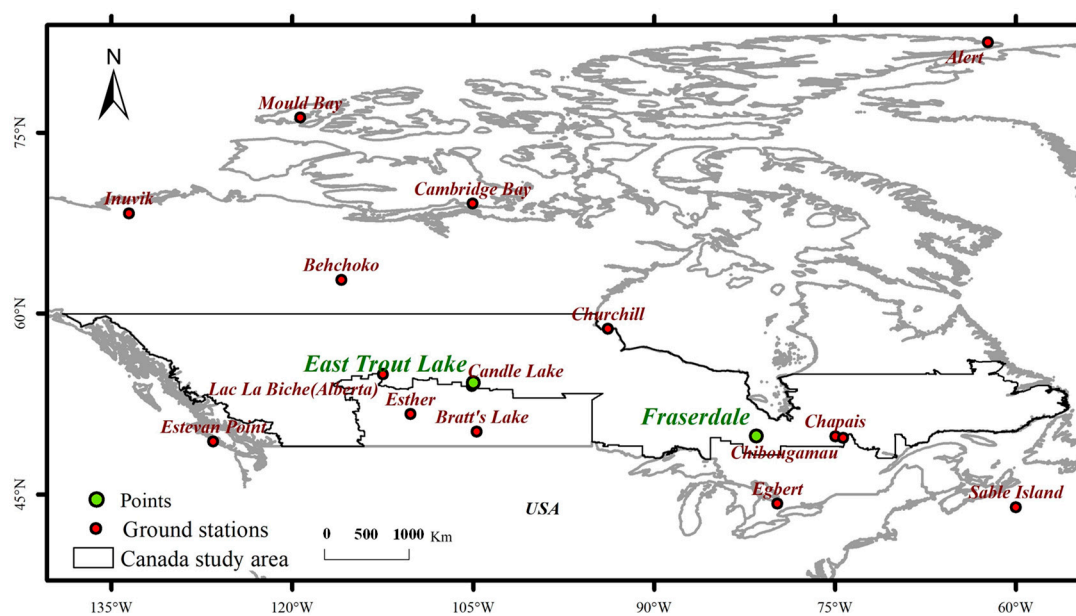


Figure. S2 Distribution map of CAN regional stations.

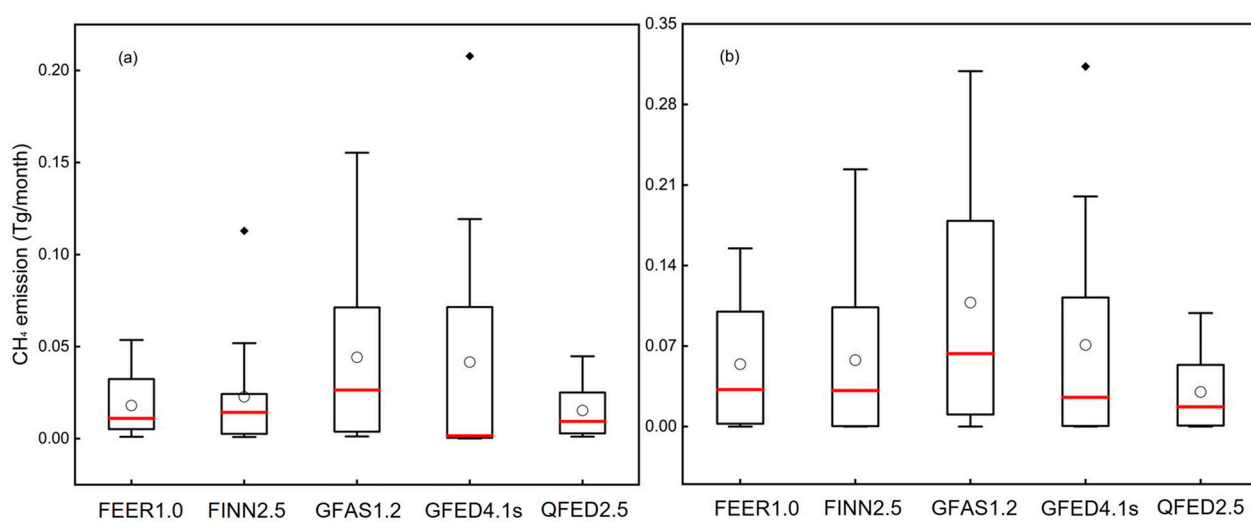


Figure. S3 Seasonal box chart of CH₄ emissions from 2010 to 2020 (black hollow circle: mean value, red solid line: median value, black diamond: Outlier): (a) CAN; (b) RUS.

