



Seamless subseasonal probabilistic streamflow forecasting

MuTHRE lets you have your
cake and eat it too!

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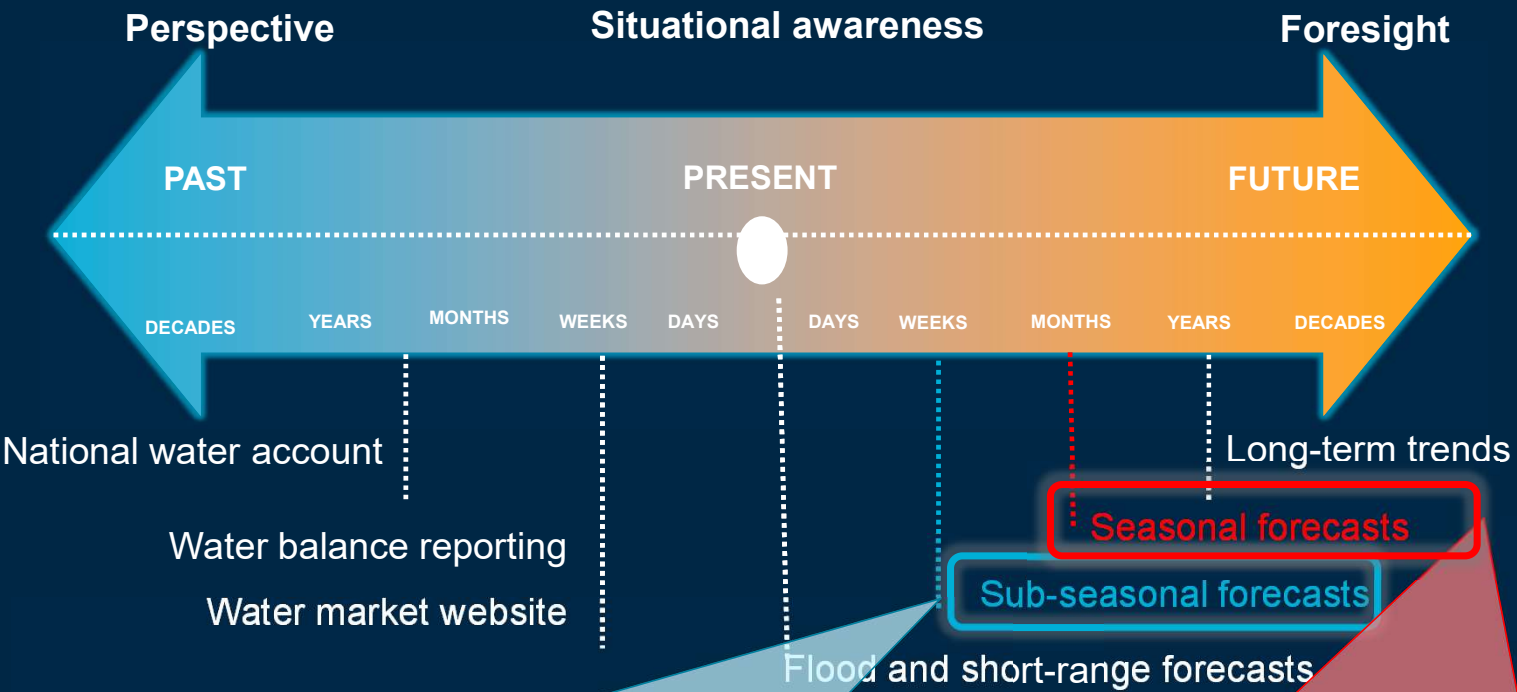
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Narendra Tuteja (WaterNSW)**

**make
history.**



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Australian Bureau of Meteorology: Water Information Products



New Capability: Seamless Subseasonal Streamflow Daily Forecasts

- Daily forecasts for 0-30 day lead time.
- Seamless forecasts with consistent quality at range of lead times and aggregation time scales (daily/weekly/monthly)

Existing Capability: Non-seamless Seasonal Streamflow Monthly Forecasts

- Available at single time scale/lead time (monthly)
- Cannot be reliably aggregated/disaggregated to longer/shorter lead times

Practical Benefits of Seamless vs Non-Seamless Forecasting

Seamless forecasts can inform decisions at a range of time scales

- Flood warning (up to 1 week)
- Managing hydropower systems (7 and 15 days)
- Managing water supply reservoirs where forecast volumes over lead times of weeks/months
- Non-seamless only works at single time scale



