

How reliable are decadal climate predictions?

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Initialised decadal predictions (INIT) vs. other sources of future information available to stakeholders

→ Different **time series**

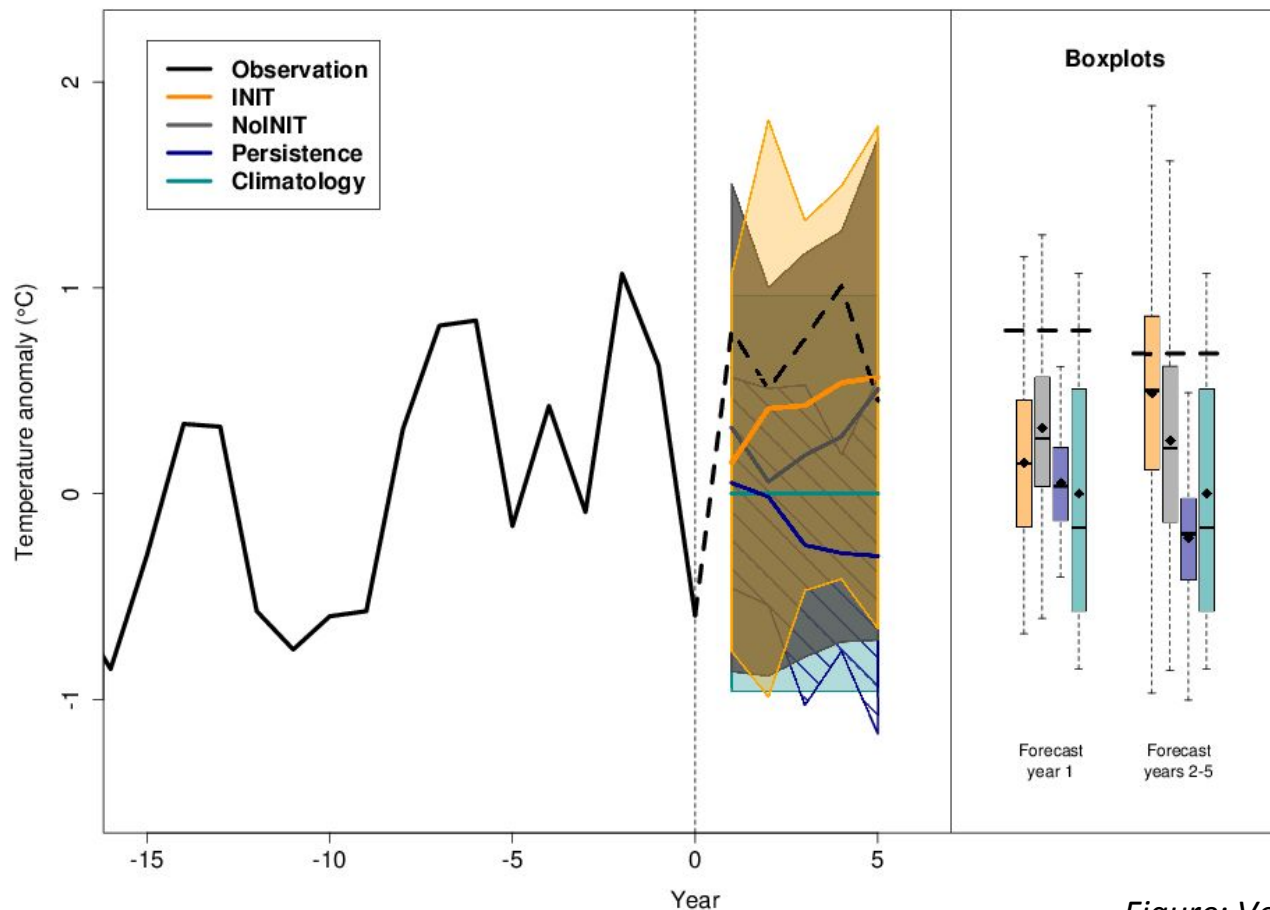


Figure: Verfaillie et al.

