codyn: an R package of community dynamics metrics

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Abstract

- 1. New analytical tools applied to long-term data demonstrate that ecological communities are highly dynamic over time. We developed an R package, codyn, to help ecologists easily implement these metrics and gain broader insights into community dynamics.
- 2. The codyn package provides temporal diversity indices and community stability metrics. All functions are designed to be easily implemented over multiple replicates.
- 3. Temporal diversity indices include species turnover, mean rank shifts and rate of community change over time. Community stability metrics calculate overall stability and patterns of species covariance and synchrony over time, and include a novel null-modeling method to test significance.
- 4. Finally, codyn contains vignettes that describe methods and reproduce figures from published papers to help users contextualize and apply functions to their own data.

Key-words: temporal analysis, biodiversity, long-term data, synchrony

Introduction

As long-term datasets increase in scope and length, new analytical tools are being developed to capture patterns of species interactions over time. This has the potential to deepen our understanding of ecological patterns and processes. For example, many traditional measurements of community structure, such as diversity indices and rank-abundance curves, represent 'snapshots in time' that poorly capture community dynamics. In contrast, the development of analogous temporal metrics, such as species turnover and rank shifts, has highlighted how much communities can vary over time (Collins et al. 2000, 2008; Cleland et al. 2013). Similarly, species dynamics can affect the stability of community properties. A growing number of synthesis studies have applied new metrics to long-term datasets to test what factors influence the stability of ecosystems (Houlahan et al. 2007; Hector et al. 2010; Grman et al. 2010; Hallett et al. 2014). As ecological records continue to grow and data storage and sharing barriers diminish, it is likely that these types of long-term studies will drive new advances in ecology.

However, there are several impediments to the broad adoption of temporal analyses by the ecological community. First, many metrics of community dynamics yet to be made publically available as functions in R. Although R has become the programming language of choice for most ecologists and environmental scientists, the majority of ecologists rely on functions published in public repositories like the Comprehensive R Archive Network (CRAN, http://cran.r-project.org) instead of self-written functions. The primary R package for vegetation analyses, vegan, provides a well-stocked toolbox for community ecologists, with an emphasis on diversity indices and ordination techniques, but does not include temporal metrics that quantify aspects of community dynamics. Second, most temporal metrics are applied to datasets that include spatial as well as temporal replication. This adds an additional layer of complexity that may stall the general use and application of these metrics. Third, many metrics of temporal dynamics are actively being developed, and so ecologists may be unfamiliar with the metrics available and their relative merits.

Here we present a new R package, codyn, that fills this niche for published functions that characterize aspects of temporal communities dynamics Short for "community dynamics," codyn functions are explicitly temporal and can be easily implemented over multiple spatial replicates or experimental units. Functions in codyn

fall into two categories: temporal diversity indices and community stability metrics. Below we describe the metrics by category and demonstrate their application with examples from the Konza Prairie Long-Term Ecological Research (LTER) Site, in Kansas, USA.

Temporal diversity indices

The diversity indices in codyn are temporal analogs to traditional diversity indices such as richness and rank-abundance curves. They include species turnover (turnover), mean shifts in rank abundance (mean_rank_shift) (Collins et al. 2008) and rate of directional change in community composition over time (rate_change) (Collins et al. 2000).

Species turnover

Species turnover represents a temporal analog to species richness (MacArthur & Wilson 1963; Diamond 1969; Collins et al. 2008; Cleland et al. 2013). The function turnover calculates three metrics of species turnover: total turnover, appearances, and disappearances. The default metric total refers to total turnover, which calculates the proportion of species that differ between time points as:

$$Total\ turnover = \frac{Species\ gained\ +\ Species\ lost}{Total\ species\ observed\ in\ both\ timepoints}$$

The turnover function includes the option to calculate only the proportion of species that appear or only those that disappear. This allows detection of differences in the time points in which many species appear versus when species drop from the system, even while the total turnover value in both scenarios may be similar.

