canaper: Categorical analysis of neo- and paleo-endemism in R

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https://github.com/joelnitta/canaper

Background

- CANAPE (Categorical Analysis of Neo- and Paleo-endemism) is a recently developed method that provides insight into the evolutionary processes underlying endemism [1]
- CANAPE integrates a phylogenetic tree with a community (species × sites) matrix to infer if endemic areas are so because of recent speciation (neoendemism), or because they harbor old lineages that have mostly gone extinct in other areas (paleoendemism), or if they are a mixture of both
- CANAPE is currently only implemented in Biodiverse [2], a program written in perl that is used via a GUI or custom scripts.
- canaper is a new R package to conduct CANAPE entirely in R

Features

- Sparse matrix encoding of community matrices to increase computing efficiency via <a href="https://physiological.com/p
- Simple implementation of parallel computing to increase speed via future

Installation and loading

canaper is currently available on GitHub

```
devtools::install_github("joelnitta/canaper")

# Load packages
library(canaper)
library(ape) # For handling phylogenies
library(future) # For parallel computing
library(tidyverse) # For data wrangling and visualization
```

Example: Australian Acacia

This demonstrates the package using the dataset of the paper where CANAPE was first published [1]: 506 species of *Acacia* in Australia distributed over 3037 sites:

Randomization test

The first step of CANAPE is to compare the observed values of phylogenetic endemism (PE) and alternative PE (PE measured on a modified tree where all branch lengths are set equal) with those from a set of random communities. The <code>cpr_rand_test()</code> conducts the randomization, using parallel computing to increase speed. The <code>picante</code> package [4] is used to generate the random communities.

```
# Set a parallel back-end, with 4 CPUs running simultaneously
plan(multisession, workers = 4)
# Run randomization test
acacia rand res <- cpr rand test(</pre>
  acacia$comm, acacia$phy,
  n reps = 100, n iterations = 100000)
#> [1] "Dropping tips from the tree because they are not present in
        the community data:"
#> [1] "Pararchidendron pruinosum" "Paraserianthes lophantha"
                                   "clunies-rossiae"
#> [3] "saligna"
# Check some of the results
acacia_rand_res[1:3, c(1,2,8,9)]
                                   pd_obs pd_rand_mean pd_obs_p_upper
        pd_obs_p_lower
#> '-1025000:-1825000'
                                            0.023
                                                                   0.01
#> '-1025000:-1875000'
                                            0.050
                                                                   0.00
#> '-1025000:-1925000'
                                            0.038
                                                                   0.57
        0.43
```

Output summary (in part):

- pe obs = observed PE
- pe rand mean = mean PE of the randomizations
- pe_obs_p_upper = percent of randomizations where observed PE was greater than random values
- pe_obs_p_lower = percent of randomizations where observed PE was lower than random values

Classify significance

The next step of CANAPE is to classify the endemism type of each site. The cpr_classify_endem() function does this automatically given output of cpr_rand_test().

```
# Classify endemism type
acacia_canape <- cpr_classify_endem(acacia_rand_res)
# Count the results
table(acacia_canape$endem_type)
#>
#> mixed neo not significant paleo
super
#> 109 11 2756 40
121
```

Endemism codes:

