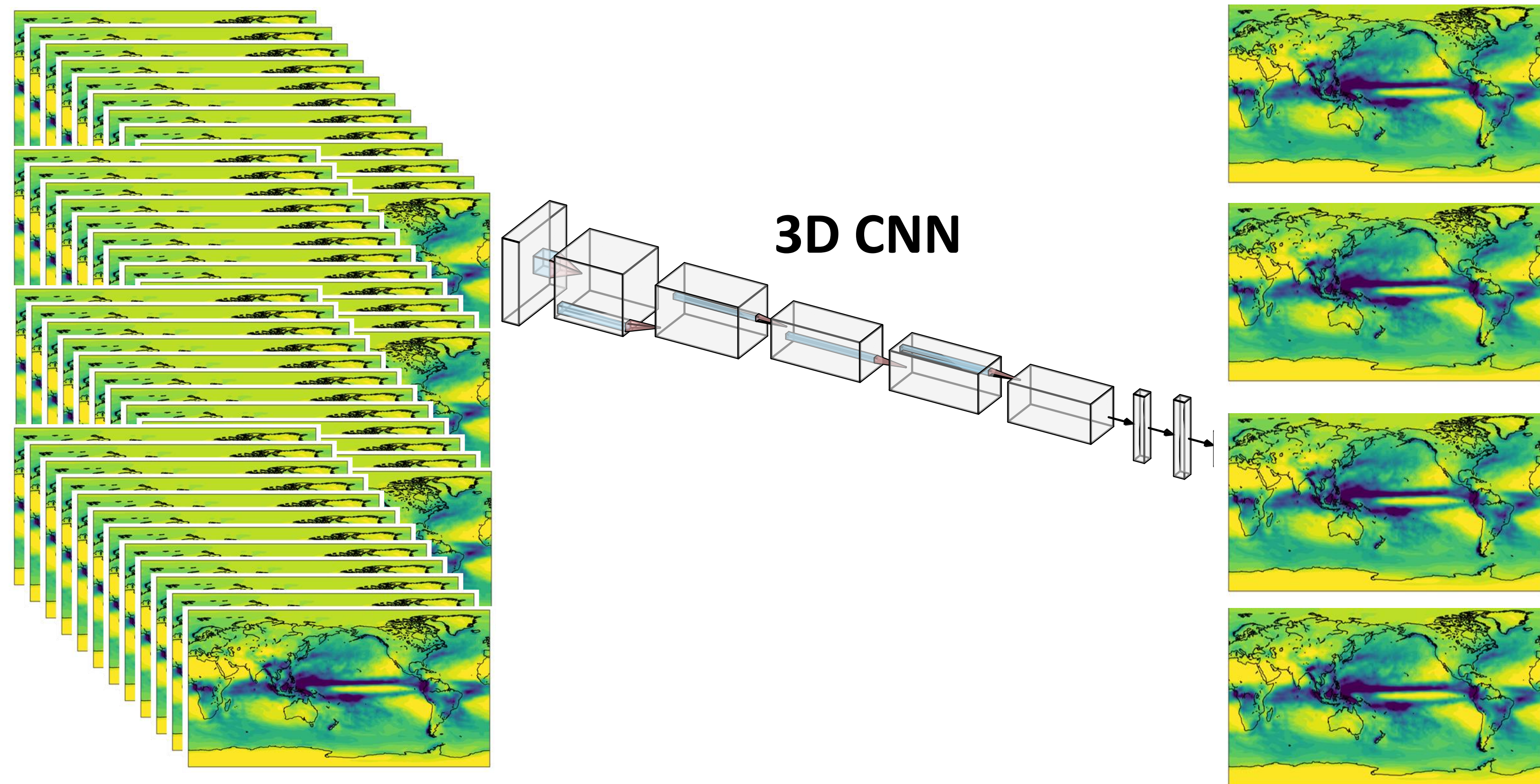


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## Method : using large ensembles and ML to model spatiotemporal forced response

3D Pr input : lat x lon x [t-10 to t years]

