

Search ORNL DAAC

Search

DAAC Home > Get Data > NASA Projects > Arctic-Boreal Vulnerability Experiment (ABOVE) > User guide

Machine learning-based Arctic-boreal terrestrial ecosystem CO₂ fluxes, 2001-2020

Get Data

Documentation Revision Date: 2024-10-03

Dataset Version: 1

Summary

This dataset provides gridded estimates of gross primary productivity (GPP), ecosystem respiration (Reco), and net ecosystem CO₂ exchange (NEE) across the circumpolar terrestrial Arctic-boreal region at a 1-km spatial resolution. Monthly CO₂ flux data from 2001 to 2020 were generated using terrestrial eddy covariance and chamber CO₂ flux observations, combined with geospatial meteorological, remote sensing, topographical and soil data, all within a random forest modeling framework. Aggregated average annual NEE, average annual NEE with direct fire emissions added based on the Global Fire Emissions Database (GFED) product, and temporal trends in annual NEE rasters over 2002-2020 are also included. The data are provided in NetCDF and GeoTIFF formats.

This dataset holds 60 NetCDFs and three files in cloud optimized GeoTIFF (COG) format.

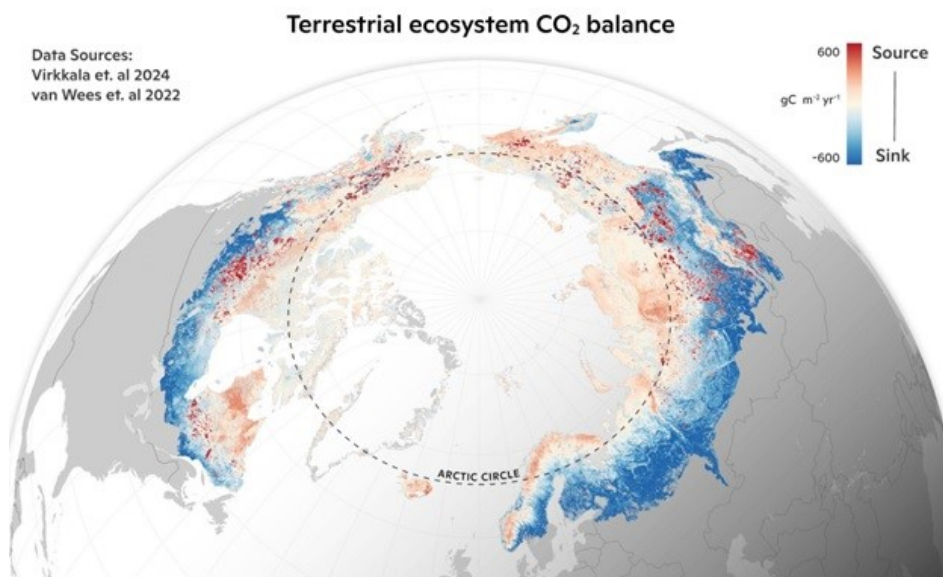


Figure 1. Annual average terrestrial net ecosystem exchange (NEE) and direct fire emissions from 2002-2020 showing areas with net CO₂ sinks (i.e. areas taking up CO₂ from the atmosphere) and net CO₂ sources (i.e. releasing net CO₂ emissions to the atmosphere) (Virkkala et al., 2024).

Citation

Virkkala, A-M., B.M. Rogers, J.D. Watts, K. Arndt, S. Potter, I. Wargowsky, and S. Natali. 2024. Machine learning-based Arctic-boreal terrestrial ecosystem CO₂ fluxes, 2001-2020. ORNL DAAC, Oak Ridge, Tennessee, USA. <https://doi.org/10.3334/ORNLDAAC/2377>

Table of Contents

1. Dataset Overview
2. Data Characteristics
3. Application and Derivation
4. Quality Assessment
5. Data Acquisition, Materials, and Methods
6. Data Access
7. References

1. Dataset Overview

This dataset provides gridded estimates of gross primary productivity (GPP), ecosystem respiration (Reco), and net ecosystem CO₂ exchange (NEE) across the circumpolar terrestrial Arctic-boreal region at a 1-km spatial resolution. Monthly CO₂ flux data from 2001 to 2020 were generated using terrestrial eddy covariance and chamber CO₂ flux observations, combined with geospatial meteorological, remote sensing, topographical and soil data, all within a random forest modeling framework. Aggregated average annual NEE, average annual NEE with direct fire emissions added based on the GFED product, and temporal trends in annual NEE rasters over 2002-2020 are also included.