Initialised decadal predictions (INIT) vs. other sources of future information available to stakeholders → Different **probability distributions**

Decadal forecasts for Year +1

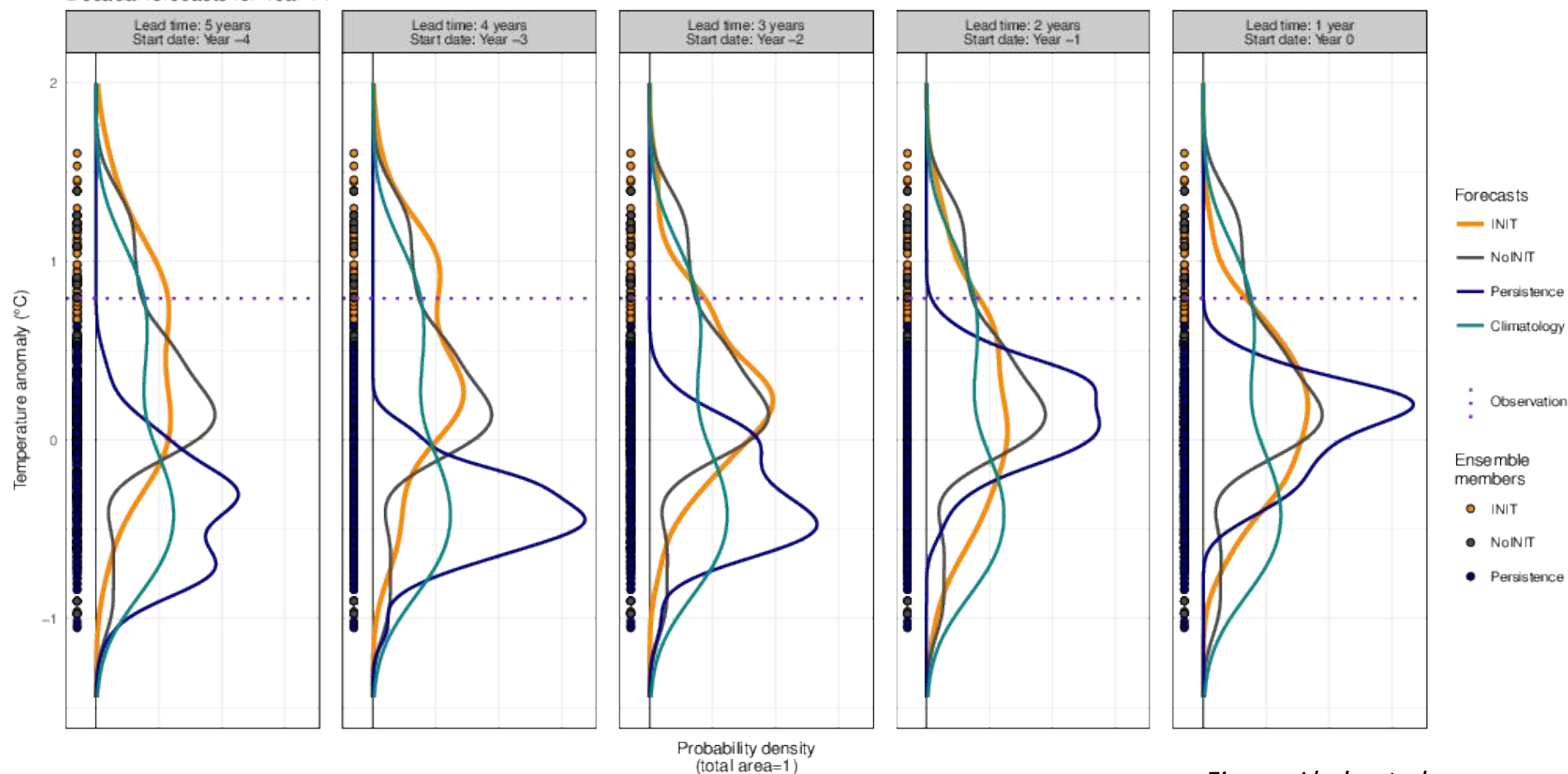


Figure: Lledo et al.

