This manuscript presents **MESMER-RCM**, a modular and computationally efficient probabilistic emulator for annual mean 2 m temperature at regional (EURO-CORDEX) scales. The approach is conceptually simple and interpretable: a Lasso-based deterministic mapping from nearby GCM grid points to each RCM grid point, combined with a data-driven, two-stage shrinkage estimator for the RCM residual covariance that enables sampling of spatially correlated internal variability. I think the illustrations in this manuscript is valuable, but there are still some issues need to be addressed before acceptance.

- 1. The authors need to clarify the MESMER method in detail, maybe in the supporting information. A schematic diagram is necessary.
- 2. The authors need to describe why you choose MESMER framework and its advantages compared with other generative AI models.
- 3. How to divide the training samples and testing samples in MESMER-RCM? And how many years for training and testing? In addition, please mark the time span of every dataset in Table S1.
- 4. Why the residual variability belongs to a multivariate Gaussian distribution? Is there any risk for underestimations under some extreme events? Please clarify this issue.
- 5. I think there should be another benchmark method to compare for further confirm the advantage of MESMER-RCM.
- 6. Could you provide more metrics on result evaluation? Such as CRPS.
- 7. Whether the residual covariance matrixes for different time ranges (e.g., 1979–2000, 2001–2050, 2051–2099) are various? If so, whether such variations are significant? Please clarify its effectiveness.