

Seasonal predictability of daily rainfall for catchment hydrology in NW South America

Using an NHMM model

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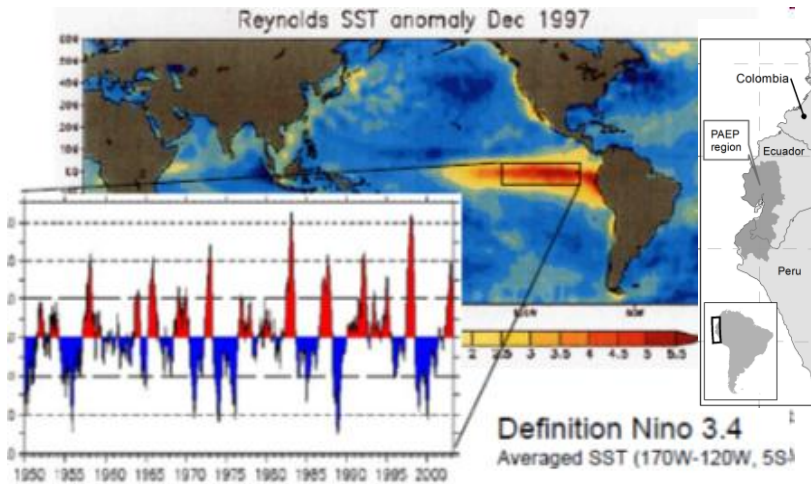
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Hydraulic Section

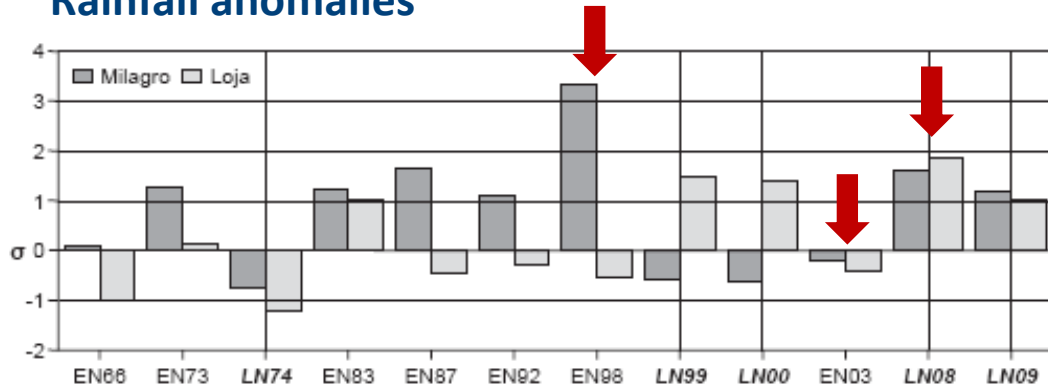
HEPEX Seasonal Hydrological Forecasting Workshop
22nd September 2015, Norrköping, Sweden



Motivation



Rainfall anomalies



Bendix et al. 2011

Water Management related problems



EN 97-98

PERU-Piura



EN 02-03 "Litigation because the absence of EN. Farmers will sue those who predicted rains"



Photo 1: Flooding close to Guayaquil (province Guayas) on 10 March 2008 (Photo: J. Bendix)

LN 07

ECUADOR-Guayas

The Pacific-Andean Basin in Ecuador and Peru

- 21 hydrological units c.a. 100,800 km²
- ~70 % Ecuadorian crop production

User needs of S2S forecast



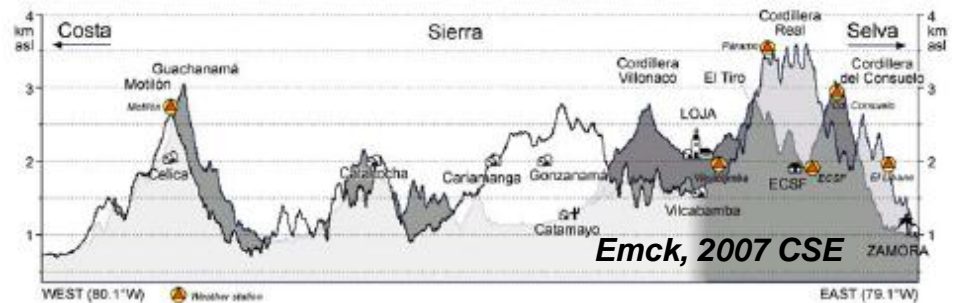
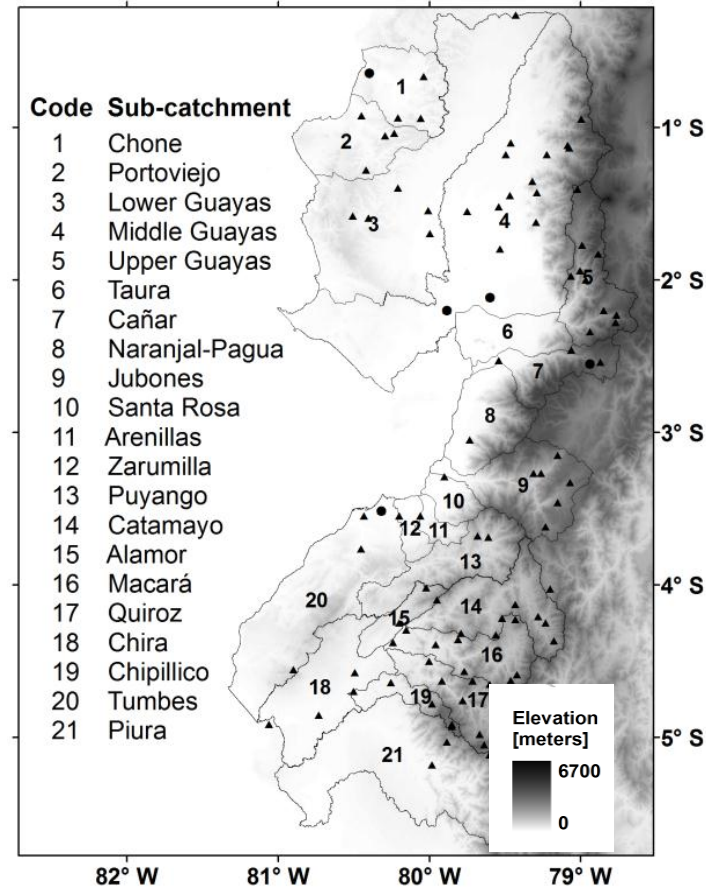
Hydro-meteo hazards