Comparing INIT and NoINIT

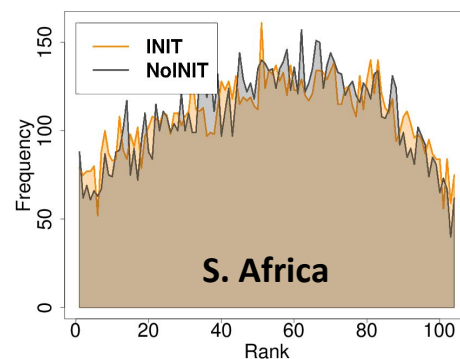
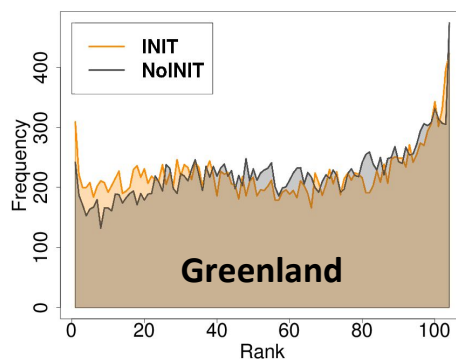
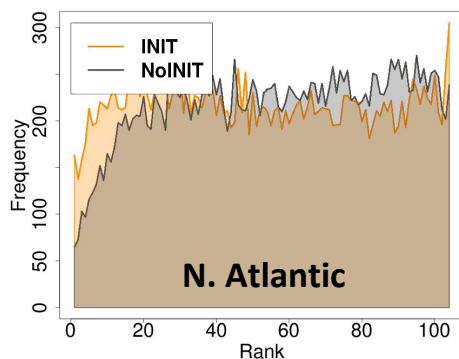
Generally done in terms of forecast accuracy (skill scores)

Here: impact of initialisation in terms of **reliability**

= agreement between the predicted probabilities and observed relative frequencies of a given event

Different tools:

- **rank histograms** → illustrate if ensemble members and verifying observation come from the same probability distribution, in which case the rank histograms are flat