Project: [Arctic-Boreal Vulnerability Experiment](#)

The Arctic-Boreal Vulnerability Experiment (ABOVE) is a NASA Terrestrial Ecology Program field campaign being conducted in Alaska and western Canada for 8 to 10 years, starting in 2015. Research for ABOVE links field-based, process-level studies with geospatial data products derived from airborne and satellite sensors, providing a foundation for improving the analysis, and modeling capabilities needed to understand and predict ecosystem responses to, and societal implications of, climate change in the Arctic and Boreal regions.

Related Publication:

Virkkala, A.-M., B.M. Rogers, J.D. Watts, K.A. Arndt, S. Potter, I. Wargowsky, E.A.G. Schuur, C. See, M. Mauritz, J. Boike, S.M. Bret-Harte, E.J. Burke, A. Burrell, N. Chae, A. Chatterjee, F. Chevallier, T.R. Christensen, R. Commane, H. Dolman, B. Elberling, C.A. Emmerton, E.S. Euskirchen, L. Feng, M. Goeckede, A. Grelle, M. Helbig, D. Holl, J. Järveoja, H. Kobayashi, L. Kutzbach, J. Liu, In. Liujkx, E. López-Blanco, K. Lunneberg, I. Mammarella, M.E. Marushchak, M. Mastepanov, Y. Matsuura, T. Maximov, L. Merbold, G. Meyer, M.B. Nilsson, Y. Niwa, W. Oechel, S.-J. Park, F.-J.W. Parmentier, M. Peichl, W. Peters, R. Petrov, W. Quinton, C. Rödenbeck, T. Sachs, C. Schulze, O. Sonnentag, V. St. Louis, E.-S. Tuittila, M. Ueyama, A. Varlagin, D. Zona, and S.M. Natali. 2024. An increasing Arctic-boreal CO₂ sink offset by wildfires and source regions. *Nature Climate Change* (in press). Preprint at <https://doi.org/10.1101/2024.02.09.579581>

Related Dataset:

Virkkala, A.-M., S. Natali, B.M. Rogers, J.D. Watts, K. Savage, S.J. Connon, M.E. Mauritz-tozer, E.A.G. Schuur, D.L. Peter, C. Minions, J. Nojeim, R. Commane, C.A. Emmerton, M. Goeckede, M. Helbig, D. Holl, H. Iwata, H. Kobayashi, P. Kolari, E. Lopez-blanco, M.E. Marushchak, M. Mastepanov, L. Merbold, M. Peichl, O. Sonnentag, T. Sachs, M. Ueyama, C. Voigt, M. Aurela, J. Boike, G. Celis, N. Chae, T. Christensen, S. Bret-Harte, S. Dengel, H. Dolman, C. Edgar, B. Elberling, S.E. Euskirchen, A. Grelle, J. Hatakka, E.R. Humphreys, J. Jaerveoja, A. Kotani, L. Kutzbach, T. Laurila, A. Lohila, I. Mammarella, Y. Matsuura, G. Meyer, M.B. Nilsson, S.F. Oberbauer, S.J. Park, F.J.W. Parmentier, R. Petrov, A.S. Prokushkin, S. Zyrianov, C. Schulze, V.L. St.louis, E.S. Tuittila, J.P. Tuovinen, W. Quinton, A. Varlagin, D. Zona, and V.I. Zyryanov. 2021. The ABCflux Database: Arctic-Boreal CO₂ Flux and Site Environmental Data, 1989-2020. ORNL DAAC, Oak Ridge, Tennessee, USA. <https://doi.org/10.3334/ORNLDAAC/1934>

- This source provided input data for training the machine learning model.

