

Regional SH Westerly Wave Variability and Cape Town's "Day Zero" Drought

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Background & Introduction

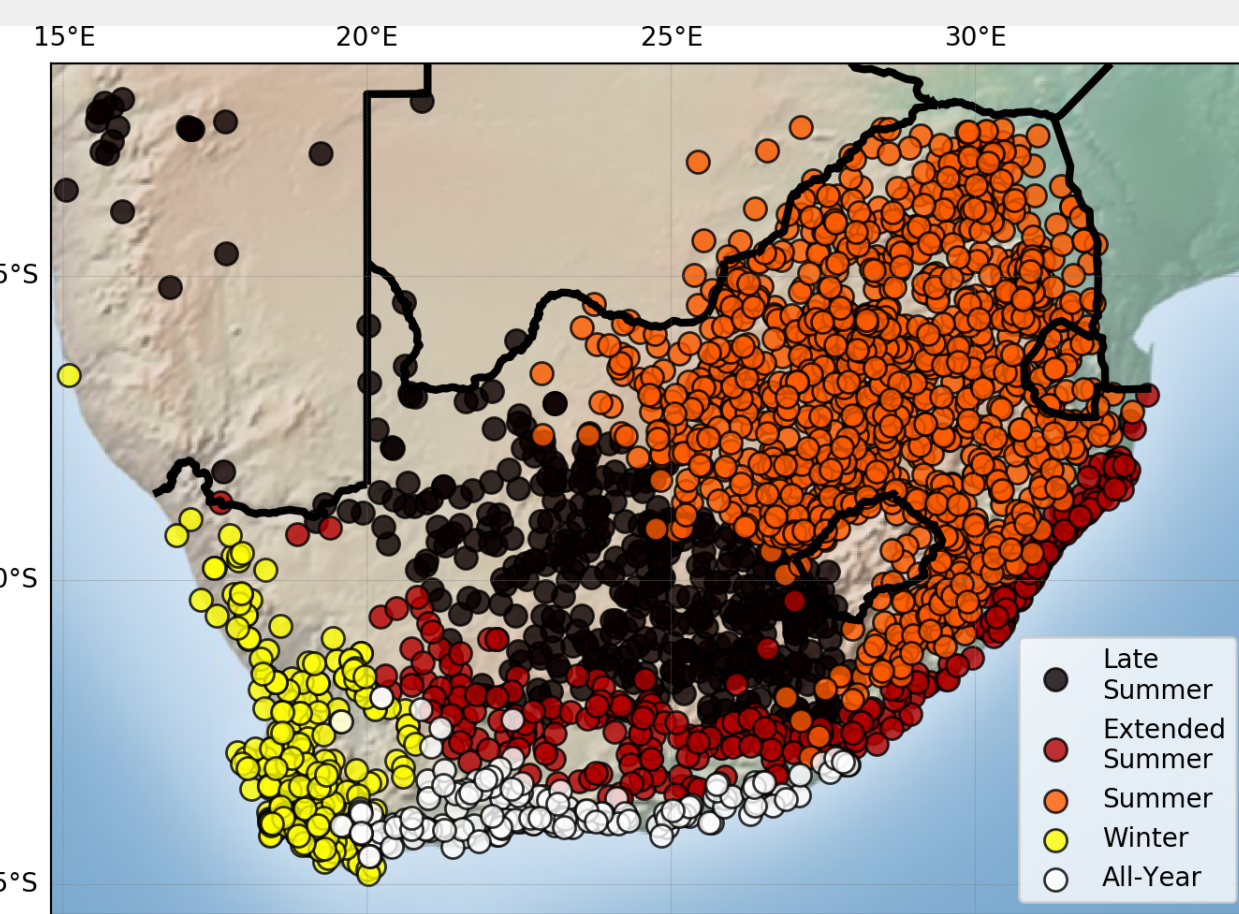


Figure 1: Ward's clustering by seasonality.

