

Ecological Dynamic Regimes

Identification, characterization, and comparison

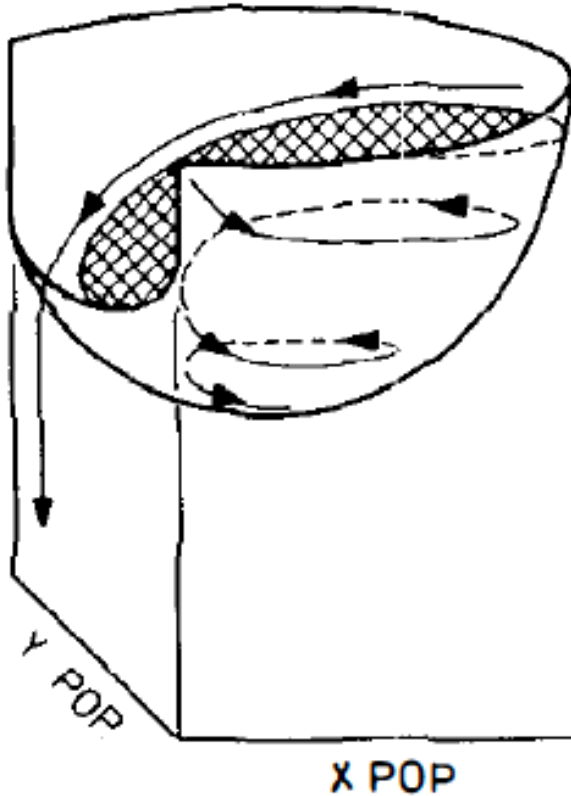
Martina Sánchez-Pinillos

Ecological Dynamic Regimes (EDR)

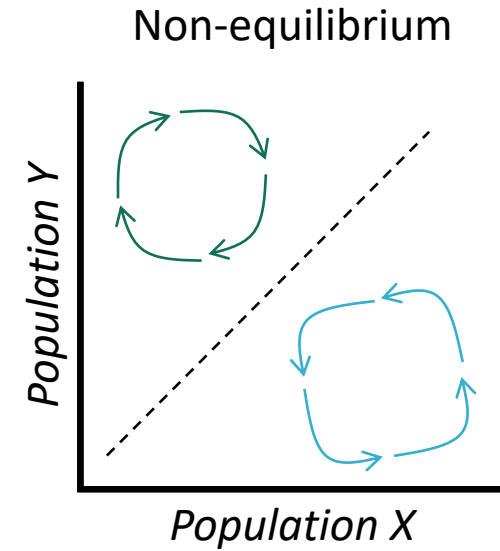
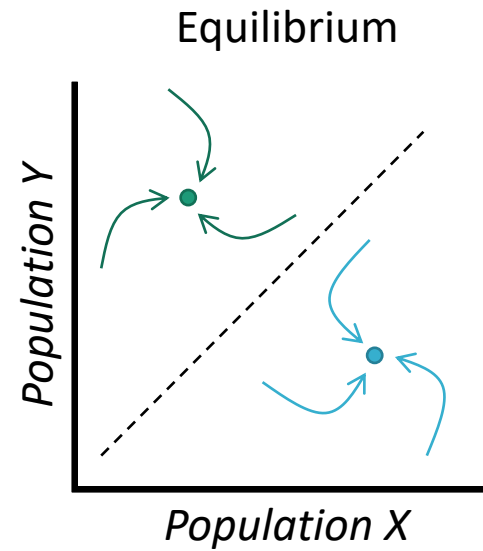
Natural fluctuations of ecosystem states around some trend or average resulting from an intricate mix of internal processes and external forces that push the system towards specific domains of attraction

Ecological Dynamic Regimes and Resilience

Theoretically...

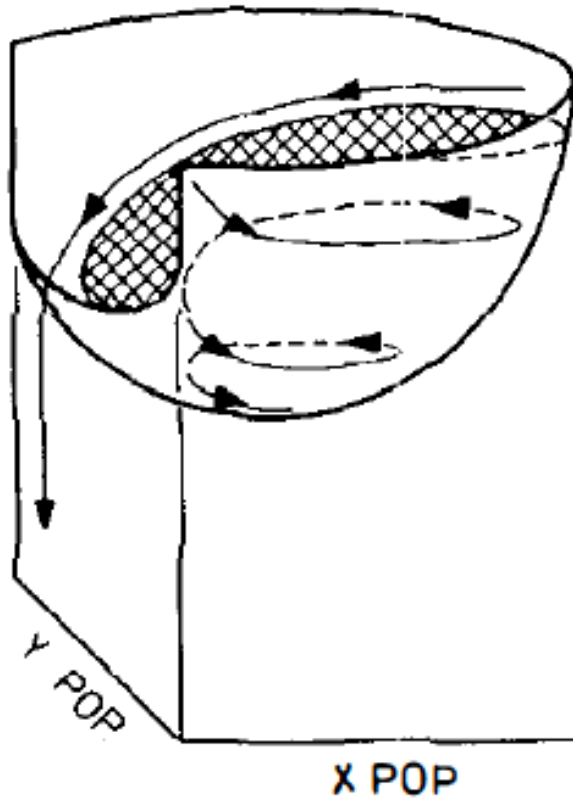


Holling (1973)



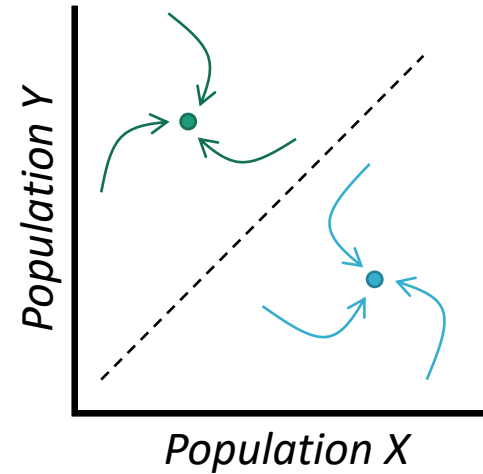
Ecological Dynamic Regimes and Resilience

Theoretically...

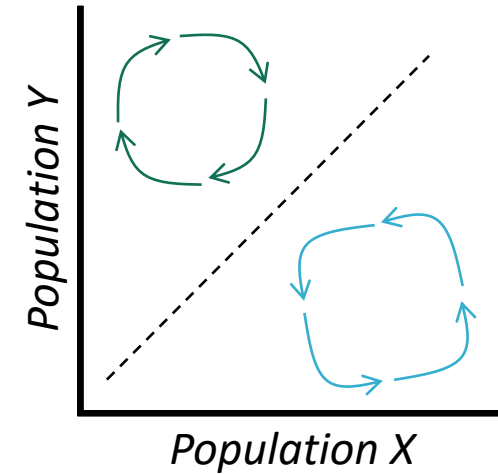


Holling (1973)

Equilibrium

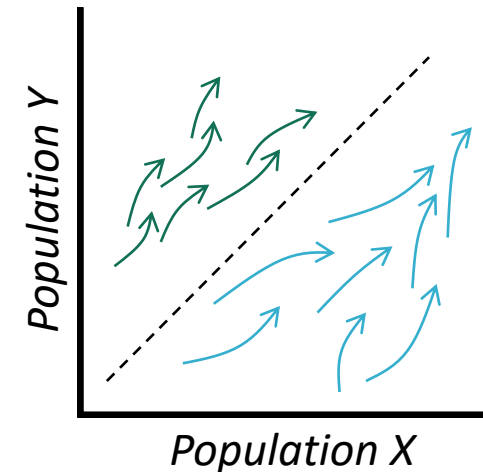


Non-equilibrium



In real ecosystems...