Mean rank shifts

Mean rank shifts represent a temporal analog of species rank abundance distributions and indicate the degree of species reording between two time points (Collins *et al.* 2008). This metric is calculated by the mean_rank_shift function as:

$$MRS = \sum_{i=1}^{N} (|R_i, t+1 - R_i, t|)/N$$

where N is the number of species in common in both time points, t is the time point, and R_i , t is the relative rank of species i in time t.

Rate of community change

The rate_change function provides a measure of the rate and direction of change in a community. Community compositional changes are described by Euclidean distances, which are calculated on pair-wise communities across the entire time series. For example, a data set with 6 time intervals will have distance values for five one-interval time lags (e.g., t_1 vs. t_2 , t_2 vs. t_3 ...), four two-interval time lags (e.g., t_1 vs. t_3 , t_2 vs. t_4 ...) and so forth. These distance values are regressed against the time lag interval. The slope of this relationship indicates the rate and direction of community change (Collins $et\ al.\ 2000$). The full dataset of lagged distances is returned using the function rate_change_interval.

Worked example: Temporal diversity patterns at Konza

The Konza Prairie LTER is a tallgrass prairie research site in which different watersheds have experienced different experimental fire frequencies. Collins et al. (2008) compared the temporal dynamics of grassland plant species in areas of Konza that were either burned annually or unburned. Species abundance data were collected over 18 years in 20 permanently marked $10\ m^2$ circular subplots at an annually burned and an unburned location. To demonstrate the codyn package we utilize the Collins et al. (2008) dataset, presented in two forms: collins08 averages species abundances across subplots within burn treatment, and knz_001d includes all subplots from the annually watershed.

The codyn examples of temporal diversity indices utilize the collins08 dataset:

```
library(codyn)
data("collins08")
```

replicate	year	species	abundance
v	1984	achimill ambrpsil amorcane	0.05 1.35 1.32

Visualizing temporal dynamics with rank clocks As one method to visualize temporal community change codyn includes a vignette to produce "rank clocks" using ggplot2, which plot the rank order of abundance of each species over time in a circle, starting with a vertical axis at 12 o'clock (Collins et al. 2008). Rank clocks highlight that there has been tremendous reordering in the relative abundance of dominant species in the annually burned but not the unburned location at Konza. For example, big bluestem (Andropogon gerardii) decreased substantially in the annually burned plot over time but remained stable and dominant in the unburned plot (Figure 1).

Quantifying temporal dynamics

Turnover Average annual species richness was higher in the unburned site at Konza, although richness varied a fair amount over time in both sites (Figure 2a). Total turnover indicates that, in addition to higher overall richness, there were also greater fluctuations in the species present in the unburned than burned location at Konza (Figure 2b). This insight helps to contextualize the richness patterns; even between consecutive years with similar species richness, the identity of the composite species could differ by 30%. Higher species richness in the unburned plots can also be related to the higher overall proportional species appearances in the unburned plots (Figure 2b). Below is an example of the code to calculate turnover; the remaining functions in codyn follow a similar syntax. Note that 'year' in the table below refers to the second year in a two-year comparison; thus '1985' refers to the turnover from 1984 to 1985.

total	year	replicate
0.237	1985	annually burned

total	year	replicate
0.302 0.242	1986 1987	annually burned annually burned

Mean rank shifts Calculating mean rank shifts indicates that the degree of species reordering at the burned and unburned communities diverged around 1992 (Figure 2c). In particular, rank shifts decreased and stabilized in the annually burned site following a substantial spike in species reordering in 1992. This spike corresponds with the marked decline in *Andropogon gerardii* highlighted by the rank clocks.

Rate of community change Rank shifts could reflect continual reshuffling of species within the same community type, or successional change from one community type to another. Calculating the rate of community change at Konza indicates that rank shifts in the annually burned site reflect directional change, whereas rank shifts in the unburned site occur despite a lack of directional change (Figure 2d). Taken together, the three codyn diversity indices tell a story of two communities with high species turnover but different degrees of directional change – both elements which are masked by non-temporal indices such as richness.

