nsink: An R package for flow path based nitrogen removal estimation

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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). `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 watershed 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 leads to several ecosystem impacts including harmful algal blooms (HABs) and hypoxia (Rabalais, Turner, and Scavia 2002). Identifying landscape N sinks (i.e., areas where N is effectively removed from the aquatic system) and analysing 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).

The first implementation of this approach was done case-by-case. Data acquisition and manipulation were mostly manual and took weeks to months to complete for a single watershed. 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 in minutes to hours with little 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:

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

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

Package Details

The `nsink` package is designed around the major steps in running a N-Sink analysis and includes the following:

1. Prepare for analysis
   - Get data

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- Calculate relative N removal layer for hydric soils, lakes and streams.
2. Run an interactive analysis
   - Calculate a flow path
   - Summarize relative N removal along a flow path
3. Run a watershed analysis
   - Develop static maps
   - Generate output datasets

Required Data

The ability to run an `nsink` analysis relies on several national scale datasets for the 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 (NHDPlus), Soil Survey Geographic Database (SSURGO), the National Land Cover Dataset (NLCD) land cover, and the National Land Cover Dataset (NLCD) impervious surface [ADD CITATION FOR DATASETS]. These datasets are all available via either 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.

Package	Task	Citation
sf	Spatial Data Handling and Analysis	E. Pebesma (2018); E. Pebesma (2021b)
raster	Spatial Data Handling and Analysis	Hijmans (2021)
stars	Spatial Data Handling and Analysis	E. Pebesma (2021c)
fasterize	Spatial Data Handling and Analysis	Ross (2020)
lwgeom	Spatial Data Handling and Analysis	E. Pebesma (2021a)
gstat	Spatial Data Handling and Analysis	E. J. Pebesma (2004); Gräler, Pebesma, and Heuvelink (2016); E. Pebesma and Graeler (2021)
sp	Spatial Data Handling and Analysis	E. J. Pebesma and Bivand (2005); Bivand, Pebesma, and Gomez-Rubio (2013); E. Pebesma and Bivand (2021)
units	Unit Transformations	E. Pebesma, Mailund, and Hiebert (2016); E. Pebesma et al. (2021)
FedData	Data Acquisition	Bocinsky (2020)
httr	Data Acquisition	Wickham (2020)
dplyr	Data Management and Analysis	Wickham et al. (2021)
zoo	Data Management and Analysis	Zeileis and Grothendieck (2005); Zeileis, Gorthendieck, and Ryan (2021)

Package	Task	Citation
`igraph`	Data Management and Analysis	Csardi and Nepusz (2006); Csardi et al. (2020)
`readr`	Data Management and Analysis	Wickham and Hester (2020)
`foreign`	Data Management and Analysis	R Core Team (2020)
`rlang`	Data Management and Analysis	Henry and Wickham (2021)
`furrr`	Parallel Processing	Vaughan and Dancho (2021)
`future`	Parallel Processing	Bengtsson (2021); Bengtsson (2020)

Table 1. R Package Dependencies for the `nsink` package

Functionality

Currently, `nsink` provides 10 exported functions to facilitate a flow path analysis of relative N removal.

- `nsink_get_huc_id()`: A function for searching the name of a USGS Watershed Boundary Dataset Hydrologic Unit (https://www.usgs.gov/core-science-systems/ngp/national-hydrography/watershed-boundary-dataset) and retrieving it's 12-digit Hydrologic Unit Code (HUC).

- `nsink_get_data()`: Using any acceptable HUC ID (e.g. 2-digit to 12-digit), this function downloads the NHDPlus, SSURGO, NLCD Land Cover, and the NLCD Impervious for that HUC.

- `nsink_prep_data()`: `nsink` needs data in a common coordinate reference system, from mutliple NHD-Plus tables, and from different portions of SSURGO. This function completes these data preparation steps and outputs all data, clipped to the HUC boundary.

- `nsink_calc_removal()`: Quantifying relative N removal across a landscape is a key aspects of an `nsink` analysis. The `nsink_calc_removal()` function takes the object returned from `nsink_prep_data()` and calculates relative N removal for each landscape sink. See Kellogg et al (2010) for details on relative N removal estimation for each sink.

- `nsink_generate_flowpath()`: This function uses a combination of flow determined by topography (e.g. via a flow-direction raster) for the land-based portions of a flow path and of downstream flow along the NHDPlus stream network.

- `nsink_summarize_flowpath()`: Summarizing removal along a specified flow path requires relative N removal and a generated flow path. This function uses these and returns a summary of relative N removal along a flow path for each sink.

- `nsink_generate_static_maps()`: The `nsink_generate_static_maps()` function analyzes N removal at the watershed scale by summarizing the results of multiple flow paths. Four static maps are returned: 1)removal efficiency; 2)loading index; 3)transport index; 4)delivery index. Removal efficiency is a rasterized version of the `nsink_calc_removal()` output. Loading index is N sources based on NLCD categories and ranges from 0 to 1. Transport index is a heat map with the cumulative relative N removal along flow paths originating from a grid of points, density set by the user, across a watershed, highlighting the gradient of downstream N retention. Delivery index is the result of multiplying the loading index and the transport index and shows potential N delivery from different sources, taking into account the relative N removal as water moves downstream.

- `nsink_plot()`: A function that plots each raster in the list returned from `nsink_generate_static_maps()`.

- `nsink_build()`: One of the drivers behind the development of the `nsink` package was to provide `n-sink` analysis output that could be used more broadly (e.g. within a GIS). The `nsink_build()` function is a wrapper for a complete `nsink` analysis and outputs R objects returned by each function

and shapefiles or TIFF versions of those objects.

- `nsink_load()`: Essentially the inverse of the `nsink_build()` function, this function takes a folder of files, likely created with `nsink_build()`, and reads them in to R.

The R package documentation contains both a detailed description of each function and 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

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