Transient states



- Interacting factors
- Heterogeneity
- Historical legacies
- Changing conditions
- Stochasticity

- Identify EDR
- Characterize EDR
- Compare EDR

... from empirical data

The EDR framework

Received: 15 February 2023

Revised: 19 June 2023

Accepted: 21 June 2023

DOI: 10.1002/ecm.1589

ARTICLE

ECOLOGICAL
MONOGRAPHS
ECOLOGICAL SOCIETY OF AMERICA

Ecological dynamic regimes: Identification, characterization, and comparison

Martina Sánchez-Pinillos¹  | Sonia Kéfi¹ | Miquel De Cáceres²  |
Vasilis Dakos¹



<https://doi.org/10.1002/ecm.1589>

The R package 'ecoregime'



Analysis of Ecological Dynamic Regimes

A **toolbox** for implementing the Ecological Dynamic Regime framework

CRAN: <https://CRAN.R-project.org/package=ecoregime>

Website: <https://mspinillos.github.io/ecoregime/>

```
# Install and load ecoregime  
install.packages("ecoregime")  
library(ecoregime)
```

WARNING!



There is no universal rule to perform the EDR framework.
Some analyses depend on the data characteristics and the objectives pursued.

Raw data

Inventory data

	Site	Obs.	sp1	sp2	sp3
1	A	1	0.8	0.2	0
2	A	2	0.6	0.4	0
3	A	3	0.2	0.8	0
4	B	1	0.1	0.3	0.6
5	B	2	0.1	0.1	0.8
:	:	:	:	:	:
N	M	n	0	1	0

State dissimilarities

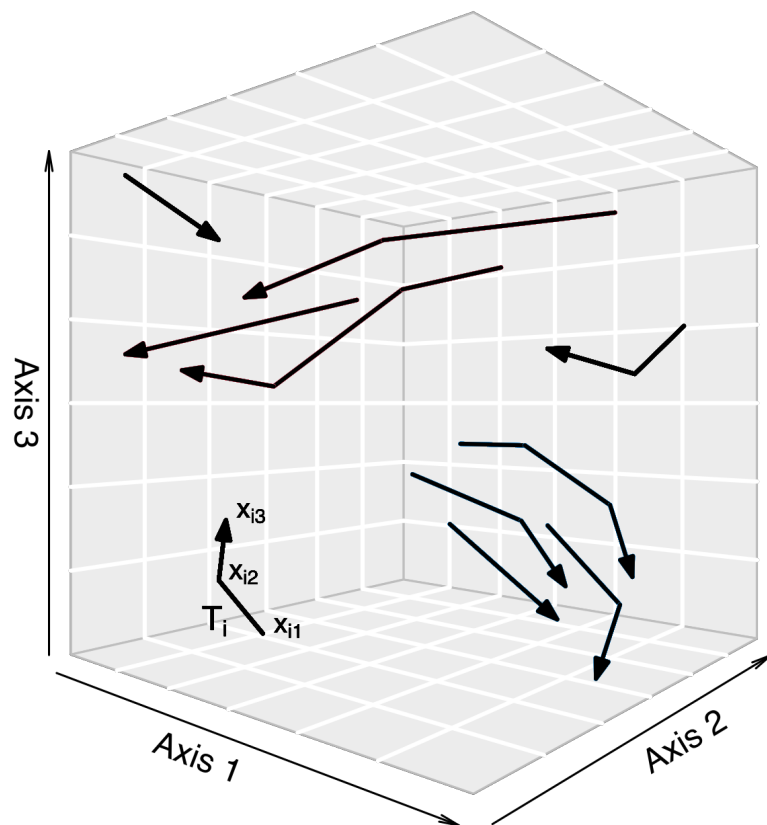
Inventory data

	Site	Obs.	sp1	sp2	sp3
1	A	1	0.8	0.2	0
2	A	2	0.6	0.4	0
3	A	3	0.2	0.8	0
4	B	1	0.1	0.3	0.6
5	B	2	0.1	0.1	0.8
:	:	:	:	:	:
N	M	n	0	1	0

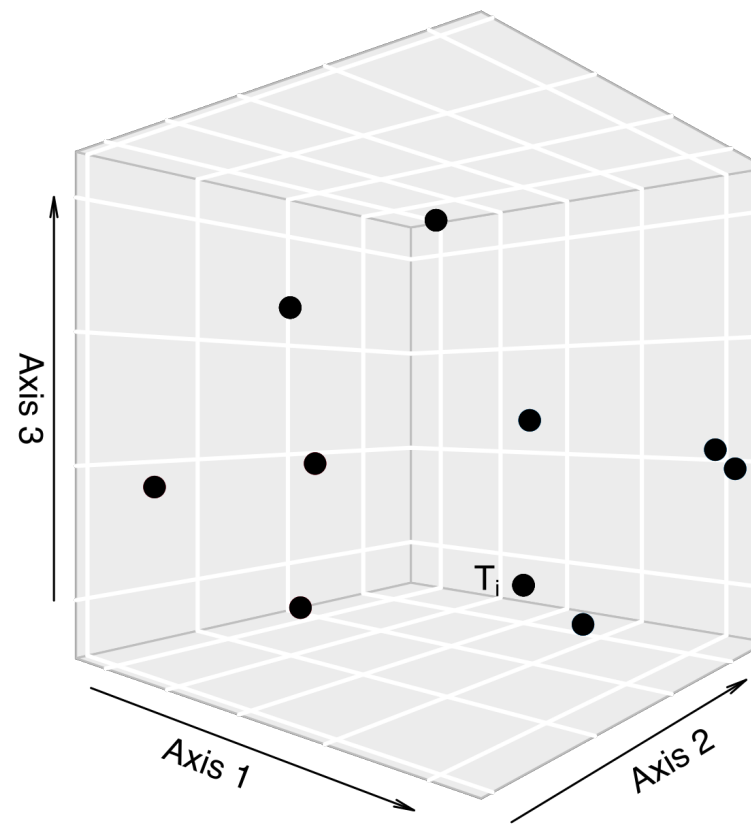
State dissimilarities

		A1	A2	A3	B1	B2	...	Mn
1	A1	$D_{P,N \times N}$						
2	A2							
3	A3							
4	B1							
5	B2							
:	:							
N	Mn							

State space



Trajectory space



Raw data

Inventory data

	Site	Obs.	sp1	sp2	sp3
1	A	1	0.8	0.2	0
2	A	2	0.6	0.4	0
3	A	3	0.2	0.8	0
4	B	1	0.1	0.3	0.6
5	B	2	0.1	0.1	0.8
:	:	:	:	:	:
N	M	n	0	1	0

```
# Species abundances
```

```
abun <- rbind(EDR_data$EDR1$abundance,  
              EDR_data$EDR2$abundance,  
              EDR_data$EDR3$abundance)
```

```
# ID trajectory
```

```
abun$ID <- paste0(abun$EDR, "_", abun$traj)
```