Community stability metrics

Ecologists have long debated the relationship between species diversity and stability (MacArthur 1955; Tilman 1996; Cottingham et al. 2001). Unstable species populations may stabilize aggregate community properties if a decrease in one species is compensated for by an increase in another. That is, abundance on one species may decline while another increases, resulting in no net change in the aggregate variable net primary production. In a time series, this should be reflected by a pattern in which species negatively covary or fluctuate asynchronously while total community abundance remains relatively stable (Yachi & Loreau 1999). codyn includes a function to characterize community stability, community_stability, and three metrics to characterize species covariance (variance_ratio) (Schluter 1984; Hallett et al. 2014) and asynchrony (synchrony with two options) (Loreau & Mazancourt 2008; Gross et al. 2014).

Community stability

The community_stability function aggregates species abundances within replicate and time period, and uses these values to calculate community stability as the temporal mean divided by the temporal standard deviation (Tilman 1999).

Variance ratio

The variance ratio was one of the first metrics to characterize patterns of species covariance (Schluter 1984) and was used in an early synthesis paper of species covariance in long time series (Houlahan *et al.* 2007). The metric compares the variance of the community (C) as a whole relative to the sum of the individual population (x_i) variances:

$$VR = \frac{Var(C)}{\sum_{i}^{N} Var(x_i)}$$

where:

$$Var(C) = \sum_{i=1}^{N} Var(x_i) + 2 \left(\sum_{i=1}^{N-1} \sum_{j=i+1}^{N} Covar(x_i, x_j) \right)$$

If species vary independently then the variance ratio will be close to 1. A variance ratio < 1 indicates predominately negative species covariance, whereas a variance ratio > 1 indicates that species generally positively covary.

Significance testing The variance ratio remains widely used but has been subject to a number of criticisms. Importantly, early uses of the variance ratio either did not include significance tests, or tested significance by comparing observed values to those returned by scrambling each species' time series. Null models using fully-scrambled species time series can generate spurious null expectations of covariance because the process disrupts within-species autocorrelation. Phase-scrambling (Grman et al. 2010) and a temporal modification of the torus-translation (adapted from Harms et al. 2001; Hallett et al. 2014) have been used to address this issue.

The variance_ratio function uses the temporal torus translation to conduct null modeling for significance tests. In this method a starting time point is randomly selected for each species' time series. This generates a null community matrix in which species abundances vary independently but within-species autocorrelation is maintained (for each species, the time series is disrupted only once). codyn also includes the option to apply the temporal torus translation for other test statistics using the temporal_torus_translation function, which returns the output of a user-specified function when applied to a null community for functions that require a species x time matrix and the temporal_torus_translation_CI, which returns confidence intervals for a user-specified function that returns a single-value test statistic and requires a species x time matrix.

Synchrony Option 1: "Loreau"

Loreau and de Mazancourt (2008) developed a metric of species synchrony that compares the variance of aggregated species abundances with the summed variances of individual species:

$$Synchrony = \frac{\sigma_{i}(x_{T})^{2}}{(\sum_{i} \sigma_{i}(x_{i}))^{2}}$$

where:

$$x_T(t) = \sum_{i=1}^{N} x_i(t)$$

This measure of synchrony is standardized between 0 (perfect asynchrony) and 1 (perfect synchrony) and is the default calculation in the synchrony function. A virtue of this metric is that it can be applied across communities of variable species richness. It can also be applied not only to species abundance but also population size and per capita growth rate. However, unlike the variance ratio it does not lend itself to significance testing.

Synchrony Option 2: "Gross"

Gross et al. (2014) developed a metric of synchrony that compares the average correlation of each individual species with the rest of the aggregated community:

$$Synchrony = (1/N) \sum_{i} Corr(x_i, \sum_{i \neq j} x_j)$$

This measure of synchrony is standardized from -1 (perfect asynchrony) to 1 (perfect synchrony) and is centered at 0 when species fluctuate independently. It can be specified in the **synchrony** function. This metric is not sensitive to richness and has the potential for null-model significance testing. It may under-perform on

short time series because it is based on correlation, and care should be taken when applying it to communities that contain very stable species (i.e., whose abundances do not change throughout the time series).