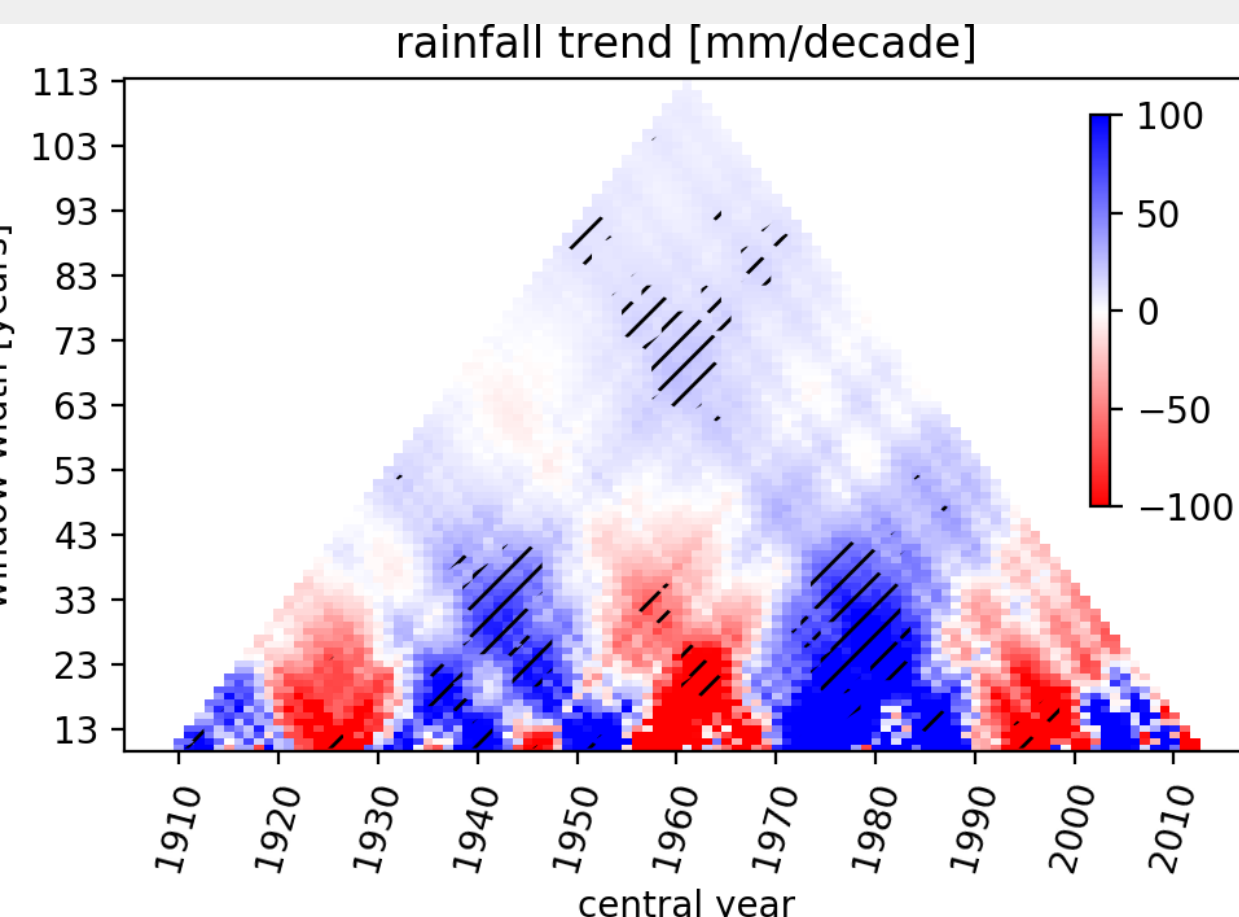


Figure 2: Theil-Sen trends near major dams.

Data & Methods

- **Self-Organising Map (SOM)**
Analysis of ERA-Interim (Dee et al., 2011) 500-hPa geopotential
- SOM: artificial neural network (ANN) commonly used for non-linear circulation clustering (Hewitson and Crane, 2002)
- SE-Atlantic domain: identify 1 trough/ridge affecting WRZ at a time $-55 \leq \phi \leq -25^\circ$; $0 \leq \theta \leq 27^\circ$
- 6 hourly, AMJJAS 1979-2017
- **Cut-Off Lows (COLs)**
► Closed, cold-cored lows at 500hPa
► COL days considered separately
- **SOM node & stations clustering**
► Ward's linkage; Euclidian distance
► SOM nodes: divergence & geostrophic vorticity near WRZ
► Stations: contribution of SOM node clusters to AMJJAS rainfall