- Slope stability
- Landslide models



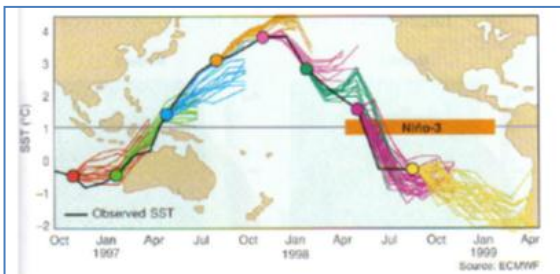
Water resources

- Inflow for reservoirs
- Flood / drought prediction



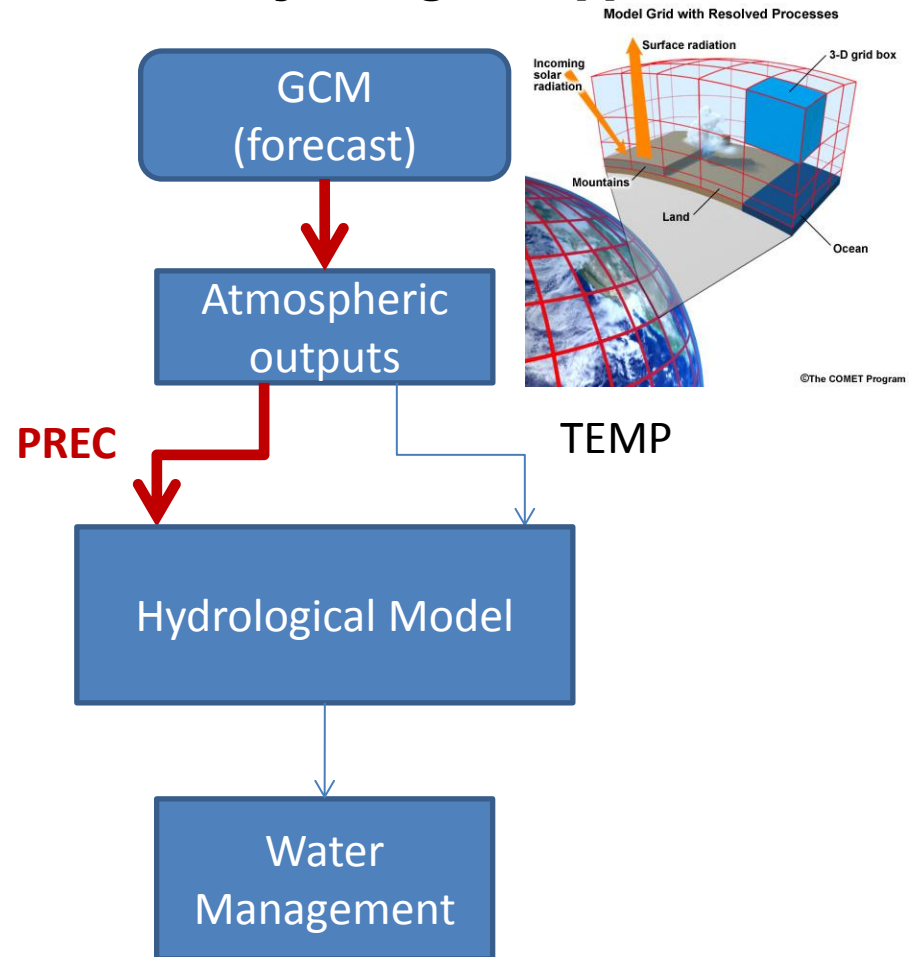
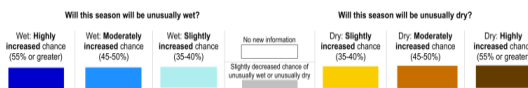
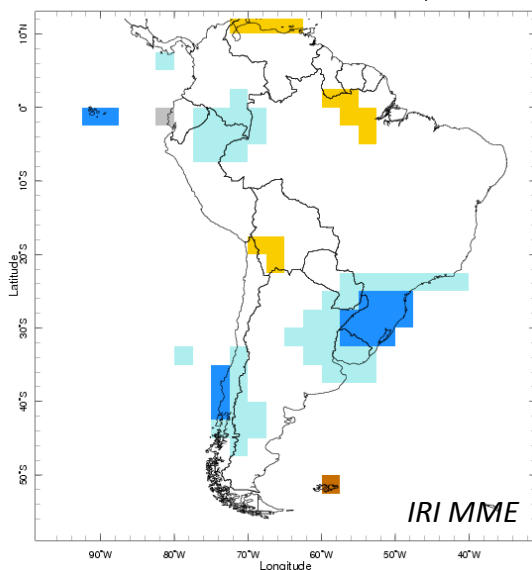
Objectives

- Provide theoretical basis for seasonal climate prediction
- Tailor GCM seasonal climate forecasts for hydrological applications



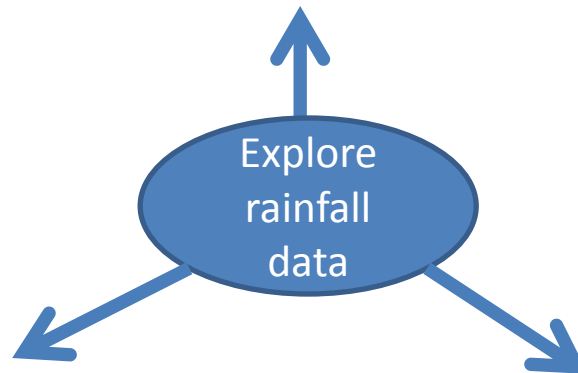
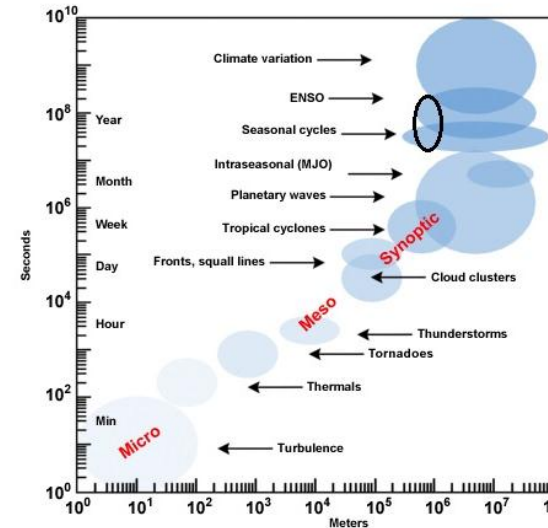
The success story: Dynamical El Niño forecast for the winter 97/98

Forecast for Oct-Dec 2012, Forecast Issued Sep 2012

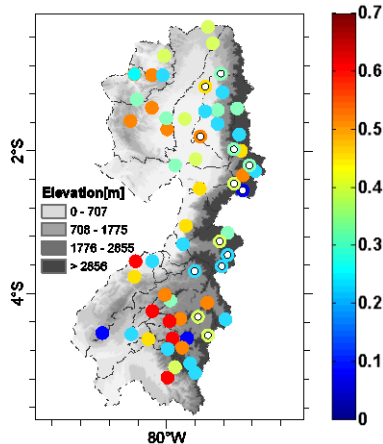


Work flow

WP1 Weather type
sub-seasonal diagnosis



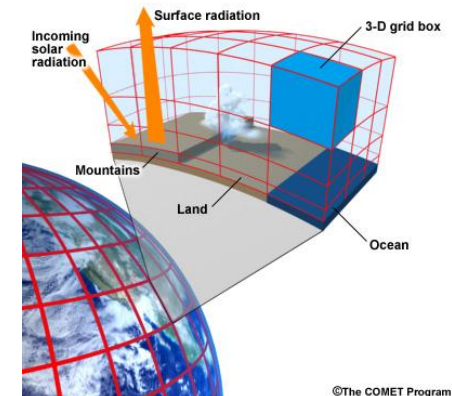
WP3 Dec-May daily rainfall
conditioned on GCM forecasts