State dissimilarities

```
# State dissimilarities (e.g. Bray-Curtis)

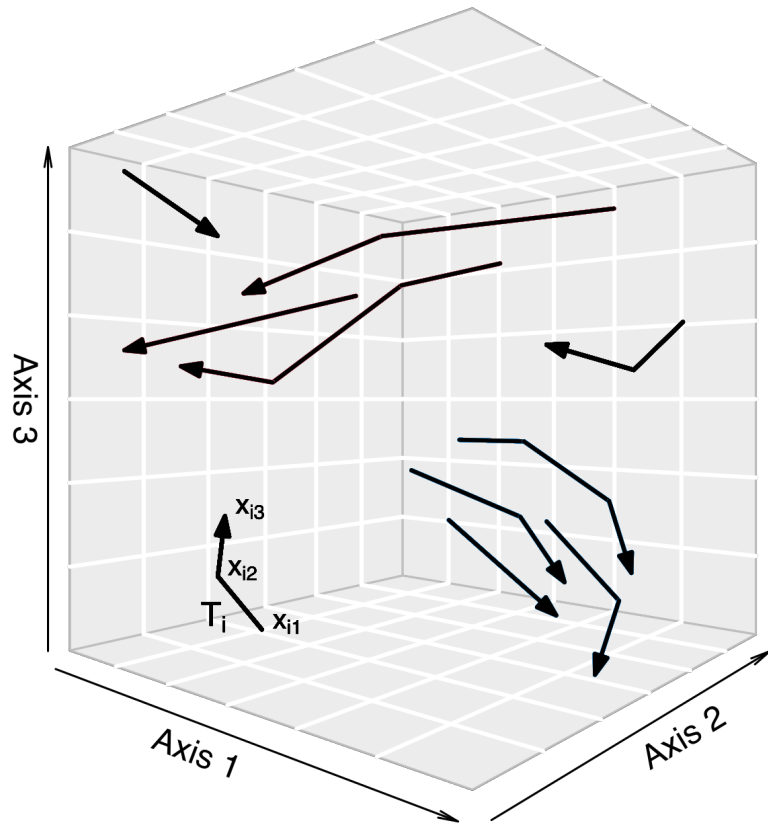
library(vegan)

dStates <- vegdist(x = abund[, paste0("sp", 1:12)],
                   method = "bray") *
```

State dissimilarities

		A1	A2	A3	B1	B2	...	Mn
1	A1	$D_{P,N \times N}$						
2	A2							
3	A3							
4	B1							
5	B2							
⋮	⋮							
N	Mn							

State space



```
# Plot trajectories
```

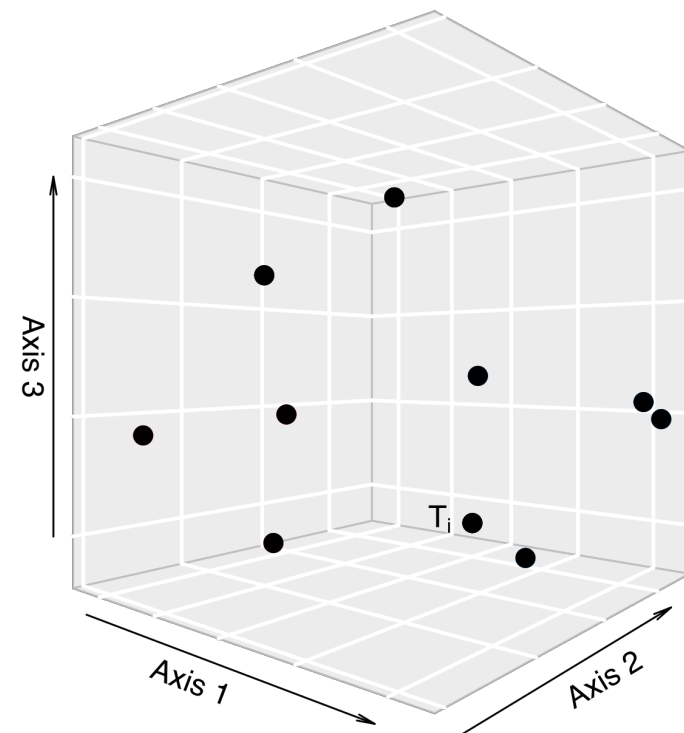
```
library(ecotraj)
```

```
trajectoryPCoA(d = dStates,  
               sites = abund$ID,  
               surveys = abund$state)
```

Trajectory space

Trajectory dissimilarities

```
dTraj <- trajectoryDistances(d = dStates,  
                              sites = abun$ID,  
                              surveys = abun$state,  
                              distance.type = "DSPD") *
```



Trajectory space

Trajectory dissimilarities

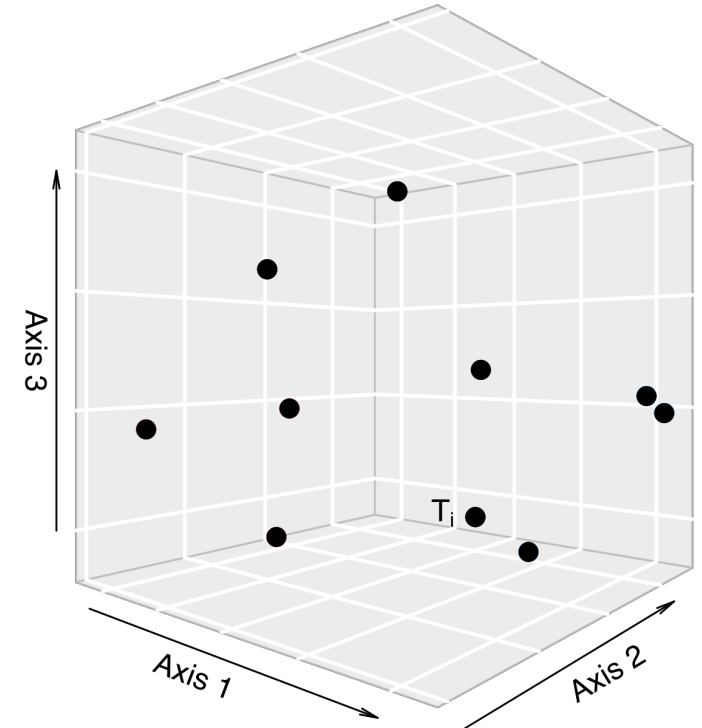
```
dTraj <- trajectoryDistances(d = dStates,  
                             sites = abun$ID,  
                             surveys = abun$state,  
                             distance.type = "DSPD") *
```

Trajectory space (PCoA)

```
pcoa_traj <- cmdscale(dTraj, k = nrow(as.matrix(dTraj)) - 1, add = T)  
traj_coord <- pcoa_traj$points
```

Plot the trajectory space

```
plot(x = traj_coord[, 1], y = traj_coord[, 2],  
     xlab = "Axis 1", ylab = "Axis 2",  
     main = "Trajectory space")
```



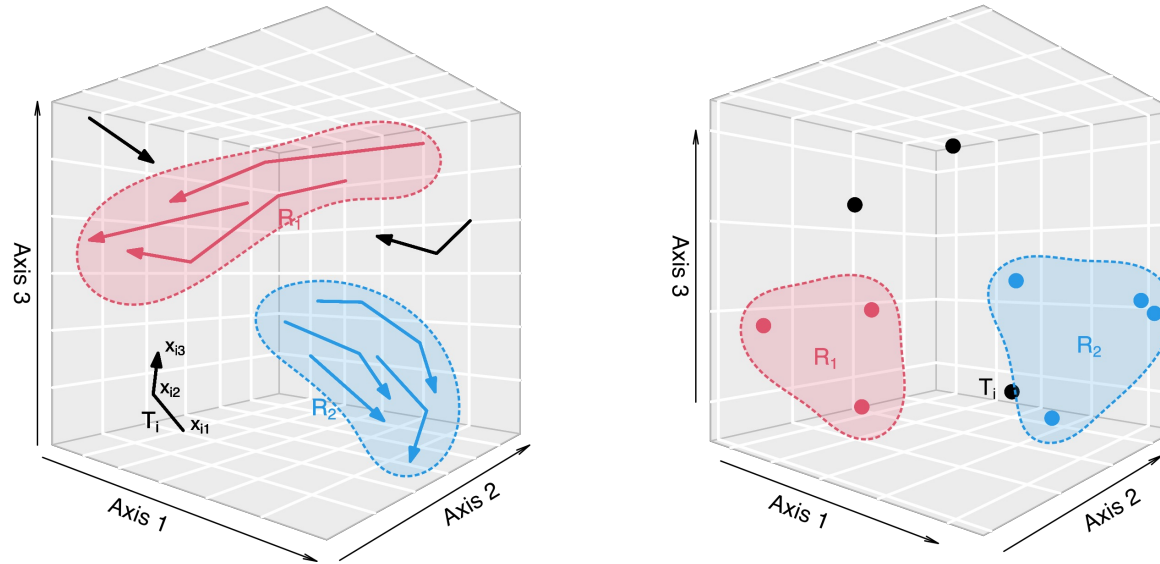
How do we identify EDRs from
empirical data?

