

Drought is Nonstationary

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"If we want everything to stay as it is, then everything must change".

Giuseppe Tomasi di Lampedusa, *The Leopard*

Google Scholar search drought “nonstationarity”: 6480 refs!

drought “nonstationary”: 14,300 refs This literature is nonstationary

Key Points

Climate exhibits Power Laws in space and time, with
higher variance at longer timescales, larger spatial areas.

It is naturally, intrinsically, **Nonstationary**.

The *subject* is drought, severe and sustained.

What is the *goal*?

- To detect change?
- To predict future change?
- To contextualize recent events?
- To guide investments in mitigation?
- To predict conditional transition to/from drought?

Tell us the goal, then we can play the game.

Only probabilistic answers are possible.

Key Points

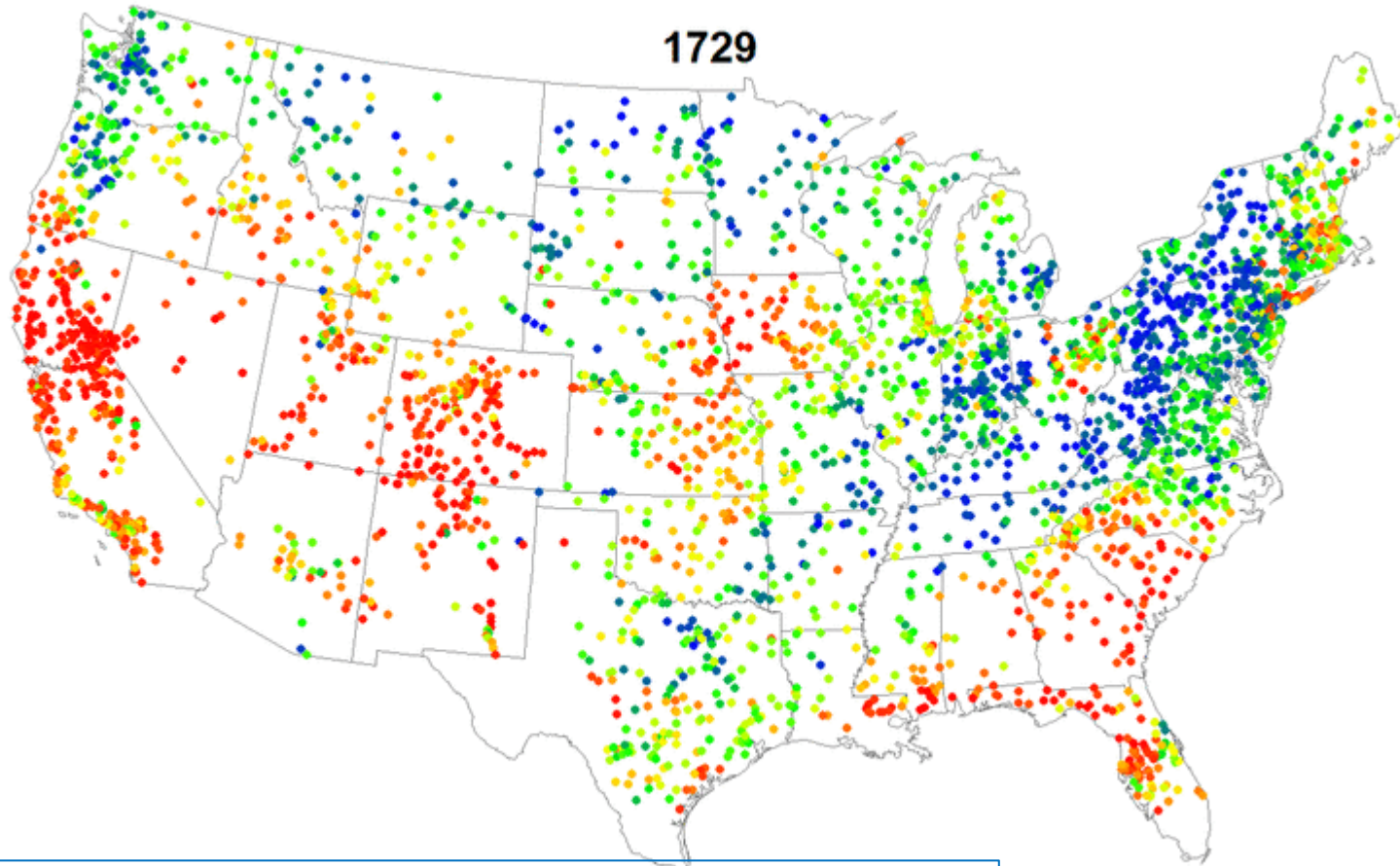
Can we detect the effect of anthropogenic forcing (changing parameters)?