WP2 Process-based
downscaling
CGCM vs. AGCM
GCM predictors

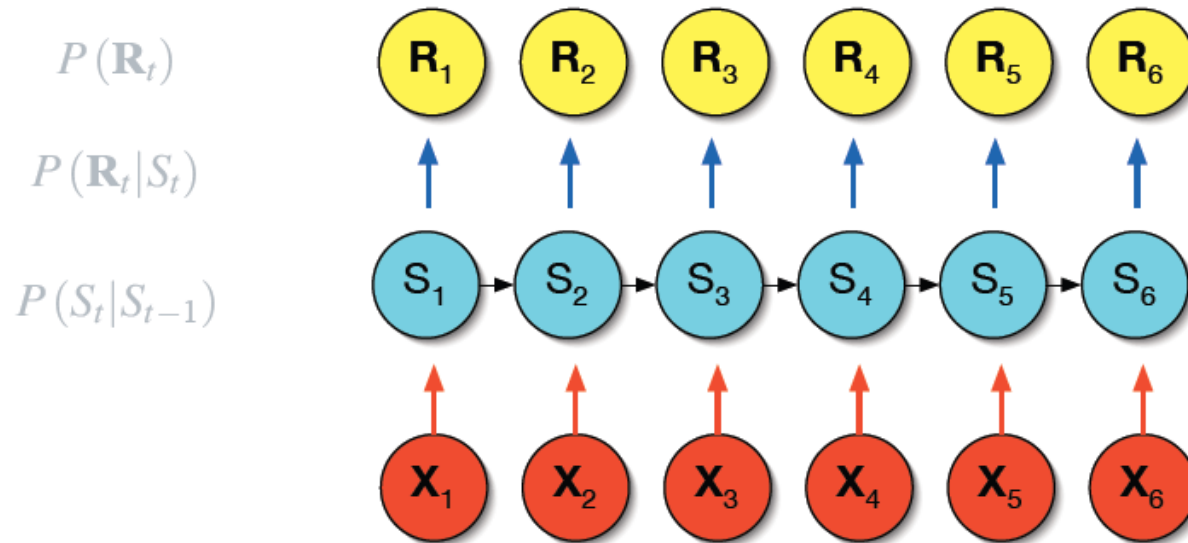
HMM

Model Grid with Resolved Processes



NHMM

Why NHMM for downscaling?



$$P(S_t | S_{1:t-1}, \mathbf{X}_{1:T}) = P(S_t | S_{t-1}, \mathbf{X}_t)$$

Robertson et al. 2004

- Describes spatio-temporal dependencies in **multi-site / catchment rainfall data**
- The hidden weather states allow “**upscaling**” of rainfall processes
- Input variable $\mathbf{X}_{1:T}$ can play the role of **predictor** on the spatial scale of **the weather state**

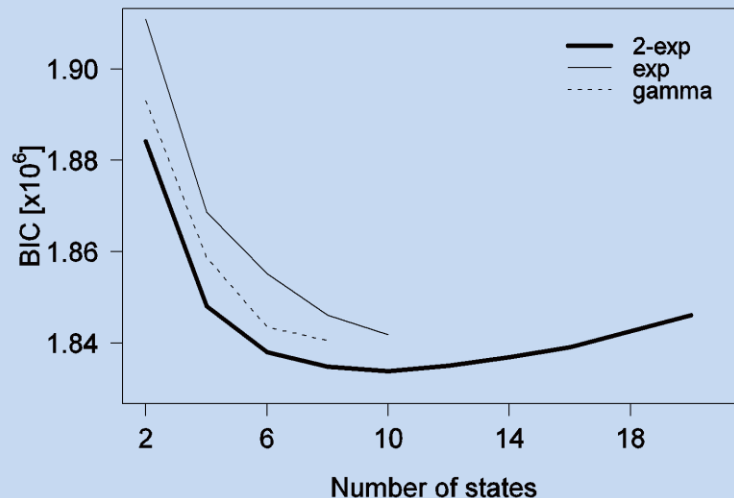
Robertson, A. W., S. Kirshner, and P. J. Smyth, 2004: Downscaling of daily rainfall occurrence over northeast Brazil using a hidden Markov model. *J. Climate*, 17, 4407-4424.

1. Dec-May weather characterization

HMM assumptions

- ✓ Number states
- ✓ Emission distributions for daily intensities and occurrences (pdf's)

→ ML parameter estimation EM¹ algorithm



Weather diagnosis

- ✓ States evolution sequence (Viterbi algorithm ²)
- ✓ Clustering of days falling into each state
- ✓ Mapping states as atmospheric composites:

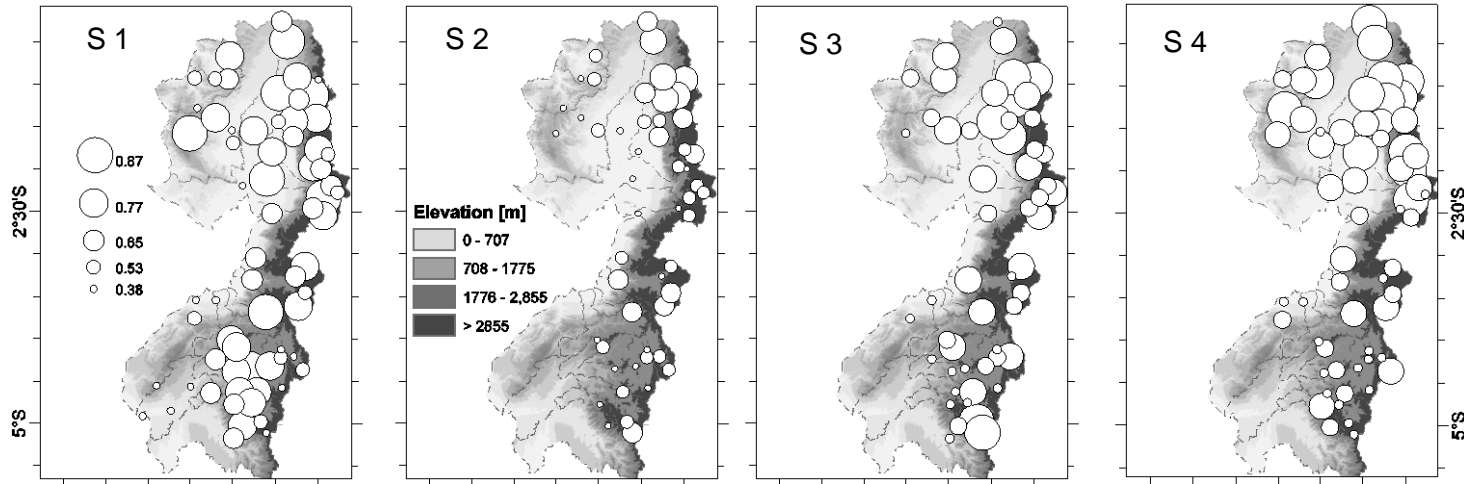
uv850, q 850 anomalies from the Dec-May climatology