Identify EDRs

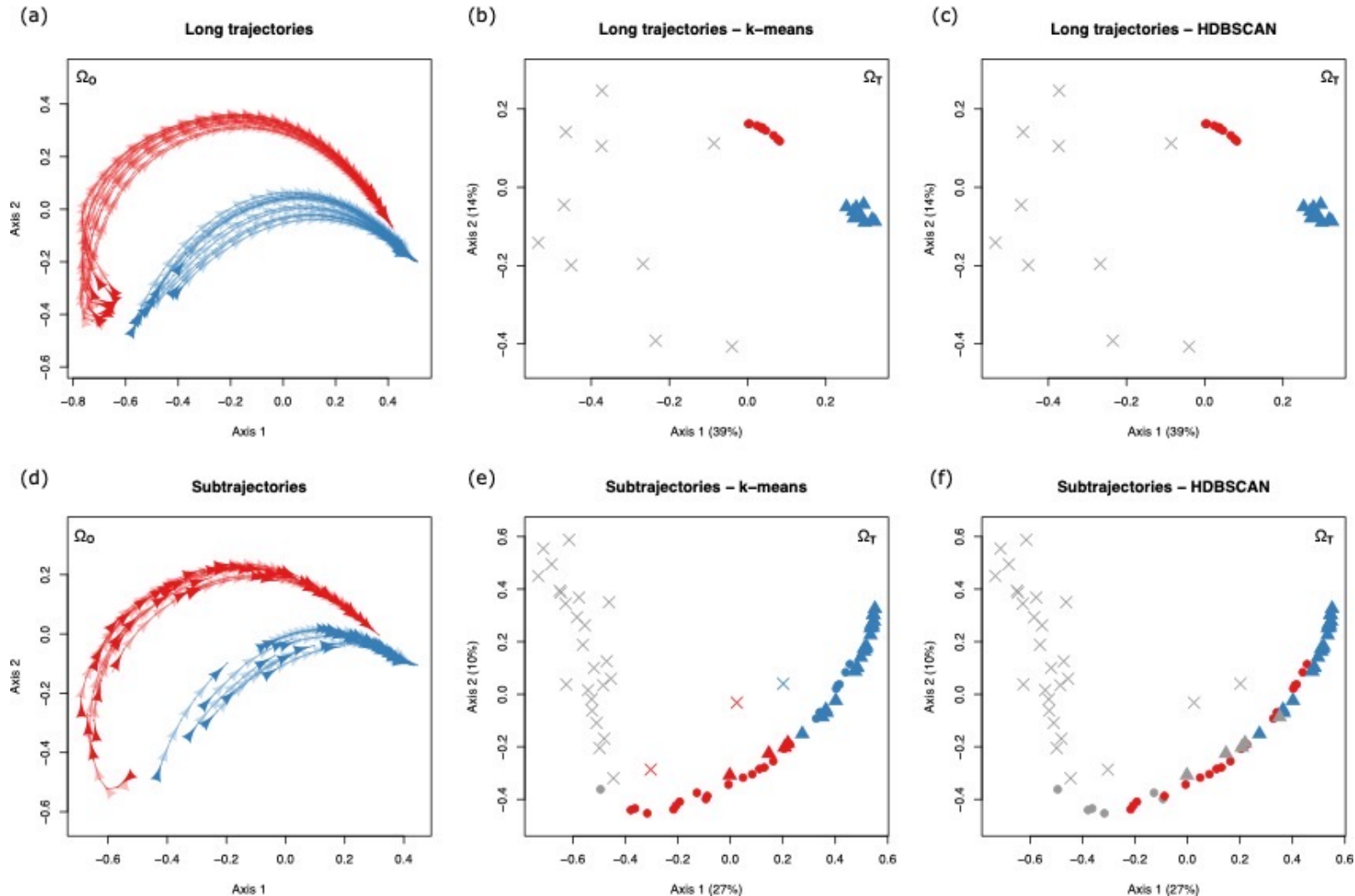
*Identifying **subsets of ecological trajectories** with more **similar geometric patterns** between each other than with any other trajectory in the same state space*

Identify EDRs

Identifying *subsets of ecological trajectories* with more *similar geometric patterns* between each other than with any other trajectory in the same state space



Clustering algorithms

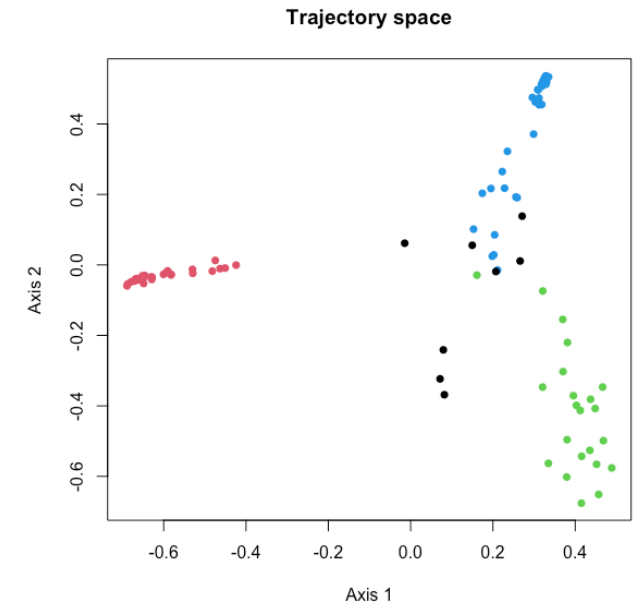


Identify EDRs

```
# Clustering analysis (e.g., HDBSCAN) *  
library(dbSCAN)  
EDR <- hdbSCAN(x = dTraj, minPts = 10)  
EDR_cluster <- data.frame(ID = unique(abun$ID),  
                           EDR_cluster = EDR$cluster)
```

Identify EDRs

```
# Clustering analysis (e.g., HDBSCAN) *  
library(dbSCAN)  
EDR <- hdbscan(x = dTraj, minPts = 10)  
EDR_cluster <- data.frame(ID = unique(abun$ID),  
                           EDR_cluster = EDR$cluster)  
  
# Plot the trajectory space  
plot(x = traj_coord[, 1], y = traj_coord[, 2],  
      xlab = "Axis 1", ylab = "Axis 2",  
      main = "Trajectory space",  
      col = EDR_cluster$EDR_cluster + 1)
```