Seamless daily forecasts integrate into real-time decision-making models

- River system models run at daily time timesteps => Seamless daily forecasts easily used as inputs
- Enables quantify impacts of forecasts
- Non-seamless seasonal models cannot be used as input => only used 'qualitatively'

Seamless forecasts simplify forecasting systems

- A single seamless product can serve a range of forecast requirements at different time scales
- Non-seamless models require multiple forecasting products at different time scales

Key Question: Do seamless forecasts (with all these extra benefits) produce similar performance as non-seamless forecasts at same time scale

Seamless Subseasonal Daily Forecasts: MUTHRE

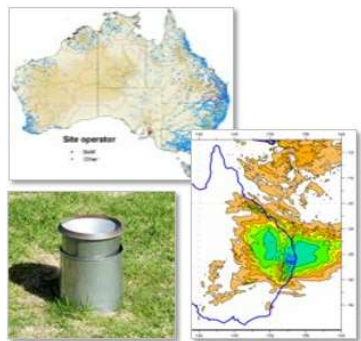
Produce daily subseasonal streamflow forecasts with high quality performance for range of lead times (0-30 day) and time scales (daily to monthly)



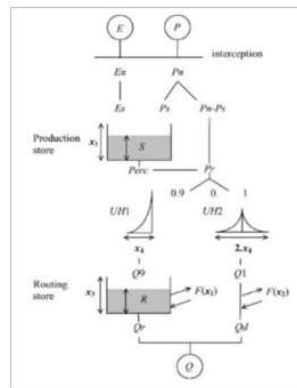
Key innovation

MuTHRE: Multi-temporal hydrological residual error model

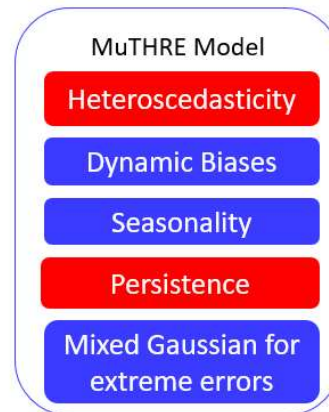
- **Seasonality:** Errors vary systematically by month
- **Dynamic Biases:** Errors vary yearly, due to non-stationarity
- **Extreme errors:** Occasional very large errors, poorly represented by Gaussian distribution



Daily Rainfall Forecasts



Rainfall-runoff model



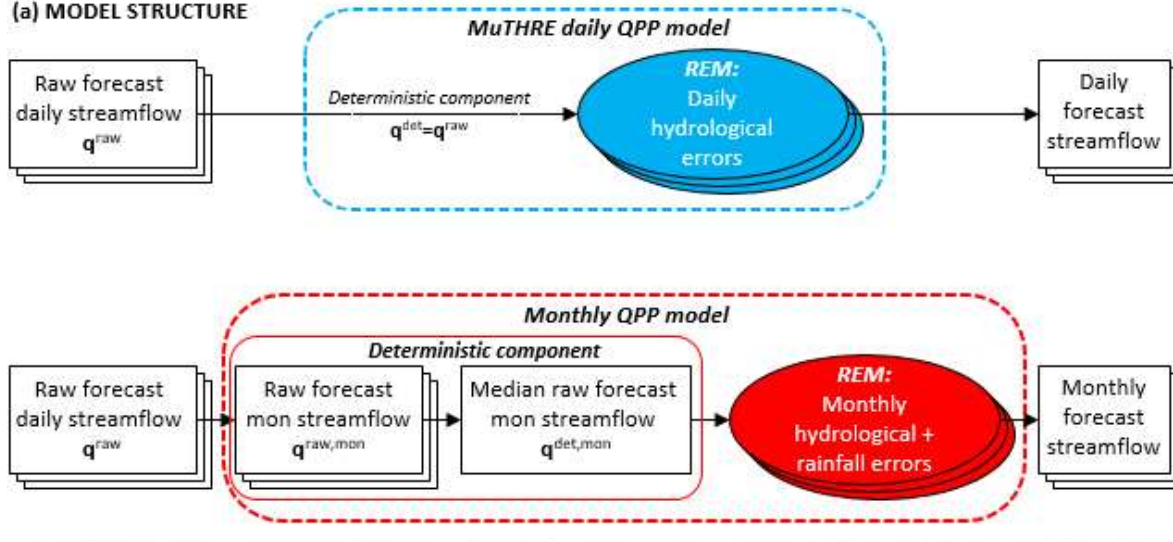
Daily streamflow post-processing

- McInerney et al, 2020, WRR: Multi-temporal Hydrological Residual Error Modeling for Seamless Subseasonal Streamflow Forecasting <https://doi.org/f8k5>
- McInerney et al, 2021, WRR, Improving the Reliability of Sub-Seasonal Forecasts of High and Low Flows by Using a Flow-Dependent Nonparametric Model, <https://doi.org/jfz6>