Acknowledgments:

This work was supported by the NASA Arctic-Boreal Vulnerability Experiment and Carbon Cycle Science programs (grant NNX17AE13G), National Science Foundation (NSF) PLR Arctic System Science Research Networking Activities (grant 1931333) and Division of Environmental Biology (grants LTREB 1354370, 2011257, DEB-0425328, DEB-0724514, and DEB-0830997), US Geological Survey Climate R&D Program, the NSF Arctic Observatory Network (grants 1936752, 1503912, 1107892), Gordon and Betty Moore Foundation (grant 8414), and funding catalyzed by the TED Audacious Project (Permafrost Pathways). Additional sources of support are listed in Virkkala et al. (2024).

2. Data Characteristics

Spatial Coverage: Arctic and boreal regions (tundra and boreal biomes), >49 degrees north latitude.

Spatial Resolution: 1 km

Temporal Coverage: 2001-01-01 to 2020-12-31

Temporal Resolution: Monthly

Site Boundaries: Latitude and longitude are given in decimal degrees.

Site	Westernmost Longitude	Easternmost Longitude	Northernmost Latitude	Southernmost Latitude
Arctic and boreal regions	-180	180	90	33.69

User note: The areas of interest have latitudes ≥ 49 degrees north; however, the data layers extend farther south due to the map projection used.

Data File Information

This dataset holds 60 NetCDFs (*.nc) and three files in cloud optimized GeoTIFF (*.tif) format.

The NetCDFs hold monthly mean estimates of CO₂ flux in units of (g C m⁻² month⁻¹ x 10,000). For net ecosystem exchange (NEE), positive values indicate net CO₂ sinks while negative values indicate CO₂ sources. For both gross primary production (GPP) and ecosystem respiration (Reco), higher values indicate larger fluxes.

The values are in a projected coordinate system (ESRI: 102017; see GeoTIFF information below) at 1-km spatial resolution. The monthly time steps are measured in days since 2001-01-01 recorded in the *time* variable. Months are denoted by the middle day of each month.

The file naming convention for NetCDFs is *CO2Fluxes_Arctic_Boreal_<variable>_<year>.nc*, where

- *<variable>*: "GPP" for gross primary productivity, "NEE" for net ecosystem exchange, or "Reco" for ecosystem respiration.
- *<year>*: the year for the estimates from 2001 to 2020.

There are three GeoTIFFs:

CO2Fluxes_Arctic_Boreal_NEE_2002_2020_avg.tif: mean annual terrestrial NEE over 2002-2020. Negative values = net ecosystem CO₂ uptake (i.e.

net ecosystem CO₂ sink); Positive values = net ecosystem CO₂ emissions (i.e. net ecosystem CO₂ source).

CO2Fluxes_Arctic_Boreal_NEEfire_2002_2020_avg.tif: mean annual terrestrial NEE plus direct fire emissions over 2002-2020.

CO2Fluxes_Arctic_Boreal_NEE_2002_2020_annual_trend_senslope.tif: temporal trend (change) in annual terrestrial NEE over 2002-2020. Negative values = increasing net ecosystem CO₂ uptake; Positive values = increasing net ecosystem CO₂ emissions. Note that a pixel might be on average a net ecosystem CO₂ sink (i.e. negative values in *NEE_gC_m2_2002_2020_avg.tif*), but its temporal trend might be positive. For example, net ecosystem CO₂ emissions were increasing, and the pixel shifted from a sink to a source.

GeoTIFF characteristics:

- Pixel values: g C m⁻² yr⁻¹
- Coordinate system: WGS 1984 Lambert Azimuthal EqArea North Pole, ESRI: 102017.
proj4 string: "+proj=laea +lat_0=90 +lon_0=0 +x_0=0 +y_0=0 +datum=WGS84 +units=m"
- Spatial resolution: nominal 1000 m
- Dimensions: 8933 columns x 7401 rows
- Nodata value: -9999

Note that there is no data layer for the temporal trend for the annual terrestrial NEE plus direct fire emissions.