Identify EDRs

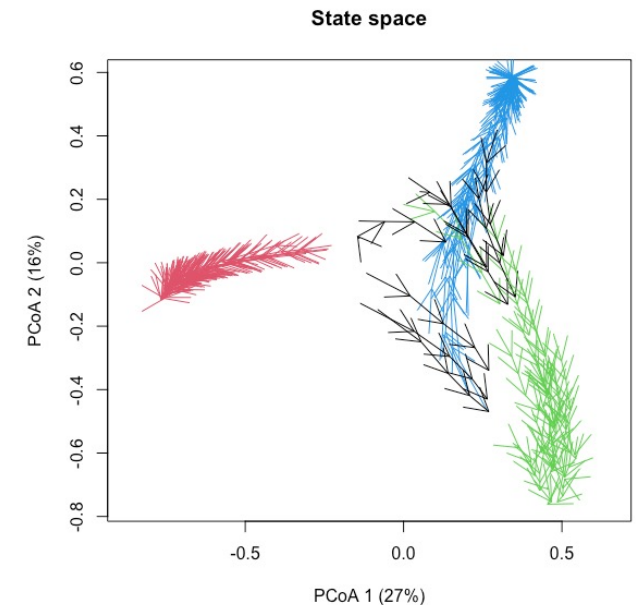
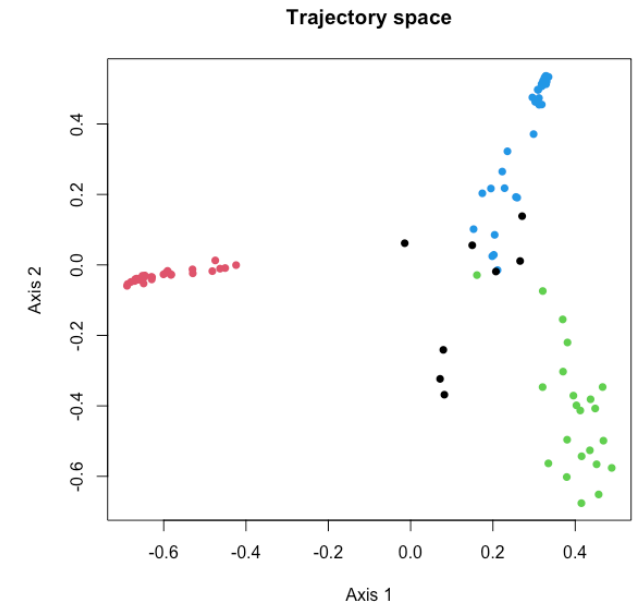
```
# Clustering analysis (e.g., HDBSCAN) *  
library(dbSCAN)  
EDR <- hdbscan(x = dTraj, minPts = 10)  
EDR_cluster <- data.frame(ID = unique(abun$ID),  
                           EDR_cluster = EDR$cluster)
```

```
# Plot the trajectory space
```

```
plot(x = traj_coord[, 1], y = traj_coord[, 2],  
     xlab = "Axis 1", ylab = "Axis 2",  
     main = "Trajectory space",  
     col = EDR_cluster$EDR_cluster + 1)
```

```
# Plot trajectories in the state space
```

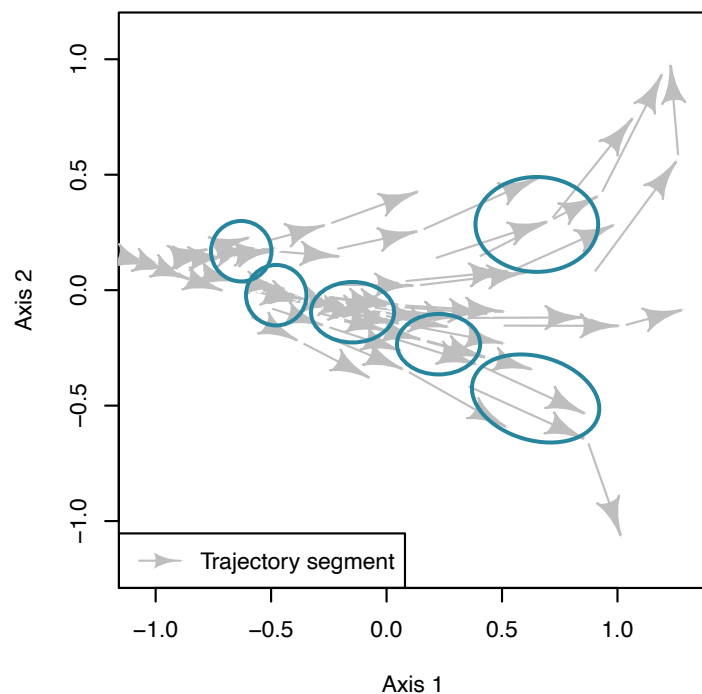
```
trajectoryPCoA(d = dStates,  
              sites = abun$ID,  
              surveys = abun$state,  
              traj.colors = EDR_cluster$EDR_cluster + 1)
```



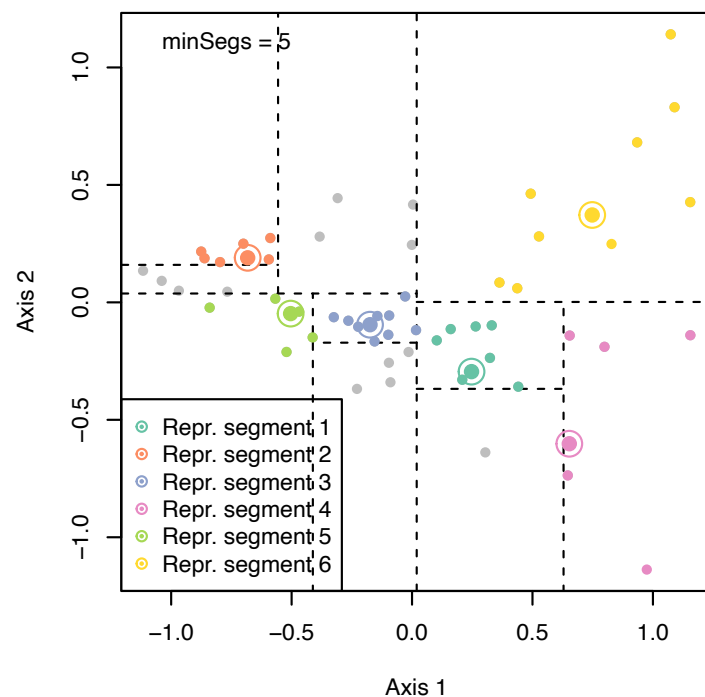
Can we summarize the **main**
dynamical patterns of an EDR?

Characterize EDRs

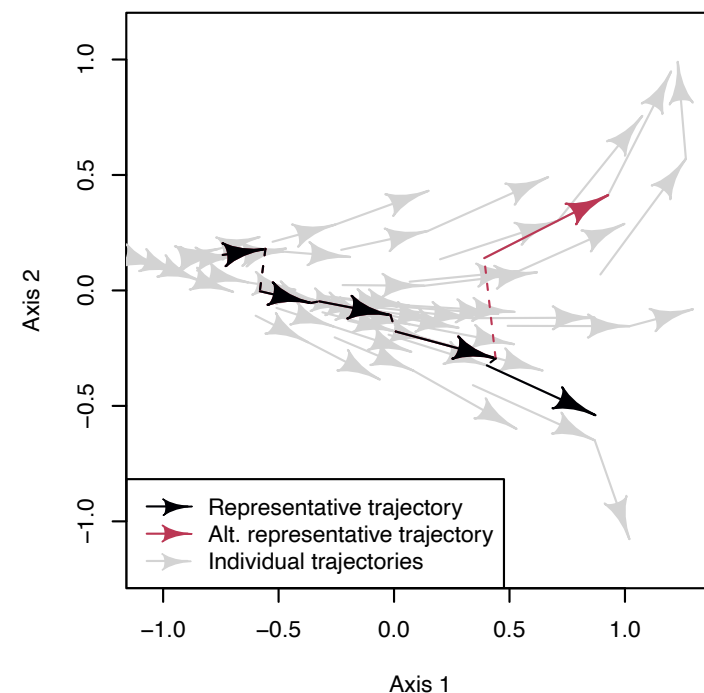
RETRA-EDR: REpresentative TRAjectories in EDRs



Look for dense regions
in the EDR



Identify representative
segments of each dense region



Merge representative
segments

Representative trajectories

```
# Select the EDR
```

```
ID_EDR <- which(abun$EDR == 1)
```

```
# Apply RETRA-EDR
```

```
RT <- retra_edr(d = as.matrix(dStates)[ID_EDR, ID_EDR],  
               trajectories = abun[ID_EDR]$traj,  
               states = abun[ID_EDR]$state,  
               minSegs = 5) *
```