Comparing INIT and NoINIT

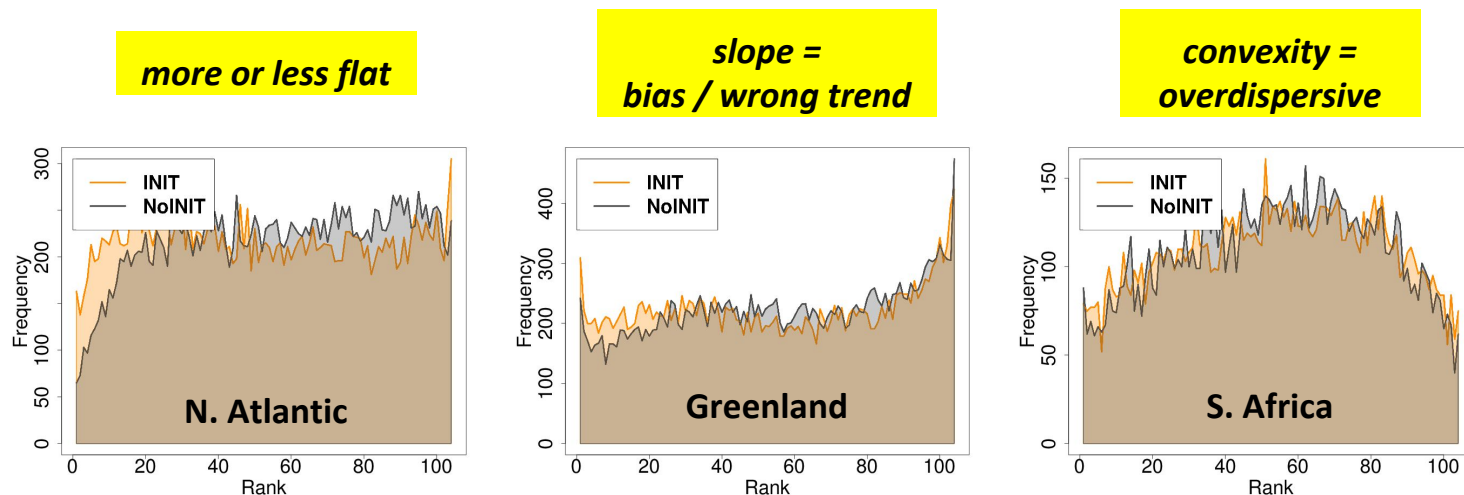
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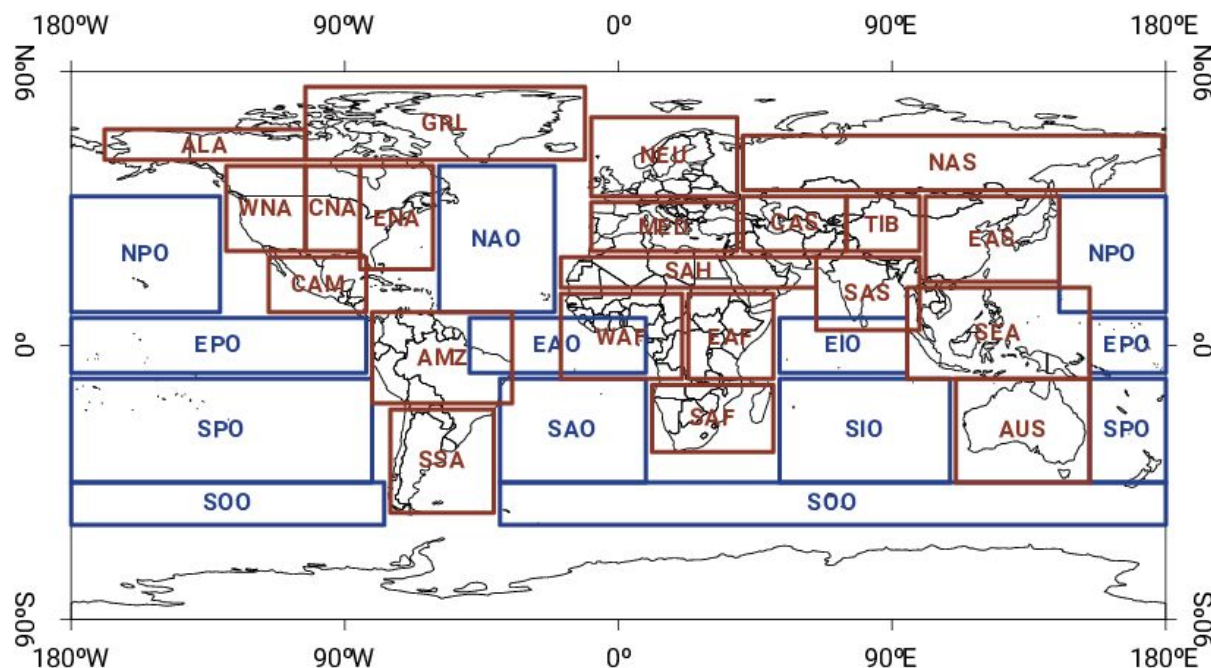
Different tools:

- rank histograms
- Joliffe and Primo (2008) test statistics: Pearson χ^2
+ decomposition into “slope” & “convexity” (& residual)



Comparing INIT and NoINIT

Near-surface air T° , multi-model set of 12 different models (CMIP5 & SPECS, both INIT & NoINIT), 30 different regions