¹ Dempster et al. 1977; Ghanhramani, 2001

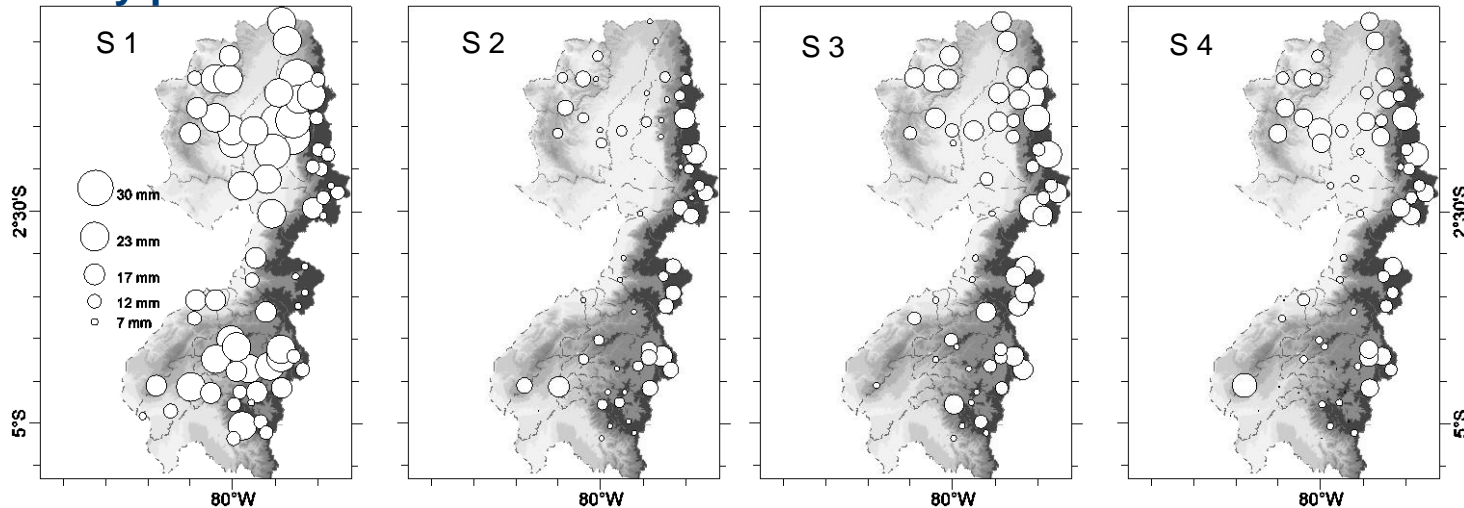
² Forney Jr., 1978

Dec-May weather characterization

Mean intensities



Wet-day probabilities



S1: Wet (+)

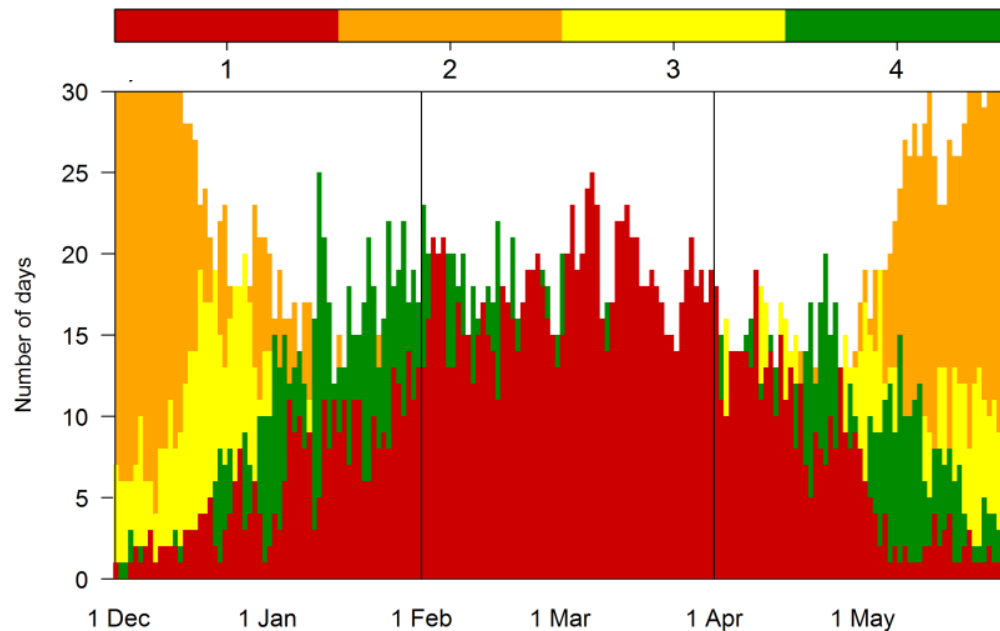
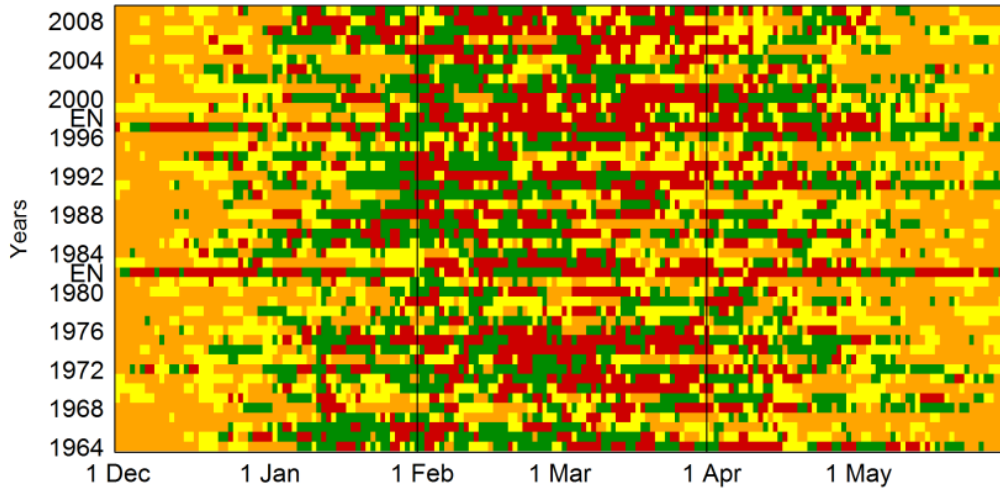
S2: Dry (+)

S3: Dry-wet (-)

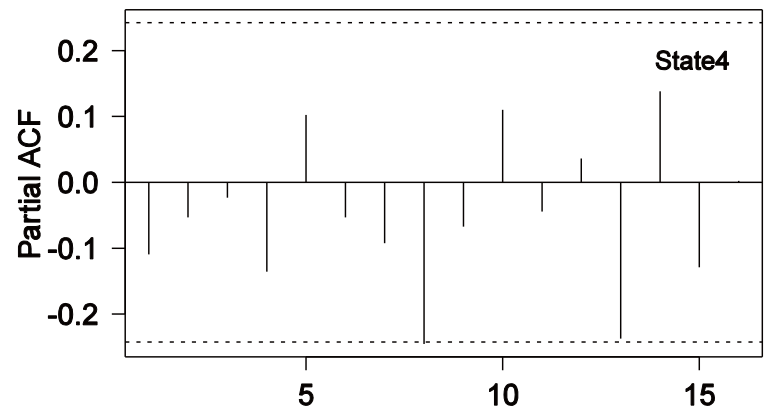
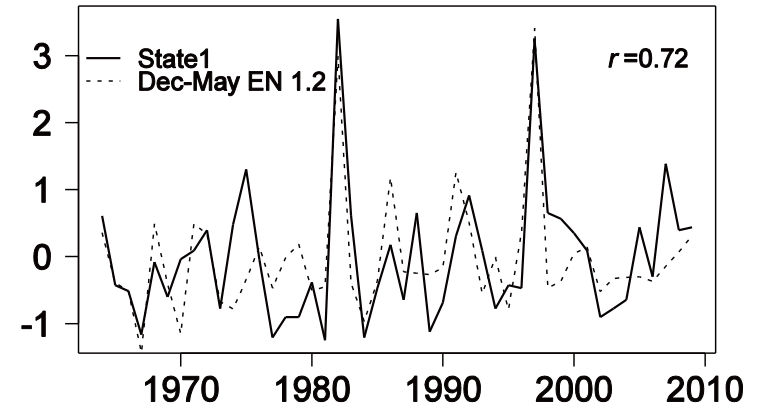
S4: Wet (++)