Overview of Forecasting Approaches



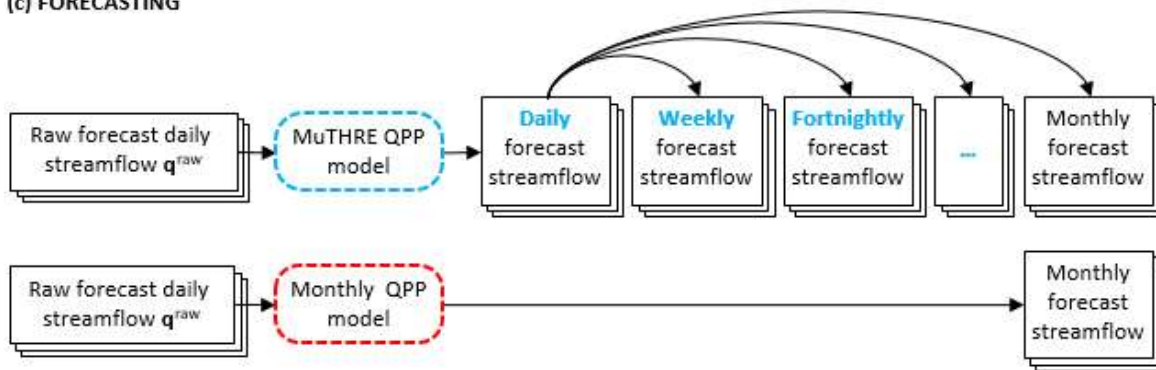
(a) MODEL STRUCTURE



Seamless Subseasonal Daily Forecasts: MUTHRE

Non-Seamless Seasonal Forecasts: Monthly QPP

(c) FORECASTING



Key Question: Do seamless forecasts produce similar performance as non-seamless forecasts at monthly time scale

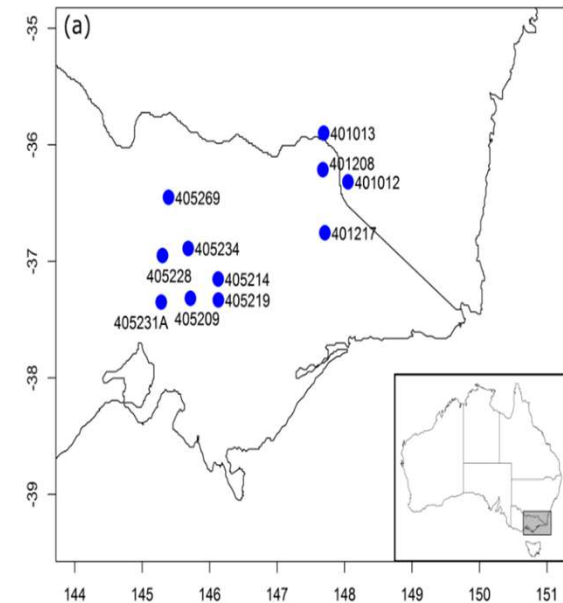
Expectation:

Non-seamless Forecasts would perform better as they only have on 'job' to do: produce forecasts at monthly time scale used for calibration

Seamless Subseasonal: Produce forecast at range of lead times and aggregation times scales