- **Physics based model simulations** based hypothesis testing
 - But the parameters of these models are tuned, and may not reflect changed dynamics
- **Paleoclimate** based hypothesis testing
 - Uncertainties, biases (e.g. variance reduction in reconstructions)
 - Spatial Transferability of sequences?
- **20th century plus observations with anthropogenic forcing as a covariate**
- **Methods:** (often require dimension reduction, another source of error)
 - **Wavelets** (Time, Space-Time) to inform changes in scaling laws?
 - **Bayesian** Framing for **Hierarchical** inference – but at the regional or global scale?
 - Latent Variable Models/Embedding
 - **Predictive ability**, especially for a combination of models and data sources?
 - A common feature is the need to project the data into a lower-dimensional embedding space and then examine the question.

Reconstructing 500 years of streamflow

availability

Annual Flows



Large areas have similar wet/dry conditions persist for many years

How to characterize drought as a spatio-temporal process AND an event process?

Ho, M., et al. (2017), WRR, doi:10.1002/2016WR019632

Formalism – Spatio-Temporal Process

- Drought measures

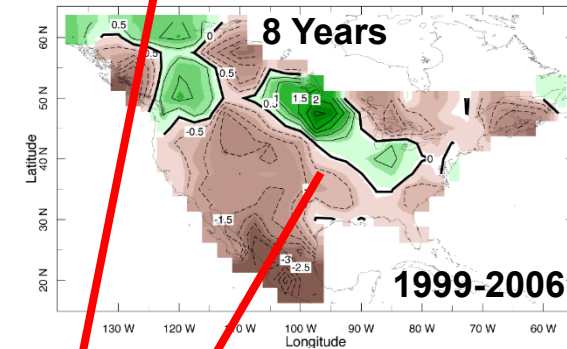
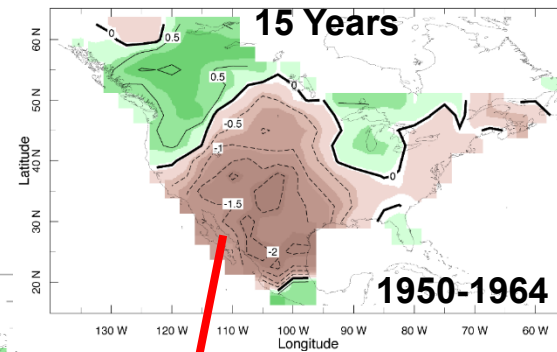
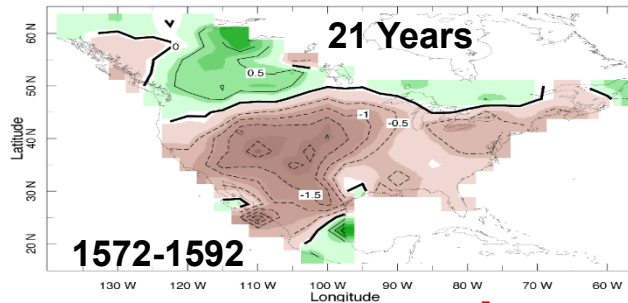
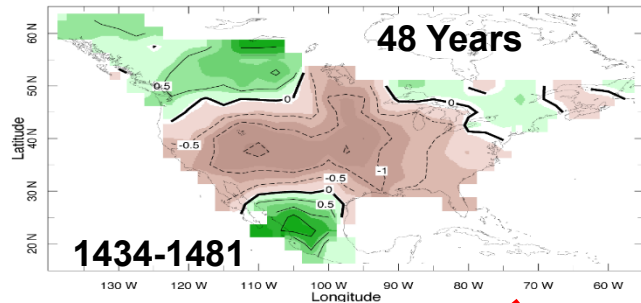
- $P(x,t)$ – **Primary variable** used to define drought: Precip, PDSI, Supply-demand cumulative deficit.
- Drought index $D(x,t) = f(P^*(x), P(x,t))$, where $P^*(x)$ is a location specific **threshold**
 - $D(x,t) = \max(0, P(x^*) - P(x,t))$ or $D(x,t) = \max(0, \sum_t (O(x,t) - I(x,t)))$
- **Drought event** measures:
 - **Contiguous Area and time** over which $D(x,t) > 0$ = Areal extent CA_i and duration Dd_i
 - Drought event **centroid** $EC_i = (\bar{x}, \bar{t})_i$
 - Drought **severity** $S_i = \int \max(D(x, t)_i, 0) dx dt$ or $S_i(x) = \int \max(D(x, t)_i, 0) dt$
- Stochastic Process of Interest at the event scale:
 - Joint/conditional probability distribution of $CA_i, Dd_i, EC_i, S_i = f(De_i)$
 - Conditional on $t, CO_2, ENSO, \dots$
- **Model $P(x,t)$ and derive $f(De_i)$**

