

QUANTIFYING THE OVERALL ADDED VALUE OF DYNAMICAL DOWNSCALING AND THE CONTRIBUTION FROM DIFFERENT SPATIAL SCALES

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Motivation of the study

- An essential requirement for RCMs is that they improve the representation of the climate compared to the global driving data (GDD), i.e., RCMs produce **added value (AV)**.

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Open question

- Which of these two situations (improvement/deterioration) is more dominant?
- Do RCMs produce in general a substantial **overall** improvement over the driving data?

Objectives of the presentation

- 1 To **evaluate the overall AV** of an ensemble of RCM simulations and its dependence on a variety of factors.
- 2 To decompose the AV according to the **contribution of different spatial scales** :
 - **large scales** : scales common to both the GDDs and the RCMs
 - **small scales** : scales only represented in the higher resolution RCMs
- 3 To explore how much of the AV can be attained using **simpler postprocessing methods**.

Two “Added Value” metrics

- 1 **Mean square error** : relative performance in terms of absolute values :

$$\begin{aligned} AV_{MSE} &= \overline{(X_{GDD} - X_{OBS})^2} - \overline{(X_{RCM} - X_{OBS})^2} \\ &= MSE_{GDD} - MSE_{RCM}. \end{aligned}$$

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- Metrics are **normalized** to allow for comparisons across regions, seasons, etc.
- Quantities vary between -1 and 1 with positive values suggesting that there is some AV.

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| 1) dx (RCM) | ES | 2 | 10 and 50-km grid spacing |
| 2) RCM | ES | 4 | R1, R2, R3, ensemble mean |
| 3) GDD | ES | 3 | NNRP, CCCMA3.1, ECHAM5 |

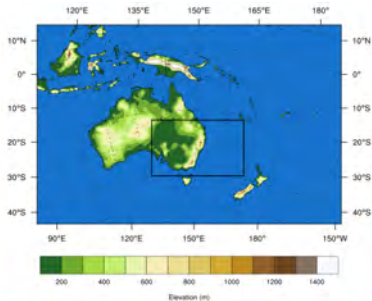
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RCM domains

- 50-km over Australasia
- 10-km over south-east Australia



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| 4) Season | CS | 4 | DJF, MAM, JJA, SON |
| 5) Region | CS | 3 | coastal, topography, flat |
| 6) Variable | CS | 3 | Tmin, Tmax, precipitation |
| 7) Measure | CS | 3 | mean, stddev, 99 th percentile |

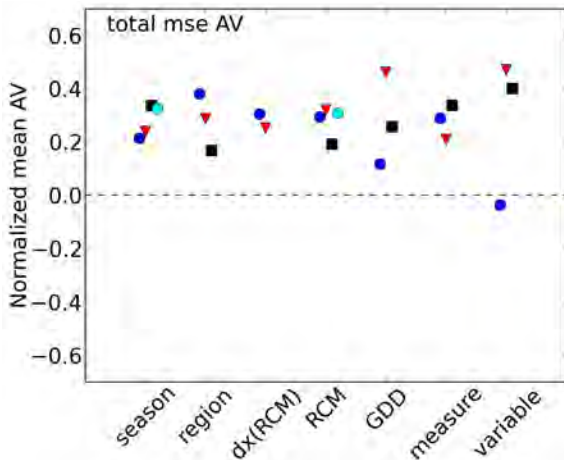
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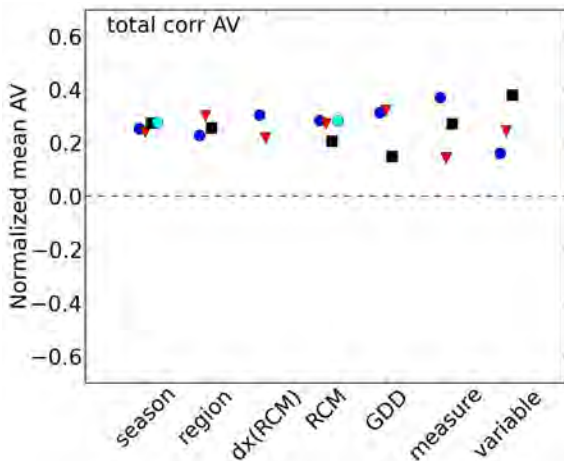
TOTAL NUMBER OF AV ESTIMATIONS : 2592