3. Application and Derivation

The dataset provides gridded monthly NEE, GPP, and Reco over the Arctic-boreal region from 2001-2020. The gridded fluxes were used to obtain regional NEE budgets, with and without direct fire emissions included (Figure 1). The data layers and budgets can be used to detect carbon flux hotspots, monitor and understand changes in Arctic-boreal carbon cycling, and for comparison against estimates from inversion and process-based models.

4. Quality Assessment

The predictive performance of the final models was assessed using the (1) R², (2) root mean square error (RMSE), (3) mean absolute error (MAE), and (4) mean bias error (MBE) between predicted and observed values using leave-one-site-out cross validation. The predictive performance of the models was good or high, with R² ranging from 0.57 to 0.73 and RMSE from 19.4 to 37.3 g C m⁻² month⁻¹. However, the performance metrics also indicate that strong sinks and sources, and high GPP and Reco were underestimated. Other known issues include the limited number of flux observations (i.e. model training data) in Siberia and Canadian Arctic, and in disturbed areas impacted by fire, thermokarst and harvesting, and the lack of accurate geospatial predictors describing soil moisture.). The uncertainty of predictions was evaluated by creating 20 bootstrapped model training datasets (with replacement; same sample size as in the original model training data) and using those to develop 20 individual models and predictions. Out of the 20 predictions, the standard deviation was calculated to represent prediction uncertainty. Similar to the predictive performance metrics, the uncertainty analysis also points towards highest uncertainties in areas with strong sinks, such as in northern Europe and southwestern Russia. However, when the uncertainty estimates were presented relative to the average flux, uncertainties were highest in tundra regions and parts of northern boreal Canada which generally have low in-situ flux data coverage. These 20 bootstrapped model runs were also used to calculate uncertainty around the regional budgets. Also, the flux data layers were compared with the Global Carbon Project atmospheric inversion ensemble and CMIP6 process model ensemble. Moderate correlation was observed between the upscaling and the inversion-based average annual NEE and temporal trends in annual NEE, providing confidence in our results.

5. Data Acquisition, Materials, and Methods

In-situ Data

Carbon dioxide flux measurements were utilized from the ABCflux dataset, which compiles terrestrial CO₂ flux measurements from Arctic-boreal ecosystems (Virkkala et al., 2021b). These fluxes were cumulative values of net ecosystem exchange (NEE), gross primary production (GPP), and ecosystem respiration (Reco), aggregated at monthly scales across 4,897 site-months. The model training dataset included measurements from eddy covariance and chamber sites (n=200), with the majority coming from eddy covariance (88% of months). Each site had from one to 213 months of measurements; the average number of months per site was 25. Fourteen percent of sites had experienced disturbances such as fire, permafrost thaw, or forest harvesting, although this proportion might underestimate overall disturbance in the Arctic-boreal zone. A similar set of sites and upscaling methods were used in Watts et al. (2019) and Virkkala et al. (2021a). This study built upon those efforts by upscaling monthly fluxes for the entire year and extending the analysis through 2020.

Geospatial Data

The study used geospatial data as predictor variables for upscaling fluxes. Predictors used to model CO₂ fluxes included MODIS-based land surface temperature, NDVI, and fractional vegetation cover, land cover, as well as permafrost occurrence, soil and topographic variables, TerraClimate-based solar radiation and vapor pressure deficit, ERA5 Land-based soil moisture, soil temperature and snow cover, and atmospheric CO₂ concentration. Fire data came from the 500-m resolution Global Fire Emissions Database (GFED) product family (van Wees et al., 2022), which used MODIS-based products for burned area detection.