Dec-May weather characterization

Intra-season variability

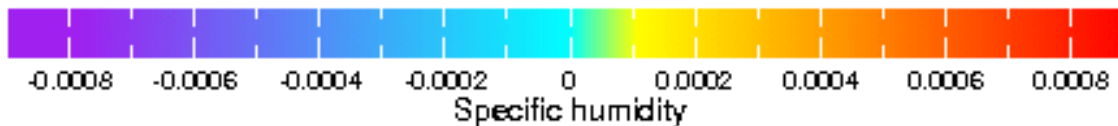
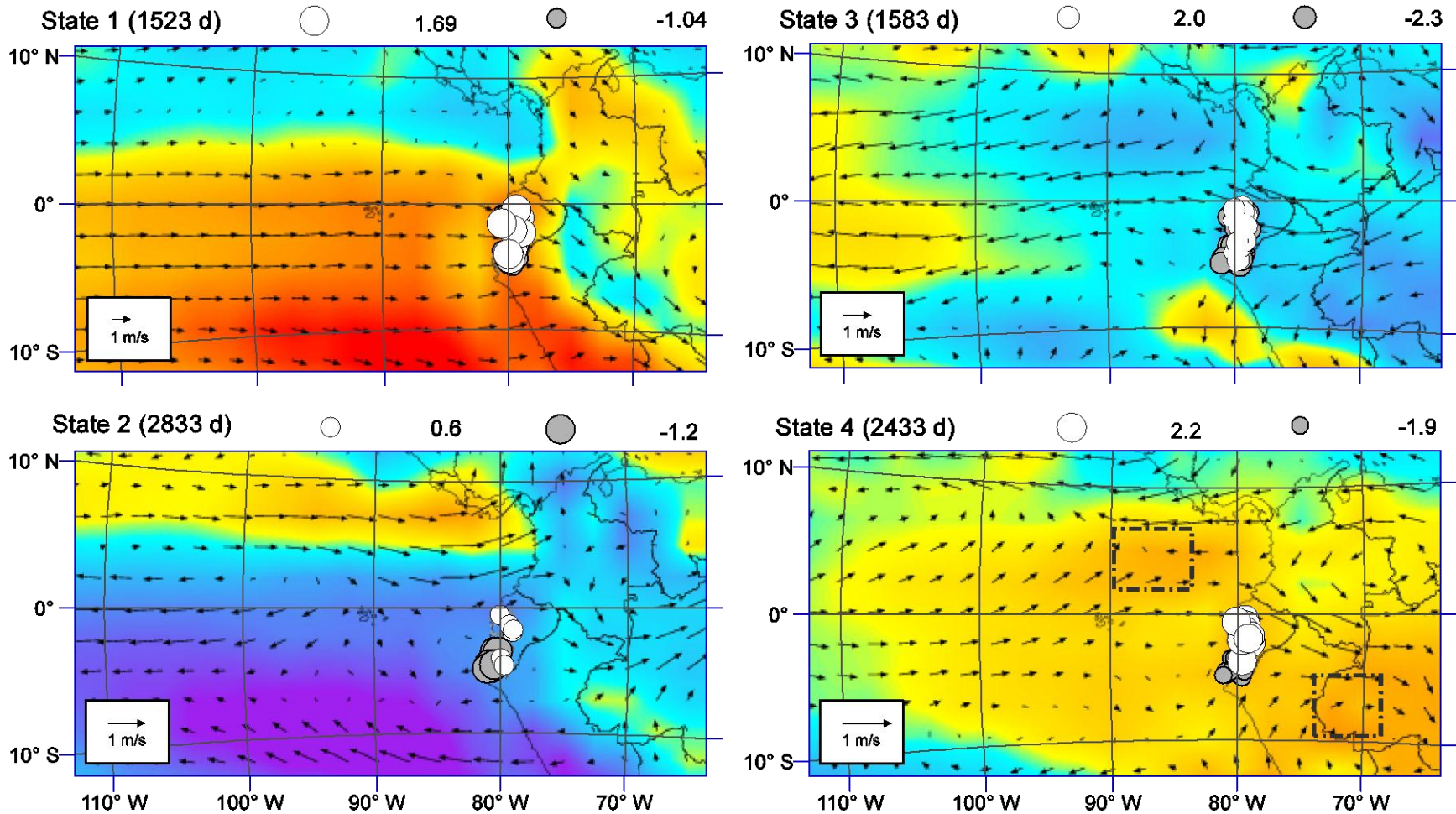