Forecast Evaluation Approach: Comprehensive and Systematic

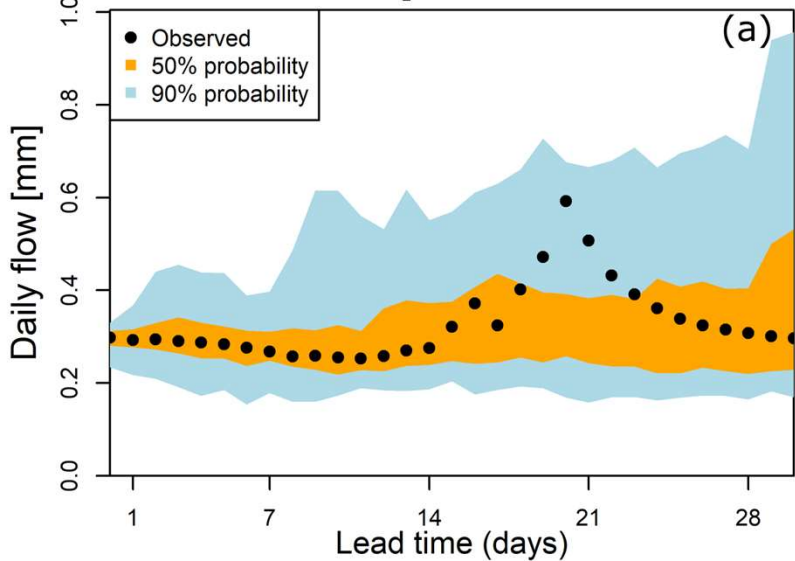
- Eleven key unregulated catchments in Murray-Darling Basin
 - >60% inflow to major storages
- Range of commonly used metrics to capture key aspects of forecast performance
 - Reliability, Sharpness, Volumetric Bias, CRPS
- Range of time scales/stratified evaluations to assess performance consistency
 - Daily and Monthly Time Scales
 - Stratified by Lead Time/Months/Years
- Leave-one-out Cross Validation Procedure



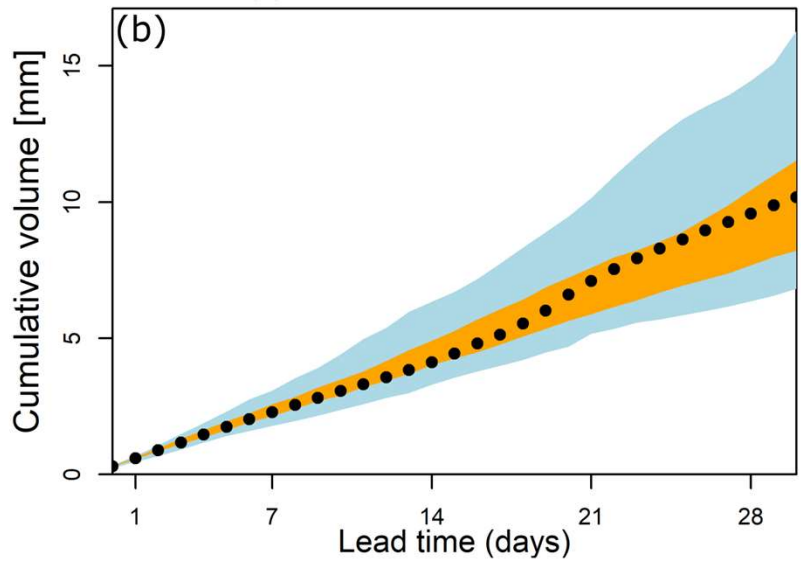
Key Outcomes: Daily Forecasts (MUTHRE)