- **Observed station rainfall:**
► > **98%** coverage 1979-2017
► Cleaned, date corrected, gap-filled

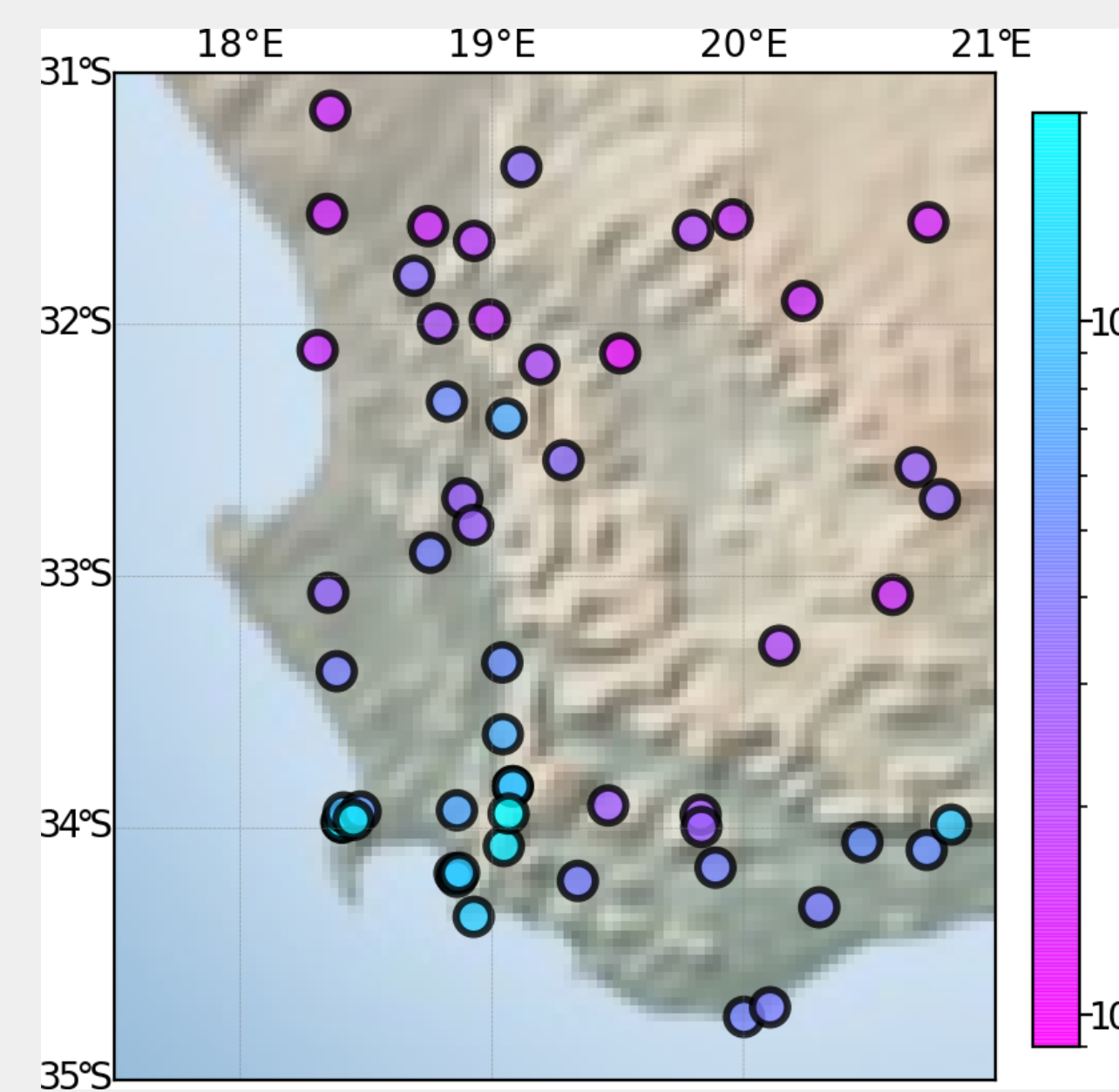


Figure 3: Mean annual station rainfall (mm)