Machine Learning Modeling

Random forest models were applied to upscale GPP, Reco, and NEE across the Arctic-boreal region from 2001 to 2020 (e.g., Figure 2). Model tuning was performed using leave-one-site-out cross-validation, minimizing root mean square error (RMSE). Models were interpreted using partial dependence plots, and variable importance was determined through permutation-based methods. With the tuned models, fluxes were upscaled across the Arctic-boreal zone for land areas north of 49°N, covering 20.69 million km². Temporal trends in the upscaled flux data were analyzed using the Mann-Kendall test, with trends summarized for key regions, biomes (tundra and boreal), and permafrost areas (Figure 3). Analysis codes are published in Virkkala et al. (2024a)

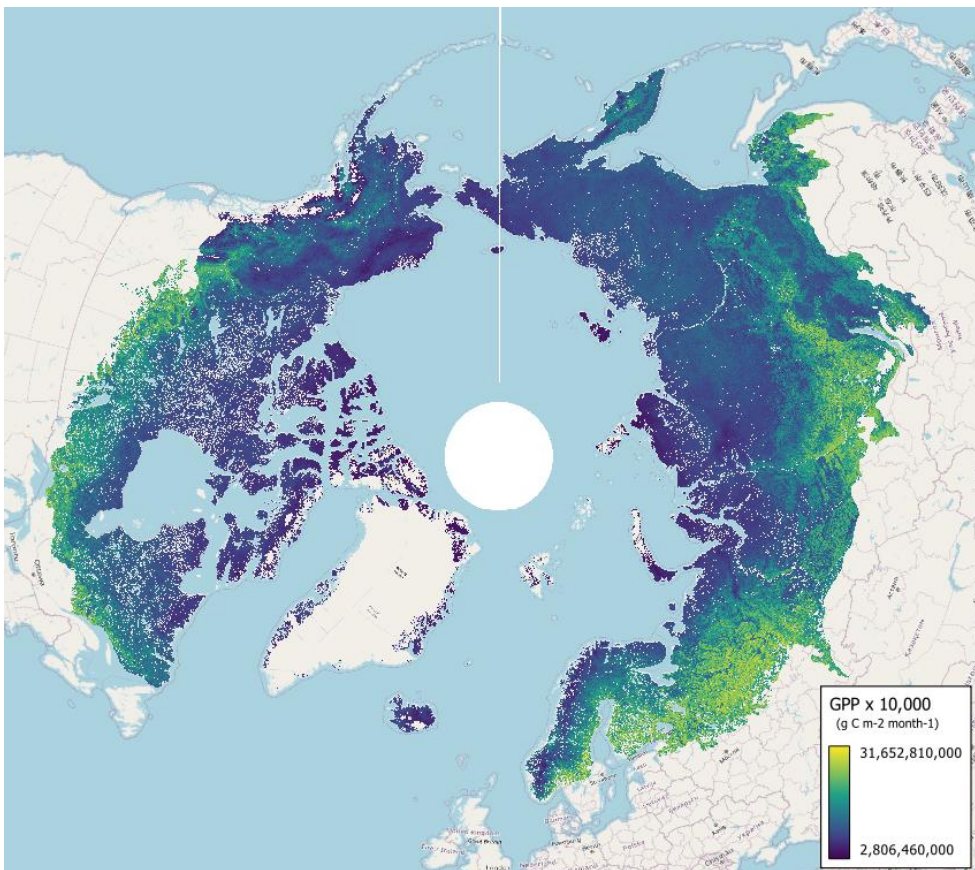


Figure 2. Upscaled estimates of gross primary productivity (GPP) for July 2001 over terrestrial portions of the Arctic-boreal region. Units are g C per m² per month; values include a scale factor of 10,000. Source: GPP_2001.nc.