Model the primary climate variables as **spatio-temporal** processes

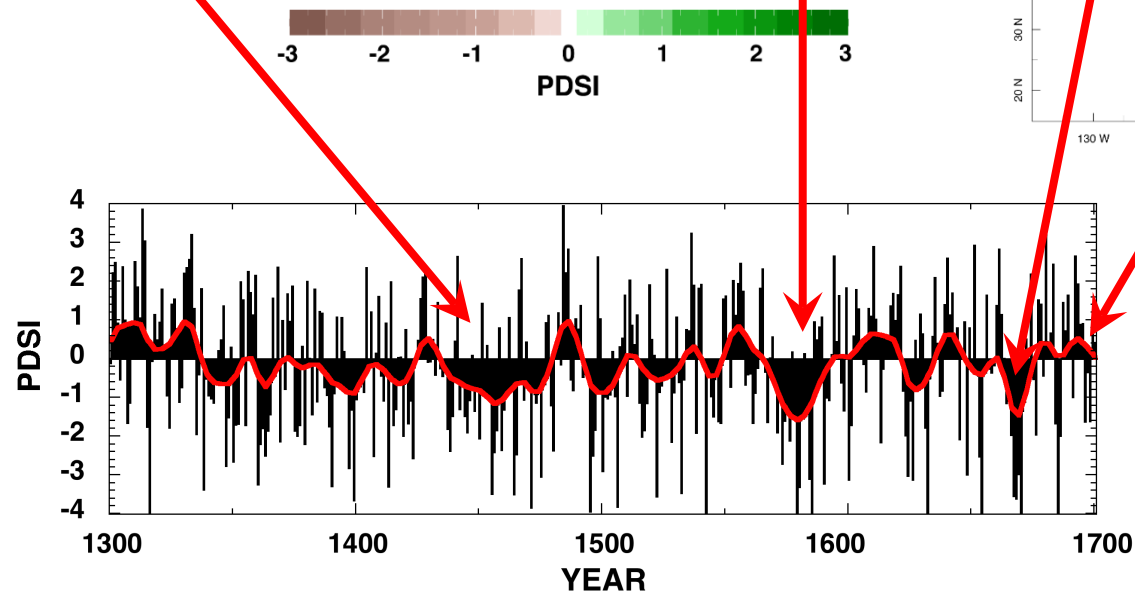
Derive the Area-Duration-Severity-Location-Event **Conditional Probabilities**

Climate Past

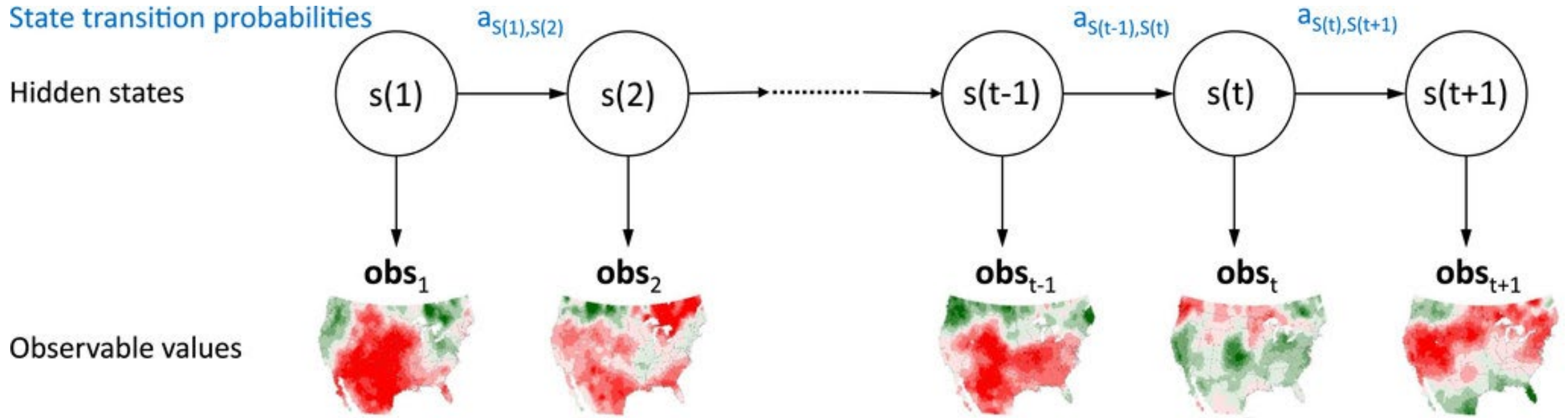
Notable Post-Medieval Megadroughts



The 20th century may have experienced more modest droughts than the prior record!