Acknowledgments

- Phillip Mukwenha for assistance with software, hardware and data
- Chris Jack & Pierre-Louis Kloppers for data formatting & cleaning
- Birgit Erni for assistance with gap-filling techniques
- South African Weather Service (SAWS) for providing long-term station records for many locations; City of Cape Town & SA Dept. of Water and Sanitation (DWS), for additional station data
- Financial assistance towards the PhD research project from the SA National Research Foundation (NRF) and CSAG
- CSAG and UCT PGFO for sponsoring attendance at this conference

Results: SOM & Station Rainfall

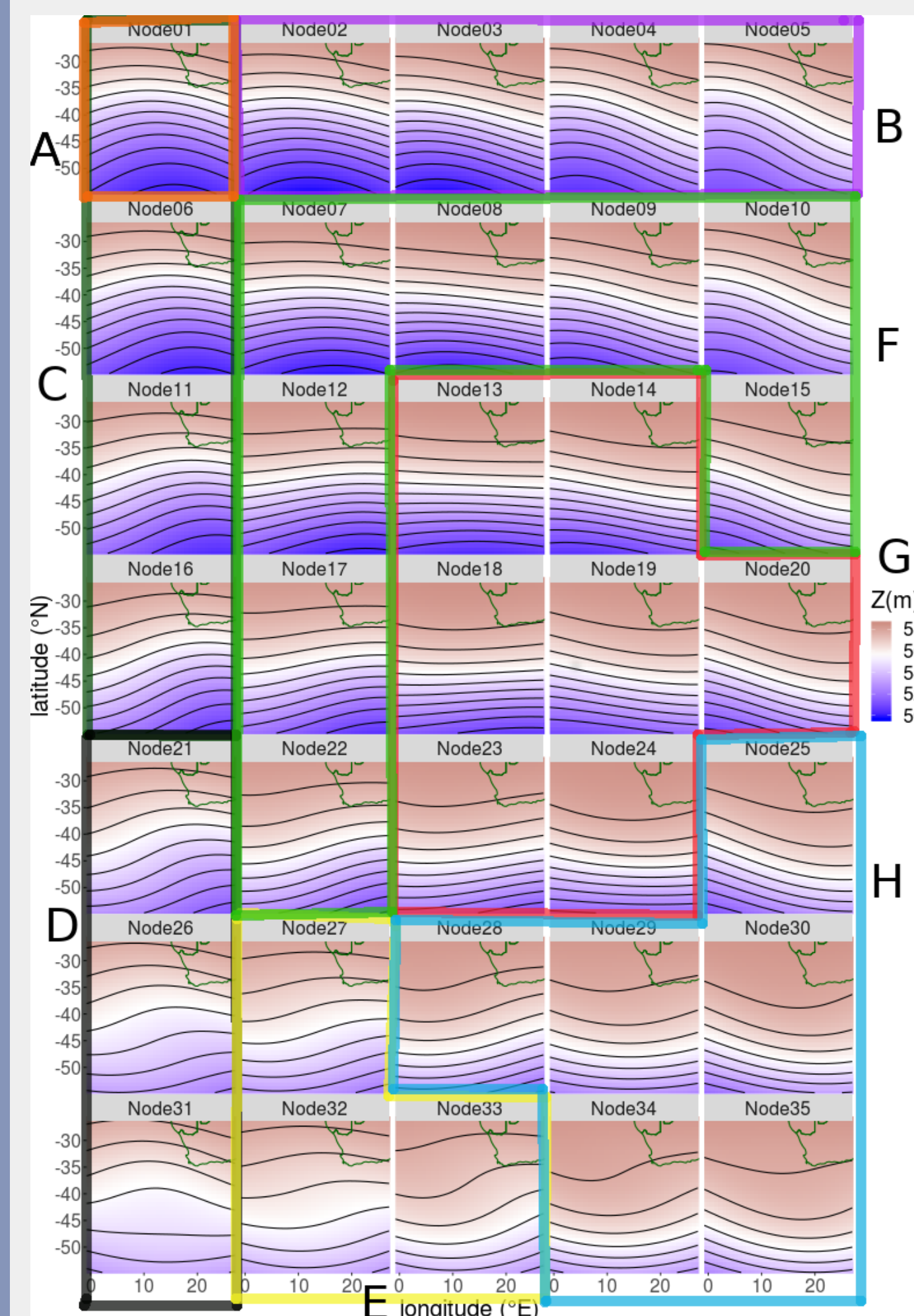


Figure 4: 5 x 7 Z500 SOM, with clusters

Node Cluster	Freq (%)	Rainfall (% range)
A: Strong trough	3.3	6.2(SE)–18.0(W)
B: NW trough	13.5	6.5(SE)–17.8(CP)
C: S steep trough	8.0	15.6(N)–22.5(SW)
D: Curved trough	8.2	14.2(SE)–24.5(NW)
E: Weak trough	6.8	4.3(N)–12.2(SE)
F: Zonal flow	20.2	5.9(NW)–10.8(CP)
G: Ridge	16.9	1.6(W)–4.7(SE)
H: Strong ridge	20.2	2.3(W)–10.7(SE)
COLs	3	7.3(SW)–18.1(SE)

Table 1: SOM node cluster frequency & contribution to AMJJAS rainfall

- **Troughs** (40% frequency) responsible for \approx **80%** of all AMJJAS rainfall in core WRZ; strongest half bring \approx **60%** of rain in mnts and NW.
- Contribution from **troughs** is highest in the **W** cluster; from **COLs** is highest in the **E & SE**; N & NW are most dependent on troughs with large curvature