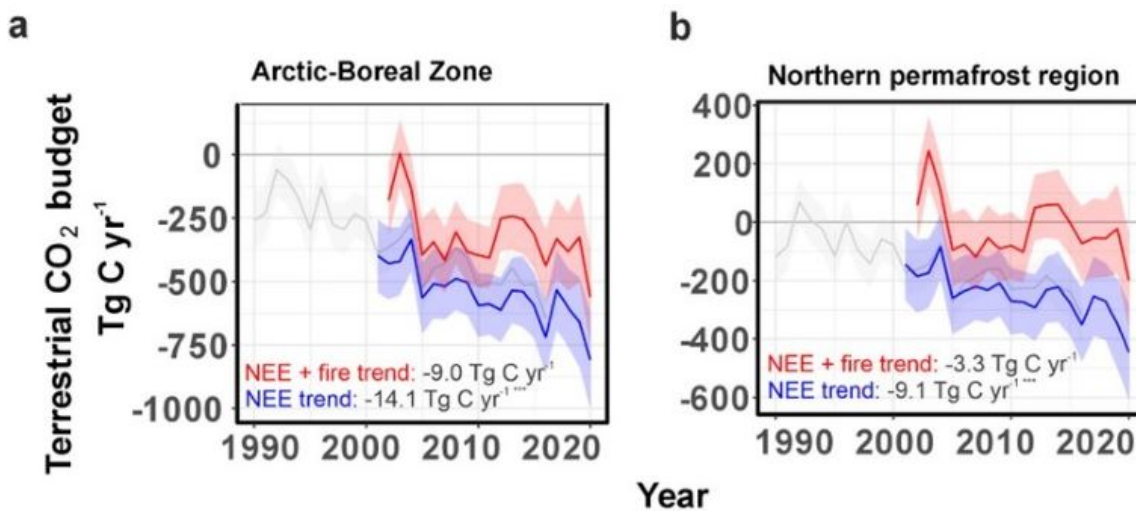


Figure 3. Time series of regional CO₂ budgets across the Arctic-boreal and northern permafrost region.

Additional details are available in Virkkala et al. (2024b).

6. Data Access

These data are available through the Oak Ridge National Laboratory (ORNL) Distributed Active Archive Center (DAAC).

[Machine learning-based Arctic-boreal terrestrial ecosystem CO₂ fluxes, 2001-2020](#)

Contact for Data Center Access Information:

- E-mail: uso@daac.ornl.gov
- Telephone: +1 (865) 241-3952

7. References

Virkkala, A.-M., J. Aalto, B.M. Rogers, Torbern Tagesson, C.C. Treat, S.M. Natali, J.D. Watts, S. Potter, A. Lehtonen, M. Mauritz, E.A. G. Schuur, J. Kochendorfer, D. Zona, W. Oechel, H. Kobayashi, Elyn Humphreys, M. Goeckede, H. Iwata, P. Lafleur, E.S. Euskirchen, S. Bokhorst, M. Marushchak, P.J. Martikainen, B. Elberling, C. Voigt, C. Biasi, O. Sonnentag, Frans-Jan Parmentier, M. Ueyama, G. Celis, V.L. St. Louis, C.A. Emmerton, M. Peichl, Jinshu Chi, Järvi Järveoja, M.B. Nilsson, S.F. Oberbauer, M.S. Torn, Sang-Jong Park, H. Dolman, I. Mammarella, Namyi Chae, R. Poyatos, Efrén López-Blanco, T.R. Christensen, M.J. Kwon, and M. Luoto. 2021a. Predictions for "Statistical upscaling of ecosystem

CO₂ fluxes across the terrestrial tundra and boreal domain: regional patterns and uncertainties". Zenodo. <https://doi.org/10.5281/zenodo.4521852>