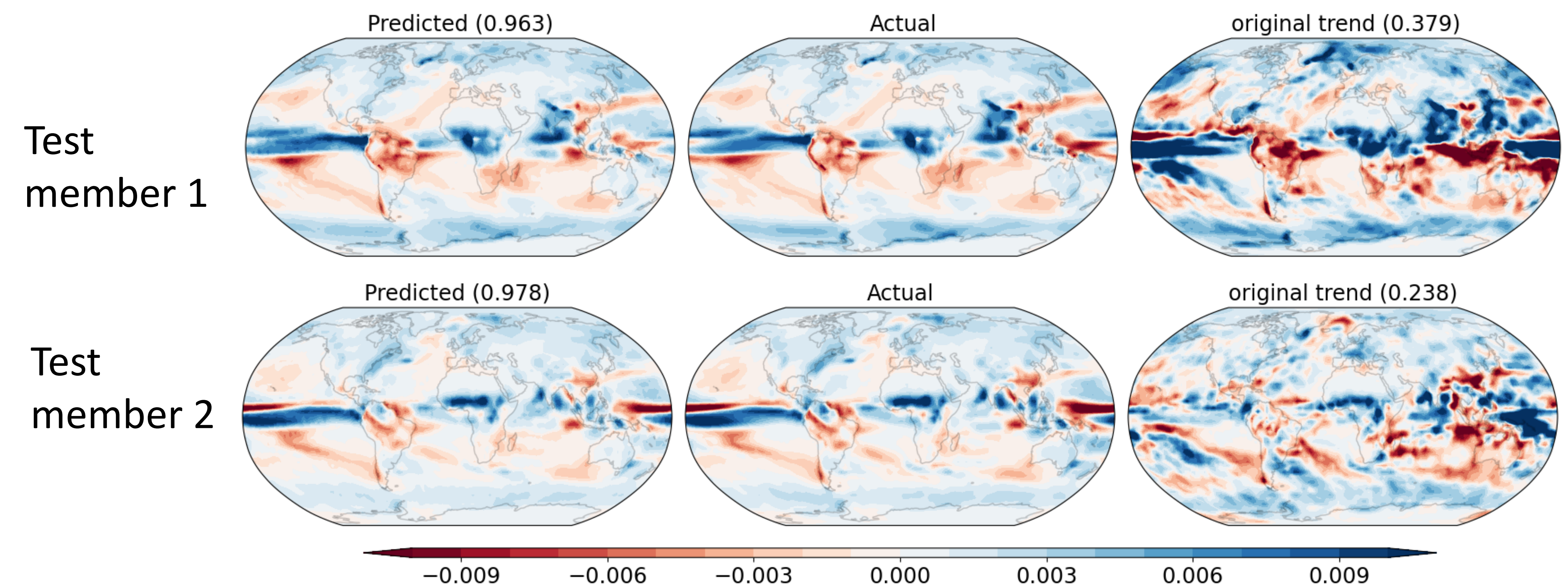
Year t forced response



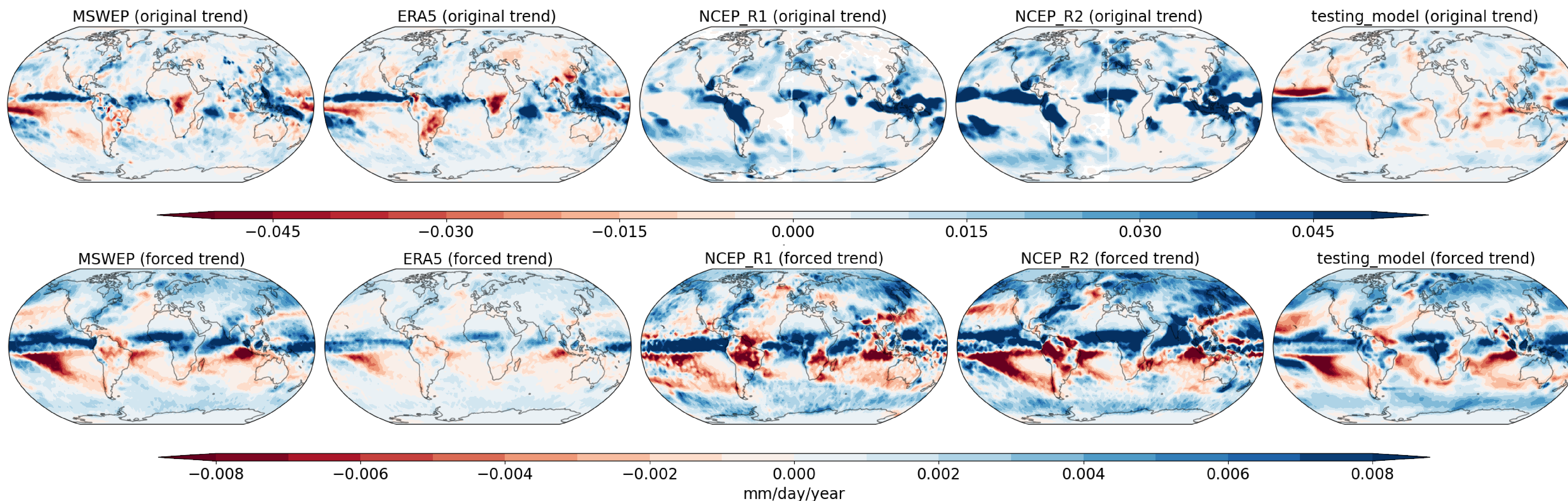
185 members x 63years x lat x lon x 10 time lags  
 (5 large ens) (1959-2022)

185 x 63 x lat x lon

## Model performance : trend calculated from the predicted and original spatiotemporal data (1959-2022) (mm/d/year)



## Observations : trend calculated from the predicted and original spatiotemporal data (1959-2022) (mm/d/year)



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