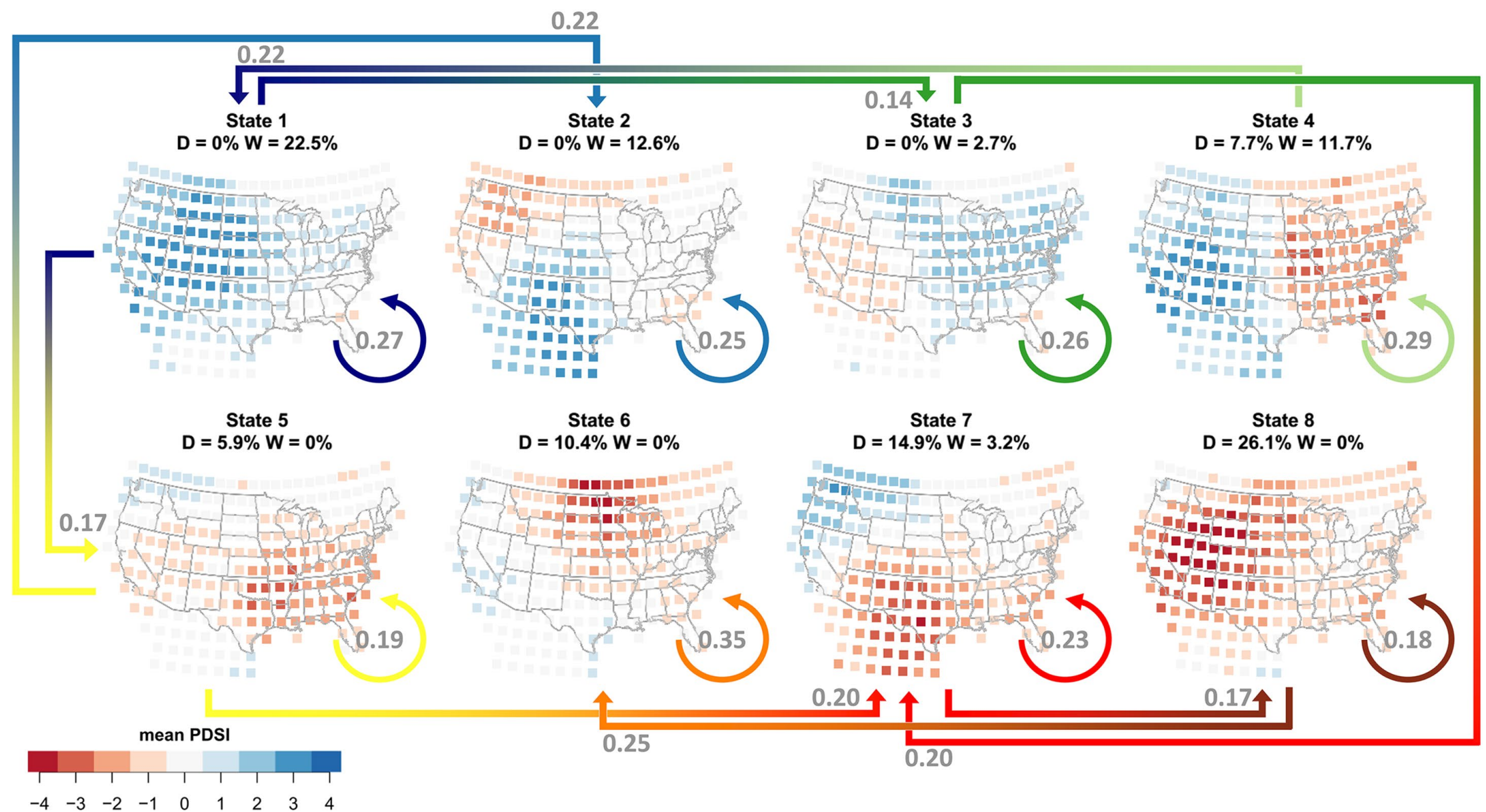
- **Hidden Markov Model – Spatio-temporal process**



Ho, M., U. Lall, and E. R. Cook, 2018: How Wet and Dry Spells Evolve across the Conterminous United States Based on 555 Years of Paleoclimate Data. *J. Climate*, **31**, 6633–6647, <https://doi.org/10.1175/JCLI-D-18-0182.1>.