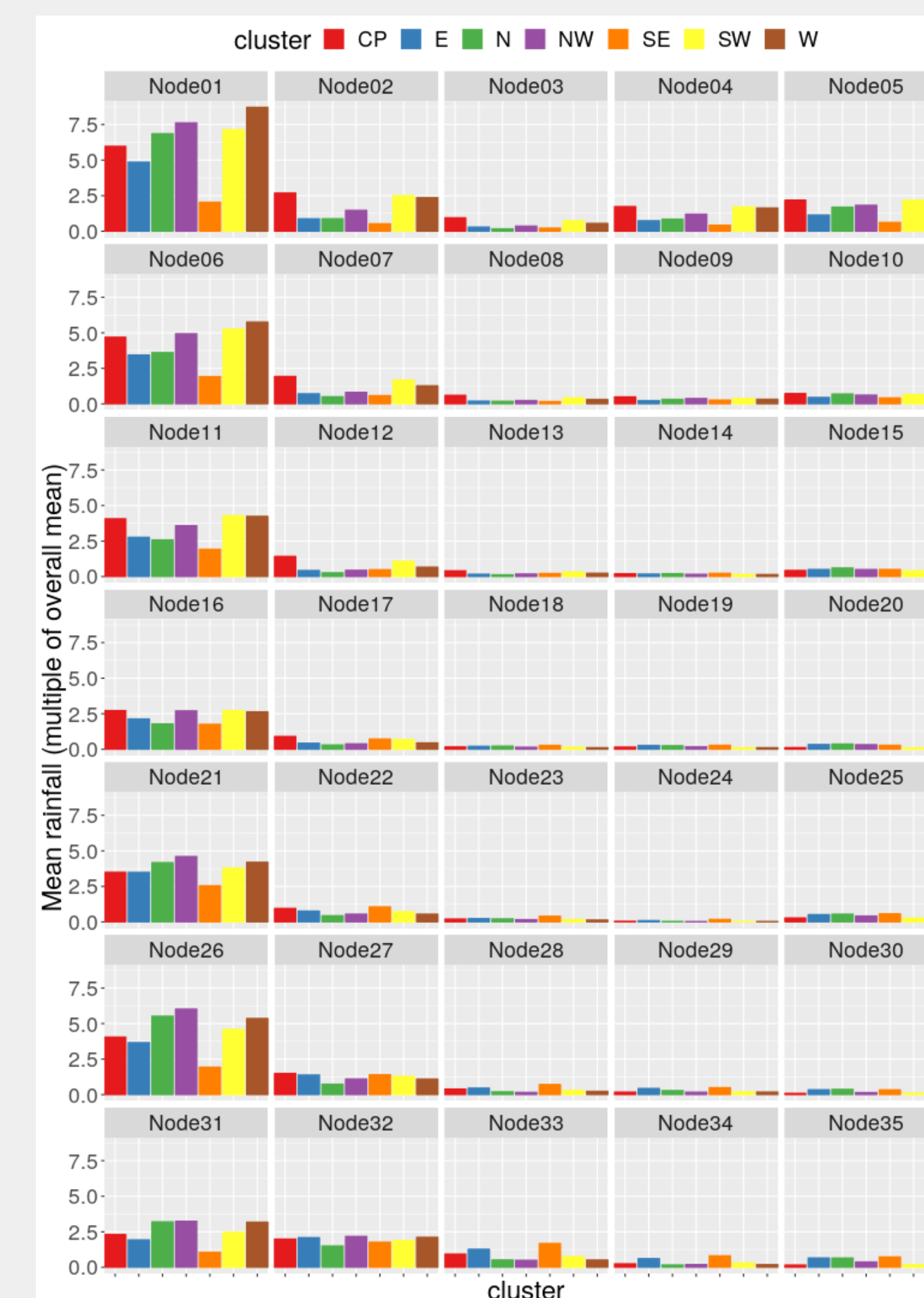


Figure 5: Normalised SOM node mean rainfall

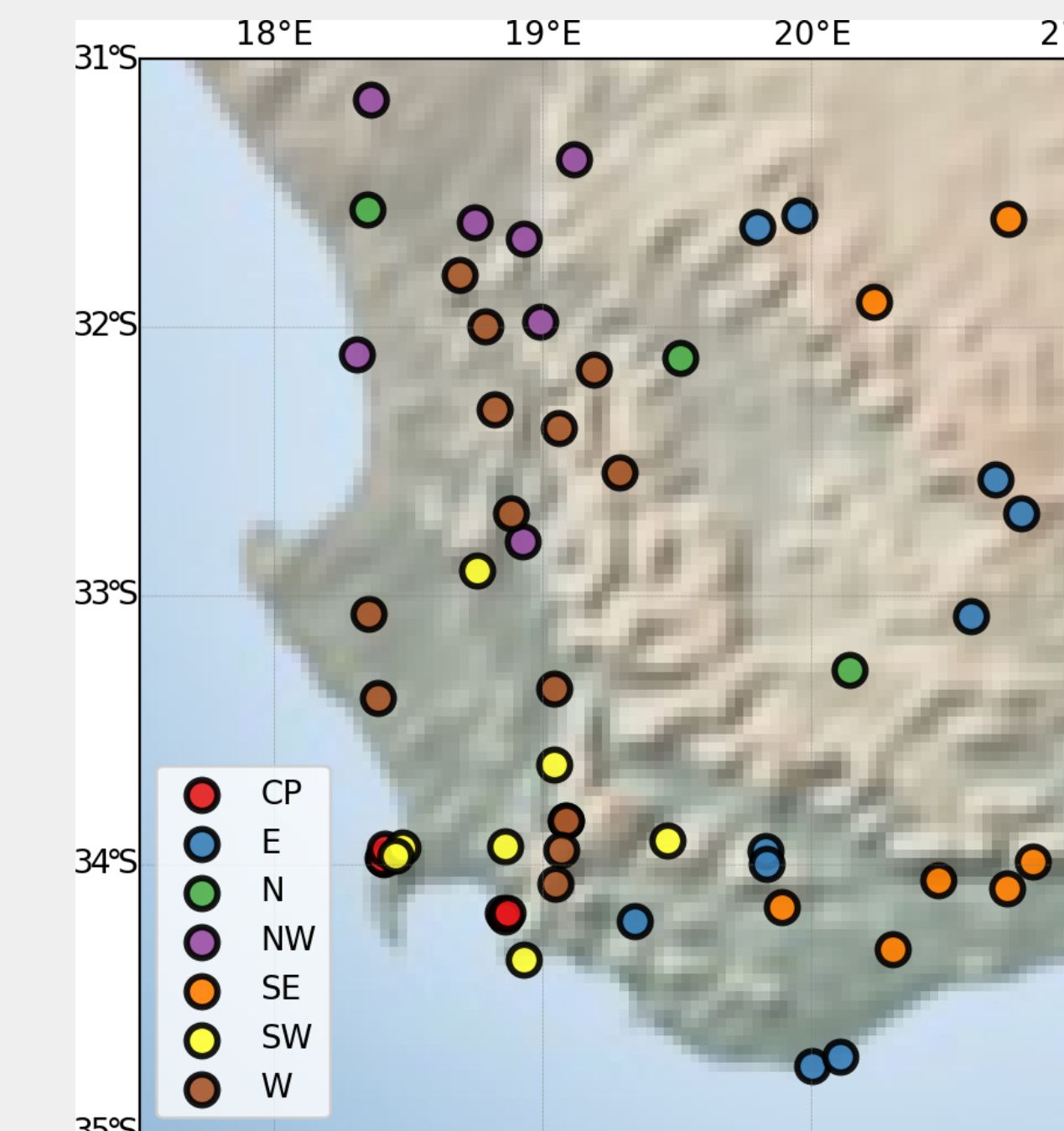


Figure 6: Station clusters obtained

Conclusion: Circulation Contribution to Drought

- Most rainfall in extended winter associated **strong troughs** in mid-tropospheric westerlies
- Circulation type frequency variation **explains 2–36%** of rainfall variability across WRZ
