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RERconverge: an R package for associating evolutionary rates with convergent traits

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Abstract

Motivation: When different lineages of organisms independently adapt to similar environments, selection often acts repeatedly upon the same genes, leading to signatures of convergent evolutionary rate shifts at these genes. With the increasing availability of genome sequences for organisms displaying a variety of convergent traits, the ability to identify genes with such convergent rate signatures would enable new insights into the molecular basis of these traits.

Results: Here we present the R package RERconverge, which tests for association between relative evolutionary rates of genes and the evolution of traits across a phylogeny. RERconverge can perform associations with binary and continuous traits, and it contains tools for visualization and enrichment analyses of association results.


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Supplementary information: Supplementary information, containing detailed vignettes for usage of RERconverge, are available at Bioinformatics online.

1 Introduction

A major motivation in evolutionary biology is to determine which genetic changes underlie phenotypic adaptations. Convergent evolution, in which the same phenotype arises independently in distinct evolutionary lineages, provides natural replicates of phenotypic adaptation that aid researchers in linking phenotypes to their underlying genetic changes. Selection repeatedly targets the same genes in several known cases of phenotypic convergence, including Prestin in echolocation in bats and marine mammals ([Li et al.](#) 2010), Mc1r in reduced pigmentation in multiple vertebrate species ([Kronforst et al.](#) 2012), and Na+ 1.4 in toxin resistance in snakes ([Feldman et al.](#) 2012). This pattern leads to the prediction that genes responding selectively to a convergent environmental or phenotypic transition may show convergent shifts in evolutionary rates (i.e., number of nucleotide or amino acid substitutions per unit time), due to reduced or increased selective constraints on those genes. For example, both genes and non-coding elements involved in eye function lose selective constraint in subterranean mammals, and the corresponding rate shift signal can be used genome-wide to identify candidate regions with eye-specific function ([Partha et al.](#) 2017). Thus, evolutionary rates provide a rich source of genome-wide molecular data that, in combination with convergent phenotypes, can be used to infer targets of convergent selection.

Several research groups have developed methods to identify patterns of convergent evolutionary rate shifts associated with convergent changes in traits or environment, including Coevol ([Lartillot and Poujol](#) 2011), Forward Genomics ([Hiller et al.](#) 2012; [Prudent et al.](#) 2016), and PhyloAcc ([Hu et al.](#) 2018). However, existing methods to identify these patterns still have
several areas for improvement, including support for both binary and continuous traits (Kowalczyk et al. 2018), compute times that enable genome-wide analyses, robust statistical treatment (Partha et al. 2019), and ease of visualization and downstream analyses. To address these issues, here we present RERconverge, a software package that implements tests of association between evolutionary rates of genetic loci and convergent phenotypic traits, both binary and continuous (pipeline illustrated in Figure 1). RERconverge incorporates and corrects for known phylogenetic relationships among included species, performs rapid analyses of genome-scale data, and implements appropriate statistical techniques to reduce the effect of outliers and correct for multiple testing. This easy-to-implement software enables biologists with extensive organismal knowledge to apply computational techniques to quickly test hypotheses and link molecular data with phenotypes.

# 2 Description and implementation

## 2.1 Basic usage of RERconverge

The RERconverge package runs within an installation of the R software (R Core Team 2017) on any platform (Linux, Windows, Mac OS). The user provides the following data as input:

- A set of phylogenetcs, all with the same topology, with branch lengths for each tree calculated from alignments of that "gene" sequence (these can also represent non-coding sequences such as enhancers) using software such as PAML (Yang 2007).
- Values for a trait of interest for extant species/tips of the tree. RERconverge can infer values for internal branches of the tree, or the user can provide these values via a single tree or by using RERconverge interactive branch selection to select foreground branches.

Detailed instructions for RERconverge are available within the Supplementary information to this article or at https://github.com/nclark-lab/RERconverge/wiki/Vignettes.

## 2.2 Rapid estimation and visualization of Relative Evolutionary Rates (RER)

RERconverge offers efficient computation of gene-specific rates of evolution on branches of phylogenetic trees in genome-scale datasets. These gene-specific rates of evolution, termed relative evolutionary rates (RER), reflect the amount of sequence divergence on a particular branch after correcting for non-specific factors affecting divergence on the branch such as time since speciation and mutation rate. Additionally, using a combination of statistical approaches including data transformation and weighted linear regression, RERconverge provides estimates of relative evolutionary rates that are robust to several factors introducing outliers in the dataset, such as the presence of distantly related species in the phylogeny (Partha et al. 2019). To accelerate the computations underlying RER calculations, key functions are written in C++ and integrated with the R code through the Rcpp package (Eddelbuettel and François 2011).

## 2.3 Flexible specification of binary and continuous trait evolution

For binary traits, RERconverge provides multiple methods for users to specify which branches are in the "foreground", meaning those that display the convergent trait of interest. Users can directly supply foreground branch assignments by providing a phylogenetic tree with appropriate branch lengths or by interactively selecting branches on a phylogeny. Alternatively, users can specify which extant species are foreground and allow the software to infer which internal branches to consider foreground under a variety of scenarios. These include whether to consider only branches at a transition to a convergent phenotype as foreground or to also consider subsequent branches that preserve the phenotype as foreground, whether the foreground trait can be lost or only gained, and whether each foreground branch should be weighted equally.

In addition to supporting analysis of binary traits, RERconverge can perform correlation analysis between evolutionary rates and continuous traits (Kowalczyk et al. 2018). Trait evolution is modeled on a trait tree constructed...
from user-provided phenotype values for extant species provided by the user and ancestral nodes inferred through a phylogenetic model. By default, trait tree branch lengths represent change in the trait between a species and its ancestor, which removes phylogenetic dependence between branches. For completeness, other options allow branch lengths to be calculated as the average or terminal trait value along a branch to represent the state of a trait rather than its change.

2.4 Genome-wide association between RER and traits, with correction for multiple testing

RERConverge rapidly computes the association between gene-specific relative evolutionary rates and the user-specified traits for large sets of genes. The program estimates the correlation between gene tree branch lengths and the values for traits along the tree for associations between genes' relative evolutionary rates and a Pearson linear correlation for continuous traits. Full gene lists are returned with correlation statistic values, the number of data points used for the correlations, and adjusted p-values (FDR) derived using the Benjamini-Hochberg correction ([Benjamini and Hochberg, 1995]).

2.5 Assessing gene set enrichment in results

Further analysis of gene lists from correlation analyses can be performed using built-in RERConverge pathway enrichment functions. These functions perform a rank-based enrichment analysis using with a Wilcoxon Rank-Sum Test to detect distribution shifts between a subset of genes and all genes in annotated pathways. Enrichment results are returned with pathway names, enrichment statistics, Benjamini-Hochberg corrected p-values, and ranked pathway genes.

3 Case study

We demonstrate the features of RERConverge using gene trees derived from the coding sequences of 19,140 genes for 62 mammal species, along with foreground branches that represent lineages of the mammalian phylogeny whose members live predominantly in marine aquatic environments (Chikina et al., [2016]; Meyer et al., [2016]; Partha et al. [in prep]). For this dataset, RERConverge ran in 1 hour and 18 minutes on RStudio with R version 3.3.0 installed on a computer running Windows 10 with 16 GB of RAM and an Intel Core i7-6500U 2.50GHz CPU. Top categories of functional enrichment included olfactory receptors, GPCR signaling (FDR=1.4e-38), transduction (FDR=7.2e-90), GPCR signaling (FDR=1.4e-38), and to explore through enrichment analyses the biological association with these traits.

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References