Virkkala, A.-M., S. Natali, B.M. Rogers, J.D. Watts, K. Savage, S.J. Connon, M.E. Mauritz-tozer, E.A.G. Schuur, D.L. Peter, C. Minions, J. Nojeim, R. Commane, C.A. Emmerton, M. Goeckede, M. Helbig, D. Holl, H. Iwata, H. Kobayashi, P. Kolari, E. Lopez-blanco, M.E. Marushchak, M. Mastepanov, L. Merbold, M. Peichl, O. Sonnentag, T. Sachs, M. Ueyama, C. Voigt, M. Aurela, J. Boike, G. Celis, N. Chae, T. Christensen, S. Bret-Harte, S. Dengel, H. Dolman, C. Edgar, B. Elberling, S.E. Euskirchen, A. Grelle, J. Hatakka, E.R. Humphreys, J. Jaerveoja, A. Kotani, L. Kutzbach, T. Laurila, A. Lohila, I. Mammarella, Y. Matsuura, G. Meyer, M.B. Nilsson, S.F. Oberbauer, S.J. Park, F.J.W. Parmentier, R. Petrov, A.S. Prokushkin, S. Zyrianov, C. Schulze, V.L. St. Louis, E.S. Tuittila, J.P. Tuovinen, W. Quinton, A. Varlagin, D. Zona, and V.I. Zyryanov. 2021b. The ABCflux Database: Arctic-Boreal CO₂ Flux and Site Environmental Data, 1989-2020. ORNL DAAC, Oak Ridge, Tennessee, USA. <https://doi.org/10.3334/ORNLDAAC/1934>

Virkkala, A.-M., B.M. Rogers, J.D. Watts, K.A. Arndt, S. Potter, I. Wargowsky, and S.M. Natali. 2024a. Code for "An increasing Arctic-boreal CO₂ sink offset by wildfires and source regions". Zenodo. <https://doi.org/10.5281/zenodo.13691585>

Virkkala, A.-M., B.M. Rogers, J.D. Watts, K.A. Arndt, S. Potter, I. Wargowsky, E.A.G. Schuur, C. See, M. Mauritz, J. Boike, S.M. Bret-Harte, E.J. Burke, A. Burrell, N. Chae, A. Chatterjee, F. Chevallier, T.R. Christensen, R. Commane, H. Dolman, B. Elberling, C.A. Emmerton, E.S. Euskirchen, L. Feng, M. Goeckede, A. Grelle, M. Helbig, D. Holl, J. Järveoja, H. Kobayashi, L. Kutzbach, J. Liu, In. Liujkx, E. López-Blanco, K. Lunneberg, I. Mammarella, M.E. Marushchak, M. Mastepanov, Y. Matsuura, T. Maximov, L. Merbold, G. Meyer, M.B. Nilsson, Y. Niwa, W. Oechel, S.-J. Park, F.-J.W. Parmentier, M. Peichl, W. Peters, R. Petrov, W. Quinton, C. Rödenbeck, T. Sachs, C. Schulze, O. Sonnentag, V. St. Louis, E.-S. Tuittila, M. Ueyama, A. Varlagin, D. Zona, and S.M. Natali. 2024b. An increasing Arctic-boreal CO₂ sink offset by wildfires and source regions. Nature Climate Change (in press). Preprint at <https://doi.org/10.1101/2024.02.09.579581>

van Wees, D., G.R. van der Werf, J.T. Randerson, B.M. Rogers, Y. Chen, S. Veraverbeke, L. Giglio, and D.C. Morton. 2022. Global biomass burning fuel consumption and emissions at 500m spatial resolution based on the Global Fire Emissions Database (GFED). Geoscientific Model Development 15:8411–8437. <https://doi.org/10.5194/gmd-15-8411-2022>

Watts, J.D., S. Natali, S. Potter, and B.M. Rogers. 2019. Gridded Winter Soil CO₂ Flux Estimates for pan-Arctic and Boreal Regions, 2003-2100. ORNL DAAC, Oak Ridge, Tennessee, USA. <https://doi.org/10.3334/ORNLDAAC/1683>



[Privacy Policy](#) | [Help](#)



Home

About Us

- Mission
- Data Use and Citation Policy
- User Working Group
- Partners

Get Data

- Science Themes
- NASA Projects
- All Datasets

Submit Data

- Submit Data Form
- Data Scope and Acceptance
- Data Authorship Policy
- Data Publication
- Timeline
- Detailed Submission
- Guidelines

Tools

- TESVIS
- THREDDS
- SDAT
- Daymet
- Airborne Data Visualizer
- Soil Moisture Visualizer

Resources

- Learning
- Data Management
- News

Help

- Earthdata Forum [↗](#)
- Email Us [✉](#)