Daily flows



Cumulative volumes

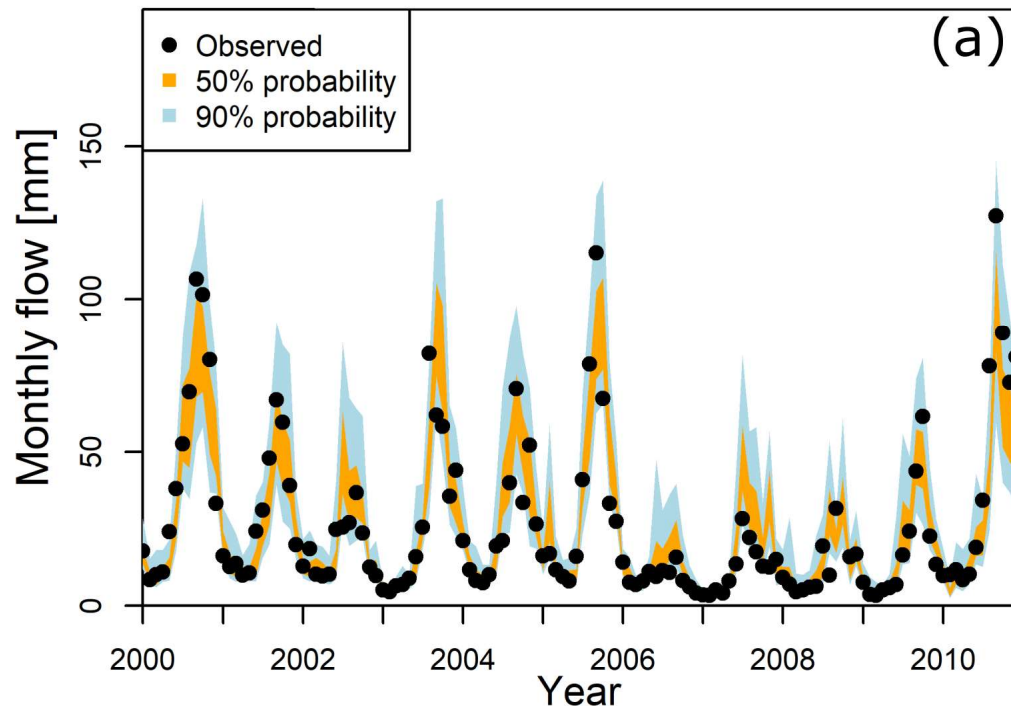


- Reliable and sharp forecasts across a range of lead times and aggregation time scales (daily/monthly) and stratifications (month/year)
- Monthly QPP cannot produce daily forecasts

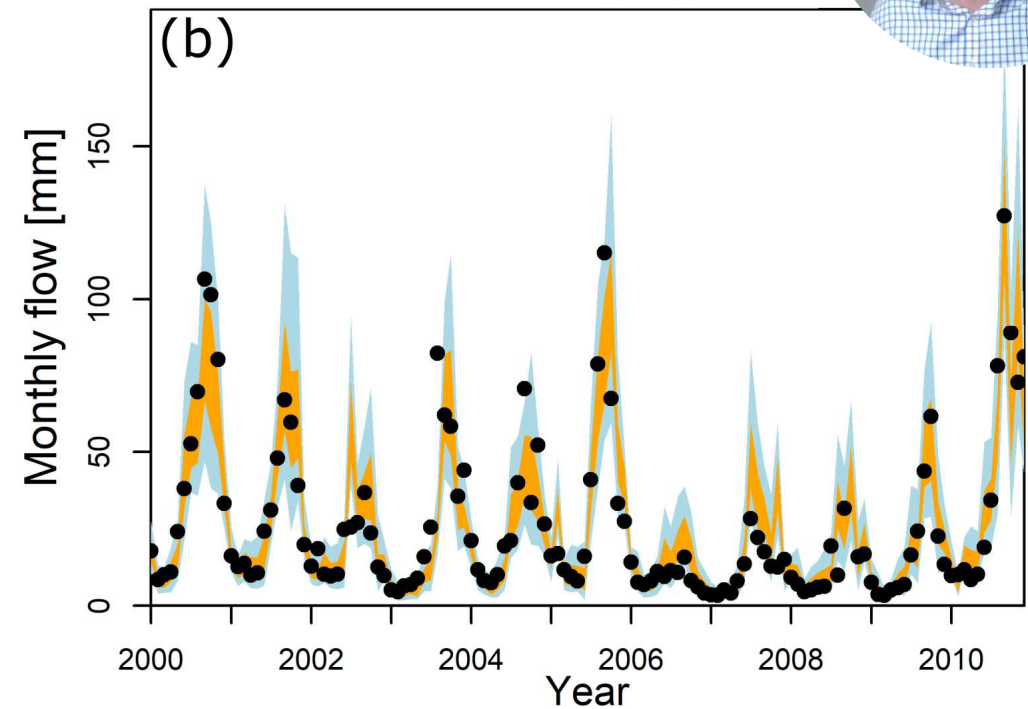
Key Outcomes: Monthly Forecast Time Series