Representative trajectories

```
# Select the EDR
```

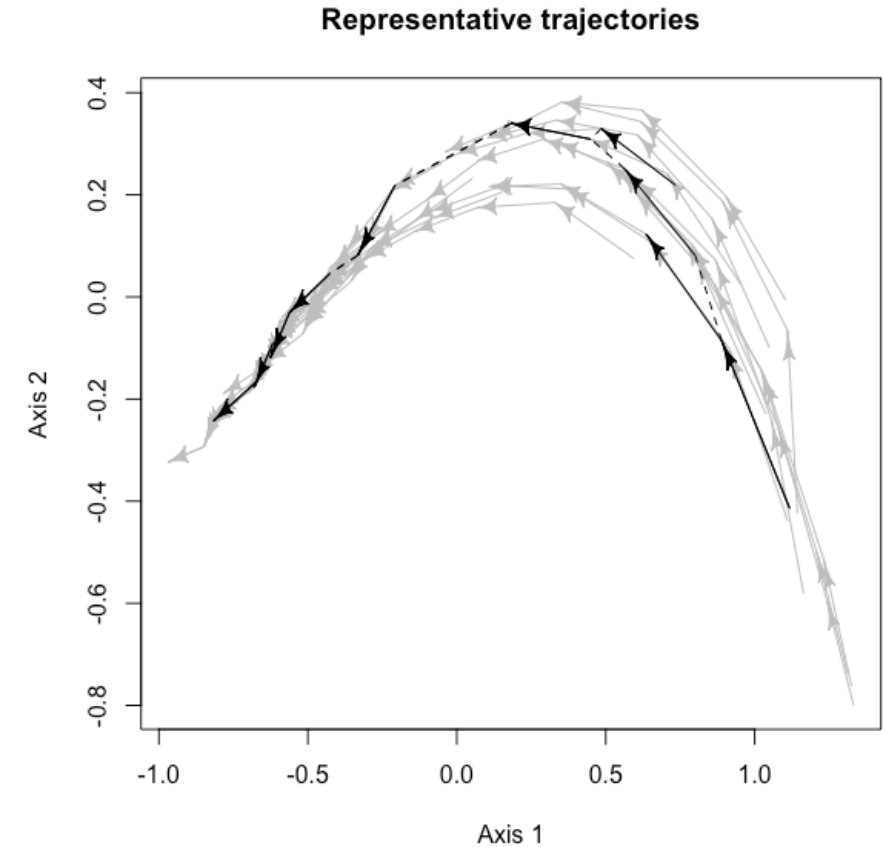
```
ID_EDR <- which(abun$EDR == 1)
```

```
# Apply RETRA-EDR
```

```
RT <- retra_edr(d = as.matrix(dStates)[ID_EDR, ID_EDR],  
               trajectories = abun[ID_EDR]$traj,  
               states = abun[ID_EDR]$state,  
               minSegs = 5) *
```

```
# Plot representative trajectories
```

```
plot(x = RT, d = as.matrix(dStates)[ID_EDR, ID_EDR],  
     trajectories = abun[ID_EDR]$traj,  
     states = abun[ID_EDR]$state,  
     main = "Representative trajectories")
```



Representative trajectories

```
# Extract field data for representative trajectories
```

```
seg_components <- strsplit(gsub("\\]", "", gsub("\\[", "-", RT$T2$Segments)), "-")
RT_data <- do.call(rbind, lapply(seg_components, function(iseg){
  data.frame(traj = rep(iseg[[1]], 2),
             state = c(iseg[[2]], iseg[[3]]))
}))
RT_data <- merge(RT_data, abund[EDR == 1], all.x = T, sort = F)
```

Representative trajectories

```
# Extract field data for representative trajectories
```

```
seg_components <- strsplit(gsub("\\]", "", gsub("\\[", "-", RT$T2$Segments)), "-")
```

```
RT_data <- do.call(rbind, lapply(seg_components, function(iseg){  
  data.frame(traj = rep(iseg[[1]], 2),  
             state = c(iseg[[2]], iseg[[3]]))  
}))
```

```
RT_data <- merge(RT_data, abund[EDR == 1], all.x = T, sort = F)
```

```
# Plot changes in species abundances
```

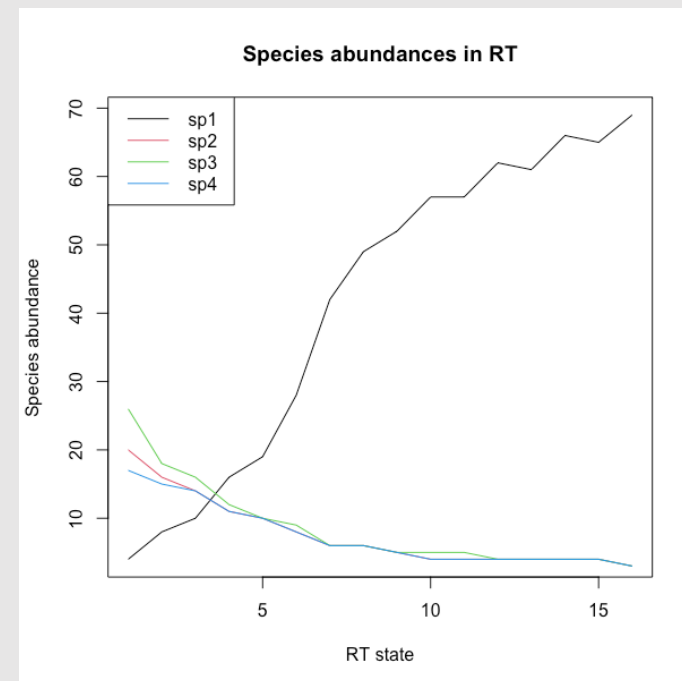
```
plot(x = 1:nrow(RT_data), y = RT_data$sp1, type = "l",  
     xlab = "RT state", ylab = "Species abundance",  
     main = "Species abundances in RT")
```

```
lines(x = 1:nrow(RT_data), y = RT_data$sp2, col = 2)
```

```
lines(x = 1:nrow(RT_data), y = RT_data$sp3, col = 3)
```

```
lines(x = 1:nrow(RT_data), y = RT_data$sp4, col = 4)
```

```
legend("topleft", paste0("sp", 1:4), lty = 1, col = 1:4)
```



How is the distribution of
individual trajectories in an EDR?

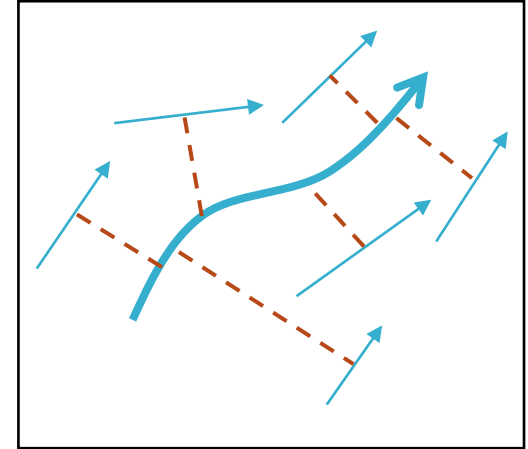
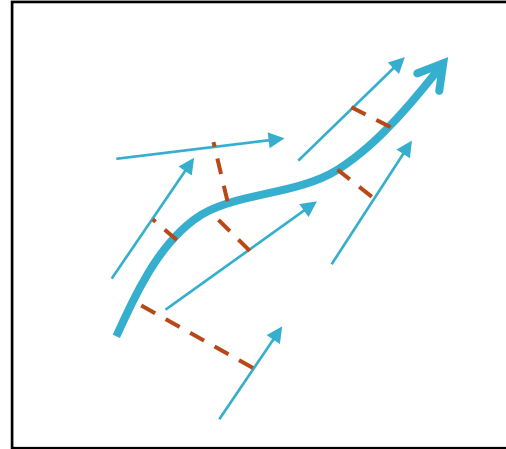
Characterize EDRs

Dynamic dispersion ($dDis$)

