

Steye Verhoeve, Sandra Hauswirth, Steven de Jong and Niko Wanders

Background

Wide-spread vegetation drought impacts can result in a regional decrease in ecosystem resilience. The spatial dependence of the post-drought vegetation recovery, i.e. the extent to which events co-occur at multiple locations simultaneously, can shed a light on the processes affecting the co-occurrence of recovery.

In our study, we identify the spatial dependence of vegetation recovery after a drought. Thereby we explain the underlying mechanisms and patterns which could support recovery forecasting in the future.

Methods

- 1. Extraction of **drought events** and the associated timing of different stages (*Fig. 1a*)
- 2. Apply event synchronization (ES) on three different stages which are related to the dynamics of the vegetation recovery (*Fig 1b,c*).
- 3. The ES results in different metrics which characterizes the spatial drought and recovery dynamics (*Fig. 1d, Fig. 2, Fig. 3*).

4. With this information and additional spatial drought characteristics we apply **k-means clustering** (*Fig. 1e, Fig. 4a*).

5. The hydro-meteorological and geographical characteristics are compared between the clusters to explain the spatial recovery dynamics (*Fig. 4b-e*).



Figure 1: Workflow illustrating the ecological drought event assessment Images retrieved from: (b) Odenweller & Donner (2020); (d) Muthuvel & Sivakumar (2024); (e) Wikipedia 'k-means clustering', Weston.pace



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Spatial dependence of vegetation recovery after drought events and the spatiotemporal characteristics





(ALD); (b) propagation potential; (c) the mean orientation of all significant links; (d) the local clustering coefficient; and (e) detail of the source and sink regions.

References

Utrecht University, Department of Physical Geography, Utrecht, Netherlands



different hydro-meteorological and geographical variables in each cluster. Shown here: (b) median annual temperature; (c) sand fraction; (d) elevation; and (e) median leaf area index (LAI) of high vegetation.

Conclusions

- recovery phase of vegetation (*Fig. 3*).
- variance (*Fig 4*).

1. Muthuvel, D., & Sivakumar, B. (2024). Spatial propagation of different drought types and their concurrent societal risks: A complex networks-based analysis. Journal of Hydrology, 636, 131247. https://doi.org/10.1016/j.jhydrol.2024.131247 2. Odenweller, A., & Donner, R. V. (2020). Disentangling synchrony from serial dependency in paired-event time series.

Physical Review E, 101 (5),052213. https://doi.org/10.1103/PhysRevE.101.052213 3. Weston.pace, CC BY-SA 3.0, via https://commons.wikimedia.org/wiki/File:K_Means_Example_Step_4.svg

4. Schwalm et al., (2017). Global patterns of drought recovery. Nature, 548(7666), 202–205. https://doi.org/10.1038/nature23021



Figure 4: (a) Clustering based on event synchronization metrics and drought characteristics, and the spread of

Different drought recovery stages have similar **connectedness patterns** (*Fig. 2*). Multiple event synchronization metrics provide insights into the dynamics of the

The **grouped recovery characteristics** are moderately different in their average annual temperature, whereas many hydro-meteorological and geographical have little

For recovery forecasting the **spatial information** is regarded subordinate compared to **temporal dynamics** when compared to other research (Schwalm et al., 2017).