MuTHRE

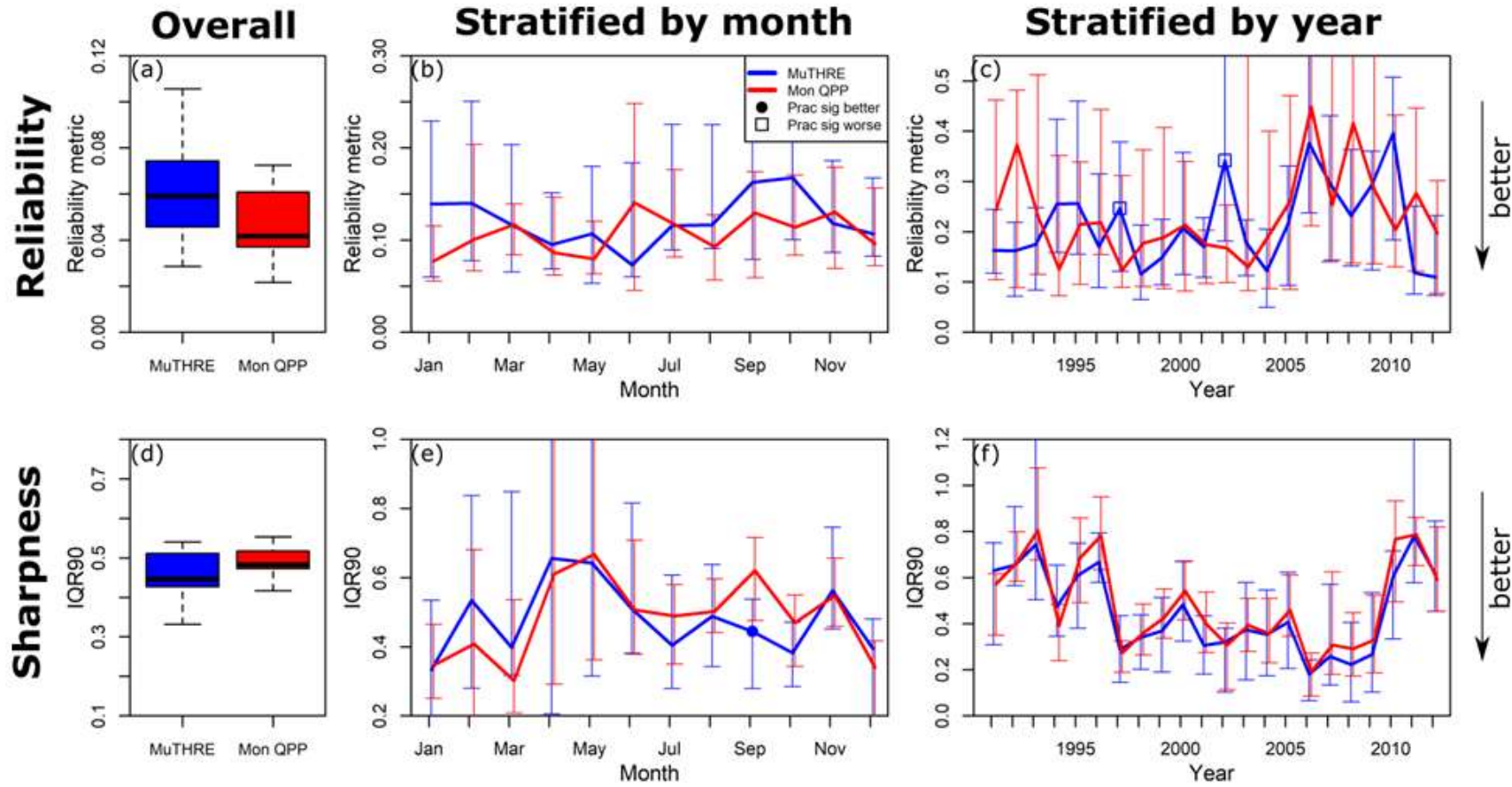


Monthly QPP



- Seamless daily forecasts from MUTHRE aggregated to monthly time scales produce similar results to non-seamless Monthly QPP

Key Outcomes: Monthly Forecasts Metrics



- Seamless daily forecasts from MUTHRE aggregated to monthly time scales produce similar results to non-seamless Monthly QPP

Summary



- Daily seamless subseasonal forecasts from MUTHRE provide essentially the same performance as the non-seamless Monthly QPP
- Seamless forecasts provides a wide range of practical benefits over non-seamless forecasts
- Modeller can proverbially ‘have their cake and eat it too’
- Key finding: Seamless forecasting is not only viable but a preferred choice for future research development and practical adoption in streamflow forecasting
- Any question: email **mark.thyer@adelaide.edu.au**
- See further details in journal paper:

McInerney et al, 2022, HESS, Seamless streamflow forecasting at daily to monthly scales: MuTHRE lets you have your cake and eat it too, <https://doi.org/j68c>