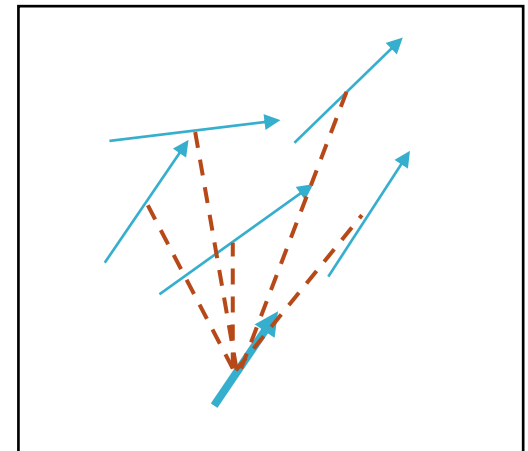
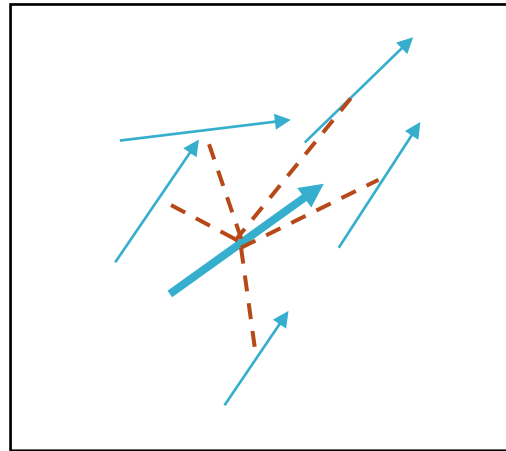
$$dDis = \frac{\sum_{i=1}^m d_{i\alpha}}{m}$$

Average distance to a trajectory of reference

$\alpha = \text{representative trajectory}$



$\alpha = \text{individual trajectory}$

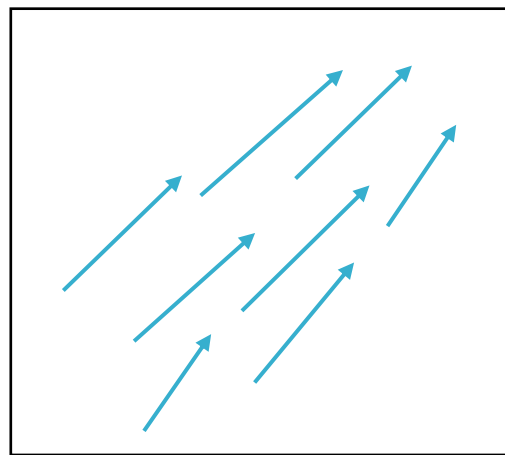
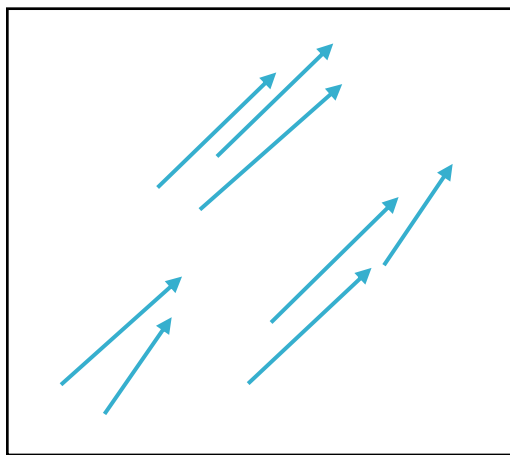


Characterize EDRs

Dynamic evenness (*dEve*)

$$dEve = \frac{\sum_{l=1}^{m-1} \min \left(\frac{d_{ij}}{\sum_{l=1}^{m-1} d_{ij}}, \frac{1}{m-1} \right) - \frac{1}{m-1}}{1 - \frac{1}{m-1}}$$

Regularity with which the EDR is filled by the individual trajectories

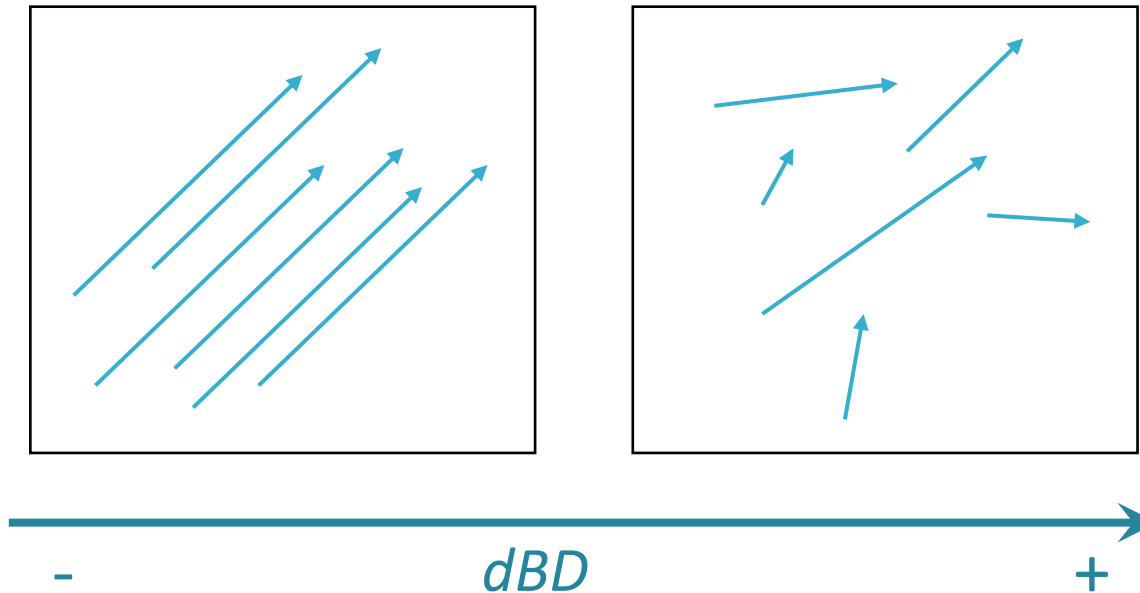


Characterize EDRs

Dynamic beta diversity (*dBD*)

$$dBD = \frac{\sum_{i=1}^{m-1} \sum_{j=i+1}^m d_{ij}^2}{m(m-1)}$$

Overall variation of ecological trajectories belonging to the same EDR



Distribution of the trajectories in the EDR

```
# Dynamic dispersion
```

```
dDis <- dDis(d = as.matrix(dStates)[ID_EDR, ID_EDR], d.type = "dStates",  
            trajectories = abund[ID_EDR]$traj,  
            states = abund[ID_EDR]$state,  
            reference = 28)
```

Distribution of the trajectories in the EDR

```
# Dynamic dispersion
```

```
dDis <- dDis(d = as.matrix(dStates)[ID_EDR, ID_EDR], d.type = "dStates",  
            trajectories = abund[ID_EDR]$traj,  
            states = abund[ID_EDR]$state,  
            reference = 28)
```

```
# Dynamic evenness
```

```
dEve <- dEve(d = as.matrix(dStates)[ID_EDR, ID_EDR], d.type = "dStates",  
            trajectories = abund[ID_EDR]$traj,  
            states = abund[ID_EDR]$state)
```

Distribution of the trajectories in the EDR

Dynamic dispersion

```
dDis <- dDis(d = as.matrix(dStates)[ID_EDR, ID_EDR], d.type = "dStates",  
            trajectories = abund[ID_EDR]$traj,  
            states = abund[ID_EDR]$state,  
            reference = 28)
```

Dynamic evenness

```
dEve <- dEve(d = as.matrix(dStates)[ID_EDR, ID_EDR], d.type = "dStates",  
            trajectories = abund[ID_EDR]$traj,  
            states = abund[ID_EDR]$state)
```