Interannual variability



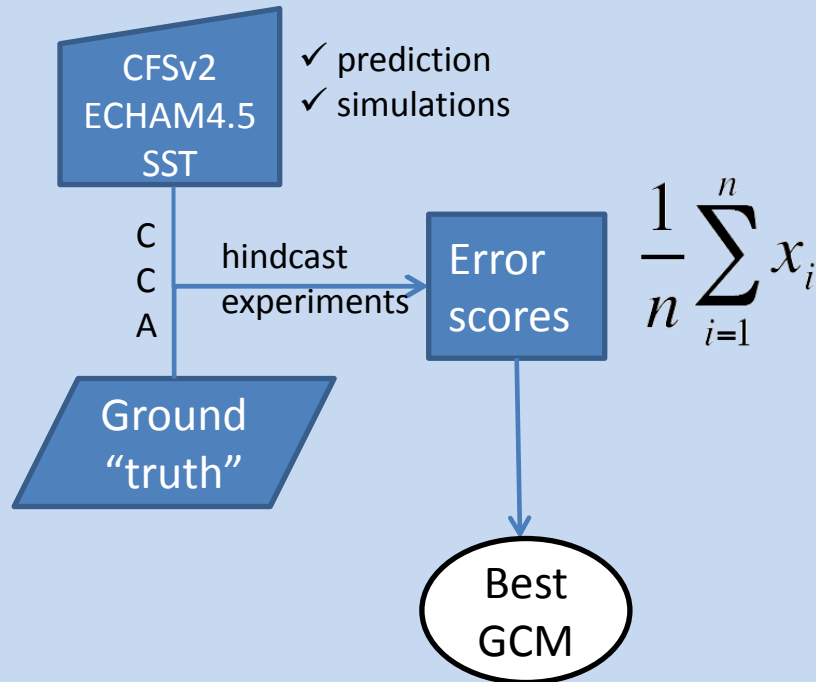
Dec-May weather characterization

Weather state - synoptic identity

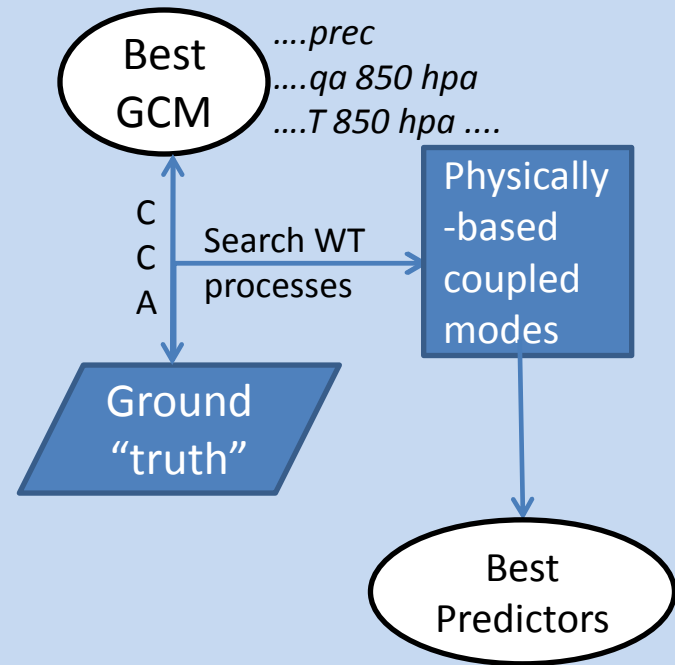


2. GCM evaluation & Predictors identification

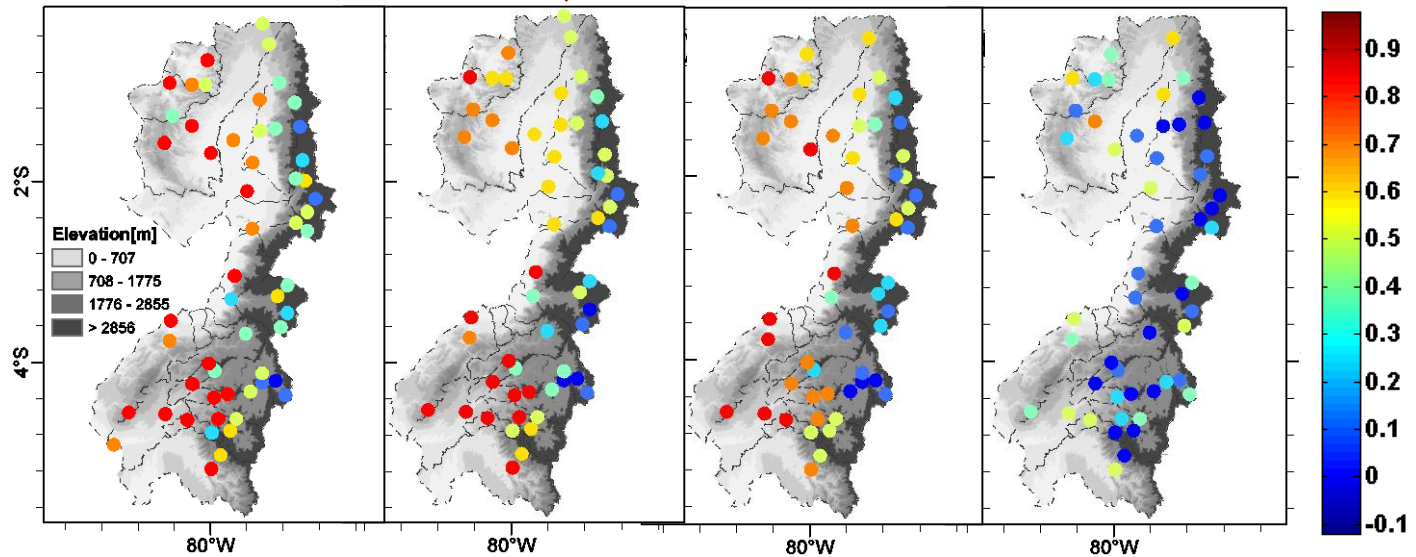
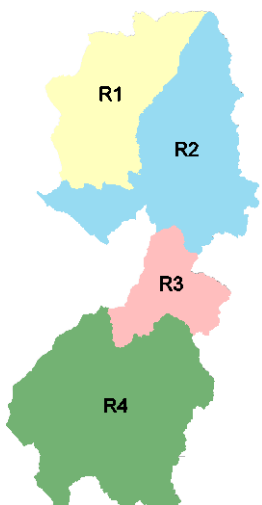
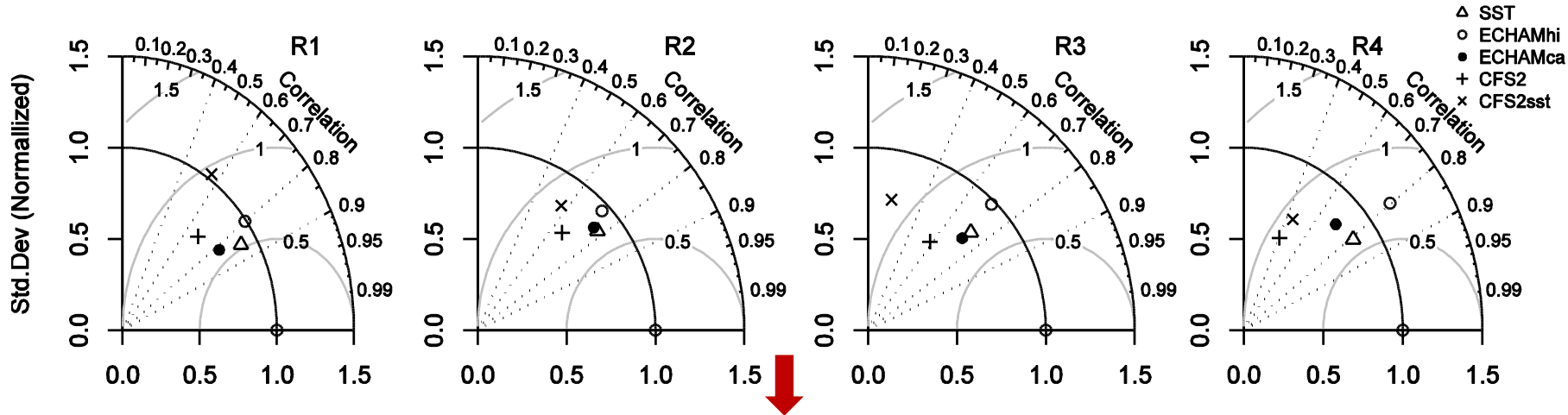
Seasonal GCM forecast evaluation



Processes Upscaling Predictor calibration



Seasonal GCM forecast evaluation

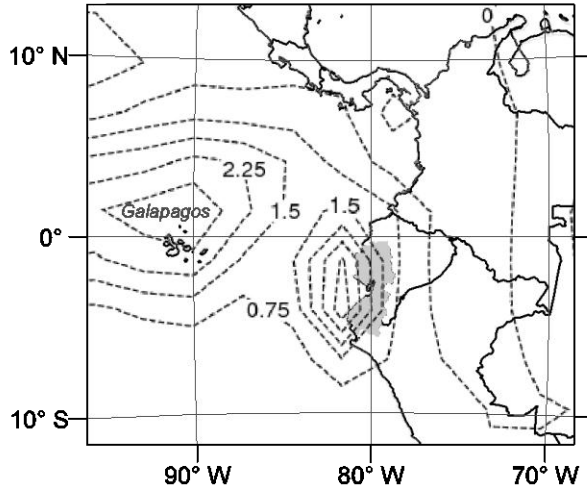


- CFSv2 skill (-) underestimation “double ITCZ syndrome”.
- ECHAM4.5ca skill(+) fails to emulate thermal rainfall as 1st mode → $qa(850hPa)$
- ECHAM4.5his skill (++) Separates thermal (signal) and dynamic (noise) → p highly skilful

