

nsink: An R package for flow path nitrogen removal estimation

Jeffrey W. Hollister¹, Dorothy Q. Kellogg², Qian Lei-Parent³, Emily Wilson³, Cary Chadwick³, David Dickson³, Arthur Gold², and Chester Arnold³

1 U. S. Environmental Protection Agency, Atlantic Coastal Environmental Sciences Division, Narragansett, RI 02882 2 University of Rhode Island, Department of Natural Resources Science, Kingston, RI 02881 3 University of Connecticut, Center for Land Use Education and Research, Storrs, CT 06268

DOI: 10.21105/joss.04039

Software

- Review ¹
- Repository 🗗
- Archive C^{*}

Editor: Chris Vernon ♂ Reviewers:

- @jmp75
- @ldecicco-usgs
- @jmp75

Submitted: 20 December 2021 Published: 11 March 2022

License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License (CC BY 4.0).

Summary

The nsink package estimates cumulative nitrogen (N) removal along a specified flow path and is based on methodologies outlined in Kellogg et al. (2010). For a user-specified watershed (i.e., hydrologic unit code (HUC)), nsink downloads all required datasets from public datasets in the United States, prepares data for use, summarizes N removal along a flow path and creates several static maps. The results of an nsink analysis may be exported to standard geospatial files for use in other applications.

Statement of need

Excess N delivery via surface water to downstream aquatic resources contributes to impaired water quality and impacts ecosystem services including harmful algal blooms (HABs) and hypoxia (Rabalais et al., 2002). Identifying landscape N sinks (i.e., areas where N is effectively removed from the aquatic system) and analyzing N delivery at the watershed scale is helpful to watershed managers, land use planners and conservation organizations. The theoretical underpinnings for identifying N sinks rely on decades of research and are explained in Kellogg et al. (2010).

Prior N-sink implementations were done case-by-case. Data acquisition and manipulation were mostly manual and took weeks to months to complete for a single 12-digit HUC. The effort required for the analysis limited it's application as scaling beyond a few pilot studies was not feasible. The goal of nsink was to address this limitation and provide an open source solution that could be run on a single small watershed (e.g., 12-digit HUC) in minutes to hours with minimal manual input.

The nsink package

Package Installation

The nsink package is available from https://github.com/usepa/nsink and may be installed in R with the following:

```
# If not installed, install remotes
install.packages("remotes")
```



Install nsink from GitHub
remotes::install_github("USEPA/nsink", dependencies = TRUE, build_vignettes = TRUE)

Package Details

The nsink package is designed around the major steps in running an N-Sink analysis and includes functions for the following tasks:

- 1. Prepare for analysis
 - Get data
 - Prepare data for analysis
 - Calculate relative N removal layer for hydric soils, lakes and streams.
- 2. Run a point-based analysis
 - Calculate a flow path
 - Summarize relative N removal along a flow path
- 3. Run a HUC-based analysis
 - Develop static maps
 - Generate output datasets

Required Data

The ability to run an nsink analysis relies on several datasets for the conterminous United States. By limiting our approach to these national datasets we are ensuring scalability of nsink because the datasets will be available for most locations in the United States. The datasets that nsink uses are the National Hydrography Dataset Plus version 2 (NHDPlus), Soil Survey Geographic Database (SSURGO), the National Land Cover Dataset (NLCD) land cover, and the National Land Cover Dataset (NLCD) impervious surface (Jin et al., 2019; Moore et al., 2019; Soil Survey Staff, 2017). These datasets are all available via an Application Programming Interface (API) or via direct download.

Dependencies

The nsink package depends on several existing R packages to facilitate spatial data handling, data acquisition, data management, data analysis and data processing. These are detailed in Table 1.

Table 1. R package dependencies for the nsink package

Package	Task	Citation
sf	Spatial Data Handling and Analysis	Pebesma (2018); Pebesma (2021b)
raster	Spatial Data Handling and Analysis	Hijmans (2021)
stars	Spatial Data Handling and Analysis	Pebesma (2021c)
fasterize	Spatial Data Handling and Analysis	Ross (2020)
lwgeom	Spatial Data Handling and Analysis	Pebesma (2021a)