- Explanatory power greater in recent years (\lesssim **50%**)
- \approx **55%** of **Day Zero Drought** shortfall explained by trough/ridge variability & trend.
- Unprecedented autumn (AM) frequency of most **intense ridge** node during severe drought year 2017, consistent with increasing trend ($p < 0.01$) over 39 yrs

Results: Circulation & Rainfall Prediction

- Annual SOM node cluster frequency anomalies with mean rainfall patterns for each are used to predict station cluster annual AMJJAS rainfall.
- Each year predicted using all others in turn.
- 2 **strongest ridge** nodes (34, 35) exhibit **blocking behaviour & unprecedented frequency** during drought; hence, they are separated in frequency time series

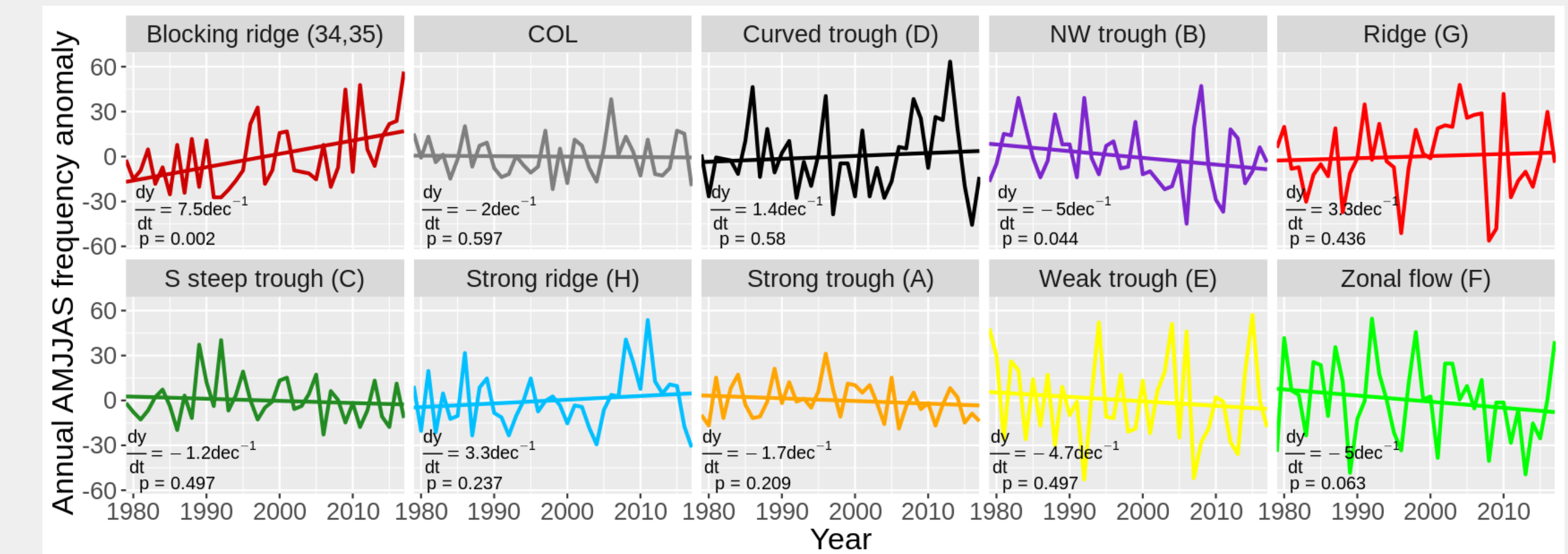


Figure 7: AMJJAS circulation type anomaly 1979-2017 with decadal trend and significance