Processes Upscaling - Predictors identification

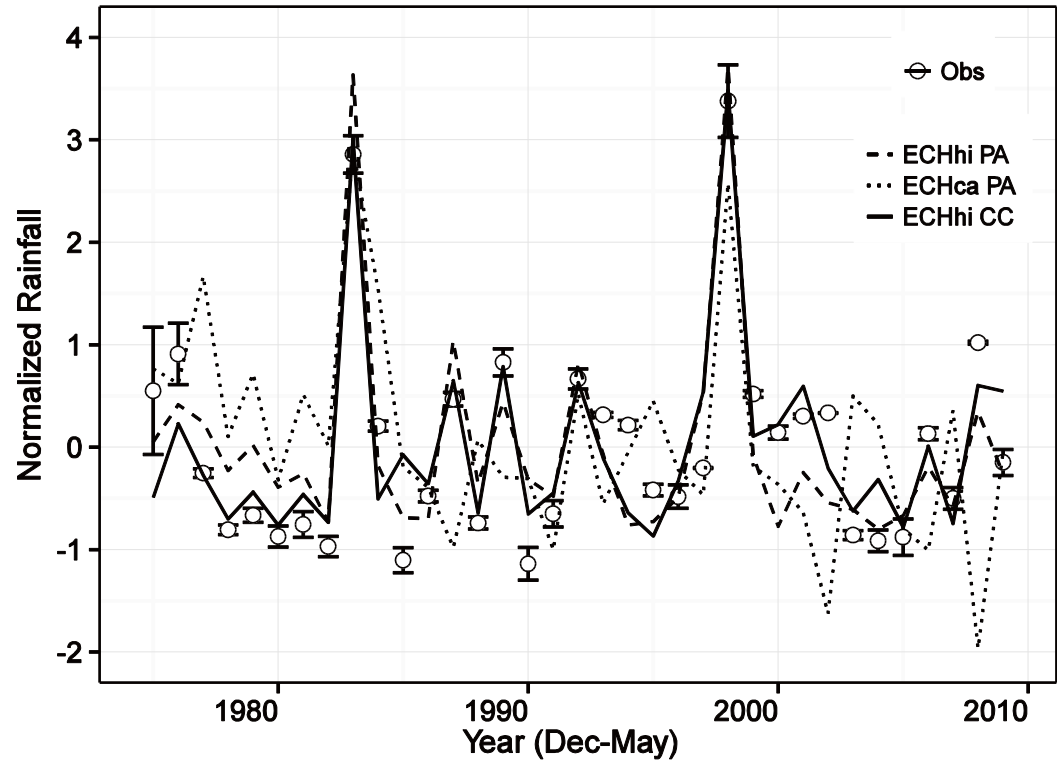
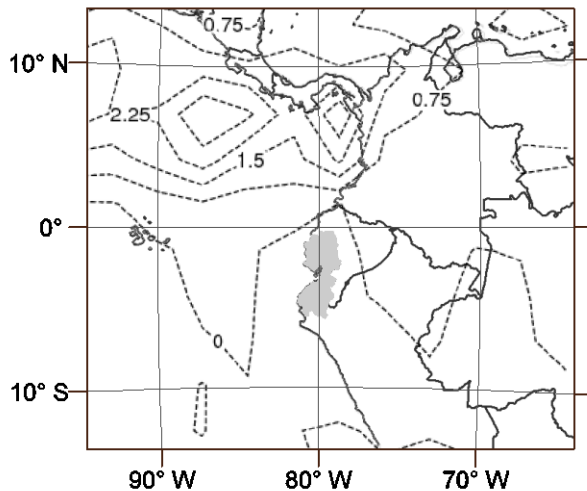
S1: Wet (+)

ECHAM4.5hi PC1 (35.31%)



S4: Wet (++)

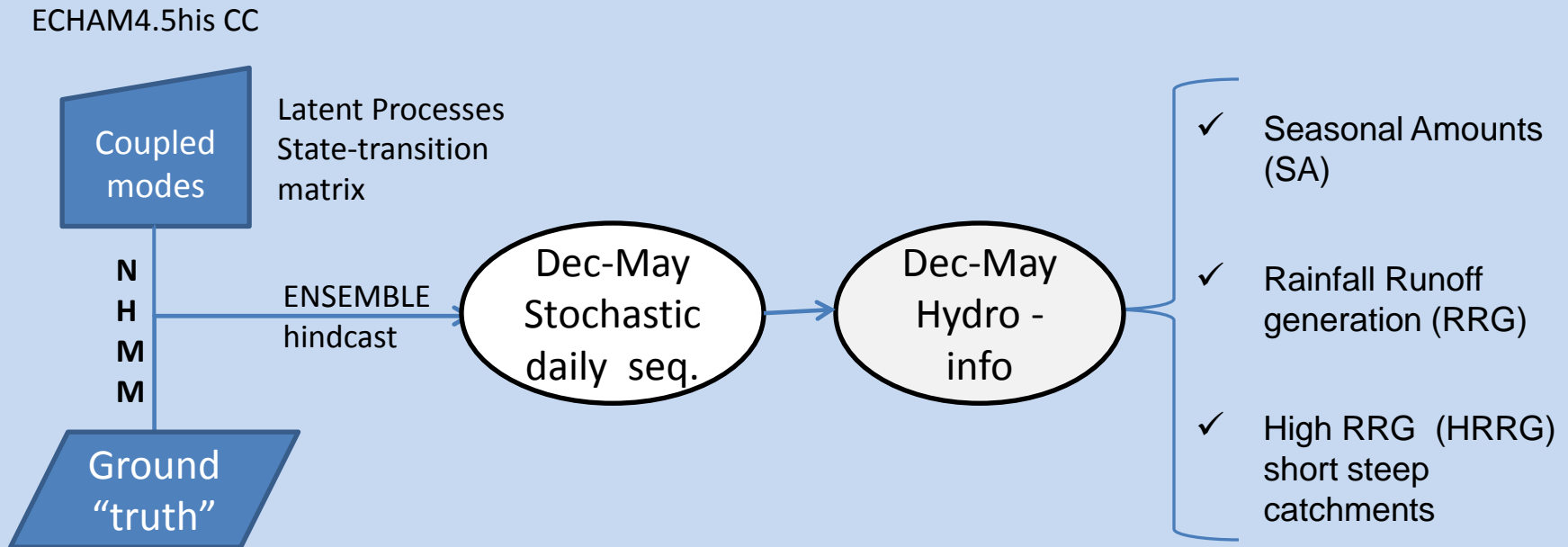
ECHAM4.5hi PC2 (25.98%)



- ECHAM4.5 (historical-simulations) pixel → spatial-bias
- ECHAM4.5 (historical-simulations)
 - 1st Canonical variate (2 PCs)
 - 1st PC ITCZ during EN years - 2nd PC ITCZ no EN years