Package	Task	Citation
gstat	Spatial Data	Pebesma (2004); Gräler et al.
	Handling and	(2016); Pebesma & Graeler (2021)
	Analysis	
sp	Spatial Data	Pebesma & Bivand (2005); Bivand
	Handling and	et al. (2013); Pebesma & Bivand
	Analysis	(2021)
units	Unit	Pebesma et al. (2016); Pebesma et
	Transformations	al. (2021)
FedData	Data Acquisition	Bocinsky (2020)
httr	Data Acquisition	Wickham (2020)
dplyr	Data Management	Wickham et al. (2021)
	and Analysis	
Z00	Data Management	Zeileis & Grothendieck (2005);
	and Analysis	Zeileis et al. (2021)
igraph	Data Management	Csardi & Nepusz (2006); Csardi et al.
	and Analysis	(2020)
readr	Data Management	Wickham & Hester (2020)
	and Analysis	
foreign	Data Management	R Core Team (2020)
	and Analysis	
rlang	Data Management	Henry & Wickham (2021)
	and Analysis	
furrr	Parallel Processing	Vaughan & Dancho (2021)
future	Parallel Processing	Bengtsson (2021); Bengtsson (2020)

Functionality

Currently, nsink provides 10 exported functions to facilitate a flow path analysis of relative N removal. The nsink repository (https://github.com/usepa/nsink) and R package documentation contain detailed documentation of each function. The pacakge also has a vignette that outlines a typical workflow for running an N-Sink analysis with the nsink package. Upon install, the vignette is accessed in R with vignette("intro", package = "nsink").

Acknowledgements

Many people have contributed in various ways to the development of the N-Sink concept. In particular, we would like to thank, Chet Arnold, Cary Chadwick, David Dickson, and Emily Wilson of the University of Connecticut's Center for Land Use Education and Research as well as Peter August, Chris Damon, and Art Gold of the University of Rhode Island's Department of Natural Resources Science. Both the UCONN and URI crews have contributed tremendously to the development of the N-Sink concept. Additionally, we are grateful to Stephen Shivers, Michael Dumelle, Justin Bousquin, Joe LiVolsi, Tim Gleason, and Wayne Munns for constructive early reviews of this paper. Lastly, Ken Forshay from the US EPA's Center for Environmental Solutions and Emergency Response deserves our thanks for shepherding the development of N-Sink for many years. The views expressed in this article are those of the authors and do not necessarily represent the views or policies of the U.S. Environmental Protection Agency. Any mention of trade names, products, or services does not imply an endorsement by the U.S. Government or the U.S. Environmental Protection Agency. The EPA does not endorse any commercial products, services, or enterprises. This contribution is identified by the tracking number ORD-044618 of the Atlantic Coastal Environmental Sciences Division, Office of Research and Development, Center for Environmental Measurement and Modeling, US Environmental Protection Agency.