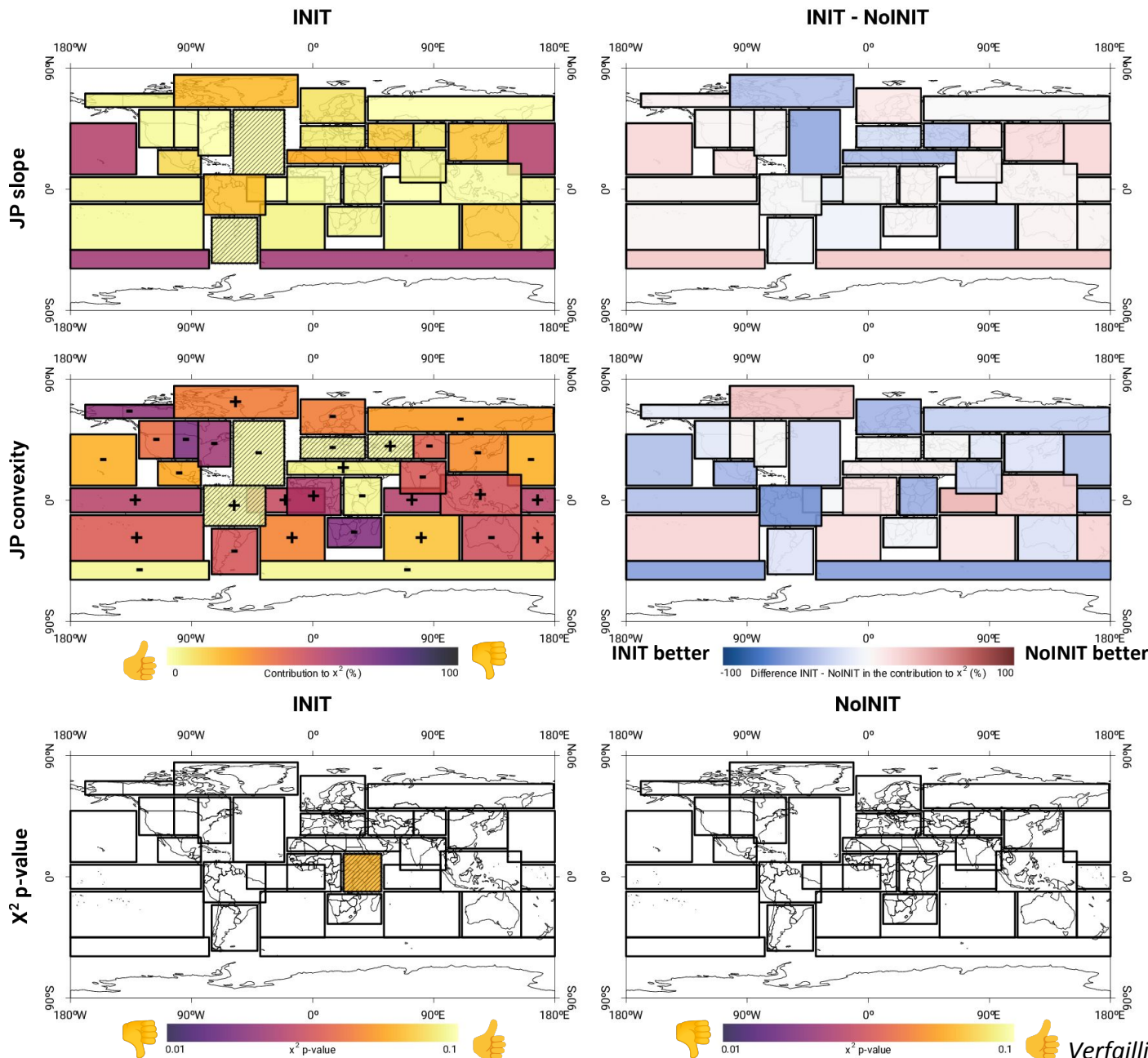


SOO: Southern Ocean
SAO: South Atlantic Ocean
EAO: Equatorial Atlantic Ocean
NAO: North Atlantic Ocean
SIO: South Indian Ocean
EIO: Equatorial Indian Ocean
SPO: South Pacific Ocean
EPO: Equatorial Pacific Ocean
NPO: North Pacific Ocean
AUS: Australia

AMZ: Amazon Basin
SSA: Southern South America
CAM: Central America
WNA: Western North America
CNA: Central North America
ENA: Eastern North America
ALA: Alaska
GRL: Greenland
MED: Mediterranean Basin
NEU: Northern Europe