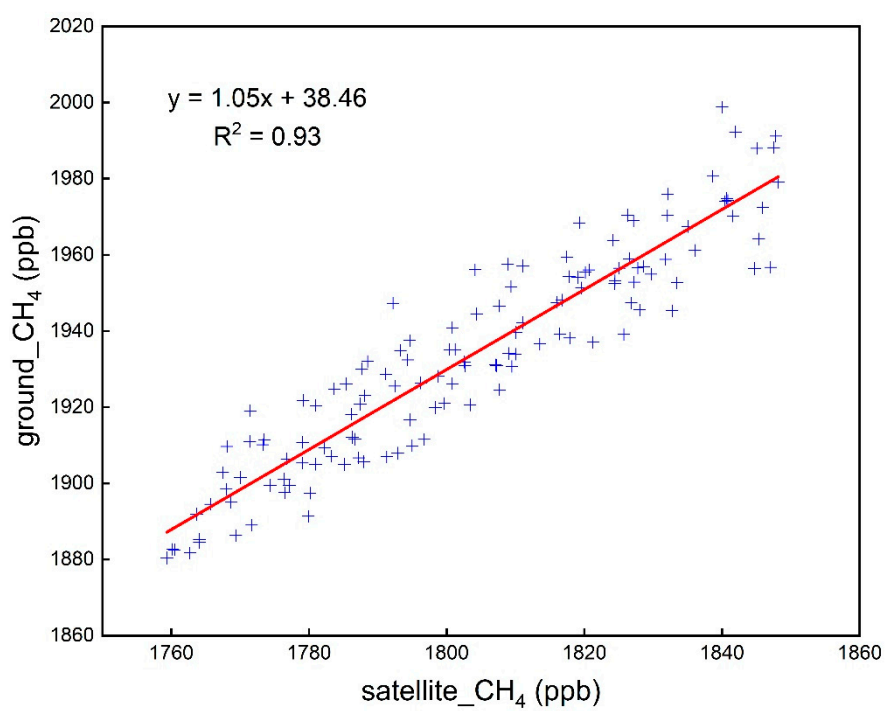


Figure. S4 Ground station data and observed remote sensing atmospheric concentration data.

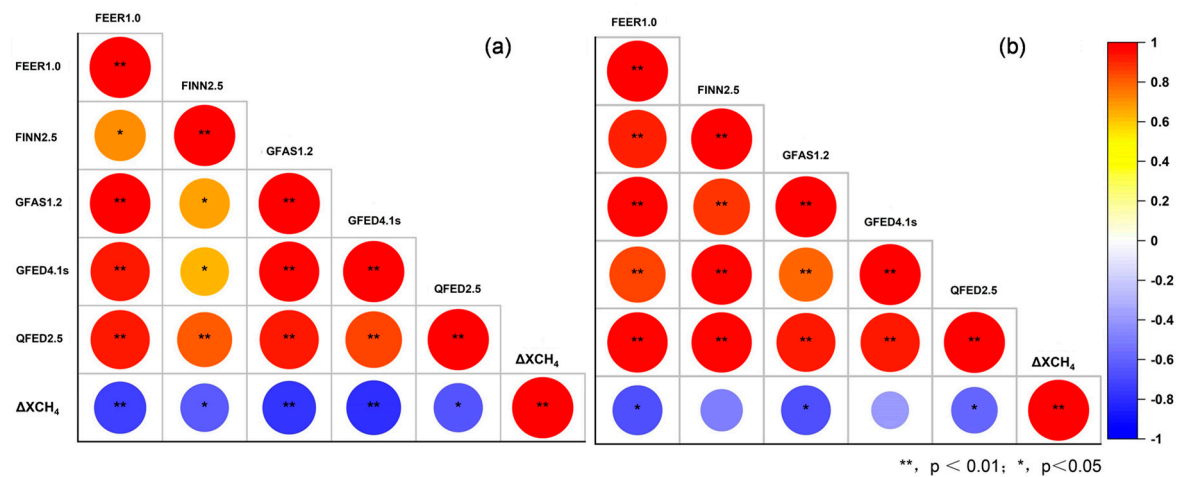


Figure. S5 Correlation analysis between inventory dataset and atmospheric CH₄ concentration data from 2010 to 2020: (a) CAN; (b) RUS.