References

- Bengtsson, H. (2020). Future: Unified parallel and distributed processing in r for everyone. https://CRAN.R-project.org/package=future
- Bengtsson, H. (2021). A Unifying Framework for Parallel and Distributed Processing in R using Futures. In *The R Journal*. https://doi.org/10.32614/RJ-2021-048
- Bivand, R. S., Pebesma, E. J., & Gomez-Rubio, V. (2013). Applied spatial data analysis with *R*, second edition. Springer, NY. https://doi.org/10.1007/978-1-4614-7618-4
- Bocinsky, R. K. (2020). FedData: Functions to automate downloading geospatial data available from several federated data sources. https://CRAN.R-project.org/package=FedData
- Csardi, G., & Nepusz, T. (2006). The igraph software package for complex network research. InterJournal, Complex Systems, 1695. https://igraph.org
- Csardi, G., Nepusz, T., Horvat, S., Traag, V., & Zanini, F. (2020). *Network analysis and visualization*. https://CRAN.R-project.org/package=zoo
- Gräler, B., Pebesma, E. J., & Heuvelink, G. (2016). Spatio-temporal interpolation using gstat. *The R Journal*, *8*, 204–218. https://doi.org/10.32614/RJ-2016-014
- Henry, L., & Wickham, H. (2021). Rlang: Functions for base types and core r and 'tidyverse' features. https://CRAN.R-project.org/package=rlang
- Hijmans, R. J. (2021). *Raster: Geographic data analysis and modeling*. https://CRAN. R-project.org/package=raster
- Jin, S., Homer, C., Yang, L., Danielson, P., Dewitz, J., Li, C., Zhu, Z., Xian, G., & Howard, D. (2019). Overall methodology design for the united states national land cover database 2016 products. *Remote Sensing*, 11(24), 2971. https://doi.org/10.3390/rs11242971
- Kellogg, D. Q., Gold, A. J., Cox, S., Addy, K., & August, P. V. (2010). A geospatial approach for assessing denitrification sinks within lower-order catchments. *Ecological Engineering*, 36(11), 1596–1606. https://doi.org/10.1016/j.ecoleng.2010.02.006
- Moore, R. B., McKay, L. D., Rea, A. H., Bondelid, T. R., Price, C. V., Dewald, T. G., Johnston, C. M.others. (2019). User's guide for the national hydrography dataset plus (NHDPlus) high resolution. *Open-File Report-US Geological Survey*, 2019-1096. https: //doi.org/10.3133/ofr20191096
- Pebesma, E. J. (2004). Multivariable geostatistics in S: The gstat package. Computers & Geosciences, 30, 683–691. https://doi.org/10.1016/j.cageo.2004.03.012
- Pebesma, E. J. (2018). Simple Features for R: Standardized Support for Spatial Vector Data. *The R Journal*, *10*(1), 439–446. https://doi.org/10.32614/RJ-2018-009
- Pebesma, E. J. (2021a). Lwgeom: Bindings to selected 'liblwgeom' functions for simple features. https://CRAN.R-project.org/package=lwgeom
- Pebesma, E. J. (2021b). Simple features for r. https://CRAN.R-project.org/package=sf
- Pebesma, E. J. (2021c). *Stars: Spatiotemporal arrays, raster and vector data cubes.* https://CRAN.R-project.org/package=stars
- Pebesma, E. J., & Bivand, R. (2021). *Sp: Classes and methods for spatial data*. https://CRAN.R-project.org/package=sp
- Pebesma, E. J., & Bivand, R. S. (2005). Classes and methods for spatial data in R. R News, 5(2), 9–13. https://CRAN.R-project.org/doc/Rnews/
- Pebesma, E. J., & Graeler, B. (2021). *Gstat: Spatial and spatio-temporal geostatistical modelling, prediction and simulation*. https://CRAN.R-project.org/package=gstat



- Pebesma, E. J., Mailund, T., & Hiebert, J. (2016). Measurement units in R. *R Journal*, 8(2), 486–494. https://doi.org/10.32614/RJ-2016-061
- Pebesma, E. J., Mailund, T., Kalinowski, T., & Ucar, I. (2021). *Units: Spatiotemporal arrays, raster and vector data cubes*. https://CRAN.R-project.org/package=units
- R Core Team. (2020). Foreign: Read data stored by 'minitab', 's', 'SAS', 'SPSS', 'stata', 'systat', 'weka', 'dBase', ... https://CRAN.R-project.org/package=foreign
- Rabalais, N. N., Turner, R. E., & Scavia, D. (2002). Beyond science into policy: Gulf of mexico hypoxia and the mississippi river: Nutrient policy development for the mississippi river watershed reflects the accumulated scientific evidence that the increase in nitrogen loading is the primary factor in the worsening of hypoxia in the northern gulf of mexico. *BioScience*, 52(2), 129–142. https://doi.org/10.1641/0006-3568(2002)052%5B0129:BSIPGO%5D2.0. CO;2
- Ross, N. (2020). Fasterize: Fast polygon to raster conversion. https://CRAN.R-project.org/ package=fasterize
- Soil Survey Staff, U. (2017). Web soil survey.
- Vaughan, D., & Dancho, M. (2021). Furr: Apply mapping functions in parallel using futures. https://CRAN.R-project.org/package=furrr
- Wickham, H. (2020). *Httr: Tools for working with URLs and HTTP*. https://CRAN.R-project. org/package=httr
- Wickham, H., François, R., Henry, L., & Müller, K. (2021). Dplyr: A grammar of data manipulation. https://CRAN.R-project.org/package=dplyr
- Wickham, H., & Hester, J. (2020). *Readr: Read rectangular text data*. https://CRAN. R-project.org/package=readr
- Zeileis, A., Gorthendieck, G., & Ryan, J. A. (2021). Zoo: S3 infrastructure for regular and irregular time series (Z's Ordered Observations). https://CRAN.R-project.org/package= zoo
- Zeileis, A., & Grothendieck, G. (2005). Zoo: S3 infrastructure for regular and irregular time series. *Journal of Statistical Software*, 14(6), 1–27. https://doi.org/10.18637/jss.v014.i06