Dynamic beta diversity

```
dBD <- dBD(d = as.matrix(dStates)[ID_EDR, ID_EDR], d.type = "dStates",  
           trajectories = abund[ID_EDR]$traj,  
           states = abund[ID_EDR]$state)
```

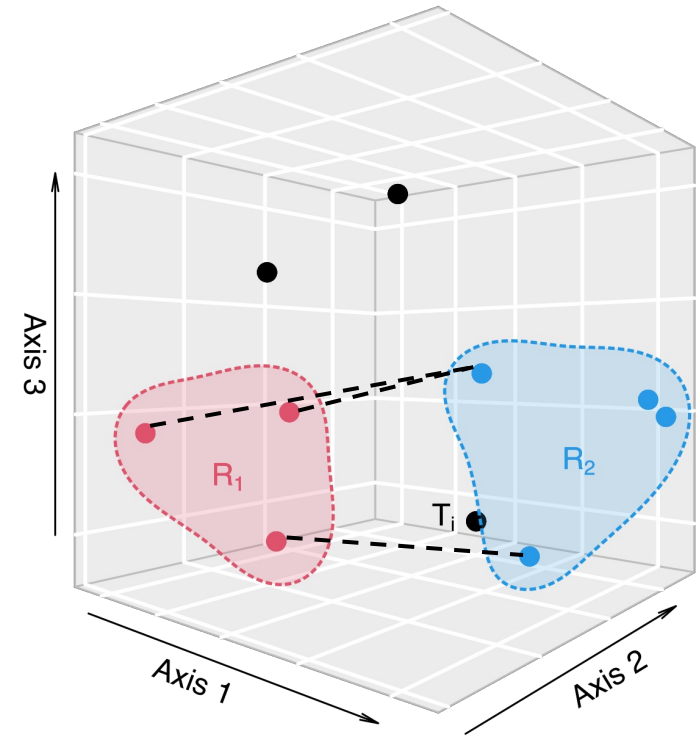
Can we compare multiple EDRs?

Compare EDRs

Dynamic regime dissimilarity (D_{DR})

$$D_{DR}(R_1, R_2) = \frac{1}{m_1} \sum_{i=1}^{m_1} D_{DSP}(T_{1i}, R_2)$$

$$D_{DSP}(T_{1i}, R_2) = \min\{D_{DSP}(T_{1i}, T_{21}), \dots, D_{DSP}(T_{1i}, T_{2m_2})\}$$



Compare EDRs

```
# Identify EDRs in the abundance matrix
```

```
abun <- merge(abun, EDR_cluster, by = "ID", all.x = T)
```

```
# EDR dissimilarity
```

```
dDR <- dist_edr(d = dStates, d.type = "dStates",  
               trajectories = abun$ID,  
               states = abun$state,  
               edr = abun$EDR_cluster)
```


What are the applications and challenges of the EDR framework?

Applications and challenges

Applications

- ✓ Ecological resilience
- ✓ Ecosystem dynamics
- ✓ Space-for-time substitution

Challenges

- Trajectory dissimilarity
- Clustering analyses
- “Curse of dimensionality”

Coming soon...

- *Ecological dynamic regimes: A key concept for assessing ecological resilience*
M. Sánchez-Pinillos, V. Dakos, S. Kéfi (*under review*)
- *Resiliencia forestal post-incendio en base a trayectorias sucesionales al nicho climático de las especies*
G. Codina, E. Batllori, F. Lloret, M. Sánchez-Pinillos (*in progress*)

To know more...

The publication

Received: 15 February 2023 | Revised: 19 June 2023 | Accepted: 21 June 2023
DOI: 10.1002/ecm.1589

ARTICLE

ECOLOGICAL
MONOGRAPHS
ECOLOGICAL SOCIETY OF AMERICA

Ecological dynamic regimes: Identification, characterization, and comparison

Martina Sánchez-Pinillos¹  | Sonia Kéfi¹ | Miquel De Cáceres²  | Vasilis Dakos¹



<https://doi.org/10.1002/ecm.1589>

Appendix S2

Appendix S2

Additional technical information

Ecological Monographs

Ecological Dynamic Regimes: Identification, characterization, and comparison

Martina Sánchez-Pinillos^{1*}, Sonia Kéfi¹, Miquel De Cáceres², Vasilis Dakos¹

¹ISEM, CNRS, Univ. Montpellier, IRD, EPHE, Montpellier, France

²CREAF, Bellaterra (Cerdanyola del Vallès), Spain

Appendix S2. Additional technical information

To know more...



The publication

Received: 15 February 2023 | Revised: 19 June 2023 | Accepted: 21 June 2023
DOI: 10.1002/ecm.1589

ARTICLE

ECOLOGICAL
MONOGRAPHS
ECOLOGICAL SOCIETY OF AMERICA

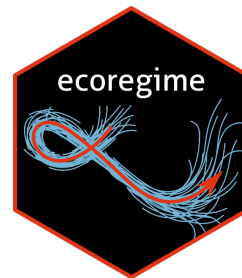
Ecological dynamic regimes: Identification, characterization, and comparison

Martina Sánchez-Pinillos¹  | Sonia Kéfi¹ | Miquel De Cáceres²  | Vasilis Dakos¹



<https://doi.org/10.1002/ecm.1589>

The R package 'ecoregime'



CRAN: <https://CRAN.R-project.org/package=ecoregime>

Website: <https://mspinillos.github.io/ecoregime/>

Appendix S2

Appendix S2

Additional technical information

Ecological Monographs

**Ecological Dynamic Regimes:
Identification, characterization, and comparison**

Martina Sánchez-Pinillos^{1*}, Sonia Kéfi¹, Miquel De Cáceres², Vasilis Dakos¹

¹ISEM, CNRS, Univ. Montpellier, IRD, EPHE, Montpellier, France

²CREAF, Bellaterra (Cerdanyola del Vallès), Spain

Appendix S2. Additional technical information

To know more...



The publication

Received: 15 February 2023 | Revised: 19 June 2023 | Accepted: 21 June 2023
DOI: 10.1002/ecm.1589

ARTICLE

ECOLOGICAL
MONOGRAPHS
ECOLOGICAL SOCIETY OF AMERICA

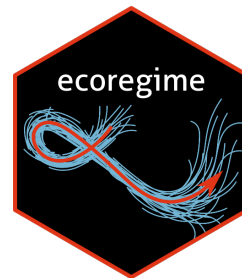
Ecological dynamic regimes: Identification, characterization, and comparison

Martina Sánchez-Pinillos¹  | Sonia Kéfi¹ | Miquel De Cáceres²  | Vasilis Dakos¹



<https://doi.org/10.1002/ecm.1589>

The R package 'ecoregime'



CRAN: <https://CRAN.R-project.org/package=ecoregime>

Website: <https://mspinillos.github.io/ecoregime/>

Appendix S2

Appendix S2

Additional technical information

Ecological Monographs

Ecological Dynamic Regimes:
Identification, characterization, and comparison

Martina Sánchez-Pinillos^{1*}, Sonia Kéfi¹, Miquel De Cáceres², Vasilis Dakos¹

¹ISEM, CNRS, Univ. Montpellier, IRD, EPHE, Montpellier, France

²CREAF, Bellaterra (Cerdanyola del Vallès), Spain

Appendix S2. Additional technical information

The corresponding author

martina.sanchez.pinillos@gmail.com



@MartinaPinillos



MSPinillos



aeet
ASOCIACIÓN
ESPAÑOLA DE
ECOLOGÍA
TERRESTRE

martina.sanchez.pinillos@gmail.com



@MartinaPinillos



MSPinillos

¡Muchas gracias!
Eskerrik asko!

Moitas grazas!
Moltes gràcies!

Co-authors:

Vasilis Dakos

Sonia Kéfi

Miquel De Cáceres



MSCA 891477 – RESET

www.resetforest.weebly.com