WAF: Western Africa
EAF: Eastern Africa
SAF: Southern Africa
SAH: Sahara
SEA: Southeast Asia
EAS: East Asia
SAS: South Asia
CAS: Central Asia
TIB: Tibet
NAS: North Asia

MM, forecast year 1, uncorrected data



underdispersive
overdispersive
no slope/convexity error ($p \geq 0.05$)

+ - ///

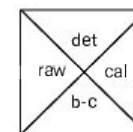
Comparing INIT and NoINIT

- Both uncorrected INIT & NoINIT generally **unreliable**
- **Biased**, incorrect **trend** and/or **dispersion** errors
- Some regions (e.g., NAO) without bias or dispersion error, yet unreliable (**residual** parameter)

MM

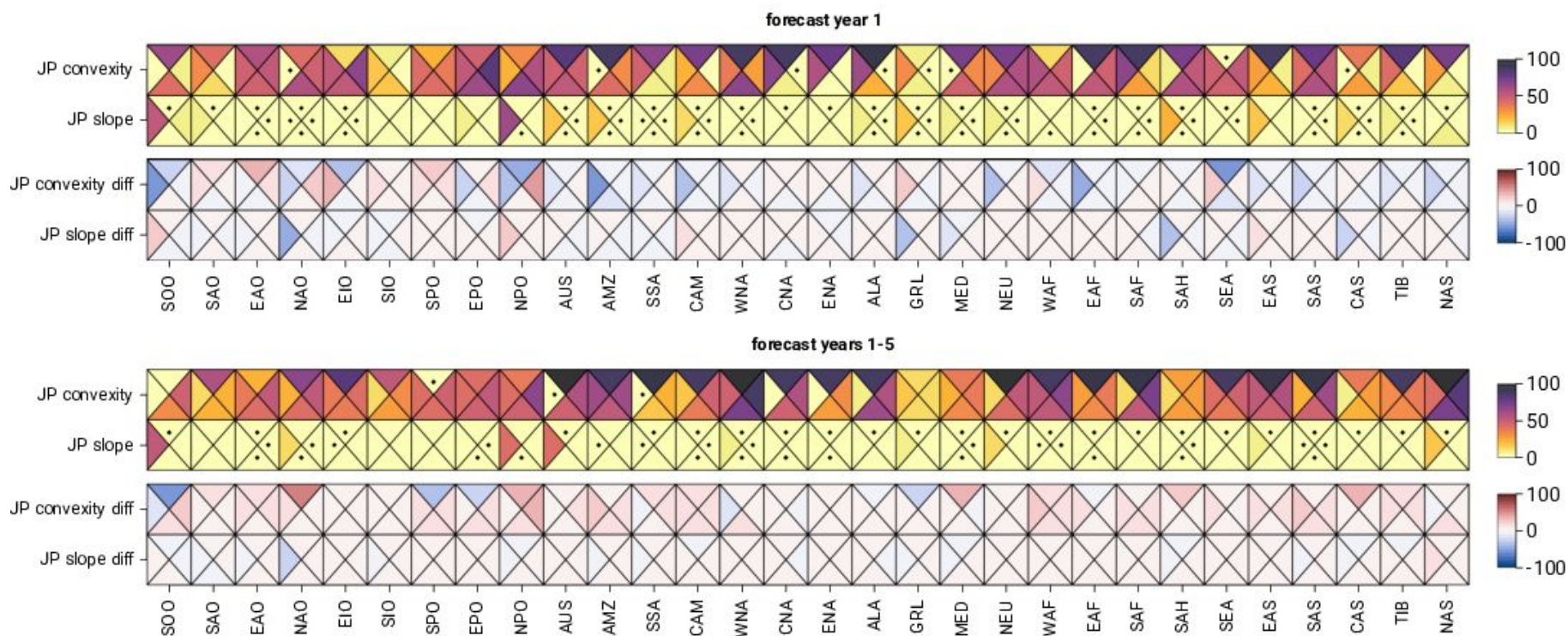
Impact of forecast time

Impact of different post-processing methods



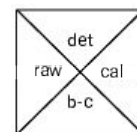
raw = uncorrected
det = detrended
b-c = bias-corrected
cal = calibrated