mean square error : $AV = (MSE_{GDD} - MSE_{RCM}) / errors_{Sum}$



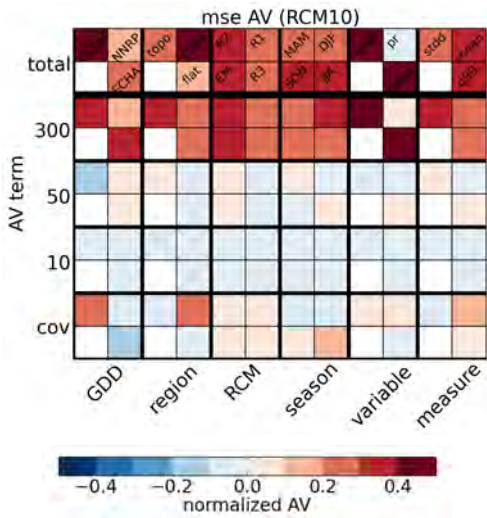
| season | region | dx (RCM) | RCM | GDD | measure | variable |
|--------|---------|----------|------|------------|----------|----------|
| ● DJF | ● coast | ● 10 | ● R1 | ● NNRP | ● mean | ● pr |
| ▼ MAM | ▼ topo | ▼ 50 | ▼ R2 | ▼ CCCMA3.1 | ▼ stddev | ▼ tmin |
| ■ JJA | ■ flat | | ■ R3 | ■ ECHAMS | ■ q99 | ■ tmax |
| ■ SON | | | ● EM | | | |

spatial correlation : $AV = (corr(RCM, OBS) - corr(GDD, OBS)) / errors_{sum}$



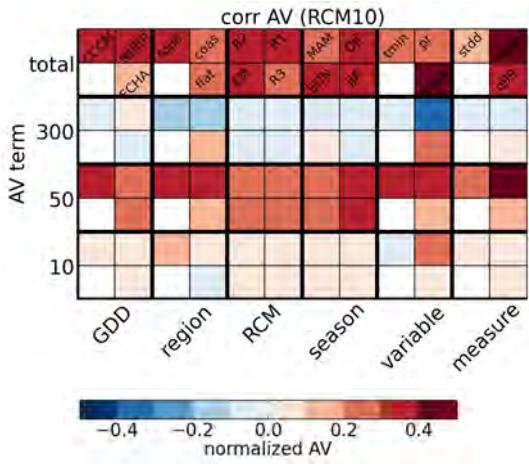
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AV as a function of spatial scales



- The AV of MSEs is dominated by the **large scale** term.

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- The AV of spatial correlations is dominated by the **small scale** term.

AV with simple GCM-postprocessing : T_{min}

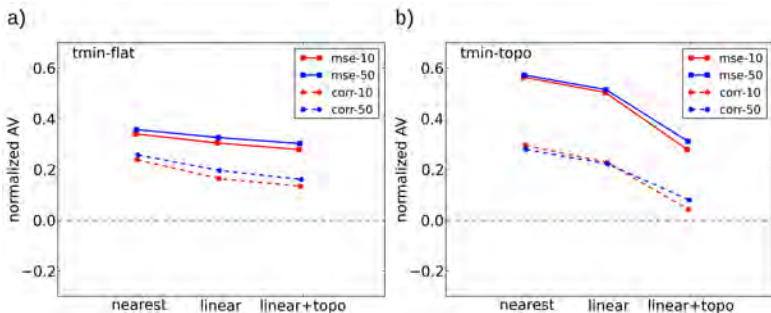


FIGURE : Overall AV for minimum temperature calculated using three different methods to interpolate GCM data.

- In general, about half of the AV in temperature variables can be obtained using simpler postprocessing.

Summary

- For both AV metrics, the overall performance of RCMs is generally superior compared with the corresponding driving data :
 - AV higher and more robust for spatial correlation metric.
- $AV_{MSE} \approx AV_{MSE}^{large\ scales}$
- $AV_{correlation} \approx AV_{correlation}^{small\ scales}$
- A significant amount of the overall AV in temperature variables was shown to be **attained using simpler post-processing** methods.

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Are these results dependent on the specific ensemble used in this study?