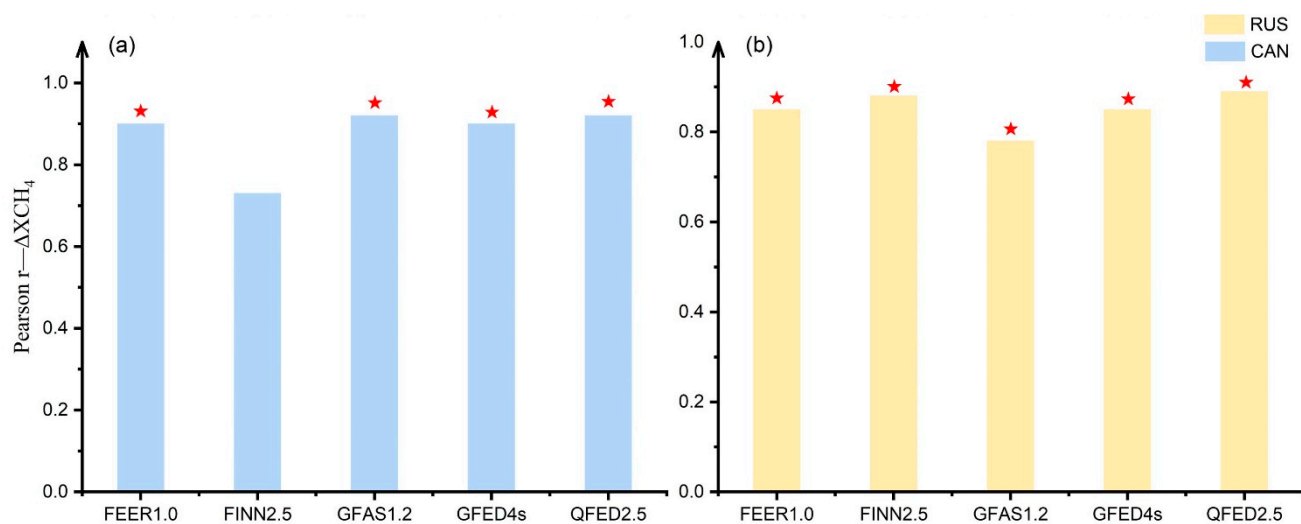


Figure. S6 Correlation number between atmospheric concentration change and gas emission under optimal lag number (red star represents $p < 0.01$): (a) CAN; (b) RUS.

Table. S1 Data collection details.

Data classification	Data name	Product name	Spatial resolution	Collection time range / Time resolution	Unit	Data format	Data sources
CH ₄ Satellite data	Mapping-XCH ₄	Atmospheric methane column concentration	0.5°	2010/01~2020/12 / Monthly	ppb	TIFF	https://data.casearth.cn/
Site data	CH ₄	Ground station data	/	2010/01~2020/12 / Monthly	ppb	TXT	https://gaw.kishou.go.jp/
GBBEI	GFED4.1s	CO ₂ /CH ₄ emission	0.25°	2010/01~2020/12 / Daily	kg/m ² /s	HDF5	https://www.geo.vu.nl/~gwerf/GFED/GFED4/
	FINN2.5		0.1°		molecules/cm ² /s	NetCDF	https://rda.ucar.edu/datasets/ds312.9/
	GFAS1.2		0.1°		kg/m ² /s		https://apps.ecmwf.int/datasets/data/cams-gfas/
	FEER1.0		0.1°		kg/m ² /s		https://feer.gsfc.nasa.gov/data/
	QFED2.5		0.1°		kg/m ² /s		http://ftp.as.harvard.edu/gcgrid/data/ExtData/HEMCO/QFED/
Auxiliary data	MCD12Q1.061	Land cover type data	500m	2015	/	TIFF	https://developers.google.com/earth-engine/datasets/catalog/MODIS_061_MCD12Q1

Table. S2 Characteristics of two atmospheric background stations and MLO stations in the CAN region.

Sites	Location	Sample	Underlying surface	Altitude/m	Observation time
ETL	54.23°N,104.99°W	bottle	forest	492	2005-now
FSD	49.84°N,81.52°W	bottle	forest	210	1999-now
MLO	19.54°N,155.58°W	bottle	bare land	3397	1984-now

Table. S3 Summary of five biomass burning emission inventories

inventory	Burned area/FRP product	Spatial resolution	Time resolution	Time range	Reference
(a) Burned-area-based approaches					
GFED4.1s	MODIS MCD64A1	0.25° × 0.25°	3-hourly, Daily, Monthly	1997-	Giglio et al. (2013)
FINNv2.5	MCD14DL	0.1° × 0.1°	Daily	2002-2021	Wiedinmyer et al. (2011)
(b) FRP-based approaches					
GFASv1.2	Assimilation of level 2 MOD14 and MYD14 FRP	0.1° × 0.1°	Daily	2003-	Whitburn et al. (2015)
FEERv1.0	From GFASv1.2	0.1° × 0.1°	Daily, Monthly	2003-	Ichoku et al. (2014)
QFEDv2.5	Level 2 fire products MOD14/MYD14	0.1° × 0.1°	Daily, Monthly	2000-	Koster et al. (2015)

Table. S4 Average CH₄ emissions and standard deviation in CAN and RUS regions from 2010 to 2020.

Area	Inventory	Period	Average (Tg CH ₄ year ⁻¹)	1 Standard Deviation (Tg CH ₄ year ⁻¹)	Coefficient of Variation (%)
CAN	FEER1.0	2010-2020	0.22	0.09	41.03
	FINN2.5		0.24	0.15	60.70
	GFAS1.2		0.52	0.29	54.53
	GFED4.1s		0.40	0.23	58.29
	QFED2.5		0.18	0.07	38.96
	Average		0.31	0.17	50.70
RUS	FEER1.0	2010-2020	0.70	0.32	45.13
	FINN2.5		0.74	0.43	58.68
	GFAS1.2		1.27	0.79	62.01
	GFED4s		0.71	0.43	60.91
	QFED2.5		0.36	0.16	43.77
	Average		0.76	0.43	54.10