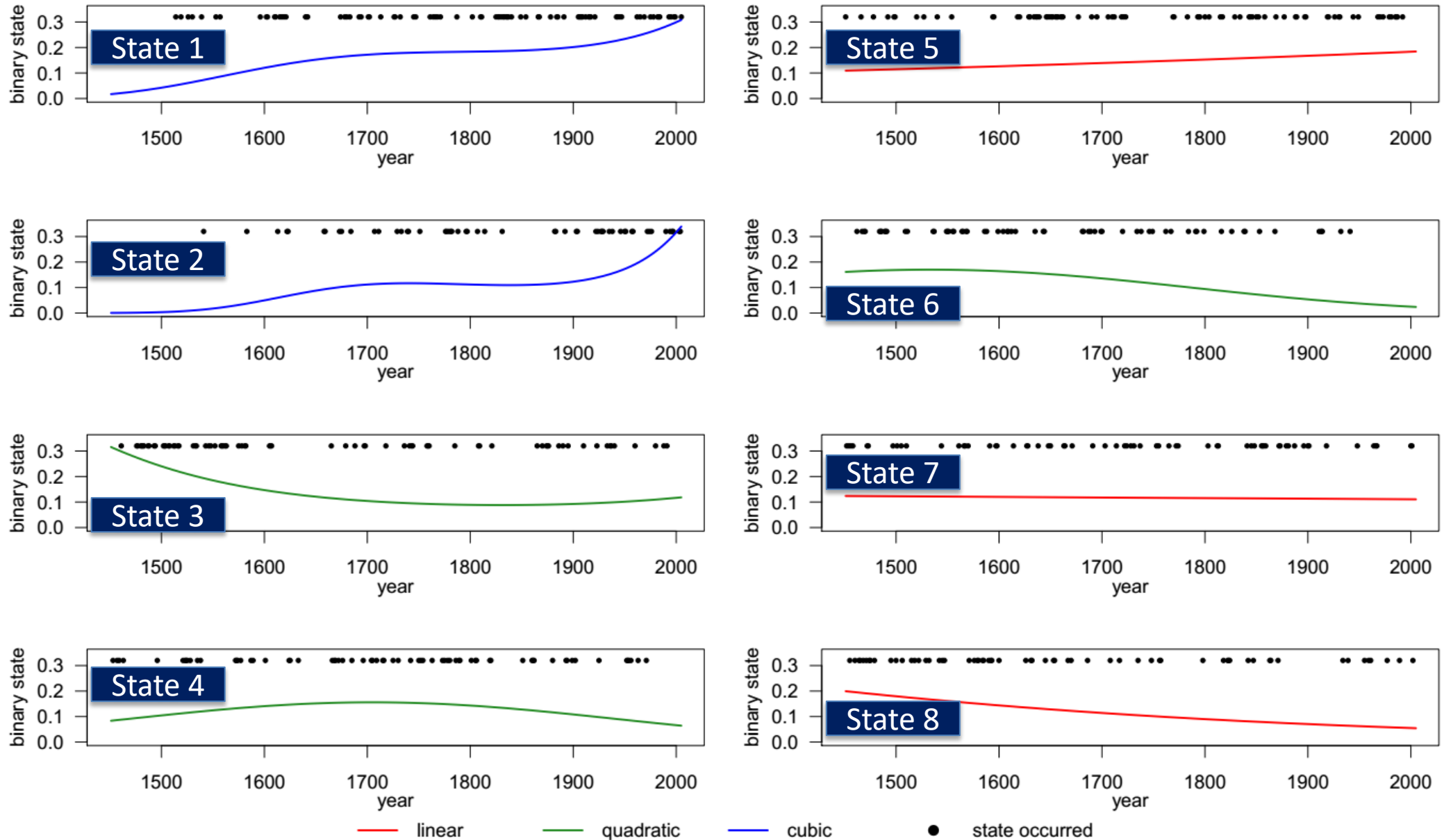
Hidden Markov Model for LBDA Dynamics: States and transition probabilities

a)



Ho, M., U. Lall, and E. R. Cook, 2018: How Wet and Dry Spells Evolve across the Conterminous United States Based on 555 Years of Paleoclimate Data. *J. Climate*, **31**, 6633–6647, <https://doi.org/10.1175/JCLI-D-18-0182.1>.

Trends in the probability of each state



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