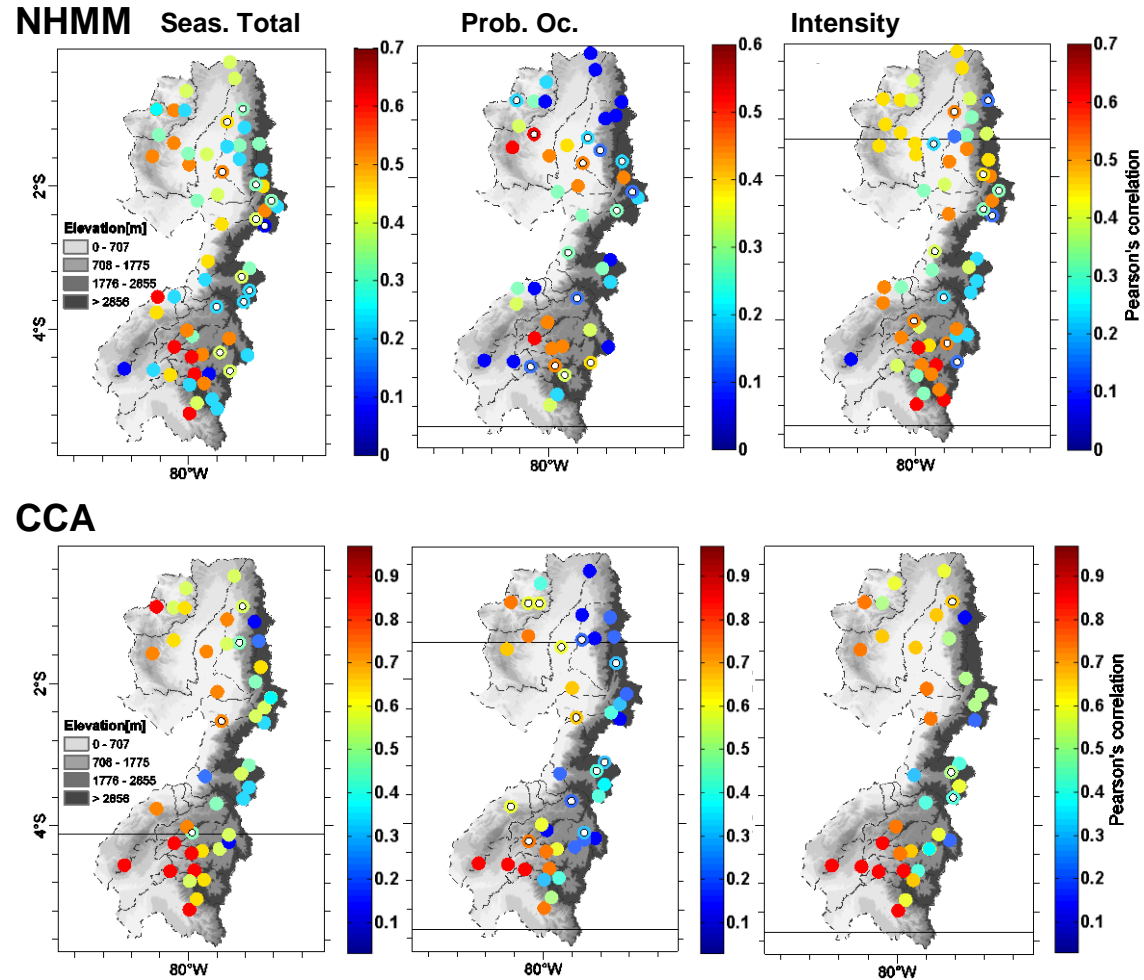
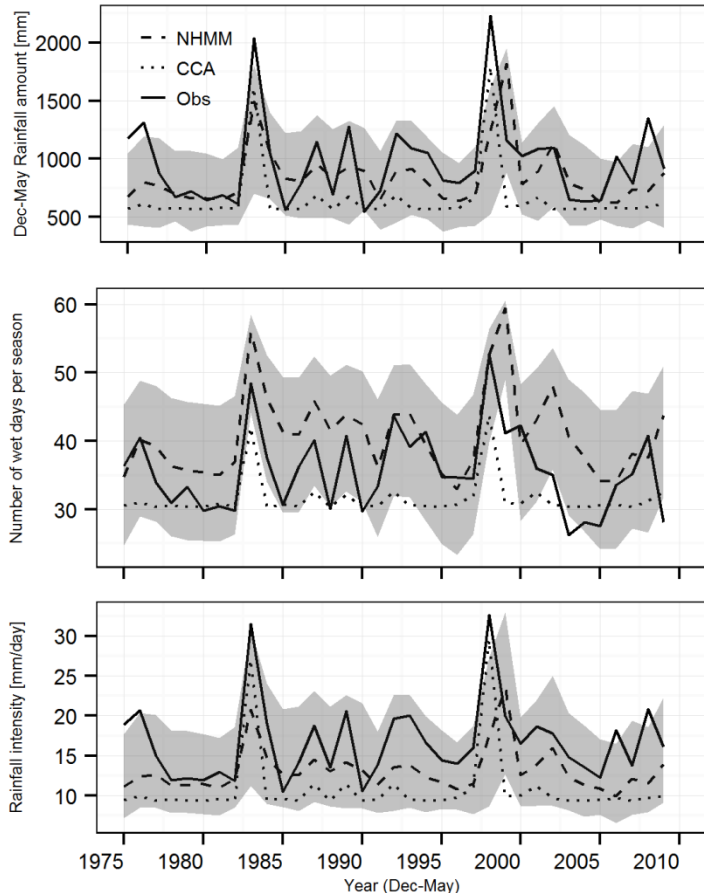
3. Dec-May process-based downscaling

Weather-type daily rainfall downscaling using a NHMM



Process-based downscaling (Dec-May)

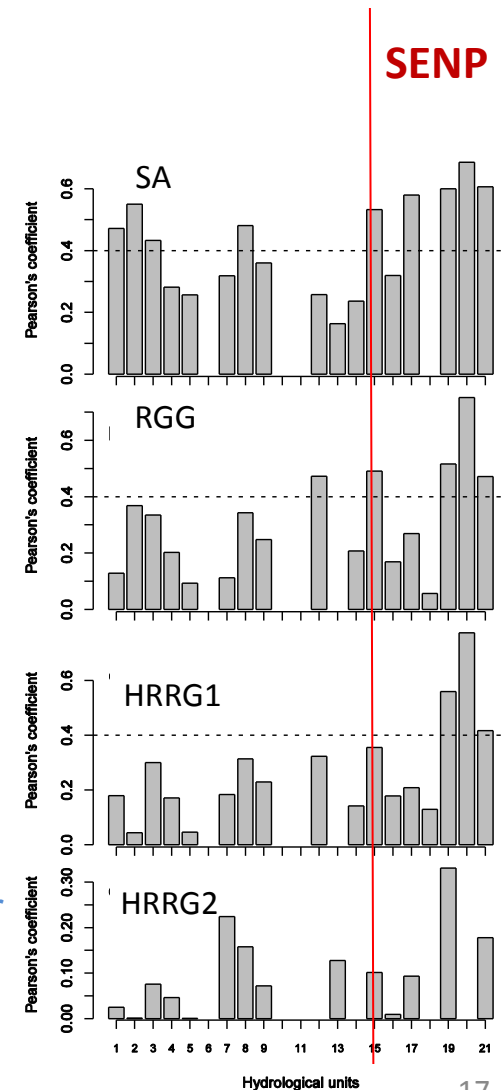
NHMM vs. CCA (correlation)



- 100 stochastic daily simulation of 36 Dec-May rainfall
- Better skills over SENP: Seas Total (+), Intensity w/d (++) , Prob. Occur. (-)

Conclusions

- ECHAM4.5 performs better than CFSv2
 - AGCM: SSTs lower boundary forcing is crucial for regional climate prediction
 - Challenge: predict SSTs (east of 105W)
- Extraction of the AGCM process-related signal is key to input the NHMM
 - Onset (Dec-Jan): dynamic (ITCZ/SAM) > thermal
 - Mature (Feb-Mar) : dynamic =~ Thermal
 - Retreat (Apr-May): thermal > dynamic (ITCZ)
- GCM-NHMM shows potential to produce daily rainfall info for catchment hydrology
 - Challenge: complex spatial dependences over inner Andean catchments





Thanks