Worked example: Community stability at Konza

The codyn examples of temporal diversity indices utilize the knz_001d dataset:

```
library(codyn)
data("knz_001d")
```

species	year	subplot	abundance
achillea millefolium	1986	A_1	0.5
ambrosia psilostachya	1988	A_1	0.5
ambrosia psilostachya	1990	A_1	3.0

Community stability (i.e., the stability of aggregate cover over time) varied considerably in the annually burned watershed (Figure 3). There was a strong negative relationship between community stability and both the variance ratio and synchrony as calculated by Loreau (Figure 3a,b), indicating that greater negative covariance/higher asynchrony was associated with increased stability. This relationship was much weaker for synchrony as calculated by Gross (Figure 3c). This is because both the Loreau method and the variance ratio are based on variances, and are therefore more heavily influenced by abundant species (which also comprise most of the aggregate cover). In contrast, the Gross method is based on correlation and consequently weights all species equally. Differences between the metrics should decrease in more even communities.

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Data accessibility

The codyn package is available on CRAN (https://cran.r-project.org/web/packages/codyn/index.html) and GitHub at (https://github.com/laurenmh/codyn). Data used in the examples presented here are included in the package.Bug reports and feature requests can be sent using the GitHub issue tracker.

Figures

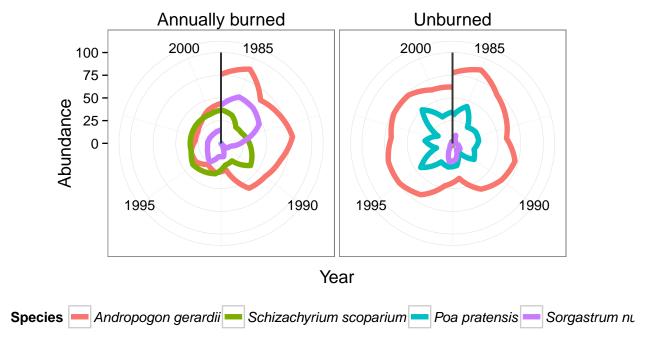


Figure 1. Rank clock of the four dominant grassland plant species in the annually burned versus unburned locations at Konza Prairie, a long-term ecological research site in Kansas, USA. Vertical black bars show the starting "12 o'clock" position on the rank clock.

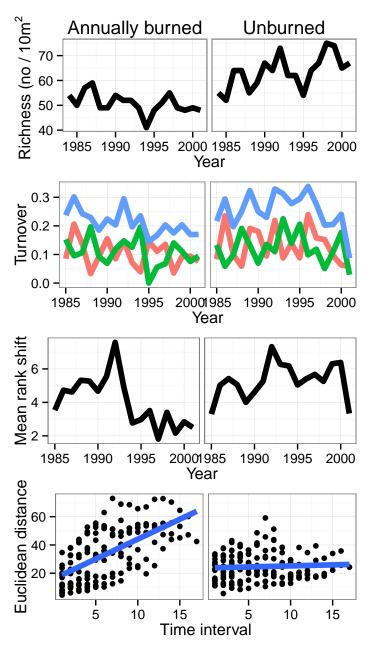


Figure 2. Species richness, turnover (total = blue, appearances = red, disappearances = green), mean rank shifts and community change over time for plant species at the annually burned and unburned location at Konza Prairie, a long-term ecological research site in Kansas, USA.

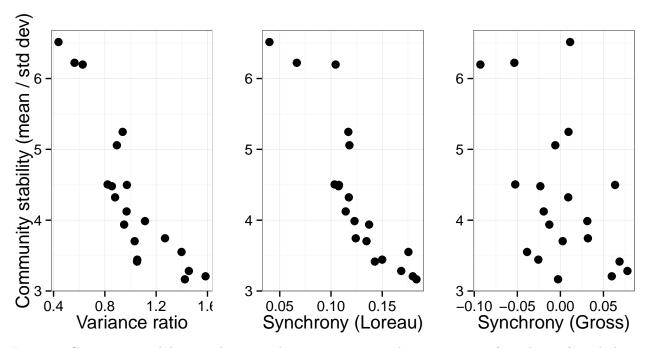


Figure 3. Community stability in relation to the variance ratio and two measures of synchrony for subplots in the annually burned location at at Konza Prairie, a long-term ecological research site in Kansas, USA.

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