• no slope/convexity error ($p \geq 0.05$)



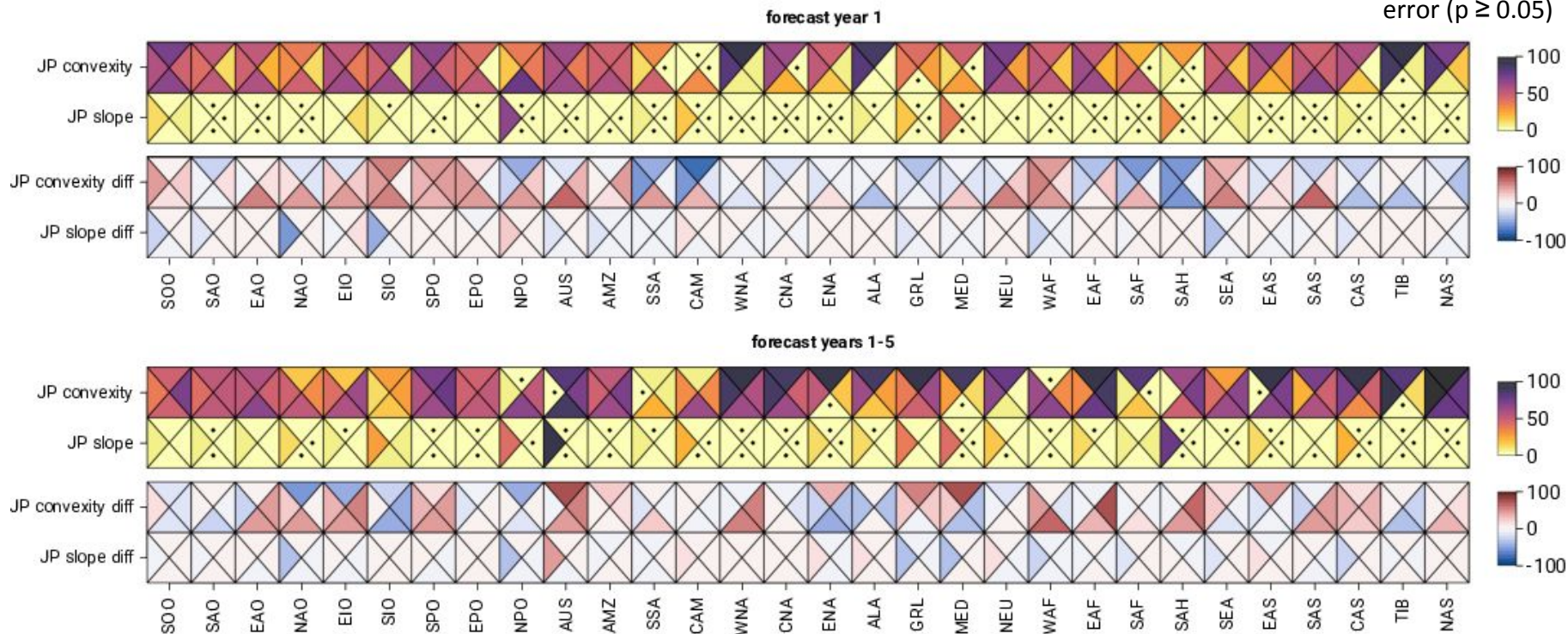
- Effect of **initialisation** on reliability small, mostly limited to forecast year 1
- **Bias correction & calibration** necessary for reliable forecasts

Impact of different model ensembles: NCAR



raw = uncorrected
det = detrended
b-c = bias-corrected
cal = calibrated

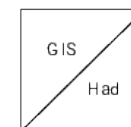
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- **Combination of models** more important than **ensemble size** of each individual forecast system
- Due to large range of **model physics & initialisation approaches** → error compensation