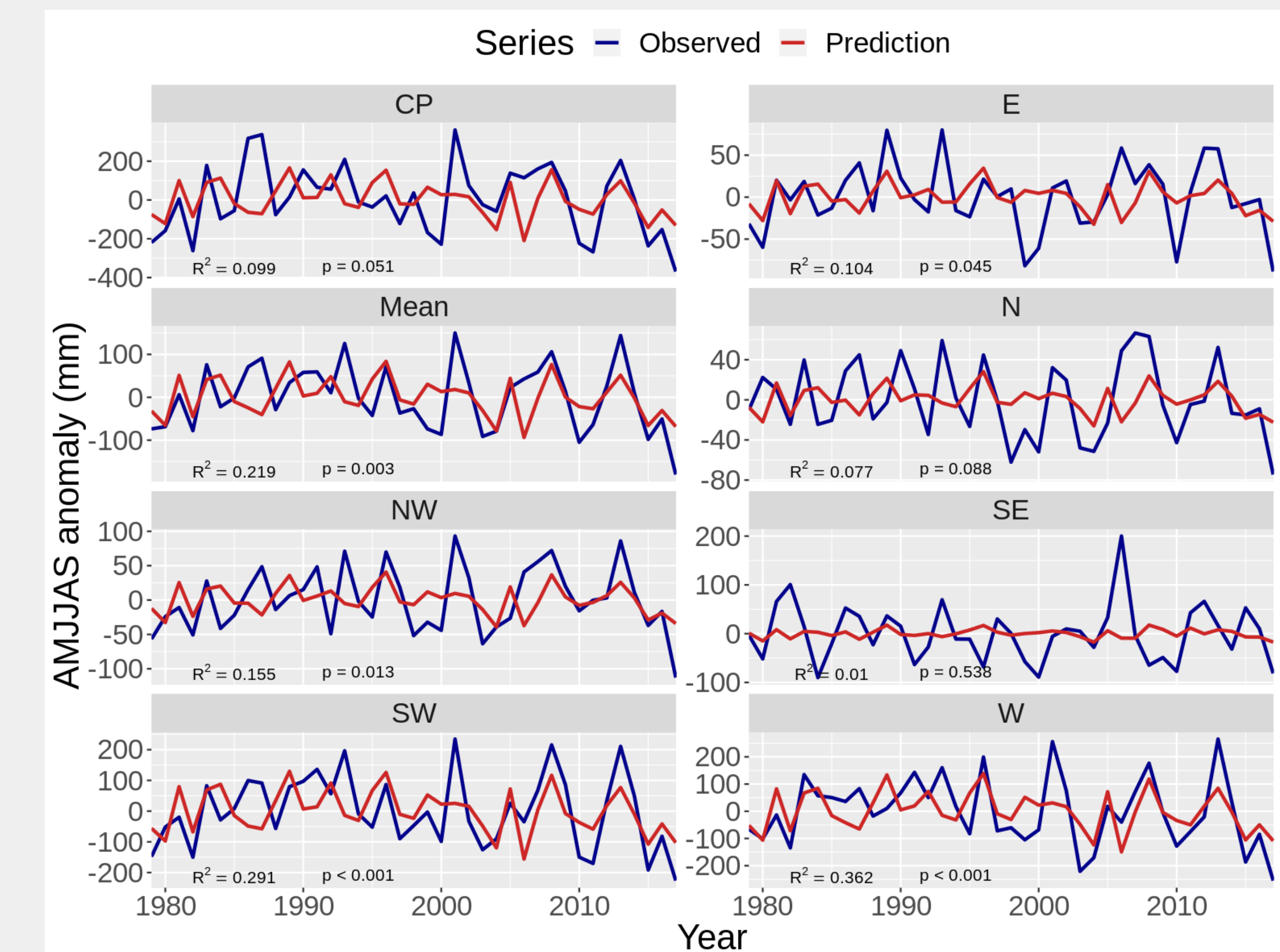


Figure 8: Observed and predicted station cluster mean anomaly 1979-2017

- Prediction **best** in the **W & SW mnts**; model explains essentially none of the variability in E & SE

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