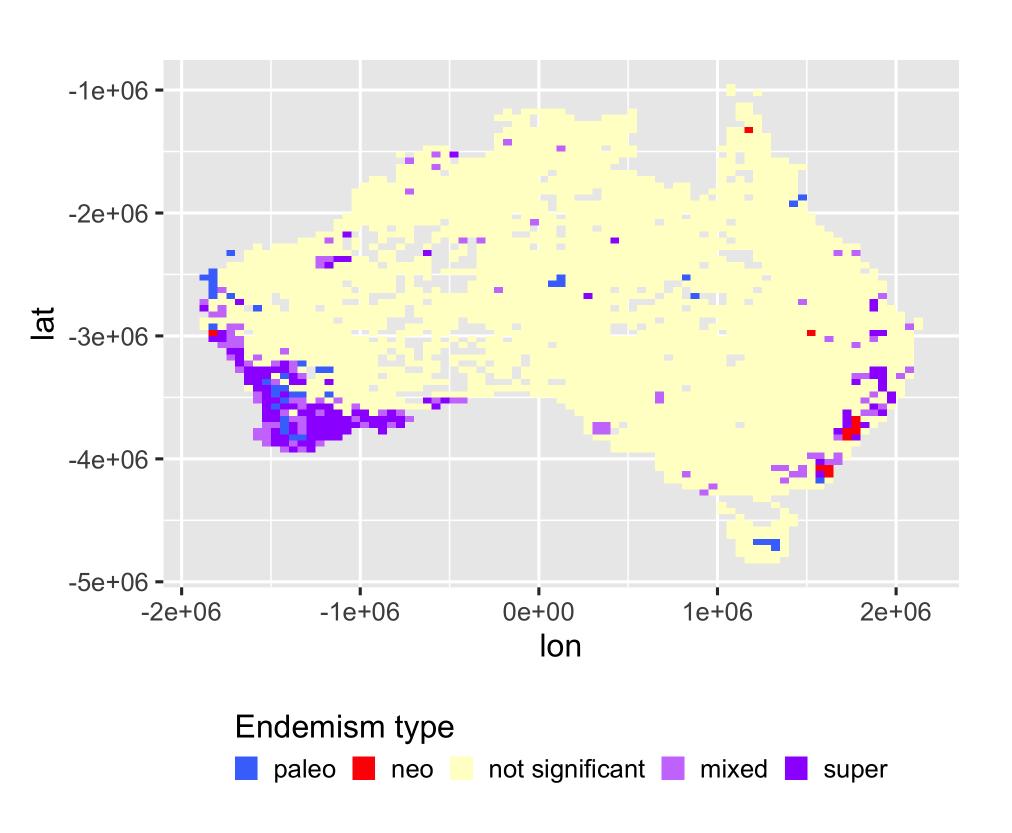
- neo = Neo-endemic
- paleo = Paleo-endemic
- mixed = Mix of neo and paleo
- super = Mixed, and highly significant (p < 0.01)

Visualize results

We can visualize the results of CANAPE using ggplot2. The cpr_endem_cols palette with colors typically used for CANAPE is provided.

```
# Fist do some data wrangling to make
# the results easier to plot (add lat/long columns)
acacia_canape <- as_tibble(acacia_canape, rownames = "site") |>
    separate(site, c("lon", "lat"), sep = ":") |>
    mutate(dplyr::across(c(lon, lat), parse_number))

# Plot the results
ggplot(acacia_canape, aes(x = lon, y = lat, fill = endem_type)) +
    geom_tile() +
    scale_fill_manual(values = cpr_endem_cols, name = "Endemism type") +
    guides(fill = guide_legend(title.position = "top")) +
    theme_gray(base_size = 24) +
    theme(legend.position = "bottom")
```



Next steps

- Implement rand_structured randomization algorithm of Biodiverse (should speed up randomizations ~10×)
- Submit to CRAN

Acknowledgements

Thanks to Shawn Laffan for providing help with the code.

References

- C.E. González-Orozco, M.C. Ebach, S. Laffan, A.H. Thornhill, N.J. Knerr, A.N. Schmidt-Lebuhn, C.C. Cargill, M. Clements, N.S. Nagalingum, B.D. Mishler, J.T. Miller, Quantifying phytogeographical regions of Australia using geospatial turnover in species composition, PLoS ONE, 9 (2014) e92558, https://doi.org/10.1371/journal.pope.0092558
- PLoS ONE. 9 (2014) e92558. https://doi.org/10.1371/journal.pone.0092558.

 S.W. Laffan, E. Lubarsky, D.F. Rosauer, Biodiverse, a tool for the spatial analysis of biological and related diversity, Ecography. 33 (2010) 643–647. https://doi.org/10.1111/j.1600-0587.2010.06237.x.
- B.H. Daru, P. Karunarathne, K. Schliep, phyloregion: R package for biogeographical regionalization and macroecology, Methods Ecol. Evol. 11 (2020) 1483–1491. https://doi.org/gkzsjp.
- [4] S.W. Kembel, P.D. Cowan, M.R. Helmus, W.K. Cornwell, H. Morlon, D.D. Ackerly, S.P. Blomberg, C.O. Webb, Picante: R tools for integrating phylogenies and ecology, Bioinformatics. 26 (2010) 1463–1464. papers://ebd7d3df-d8f1-4be8-bcf7-4fb07ababf30/Paper/p1990.

Source code: https://github.com/joelnitta/botany_poster_2021