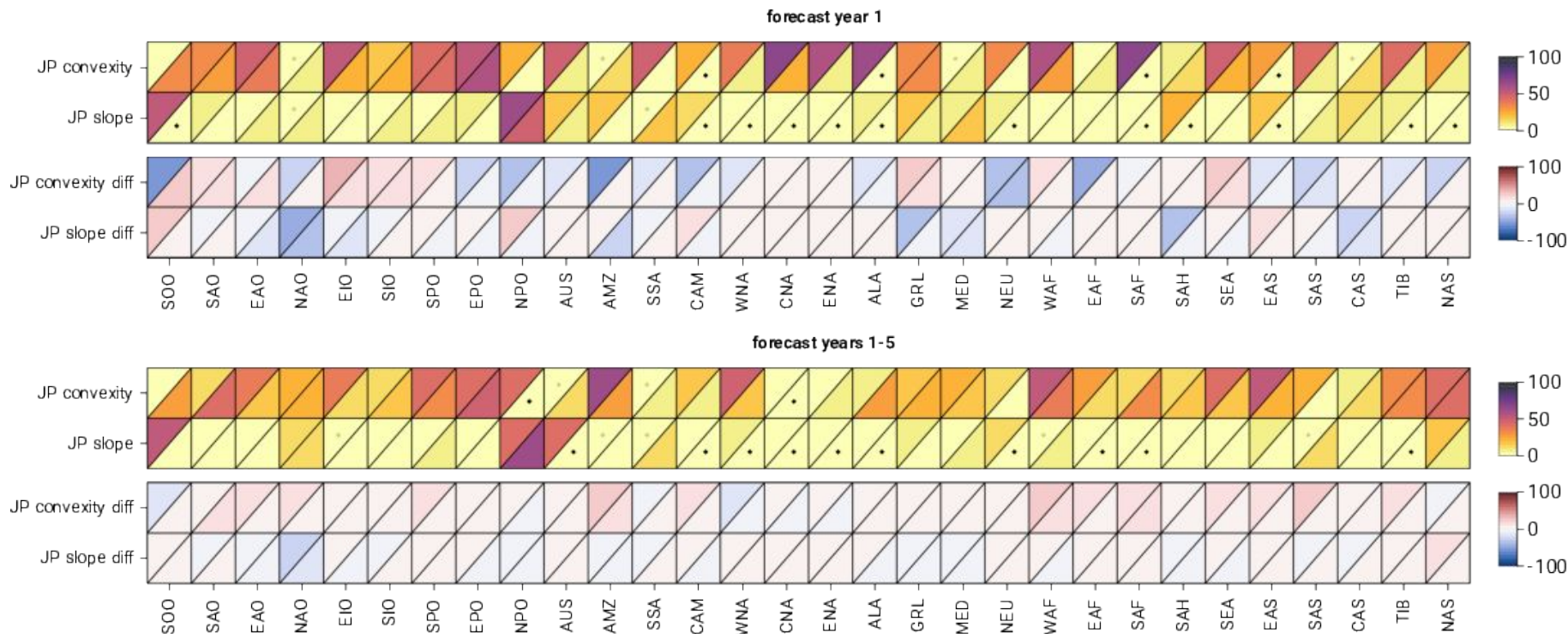
MM, uncorrected data

Impact of the reference obs dataset



GIS = GISTEMP
Had = HadCRUT4

- no slope/convexity error ($p \geq 0.05$)



- Impact of different reference datasets
- Measure of **observational uncertainty**

Main conclusions

- Both uncorrected INIT & NoINIT generally **unreliable**
- **Biased**, incorrect **trend** and/or **dispersion** errors
- Some regions (e.g., NAO) without bias or dispersion error, yet unreliable (**residual** parameter)
- Effect of **initialisation** on reliability small, mostly limited to forecast year 1
- **Bias correction & calibration** necessary for reliable forecasts
- **Combination of models** more important than **ensemble size** of each individual forecast system
- Due to large range of **model physics & initialisation approaches** → error compensation
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Thanks!

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