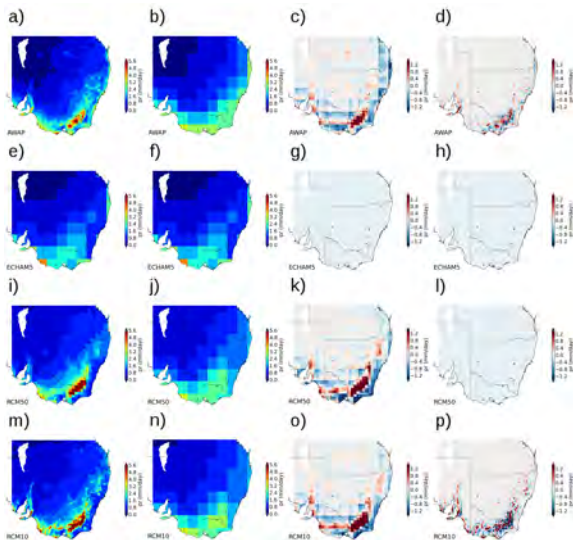
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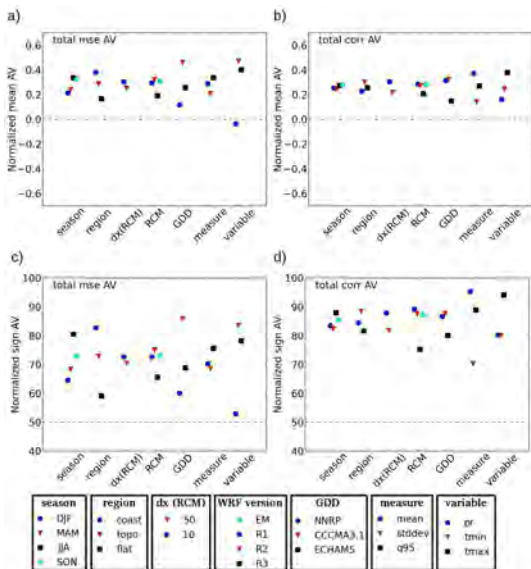
THANKS FOR YOUR ATTENTION!!!

QUESTIONS ? COMMENTS ?

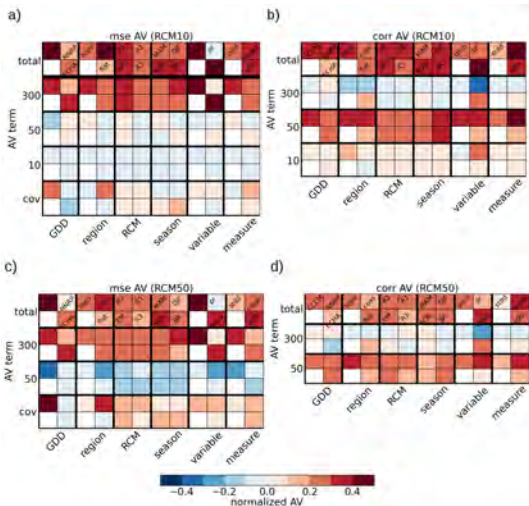
Spatial scale decomposition method



AV measures : mean Vs sign



AV as a function of spatial scales



10-km RCM vs 50-km RCM

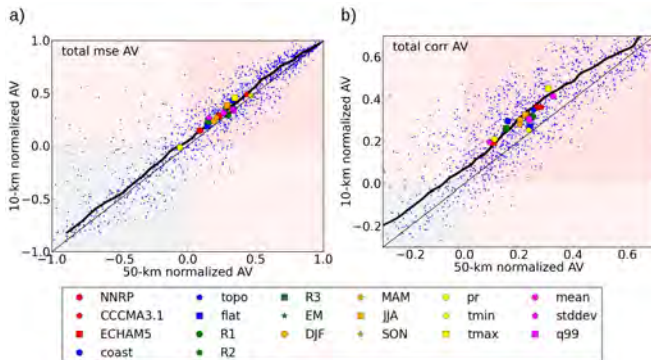


FIGURE : AV values for the **50-km RCM** simulations compared with the global driving data for different spatial scale terms.

Raw Vs. Corrected : minimum temperature

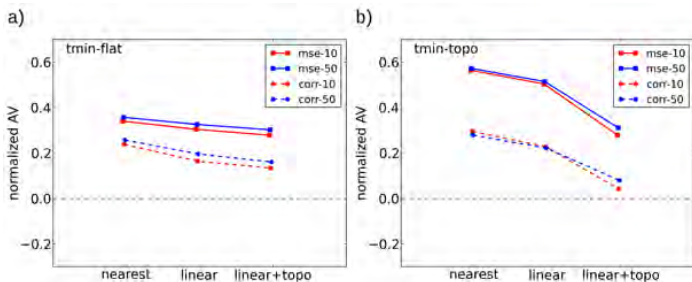


FIGURE : Overall AV for minimum temperature calculated using three and different interpolation : nearest neighbor, bilinear and bilinear plus topographic correction.

Raw Vs. Corrected : maximum temperature

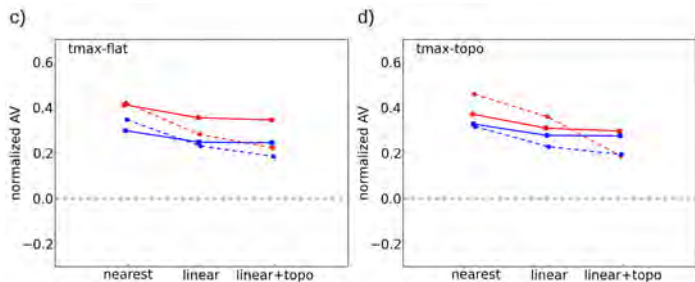


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